Computational methods for multi-parameter persistence

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Computational methods for multi-parameter persistence

PhD thesis

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Chapter 1

Introduction

Topological data analysis [31, 66] is a mathematical field that seeks to apply methods from topology to analyze shape and topological features of data sets. *Persistent homology* [58, 59, 61, 64, 107, 108], a central and arguably one of its most popular tools, is a multi-scale approach that captures the changes of the homology of a filtered topological space along the filtration. This can be interpreted as how clusters, holes and higher dimensional voids appear and vanish during the filtration. The theory of persistent homology touches upon algebraic and computational topology, discrete Morse theory [6, 69], representation theory [108] and others.

A common pipeline for the application of persistent homology is the following [76]:

- 1. Starting from a point cloud $S \subseteq Y$ lying in a topological space Y, build a topological subspace of Y filtered by a scale parameter. For example, if (Y,d) is a metric space, the spaces $S^{(r)} := \bigcup_{s \in S} B_r(s)$ for $r \in \mathbf{R}$ (where $B_r(s) = \{y \in Y \mid d(s, y) \leq r\}$) satisfy $S^{(r)} \subseteq S^{(r')}$ if $r \leq r'$, and thus assemble to an **R**-indexed filtration, called the *offset filtration* of S.
- 2. This filtered space is replaced by a finite filtered simplicial complex K_* , usually with the same (or similar) homology as the filtered space. For example, let Y be a Euclidean space, and define the *Voronoi* domain of $s \in S$ as

$$\operatorname{Vor}(s) = \{ y \in Y \mid d(y, s) \le d(y, t) \; \forall t \in S \}.$$

Then $\{B_r(s) \cap \operatorname{Vor}(s) | s \in S\}$ is a cover of $S^{(r)}$, and its nerve is a simplicial complex filtered by r, called the *Delaunay* or *alpha filtration* [60]; see Figure 1.1. By the (functorial) nerve theorem [13], the Delaunay filtration and the offset filtration are (functorially) homotopy equivalent. Other common choices are the *Čech-filtration* and the the *Vietoris-Rips filtration* with vertex set S; see Examples 2.1.7 and 2.1.8.

3. Computing the simplicial homology $H_{\bullet}(K_r)$ (with coefficients in a field k) for every r yields a collection $H_{\bullet}(K_*)$ of finite dimensional vector spaces $H_{\bullet}(K_r)$ for every r, connected by the morphisms $H_{\bullet}(K_r) \to H_{\bullet}(K_{r'})$ induced by $K_r \hookrightarrow K_{r'}$ for all $r \leq r'$. This collection is called the *persistent homology* of K_* .



Figure 1.1: Offset filtration (blue) $\bigcup_{s \in S} B_r(s)$, of a point cloud S (black) at different values of r, superimposed with the Delaunay filtration (black, light blue).

Figure 1.2: Persistence diagram of the Delaunay filtration K_* from Figure 1.1. Each point (x, y) in the diagram stands for a homology class such that for each r, the classes with $x \le r < y$ form a basis of $H_{\bullet}(K_r)$. There is a single point for $H_1(K_*)$ that is far from the diagonal. This corresponds to the homology of the annulus the points are sampled from.

Figure 1.3: Point cloud obtained from adding three outliers to Figure 1.1 (left) and persistence diagram of the resulting Delaunay filtration (right). The actual homology of the sampled annulus cannot be seen from in the diagram anymore.



4. The persistent homology $H_{\bullet}(K_*)$ can be compactly described by its *barcode* [76, 128], or by its *persistence diagram*, see Figure 1.2. Each point (x, y) of the persistence diagram represents a cycle in K_z for all $z \ge x$. It is homologous to older cycles if $z \ge y$. Thus, for each z, the points (x, y) of the diagram with $x \le z < y$ represent a basis of $H_z(K_*)$. The barcode can be easily computed by a Gaussian column reduction scheme [128], and efficient implementations are widely available [1, 8, 81, 111, 122].

If S is sampled from a subspace $X \subset Y$, then under certain conditions on S, one can infer the homology of X from the persistent homology of S [43, 48, 106]. In our example, we have $Y = \mathbf{R}^2$, $X \supset S$ is an annulus, and the long lived homology classes in Figure 1.2 correspond to the homology of X.

Particular strengths of this approach lie in the fact that it captures topological features of a space at different scales at the same time. If X is a submanifold of \mathbb{R}^n , then representatives of the homology classes can be used to reconstruct a triangulation of X [18, 49, 57]. A vectorized version of persistence diagrams can be used as input for other (e.g., machine learning) data analysis methods [92, 114, 115]. Furthermore, persistent homology is stable [16, 28, 40, 48], in the sense that small perturbations to the input lead to only small perturbations of the output, in a way that can be made precise.

Despite these stability results, however, traditional one-parameter persistent homology is susceptible to outliers. To make this precise, assume that $S \subset Y$ is a sample of a subspace $X \subset Y$ that also contains (few) outliers not in X. Already for relatively small values of $|S \setminus X|/|S|$, the homology $H_{\bullet}(X)$ may become unintelligible from $H_{\bullet}(S^{(r)})$; see Figure 1.3. Additionally, correctly capturing the topology of X from S is difficult if X has topological features of different scales.

Multi-parameter persistent homology [34, 91] is seen as a possible remedy for the above shortcomings. It extends (one-parameter) persistent homology by introducing additional parameters that control, for instance, the density of the sampled points. A common problem is to compute a *minimal free resolution* of $H_{\bullet}(K_*)$, which generalizes the construction of the barcode to more than one parameter.

Although, contrary to the situation in one-parameter persistence, the indecomposable multiparameter persistence modules cannot be classified as easily as in the one-parameter case [34, 74] and give no stable invariant [20], many other invariants, such as graded Betti numbers [90, 99], the (generalized) rank invariant [56, 89], signed [27] and fibered barcodes [100], shift dimension [37] and others, can be computed from a minimal free resolution, which motivates the interest in computing these. An example for such a minimal free resolution is shown in Figure 1.4. The



Figure 1.4: The point cloud from Figure 1.3, equipped with a density function. The right two plots show the Hilbert function (blue) and the graded Betti numbers (teal: β_0 , red: β_1 , orange: β_2) of the associated function-Rips homology. See Figure 2.6a for details. A horizontal slice at density value ρ through the Betti diagram corresponds to the one-parameter persistent homology of the subsample having only points with density value at most ρ . For example, a horizontal slice through the diagram for H_1 at $\rho \approx 70$ would produce one long bar in the barcode, corresponding to the region in the parameter space for which H_1 is one-dimensional. This is precisely the long bar in the barcode of H_1 one would see if one removes the points with density greater than 70 from the pointcloud (which comprises also the outliers inside the circle).

picture shows the Hilbert function and graded Betti numbers of the function-Rips persistent homology of the same point cloud as above.

For one- and two-parameter persistence, the computational worst-case complexity for computing the persistent homology of a (one-critical) filtered simplicial complex is cubic in the number of simplices. In practice, however, computing the minimal free resolution of two-parameter persistent homology is much more difficult than computing the barcode of a one-parameter filtered complex [33, 73, 85, 86, 98]. For more than two parameters, no method with cubic complexity is known.

Since the first algorithm for one-parameter persistent homology [128] has been described, substantial performance improvements have been achieved, some of which rely on computing (relative) persistent cohomology instead of (absolute) persistent homology. A duality principle allows to relate barcodes of absolute and relative homology and cohomology [19, 55]. It has been unclear so far, however, how this approach can be generalized beyond one-parameter persistent homology.

Contribution

Summary In this thesis, we explore different duality principles extending the known dualities in one-parameter persistence [55] to two- and multi-parameter persistent (co)homology. We present different ways to compute a minimal free resolutions of the persistent cohomology of a finite, one-critically two-parameter-filtered simplicial complex, and explain how this can be related to minimal free resolutions of its persistent homology. We devise an algorithm that allows for an optimization scheme similar to *clearing* in one-parameter persistence. A C++-implementation of our algorithm is publicly available [94]. Experiments demonstrate the practicability of our approach.

Cohomology computation Let K_* be a finite *n*-parameter filtered simplicial complex that is *one-critical*; that is, every simplex enters the filtration at a unique minimal value. Let $K = \operatorname{colim}_{z \in \mathbb{Z}^n} K_z$. In the following, when talking about (co)homology, we always mean reduced (co)homology with coefficients in a fixed field k.

$$C_{\bullet}(K_{*}) = \bigoplus_{\sigma \in K_{*}} \xrightarrow{0} Z \qquad C^{\bullet}(K_{*}) = \bigoplus_{\sigma \in K_{*}} \xrightarrow{0} Z \qquad C^{\bullet}(K_{*}) = \bigoplus_{\sigma \in K_{*}} \xrightarrow{0} Z \qquad Z^{\bullet}(K_{*}) = \bigcup_{\sigma \in K_{*}} \xrightarrow{0} Z \qquad Z^{\bullet}(K, K_{*}) = \bigcup_{\sigma \in K_{*}} \xrightarrow{0} Z \qquad Z^{\bullet}(K, K_{*}) = \bigcup_{\sigma \in K_{*}} \xrightarrow{0} Z \qquad Z^{\bullet}(K, K_{*}) = \bigcup_{\sigma \in K_{*}} \xrightarrow{0} Z \qquad Z^{\bullet}(K, K_{*}) = \bigcup_{\sigma \in K_{*}} \xrightarrow{0} Z \qquad Z^{\bullet}(K, K_{*}) = \bigcup_{\sigma \in K_{*}} \xrightarrow{0} Z \qquad Z^{\bullet}(K, K_{*}) = \bigcup_{\sigma \in K_{*}} \xrightarrow{0} Z \qquad Z^{\bullet}(K, K_{*}) = \bigcup_{\sigma \in K_{*}} \xrightarrow{0} Z \qquad Z^{\bullet}(K, K_{*}) = \bigcup_{\sigma \in K_{*}} \xrightarrow{0} Z \qquad Z^{\bullet}(K, K_{*}) = \bigcup_{\sigma \in K_{*}} \xrightarrow{0} Z \qquad Z^{\bullet}(K, K_{*}) = \bigcup_{\sigma \in K_{*}} \xrightarrow{0} Z \qquad Z^{\bullet}(K, K_{*}) = \bigcup_{\sigma \in K_{*}} \xrightarrow{0} Z \qquad Z^{\bullet}(K, K_{*}) = \bigcup_{\sigma \in K_{*}} \xrightarrow{0} Z \qquad Z^{\bullet}(K, K_{*}) = \bigcup_{\sigma \in K_{*}} \xrightarrow{0} Z \qquad Z^{\bullet}(K, K_{*}) = \bigcup_{\sigma \in K_{*}} \xrightarrow{0} Z \qquad Z^{\bullet}(K, K_{*}) = \bigcup_{\sigma \in K_{*}} \xrightarrow{0} Z \qquad Z^{\bullet}(K, K_{*}) = \bigcup_{\sigma \in K_{*}} \xrightarrow{0} Z \qquad Z^{\bullet}(K, K_{*}) = \bigcup_{\sigma \in K_{*}} \xrightarrow{0} Z \qquad Z^{\bullet}(K, K_{*}) = \bigcup_{\sigma \in K_{*}} \xrightarrow{0} Z \qquad Z^{\bullet}(K, K_{*}) = \bigcup_{\sigma \in K_{*}} \xrightarrow{0} Z \qquad Z^{\bullet}(K, K_{*}) = \bigcup_{\sigma \in K_{*}} \xrightarrow{0} Z \qquad Z^{\bullet}(K, K_{*}) = \bigcup_{\sigma \in K_{*}} \xrightarrow{0} Z \qquad Z^{\bullet}(K, K_{*}) = \bigcup_{\sigma \in K_{*}} \xrightarrow{0} Z \qquad Z^{\bullet}(K, K_{*}) = \bigcup_{\sigma \in K_{*}} \xrightarrow{0} Z \qquad Z^{\bullet}(K, K_{*}) = \bigcup_{\sigma \in K_{*}} \xrightarrow{0} Z \qquad Z^{\bullet}(K, K_{*}) = \bigcup_{\sigma \in K_{*}} \xrightarrow{0} Z \qquad Z^{\bullet}(K, K_{*}) = \bigcup_{\sigma \in K_{*}} \xrightarrow{0} Z \qquad Z^{\bullet}(K, K_{*}) = \bigcup_{\sigma \in K_{*}} \xrightarrow{0} Z \qquad Z^{\bullet}(K, K_{*}) = \bigcup_{\sigma \in K_{*}} \xrightarrow{0} Z \qquad Z^{\bullet}(K, K_{*}) = \bigcup_{\sigma \in K_{*}} \xrightarrow{0} Z \qquad Z^{\bullet}(K, K_{*}) = \bigcup_{\sigma \in K_{*}} \xrightarrow{0} Z \qquad Z^{\bullet}(K, K, K_{*}) = \bigcup_{\sigma \in K_{*}} \xrightarrow{0} Z \qquad Z^{\bullet}(K, K, K) = \bigcup_{\sigma \in K_{*}} \xrightarrow{0} Z \qquad Z^{\bullet}(K, K) = \bigcup_{\sigma \in K_{*}} \xrightarrow{0} Z \qquad Z^{\bullet}(K, K) = \bigcup_{\sigma \in K_{*}} \xrightarrow{0} Z \qquad Z^{\bullet}(K, K) = \bigcup_{\sigma \in K_{*}} \xrightarrow{0} Z \qquad Z^{\bullet}(K, K) = \bigcup_{\sigma \in K_{*}} \xrightarrow{0} Z \qquad Z^{\bullet}(K, K) = \bigcup_{\sigma \in K_{*}} \xrightarrow{0} Z \qquad Z^{\bullet}(K, K) = \bigcup_{\sigma \in K_{*}} \xrightarrow{0} Z \qquad Z^{\bullet}(K, K) = \bigcup_{\sigma \in K_{*}} \xrightarrow{0} Z \qquad Z^{\bullet}(K, K) = \bigcup_{\sigma \in K_{*}} \xrightarrow{0} Z \qquad Z^{\bullet}(K, K) = \bigcup_{\sigma \in K_{*}} \xrightarrow{0} Z \qquad Z^{\bullet}(K, K) = \bigcup_{\sigma \in K_{*}} \xrightarrow{0} Z \qquad Z^{\bullet}(K, K) = \bigcup_{\sigma \in K_{*}} \xrightarrow{0} Z \qquad Z^{\bullet}(K, K) = \bigcup_{\sigma \in K_{*}} \xrightarrow{0} Z \qquad Z^{\bullet}(K, K) = \bigcup_{\sigma \in K_{*}} \xrightarrow{0} Z \qquad Z^{\bullet}(K, K) = \bigcup_{\sigma \in K_{*}} \xrightarrow{0} Z \qquad Z^{\bullet}(K, K) = \bigcup_{\sigma \in K_{*}} \xrightarrow{0} Z \qquad Z^{\bullet}(K, K) = \bigcup_{\sigma \in K_$$

Figure 1.5: Absolute and relative simplicial (co)chains of a Z-filtered simplicial complex K_* . A persistence module is free if it is a direct sum of interval modules supported on intervals of the form $[b, \infty)$ for b in Z. In particular, $C_{\bullet}(K_*)$ and $C^{\bullet}(K, K_*)$ are (co)chain complexes of free modules, where the red line indicate the support of a simplex $\sigma \in K_*$ in the respective (co)chain complexes.

The main challenge in computing the absolute and relative cohomology of K_* lies in the fact that in multi-parameter persistence, $C^{\bullet}(K_*)$ and $C^{\bullet}(K, K_*)$ are no cochain complexes of free modules. This is different from one-parameter persistence, where $C^{\bullet}(K, K_*)$ is a cochain complex of free modules; see Figures 1.5 and 1.6 for an illustration. Working with (co)chain complexes of free modules is necessary to ensure applicability of the common column reduction schemes to compute persistent (co)homology. In this thesis, we propose two different ways to remedy this, which we summarize in the following.

First approach As said above, in one-parameter persistence, $C^{\bullet}(K, K_*)$ is a cochain complex of free modules. The approach presented in Chapter 3 generalizes this to the multi-parameter case in the following way. We define a certain cochain complex $N^{\bullet}(K_*) := (\nu C_{\bullet}(K_*))^*$, where $(-)^*$ denotes the pointwise dual persistence module and ν the Nakayama functor; see Definitions 2.1.10, 3.1.2 and 3.2.2. In the one-parameter case, we have $C^{\bullet}(K, K_*) = (\nu C_{\bullet}(K_*))^*$. In persistence over more than one parameter, $C^{\bullet}(K, K_*)$ is not a cochain complex of free modules; however, $N^{\bullet}(K_*)$ is. In Chapter 3, we work with the latter complex and show:

Theorem A (page 42). Let K_* be a one-critically \mathbb{Z}^n -filtered simplicial complex, such that $H_{\bullet}(K_*)$ is finitely supported. Then there is a natural isomorphism

$$H^d(K_*) \cong H^{d+n}(N^{\bullet}(K_*))$$

for all d.

Here, we say that a persistence module M is finitely supported if its components M_z are zero for all but finitely many $z \in \mathbb{Z}^n$. This theorem is a generalization of the fact that if K is acyclic, then $H_d(K_*) \cong H^{d+1}(K, K_*)^*$, which can be seen from the long exact sequence of the pair (K, K_*) . If $H_*(K_*)$ is not finitely supported, one can replace $C_{\bullet}(K_*)$ by a complex \hat{C}_* finitely supported homology, which can be used to compute $H^d(K_*)$; see Section 3.7. Furthermore, for any number of parameters, the cohomology of $N^{\bullet}(K_*)$ can be related to $H^{\bullet}(K_*)$ and $H^{\bullet}(K, K_*)$ (see Theorem A and Corollary 3.2.13).

Second approach The second approach, presented in Chapter 4, is to replace $C^{\bullet}(K_*)$ and $C^{\bullet}(K, K_*)$ by quasi-isomorphic cochain complexes of free modules. We do this by choosing free resolutions of $C^d(K_*)$ (resp. $C^d(K, K_*)$) for each d and deriving an explicit formula for a minimal free resolution of $H^d(K_*)$ (resp. $H^d(K, K_*)$) using these resolutions.

Necessity of the finite-support condition In one-parameter persistence, it is known that $H^{\bullet}(K_*)$ and $H^{\bullet}(K, K_*)$ uniquely determine each other up to isomorphism even if $H^{\bullet}(K) \neq 0$ (see Corollary 2.2.11). In Sections 3.6 and 4.6, we show by counterexamples that corresponding statements do not hold in two-parameter persistence, namely:

Theorem B (pages 55 and 98). Let K_* be a one-critically 2-parameter filtered simplicial complex.

(a) Unless $H_{\bullet}(K_*)$ is finitely supported, $H^{\bullet}(K_*)$ and $H^{\bullet}(N^{\bullet}(K_*))$ need not determine each other up to isomorphism.



Figure 1.6: Absolute and relative simplicial (co)chains of a Z^2 -filtered simplicial complex K_* . A module is free if it is a direct sum of modules F(z), each of which has components $F(z)_w = \begin{cases} k & \text{if } w \ge z, \\ 0 & \text{otherwise} \end{cases}$ i.e., the module of the upper right quadrant with minimal element z. In particular, if K_* is one-critically Z^2 -filtered, then $C_{\bullet}(K_*)$ is a chain complex of free modules. None of $C_{\bullet}(K, K_*)$, $C^{\bullet}(K_*)$, $C^{\bullet}(K, K_*)$ is a (co)chain complex of free modules. This is different from 1-parameter persistence; see Figure 1.5.

(b) Unless H_●(K) = 0, H[●](K_{*}) and H[●](K, K_{*}) need not determine each other up to isomorphism.

Higher coboundary morphisms In one-parameter persistent homology, one can show that if $H^{\bullet}(K) = 0$, then $H^{d+1}(K, K_*)$ can be computed from the (matrix representing the) coboundary morphism

$$\delta^{d+1}: C^d(K, K_*) \to C^{d+1}(K, K_*)$$

alone. That is, it is not necessary to consider the coboundary morphism δ^{d+2} , because the assumption ensures that all relevant information is already contained in δ^{d+1} ; see Section 2.2.1. This is of practical importance, because for Vietoris–Rips complexes, a matrix representing δ^{d+2} would be prohibitively large. By providing explicit algorithms, we show the following analogue statements for two-parameter persistence:

Theorem C (page 54). If K_* is a finite two-parameter filtered complex such that $H_{\bullet}(K_*)$ is finitely supported, then a minimal free resolution of $H^{d+2}(N^{\bullet}(K_*))$ can be computed from the coboundary morphism $N^d(K_*) \to N^{d+1}(K_*)$ alone.

Theorem D (page 90). If K_* is a finite two-parameter filtered complex such that $H_{\bullet}(K) = 0$, where $K = \operatorname{colim}_z K_z$, then a minimal free resolution $H^{d+1}(K, K_*)_{\bullet}$ of $H^{d+1}(K, K_*)$ can be computed from the coboundary morphism $C^d(K, K_*) \to C^{d+1}(K, K_*)$ alone.

Duality and free resolutions So far, we have concentrated on computing a minimal free resolution of $H^{d}(K_*)$, either directly, or by computing a minimal free resolution of $H^{d+1}(K_*)$ or $H^{d+n}(N^{\bullet}(K_*))$ and showing that these cohomology modules are isomorphic $H^{d}(K_*)$ under certain conditions. In applications, however, we might be interested in a free resolution of persistent homology $H_d(K_*)$. We show that if $H^d(K_*)$ has finite total dimension, then there is a simple correspondence between minimal free resolutions of $H^d(K_*)$ and $H_d(K_*)$. More generally, we show:

Theorem E (page 48). Let M be a finitely generated n-parameter persistence module with bounded support. For graded matrices U_1, \ldots, U_n , the following are equivalent:

- (i) U_1, \ldots, U_n represent a free resolution of module M,
- (ii) U_1, \ldots, U_n represent an injective resolution of the shifted module $M\langle -\epsilon \rangle$,
- (iii) the graded transposes $U_1^{\top}, \ldots, U_n^{\top}$ represent a free resolution of the dual module $M\langle -\epsilon \rangle^*$.

Computation Combining Theorems A and E allows us to obtain a minimal free resolution of $H_d(K_*)$ for a two-parameter filtered complex K_* in the following steps:

- 1. If using b) or c) in the following, ensure before that K_* has finitely supported homology, e.g. by replacing K_* by a suitable complex with finitely supported homology as described in Section 3.7.
- 2. Compute a free resolution of either
 - a) $H^{d}(K_{*})$ (see Section 4.2.1);
 - b) $H^{d+1}(K, K_*)$ (see Section 4.2.2) and obtain a free resolution of $H^d(K_*)$ by the long exact sequence of cohomology; or
 - c) $H^{d+2}(N^{\bullet}(K_*))$ (see Chapter 3) and obtain a free resolution of $H^d(K_*)$ by Theorem A.
- 3. From that, obtain a free resolution of $H_d(K_*)$, using Theorem E.

Only the second step is computationally involved. Without loss of generality, assume that K_* has finitely supported homology. We propose strategies algorithms to compute minimal free resolutions of $H^{d+2}(N^{\bullet}(K_*))$, $H^d(K_*)$ and $H^{d+1}(K, K_*)$ in the two-parameter case, see Sections 3.5, 4.2 and 4.4.

Pulling back from the colimit For $H^{d+2}(N^{\bullet}(K_*))$, the core of the proposed algorithm (see Algorithm 11) is the following insight, which also underlies the proofs of Theorems C and D. For a persistence module M, we define the vector space colim $M := \operatorname{colim}_{z \in \mathbb{Z}^n} M_z$.

Theorem F (page 50). Let M, N be free persistence modules and $f: M \to N$ be a morphism, and let U be the unique maximal submodule of M with colim U = colim ker f. Then U = ker f.

For a general submodule $V \subseteq M$, it is not true that for the unique maximal submodule $U \subseteq M$ with $\operatorname{colim} U = \operatorname{colim} V$, the inclusion $U \subseteq V$ is an equality, even if U and/or M are free. The lemma shows that this is the case for kernels indeed.

If M and N are free two-parameter modules, then also ker f is free, which can be derived from Hilbert's syzygy theorem; see Corollary 2.3.20. We propose an algorithm (see Theorem 3.4.7) that computes a basis of ker f, given basis of colim ker f. Using this, we show how to compute $H^{d+1}(K, K_*)$ and $H^{d+2}(N^{\bullet}(K_*))$; see Algorithms 11 and 13. Both algorithms allow for a optimization that is analogous to the *clearing* scheme in one-parameter persistence [11, 46]; see Remark 3.5.3.

Implementation and experiments Earlier experiments that we will not report here have shown that that the approach using the complex $N^{\bullet}(K_*)$ from Chapter 3 is much more efficient than working with free resolutions of $C^{\bullet}(K_*)$ or $C^{\bullet}(K, K_*)$ as described in Chapter 4; see also Section 4.5. For this reason, for the implementation and runtime experiments that we describe in Chapter 5, we concentrate on the former approach. The implementation is open source and publicly available [94]. This implementation was also used to produce the diagram in Figure 1.4.

We use our implementation to compute minimal free resolutions of $H_{\bullet}(K_*)$ for density-Rips complexes and other two-parameter filtered complexes of different sizes; see Section 5.2. We compare our approach to the state-of-the-art algorithm from [73], both with respect to run time and memory footprint. We analyze the effect of the well-known chunk preprocessing [72] and implementation details (such as the sparse matrix representation employed) on the performance.

The experiments show that our approach is able to outperform the method from [73], both with respect to run time and memory. This is true both for the heap and the vector format [12] for sparse matrices. When computing $H_d(K_*)$ for $d \ge 2$, our algorithm was faster on almost all instances, by a factor of often up to 15. There were instances (for instances the cyclooctane data set or points clouds samples orthogonal groups, see Section 5.2) that were tractable only with the cohomology algorithm, while the homology algorithm did not terminate within acceptable time. Our algorithm can be seen as an alternative to chunk preprocessing, since both serve to efficiently remove trivial summands from the input complex. Consequently, while the algorithm from [73] is known to strongly benefit from chunk preprocessing, our algorithm does not require chunk preprocessing to be efficient. Independently of our cohomology algorithm, we propose a cohomological version of chunk preprocessing for the homology algorithm [73] that we call chunk* preprocessing.

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Chapter 2

Background

In this section, we give an overview of the relevant preliminaries of single and multi-parameter persistent homology. We introduce persistence modules, both as functors over a poset and as graded modules over an algebra, and explain how they arise from data. We explain the algorithmic details of the computation of one-parameter persistent homology and cohomology, and explain how minimal free resolutions of two-parameter persistent homology are computed.

2.1 Persistence modules

Let k be a field, and let Vec (respectively, vec) denote the category of k-vector spaces (resp., finite dimensional k-vector spaces). Let P be a poset, viewed as a category with one object p for every $p \in P$ and precisely one morphism from p to q for each $p \leq q$ in P.

Definition 2.1.1. A *P*-persistence module is a functor $M: P \to \text{Vec}$ that assigns to every $p \in P$ a k-vector space M_p , and to every pair $p \leq q$ of comparable elements in P a morphism $M_{qp}: M_p \to M_q$. The vector spaces M_p are called the *components* of M, and he maps M_{qp} are called the *structure maps* of M. For $m \in M_p$, we denote by $g(m) \coloneqq p$ the grade of m. The module M is called *pointwise finite dimensional* if M_p is finite dimensional for all $p \in P$. The support of M is supp $M \coloneqq \{p \in P \mid M_p \neq 0\}$. A morphism of persistence modules is a natural transformation of functors. We write Vec^P (resp., vec^P) for the category of P-persistence modules (resp., pointwise finite dimensional modules).

If M and N are persistence modules, then $M \oplus N$ is the persistence module with $(M \oplus N)_p = M_p \oplus N_p$. If $f: M \to N$ is a morphism of persistence modules, then ker f and coker f are the persistence modules with $(\ker f)_p = \ker f_p$ and $(\operatorname{coker} f)_p = \operatorname{coker} f_p$. These render Vec^P and vec^P an abelian category. For example,

$$0 \rightarrow \begin{pmatrix} 0 \rightarrow 0 \rightarrow k \\ \uparrow & \uparrow & \uparrow \\ 0 \rightarrow 0 \rightarrow 0 \\ \uparrow & \uparrow & \uparrow \\ 0 \rightarrow 0 \rightarrow 0 \end{pmatrix} \rightarrow \begin{pmatrix} k \xrightarrow{\begin{pmatrix} 1 \\ 0 \\ 0 \\ 1 \\ 1 \end{pmatrix}} k^2 \xrightarrow{\downarrow 4} k^3 \\ \uparrow & \begin{pmatrix} 1 \\ 0 \\ 0 \\ 1 \\ 1 \end{pmatrix} \uparrow \begin{pmatrix} 0 \\ 0 \\ 0 \\ 1 \\ 0 \\ 1 \end{pmatrix} \uparrow \begin{pmatrix} 0 \\ 0 \\ 0 \\ 1 \\ 0 \\ 1 \end{pmatrix} \uparrow \begin{pmatrix} 0 \\ 0 \\ 0 \\ 1 \\ 0 \\ 1 \end{pmatrix} \uparrow \begin{pmatrix} 0 \\ 0 \\ 0 \\ 1 \\ 0 \\ 1 \end{pmatrix} \end{pmatrix} \rightarrow \begin{pmatrix} k \xrightarrow{\begin{pmatrix} 1 \\ 0 \\ 1 \\ 0 \\ 1 \\ 1 \end{pmatrix} \uparrow \\ \begin{pmatrix} 1 \\ 0 \\ 1 \\ 1 \end{pmatrix} \uparrow \\ \begin{pmatrix} 1 \\ 0 \\ 1 \\ 1 \end{pmatrix} \uparrow \\ \begin{pmatrix} 1 \\ 0 \\ 1 \\ 1 \end{pmatrix} \uparrow \\ \begin{pmatrix} 1 \\ 0 \\ 1 \\ 1 \end{pmatrix} \uparrow \\ \begin{pmatrix} 1 \\ 0 \\ 1 \\ 1 \end{pmatrix} \uparrow \\ \begin{pmatrix} 1 \\ 0 \\ 1 \\ 1 \end{pmatrix} \uparrow \\ \begin{pmatrix} 1 \\ 0 \\ 1 \\ 1 \end{pmatrix} \uparrow \\ \begin{pmatrix} 1 \\ 0 \\ 1 \\ 1 \end{pmatrix} \uparrow \\ \begin{pmatrix} 1 \\ 0 \\ 1 \\ 1 \end{pmatrix} \uparrow \\ \begin{pmatrix} 1 \\ 0 \\ 1 \\ 1 \end{pmatrix} \uparrow \\ \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix} \end{pmatrix} \rightarrow 0$$

is a short exact sequence of modules over the poset $\{0, 1, 2\}^2$.

2.1.1 Filtered spaces and complexes

Typically, persistence modules arise as the homology of filtered topological spaces or filtered simplicial complexes. We write Top for the category of topological spaces.

Definition 2.1.2. A *P*-filtered space is a functor $X_* \colon P \to \text{Top}, p \mapsto X_p$, such that $X_{qp} \colon X_p \to X_q$ is an inclusion for all $p \leq q$ in *P*. We write $\text{Top}^{\subseteq P}$ for the category of *P*-filtered topological spaces.

Example 2.1.3 (Sublevel sets). Let X be a topological space, equipped with a function $f: X \to P$. The sublevel set filtration $X_{f\leq *}$ is the P-filtered topological subspace of X with $X_{f\leq p} = f^{-1}(\{q \in P \mid q \leq p\})$.

Example 2.1.4 (Offset filtration [79]). The offset filtration $S^{(*)}$ of a subset S of a metric space X is the **R**-filtered subspace of X with $S^{(r)} := \bigcup_{s \in S} B_r(x)$. Here $B_r(s)$ denotes the open ball of radius r centered at s. If $M \subseteq \mathbf{R}^n$ is a submanifold and $S \subset M$ a finite sample, then under certain conditions, the homology of M can be obtained from $M^{(r)}$ and $S^{(r)}$ [61, p. 12]. See also [48].

For computational aspects, it is more convenient to work with a discrete model of topological spaces, namely, filtered simplicial complexes.

Definition 2.1.5. An (abstract) simplicial complex K is a set of nonempty finite sets such that if $\sigma \in K$ and $\rho \subseteq \sigma$ is non-empty, then $\rho \in K$. For any $d \ge 0$, the elements of $K^d = \{\sigma \in K \mid |\sigma| = d + 1\}$ are called the *d*-simplices of K. The 0-simplices are called the *vertices* of K. A morphism $f: K \to L$ of simplicial complexes is a morphism $K \to L$ of sets such that $f(K^0) \subseteq L^0$, and $f(\sigma) = \bigcup_{v \in \sigma} f(\{v\})$. We denote by Simp the category of simplicial complexes. A *P*-filtered simplicial complex is a functor $K_*: P \to \text{Simp}, p \mapsto K_p$ such that $K_{qp}: K_p \to K_q$ is injective for all $p \le q$. If K_* is a filtered simplicial complex, we write $K := \text{colim}_{p \in P} K_p$. We call K_* finite if K is finite. We write $\text{Simp}^{\subseteq P}$ for the category of *P*-filtered simplicial complexes.

Example 2.1.6 (Delaunay complex [60]). Let S be a finite subset of a metric space X. For $s \in S$, let $\operatorname{Vor}(s) = \{x \in X \mid \forall t \in S : d(x,t) \ge d(x,s)\}$ be the Voronoi domain of s with respect to S, and let $\operatorname{Vor}_r(s) = \operatorname{Vor}(s) \cap B_r(s)$. Then the Delaunay filtration or α -filtration $\operatorname{Del}_*(S)$ is the **R**-filtered simplicial complex with $\operatorname{Del}_r(S) = \{\sigma \subseteq S \mid 0 < |\sigma| < \infty$ and $\bigcap_{s \in \sigma} \operatorname{Vor}_r(s) \neq \emptyset\}$; see Figure 1.1.

Example 2.1.7 (Čech complex). Let S be a subset of a metric space X. The Čech complex $\check{C}_*(S)$ is the **R**-filtered abstract simplicial complex given by $\check{C}_r(S) = \{\sigma \subseteq S \mid 0 < |\sigma| < \infty \text{ and } \bigcap_{s \in \sigma} B_r(s) \neq \emptyset\}.$

See Figures 1.1 and 2.1 for examples. The Delaunay complex $\text{Del}_r(S)$ is the nerve of the closed cover $\{\text{Vor}_r(s) \mid s \in S\}$ of the offset space $S^{(r)}$. The Čech complex $\check{C}_r(S)$ is the nerve of the closed cover $\{B_r(s) \mid s \in S\}$ of $S^{(r)}$. Under suitable conditions (for example, if X is Euclidean), the functorial nerve theorem [13] guarantees that $S^{(r)}$ and the geometric realizations $|\text{Del}_r(S)|$ and $|\check{C}_r(S)|$ are homotopy equivalent to $S^{(r)}$ in an **R**-filtered way; see also [10]. Therefore, both can be seen as a combinatorial model of the filtered space $r \mapsto S^{(r)}$. However, both complex are computationally involved in high dimensions or in non-Euclidean spaces. A common alternative is the following:

Example 2.1.8 (Vietoris–Rips complex [125]). Let S be a metric space. For $\sigma \subseteq S$, let diam $\sigma = \max_{s,t\in\sigma} d(s,t)$. The Vietoris–Rips complex $VR_*(S)$ associated to S is the **R**-filtered abstract simplicial complex given by $VR_r(S) = \{\sigma \subseteq S \mid 0 < |\sigma| < \infty \text{ and } \dim \sigma \leq r\}.$

The Vietoris–Rips complex is a *clique complex*; meaning that for d > 2, a *d*-simplex $\sigma \in VR_r(S)$ if $\rho \in VR_r(S)$ for all its 2-faces $\rho \in \binom{\sigma}{2}$. See Figure 2.2 for an example. For $S \subseteq \mathbf{R}^n$, the Vietoris–Rips complex relates to the Čech complex via mutual inclusions $VR_r(S) \subseteq \check{C}_{r'/2}(S) \subseteq VR_{r'}(S)$ whenever $\frac{r'}{r} \geq \sqrt{2n/(n+1)}$ [54, Theorem 2.5, 4, 41]. If *n* is large, it is considerably easier to decide if a set $\sigma \subset S$ lies in $VR_r(S)$ than to decide of σ lies in $Del_r(S)$ or $\check{C}_r(S)$, which motivates the interest in VR(S).



Figure 2.1: Offset filtration (blue) $\bigcup_{s \in S} B_r(s)$, of the point cloud from Figure 1.1a, superimposed with a projection of the 2-skeleton of the *Čech filtration* (black, light blue).



Figure 2.2: Like Figure 2.1, but with the Vietoris-Rips filtration VR_{2r}.

If S is finite, let $r_1 < r_2 < \cdots < r_n$ be the distinct values in $\{\operatorname{diam} \sigma \mid \sigma \subseteq S, \sigma \neq \emptyset\}$. By setting

$$K_{i}(S) = \begin{cases} VR(S)_{r_{1}} & \text{if } i \leq 1, \\ VR(S)_{r_{i}} & \text{if } 1 \leq i \leq n, \\ VR(S)_{r_{n}} & \text{if } n \leq i, \end{cases}$$
(2.1)

we obtain a **Z**-filtered simplicial complex K_* that we also call the Vietoris–Rips complex of S and also denote by $VR_*(S)$.

2.1.2 Persistent homology

Since Vec^{P} is an abelian category, it makes sense to consider chain complexes of persistence modules. A *chain complex* C_{\bullet} is a sequence

$$\cdots \to C_{d+1} \xrightarrow{\partial_{d+1}} C_d \xrightarrow{\partial_d} C_{d-1} \to \cdots$$

of modules and morphisms, such that $\partial_d \partial_{d+1} = 0$ for all d. Its d-cycles $Z_d(C_{\bullet}) := \ker \partial_d$, d-boundaries $B_d(C_{\bullet}) := \operatorname{im} \partial_{d+1}$ and d-homology $H_d(C_{\bullet}) := Z_d(C_{\bullet})/B_d(C_{\bullet})$ are persistence modules. Dually, a cochain complex C^{\bullet} is a sequence

$$\cdots \to C^{d-1} \xrightarrow{\delta^d} C^d \xrightarrow{\delta^{d+1}} C^{d+1} \to \cdots$$

of modules and morphisms, such that $\delta^{d+1}\delta^d = 0$ for all d. Its d-cocycles $Z^d(C^{\bullet}) := \ker \delta^{d+1}$, d-boundaries $B^d(C^{\bullet}) := \operatorname{im} \delta^d$ and d-cohomology $H^d(C^{\bullet}) := Z^d(C^{\bullet})/B^d(C^{\bullet})$ are persistence modules. We say that a (co)chain complex C_{\bullet} (resp., C^{\bullet}) has finite (total) dimension if $\bigoplus_d C_d$ (resp. $\bigoplus_d C^d$) is finite dimensional. A (co)chain complex is acyclic if its (co)homology is zero in all dimensions. (Co)chain complexes of persistence modules form an abelian category.

Remark. We silently require that all (co)chain complexes we work with are bounded below.

We assume that P has an involution $P \to P, p \mapsto -p$ that identifies P with its opposite poset P^{op} ; that is, $p \leq q$ if and only if $-q \leq -p$. We also assume that P is a lattice; that is, each $p, p' \in P$ have a unique *least upper bound* or *join*, denoted by $p \vee p'$, and a unique greatest lower bound or meet, denoted by $p \wedge p'$.

Example 2.1.9. The sets \mathbf{Z}^n or \mathbf{R}^n are posets with the ordering $(z_1, \ldots, z_n) \leq (z'_1, \ldots, z'_n)$ if $z_i \leq z'_i$ for all *i*. Negation identifies both with their opposite poset. Both are lattices, where

$$(z_1, \dots, z_n) \lor (z'_1, \dots, z'_n) = (\max\{z_1, z'_1\}, \dots, \max\{z_n, z'_n\}), (z_1, \dots, z_n) \land (z'_1, \dots, z'_n) = (\min\{z_1, z'_1\}, \dots, \min\{z_n, z'_n\}).$$

Let the k-dual vector space of a k-vector space V be denoted by V^* .

Definition 2.1.10. The dual module M^* of a persistence module $M \in \text{Vec}^P$ is the *P*-persistence module with $(M^*)_p = (M_{-p})^*$ and $(M^*)_{qp} = (M_{-p,-q})^*$.

Let C_{\bullet} be a chain complex, then its dual complex $C^{\bullet} \coloneqq (C_{\bullet})^*$ is the cochain complex with $C^d = (C_d)^*$ and $\delta^d = (\partial_d)^*$. The functor $(-)^* \colon \operatorname{Vec}^P \to \operatorname{Vec}^P$ is exact contravariant. Therefore, there is a natural isomorphism

$$H^{\bullet}((C_{\bullet})^*) \cong H_{\bullet}(C_{\bullet})^*.$$

Definition 2.1.11. Let $K_* \in \text{Simp}^{\subseteq P}$. The *(absolute) persistent chain complex* of K_* is the chain complex $C_{\bullet}(K_*)$ of *P*-persistence modules, with components

$$C_d(K_*): P \to \text{Vec}, \ p \mapsto C_d(K_p)$$

where $C_{\bullet}(K_p)$ denotes the simplicial chain complex of the simplicial complex K_p . We denote its *d*-cycles, *d*-boundaries and *d*th homology by $Z_d(K_*)$, $B_d(K_*)$ and $H_d(K_*)$ The module $H_d(K_*)$ is called the *d*th (absolute) persistent homology of K_* .

We can see a persistence module $M \in \text{Vec}^P$ as a *P*-indexed diagram of vector spaces. In that sense, we define the functors

$$\begin{array}{cc} \operatorname{colim} \colon \operatorname{Vec}^P \to \operatorname{Vec}, & \Delta \colon \operatorname{Vec} \to \operatorname{Vec}^P, & \operatorname{lim} \colon \operatorname{Vec}^P \to \operatorname{Vec}, \\ & M \mapsto \operatorname{colim}_P M, & V \mapsto \left\{ \begin{smallmatrix} p & \mapsto V, \\ (p \leq q) & \mapsto \operatorname{id}_V \end{smallmatrix} \right\}, & M \mapsto \operatorname{lim}_P M. \end{array}$$

The functor Δ , called the *diagonal*, is right adjoint to colim and left adjoint to lim. For a vector space V, ΔV is the module that is constantly V.

Definition 2.1.12. Let $K_* \in \text{Simp}^{\subseteq P}$ and $K \coloneqq \bigcup K_*$. The relative persistent chain complex of K_* is the chain complex

$$C_{\bullet}(K, K_*) \coloneqq \frac{\Delta C_{\bullet}(K)}{C_{\bullet}(K_*)}$$

of persistence modules. We denote the relative *d*-cycles, *d*-boundaries and *d*th homology by $Z_d(K, K_*)$, $B_d(K, K_*)$ and $H_d(K, K_*)$, respectively.

Remark. For every $p \in P$, the absolute chain complex $C_{\bullet}(K_p)$ is spanned by the simplices $\sigma \in K_p$, while the relative chain complex $C_{\bullet}(K, K_p)$ is spanned by the simplices $\sigma \in K \setminus K_p$.

Definition 2.1.13. For $K_* \in \text{Simp}^{\subseteq P}$ and $K := \bigcup K_*$, we define the absolute persistent cochain complex $C^{\bullet}(K_*) := C_{\bullet}(K_*)^*$ and the relative persistent cochain complex $C^{\bullet}(K, K_*) := C_{\bullet}(K, K_*)^*$ of K_* . We denote the respective cocycles, coboundaries and cohomology by Z(-), $B^{\bullet}(-)$ and $H^{\bullet}(-)$. The modules $H^d(K_*)$ and $H^d(K, K_*)$ are called the *dth absolute* (resp., relative) persistent cohomology of K_* .

The inclusion $i: K_* \hookrightarrow K$ induces an inclusion $i_{\bullet}: C_{\bullet}(K_*) \to C_{\bullet}(K)$ whose cokernel is $C_{\bullet}(K, K_*)$, and a restriction morphism $i^{\bullet}: C^{\bullet}(K) \to C^{\bullet}(K_*), \gamma \mapsto \gamma|_{K_*}$ whose kernel equals $C^{\bullet}(K, K_*)$. There is a short exact sequence

$$0 \to C_{\bullet}(K_*) \xrightarrow{i_{\bullet}} \Delta C_{\bullet}(K) \xrightarrow{p_{\bullet}} C_{\bullet}(K, K_*) \to 0$$

that gives rise to a long exact sequence of P-persistence modules

$$\cdots \to H_{d+1}(K, K_*) \xrightarrow{\delta} H_d(K_*) \xrightarrow{i_d} \Delta H_d(K) \xrightarrow{p_d} H_d(K, K_*) \xrightarrow{\delta_d} H_{d-1}(K_*) \to \cdots .$$
(2.2)

Dually, the short exact sequence

$$0 \to C^{\bullet}(K, K_*) \xrightarrow{p^{\bullet}} \Delta C^{\bullet}(K) \xrightarrow{i^{\bullet}} C^{\bullet}(K_*) \to 0$$

gives rise to a long exact sequence

$$\cdots \to H^{d-1}(K_*) \xrightarrow{\delta^d} H^d(K, K_*) \xrightarrow{p^d} \Delta H^d(K) \xrightarrow{i^d} H^d(K_*) \xrightarrow{\delta^{d+1}} H^{d+1}(K, K_*) \to \cdots .$$
(2.3)

Remark 2.1.14. As mentioned above, we work with reduced (co)homology where not stated otherwise. In particular, if $H_d(K) = 0$ for all d, then the above exact sequences show that for all d, we have

$$H_d(K_*) \cong H_{d+1}(K, K_*), \qquad H^d(K_*) \cong H^{d+1}(K, K_*).$$

2.1.3 Free modules

Let M be a P-persistence module.

Definition 2.1.15. A (homogeneous) generating set of M is a collection $\{m_i \in M; i \in I'\}$ of elements of M, indexed by some indexing set I, such that for every $m \in M$, there exist coefficients $\lambda_i \in k$ for $i \in I$, such that only finitely many are non-zero and

$$m = \sum_{i \in I} \lambda_i M_{g(m),g(m_i)}(m_i)$$

The generating system is *minimal* if no proper subset of it is a generating system. It is called a *basis* of M if for all $p \in P$,

$$\sum_{i \in I} \lambda_i M_{p,g(m_i)}(m_i) = 0$$

implies that $\lambda_i = 0$ for all $i \in I$. A module is *finitely generated* if it has a finite generating system, and *free* if it has a basis.

A finitely generated module has a minimal generating system. If M is finitely generated (resp., free), then the multiset $\{g(m_i) | i \in I\}$ does not depend on the chosen minimal generating system (resp., basis) $\{m_i | i \in I\}$ of M. If M is free, we call the multiset $\{g(m_i) | i \in I\}$ the graded rank of M, denoted by rk M.

For every $p \in P$, the persistence module F(p) with

$$F(p)_q = \begin{cases} k \text{ if } p \le q, \\ 0 \text{ otherwise,} \end{cases} \qquad F(p)_{rq} = \begin{cases} \text{id if } p \le q \le r, \\ 0 \text{ otherwise.} \end{cases}$$

is free of rank rk $F(p) = \{p\}$. If $p_i \in P$ for all $i \in I$ for some indexing set i, then $F := \bigoplus_{i \in I} F(p_i)$ is free of rank $\{p_i \mid i \in I\}$. The standard basis of F is the basis $\{e_i \mid i \in I\}$, where the *i*th standard basis vector e_i is the element $e_i = 1 \in F(p_i)_{p_i}$ of the *i*th summand. There is a one-toone correspondence between generating systems (resp., bases) $\{m_i \mid i \in I\}$ of a module M and surjections (resp., isomorphisms)

$$\bigoplus_{i\in I} F(g(m_i)) \to M, e_i \mapsto m_i.$$

Example 2.1.16. The \mathbb{Z}^2 -persistence module on the left of

$$\begin{bmatrix} \vdots & \vdots & \vdots \\ \uparrow \begin{pmatrix} 1 & 0 \\ 0 \end{pmatrix} \begin{pmatrix} 1 & 0 \\ 0 \end{pmatrix} \uparrow \\ k \xrightarrow{(1)} & \uparrow \begin{pmatrix} 1 & 0 \\ 0 \end{pmatrix} \uparrow \\ \uparrow & (1) & \uparrow \\ 0 \xrightarrow{(1)} & \uparrow \begin{pmatrix} 1 & 0 \\ 0 \end{pmatrix} \end{pmatrix} \\ 0 \xrightarrow{(1)} & \uparrow \begin{pmatrix} 1 & 0 \\ 0 \end{pmatrix} \end{pmatrix} \\ 0 \xrightarrow{(1)} & \uparrow \begin{pmatrix} 1 & 0 \\ 0 \end{pmatrix} \end{pmatrix} \\ \cong \begin{bmatrix} \vdots & \vdots & \vdots \\ \parallel & \parallel \\ k \xrightarrow{(1)} & k \xrightarrow{(1)} & k \xrightarrow{(1)} \\ 0 \xrightarrow{(1)} & \uparrow \begin{pmatrix} 1 & 0 \\ 0 \end{pmatrix} \end{pmatrix} \\ 0 \xrightarrow{(1)} & \uparrow \begin{pmatrix} 1 & 0 \\ 0 \end{pmatrix} \end{pmatrix} \\ \cong \begin{bmatrix} \vdots & \vdots & \vdots \\ \uparrow & \parallel & \parallel \\ 0 \xrightarrow{(1)} & k \xrightarrow{(1)} & k \xrightarrow{(1)} \\ 0 \xrightarrow{(1)} & \uparrow \begin{pmatrix} 1 & 0 \\ 0 \end{pmatrix} \end{pmatrix} \\ 0 \xrightarrow{(1)} & \uparrow \begin{pmatrix} 1 & 0 \\ 0 \end{pmatrix} \end{pmatrix} \\ = \begin{bmatrix} \vdots & \vdots & \vdots \\ \uparrow & \uparrow & \parallel \\ 0 \xrightarrow{(1)} & k \xrightarrow{(1)} & k \xrightarrow{(1)} \end{pmatrix} \\ \oplus \begin{bmatrix} \vdots & \vdots & \vdots \\ \uparrow & \uparrow & \parallel \\ 0 \xrightarrow{(1)} & k \xrightarrow{(1)} & k \xrightarrow{(1)} \end{pmatrix} \\ \oplus \begin{bmatrix} \vdots & \vdots & \vdots \\ \uparrow & \uparrow & \parallel \\ 0 \xrightarrow{(1)} & h \xrightarrow{(1)} & h \xrightarrow{(1)} \end{pmatrix} \\ \oplus \begin{bmatrix} \vdots & \vdots & \vdots \\ \uparrow & \uparrow & \parallel \\ 0 \xrightarrow{(1)} & h \xrightarrow{(1)} & h \xrightarrow{(1)} \end{pmatrix} \\ \oplus \begin{bmatrix} \vdots & \vdots & \vdots \\ \uparrow & \uparrow & \parallel \\ 0 \xrightarrow{(1)} & h \xrightarrow{(1)} & h \xrightarrow{(1)} \end{pmatrix} \\ \oplus \begin{bmatrix} \vdots & \vdots & \vdots \\ \uparrow & \uparrow & \parallel \\ 0 \xrightarrow{(1)} & h \xrightarrow{(1)} & h \xrightarrow{(1)} \end{pmatrix} \\ \oplus \begin{bmatrix} \vdots & \vdots & \vdots \\ 0 \xrightarrow{(1)} & h \xrightarrow{(1)} & h \xrightarrow{(1)} \end{pmatrix} \\ \oplus \begin{bmatrix} \vdots & \vdots & \vdots \\ 0 \xrightarrow{(1)} & h \xrightarrow{(1)} & h \xrightarrow{(1)} \end{pmatrix} \\ \oplus \begin{bmatrix} \vdots & \vdots & \vdots \\ 0 \xrightarrow{(1)} & h \xrightarrow{(1)} & h \xrightarrow{(1)} \end{pmatrix} \\ \oplus \begin{bmatrix} \vdots & \vdots & \vdots \\ 0 \xrightarrow{(1)} & h \xrightarrow{(1)} & h \xrightarrow{(1)} \end{pmatrix} \\ \oplus \begin{bmatrix} \vdots & \vdots & \vdots \\ 0 \xrightarrow{(1)} & h \xrightarrow{(1)} & h \xrightarrow{(1)} \end{pmatrix} \\ \oplus \begin{bmatrix} \vdots & \vdots & \vdots \\ 0 \xrightarrow{(1)} & h \xrightarrow{(1)} & h \xrightarrow{(1)} \end{pmatrix} \\ \oplus \begin{bmatrix} \vdots & \vdots & \vdots \\ 0 \xrightarrow{(1)} & h \xrightarrow{(1)} & h \xrightarrow{(1)} \end{pmatrix} \\ \oplus \begin{bmatrix} \vdots & \vdots & \vdots \\ 0 \xrightarrow{(1)} & h \xrightarrow{(1)} & h \xrightarrow{(1)} \end{pmatrix} \\ \oplus \begin{bmatrix} \vdots & \vdots & \vdots \\ 0 \xrightarrow{(1)} & h \xrightarrow{(1)} & h \xrightarrow{(1)} \end{pmatrix} \\ \oplus \begin{bmatrix} \vdots & \vdots & \vdots \\ 0 \xrightarrow{(1)} & h \xrightarrow{(1)} & h \xrightarrow{(1)} & h \xrightarrow{(1)} \end{pmatrix} \\ \oplus \begin{bmatrix} \vdots & \vdots & \vdots \\ 0 \xrightarrow{(1)} & h \xrightarrow{(1)} \end{pmatrix} \\ \oplus \begin{bmatrix} \vdots & \vdots & \vdots \\ 0 \xrightarrow{(1)} & h \xrightarrow{(1)} & h$$

is a direct sum of the three free modules on the right. Because this is tedious to write and read, we usually draw two-parameter modules with diagrams such as



Thus, a module is free if it is a direct sum of modules whose support is an upper right quadrant.

Definition 2.1.17 (One-critical filtered complexes). We call a *P*-filtered complex K_* onecritical if $\{p \in P \mid \sigma \in K_p\}$ has a single minimal element for every $\sigma \in K$, denoted by $g(\sigma)$ and called the grade of σ .

If P is totally ordered, then all every P-filtered simplicial complexes K_* with $\bigcap K_* = \emptyset$ is one-critical. If K_* is one-critical, then $C_d(K_*)$ is free. The standard basis of $C_d(K_*)$ is the basis $\{\sigma \mid \sigma \in K_*^d\}$.

Definition 2.1.18 (*P*-graded matrix). Let *P* be a poset. A *P*-graded $m \times n$ -matrix consists of an ordinary $m \times n$ -matrix u(M) with entries in *k*, called its underlying matrix, and two tuples $\operatorname{rg}^M \in P^m$ and $\operatorname{cg}^M \in P^n$, called the row and column grades of *M*. In this case, we also say that *M* is a graded $\operatorname{rg}^M \times \operatorname{cg}^M$ -matrix. It is called valid if $M_{ij} \neq 0$ only if $\operatorname{rg}_i^M \leq \operatorname{cg}_j^M$. The sum M + M' of two graded matrices is defined if $\operatorname{rg}^M = \operatorname{rg}^{M'}$ and $\operatorname{cg}^M = \operatorname{cg}^{M'}$. In this case, it has $\operatorname{rg}^{M+M'} = \operatorname{rg}^M = \operatorname{rg}^{M'}$ and $\operatorname{cg}^{M+M'} = \operatorname{cg}^M = \operatorname{cg}^{M'}$. The product MM' of two graded matrices is defined if $\operatorname{cg}^M = \operatorname{rg}^M$. The product MM' of two graded matrices is defined if $\operatorname{cg}^M = \operatorname{rg}^M$ and $\operatorname{cg}^{MM'} = \operatorname{cg}^M$. The graded matrices is defined if $\operatorname{rg}^M = \operatorname{rg}^M$ and $\operatorname{cg}^{MM'} = \operatorname{cg}^M$. The graded $\operatorname{ranspose}$ of *M* is the graded $n \times m$ -matrix M^{\top} with row grades $\operatorname{rg}_i^{M^{\top}} = -\operatorname{cg}_i^M$, column grades $\operatorname{cg}_i^{M^{\top}} = -\operatorname{rg}_i^M$ and entries $[M^{\top}]_{ij} = [M]_{ji}$.

Remark 2.1.19. A graded matrix is valid if and only if its transpose is valid.

Remark 2.1.20. For algorithmic purposes, e.g., when maintaining a certain order (such as lexicographic or colexicographic order) on the row and column grades, it may be convenient to use the graded *anti-transpose* M^{\top} with row grades $\operatorname{rg}_{i}^{M^{\perp}} = -\operatorname{rg}_{n+1-i}^{M}$, column grades $\operatorname{cg}_{j}^{M^{\perp}} = -\operatorname{rg}_{m+1-j}^{M}$ and entries $[M^{\perp}]_{ij} = [M]_{m+1-j,n+1-i}$ instead of the graded transpose.

The following follows from the fact that $\operatorname{Hom}(F(q), F(p)) \cong \begin{cases} k & \text{if } p \leq q, \\ 0 & \text{otherwise} \end{cases}$:

Lemma 2.1.21. Let F and F' be free persistence modules with bases $(b_j)_{j=1}^n$ and $(b'_i)_{i=1}^m$, respectively. Then there is a one-to-one correspondence between morphisms $f: F \to F'$ and valid graded $(g(b'_i))_i \times (g(b_j))_j$ -matrix. The morphism f corresponds to the matrix M with entries

$$f(b_j) = \sum_{g(b'_i) \le g(b_j)} M_{ij} F'_{g(b_j),g(b'_i)}(b'_i)$$

for all $j \leq n$.

In this case, we identify valid graded $(g(b'_i))_i \times (g(b_j))_j$ -matrices and morphisms from F to F' without making the distinction explicit.

Definition 2.1.22. A module $M \in \text{Vec}^P$ is projective (resp., *injective*) if the contravariant functor $\text{Hom}_{\text{Vec}^P}(-, M)$: $\text{Vec}^P \to \text{Vec}$ (resp., the covariant functor $\text{Hom}_{\text{Vec}^P}(M, -)$: $\text{Vec}^P \to \text{Vec}$) is exact.

Equivalently, a module M is projective (resp., injective) if and only if every short exact sequence of modules of the form $0 \to A \to B \to M \to 0$ (resp., $0 \to M \to A \to B \to 0$) splits.

Proposition 2.1.23 ([82, Proposition 5]). Every projective module $M \in \text{Vec}^P$ is free.

This is a generalization of Kaplansky's theorem [7, 83] and the Quillen–Suslin-theorem [113, 120].

2.2 One-parameter persistence

Let $P = \mathbf{Z}$ or $P = \mathbf{R}$. For $-\infty \leq b < d \leq \infty$, we let $I(b, d) \in \operatorname{Vec}^{P}$ be the *interval module* with

$$I(b,d)_p = \begin{cases} k \text{ if } b \le p < d, \\ 0 \text{ otherwise,} \end{cases} \qquad \qquad I(b,d)_{qp} = \begin{cases} \text{id if } b \le p \le q < d, \\ 0 \text{ otherwise.} \end{cases}$$

In particular, $I(b, \infty) = F(b)$ for $b > -\infty$.

Theorem 2.2.1 (Structure Theorem [26, 52, 126], see also [70]). If $M \in \text{vec}^P$ for $P = \mathbf{Z}$ or $P = \mathbf{R}$, then

$$M \cong \bigoplus_{i \in I} I(b_i, d_i)$$

for a uniquely determined finite indexing set I and $-\infty \leq b_i < d_i \leq \infty$ for all $i \in I$.

The multiset barc $M = \{(b_i, d_i) \mid i \in I\}$ is called the *barcode* of M. A barcode is commonly depicted by a *persistence diagram* plotting the pairs (b_i, d_i) as points; see Figure 2.3b. Pairs $(b_i, d_i) \in \text{barc } M$ are called *essential* if $d_i = \infty$ and *finite* or *non-essential* otherwise. Thus, the essential pairs of barc M correspond to a basis of the vector space colim M.

From now on, let $P = \mathbf{Z}$. Persistence modules over \mathbf{Z} are equivalent to graded modules over the principal ideal domain (PID) k[x] (see Proposition 2.3.10). Therefore, if M is finitely generated, then Theorem 2.2.1 is a graded version of the invariant factor decomposition of finitely generated modules over PIDs [128, Theorem 2.1], which is essentially a manifestation of Gabriel's theorem [74].

Let C_{\bullet} be a chain complex of finite rank free persistence modules. Then $H_d(C_{\bullet}) \in \text{vec}^{\mathbb{Z}}$ for each d, and we consider barc $H_d(C_{\bullet})$. Choose bases of C_d for each d, and let D_d be the graded matrix representing the boundary morphism ∂_d of C_{\bullet} with respect to the standard basis. Since k[x] is a PID, every \mathbb{Z} -graded matrix has a Smith normal form, and barc $H_d(C_{\bullet})$ can be determined from the graded Smith normal forms of D_d and D_{d-1} [128]. In practical computations, it is more convenient to work with the *Standard Algorithm* instead, which we recall now.

Definition 2.2.2. The *pivot* of a column M_j of a matrix M is piv $M_j := \max\{i \mid M_{ij} \neq 0\}$. A matrix M is *reduced* if all its non-zero columns have distinct pivots.

Let $C_{\bullet} = \bigoplus_{i=1}^{N} F(z_i)$ be a chain complex of free **Z**-persistence modules of finite rank, with $z_i \leq z_j$ if $i \leq j$, and D_d be the matrix representing ∂_d with respect to the standard basis.

Proposition 2.2.3 (Persistence Pairing [64, 128]). If V_d be an invertible valid graded matrix such that $R_d := D_d V_d$ is reduced for all d, then

$$\operatorname{barc} H_d(C_{\bullet}) = \{ (z_i, z_j) \mid [R_d]_i = 0 \text{ and } i = \operatorname{piv}[R_{d+1}]_j \} \\ \cup \{ (z_i, \infty) \mid [R_d]_i = 0 \text{ and } \nexists j \colon i = \operatorname{piv}[R_{d+1}]_j \}.$$

$$(2.4)$$

Figure 2.3: Left: simplex-wise Z-filtration of a simplicial complex K_* The numbers indicate the value in Z at which a simplex enters the filtration. Right: Persistence diagram showing the barcode of the reduced persistent homology of K_* . Each point (x, y) in the diagram denotes a pair $(x, y) \in \text{barc } H_{\bullet}(K_*)$.



The matrices R_{\bullet} and V_{\bullet} can be computed from D_{\bullet} using the Standard Algorithm; see Algorithm 1. If $[R_d]_i = 0$, then *i* is called a *birth index* of R_d ; otherwise, *i* is called a *death index*. Every column index *i* of R_d is either birth or death. To see this, note that every submodule of a free one-parameter persistence module of finite rank is free. In particular, $0 \to Z_{\bullet}(C_{\bullet}) \to C_{\bullet} \to B_{\bullet}(C_{\bullet}) \to 0$ is an exact sequence of free modules and thus splits. Assuming that all columns of R_d have distinct grade, this splitting is uniquely determined and fixes a partition of the basis of C_{\bullet} into birth and death indices.

Lemma 2.2.4. If R_d , R_{d+1} are as in Proposition 2.2.3, then $[R_d]_i = 0$ whenever $i = piv[R_{d+1}]_j$. In particular, we can simplify (2.4) to

barc
$$H_d(C_{\bullet}) = \{(z_i, z_j) \mid i = \text{piv}[R_{d+1}]_j\}$$

 $\cup \{(z_i, \infty) \mid [R_d]_i = 0 \text{ and } \nexists j : i = \text{piv}[R_{d+1}]_j\}$

Example 2.2.5. Consider the filtered complex K_* in Figure 2.3. We compute a barcode for its reduced homology. Its augmented chain complex has the graded boundary matrices

$$D_0 = {}^{1}\begin{bmatrix} 1 & 2 & 3 & 7\\ 1 & 1 & 1 & 1 \end{bmatrix}, \qquad D_1 = {}^{1}_{3}\begin{bmatrix} 1 & 5 & 6 & 8 & 9\\ 0 & 1 & \overline{1} & 0 & 0\\ 0 & 1 & \overline{1} & 0 & 1\\ 1 & 0 & 1 & \overline{1} & 0\\ 0 & 0 & 0 & 1 & 1 \end{bmatrix}, \qquad D_2 = {}^{4}_{5}\begin{bmatrix} 1\\ 1\\ 1\\ 0\\ 8\\ 9\\ 0 \end{bmatrix}$$

where we write $\overline{1}$ for -1. The Standard Algorithm computes the matrices

According to (2.4), we obtain

barc
$$H_0(K_*) = \{(2,5), (3,4), (7,8)\},$$
 barc $H_1(K_*) = \{(6,10), (9,\infty)\},$

and no non-zero higher dimensional persistent homology modules; see Figure 2.3.

Definition 2.2.6 (Persistence basis). Let C_{\bullet} be a chain complex of finite rank free modules, and assume for a d that barc $H_d(C_{\bullet}) = \{(b_i, d_i) \mid i \in I\}$ for some indexing set I. We let $I_f = \{i \in I \mid d_i < \infty\}$ and $I_e = I \setminus I_f$. A persistence basis of $H_d(C_{\bullet})$ is a system

$$\{(z_i, c_i) \mid i \in I_{\rm f}\} \cup \{z_i \mid i \in I_{\rm e}\},\tag{2.5}$$

where $z_i \in Z_d(C_{\bullet})$ and $c_i \in C_{d+1}$, such that for $i \in I_f$, we have $I(b_i, d_i)_{d_i, b_i}(z_i) = \partial c_i$, and such that the assignment

$$\bigoplus_{i \in I} I(b_i, d_i) \longrightarrow H_d(C_{\bullet}),$$
$$e_i \longmapsto [z_i]$$

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is an isomorphism, where e_i denotes the generator of the interval module indexed by *i*. The pairs in the left set in (2.5) are called *persistence pairs*, the cycles in the second pairs are representatives of the *essential homology classes*.

Example 2.2.7. If R_{\bullet} and V_{\bullet} are as in Proposition 2.2.3, then

 $\{([R_{d+1}]_j, [V_{d+1}]_j) \mid [R_{d+1}]_j \neq 0\} \cup \{[V_d]_i \mid [R_d]_i = 0 \text{ and } \nexists j : i = \operatorname{piv}[R_{d+1}]_j\}$

represent a persistence basis of $H_d(C_{\bullet})$.

Algorithm 1 has complexity $\mathcal{O}(N^3)$ [128], which can be attained even if $C_{\bullet} = C_{\bullet}(K_*)$ for a filtered simplicial complex K_* [105], although worst-case complexity is rarely observed in practice; see also [78]. If K = VR(S) is the Vietoris–Rips complex of a finite metric space S, then $|K^d| = \binom{|S|}{d+1}$. Therefore, computing the barcode of $H_d(VR_*(S))$ becomes impractical already for relatively small values of |S|. The following standard approach remedies this shortcoming.

```
Algorithm 1: Standard algorithm for one-parameter persistent homology.
Input:
             A graded m \times n-matrix D, rows and columns sorted by grade.
Output: A Z-graded invertible n \times n-matrix V such that DV is reduced.
function Reduce(D):
     p \leftarrow (0, \ldots, 0) \in \mathbf{N}^n
     V \leftarrow 1 \in k^{n 	imes n} with \mathrm{rg}^V = \mathrm{cg}^V = \mathrm{cg}^D
                                                                                                         ▷ graded unit matrix
     foreach j = 1, \ldots, n do
          while piv[D]_i \neq 0 do
                i \leftarrow \operatorname{piv}[D]_i
                k \leftarrow p_i
                if k \neq 0 then
                     [D]_j \leftarrow [D]_j - [D]_{ij}/[D]_{ik}[D]_k
[V]_j \leftarrow [V]_j - [D]_{ij}/[D]_{ik}[V]_k
                else p_i \leftarrow j; break
     return V
```

2.2.1 Relative cohomology and clearing

As before, let K_* be a finite one-critical **Z**-filtered simplicial complex. Consider the relative cochain complex $C^{\bullet}(K, K_*)$. Because $C^{\bullet}(K, K_*)$ is the kernel of the restriction $\Delta C^{\bullet}(K) \rightarrow C^{\bullet}(K_*)$, each vector space $C^{\bullet}(K, K_z)$ is spanned by the dual simplices σ^* with $\sigma \notin K_z$. Therefore,

$$C^{\bullet}(K, K_*) \cong \bigoplus_{\sigma \in K} F(-g(\sigma) + 1)$$

is a cochain complex of free **Z**-persistence modules; see Figure 1.5. With respect to the standard basis of $C_{\bullet}(K, K_*)$, the coboundary operator $\delta^d \colon C^{d-1}(K, K_*) \to C^d(K, K_*)$ is represented by the graded matrix $D^d \coloneqq D_d^{\top}$.

Remark. We assumed that the rows and columns of D_d are ordered such that their grades are monotonically increasing. The reversal of the row and column order in Definition 2.1.18 ensures that D^d has the same property.

One can compute the reduced form $R^d = D^d V^d$ from D^d using Algorithm 1, and, one obtains the barcode of $H^{d+1}(K, K_*)$ from R^d and R^{d+1} analogously to Proposition 2.2.3. The following is dual to Lemma 2.2.4 Algorithm 2: A variant of Algorithm 1 that incorporates clearing.

A valid graded $I \times m$ -matrix D and a reduced valid graded $m \times I$ -matrix R such that DR = 0; Input: both with rows and columns sorted by grade. **Output**: A **Z**-graded invertible $n \times n$ -matrix V such that DV is reduced. function ReduceC(D, R'): $p \leftarrow (0, \ldots, 0) \in \mathbf{N}^m$ $V \leftarrow 1 \in k^{n \times n}$ with $rg^V = cg^V = cg^D$ for j = 1, ..., n do ▷ clearing if $[R'_d] \neq 0$ then $[D]_{\operatorname{piv}[R_d]} \leftarrow 0$ $[V]_{\mathsf{piv}[R_d]} \leftarrow [R_d]$ for j = 1, ..., m do while $piv[D]_i \neq 0$ do $i \leftarrow \operatorname{piv}[D]_i$ $k \leftarrow p_i$ if $k \neq 0$ then $[D]_j \leftarrow [D]_j - [D]_{ij}/[D]_{ik}[D]_k$ $[V]_j \leftarrow [V]_j - [D]_{ij}/[D]_{ik}[V]_k$ else $p_i \leftarrow j$; break return V

Lemma 2.2.8. Let $R^d = D^d V^d$ and $R^{d-1} = D^{d-1}V^{d-1}$ be reduced, where V^d and V^{d-1} are invertible upper triangular matrices. Then $[R^d]_i = 0$ whenever $i = \operatorname{piv}[R^{d-1}]_j$. In particular,

barc $H^{d+1}(K, K_*) = \{(z_i, z_j) \mid i = \operatorname{piv}[R^d]_j\}$ $\cup \{(z_i, \infty) \mid [R^{d+1}]_i = 0 \text{ and } \nexists j \colon i = \operatorname{piv}[R^d]_j\}.$

If Algorithm 1 is applied to R_d or R^d , it will spend most of its run time on those columns of R_d (resp., R^d) that are reduced to zero. Experience shows that if a column of R_d (resp., R^d) is not reduced to zero, it will typically reach its final state already after relatively few additions. If computing the (co)homology of Vietoris–Rips complexes, this is true in particular for cohomology, because in this case, R^d has few columns and many rows, so chances are low that two columns have the same pivot.

Looking again at Lemmas 2.2.4 and 2.2.8, we see that it is not necessary in all cases to reduce all columns of R_d (resp., R^d). Namely, if $i = \operatorname{piv}[R_{d+1}]_j$ (resp. $i = \operatorname{piv}[R^{d-1}]_j$) for some j, then we know that Algorithm 1 will result in $[R_d]_i = 0$ (resp. $[R^d]_i = 0$). We can therefore skip reducing these columns and set $[R_d]_i \coloneqq 0$, $[V_d]_i \coloneqq [R_{d+1}]_j$ (resp. $[R^d]_i \coloneqq 0$, $[V^d]_i \coloneqq [R^{d-1}]_j$) immediately. This optimization scheme is called *clearing* or the *twist algorithm* [11, 46]; see Algorithm 2. It has a great share in the efficiency of implementations such as [1, 9, 111].

We note that for Vietoris-Rips complexes, clearing is effective only when computing the barcode of $C^{\bullet}(K, K_*)$, rather than the barcode of $C_{\bullet}(K_*)$ [12, 9, p. 403]. Namely, when reducing $D_1, \ldots, D_{d_{\max}}$, we have to compute $R_{d_{\max}}$ using Algorithm 1, before we can compute $R_{d_{\max}-1}, \ldots, R_1$ using Algorithm 2. For Vietoris-Rips complexes, applying Algorithm 1 to $D_{d_{\max}}$ will dominate the run time. For cohomology, however, we may compute R^1 using Algorithm 1, and then use Algorithm 2 to compute $R^2, \ldots, R^{d_{\max}}$. As D^1 is very small compared to the higher dimensional coboundary matrices, this is much more efficient. Furthermore, there are specialized algorithms to compute the barcode of $H^0(K, K_*)$ more efficiently [9].

It remains to relate the barcodes of $H^{\bullet}(K, K_*)$ and $H_{\bullet}(K_*)$. The first step follows from $H^{\bullet}((C_{\bullet})^*) \cong H_{\bullet}(C_{\bullet})^*$:

Lemma 2.2.9. If $M \in \text{vec}^{\mathbb{Z}}$, then barc $M^* = \{(-d, -b) \mid (b, d) \in \text{barc } M\}$.

Next, we relate relative and absolute homology. Recall the long exact homology sequence

$$\cdots \to H_{d+1}(K, K_*) \xrightarrow{\partial_{d+1}} H_d(K_*) \xrightarrow{i_d} \Delta H_d(K) \xrightarrow{p_d} H_d(K, K_*) \xrightarrow{\partial_d} H_{d-1}(K_*) \to \cdots$$

Proposition 2.2.10 ([19]). For every d, the induced short exact sequences

$$0 \to \operatorname{coker} p_{d+1} \to H_d(K_*) \to \ker p_d \to 0 \tag{2.6}$$

$$0 \to \operatorname{coker} i_{d+1} \to H_{d+1}(K, K_*) \to \ker i_d \to 0$$
(2.7)

split.

Proof (sketch). For any module M, let

$$\operatorname{barc}_{-\infty} M \coloneqq \{(b,d) \in \operatorname{barc} M \mid b = -\infty\}, \qquad \operatorname{barc}^{\infty} M \coloneqq \{(b,d) \in \operatorname{barc} M \mid d = \infty\}.$$

The barcode barc $\Delta H_d(K)$ consists exclusively of bars $(-\infty, \infty)$. Therefore, the barcode of any submodule of $\Delta H_d(K)$ contains only bars of the form (b, ∞) for some b. This implies that

$$\operatorname{barc} \operatorname{ker} p_d \subseteq \operatorname{barc}^{\infty} H_d(K_*),$$
(2.8)

so ker p_d is free; hence (2.6) splits. Likewise, the barcode of any quotient of $\Delta H_{d+1}(K)$ contains only bars of the form $(-\infty, d)$ for some d. This implies that

$$\operatorname{barc}\operatorname{coker} i_{d+1} \subseteq \operatorname{barc}_{-\infty} H_{d+1}(K, K_*), \tag{2.9}$$

so coker i_{d+1} is injective; hence (2.7) splits. See [19] for details.

One may also check that barc $H_{d+1}(K, K_*)$ and thus also barc coker p_{n+1} contains no bars of the form (b, ∞) for any b. Similarly, barc $H_d(K_*)$ and thus also barc ker i_d contains no bars of the form $(-\infty, d)$ for any d. Exactness of (2.6) and (2.7) implies that (2.8) and (2.9) are equalities.

Corollary 2.2.11 ([55, §2.4]). Let K_* be a finite one-critical **Z**-filtered simplicial complex. For all d,

$$\operatorname{barc} H_d(K_*) = \{ (b, d) \in \operatorname{barc} H_{d+1}(K, K_*) \mid -\infty < b \} \\ \cup \{ (d+1, \infty) \mid (-\infty, d) \in H_d(K, K_*) \},$$
(2.10)
$$\operatorname{barc} H_{d+1}(K, K_*) = \{ (b, d) \in \operatorname{barc} H_d(K_*) \mid d < \infty \}$$

$$\cup \{(-\infty, b-1) \mid (b, d) \in H_{d+1}(K_*)\}.$$
(2.11)

Therefore, the **Z**-persistence modules $\bigoplus_d H_d(K_*)$ and $\bigoplus_d H_d(K, K_*)$ determine each other uniquely even if K is not acyclic. We can convert a barcode of $H^{\bullet}(K, K_*)$ (which is efficiently computable using Algorithm 2) to a barcode of $H_{\bullet}(K_*)$ using (2.10). This is also the approach followed in [9].

Remark. When looking at (2.10), it might seem surprising that $H_{\bullet}(K_*)$ can be determined more efficiently by computing $H^{\bullet}(K, K_*)$, given that one needs a barcode of $H^{d+1}(K, K_*)$ to determine the barcode of $H_d(K_*)$. It seems that one needs to compute the reduced matrix R^{d+1} to obtain barc $H^{d+1}(K, K_*)$, which would be unfeasible for Vietoris–Rips complexes. A closer look at (2.10) reveals that only the *finite* bars of barc $H^{d+1}(K, K_*)$ are needed to determine barc $H_d(K_*)$. According to Lemma 2.2.8, these can be determined from R^d .



Figure 2.4: Persistence diagram of the Vietoris–Rips complex of a point cloud sampled from an annulus (a) without and (b) with three outliers outliers. The two point clouds are the same as in Figures 1.2 and 1.3.

2.3 Multi-parameter persistence

Multi-parameter persistence is the theory of \mathbb{Z}^{n} - or \mathbb{R}^{n} -persistence modules for n > 1. To motivate multi-parameter persistence, recall the stability results of one-parameter persistence. These assert that: a) for modules, the assignment $M \mapsto \text{barc } M$ is an isometry [15, 28, 40, 96]; b) for a triangulable space X, the assignment { \mathbb{R} -valued continuous functions on X} $\rightarrow \text{Vec}^{\mathbb{R}}$, $f \mapsto H_d(X_{f \leq *})$ is Lipschitz with respect to the supremum norm and the interleaving distance on persistence modules [48]; and c) for metric spaces X, the assignments $X \mapsto H_d(VR_*(X))$ and $X \mapsto H_d(\check{C}_*(X))$ are Lipschitz with respect to the Gromov–Hausdorff distance and the interleaving distance [38, 41, §5.2].

Despite these stability results, however, one-parameter persistent homology is known to be susceptible to outliers [23]. For example, Figure 2.4 shows that already adding very few outliers inside the annulus drastically changes the persistence diagram. In particular, while the the persistence diagram in (b) exhibits the actual homology of the annulus, this is not the case for the persistence diagram in (a), Figure 1.2 shows the persistence diagrams of the Delaunay-filtration for the same point cloud.

According to [24, §1.7], common approaches to address this in the framework of one-parameter persistence are removing points that have density below a fixed threshold [32], considering the filtration by density for a fixed scale parameter [44, 45], or the development of new filtrations robust to outliers [25, 29, 39, 42, 112], most of which rely on fixing a density, scale or bandwidth parameter, which arguably contradicts the explorative approach of persistent homology. A natural remedy for this is to introduce one or more additional parameters to control for these properties, which leads to the notion of multi-parameter persistent homology. A common choice is to extend constructions such as the α -complex, the Čech complex and the Vietoris–Rips complex by a second parameter that controls for the local density of the point cloud.

Example 2.3.1 (Function-Rips bifiltration). Let X be a metric space and $f: X \to \mathbf{R}$ be a continuous function. The function-Rips bifiltration $VR_*(X, f)$ is the \mathbf{R}^2 -filtered simplicial complex with $VR_{r,s}(X, f) \coloneqq VR_r(X_{f\leq s})$. It is a one-critical bifiltration. If f(v) measures the density of S around v, we call $VR_*(X, f)$ also the density-Rips bifiltration.

Example 2.3.2 (Degree-bifiltrations [24, §2.3]). The degree of a vertex v in a simplicial complex K is the number of edges incident to v. For a P-filtered simplicial complex K_* , let $D_*(K_*)$ be the $\mathbb{Z} \times P$ -filtered simplicial complex such that $D_k(K_p)$ is the maximal subcomplex of K_p on the vertices of degree at least -k - 1. In particular, we define the degree-Rips bifiltration $D_*(VR_*(S))$ and degree-Čech bifiltration $D_*(\tilde{C}_*(S))$, which are bifiltrations over $\mathbb{R} \times \mathbb{Z}$. These need not be one-critical in general. For example, Figure 2.5a contains N = 51 points sampled from a unit circle, with a single outlier in the origin. Every simplex has a curve describing the pairs (-degree, diameter) at which the simplex enters the filtration; see Figure 2.5. For the vertices, each of these curves has N steps; one for each edge connecting the vertex to one other simplex.



Figure 2.5: Generator curves of a degree-Rips bifiltration. (a) The underlying point cloud consists of 50 points sampled from a perturbed unit circle (blue), plus 25 additional points sampled uniformly from the square (red). (b)–(d) Generator curves for the 0-, 1- and 2-simplices of the associated degree-Rips complex. The generator curve for simplices spanned by "red" vertices are red, curves for simplices involving "blue" and "red" samples are drawn magenta.

Another common bifiltrations of spaces is the *multi-cover bifiltration* [118]. Other common bifiltered simplicial complexes are the *subdivision-Rips* and *subdivision-Čech bifiltration* and the the *rhomboid tilings* [51, 62, 63]. Multi-parameter persistence modules can be equipped with an interleaving metric [96], and under suitable conditions, assigning to a finite metric space its multi-cover bifiltration, subdivision- or degree-Rips or -Čech bifiltration is Lipschitz with respect to the (Gromov-)Prokhorov and the (generalized, homotopy-)interleaving distance [24, theorem 1.6, 1.7, 116].

Besides remedying the impact of outliers, multi-parameter persistent homology has been used in image classification [35], analysis of time-dependent spaces [88], time series [87] and clustering [116].

2.3.1 Invariants of multi-parameter persistence modules

In general, multi-parameter persistence modules do not admit a classification as simple as the barcode in one-parameter persistence. To explain this, we note that by virtue of Theorem 2.2.1, the barcode of a one-parameter persistence module is a way to describe its decomposition into indecomposable direct summands. For any P, a module in $M \in \text{vec}^P$ is *indecomposable* if $M = M' \oplus M''$ implies that one of M', M'' is zero. Every module $M \in \text{vec}^P$ is isomorphic to a direct sum of indecomposable modules that have local endomorphism rings [26]. The Krull–Remak–Schmidt–Azumaya Theorem [5, thm. 1.(ii), 68, thm. 2.12] implies that the decomposition of M is essentially unique. For $P = \mathbb{Z}$, the indecomposable modules in $\text{vec}^{\mathbb{Z}}$ are precisely the interval modules I(b, d) for $-\infty \leq b < d \leq \infty$. For n > 1, the indecomposable \mathbb{Z}^n -persistence modules cannot be classified by such a simple structure. In particular, an indecomposable \mathbb{Z}^n -persistence modules modules need not be determined by its support.

Example 2.3.3. The \mathbb{Z}^2 -persistence module

$$\begin{array}{c} \vdots & \vdots & \vdots \\ \parallel & \begin{pmatrix} 1 \\ 0 \end{pmatrix} & k^2 = k^2 = \cdots \\ \uparrow & \begin{pmatrix} \alpha \\ \beta \end{pmatrix} \uparrow & \begin{pmatrix} a \\ b \end{pmatrix} & k^2 = \cdots \\ \uparrow & \uparrow & \begin{pmatrix} a \\ b \end{pmatrix} & k^2 = \cdots \\ \uparrow & \uparrow & \begin{pmatrix} a \\ 1 \end{pmatrix} \uparrow & \cdots \\ 0 \longrightarrow 0 \longrightarrow k = \cdots \end{array}$$

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is indecomposable for every $a, b \neq 0$, because it contains the indecomposable representation

$$M = \begin{pmatrix} k & \stackrel{\stackrel{(1)}{\longrightarrow}}{\xrightarrow{\nearrow}} k^2 \\ k {a \choose b} & \uparrow {0 \choose 1} \\ k \end{pmatrix} \text{ of the quiver } Q = \begin{pmatrix} u & \xrightarrow{\longrightarrow} x \\ v & \uparrow \\ w \end{pmatrix}$$

One way to see that M is indecomposable is by listing all isomorphism classes of submodules and quotients of M and observing that no non-trivial submodule of M is isomorphic to a quotient of M. Another way to see this is by considering the endomorphism ring of M. Explicitly, End M can be seen as the set of those matrices

$$\begin{pmatrix} M_u & M_v & M_w & M_x \\ M_v & & & \\ M_w & & & \\ M_x & & & \begin{bmatrix} * & * \\ & * & & \\ & & & \begin{bmatrix} * & * \\ * & * \end{bmatrix} \end{pmatrix} \in k^{5 \times 5}$$

that commute with the matrices

$$\binom{0}{1} \begin{pmatrix} 0 & 0 \\ 0 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}, \binom{0}{0} \begin{pmatrix} 0 & 0 \\ 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}, \binom{0}{0} \begin{pmatrix} 0 & 0 \\ 0 & 0 \\ 0 & 0 & 0 \\ 1 & 0 & 0 \end{pmatrix}$$

corresponding to the structure maps of M. This defines a linear system. Solving this shows that End $M \cong k$ unless a = 0 or b = 0. In particular, End M is a local ring unless a = 0 or b = 0, hence M is indecomposable by the Azumaya–Krull–Remak–Schmidt theorem [5, 26, §2, 68]. In particular, this gives a family of indecomposable \mathbb{Z}^2 -persistence modules parametrized by $k\mathbb{P}^1 \setminus \{[0:1], [1:0]\}$. This implies that there is no discrete invariant (in the sense of [34]) classifying all indecomposable persistence modules.

Gabriel's theorem [74] implies that there exist infinitely many indecomposable $\{1, \ldots, k\}^n$ persistence modules for any $k \ge 1$ and n > 1. In fact, for any n > 1, there are indecomposable \mathbf{Z}^n -persistence modules of arbitrarily large dimension [30]. For multi-parameter persistence modules, instead of a barcode, one seeks to define and compute other meaningful invariants from a module $M \in \text{vec}^{\mathbf{R}^n}$ or $\text{vec}^{\mathbf{Z}^n}$, such as the graded Betti diagrams $\beta_i(M)$ (see Definition 2.3.6), the *Hilbert function* $\mathbf{Z}^n \to \mathbf{N}, p \mapsto \dim M_p$, Hilbert series, associated primes and local cohomology [80], the rank invariant $\{(p,q) \in (\mathbf{Z}^n)^2 \mid p \le q\} \to \mathbf{N}, (p,q) \mapsto \text{rk } M_{qp}$ for $p \le q$ [34], the generalized rank invariant [56, 88, 90, 109], the signed barcode [27], the fibered barcode [22, 100, 123], the shift dimension [37, 75], or various others.

Of particular interest is computing a minimal free resolution or a minimal free presentation of a given finitely generated persistence module $M \in \text{vec}^{\mathbb{Z}^n}$, because finitely generated modules are in bijection with isomorphism classes of minimal free resolution. Additionally, several of the above invariants can computed from a minimal free resolution of M. Therefore, it is a natural question how to compute a minimal free resolution of $H_{\bullet}(K_*)$ for a finite \mathbb{Z}^n -filtered simplicial complex K_* .

Definition 2.3.4. A free presentation of a persistence module M is a morphism $\partial: F_1 \to F_0$ of free modules such that $M \cong \operatorname{coker} \partial$. It is called *finite* if F_1 , F_0 have finite rank. A free resolution of a persistence module M is a chain complex F_{\bullet} of free modules concentrated in non-negative degrees such that there is an exact sequence

$$\dots \to F_1 \to F_0 \xrightarrow{f_0} M \to 0 \tag{2.12}$$

It is called *finite* if all F_i have finite rank, and only finitely many F_i are non-zero. The *length* of a finite free resolution F_{\bullet} is the smallest integer ℓ such that $F_i = 0$ for all $i > \ell$. A morphism of free resolutions is a morphism of chain complexes. This renders $Ch(Vec^P)$ an abelian category. The morphism f_0 is called the *augmentation map* of the resolution, and the sequence in (2.12) is sometimes called the *augmented resolution*; usually, however, by a slight abuse of language that should not confuse the reader, the sequence (2.12) is also called a free resolution of M.

Example 2.3.5. If $M \in \text{vec}^{\mathbb{Z}}$ is finitely generated and barc $M = \{(b_i, d_i); i \in I\}$ for some index set I, then $0 \to \bigoplus_{i \in I, d_i \neq \infty} F(d_i) \to \bigoplus_{i \in I} F(b_i)$ is a free resolution of M.

Every persistence module has a (not necessarily finite) free resolution. Finite free resolutions (resp., free presentations) are a convenient way to describe a module. Namely, after having chosen bases for the modules in the resolution (presentation), the morphisms in the resolution (presentation) can be represented by graded matrices, which can be stored and processed by a computer program.

Remark (Lifts of morphisms). If $F_{\bullet} \to M$ and $G_{\bullet} \to N$ are free resolutions of two modules M and N, and $\phi: M \to N$ is a morphism, then there exists a (not necessarily unique) morphism $\phi_{\bullet}: F_{\bullet} \to G_{\bullet}$ such that the diagram

$$egin{array}{ccc} F_{ullet} & o M \ \phi_{ullet\downarrow} & & \downarrow \phi \ G_{ullet} & o N \end{array}$$

commutes. This follows from exactness of the rows, and from the universal property of the free (or, more generally, projective) modules involved. The morphism f_{\bullet} is called a *lift* of f.

Finite resolutions need not be unique. However, any two free resolutions of the same module are chain homotopy equivalent. Finitely generated \mathbf{Z}^n -persistence modules have an essentially unique smallest free resolution, called their *minimal free resolution*; see Theorem 2.3.18. The minimal free resolution of a module M is minimal in the sense that it is a direct summand of any other free resolution of M. We explain this in detail in Section 2.3.3.

Definition 2.3.6. The *i*th graded Betti number $\beta_i(M)$ of a finitely generated module $M \in \text{vec}^{\mathbb{Z}^n}$ is the graded rank $\beta_i(M) = \text{rk } F_i$ for any minimal free resolution F_{\bullet} of M.

Theorem 2.3.18 below shows that a finitely generated module has an essentially unique free resolution, which implies that its graded Betti numbers are independent of the chosen minimal free resolution and hence well-defined.

Example 2.3.7. The barcode of a one-parameter persistence module $M \in \text{vec}^Z$ represents a minimal free resolution. If $M \in \text{vec}^Z$ is finitely generated and barc $M = \{(b_i, d_i) \mid i \in I\}$ for some finite indexing set I, then $\beta_0 = \{b_i \mid i \in I\}$ and $\beta_1(M) = \{d_i \mid i \in I, d_i < \infty\}$.

2.3.2 Examples

Density-Rips filtrations We equip the point cloud S from Figure 2.4b with the density function $\overline{}$

$$\rho: S \to \mathbf{R}_{\geq 0}, \quad p \mapsto \sum_{p \neq q \in S} \exp(-\frac{\|p-q\|^2}{2\sigma^2})$$
(2.13)

for a manually chosen bandwidth parameter σ , see Figure 2.6a for an example. We have computed a minimal free resolution of $H_{\bullet}(VR_*(S,\rho))$, using our implementation [94]. The corresponding Betti diagrams and Hilbert functions are shown in Figure 2.6. Note that the diagrams show reduced homology.

For a second example, let S be the point cloud shown Figure 2.7a. Despite the fact that the circle is clearly visible in the picture, the homology type of S^1 would be invisible in the corresponding Vietoris–Rips persistence diagram. We equip the point cloud with a density function ρ as in (2.13). Since ρ takes lower values on the supposed outliers, we filter the point cloud by $-\rho$. The graded Betti diagrams and Hilbert function of $H_{\bullet}(VR_*(S, -\rho))$ are shown in Figure 2.7.

From this diagram, one can obtain an estimate for a value of ρ that discerns the outliers. Each horizontal slice through the resolution at a value $-\rho_0$ corresponds to a resolution of the one-parameter persistent homology of $\{p \in S \mid \rho(p) \geq \rho_0\}$; that is, a barcode. One sees that for $\rho_0 \approx 14$, the corresponding horizontal slice contains a large region where $H_1(-)$ is one-dimensional. The corresponding barcode is shown in Figure 2.8.



Figure 2.6: (a) The point cloud *S* from Figure 2.4b, equipped with the Gaussian density function $\rho: \rho \mapsto \sum_{p \neq q \in S} \exp(-\frac{\|\rho-q\|^2}{2\sigma^2})$ with $\sigma = 2.5$. The function ρ that takes larger values on the outliers, and lower values on the desired points. (b), (c) Graded Betti numbers (teal: β_0 , red: β_1 , orange: β_2) and Hilbert function (shades of blue increasing from dimension = 0 to dimension ≥ 10) of the reduced homology $H_{\bullet}(VR_*(X, \rho))$ of the density-Rips complex of the point cloud from (a).



Figure 2.7: (a) Uniform sample (250 points) from a unit 1-sphere, superimposed with uniform sample (150 points) of the rectangle $[-1.5, 1.5]^2$. Each point is assigned the density from (2.13) with $\sigma = 0.2$. The right picture shows the (non-reduced) persistent homology of the associated one-parameter Vietoris–Rips complex. (b), (c) Associated graded Betti numbers.

Figure 2.8: Barcode representing the horizontal slice through the diagrams in Figure 2.7 at density $\rho = 14$. This corresponds to taking the barcode of $H_{\bullet}(VR_*(S'))$, where S' consists of the points from Figure 2.7a that satisfy $\rho(-) \ge 14$.





Figure 2.9: Graded Betti numbers and Hilbert function of (non-reduced) persistent homology of the degree-Rips filtration on the points from Figure 2.5a.

Degree-Rips filtrations Recall the point-cloud S from Figure 2.5a. Figure 2.9 shows the Hilbert function and graded Betti numbers of $H_{\bullet}(K_*)$ of the corresponding degree-Rips filtration.

2.3.3 Persistence modules as modules over a graded algebra

For explaining the theory of minimal free resolutions, it is convenient to talk about persistence modules using the language of graded algebras.

Definition 2.3.8. A \mathbb{Z}^n -graded k-algebra is a k-algebra A, such that the underlying k-vector space of A is a direct sum $A = \bigoplus_{z \in \mathbb{Z}^n} A_z$ of vector spaces, and multiplication satisfies $A_z A_{z'} \subseteq A_{z+z'}$ for all $z, z' \in \mathbb{Z}^n$. In particular, $1 \in A_0$.

A \mathbb{Z}^n -graded A-module is an A-module M, such that the underlying k-vector space of M is a direct sum $M = \bigoplus_{z \in \mathbb{Z}^n} M_j$ of vector spaces, and multiplication satisfies $A_z M_{z'} \subseteq M_{z+z'}$ for all $z, z' \in \mathbb{Z}^n$. A morphism of graded A-modules M and N is a morphism $f: M \to N$ of ordinary A-modules, with the additional property that $f(M_z) \subseteq N_z$ for all $z \in \mathbb{Z}^n$. We write AgMod_{\mathbb{Z}^n} for the category of \mathbb{Z}^n -graded A-modules. The components A_z and M_z are called the graded or homogeneous components of A and M, respectively. An element of A or M is called homogeneous if it lies in a homogeneous component.

We require all graded algebras A we work with to be non-negatively graded, that is, $A_z = 0$ unless $z \ge 0$.

Example 2.3.9. The polynomial algebra $k[x_1, \ldots, x_n]$ is a (non-negatively) \mathbb{Z}^n -graded k-algebra, where for $z \in \mathbb{Z}^n$, the z-graded component is the one-dimensional vector space spanned by the monomial $x_1^{z_1} \cdots x_n^{z_n}$. It is also \mathbb{Z} -graded, where for $m \in \mathbb{Z}$, the vector space A_m is spanned by all monomials of total degree n. For convenience, we use the notation $x^z := x_1^{z_1} \cdots x_n^{z_n}$ for $z = (z_1, \ldots, z_n) \in \mathbb{Z}^n$.

Proposition 2.3.10 ([34, Theorem 1]). The functor

$$F: \operatorname{Vec}^{\mathbf{Z}^n} \to k[x_1, \dots, x_n] \operatorname{-gMod}_{\mathbf{Z}^n}, \qquad F(M) = \bigoplus_{z \in \mathbf{Z}^n} M_z, \tag{2.14}$$

is an equivalence of categories, where for $z, z' \in \mathbb{Z}^n$, the element x^z acts on $m \in F(M)_{z'}$ by $x^z m \coloneqq M_{z+z',z'}(m)$. Its quasi-inverse is the functor

$$G: k[x_1, \dots, x_n] \operatorname{-gMod}_{\mathbf{Z}^n} \to \operatorname{Vec}^{\mathbf{Z}^n}, \qquad G(M)_z = M_z, \qquad (2.15)$$
$$G(M)_{z,z'} = x^{z-z'}|_{M_{z'}}.$$

Every equivalence of abelian categories is exact. In particular, limits and colimits in $\operatorname{Vec}^{\mathbb{Z}^n}$ correspond to limits and colimits in $k[x_1, \ldots, x_n]$ -gMod_{\mathbb{Z}^n}, and free (projective, injective) modules in $\operatorname{Vec}^{\mathbb{Z}^n}$ correspond to free (projective, injective) modules in $k[x_1, \ldots, x_n]$ -gMod_{\mathbb{Z}^n}.

Remark. One can define the functors F and G from Proposition 2.3.10 in greater generality. Namely, if S is a commutative partially ordered monoid and P a partially ordered set with a monotonic action $S \times P \to P$, then there is a fully faithful functor $F: \operatorname{Vec}^P \to k[S]$ -gMod_P defined analogously to (2.14), where k[S] denotes the monoid algebra of S. See [50] for details.

Minimal free resolutions

Definition 2.3.11. A (homological) d-ball is a chain complex of the form

 $\cdots \to 0 \to F(z) \xrightarrow{\mathrm{id}} F(z) \to 0 \to \cdots$

for some z, concentrated in degrees d, d-1. A chain complex of free modules is *trivial* if it is isomorphic to a direct sum of homological balls. It is *minimal* if it contains no homological ball as a direct summand.

Before we explore different equivalent ways of characterizing minimality of chain complexes, we establish the following lemma.

Lemma 2.3.12 (Eliminating balls). Consider a chain complex

$$C_{\bullet}: \qquad \cdots \to K \xrightarrow{\begin{pmatrix} a \\ b \end{pmatrix}} L \oplus M \xrightarrow{\begin{pmatrix} c & a \\ e & f \end{pmatrix}} N \oplus O \xrightarrow{(g h)} P \to \cdots$$

If c is invertible, then $C_{\bullet} \cong C'_{\bullet} \oplus B_{\bullet}$, where

$$C'_{\bullet}: \qquad \cdots \to K \xrightarrow{b} M \xrightarrow{f-ec^{-1}d} O \xrightarrow{h} P \to \cdots$$
$$B_{\bullet}: \qquad \cdots \to 0 \longrightarrow L \xrightarrow{c} N \longrightarrow 0 \to \cdots$$

and the projection $C_{\bullet} \to C'_{\bullet}$ is a quasi-isomorphism.

Remark 2.3.13. Every quasi-isomorphism of bounded below complexes of free modules is a chain homotopy equivalence. The same is true for bounded below complexes of projective or injective modules. This follows from [127, Corollary 1.5.4] and the lifting property of free (resp. projective, injective) modules.

Proof. Consider the automorphisms

$$\begin{split} \eta &= \begin{pmatrix} 1 & c^{-1}d \\ 0 & 1 \end{pmatrix} \in \operatorname{Aut}(L \oplus M), \qquad \qquad \eta^{-1} &= \begin{pmatrix} 1 & -c^{-1}d \\ 0 & 1 \end{pmatrix} \in \operatorname{Aut}(L \oplus M), \\ \theta &= \begin{pmatrix} 1 & 0 \\ -ec^{-1} & 1 \end{pmatrix} \in \operatorname{Aut}(N \oplus O), \qquad \qquad \theta^{-1} &= \begin{pmatrix} 1 & 0 \\ ec^{-1} & 1 \end{pmatrix} \in \operatorname{Aut}(N \oplus O). \end{split}$$

Then $\theta \begin{pmatrix} c & d \\ e & f \end{pmatrix} \eta^{-1} = \begin{pmatrix} c & 0 \\ 0 & f - ec^{-1}d \end{pmatrix}$. Since C_{\bullet} is a chain complex, ca + db = 0 and gc + he = 0. This implies that

$$\eta(\begin{smallmatrix} a \\ b \end{smallmatrix}) = (\begin{smallmatrix} a + c^{-1} db \\ b \end{smallmatrix}) = (\begin{smallmatrix} 0 \\ b \end{smallmatrix}) \qquad (g, h) \theta^{-1} = (g + hec^{-1}, h) = (0, h).$$

This shows that the diagram

$$\begin{array}{cccc} C_{\bullet} \colon & \cdots \to K & & \stackrel{\begin{pmatrix} a \\ b \end{pmatrix}}{\longrightarrow} & L \oplus M & \stackrel{\begin{pmatrix} c & d \\ e & f \end{pmatrix}}{\longrightarrow} & N \oplus O & \stackrel{(g \ h)}{\longrightarrow} & P \to \cdots \\ & & & & & \\ & & & & & \\ & & & & & \\ B_{\bullet} \oplus C'_{\bullet} \colon & \cdots \to K & \stackrel{\begin{pmatrix} 0 \\ b \end{pmatrix}}{\longrightarrow} & L \oplus M & \stackrel{\begin{pmatrix} c & d \\ 0 & f - ec^{-1}d \end{pmatrix}}{\longrightarrow} & N \oplus O & \stackrel{(0,h)}{\longrightarrow} & P \to \cdots . \end{array}$$

commutes and is a pair of mutually inverse isomorphisms of chain complexes. Since $B_{\bullet} = (\cdots 0 \to L \xrightarrow{c} N \to 0 \to \cdots)$ is trivial, the projection $C_{\bullet} \to C'_{\bullet}$ is a quasi-isomorphism. \Box

A graded k-algebra A is called graded local if it has a unique maximal homogeneous ideal \mathfrak{m} , and connected if $A_0 = k$. Every connected k-algebra A is graded local with $\mathfrak{m} = \bigoplus_{i>0} A_i$. As a **Z**- or **Z**ⁿ-graded algebra, the polynomial algebra $k[x_1, \ldots, x_n]$ is connected and thus graded local, with the unique maximal homogeneous ideal $\mathfrak{m} = (x_1, \ldots, x_n)$.

We are now ready to collect some equivalent ways to characterize minimal chain complexes and minimal free resolutions. The idea behind these is that complex of free modules is nonminimal if and only if its boundary morphisms contain an invertible component. Since boundary morphisms are homogeneous and m is the ideal of all homogeneous non-invertible elements $k[x_1, \ldots, x_n]$, we get that a complex is minimal if and only if all components of its boundary morphisms lie in m. The following makes this precise.

Lemma 2.3.14 (Characterizations of minimality). Let A be a connected graded algebra and $(F_{\bullet}, \partial_{\bullet})$ be chain complex of graded free A-modules of finite rank. Then the following are equivalent:

- (i) F_{\bullet} is minimal.
- (ii) For every d, we have $\operatorname{im} \partial_{d+1} \subseteq \mathfrak{m} F_d$.
- (iii) All boundary operators of the complex $F_{\bullet}/\mathfrak{m}F_{\bullet}$ are zero.
- (iv) All boundary operators of the complex $F_{\bullet} \otimes_{k[x_1,...,x_n]} k$ are zero.

In (*iv*), we regard k as an $k[x_1, \ldots, x_n]$ -module with the only non-zero component k in grade 0, on which all indeterminates act by zero. The tensor product is the usual tensor product of graded modules.

Proof. Choose homogeneous bases $(e_j)_{j \in J}$ and $(e'_i)_{i \in I}$ of the free modules F_{d+1} and F_d , respectively, for indexing sets J and I. For every $i \in I$ and $j \in J$, let $f_{ij} \in k[x_1, \ldots, x_n]$ be such that $\partial_{d+1}(e_j) = \sum_{i \in I} f_{ij} e'_j$

 $(i) \Rightarrow (ii)$ Assume that $\partial_{d+1}(a) \notin \mathfrak{m}F_d$ for some $a \in F_{d+1}$. Then there exist also $i_0 \in I$ and $j_0 \in J$ such that $f_{i_0,j_0} \in A \setminus \mathfrak{m} = k^*$. In particular, e_{i_0} and e'_{i_0} are homogeneous of the same degree p, and $f_{i_0,j_0} \colon F(p) \to F(q)$ is an isomorphism. Then Lemma 2.3.12 shows that the d + 1-ball

$$\cdots \to 0 \to F(p) \xrightarrow{J_{i_0,j_0}} F(p) \to 0 \to \cdots$$

is a direct summand of F_{\bullet} , so F_{\bullet} is not minimal.

- $(ii) \Rightarrow (i)$ Assume that there are complexes $(B_{\bullet}, \partial_{\bullet}^B)$ and $(F'_{\bullet}, \partial_{\bullet}^{F'})$ such that $F_{\bullet} \cong B_{\bullet} \oplus F'_{\bullet}$, where B_{\bullet} is a homological d + 1-ball. This means that im $\partial_{d+1}^B = B_d$. Because B_d is finitely generated, supp B_d contains at least one minimal element g. Because all elements of \mathfrak{m} have strictly positive grade, $(\mathfrak{m}B_d)_g = 0$. Therefore, im $\partial_d^B \not\subseteq \mathfrak{m}B_{d-1}$.
- $(ii) \Leftrightarrow (iv)$ If $B_{\bullet} = (\dots \to 0 \to F(z) \xrightarrow{=} F(z) \to 0 \to \dots)$ is a homological ball, then $B_{\bullet} \otimes k = (\dots \to 0 \to k\langle -z \rangle \xrightarrow{=} k\langle -z \rangle \to 0 \to \dots)$, where $k\langle -z \rangle$ is the persistence module with $(k\langle -z \rangle)_w = k$ if w = z, and zero otherwise. In particular, the boundary operator of $B_{\bullet} \otimes k$ is not zero. Thus, F_{\bullet} contains no homological ball as a direct summand if and only if $F_{\bullet} \otimes k$ has no non-zero boundary operator.
- $(iv) \Leftrightarrow (iii)$ This follows by recalling that \mathfrak{m} is the ideal such that $k = k[x_1, \ldots, x_n]/\mathfrak{m}$.

Definition 2.3.15. A free resolution is *minimal* if it is minimal as a chain complex. A free presentation of M is *minimal* if it extends to a minimal free resolution.
Example 2.3.16 (Barcodes). Let $M \in \text{vec}^{\mathbb{Z}^2}$ be finitely generated with generating set $\{m_i | i \in I\}$ for some finite indexing set I, and let barc $M = \{(b_i, d_i) | i \in I\}$. Then

$$\begin{array}{ccc} 0 \to \bigoplus_{i \in I, d_i < \infty} F(d_i) \longrightarrow & \bigoplus_{i \in I} F(b_i) & \stackrel{\varepsilon}{\longrightarrow} M, \\ & e_i \longmapsto F(b_i)_{d_i, b_i}(e'_i) \\ & e'_i \longmapsto g_j \end{array}$$

is a minimal free resolution of M, where e_i and e'_i denote the *i*th standard basis vector of the respective free module. See also Definition 2.2.6. A non-minimal free resolution of M would correspond to a barcode with $b_i = d_i$ for some $i \in I$. In this case, the corresponding generator m_i is not necessary.

Lemma 2.3.17. Let A be connected and $(F_{\bullet}, \partial_{\bullet})$ be a free resolution of an A-module M. Then F_{\bullet} is minimal if and only if the augmented resolution $F_{\bullet} \xrightarrow{\partial_0} M \to 0$ (see Definition 2.3.4) has the property that for every $d \geq 0$, a basis of F_d is mapped to a minimal generating system of $\operatorname{im} \partial_d$.

Proof. According to Lemma 2.3.14(*ii*), F_{\bullet} is minimal if and only if $\operatorname{im} \partial_{d+1} \subseteq \mathfrak{m} F_d$ for all $d \ge 0$. Fix a $d \ge 0$, and let $(e_i)_{i \in I}$ be a homogeneous basis of F_d for some indexing set I.

Assume that $(\partial_d(e_i))_{i\in I}$ is a non-minimal generating set of $\operatorname{im} \partial_d$. Then there exists an $i_0 \in I$ and homogeneous polynomials $f_i \in k[x_1, \ldots, x_n]$ for all $i \neq i_0$ such that $\partial_d(e_{i_0}) = \sum_{i\neq i_0} f_i \partial(e_i)$. Then $g \coloneqq e_{i_0} - \sum_{i\neq i_0} f_i e_i$ satisfies $\partial_d(g) = 0$. By exactness, we get $g \in \operatorname{im} \partial_{d+1}$, but $g \notin \mathfrak{m} F_d$ because $e_{i_0} \in F_d \setminus \mathfrak{m} F_d$.

Conversely, assume $\operatorname{im} \partial_{d+1} \nsubseteq \mathfrak{m} F_d$. Then there exists a non-zero homogeneous element $g \in \operatorname{im} \partial_{d+1} \setminus \mathfrak{m} F_d$. For $i \in I$, let $f_i \in k[x_1, \ldots, x_n]$ be the homogeneous polynomials such that $g = \sum_{i \in I} f_i e_i$. Since $g \in F_d \setminus \mathfrak{m} F_d$, there exists an $i_0 \in I$ such that $f_{i_0} \in A \setminus \mathfrak{m} = k^*$. Since $\partial_g(g) = 0$, we get $\partial_d(e_{i_0}) = -f_{i_0}^{-1} \sum_{i \neq i_0} f_i \partial_d(e_i)$, so the generating system $(\partial_d(e_i))_{i \in I}$ of $\operatorname{im} \partial_d$ is not minimal.

The algebra $k[x_1, \ldots, x_n]$ is Noetherian, which implies that every submodule of a finitely generated module is finitely generated again. Therefore, every finitely generated module has a minimal free resolution, obtained by choosing a minimal generating system $\partial_0: F_0 \to M$ of M and a minimal generating system $\partial_{d+1}: F_{d+1} \to \ker \partial_d$ of $\ker \partial_d$ for every $d \ge 0$. The thus obtained resolution is essentially unique:

Theorem 2.3.18 (Uniqueness of minimal free resolutions [65, p. 491]). Let $M \in k[x_1, \ldots, x_n]$ gMod_{Zⁿ} be finitely generated and F_{\bullet} be a minimal free resolution of M. For every other free resolution F'_{\bullet} of M, there exists a trivial chain complex F''_{\bullet} such that $F'_{\bullet} \cong F_{\bullet} \oplus F''_{\bullet}$. In particular, all minimal free resolutions of M are isomorphic.

Theorem 2.3.19 (Hilbert's syzygy theorem [110, p. 56, 65, p. 474]). Every minimal free resolution of a finitely generated module $M \in k[x_1, \ldots, x_n]$ -gMod_{**Z**ⁿ} has length at most n.

Proof (sketch). Let $R = k[x_1, \ldots, x_n]$. For $i = 1, \ldots, n$, consider the chain complexes $\Omega_i : 0 \to F(e_i) \xrightarrow{x_i} F(0) \to 0$ of \mathbb{Z}^n -graded free R-modules concentrated in degrees 1 and 0, where $e_i \in \mathbb{Z}^n$ is the *i*th unit vector. The Koszul complex $\Omega_{\bullet} := \bigotimes_{i=1}^n \Omega_i$ is a \mathbb{Z}^n -graded minimal free resolution of length n of the graded R-module $k \cong F(0)/\mathfrak{m}F(0)$; here, \otimes denotes the usual tensor product of chain complexes of modules. For a finitely generated module $M \in k[x_1, \ldots, x_n]$ -gMod_{\mathbb{Z}^n} and a \mathbb{Z}^n -graded minimal free resolution F_{\bullet} of M, we obtain that

$$F_i \otimes_R k \stackrel{2.3.14(iv)}{\cong} H_i(F_{\bullet} \otimes_R k) \stackrel{(*)}{=} \operatorname{Tor}_i^R(M,k) \stackrel{(*)}{=} H_i(M \otimes_R \Omega_{\bullet}) = 0$$

if i > n because Ω_{\bullet} has length n and M (as a chain complex) has length 0, using the definition of Tor in (*). Now each F_i is finitely generated, and $F_i \otimes_R k = 0$ for all i > n. This implies that $F_i = 0$ for all i > n, so F_{\bullet} has length n.

Corollary 2.3.20. If $F_{\bullet} = (F_{n-1} \xrightarrow{\partial_{n-1}} F_{n-2} \rightarrow \cdots \xrightarrow{\partial_1} F_0)$ is an exact sequence of graded free $k[x_1, \ldots, x_n]$ -modules, then ker ∂_{n-1} is free.

Proof. The sequence $0 \to \ker \partial_{n-1} \to F_{n-1} \to F_{n-2} \to \cdots \to F_0 \to \operatorname{coker} \partial_1 \to 0$ is exact. We show that it is a free resolution of $M \coloneqq \operatorname{coker} \partial_1$. We extend F_{\bullet} to a free resolution $F_{\bullet} \colon \cdots \to F_n \to F_{n-1} \to \cdots \to F_0$ of M. According to Theorem 2.3.18, F_{\bullet} is isomorphic to a direct sum of a minimal free resolution F'_{\bullet} of M and a trivial complex, so

$$F_{\bullet} \cong (\dots \to \underbrace{B_3 \oplus F_2' \oplus B_2}_{F_2} \to \underbrace{B_2 \oplus F_1' \oplus B_1}_{F_1} \to \underbrace{B_1 \oplus F_0'}_{F_0} \to 0)$$

for free modules B_1, B_2, \ldots Then

$$0 \to F'_n \oplus B_n \to \underbrace{F'_{n-1} \oplus B_n \oplus B_{n-1}}_{F_{n-1}} \to F_{n-2} \to \dots \to F_0 \to M \to 0$$

is a free resolution of M. By the five lemma, ker $\partial_{n-1} \cong F'_n \oplus B_n$, which is free.

2.3.4 Computing minimal free resolutions

Let C_{\bullet} be a chain complex of finite rank free \mathbb{Z}^n -persistence modules. We describe an algorithm that computes a minimal complex quasi-isomorphic to C_{\bullet} . Let $C_d = \bigoplus_{k=1}^{n_d} F(z_{dk})$ for every d, and let D_d be the graded matrix representing ∂_d .

Definition 2.3.21. Let D be a valid graded $m \times n$ -matrix and $i \leq m$, and $j \leq n$. We call (i, j) a *local pair* of D if $[D]_{ij} \neq 0$ and $\operatorname{rg}_i^D = \operatorname{cg}_j^D$. In this case, we call i a local row index and j a local column index of D. We call a row or column *local* if it is part of a local pair. We call D minimal if it has no local pairs.

This definition is motivated by the following lemma:

Lemma 2.3.22. The complex C_{\bullet} is minimal if and only if D_d is minimal for every d.

Proof. Let $D: F \to F'$ be a valid graded matrix representing a morphism of free modules F and F' with bases $(e_j)_{j\in J}$ and $(e'_i)_{i\in I}$ for index sets I and J, respectively. Then $D \subseteq \mathfrak{m}F'$ if and only if D has no local pairs. To see this, note that $De_j = \sum_i f_{ij}e'_i$ for all i, j, for polynomials $f_{ij} = x^{\operatorname{cg}_j^D - \operatorname{rg}_i^D} D_{ij} \in k[x_1, \ldots, x_n]$. If D has a local pair (i_0, j_0) , then f_{ij} is invertible, so im $D \not\subseteq \mathfrak{m}F'$. The converse is similar. The statement then follows from Lemma 2.3.14(ii). \Box

If D_d has a local pair (i_0, j_0) , then we apply Lemma 2.3.12 with $c = [D_d]_{i_0, j_0}$. Define the graded matrices

$$\begin{aligned} D'_{d+1} &\coloneqq ([D_{d+1}]_{ij})_{i \neq i_0}, \\ D'_{d} &\coloneqq ([D_d]_{ij} - \frac{[D_d]_{i_0j}[D_d]_{ij_0}}{[D_d]_{i_0j_0}})_{i \neq i_0, j \neq j_0}, \\ D'_{d-1} &\coloneqq ([D_{d-1}]_{ij})_{j \neq j_0}, \end{aligned}$$

and $D'_d \coloneqq D_d$ in all other cases. Lemma 2.3.12 shows that these represent a chain complex

$$\cdots \to C_{d+1} \xrightarrow{D'_{d+1}} \bigoplus_{k \neq j_0} F(z_{dk}) \xrightarrow{D'_d} \bigoplus_{k \neq i_0} F(z_{d-1,k}) \xrightarrow{D'_{d-1}} C_{d-2} \to \cdots$$
(2.16)

Algorithm 3: Minimization: removes all local pairs from a valid graded matrix M. [71, 86, 99].

Input: A valid graded $m \times n$ -matrix M such that $\operatorname{rg}_i^M \not\geq \operatorname{rg}_{j'}^M$ and $\operatorname{cg}_j^M \not\geq \operatorname{cg}_{j'}^M$ for all i < i' and j < j'. **Output:** A graded matrix M', non-local row and column indices r, c of M.

function Minimize(M): $p \leftarrow 0 \in \mathbf{N}^m$ $c \leftarrow \emptyset$ for j = 1, ..., n do identify local pairs forever do $i \leftarrow \mathsf{piv}(M_j)$ if $rg_i^M \neq cg_j^M$ then $c \leftarrow c \cup \{j\}$; break else if $p_i = 0$ then $p_i \leftarrow j$; break ▷ column j non-local ▷ column j local with pivot i else $[M]_j \leftarrow [M]_j - \frac{[M]_{ij}}{[M]_{ip_i}} [M]_{p_i}$ for $j \in c$ do embarrassingly parallel while $\{i \mid D_{ij} \text{ and } p_i \neq 0\} \neq \emptyset$ do > remove entries in local rows $i \leftarrow \max\{i \mid D_{ij} \neq 0 \text{ and } p_i \neq 0\}$ $i \leftarrow \max\{i \mid D_{ij} \neq 0 \text{ and } p_i \neq 0\}$ $[M]_j \leftarrow [M]_j - \frac{[M]_{ij}}{[M]_{ip_j}}[M]_{p_i}$ $r \leftarrow \{i \mid p_i = 0\}$ return $(M_{ij})_{i \in r, j \in c}$, r, c

quasi-isomorphic to C_{\bullet} . Repeating this procedure for every matrix D_d until there are no more local pairs in the chain complex yields a minimal complex C'_{\bullet} quasi-isomorphic to C_{\bullet} .

For the following, assume that the columns of D_d are ordered non-decreasingly by grade; that is, D_d satisfies $\operatorname{rg}_i^{D_d} < \operatorname{rg}_{i'}^{D_d}$ implies i < i' and $\operatorname{cg}_j^{D_d} < \operatorname{cg}_{j'}^{D_d}$ for all j < j'.

Proposition 2.3.23 (Elimination of homological d-balls [71, 86, 99]). Let C_{\bullet} be a chain complex of finite rank free modules, represented by graded matrices D_{\bullet} that have the row and column grades in non-decreasing order. Then Algorithm 3 computes $(D'_d, r, c) \coloneqq \text{Minimize}(D_d)$, where D'_d is a minimal graded matrix and r and c are subsets of \mathbf{N} , such that the graded matrices

 $\dots, D_{d+2}, \quad D_{d+3}, \quad ([D_{d+1}]_{ij})_{i \in c}, \quad D'_d, \quad ([D_{d-1}]_{ij})_{j \in r}, \quad D_{d-2}, \quad D_{d+3}, \dots$ (2.17)

represent a chain complex C'_{\bullet} quasi-isomorphic to C_{\bullet} .

Remark. Besides the minimal matrix D', the algorithm returns the set r of all row indices i and the set c of all column indices j that do not belong to a local pair (i, -) or (-, j). These sets are necessary to truncate the matrices adjacent to D in a chain complex as in (2.17).

Proof of Proposition 2.3.23 (idea). Since the algorithm only acts on D_d , we only write D for D_d as in the algorithm. That the rows columns of D are non-decreasingly ordered by grade ensures that a column $[D]_j$ is local if and only if $(\text{piv}[D]_j, j)$ is a local pair. Namely, by validity, no non-zero entry in $[D]_j$ can have a row grade not smaller or equal than cg_j^D , and by the assumption, no non-zero entry in $[D]_j$ can have a row grade larger than the pivot grade.

The algorithm proceeds in two phases. In the first phase, the algorithm identifies a maximal set of pairwise disjoint local pairs, which corresponds to a maximal trivial direct summand of D. To do so, it reduces each column $[D]_j$ using column additions from columns $[D]_{j_0}$ that are already identified as local. Let $i = \text{piv}[D]_j$. The reduction of $[D]_j$ stops either if $\text{rg}_i^D \neq \text{cg}_j^D$ (in this case, $[D]_j$ is non-local), or $\text{rg}_i^D = \text{cg}_j^D$ but no other local column has the same pivot (in this case, (i, j) is identified as a local pair). In the second phase, the algorithm performs column additions from the local to the non-local columns in order to remove all entries in the local rows.

To verify correctness, we note that the algorithm repeatedly performs the construction from (2.16). Lemma 2.3.12 implies that the computed chain complex C'_{\bullet} is quasi-isomorphic and hence (by Remark 2.3.13) homotopy equivalent to C_{\bullet} . One verifies that D' = Minimize(D) has no local pairs and thus is minimal.

- **Corollary 2.3.24.** (i) Let D_1, D_2, \ldots, D_n be graded matrices that represent a chain complex C_{\bullet} of free n-parameter persistence modules. Then Algorithm 4 computes graded matrices D'_1, D'_2, \ldots that represent a minimal chain complex homotopy equivalent to C_{\bullet} .
 - (ii) Let D^1, D^2, \ldots, D^n be graded matrices that represent a cochain complex C^{\bullet} of free nparameter persistence modules. Then Algorithm 5 computes graded matrices D'^1, D'^2, \ldots that represent a minimal cochain complex homotopy equivalent to C^{\bullet} .

Proof. Algorithm 4 applies Algorithm 3 to repeatedly. According to Proposition 2.3.23, this produces a sequence of homotopy equivalent chain complexes

for free modules $C'_d = F(\operatorname{cg}^{D''_d})$ and $\tilde{C}_d = F(\operatorname{cg}^{\tilde{D}_d})$, where

$$\begin{split} D'_d &\coloneqq \begin{cases} D_d & \text{if } d = 1, \\ ([D_d]_{ij})_{d \in r_{d-1}} & \text{if } d > 1 \end{cases} \\ (D''_d, r_d, c_d) &\coloneqq \texttt{Minimize}(D'_d), \\ \tilde{D}_d &\coloneqq ([D'_d]_{ij})_{j \in r_{d+1}} \end{split}$$

It follows from Proposition 2.3.23 that each of the vertical maps of chain complexes is a homotopy equivalence indeed, and that each computed matrix D''_d is minimal, in the sense of Definition 2.3.21. The matrices \tilde{D}_d are obtained from D''_d by deleting certain columns. This cannot introduce new local pairs, so also all of the matrices \tilde{D}_d are minimal. By Lemma 2.3.22, the matrices \tilde{D}_d represent a minimal chain complex \tilde{C}_{\bullet} . For cochain complexes, the proof is analogous.

Remark. A chain complex C_{\bullet} of free modules is minimal if and only if its dual cochain complex C^{\bullet} is. If C_{\bullet} is a cochain complex, we two have two alternatives to produce a minimal chain complex homotopy equivalent to C_{\bullet} ; namely, by applying Algorithm 4 to the boundary matrices D_d of C_{\bullet} (which performs column operations on the matrices D_d), or by applying Algorithm 5 to the coboundary matrices $(D_d)^{\top}$ of C^{\bullet} (which can be seen as performing row operations on the matrices D_d).

Algorithm 4: Minimization of chain complexes. Input: Graded matrices D_1, D_2, \ldots forming a chain complex C_{\bullet} . **Output:** Graded matrices D'_1, D'_2, \ldots forming a minimal chain complex C'_{\bullet} quasi-isomorphic to C_{\bullet} . $D'_1, r, c \leftarrow \texttt{Minimize}(D_1)$ for d = 2, ..., do D'_d $\leftarrow ([D_d]_{ij})_{i \in r}$ $D'_d, r, c \leftarrow \texttt{Minimize}(D'_d)$ $D'_{d-1} \leftarrow ([D_{d-1}]_{ij})_{j \in r}$ yield D'_{d-1}

Algorithm 5: Minimization of cochain complexes.

Input: Graded matrices D^1, D^2, \ldots forming a cochain complex C^{\bullet} **Output**: Graded matrices D'^1, D'^2, \ldots forming a minimal cochain complex C'^{\bullet} quasi-isomorphic to C^{\bullet} . $D^{\prime 1}, r, c \leftarrow \texttt{Minimize}(D^1)$ for d = 2, ... do D'^d $\leftarrow ([D^d]_{ij})_{j \in c}$ $D^{\prime d}, r, c \leftarrow \texttt{Minimize}(D^{\prime d})$ $D'^{d-1} \leftarrow ([D_{d-1}]_{ij})_{i \in c}$ yield D'^{d-1}

2.4 Computing two-parameter persistent homology

In this section, we explain how to compute graded matrices representing a minimal free resolution of $H_d(K_*)$ for each d, where $K_* \in \text{Simp}^{\subseteq \mathbb{Z}^n}$ is a finite two-parameter filtered simplicial complex. If K_* is one-critically filtered, $C_{\bullet}(K_*)$ is a chain complex of free modules in vec^{**Z**ⁿ}. Now, the task is to compute a minimal free resolution of $H_d(C_{\bullet})$ for each d, where C_{\bullet} is a chain complex of free \mathbb{Z}^n -persistence modules.

This is commonly solved by Gröbner basis methods; for example, the algorithms [21, 67, 93] can be used for this task. Gröbner base methods have been previously studied in the context of persistent homology [33, 119]. A specific algorithm to compute a minimal free presentation or resolution of $H_d(C_{\bullet})$ for the case n=2 has been proposed in [99]. It relies on the fact that in this case, according to Corollary 2.3.20, $Z_d(C_{\bullet})$ is free for every d. The algorithm [99], which is part of the software package Rivet [123], has been reported to outperform the implementation of the Gröbner base algorithm [67, 93] in SINGULAR, MACAULAY 2 and COCOA [99]. The algorithm has been further improved in [73, 86] and implemented in the software mpfree [85]. We refer to this algorithm as the Lesnick-Wright- or LW-algorithm.

In the rest of this chapter, we explain how to compute a minimal free resolution of $H_d(C_{\bullet})$, following [73, 86, 99].

Remark 2.4.1 (Multi-critical filtrations). If K_* is m-critically \mathbb{Z}^2 -filtered for m > 1, then $C_{\bullet}(K_*)$ need not be a chain complex of free modules. However, one can define a chain complex C'_{\bullet} of free modules that is quasi-isomorphic to $C_{\bullet}(K_*)$ [36], where each d-simplex of K_* contributes m generators of C'_d and m-1 generators of C'_{d+1} . An analogous complex can be constructed if K_* is m-critically \mathbb{Z}^n -filtered. Namely, assume

the d-simplex σ enters the filtration at $g_{\sigma,1}, \ldots, g_{\sigma,m} \in \mathbb{Z}^n$. Define the Koszul complex

$$\Omega_{\bullet}(\sigma) \coloneqq \bigotimes_{i=1}^{m} (0 \to F(g_{\sigma,i}) \to F(0) \to 0),$$

of σ , where the two free modules inside the parentheses are placed in degrees zero and one. Here, the tensor product denotes the usual tensor product of chain complexes of graded modules.

For a chain complex C_{\bullet} , let $\operatorname{tr}_{\geq i} C_{\bullet}$ be the truncated chain complex with $(\operatorname{tr}_{\geq i} C_{\bullet})_d \coloneqq C_d$ if $d \geq i$, zero otherwise, and the boundary morphisms from C_{\bullet} . Then

$$\operatorname{tr}_{\geq 1}(\Omega_{\bullet}(\sigma))$$

is a chain complex of free modules quasi-isomorphic to the submodule of $C_d(K_*)$ generated by σ . Then one may define a double complex \tilde{C} of persistence modules, whose rows are

$$\tilde{C}_p \bigoplus_{\sigma \in K^p_*} \operatorname{tr}_{\geq 1}(\Omega_{\bullet}(\sigma))$$

and whose vertical boundary maps come from the ones of $C_{\bullet}(K_*)$. Taking its total complex yields a free chain complex quasi-isomorphic to $C_{\bullet}(K_*)$.

2.4.1 The Lesnick–Wright algorithm

Let C_{\bullet} be a chain complex of finite rank free \mathbb{Z}^2 -persistence modules with boundary morphisms $\partial_d \colon C_d \to C_{d-1}$ represented by valid graded matrices D_d . For every d, let $i_d \colon Z_d(C_{\bullet}) \hookrightarrow C_d$ be the canonical inclusion. Because C_{\bullet} is a chain complex, there exists a unique morphism $p_d \colon C_d \to Z_{d-1}(C_{\bullet})$ such that $\partial_d = i_{d-1}p_d$. Corollary 2.3.20 implies that the kernel of a morphism of free modules of finite rank is free. In particular, the module $Z_d(C_{\bullet}) = \ker \partial_d$ is free for every d, so

$$0 \to Z_{d+1}(C_{\bullet}) \xrightarrow{\imath_{d+1}} C_{d+1} \xrightarrow{p_{d+1}} Z_d(C_{\bullet})$$
(2.18)

is a free resolution of $H_d(C_{\bullet})$. We compute matrices I_{d+1} and P_{d+1} that represent the morphisms i_{d+1} and p_{d+1} .

Definition 2.4.2. The *lexicographic order* \leq_{lex} and the *colexicographic order* \leq_{colex} on \mathbb{Z}^2 are the total orders given by

$$\begin{array}{l} z \preceq_{\mathrm{lex}} z' & :\Leftrightarrow x \leq x' \text{ or } (x = x' \text{ and } y \leq y'), \\ z \preceq_{\mathrm{colex}} z' & :\Leftrightarrow y \leq y' \text{ or } (y = y' \text{ and } x \leq x'). \end{array}$$

for z = (x, y) and z' = (x', y').

We have $z \leq z'$ if and only if $z \preceq_{\text{lex}} z'$ and $z \preceq_{\text{colex}} z'$.

Theorem 2.4.3 ([86, 99]). Let $M: F \to G$ be a valid \mathbb{Z}^2 graded matrix, such that the column grades of M are colexicographically ordered. Then Algorithm 6 computes a graded matrix K that represents a basis of the free submodule ker M of F.

Algorithm 6 computes the columns of K ordered lexicographically by grade. The role of the lexicographic and the colexicographic ordering in the algorithm can be exchanged.

Remark. Algorithm 6 does not consider the basis grades of G. This makes sense, because ker $M = \ker JM$ for any free module H and every valid injective graded matrix $J: G \to H$.

We may compute the matrices I_{d+1} and P_{d+1} as follows. Consider the commutative diagram

$$0 \longrightarrow Z_{d+1}(C_{\bullet}) \xrightarrow{i_{d+1}} C_{d+1} \xrightarrow{p_{d+1}} Z_d(C_{\bullet}) \rightarrow H_d(C_{\bullet}) \longrightarrow 0$$

$$\downarrow i_d (1)$$

$$\partial_{d+1} \rightarrow C_d.$$
(2.19)

Applying Algorithm 6 to D_d and D_{d+1} yields graded matrices $I_d = \text{Ker}(D_d)$ and $I_{d+1} = \text{Ker}(D_{d+1})$ representing i_d and i_{d+1} , respectively; see (1). We have $D_d D_{d+1} = 0$. Because I_{d+1} represents the inclusion ker $D_d \to C_d$, there is a unique valid graded matrix $P_{d+1} : C_{d+1} \to C_d$ such that $D_{d+1} = I_d P_{d+1}$. This linear system can be solved for P_{d+1} , for example, using Algorithm 7; see (2). Then the matrices P_{d+1} and I_{d+1} represent the free resolution (2.18). Applying Algorithm 4 gives a minimal free resolution of $H_d(C_{\bullet})$.

Remark. Although Algorithm 7 does not take the grading into account, it follows from existence and uniqueness of P_{d+1} that the result is a valid graded matrix.

Computing a minimal generating system of im ∂_{d+1} **first** It turns out that solving the linear system $D_{d+1} = I_d P_{d+1}$ for P_{d+1} does some unnecessary computation, as we will now see. Namely, if D_{d+1} is non-minimal (in the sense of Definition 2.3.21), then local pairs of D_{d+1} will give local pairs of P_{d+1} . Therefore, in this case, solving $D_{d+1} = I_d P_{d+1}$ computes columns of P_{d+1} that are discarded again when minimizing the resolution (Algorithm 4). [99] mentions the following construction that avoids this.

Consider again the commutative diagram in (2.19). We observe that $0 \to Z_{d+1}(C_{\bullet}) \to C_{d+1}$ is a free resolution of im ∂_{d+1} . Using Algorithm 3, one may compute a minimal free resolution

$$0 \to Z'_{d+1} \xrightarrow{i'_{d+1}} C'_{d+1} \to \operatorname{im} \partial_{d+1}$$
(2.20)

of $\operatorname{im} \partial_{d+1}$ and let $\partial'_{d+1} \colon C'_{d+1} \to C_d$ such that $\operatorname{im} \partial'_{d+1} = \operatorname{im} \partial_{d+1}$; see 2 in (2.21) below. Actually, computing i_{d+1} and passing to the minimal free resolution (2.20) of $\operatorname{im} \partial_{d+1}$ can be done in one step by a variant of Algorithm 6:

Proposition 2.4.4 ([86, 99]). Let $M: F \to G$ be a valid graded matrix representing a morphism of free \mathbb{Z}^2 -persistence modules. Then Algorithm 8 computes graded matrices M', K' such that M' represents a minimal generating system $M': F' \to G$ of im M, and K' represents a basis of ker M'.

Because im $\partial'_{d+1} = \operatorname{im} \partial_{d+1}$, we have $\partial_d \partial'_{d+1} = 0$, so there exists a unique morphism p'_{d+1} such that $\partial'_{d+1} = i_d p'_{d+1}$; see \mathfrak{F} in the following commutative diagram:

$$0 \to Z_{d+1}(C_{\bullet}) \xrightarrow{i_{d+1}} C_{d+1} \xrightarrow{p_{d+1}} Z_d(C_{\bullet}) \longrightarrow H_d(C_{\bullet}) \to 0$$

$$0 \longrightarrow Z'_{d+1} \xrightarrow{\swarrow} C'_{d+1} \xrightarrow{\swarrow} Z_d(C_{\bullet}) \xrightarrow{\uparrow} H_d(C_{\bullet}) \to 0$$

$$(2.21)$$

$$(2.21)$$

$$(2.21)$$

Now, the sequence

$$0 \to Z'_{d+1} \xrightarrow{i'_{d+1}} C'_{d+1} \xrightarrow{p'_{d+1}} Z_d(C_{\bullet})$$
(2.22)

is a free resolution of $H_d(C_{\bullet})$.

Because (2.20) is a minimal resolution, (2.22) contains no homological 2-balls. Analogously to [98], we call such a resolution a *semi-minimal free resolution* of $H_d(C_{\bullet})$. To obtain a minimal free resolution of $H_d(C_{\bullet})$, it remains to eliminate all homological 1-balls by applying Algorithm 3 to P'_{d+1} . Together, we obtain the following strategy:

The following steps compute a free resolution of $H_d(C_{\bullet})$:

- 1. Compute a graded matrix $I_d := \text{Ker}(D_d)$ (Algorithm 6) representing the kernel inclusion $i_d: Z_d(C_{\bullet}) \hookrightarrow C_d(C_{\bullet}); see (1).$
- 2. Compute graded matrices $(D'_{d+1}, I'_{d+1}) := \text{MGSWithKer}(D_{d+1})$ (Algorithm 8) such that D'_{d+1} represents a minimal generating system of $B_d(C_{\bullet})$, and I'_{d+1} represents ker D'_{d+1} ; see (2').
- 3. Compute the unique graded matrix $P'_{d+1} \coloneqq \texttt{Factorize}(D'_{d+1}, I_d)$ (Algorithm 7) such that $I_d P'_{d+1} = D'_{d+1}$ that represents p'_{d+1} ; see (3).

Algorithm 6: LW-algorithm. For a valid matrix $M: F \to G$ representing a morphism of free \mathbb{Z}^2 -persistence modules, the algorithm computes a basis K of ker $M \subseteq F$ [86, 99].

A valid graded $m \times n$ -matrix M with cg^M ordered colexicographically. Input: **Output**: A reduced graded matrix K representing a basis of ker M with cg^{K} ordered lexicographically. function Ker(M): $Q \leftarrow \{(\mathsf{cg}_j^M, j)\} \text{ as priority queue with } (z, j) \leq (z', j') :\Leftrightarrow z \preceq_{\mathsf{lex}} z' \lor (z = z' \land j \leq j')$ $V \leftarrow E \in k^{n \times n}$ ▷ reduction matrix $K \leftarrow 0 \in k^{n \times 0}$ with $rg^K = cg^M$ $p \leftarrow 0 \in \mathbf{N}^m$ \triangleright if $p_i \neq 0$, then $i = \text{piv } M_{p_i}$ while $Q \neq \emptyset$ do $(z,j) \leftarrow \texttt{PopMin}(Q)$ forever do $i \leftarrow \text{piv} M_i$ if i = 0 then append V_i to K and append z to cg^K; break else if $p_i = 0$ then $p_i \leftarrow j$; break else if $\operatorname{cg}_{p_i}^M \not\leq z$ then append $(\operatorname{cg}_{p_i}^M \lor z, j)$ to Q and $p_i \leftarrow j$; break $V_j \leftarrow V_j - M_{ij}/M_{ip_i}V_{p_i}$ $M_j \leftarrow M_j - M_{ij}/M_{ip_i}M_{p_i}$ return K

Algorithm 7: Factorization of matrices

Input:A matrix $L \in k^{\ell \times n}$, and a reduced matrix $M \in k^{\ell \times m}$ Output:The matrix $N \in k^{m \times n}$ such that M = LN, if it exists.function Factorize(L, M): $p \leftarrow 0 \in \mathbb{N}^{\ell}$ $p \leftarrow 0 \in \mathbb{N}^{\ell}$ $N \leftarrow 0 \in k^{m \times n}$ for $j = 1, \dots, m$ do $p_{piv M_j} \leftarrow j$ \triangleright embarrassingly parallelwhile $L_j \neq 0$ do \triangleright if $p_i = 0$, then $\nexists N : L = MN$ $N_j \leftarrow N_j - L_{ij}/M_{ip_i}M_{p_i}$ \triangleright if $p_i = 0$, then $\nexists N : L = MN$ return NN

Then I'_{d+1} , P'_{d+1} represent the free resolution of $H_{d+1}(C_{\bullet})$ from (2.22). It contains no homological 2-balls as direct summands.

 To obtain a minimal free resolution, split off all homological 1-balls using Minimize() (Algorithm 3).

The procedures Ker() and MGSWithKer() (Algorithms 6 and 8) can be combined into a single algorithm KerAndMgsWithKer(). Given a valid graded matrix M, it computes a matrix K representing kerM, a matrix M' representing a minimal generating system of im M, and a matrix K' representing ker M'. This can be used to compute minimal free resolutions for $H_0(C_{\bullet}), H_1(C_{\bullet}), \ldots$; see Algorithm 9.

```
Algorithm 8: A variant of Algorithm 6. For a morphism f: F \to G of free modules, computes minimal
generating system f' \colon F' \to G of im f, together with ker f'.
              A graded m \times n-matrix M representing f.
Input:
Output: Graded matrices M' representing f' and K' representing a basis of ker f'.
function MGSWithKer(M):
     Q \leftarrow \{(cg_j^M, j)\} as priority queue with (z, j) \leq (z', j') if z \preceq_{lex} z' or z = z' and j \leq j'
     M' \leftarrow \mathsf{graded} \ m \times 0\text{-matrix with } \mathrm{rg}^{M'} = \mathrm{rg}^{M}
     V' \leftarrow \mathsf{empty matrix}
     K' \leftarrow \mathsf{graded} \ \mathsf{0} \times \mathsf{0}\text{-matrix} \ \mathsf{with} \ \mathsf{rg}^{K'} = \mathsf{cg}^{M'}
     m \leftarrow 0 \in \mathbf{N}^n
                                                                                                                 \triangleright index of M_i in M'
     p \leftarrow 0 \in \mathbf{N}^m
                                                                                                                 ▷ pivot row to column
     while Q \neq \emptyset do
                                                                                                                    assignment
           (z,j) \leftarrow \texttt{PopMin}(Q)
           forever do
                 i \leftarrow \operatorname{piv} M_i
                 if i = 0 then
                       if m_j \neq 0 then append V'_{m_j} to K' and z to cg^{K'}
                                                                                                                \triangleright K_{m_i} represents a basis
                                                                                                                    element of ker M'
                       break
                 else if p_i = 0 then p_i \leftarrow j; break
                 else if cg_{p_i}^M \not\leq z then
add (cg_{p_i}^M \lor z, j) to Q
                       p_i \leftarrow j
                       break
                 else
                       if z \neq \mathrm{cg}_j^M then V'_{m_j} \leftarrow V'_{m_j} - M_{ij}/M_{i p_i} V'_{m_{p_i}}
                      M_j \leftarrow M_j - M_{ij}/M_{ip_i}M_{p_i}
           if M_j \neq 0 and z = cg_j^M then
                                                                                                                 \triangleright M_j cannot be reduced to
                 m_j \leftarrow \# \text{ columns in } M' + 1
                                                                                                                    zero at cg_i^M and therefore is
                 append M_j to M'_{m_j}
                                                                                                                    an element of the min. gen.
                 V' \leftarrow \begin{pmatrix} V' & 0 \\ 0 & 1 \end{pmatrix} \\ K' \leftarrow \begin{pmatrix} K' \\ 0 \end{pmatrix}
                                                                                                                    system.
                 append z to cg^{M'} and rg_{m_i}^{K'}
     return M', K'
```

```
Algorithm 9: Homology algorithm. Given valid graded matrices D_1, \ldots, D_{\ell+1} representing a chain complex C_{\bullet} of free \mathbb{Z}^2-modules, computes matrices F_{d,1}, F_{d,0} representing a minimal free resolution of H_d(C_{\bullet}) for each d = 0, \ldots, \ell.
```

 $\begin{array}{ll} I_1, D_1', I_1' \leftarrow \texttt{KerAndMgsWithKer}(D_1) \\ F_{0,1}, F_{0,0} \leftarrow \texttt{MinimizeCpx}(I_1', D_1') \\ \texttt{for } d = 1, \dots, \ell \texttt{ do} \\ & \left| \begin{array}{c} I_{d+1}, D_{d+1}', I_{d+1}' \leftarrow \texttt{KerAndMgsWithKer}(D_{d+1}) \\ P_{d+1} & \leftarrow \texttt{Factorize}(D_{d+1}', I_d) \\ F_{d,1}, F_{d,0} & \leftarrow \texttt{MinimizeCpx}(I_{d+1}', P_{d+1}) \end{array} \right. \end{array}$

▷ see Algorithm 4

2.4.2 Notes on implementations

Currently, the fastest implementation of the computation of minimal free presentations or resolutions of two-parameter persistent homology is mpfree [85]. This software applies chunk preprocessing (see below) to its input, before computing a minimal free presentation or resolution using the above approach [73, 86].

Chunk preprocessing The procedure MinimizeCpx() (Algorithm 4) can be used as an efficient preprocessing step to the actual computation of a resolution of persistent homology. This preprocessing is called *chunk preprocessing*. Its benefit lies in the fact that many complexes, such as Vietoris-Rips complexes or function-Rips complexes, contain many local pairs. The procedure Minimize() (Algorithm 3), which does the heavy lifting inside MinimizeCpx(), can be efficiently parallelized (see below), which is not the case for the LW-algorithm. In two parameter persistence, this preprocessing considerably increases the efficiency of the entire computation [72, 73]. In [72, Table 1], chunk preprocessing has been observed to outperform an older approach (see [117]) to decreasing the input size based on discrete Morse theory.

We propose and evaluate an analogous preprocessing scheme based on Algorithm 5; see the procedure MinimizeCpx*() in Section 5.1.6. Recently, other preprocessing steps that simplify the input simplicial complex have been proposed [2].

Parallelization The second phase in Minimize() (Algorithm 3) can be implemented in an embarrassingly parallel way since no two non-local columns affect each other. The first phase only performs column operations $[D]_j \leftarrow [D_j] + [D_{p_i}]$ if $cg_j^D = cg_{p_i}^D$. This implies that if the column indices of D are partitioned into *chunks* of common column grade, the chunks can be processed in parallel in the first phase, hence the name *chunk preprocessing* [11]. In practice, however, the first phase of Minimize() has been observed to account for only a small fraction of the total run time of Minimize(), and the benefit from parallelizing it has been observed to be relatively limited. The procedure Factorize() (Algorithm 7) can be implemented in an embarrassingly parallel way, too, although it only accounts for a small fraction of the total run time of the entire homology resolution computation (Algorithm 9). There is no known way to parallelize Ker() or KerAndMgsWithKer() (Algorithms 6 and 8), which dominate the runtime of Algorithm 9.

Minimal free resolutions vs. presentations Often (e.g., in [73, 86, 98]), one is interested in computing a minimal free presentation of persistent homology instead of a minimal free resolution. However, for \mathbb{Z}^2 -persistence modules, it is not less work to compute the latter than to compute the former. To see this, consider the free resolution (2.18) of $H_d(C_{\bullet})$. A free presentation of $H_d(C_{\bullet})$ is given by the morphism $p_{d+1}: C_{d+1} \to Z_d(C_{\bullet})$. To remove all direct summands of p_{d+1} of the form $G \to 0$, it is necessary to compute ker $p_{d+1} = Z_{d+1}(C_{\bullet})$ and use this to remove certain columns of the matrix representing p_{d+1} (which is what KerAndMgsWithKer() (Algorithm 8) does). Therefore, to compute a minimal free presentation of a the persistent homology of \mathbb{Z}^2 -filtered complex one has to carry out the same computations as for computing a minimal free resolution.

Row and column orders As mentioned above (see Proposition 2.3.23), Minimize() (Algorithm 3) requires the input graded matrix M to have row and column grades in non-descending order. This requirement is satisfied if rg^M and cg^M are ordered in a total order refining the partial order \leq on \mathbb{Z}^2 . In the homology resolution computation (Algorithm 9), Minimize() is applied to the matrix P'_d . which in turn is computed from I_d and D'_{d+1} using Factorize() (Algorithm 7). The matrices I_d and D'_{d+1} are computed using Ker() and KerAndMgsWithKer() (Algorithms 6 and 8). Ker() ensures that I_d has columns in lexicographic order by grade. Ker() and KerAndMgsWithKer() do not require any specific order on the row and column grades of its input D_d and D_{d+1} . We may thus assume that all boundary matrices have their rows and and columns ordered colexicographically by grade. Then KerAndMgsWithKer() computes D'_{d+1} with columns in colexicographic order, too, which together implies that Factorize() computes

 P'_d with columns in colexicographic order (from D'_{d+1}) and rows in lexicographic order (from I_d). Thus, Minimize() can be applied directly to P'_d ; i.e., there is no re-ordering needed in Algorithm 9.

Two-parameter persistent cohomology

In the following two chapters, we introduce different approaches to computing the persistent cohomology of a finite one-critical two-parameter filtered simplicial complex K_* . Specifically, we explore different ways to compute a minimal free resolution of $H_{\bullet}(K_*)$, using cohomology as an intermediate step. Our goal is to apply this to two-parameter clique filtrations such as function-Rips complexes. Using cohomology is motivated by the effectiveness of clearing in the computation of one-parameter persistent cohomology of Vietoris-Rips complexes.

Chapter 3 presents an approach that computes two-parameter persistent (co)homology by computing the cohomology of a certain cochain complex $N^{\bullet}(K_*)$ of free modules. In general, $N^{\bullet}(K_*)$ is different from $C^{\bullet}(K_*)$ and $C^{\bullet}(K, K_*)$, but relates to these by a certain property of vec^{\mathbb{Z}^n} called the *Calabi-Yau property*, see Section 3.2. This property is crucial both for an efficient computation of a minimal free resolution of $H^{\bullet}(N^{\bullet}(K_*))$, and for obtaining a minimal free resolution of $H_{\bullet}(K_*)$ from it. Parts of this approach can be generalized to an arbitrary number of parameters. In particular, for one parameter, this approach gives back the usual computation of relative persistent cohomology. The approach can be implemented efficiently; see Section 5.1.

In Chapter 4, we explain how minimal free resolutions of $H^{\bullet}(K_*)$ and $H^{\bullet}(K, K_*)$ can be computed directly from the cochain complexes $C^{\bullet}(K_*)$ and $C^{\bullet}(K, K_*)$. For two and more parameters, these are no complexes of free modules, and the main part of Chapter 4 lies in remedying this. While this approach arguably is conceptually simpler than the aforementioned one, when it comes to implementing it, it comes with certain drawbacks that we explain in Section 4.5.

Chapter 3

Persistent cohomology using the Calabi–Yau property

Let K_* be a finite one-critical \mathbb{Z}^n -filtered simplicial complex, for any $n \geq 1$. Recall from Section 2.2.1 that in one-parameter persistence, the relative cochain complex $C^{\bullet}(K, K_*)$ is a cochain complex of free modules. In this section, we construct a cochain complex $N^{\bullet}(K_*)$ of free persistence modules such that n = 1 yields $N^{\bullet}(K_*) = C^{\bullet}(K, K_*)$. The construction of $N^{\bullet}(K_*)$ can thus be seen as a way to generalize the definition of $C^{\bullet}(K, K_*)$ to more than one parameter.

We prove that if $H_d(K_*)$ is finitely supported for all d, then $H_d(K_*) \cong H^{d+n}(N^{\bullet}(K_*))^*$; see Theorem A. This implies that duality gives a correspondence between injective resolutions of $H^{d+n}(N^{\bullet}(K_*))$ and free resolutions of $H_d(K_*)$. We show that under the same conditions, there is a direct correspondence between injective and free resolutions of $H^{d+n}(N^{\bullet}(K_*))$ (or any other finite dimensional module); see Corollary 3.2.11 and Theorem E. For n = 2, we devise an efficient method to compute a minimal free resolution of $H^{d+2}(N_{\bullet})$ that involves a scheme akin to clearing, see Section 3.5. This method underlies the implementation presented in Section 5.1.

Parts of the results presented in this chapter are joint work with Ulrich Bauer and Michael Lesnick, and have appeared in [14].

3.1 The free cochain complex $N^{\bullet}(K_*)$

Definition 3.1.1. For a module M and $z \in \mathbb{Z}^n$, let $M\langle z \rangle$ be the module with graded components $M\langle z \rangle_w = M_{z+w}$. For $z \ge 0$, the structure maps of M give a morphism $M \to M\langle z \rangle$. Note that $M\langle z \rangle^* = M^*\langle -z \rangle$. For a graded matrix A, let $A\langle z \rangle$ be the graded matrix with $A\langle z \rangle_{ij} = A_{ij}$, $\operatorname{rg}_i^{A\langle z \rangle} = \operatorname{rg}_i^A - z$ and $\operatorname{cg}_j^{A\langle z \rangle} = \operatorname{rg}_j^A - z$ for all i, j.

This ensures that if $A: F \to F'$ is a valid graded matrix representing a morphism of free modules, then $A\langle z \rangle \colon F\langle z \rangle \to F'\langle z' \rangle$. Let K_* be a one-critical \mathbb{Z}^n -filtered simplicial complex. We fix an ordering on the simplices of K_* and endow $C_{\bullet}(K_*) = \bigoplus_{\sigma \in K_*} F(g(\sigma))$ with the standard basis. Let the graded matrix D_{\bullet} represent the boundary operator ∂_{\bullet} . Let $\epsilon = (1, \ldots, 1) \in \mathbb{Z}^n$.

Definition 3.1.2. Let $N^{\bullet}(K_*)$ be the cochain complex of free modules

$$N^d(K_*) = \bigoplus_{\sigma \in K^d_*} F(\epsilon - g(\sigma)),$$

such that the coboundary operator $\delta^d \colon N^{d-1}(K_*) \to N^d(K_*)$ is represented by the graded matrix $D^d \coloneqq D_d^{\top}$ for every d.

Example 3.1.3. If n = 1, then $N^{\bullet}(K_*) = C^{\bullet}(K, K_*)$.

According to Remark 2.1.19, D^d is valid, so δ^d is a well-defined morphism for every d. Our first goal is to prove the following statement:



Figure 3.1: A free module $F = F(z_1) \oplus F(z_2) \oplus F(z_3)$ (left) and its image $\nu F = I(z_1) \oplus I(z_2) \oplus I(z_3)$ under the Nakayama functor (right)

Theorem A. A If $H_{\bullet}(K_*)$ has finite support, then there is a natural isomorphism $H_d(K_*) \cong H^{d+n}(N^{\bullet}(K_*))^*$ for all d.

We defer the proof of Theorem A to the next subsection; see page 45. The condition that $H_{\bullet}(K_*)$ has finite total dimension equivalently means that $H_d(K_*)$ is pointwise finite dimensional for all d and has finite support in \mathbb{Z}^n . Once we have proven Theorem A, we obtain:

Corollary 3.1.4. If $H_{\bullet}(K_*)$ has finite support and F_{\bullet} is a free resolution of $H^{d+n}(N^{\bullet}(K_*))$, then $(F_{\bullet})^*$ is an injective resolution of $H_d(K_*)$.

Proof. Since the duality $(-)^* : \operatorname{Vec}^{\mathbb{Z}^n} \to \operatorname{Vec}^{\mathbb{Z}^n}$ is contravariant exact, it maps free resolutions to injective resolutions and vice versa.

3.2 The Calabi-Yau-property of persistence modules

Assuming for now that Theorem A has been established, Corollary 3.1.4 gives us an injective resolution of $H_d(K_*)$. It remains to compute from this the *free* resolution of $H_d(K_*)$ we are interested in. Both Theorem A and a simple correspondence between free and injective resolutions of the same module will follow from Theorem 3.2.8, which is closely related to a property known in some areas of algebra as the *Calabi–Yau property* of a triangulated category [77].

Definition 3.2.1. For persistence modules M, N, let Hom(M, N) be the persistence module with components $Hom(M, N)_z = Hom(M, N\langle z \rangle)$. The structure maps of Hom(M, N) are induced by $N \to N\langle z \rangle$.

For $z \in \mathbb{Z}^n$, we define the injective module $I(z) = F(-z)^*$; that is,

$$I(z)_w = \begin{cases} k & \text{if } w \leq z, \\ 0 & \text{otherwise,} \end{cases} \qquad \qquad I(z)_{w \leq w'} = \begin{cases} \text{id if } w' \leq z, \\ 0 & \text{otherwise.} \end{cases}$$

Let $\mathcal{P}_{\text{Vec}\mathbf{z}^n}$ and $\mathcal{I}_{\text{Vec}\mathbf{z}^n}$ be the full subcategories of $\text{Vec}^{\mathbf{Z}^n}$ consisting of finite rank free modules and their dual modules, respectively. Note that $\mathcal{P}_{\text{Vec}\mathbf{z}^n}$ contains all point-wise finite dimensional projective modules modules. Note also that all modules in $\mathcal{I}_{\text{Vec}\mathbf{z}^n}$ are injective, but $\mathcal{I}_{\text{Vec}\mathbf{z}^n}$ does not comprise all injective modules, not even all point-wise finite dimensional ones. Both categories are closed under finite direct sums.

Definition 3.2.2. The functors

$$\nu = \operatorname{Hom}(-, F(0))^* \colon \mathcal{P}_{\operatorname{Vec}} \mathbf{z}^n \longrightarrow \mathcal{I}_{\operatorname{Vec}} \mathbf{z}^n, \\ \nu' = \operatorname{Hom}(I(0)^*, -) \colon \mathcal{I}_{\operatorname{Vec}} \mathbf{z}^n \longrightarrow \mathcal{P}_{\operatorname{Vec}} \mathbf{z}^n$$

are called the Nakayama and the inverse Nakayama functor.

Lemma 3.2.3 ([3, p. 84]). The functors ν and ν' are mutually quasi-inverse equivalences of categories.

This follows from the following:

Lemma 3.2.4. We have $\nu F(z) \cong I(z)$ and $\nu^{-1}I(z) \cong F(z)$, cf. Figure 3.1.

Proof. We note

$$\left(\operatorname{Hom}(F(z),F(0))^*\right)_w = \left(\operatorname{Hom}(F(z),F(0))_{-w}\right)^* = \operatorname{Hom}(F(z),F(w))^* \cong \left\{\begin{smallmatrix} k & \text{if } z \ge w, \\ 0 & \text{otherwise.} \end{smallmatrix}\right\}$$

For $z \ge w$, let $\iota_{wz} \colon F(z) \hookrightarrow F(w)$ be the morphism with $(\iota_{wz})_v = \mathrm{id}_k$ for all $v \ge z$. Then the isomorphism $\nu F(z) \cong I(z)$ is given in graded components by

$$\phi \colon \operatorname{Hom}(F(z), F(w))^* \longrightarrow I(z)_w = k,$$
$$\psi \longmapsto \psi(\iota_{wz}).$$

The statement $\nu^{-1}I(z) \cong F(z)$ follows analogously.

In particular, we obtain

$$N^{\bullet}(K_*) = (\nu C_{\bullet}(K_*) \langle \epsilon \rangle)^*.$$

Remark. If n = 1 and $K_* \in \text{Simp}^{\subseteq \mathbb{Z}}$ is a finite filtered complex, then $C^{\bullet}(K_*, K) = \nu C_{\bullet}(K_*)$.

Definition 3.2.5. For a chain complex C_{\bullet} and $i \in \mathbb{Z}$, let $C_{\bullet}[i]$ be the chain complex with d-dimensional component $(C_{\bullet}[i])_d = C_{i+d}$. Analogously, for a cochain complex C^{\bullet} , let $C^{\bullet}[i]$ be the cochain complex with $(C^{\bullet}[i])^d = C^{i+d}$. We have $C_{\bullet}[i]^* = (C_{\bullet})^*[i]$.

For $z \in \mathbb{Z}^n$, we write $z = (z_1, \ldots, z_n)$. For $n \in \mathbb{N}$, let $[n] \coloneqq \{1, \ldots, n\}$ and let $\binom{[n]}{k} \coloneqq \{S \subseteq [n] \mid |S| = k\}$.

Definition 3.2.6. For $S = \{s_1 < \ldots < s_k\} \in {\binom{[n]}{k}}$, we define the functors

$$\operatorname{colim}_{S} \colon \operatorname{Vec}^{\mathbf{Z}^{n}} \to \operatorname{Vec}^{\mathbf{Z}^{n-k}}, \quad (\operatorname{colim}_{S} M)_{(z_{1},...,\hat{z}_{s_{1}},...,\hat{z}_{s_{k}},...,z_{k})} = \operatorname{colim}_{(z_{s_{1}},...,z_{s_{k}})\in\mathbf{Z}^{k}} M_{z},$$

$$\Delta_{S} \colon \operatorname{Vec}^{\mathbf{Z}^{n-k}} \to \operatorname{Vec}^{\mathbf{Z}^{n}}, \quad (\Delta_{S} M)_{(z_{1},...,z_{n})} = M_{(z_{1},...,\hat{z}_{s_{1}},...,\hat{z}_{s_{n-k}},...,z_{n})},$$

$$\lim_{S} \colon \operatorname{Vec}^{\mathbf{Z}^{n}} \to \operatorname{Vec}^{\mathbf{Z}^{n-k}}, \quad (\lim_{S} M)_{(z_{1},...,\hat{z}_{s_{1}},...,\hat{z}_{s_{k}},...,z_{k})} = \lim_{(z_{s_{1}},...,z_{s_{k}})\in\mathbf{Z}^{k}} M_{z},$$

where $\hat{z}_{s_1}, \ldots, \hat{z}_{s_k}$ are omitted. We also define $\operatorname{Colim}_S \coloneqq \Delta_S \operatorname{colim}_S$ and $\operatorname{Lim}_S \coloneqq \Delta_S \operatorname{Lim}_S$.

That is, $\operatorname{colim}_S M$ is the (n-|S|)-parameter module obtained by taking the colimit of M with respect to the axes indexed by S. For example, if n = 3 and $S = \{1, 3\}$, then $(\operatorname{colim}_{\{1,3\}} M)_z = \operatorname{colim}_{(z_1, z_3) \in \mathbb{Z}^2} M_{(z_1, z, z_3)}$. The module $\Delta_S M$ (and thus also $\operatorname{Colim}_S M$ and $\operatorname{Lim}_S M$) is constant along the coordinate axes specified by S. The functor Δ_S , called the *S*-diagonal, is exact, right adjoint to colim_S and left adjoint to lim_S .

Definition 3.2.7. A chain complex F_{\bullet} in $\operatorname{Vec}^{\mathbb{Z}^n}$ is called *eventually acyclic* if $\operatorname{colim}_S F_{\bullet}$ is acyclic for all non-empty $S \subseteq [n]$. A \mathbb{Z}^n -filtered simplicial simplicial complex K_* is *eventually acyclic* if $\operatorname{colim}_S K_*$ is acyclic for all non-empty $S \subseteq [n]$.

A chain complex of free persistence modules is eventually acyclic if and only if supp $H_d(F_{\bullet})$ is bounded for all d.

Theorem 3.2.8 (Calabi–Yau property of persistence modules). If F_{\bullet} is an eventually acyclic complex of free \mathbb{Z}^n -persistence modules, then F_{\bullet} and $\nu F_{\bullet}[n]\langle \epsilon \rangle$ are naturally quasi-isomorphic.

Before proving Theorem 3.2.8, we introduce some additional terminology that we will also use later.

For a module M, we define the modules

$$\Omega_k M = \bigoplus_{S \in \binom{[n]}{n-k}} \operatorname{Colim}_S M$$



Figure 3.2: The Koszul complex from (3.2) for n = 2. The sequences are exact, and the module above the arrow denotes the image of that morphism.

for each $0 \leq k \leq n$. If $S' \subseteq S$, then there is a canonical morphism $c_{S,S'}$ Colim_{S'} $M \to \text{Colim}_S M$. These give rise to morphisms

$$\Omega_k M = \bigoplus_{S' \in \binom{[n]}{n-k}} \operatorname{Colim}_{S'} M \xrightarrow{\kappa_k} \bigoplus_{S \in \binom{[n]}{n-k+1}} \operatorname{Colim}_S M = \Omega_{k-1} M$$

that are defined by their non-zero components

$$(-1)^{j}c_{S,S\setminus\{s_{i}\}}\colon \operatorname{Colim}_{S\setminus\{s_{i}\}}M\to\operatorname{Colim}_{S}M$$

for $S = (s_1 < \cdots < s_j < \cdots < s_{n-k})$. The sign rule ensures that $(\Omega_{\bullet}M, \kappa_{\bullet})$ is a chain complex.

Lemma 3.2.9. The assignment $M \mapsto \Omega_{\bullet} M$ is a functor.

Definition 3.2.10. We call the functor Ω_{\bullet} the Koszul complex functor.

Remark. The complex from Remark 2.4.1 is not an instance of this functor, but follows the same combinatorics. There exist several different but closely related notions of "the" Koszul complex.

Proof of Theorem 3.2.8. For a free module F(z), the sequence

$$0 \to \Omega_n F(z) \to \Omega_{n-1} F(z) \to \dots \to \Omega_0 F(z) \to \nu F(z) \langle \epsilon \rangle.$$
(3.1)

is exact. The last morphism is the canonical morphism

$$\Omega_0 F(z) = \operatorname{Colim} F(z) = \operatorname{Lim} I(z) \langle \epsilon \rangle \to I(z) \langle \epsilon \rangle = \nu F(z) \langle \epsilon \rangle.$$

One may check (e.g., using [65, Corollary 6.6 or Theorem A6.6]) that all modules $\Omega_k F(z)$ are flat¹, so (3.1) sequence is a flat resolution of $\nu F(z)\langle\epsilon\rangle$. see Figure 3.2. Let F_{\bullet} be a bounded complex of free modules. Using functoriality of Ω_{\bullet} and ν , we get an exact sequence

$$\Omega_{\bullet}F_{\bullet}: \quad 0 \to \underbrace{\Omega_{n}F_{\bullet}}_{F_{\bullet}} \to \Omega_{n-1}F_{\bullet} \to \dots \to \Omega_{0}F_{\bullet} \to \nu F_{\bullet}\langle\epsilon\rangle \to 0 \tag{3.2}$$

of chain complexes, given by taking a shifted copy of (3.1) for every summand F(z) in a direct sum decomposition of F_{\bullet} . We unsplice (3.2) into short exact sequences



¹A graded module F is flat if the graded module tensor product functor $-\otimes F$ is exact.

with chain complexes $U_{\bullet}^{(k)}$ for each k. In the derived category $\mathcal{D}^{\mathrm{b}}(\mathrm{Vec}^{\mathbf{Z}^n})$ (see Section 3.2.1 below for the notion of the derived category), each of these short exact sequences gives a triangle [127, §10.4.9]. We obtain connecting homomorphisms

$$\partial^{(n-1)} \colon U_{\bullet}^{(n-1)}[1] \to F_{\bullet}, \quad \partial^{(n-2)} \colon U_{\bullet}^{(n-2)}[1] \to U_{\bullet}^{(n-1)}, \quad \dots \quad \partial^{(0)} \colon \nu F_{\bullet}\langle \epsilon \rangle[1] \to U_{\bullet}^{(1)}$$

in $\mathcal{D}^{\mathrm{b}}(\mathrm{Vec}^{\mathbf{Z}^n})$. These descend to maps fitting into the long exact sequences

$$\cdots \to H_{d+1}(\Omega_{n-1}F_{\bullet}) \to H_{d+1}(U_{\bullet}^{(n-1)}) \xrightarrow{\partial^{(n-1)}} H_d(F_{\bullet}) \longrightarrow H_d(\Omega_{n-1}F_{\bullet}) \longrightarrow \cdots,$$

$$\cdots \to H_{d+2}(\Omega_{n-2}F_{\bullet}) \to H_{d+2}(U_{\bullet}^{(n-2)}) \xrightarrow{\partial^{(n-2)}} H_{d+1}(U_{\bullet}^{(n-1)}) \to H_{d+1}(\Omega_{n-2}F_{\bullet}) \to \cdots,$$

$$\cdots \to H_{d+n}(\Omega_0F_{\bullet}) \longrightarrow H_{d+n}(\nu F_{\bullet}\langle\epsilon\rangle) \xrightarrow{\partial^{(0)}} H_{d+n-1}(U_{\bullet}^{(1)}) \to H_{d+n-1}(\Omega_0F_{\bullet}) \to \cdots,$$

$$(3.4)$$

induced by the short exact sequences (3.3). By assumption, $\operatorname{Colim}_S H_d(F_{\bullet}) = 0$ for all d and for all S with |S| > 0. The functor Colim_S is exact for all S because it is a directed colimit. In particular,

$$H_d(\Omega_k F_{\bullet}) = \bigoplus_{|S|=k} H_d(\operatorname{Colim}_S F_{\bullet}) = 0$$

for all k < n. Therefore, the long exact sequences (3.4) show that all connecting homomorphisms $\partial^{(k)}$ are quasi-isomorphisms. Thus,

$$\partial^{(n-1)} \circ \cdots \circ \partial^{(0)} \colon \nu F_{\bullet}[n] \langle \epsilon \rangle \longrightarrow F_{\bullet}$$

is a quasi-isomorphism.

A proof in a more general context can be found in [84, Lemma 4.1].

Proof of Theorem A. With $N^{\bullet}(K_*) = (\nu C_{\bullet}(K_*)\langle \epsilon \rangle)^*$, Theorem 3.2.8 gives

$$H^{d+n}(N^{\bullet}(K_*))^* \cong H_d(N^{\bullet}(K_*)^*[n]) \cong H_d(\nu C_{\bullet}(K_*)\langle \epsilon \rangle [n]) \cong H_d(C_{\bullet}(K_*)) = H_d(K_*). \quad \Box$$

Corollary 3.2.11. Let $M \in \text{Vec}^{\mathbb{Z}^n}$ be finitely supported.

- (i) If F_{\bullet} is a free resolution of M, then $\nu F_{\bullet}[n]\langle \epsilon \rangle$ is an injective resolution of M.
- (ii) If I^{\bullet} is an injective resolution of M, then $\nu^{-1}I^{\bullet}[-n]\langle -\epsilon \rangle$ is a free resolution of M.
- (iii) If F_{\bullet} is a free resolution of M, then $(\nu F_{\bullet}[n]\langle \epsilon \rangle)^*$ is a free resolution of M^* .

3.2.1 Remarks on Theorem 3.2.8

One-parameter persistence For n = 1, we have $C_{\bullet}(K, K_*) \cong \nu C_{\bullet}(K_*)\langle 1 \rangle$. If K is acyclic, then Theorem 3.2.8 gives back the isomorphism $H_d(K_*) \cong H_{d+1}(K, K_*)$ we already obtained from (2.2).

Relation to derived categories Recall that the *derived category* $\mathcal{D}(\mathcal{C})$ of an abelian category \mathcal{C} is the triangulated category whose objects are chain complexes of persistence modules, and whose morphisms are $\operatorname{Hom}_{\mathcal{D}(\mathcal{C})}(X, Z) = \{X_{\bullet} \stackrel{\simeq}{\leftarrow} Y_{\bullet} \to Z_{\bullet}\}/\sim$, where $X_{\bullet} \stackrel{\simeq}{\leftarrow} Y_{\bullet}$ is a quasiisomorphism and \sim is a certain equivalence relation. That is, the derived category is obtained via formally inverting quasi-isomorphisms by imposing a calculus of fractions [124], analogous to the construction of localization of rings; see [127, §10] for details. The *bounded derived category* $\mathcal{D}^{b}(\operatorname{Vec}^{\mathbb{Z}^{n}})$ is the full subcategory of $\mathcal{D}(\operatorname{Vec}^{\mathbb{Z}^{n}})$ of chain com-

The bounded derived category $\mathcal{D}^{\mathrm{b}}(\operatorname{Vec}^{\mathbb{Z}^{-}})$ is the full subcategory of $\mathcal{D}(\operatorname{Vec}^{\mathbb{Z}^{-}})$ of chain complexes C_{\bullet} whose homology has finite total dimension [84, §4.1]; that is, $H_d(C_{\bullet})_z = 0$ for all but finitely many d and z. Let $\mathcal{K}^{\mathrm{b}}(\mathcal{P}_{\operatorname{Vec}^{C}})$ and $\mathcal{K}^{\mathrm{b}}(\mathcal{I}_{\operatorname{Vec}^{C}})$ denote the respective bounded homotopy



(a) Same as Figure 3.2, but shifted by $2z - \epsilon$. The coordinates drawn in opposite direction. If $C^{\bullet}(K_*) = \bigoplus_i I(-z_i)$, then $C^{\bullet}(K, K_*)$ is a direct sum of shifted copies of the module drawn over the middle arrow.



(b) The Koszul complex from (3.6). Dual of (a). If $C_{\bullet}(K_*) = \bigoplus_i F(z_i)$, then $C_{\bullet}(K, K_*)$ is a direct sum of shifted copies of the module drawn over the middle arrow.

Figure 3.3: Illustration of the variations of the Koszul complex for n = 2. The sequences are exact, and the module above the arrow denotes the image of that morphism.

categories; that is, $\mathcal{K}^{\mathrm{b}}(\mathcal{P}_{\mathrm{Vec}^{C}})$ (and analogously, $\mathcal{K}^{\mathrm{b}}(\mathcal{I}_{\mathrm{Vec}^{C}})$) is the triangulated category whose objects are chain complexes of projective (resp., injective) persistence modules with homology of finite total dimension, and whose morphisms are chain homotopy classes of morphisms of chain complexes. Recall that $\mathcal{D}^{\mathrm{b}}(\mathcal{C})$, $\mathcal{K}^{\mathrm{b}}(\mathcal{P}_{\mathrm{Vec}^{C}})$ and $\mathcal{K}^{\mathrm{b}}(\mathcal{I}_{\mathrm{Vec}^{C}})$ are equivalent as triangulated categories. We note:

- 1. Theorem 3.2.8 states that $\nu \cong [-n]\langle -\epsilon \rangle$ as endo-functors of $\mathcal{D}^{\mathrm{b}}(\mathrm{Vec}^{\mathbb{Z}^n})$.
- 2. The Koszul complex $\Omega_{\bullet}F(0)$ concentrated in degrees $n, \ldots, 0$ satisfies $-\otimes \Omega_{\bullet}F(0) \cong \nu \langle \epsilon \rangle$ as endo-functors of $\mathcal{D}^{\mathrm{b}}(\operatorname{Vec}^{\mathbb{Z}^n})$. Note that there exist different but similar definitions of "the" Koszul complex, which all follow the same combinatorics, e.g., the complex defined in Remark 2.4.1 or the one from [84, §4.2].
- 3. A triangulated category \mathcal{T} is n-Calabi-Yau if [n] is a Serre functor [47, §2, 77]; that is, if $\operatorname{Hom}_{\mathcal{T}}(A_{\bullet}, B_{\bullet})^* \cong \operatorname{Hom}_{\mathcal{T}}(B_{\bullet}, A_{\bullet}[n])$ for all A_{\bullet}, B_{\bullet} . One can show that [n] is a Serre functor if $\nu \cong [-n]$; see [84, Lemma 4.1]. In [84, §4.2], it is shown that $\mathcal{D}^{\mathrm{b}}(k[x_1, \ldots, x_n])$ is *n*-Calabi-Yau, using a slightly different Koszul complex than we do. Theorem 3.2.8 can be viewed as a \mathbb{Z}^n -graded version of this.

Relation to relative cochains We can also state a dual version of Theorem 3.2.8:

Theorem 3.2.12. If F^{\bullet} is a cochain complex of free modules such that $H^{\bullet}(F^{\bullet})$ is finitely supported, then $F^{\bullet} \simeq \nu F^{\bullet} \langle \epsilon \rangle [-n]$.

Proof. The proof is analogous to the proof of Theorem 3.2.8, with the commutative diagram



of cochain complexes.

Corollary 3.2.13. If $\operatorname{Colim}_S K_*$ is acyclic for all $S \in {\binom{[n]}{k}}$ for 0 < k < n, then

$$H^{d+n}(N^{\bullet}(K_*)) \cong H^{d+1}(K, K_*)$$

Proof. With $F^{\bullet} = N^{\bullet}(K_*)$ and $\nu N^{\bullet} \langle \epsilon \rangle = C^{\bullet}(K_*)$, we obtain $U^{\bullet}_{(1)} = C^{\bullet}(K, K_*)$; see Figure 3.3a. From (3.5), we obtain the sequence

$$H^{d+n}(N^{\bullet}(K_*)) \longrightarrow H^{d+n-1}(U^{\bullet}_{(n-1)}) \longrightarrow \cdots \longrightarrow H^{d+1}(K,K_*) \longrightarrow H^d(K_*)$$

of connecting homomorphisms. If $\operatorname{Colim}_S K_*$ is acyclic for all $S \in {\binom{[n]}{k}}$ for 0 < k < n, then $\Omega_k F^{\bullet}$ is acyclic for all 0 < k < n, so this is a sequence of isomorphisms.

Proof using limits Alternatively, Theorem 3.2.8 can be proven using limits instead of colimits. For $1 \le k \le n$, let $\Omega^k M$: $\bigoplus_{S \in \binom{[n]}{n-k}} \lim_{S \to \infty} M$. Then there is an exact sequence

$$0 \to \nu^{-1} I(z) \langle \epsilon \rangle \to \Omega^0 I(z) \langle \epsilon \rangle \to \dots \to \Omega^{n-1} I(z) \langle \epsilon \rangle \to \underbrace{\Omega^n I(z) \langle \epsilon \rangle}_{I(z) \langle \epsilon \rangle} \to 0; \tag{3.6}$$

see Figure 3.3b. Let F_{\bullet} be a chain complex of free modules, and let $I_{\bullet} = \nu F_{\bullet}[n]\langle \epsilon \rangle$ such that $I_{\bullet} \simeq F_{\bullet}$. Taking shifted copies of (3.6) and unsplicing the sequence into short exact sequence analogously to (3.4) yields a commutative diagram



for suitable chain complexes $U'^{(k)}_{\bullet}$. Note that (3.7) is the dual of (3.5).

3.3 The Nakayama functor and matrices

It remains to describe the injective resolution $\nu F_{\bullet}[n]\langle \epsilon \rangle$ of M in terms of matrices representing F_{\bullet} . To that end, we use matrices also to describe morphisms of *injective* modules. We say an injective module I is of *finite type* if $I \in \mathcal{I}_{\text{Vec}^{\text{Vec}}\mathbf{z}^n}$. Note that such a module is of the form $I \cong \bigoplus_{i \in I} I(z_i)$ for some finite indexing set I.

Recall that a basis of a free module F can be identified with the choice of an isomorphism $\bigoplus_{z \in \operatorname{rk} F} F(z) \to F$.

Definition 3.3.1. An *injective basis* of an injective module I of finite type is an isomorphism $I \cong \bigoplus_{i \in I} I(z_i)$.

Lemma 3.3.2. A graded matrix M represents a morphism $\bigoplus_j I(\operatorname{cg}_j^M) \to \bigoplus_i I(\operatorname{rg}_i^M)$ if and only if M is valid.

Proof. This follows from $\text{Hom}(I(z), I(z')) = \begin{cases} k \text{ if } z' \leq z \\ 0 \text{ otherwise,} \end{cases}$.

In particular, a graded matrix represents a morphism of free modules if and only if it represents a morphism of injective modules.

Lemma 3.3.3. A graded matrix represents a morphism $f: \bigoplus_{j=1}^{n} F(z_j) \to \bigoplus_{i=1}^{m} F(z'_i)$ if and only if it represents the morphism $\nu f: \bigoplus_{j=1}^{n} I(z_j) \to \bigoplus_{i=1}^{m} I(z'_i)$.

Proof. The functor ν is additive. Therefore, it suffices to check the statements for morphisms $f: F(z) \to F(z')$. If $z' \leq z$, then $\operatorname{Hom}(F(z), F(z') = \operatorname{Hom}(I(z), I(z')) = 0$, so let $z' \leq z$. Any morphism $f: F(z) \to F(z')$ is of the form $f = \lambda \iota_{z'z}$ for some $\lambda \in k$, where $\iota_{z'z} \iota_{z'z} : F(z) \to F(z')$ is the injective morphism with $(\iota_{z'z})_v = \operatorname{id}_k$ for all $v \geq z$, and zero otherwise. Note that $\iota_{z''z'}\iota_{z'z} = \iota_{z''z}$ for some $\gamma \in k$, where $\pi_{z'z} : I(z) \to I(z')$ is the injective morphism with $(\pi_{z'z})_v = \operatorname{id}_k$ for all $v \geq z$, and zero otherwise. Note that $v \geq z'$, and zero otherwise. Given a morphism $f = \lambda \iota_{z',z}$ for some $\lambda \in k$, we calculate

$$\begin{split} I(z)_w & \xrightarrow{\phi} \nu F(z)_w = \operatorname{Hom}(F(z), F(w))^* \xrightarrow{\nu f} \operatorname{Hom}(F(z'), F(w))^* = \nu F(z') \xrightarrow{\phi^{-1}} I(z'), \\ \mu & \longmapsto \qquad [\phi(\mu) \colon \iota_{wz} \mapsto \mu] \longmapsto [\iota_{wz'} \mapsto \phi(\mu)(\iota_{wz'}f) = \lambda \mu] \quad \longmapsto \lambda \mu, \end{split}$$

where we use the isomorphism $\phi: I(z) \to \nu F(z)$ from Lemma 3.2.4. Hence, $\nu f = \lambda \pi_{z',z}$. \Box

Theorem E. Let M be a finitely generated n-parameter persistence module with bounded support. For graded matrices U_1, \ldots, U_n , the following are equivalent:

- (i) U_1, \ldots, U_n represent a free resolution $0 \to F_n \xrightarrow{U_n} \cdots \xrightarrow{U_1} F_0$ of M,
- (ii) U_1, \ldots, U_n represent an injective resolution of $I^0 \xrightarrow{U_n} \cdots \xrightarrow{U_1} I^n \to 0$ of $M\langle -\epsilon \rangle$,
- (iii) the graded transposes $U_1^{\top}, \ldots, U_n^{\top}$ represent a free resolution $0 \to G_0 \xrightarrow{U_n^{\top}} \cdots \xrightarrow{U_1^{\top}} G_0$ of $M^*\langle \epsilon \rangle$.

In this case $I^q = \nu F_{n-q} = G_q^*$.

Proof. With Lemma 3.3.3, this follows from Theorem 3.2.8.

Example 3.3.4. Let $M \in \text{Vec}^{\mathbb{Z}^2}$ be the module



with components $M_z = k$ if z lies in the shaded region in the following picture, and $M_z = 0$ otherwise. All structure morphisms between non-zero components of M are identities. Here, the symbols \cdot , \bullet and \star denote the grades of the 0-, 1- and 2-syzygies of M, respectively. The sequence



is exact, so G_{\bullet} is a free resolution of M. With the same matrices, the sequence



is exact, so $\nu G_{\bullet}[2]\langle \epsilon \rangle$ is an injective resolution of M.

Lemma 3.3.5. A chain complex C_{\bullet} of free modules is minimal if and only if $(\nu C_{\bullet})^*$ is.

Proof. This follows from the additivity of ν and $(-)^*$. Namely, if $C_{\bullet} \cong C'_{\bullet} \oplus B_{\bullet}$ for a homological d-ball $B = (\cdots \to 0 \to F(z) \xrightarrow{\cong} F(z) \to 0 \to \cdots)$, then we obtain the cochain complex $(\nu C_{\bullet})^* \cong (\nu C'_{\bullet})^* \oplus (\nu B_{\bullet})^*$ with the cohomological d – 1-ball $(\nu B_{\bullet})^* = (\cdots \to 0 \to F(-z) \xrightarrow{\cong} F(-z) \to 0 \to \cdots)$.

3.4 Pulling back modules from the colimit

From now on, we consider \mathbb{Z}^2 -persistence modules only. In this and the following section, we explain how we actually compute $H^{d+2}(N^{\bullet}(K_*))$. In principle, this could be done by a procedure analogous to the one described in Section 2.4. Namely, the dashed sequence in the commutative diagram

is a free resolution of $H^{d+2}(N^{\bullet}(K_*))$. We can obtain matrices I^{d+1}, P^{d+1} representing the morphisms i^{d+1}, p^{d+1} of this resolution with the method described in Section 2.4. However, this would involve the coboundary maps δ^{d+2} and δ^{d+3} , which is not feasible for clique complexes. Instead, we propose a method that computes a free resolution of $H^{d+2}(N^{\bullet}(K_*))$ from δ^{d+1} only.

At the core of this approach lies the observation that the kernel of a map $f: M \to N$ of free persistence modules is determined by the kernel of the morphism colim f of vector spaces, such that ker f can be reconstructed by "pulling back" colim f along the canonical morphism to the colimit. The following makes this precise.

Recall the Functors $\operatorname{Colim} = \Delta \operatorname{colim}$ and $\operatorname{Lim} = \Delta \operatorname{lim} : \operatorname{Vec}^{\mathbb{Z}^n} \to \operatorname{Vec}^{\mathbb{Z}^n}$. For a persistence module M, let $\operatorname{Colim} M = \Delta \operatorname{colim} M$, and let $\eta_M : M \to \operatorname{Colim} M$ be the canonical morphism. Note that η_M is induced by the unit of the adjunction $\operatorname{colim} \dashv \Delta$.

Definition 3.4.1. For a module $M \in \text{Vec}^{\mathbb{Z}^2}$ and a vector space $V \subseteq \text{colim } M$, let $[V]_M := \eta_M^{-1}(\Delta V) \subseteq M$.

The submodule $[V]_M$ is the unique largest submodule of M whose colimit is V:

Lemma 3.4.2. For $M \in \text{Vec}^{\mathbb{Z}^2}$ and $V \subseteq \text{colim } M$, we have

$$[V]_M = \sum \{N \subseteq M \mid \operatorname{colim} N \subseteq V\}.$$

Proof. Let $\mathcal{U} := \{N \subseteq M \mid \operatorname{colim} N \subseteq V\}$ If $x \in [V]_M$, then $\eta_M(x) \in V$. Therefore, the submodule $\langle x \rangle_M$ of M spanned by x satisfies $\operatorname{colim} \langle x \rangle_M = \langle \eta_M(x) \rangle_V \subseteq V$, so $\langle x \rangle_M \in \mathcal{U}$ and thus $x \in \sum \mathcal{U}$. This shows $[V]_M \subseteq \sum \mathcal{U}$. Conversely, let $x \in N \subseteq M$ such that $\operatorname{colim} N = \eta_M(N) \subseteq V$. Then $\eta_M(x) \in V$, so $x \in \eta_M^{-1}(V)$. Because V is closed under addition, this shows $\sum \mathcal{U} \subseteq [V]_M$.

A persistence module is called *torsion free* if all its structure maps are injective. A free module is torsion free.

Theorem F. If $f: M \to N$ is a morphism and N is torsion free, then ker $f = [\operatorname{colim} \ker f]_M$.

Proof. If N is torsion free, then $\eta_N \colon N \to \operatorname{Colim} N$ is injective. Every submodule $L \subseteq M$ satisfies $L \subseteq [\operatorname{colim} L]_M$, so in particular ker $f \subseteq [\operatorname{colim} \ker f]_M$. It remains to show the other inclusion $[\operatorname{colim} \ker f]_M \subseteq \ker f$. Consider the commutative diagram



The functor Colim is a directed colimit and thus exact. Therefore, $\operatorname{Colim} \ker f = \ker \operatorname{Colim} f$. This implies

$$\eta_N f j = (\operatorname{Colim} f)(\operatorname{Colim} i)\eta_{[\operatorname{colim} \ker f]_M} = 0.$$

Since η_N is injective, we obtain fj = 0, so the injective morphism j factors uniquely through ker f. This proves the claim.

The lemma shows that ker f is the largest submodule of M whose colimit is colim ker f. In particular, one can reconstruct ker f from colim ker f.

Corollary 3.4.3. If M is free of finite rank and $V \subseteq \operatorname{colim} M$, then $[V]_M$ is free.

Proof. Consider the canonical projection \tilde{p} : colim $M \to \operatorname{colim} M/V$. Let $z_0 \leq z$ for all $z \in \operatorname{rk} M$, and let N be the persistence module with $N_z = \begin{cases} \operatorname{colim} M/V & \text{if } z_0 \leq z, \\ 0 & \text{otherwise} \end{cases}$ and $N_{z',z} = \operatorname{id}$ for all $z_0 \leq z \leq z'$. Then N is free of graded rank $\{z_0 \mid z \in \operatorname{rk} M\}$ and satisfies colim $N = \operatorname{colim} M/V$. The morphism $p \coloneqq (\Delta \tilde{p})\eta_M \colon M \to N$ satisfies colim $p = \tilde{p}$. From Theorem F, we get that $[V]_M = [\ker \tilde{p}]_M = \ker p$, which is free because it is a kernel of free modules in $\operatorname{Vec}^{\mathbf{Z}^2}$; see Corollary 2.3.20.

In more than two parameters, Corollary 3.4.3 may fail; that is, the pullback $[V]_M$ of $V \subseteq$ colim F may not be free even if M is. If $[V]_M$ is not free, it has a minimal generating system of larger cardinality than dim V.

Recall the lexicographic and colexicographic ordering \leq_{lex} and \leq_{colex} on \mathbb{Z}^2 from Definition 2.4.2. Fix a tuple $\vec{z} \in (\mathbb{Z}^2)^m$ and let $z_i = (x_i, y_i)$ for all i.

Definition 3.4.4. With respect to \vec{z} , the *lex pivot* of a non-zero vector $b \in k^m$, denoted by l-piv_{\vec{z}}(b), is the largest index in $\{i | b_i \neq 0\}$ for which z_i attains its \leq_{lex} -maximal value. The *colex pivot*, denoted by c-piv_{\vec{z}}(b), is defined analogously. For $0 \in k^m$, we let l-piv_{\vec{z}}(0) = c-piv_{\vec{z}}(0) = 0. A $m \times n$ -matrix is *bireduced* with respect to \vec{z} if all its non-zero columns have distinct lex pivots and distinct colex pivots with respect to \vec{z} .

Notation 3.4.5. For a $m \times n$ -matrix B such that $B_j \neq 0$ for all j, let $[B]_{\vec{z}}$ denote the graded matrix with entries $u([B]_{\vec{z}}) = B$, row grades $\operatorname{rg}^{[B]_{\vec{z}}} = \vec{z}$ and column grades $\operatorname{cg}^{[B]_{\vec{z}}}_j = \bigvee_{B_{ij} \neq 0} z_i$.

The graded matrix $[B]_{\vec{z}}$ is valid, and any valid graded matrix \tilde{B} with $u(\tilde{B}) = B$ and $\operatorname{rg}^{\tilde{B}} = \vec{z}$ satisfies $\operatorname{cg}_{j}^{\tilde{B}} \geq \operatorname{cg}_{j}^{[B]_{\vec{z}}}$ for all j. Recall that for any $z, z' \in \mathbb{Z}^{2}$, we have $z \leq z'$ if and only if $z \leq_{\operatorname{lex}} z'$ and $z \leq_{\operatorname{colex}} z'$. This implies that $\operatorname{cg}_{j}^{[B]_{\vec{z}}}$ is determined only by the lex and colex pivot of B_{j} :

$$cg_{j}^{[B]_{\vec{z}}} = \bigvee_{\substack{B_{i,j} \neq 0}} z_{i} = (x_{1\text{-piv}\,B_{j}}, y_{\text{c-piv}\,B_{j}}).$$
(3.9)

Let $\vec{z} = (z_i)_i \in (\mathbb{Z}^2)^n$ be a tuple of grades, let $M = \bigoplus_{i=1}^m F(z_i)$ be a free module, let $(e_i)_i$ denote the standard basis of M, let $V \subseteq \operatorname{colim} M$ be a vector subspace, and let B be a $m \times n$ -matrix that represents a basis of V with respect to the basis $(\eta_M(e_i))_i$ of colim M.

Lemma 3.4.6. If B is bireduced, then $[B]_{\vec{z}}$ represents a basis of $[V]_M$.

Proof. The graded matrix $[B]_{\vec{z}}$ represents a basis of a free submodule $N \subseteq M$ with $\eta_M(N) = V$. We have to show that $N = [V]_M$. Since B is a basis of V, we have $N \subseteq [V]_M$.

For the converse inclusion, let $v \in [V]_M$. We identify v with the unique valid graded column vector that represents v with respect to the basis $(e_i)_i$ of M and $\bar{v} \coloneqq \eta_M(v) \in V$ with the ungraded column vector that represents \bar{v} with respect to the basis $(\eta_M(e_i))_i$ of colim M. Because B is a basis of V, there exists a unique column vector \bar{w} such that $\bar{v} = B\bar{w}$. Let w be the graded column vector with $u(w) = \bar{w}$, row grades $\operatorname{rg}^w = \operatorname{cg}^{[B]_{\bar{z}}}$ and column grade $\operatorname{cg}^w = g(v)$. Then $v = [B]_{\bar{z}}w$. To show that $v \in N$, we have to show that w is valid. Because all columns B have pairwise distinct lex and colex pivots, we have

$$\begin{aligned} &\text{l-piv}_{\vec{z}} v = \text{l-piv}_{\vec{z}} B \bar{w} = \max\{\text{l-piv}_{\vec{z}} B_j \mid w_j \neq 0\}, \\ &\text{c-piv}_{\vec{z}} v = \text{c-piv}_{\vec{z}} B \bar{w} = \max\{\text{c-piv}_{\vec{z}} B_j \mid w_j \neq 0\}. \end{aligned}$$

Therefore, if j is an index such that $w_j \neq 0$, then $l \operatorname{-piv}_{\vec{z}} \bar{v} \geq l \operatorname{-piv}_{\vec{z}} B_j$ and $\operatorname{c-piv}_{\vec{z}} \bar{v} \geq \operatorname{c-piv}_{\vec{z}} B_j$. Because v is a valid graded column vector, we have

$$\operatorname{cg}^{w} = \operatorname{cg}^{v} \ge \bigvee_{v_{i} \neq 0} z_{i} \stackrel{(3.9)}{=} (x_{\operatorname{l-piv} v}, y_{\operatorname{c-piv} v}) \ge (x_{\operatorname{l-piv} B_{j}}, y_{\operatorname{c-piv} B_{j}}) \stackrel{(3.9)}{=} \operatorname{cg}_{j}^{[B]_{\vec{z}}} = \operatorname{rg}_{j}^{w}$$

for every j with $w_j \neq 0$, so w is valid.

Theorem 3.4.7. Let M be a free \mathbb{Z}^2 -persistence module of finite rank with a fixed basis, let $V \subseteq \operatorname{colim} M$ be a subspace, and let B be a matrix representing a generating set of V. Then Algorithm 10 calculates a graded matrix representing a basis of $[V]_M$.

Proof. In each iteration of the first for-loop, the colex-pivot index of one column decreases, so the loop terminates. When it does, all columns have distinct colex-pivots. Line (a) ensures that no column with a larger lex-pivot is added to a column with a smaller one. Therefore, the lex-pivots of the nonzero columns of B do not increase during the first for-loop. Analogously, during each iteration of the second for-loop, the lex-pivot of a column decreases. When the loop terminates, all nonzero columns have distinct lex-pivots. Line (c) ensures that no column

Algorithm 10: Computes a basis of $[V]_M$, where M is free of rank \vec{z} , and $V \subseteq \operatorname{colim} M$.

Input: An $m \times n$ -matrix B representing a generating set of V, $\vec{z} = (z_i)_i \in (\mathbb{Z}^2)^m$. **Output**: A graded $m \times n$ -matrix whose nonzero columns represent a basis of $[V]_M$.

function Bireduce(B): $p \leftarrow 0 \in \mathbf{N}^n$ for j' = 1, ..., n do *i* ← *i*′ while $i \leftarrow \text{c-piv}_{\neq}(B_i) \neq 0$ do if $p_i = 0$ then $p_i = j$; break if $l-piv_{\vec{z}}(B_i) < l-piv_{\vec{z}}(B_{p_i})$ then swap p_i and j (a) $B_i \leftarrow B_i - B_{ij} / B_{ip_i} B_{p_i}$ $p \leftarrow 0 \in \mathbf{N}^n$ (b) for j' = 1, ..., n do $j \leftarrow j'$ while $i \leftarrow 1-\text{piv}_{\neq}(B_i) \neq 0$ do if $p_i = 0$ then $p_i = j$; break if $c-piv_{\vec{z}}(B_j) < c-piv_{\vec{z}}(B_{p_i})$ then swap p_i and j(c) $B_j \leftarrow B_j - B_{ij}/B_{ip_i}B_{p_i}$ return $[B]_{\vec{z}}$

with a larger colex-pivot is added to a column with a smaller one. Since all nonzero columns have distinct colex-pivots after the first for-loop, the colex-pivots of the columns do not change during the second for-loop. Therefore, when the algorithm terminates, all nonzero columns of B have pairwise distinct lex- and colex-pivots, so B is bireduced. The statement then follows from Lemma 3.4.6.

Remark. Let $B \in k^{l \times m}$ and $\vec{z} \in (\mathbf{Z}^2)^l$ as in Notation 3.4.5. The procedure Bireduce() (Algorithm 10) uses the following particularity of \mathbf{Z}^2 : if two columns B_j and B_k have the same lex- or colex-pivot w.r.t. \vec{z} , then the respective columns of $[B]_{\vec{z}}$ have comparable grades. This need not be true for $\vec{z} \in (\mathbf{Z}^n)^l$ for n > 2.

3.5 Computing a free resolution of $H^{\bullet}(N^{\bullet}(K_*))$

In this section, we explain how to use Bireduce() (Algorithm 10) to compute a minimal free resolution of $H^{d+2}(N^{\bullet})$, where N^{\bullet} is an eventually acyclic cochain complex of free twoparameter persistence modules. For example, $N^{\bullet} = N^{\bullet}(K_*)$ if K_* is eventually acyclic. The dashed sequence in the commutative diagram

$$0 \xrightarrow{p^{d}} Z^{d+1}(N^{\bullet}) \xrightarrow{i^{d+1}} N^{d+1} \xrightarrow{p^{d+1}} Z^{d+2}(N^{\bullet}) \xrightarrow{\dots} H^{d+2}(N^{\bullet})$$
(3.10)

is a free resolution of $H^{d+2}(N^{\bullet})$. Chose a basis of N^{\bullet} , and let D^{\bullet} be the graded matrix representing δ^{\bullet} . Because N^{\bullet} is eventually acyclic, the morphism $\operatorname{colim} p^d$: $\operatorname{colim} N^d \to \operatorname{colim} Z^{d+1}(N^{\bullet})$ is surjective. In other words, the ungraded matrix underlying D^{d+1} represents a generating system of $\operatorname{colim} Z^{d+1}(N^{\bullet})$. When applied to D^{d+1} , Algorithm 10 thus computes a graded matrix ${\cal I}^{d+1}$ that represents a basis of the free submodule

$$[\operatorname{colim} Z^{d+1}(N^{\bullet})]_{N^{d+1}} \subseteq N^{d+1}.$$

According to Theorem F, we have

$$[\operatorname{colim} Z^{d+1}(N^{\bullet})]_{N^{d+1}} = Z^{d+1}(N^{\bullet}).$$

Therefore, Bireduce() (Algorithm 10) allows us to compute a basis

 $I^{d+1} := \texttt{Bireduce}(D^{d+1})$

of $Z^{d+1}(N^{\bullet})$ from D^{d+1} ; i.e., a matrix representing the morphism i^{d+1} . It remains to compute a graded matrix P^{d+1} representing the morphism p^{d+1} . This can be done using the following consequence of Corollary 3.2.11:

Corollary 3.5.1. If $0 \to F_n \xrightarrow{f_n} \cdots \xrightarrow{f_1} F_0 \to M$ is a free resolution of a finitely supported \mathbb{Z}^n -persistence module M, then $(\nu F_0)^* = \ker(\nu f_2)^*$.

Proof. Theorem 3.2.8 If

$$0 \longrightarrow F_n \xrightarrow{f_n} \cdots \xrightarrow{f_2} F_1 \xrightarrow{f_1} F_0 \longrightarrow M$$

is a free resolution of M, then by Corollary 3.2.11,

$$M \longrightarrow \nu F_n \xrightarrow{\nu f_n} \cdots \xrightarrow{\nu f_2} \nu F_1 \xrightarrow{\nu f_1} \nu F_0 \longrightarrow 0$$

is an injective resolution of M and the dual sequence

$$0 \longrightarrow (\nu F_0)^* \xrightarrow{(\nu f_1)^*} (\nu F_1)^* \xrightarrow{(\nu f_2)^*} \cdots \xrightarrow{(\nu f_n)^*} (\nu F_n)^* \longrightarrow M^*$$

is a free resolution of M^* . In particular, $(\nu F_0)^* = \ker((\nu f_2)^*)$.

This shows that in the situation of (3.10), we have

$$(\nu Z^{d+2}(N^{\bullet}))^* = \ker(\nu i^{d+1})^*,$$

where $(\nu i^{d+1})^*$ is a morphism of free modules. The morphism $(\nu p^{d+1})^*$ is the inclusion

$$(\nu p^{d+1})^* \colon \ker(\nu i^{d+1})^* \hookrightarrow (\nu N^{d+1})^*.$$

If i^{d+1} is represented by the graded matrix I^{d+1} , then according to Lemma 3.3.3, the morphism $(\nu i^{d+1})^*$ is represented by the graded matrix $(I^{d+1})^\top$, and $(\nu p^{d+1})^*$ is represented by some matrix $(P^{d+1})^\top$, such that P^{d+1} represents p^{d+1} . Because $(\nu p^{d+1})^*$ is the kernel inclusion of a morphism of free modules, the matrix $(P^{d+1})^\top$ can be computed by using Ker() (Algorithm 6).

Theorem 3.5.2. Let N^{\bullet} be an eventually acyclic cochain complex of free modules. Then Algorithm 11 computes free resolutions of $H^d(N^{\bullet})$.

Remark 3.5.3 (Clearing). In general, D^{d+1} is not injective. As in one-parameter persistent cohomology, the first loop in Algorithm 10 would spend some time on reducing the columns of D^{d+1} that are eventually reduced to zero if Algorithm 10 was applied to D^{d+1} . The clearing scheme in Algorithm 11, line (a) avoids this.

Proof of Theorem 3.5.2. For legibility, we omit the dimension superscripts in the following; every matrix I and P and every morphism i and p has a superscript d + 1.

Because N^{\bullet} is eventually acyclic, we have that

$$\operatorname{colim} B^d(N^{\bullet}) = B^d(\operatorname{colim} N^{\bullet}) = Z^d(\operatorname{colim} N^{\bullet}) = \operatorname{colim} Z^d(N^{\bullet})$$

for all d. The underlying matrix $u(D^{d+1})$ of D^{d+1} thus represents a generating system of $\operatorname{colim} Z^d(C_{\bullet})$.

Claim. After the clearing step in line (a), we still have that $u(D^{d+1})$ represents a generating system of colim $Z^d(C_{\bullet})$.

Proof of claim. Let \hat{I} denote the matrix $\hat{I} := \text{Bireduce}(D^d)$ obtained as I in the previous iteration. Because $D^{d+1}D^d = 0$ and $\text{Bireduce}(D^d)$ performs only column operations, we have $D^{d+1}\hat{I} = 0$. Then every column \hat{I}_j with pivot i represents a linear relation $[D^{d+1}]_i = \sum_{i' < i} \hat{I}_{i'j}[D^{d+1}]_{i'}$ of the columns of D^{d+1} . Setting the columns $[D^{d+1}]_i$ to zero for every pivot $i \in q$ of \hat{I}_j thus does not change the image of $u(D^{d+1})$.

It follows from Theorem 3.4.7 and Corollary 3.5.1 that the matrices

$$I\coloneqq \texttt{Bireduce}(D^{d+1})$$

 $P\coloneqq (\texttt{Ker}(I^{ op}))^{ op}$

represent the free resolution

$$0 \to Z^{d+1}(N^{\bullet}) \xrightarrow{i} N^{d+1} \xrightarrow{p} Z^{d+2}(N^{\bullet}) \to H^{d+2}(N^{\bullet})$$

of $H^{d+2}(N^{\bullet})$. By Proposition 2.3.23, replacing I by the minimal (in the sense of Definition 2.3.21) graded matrix I' := Minimize(I) fits into a free resolution

$$0 \to Z'^{d+1}(N^{\bullet}) \xrightarrow{I'} C'^{d+1} \xrightarrow{P'} Z^{d+2}(N^{\bullet}) \to H^{d+2}(N^{\bullet})$$

of $H^{d+2}(N^{\bullet})$ for some matrix P'. We could compute P' by

$$P' \coloneqq (\operatorname{Ker}((I')^{\top}))^{\top}$$

However, the thus computed matrix P' need not be minimal. To address this, recall that MGSWithKer() (Algorithm 8) can be used to compute matrices

$$((\tilde{I})^{\top}, (\tilde{P})^{\top}) \coloneqq \texttt{MGSWithKer}((I')^{\top}),$$

where im $\tilde{I}^{\top} = \operatorname{im}(I)^{\top}$, and \tilde{P}^{\top} is a minimal matrix (in the sense of Definition 2.3.21) represents a basis of ker \tilde{I}^{\top} . These fit into a free resolution

$$0 \to \tilde{Z}^{d+1} \xrightarrow{\tilde{I}} \tilde{C}^{d+1} \xrightarrow{\tilde{P}} \tilde{Z}^{d+2} \to H^{d+1}(N^{\bullet})^*$$

of $H^{d+1}(N^{\bullet})^*$ for some free modules \tilde{Z}^{d+1} , \tilde{C}^{d+1} and \tilde{Z}^{d+2} . The matrix \tilde{P} is minimal because \tilde{P}^{\top} is, and \tilde{I} is minimal because already I' was minimal. Hence, this is a minimal free resolution of $H^{d+1}(N^{\bullet})^*$.

Theorem 3.5.2 also shows:

Theorem C. Let K_* be an eventually acyclic one-critically filtered simplicial complex. Then $H^{d+2}(N^{\bullet}(K_*))$ can be computed from the coboundary map $\delta^{d+1} \colon N^d(K_*) \to N^{d+1}(K_*)$.

Algorithm 11: Cohomology algorithm. Computes a minimal free resolution of $H^{\bullet}(N^{\bullet})$ for a cochain complex N^{\bullet} of free Z^{2} -modules, using clearing.

(nput : Graded matrices D^{\bullet} representing N^{\bullet} . Dutput : Pairs of graded matrices representing a fr	ee resolution of $H^d(N^ullet)$ for $d=0,1,\ldots$
0	$\emptyset \to \emptyset$	▷ pivots for clearing
f	for $d = 0, 1,$ do	
(a)	for $j \in q$ do $[D^{d+1}]_j \leftarrow 0$	▷ <i>clearing</i>
(b)	$I \qquad \leftarrow \texttt{Bireduce}(D^{d+1})$	▷ see Algorithm 10
	$q \qquad \leftarrow \{ piv I_j \mid I_j \neq 0 \}$	pivots for next iteration
(c)	$I', r, c \ \leftarrow \texttt{Minimize}(I)$	▷ see Algorithm 3
	$ ilde{I}^ op, ilde{P}^ op \leftarrow extsf{MGSWithKer}((I')^ op)$	▷ see Algorithm 8
	yield \tilde{I}, \tilde{P}	\triangleright min. free res. of $H^{d+2}(N^{\bullet})$

Remark. Recall that a matrix is minimal if and only if its graded transpose is. Since we need $(I')^{\top}$ anyway, it makes sense to replace line (c) by

$$((I')^{\top}, r, c) \leftarrow \texttt{Minimize}(I^{\top}).$$

In other words, we can either first minimize I and then take the graded transpose, or first take the transpose and then minimize. Experiments have shown that the version in line (c) is slightly faster.

Remark (Base change). It is possible to change the basis of N^d before or after line (b). In this case, the algorithm computes a different but equivalent free resolution. In particular, one may choose a basis of N^d that makes later computations less expensive. We will elaborate on this idea in Section 5.1.5.

3.6 Eventual acyclicity of K_* is necessary

Recall that in one-parameter persistence (where $N^{\bullet}(K_*) = C^{\bullet}(K, K_*)$), the absolute and relative cohomology $H^{\bullet}(K_*)$ and $H^{\bullet}(K, K_*)$ determine each other uniquely even if K_* is not eventually acyclic, see Proposition 2.2.10 and Corollary 2.2.11. In two and more parameters, this is not true anymore, as we will see in this section. Specifically, we show:

Theorem B(a). There exist one-critically two-parameter filtered simplicial complexes K_* , L_* and M_* that are not eventually acyclic, such that

- (i) $H^{\bullet}(K_*) \cong H^{\bullet}(L_*)$, but $H^{\bullet}(N^{\bullet}(K_*)) \cong H^{\bullet}(N^{\bullet}(L_*))$, and
- (ii) $H^{\bullet}(K_*) \cong H^{\bullet}(M_*)$, but $H^{\bullet}(N^{\bullet}(K_*)) \ncong H^{\bullet}(N^{\bullet}(M_*))$.

Proof. Let K_* , L_* and M_* be the one-critically filtered simplicial complexes in Figure 3.4. These are not eventually acyclic. We compute $H_{\bullet}(-)$ and $H^{\bullet}(N^{\bullet}(-))$ for each of these complexes. For simplicity, we compute non-reduced (co)homology; an example using reduced homology can be obtained by adding one additional vertex.

The homology and cohomology modules are shown in Figures 3.5 to 3.10. One sees that $H_d(K_*) \cong H_d(M_*)$ (and thus also $H^d(K_*) \cong H^d(M_*)$) and $H^d(N^{\bullet}(K_*)) \cong H^d(N^{\bullet}(L_*))$ for all d. The module $H_0(K_*)$ is indecomposable because it contains an indecomposable quiver representation, while $H_0(L_*)$ is a direct sum of nonzero submodules, so $H_0(K_*) \ncong H_0(L_*)$ (and thus also $H^0(K_*) \ncong H^0(L_*)$). Similarly, $H^2(N^{\bullet}(K_*))$ is indecomposable, while $H^2(N^{\bullet}(M_*))$ is a direct sum of nonzero submodules, so $H^2(N^{\bullet}(K_*)) \ncong H^2(N^{\bullet}(M_*))$. This establishes the desired example.

This example also shows that $H^{\bullet}(K_*)$ and $H^{\bullet}(N^{\bullet}(K_*))$ need not determine each other uniquely even if colim K_* is acyclic (but colim_{1} K_* or colim_{{2}} K_* is not).



Figure 3.4: The filtered simplicial complexes used in Theorem B(a). The filtration goes left to right and bottom to top.

Figure 3.5: Homology of the complex K_* from Figure 3.4a. The module $H_0(K_*)$ is generated by x, y. These are subject to the relations $\partial e = \partial f = x + y$. Hence, the generators x and y are identified when e or y are present in the filtration. Analogously for $H_1(K_*)$. The module $H_0(K_*)$ is indecomposable, because it contains an indecomposable quiver representation.



Figure 3.6: Cohomology of $N^{\bullet}(K_*)$ for the complex K_* from Figure 3.4a. The coordinate axes are drawn pointing downward. Therefore, the generator labeled, e.g., "*i*" really means the basis element of $N^{\bullet}(K_*)$ of grade -g(i) corresponding to the cell *i*. The symbols \circ denote the negated grades of the respective cells of K_* . That is, the label "x + y" is placed at $-g(x + y) = -g(x) \vee -g(y)$.

Figure 3.7: Homology of the complex L_* from Figure 3.4b. The generators x, y of $H_0(K_*)$ are identified by $\partial e = \partial f = x + f$ at $g(x) \lor g(y)$, and z + x generates a direct summand "killed" by ∂g and ∂h .

 $(a) H^{0}(N^{\bullet}(K_{*})) \qquad (b) H^{1}(N^{\bullet}(K_{*})) \qquad (c) H^{2}(N^{\bullet}(K_{*}))$ $(a) H^{0}(L_{*}) \qquad (b) H^{1}(N^{\bullet}(K_{*})) \qquad (c) H^{2}(N^{\bullet}(K_{*}))$ $(a) H_{0}(L_{*}) \qquad (b) H_{1}(L_{*})$ $(b) H_{1}(L_{*}) \qquad (b) H_{1}(L_{*})$ $(a) H^{0}(N^{\bullet}(L_{*})) \qquad (b) H^{1}(N^{\bullet}(L_{*})) \qquad (c) H^{2}(N^{\bullet}(L_{*}))$

 ${\mathop{\rm e}\limits_{\scriptscriptstyle 0}}$

Figure 3.8: Cohomology of $N^{\bullet}(L_*)$ for the complex L_* from Figure 3.4b. Note that $e + f + g \equiv e + h + f \pmod{\delta z}$.



3.7 Making K_{*} eventually acyclic

For a given filtered complex K_* and $z \in \mathbb{Z}^n$, we consider two ways to construct an eventually acyclic chain complex \tilde{C}_{\bullet} such that $(\tilde{C}_{\bullet})_w = C_{\bullet}(K_w)$ for all $w \leq z$. Inspired by the first way presented in Section 3.7.1, we refer to this as *coning off* K_* , although we do not build an actual cone in all cases. The motivating situation for this is that we choose $z > g(\sigma)$ for all $\sigma \in K_*$, so that everying "interesting" in K_* happens below z.

3.7.1 Simplicially coning off K_*

Assume that $z = (z_1, \ldots, z_n) > g(\sigma)$ for all $\sigma \in K_*$, and let $1 \le k \le n$. For $w = (w_1, \ldots, w_n) \in \mathbb{Z}^n$ let $w|_z^k \in \mathbb{Z}^n$ be obtained by replacing the *k*th coordinate of *w* by z_k . We define the one-critically filtered simplicial complex \tilde{K}_* with

$$K_*^{(d)} = K_*^{(0)} \cup \begin{cases} \{a\} & \text{if } d = 0, \\ \{\tilde{\sigma} \mid \sigma \in K_*^{(d-1)}\} & \text{if } d > 0, \end{cases}$$

for $\tilde{\sigma} \coloneqq \sigma \cup \{a\}$, equipped with the grading $g(a) \coloneqq \bigwedge_{\sigma \in K_*} g(\sigma)|_z^k$ and $g(\tilde{\sigma}) \coloneqq g(\sigma)|_z^k$. If $w_k \ge z_k$, then \tilde{K}_w is a cone with apex a and thus contractible. If $w_k < z_k$, then $\tilde{K}_w = K_w$. The same construction can be applied on the level of chains. For $w \in \mathbb{Z}^n$ let $F(z)|_z^k \coloneqq F(w|_z^k)$. Then $\tilde{C}_{\bullet} \coloneqq C_{\bullet}(\tilde{K}_*)$ is the chain complex with

$$\widetilde{C}_d \coloneqq C_d(K_*) \oplus C_{d-1}(K_*)|_z^k, \qquad \partial_d \coloneqq \begin{pmatrix} \partial_d & (-1)^d \\ \partial_{d-1} \end{pmatrix}.$$

Repeatedly applying this construction to \tilde{C}_{\bullet} for all $1 \leq k \leq n$ gives an eventually acyclic chain complex \tilde{C}_{\bullet} such that $(\tilde{C}_{\bullet})_w = (C_{\bullet})_w$ if w < z.

3.7.2 Coning off K_* using homology resolutions

Alternatively, if C_{\bullet} is a chain complex of free modules, one can obtain a new complex \hat{C}_{\bullet} by adjoining only as many new basis elements to C_{\bullet} as necessary to turn \hat{C}_{\bullet} eventually acyclic. Let $z \in \mathbb{Z}^n$ with z > w for all $w \in \operatorname{rk} C_d$ for all d. For $k \leq n$, recall the exact functors $\operatorname{colim}_{\{k\}}: \operatorname{Vec}^{\mathbb{Z}^n} \to \operatorname{Vec}^{\mathbb{Z}^{n-1}}$ from Definition 3.2.6. We also define the exact functor

$$E_k \colon \operatorname{Vec}^{\mathbf{Z}^{n-1}} \longrightarrow \operatorname{Vec}^{\mathbf{Z}^n},$$
$$M \longmapsto (w_1, \dots, w_n) \mapsto \left\{ \begin{smallmatrix} M_{(w_1, \dots, w_{k-1}, w_{k+1}, \dots, w_n)} & \text{if } w_k \ge z_k \\ 0 & \text{otherwise.} \end{smallmatrix} \right.$$

A module $F \in \text{Vec}^{\mathbb{Z}^{n-1}}$ is free if and only if $E_k F$ is. In this case,

$$E_kF((w_1,\ldots,w_{k-1},w_{k+1},\ldots,w_n)) = F((w_1,\ldots,w_{k-1},z_k,w_{k+1},\ldots,w_n)).$$

Two parameters We first describe the construction for the case n = 2. Let $z \in \mathbb{Z}^2$ such that z > w for all $z \in \operatorname{rk} C_d$ for all d as above, and choose $k \in \{1, 2\}$. W.l.o.g., we assume that k = 1. Consider the chain complex $C'_{\bullet} := \operatorname{colim}_{\{k\}} C_{\bullet}$ of free one-parameter persistence modules. For each d, choose a free resolution $F_{d,\bullet}$ of $H_d(C'_{\bullet}) \cong \operatorname{colim}_{\{k\}} H_d(C_{\bullet})$ of length 1, and let $f_{d,0} \colon F_{d,0} \to H_d(C'_{\bullet})$ be the augmentation morphism. Consider the diagram

of one-parameter persistence modules. Each of the two rows is exact and thus forms (in the case of the lower row, the beginning of a) free resolution of $H_d(C_{\bullet})$. We construct the morphisms $\phi_{d,0}$ and $\phi_{d,1}$ that make the diagram commute. We will use these to construct a chain complex \tilde{C}_{\bullet} with $(\tilde{C}_{\bullet})_w = (C_{\bullet})_w$ if $w_k < z_k$ and $(\tilde{C}_{\bullet})_w = (\tilde{C}'_{\bullet})_w$ otherwise.

Since q_d is surjective and $F_{d,0}$ free, there exists a morphism $\phi_{d,0}$ such that $q_d\phi_{d,0} = f_{d,0}$. This implies $q_d\phi_{d,0}f_{d,1} = 0$. Because the lower row is exact, ∂_{d+1} surjects onto $B_d(C'_{\bullet}) = \ker q_d$. Since $F_{d,1}$ is free, there exists a morphism $\phi_{d,1}$ such that $\partial_{d+1}\phi_{d,1} = \phi_{d,0}f_{d,1}$. It follows from commutativity of (3.11) than

$$\tilde{C}'_{d} \coloneqq C'_{d} \oplus F_{d-1,0} \oplus F_{d-2,1}, \qquad \tilde{\partial}_{d} \coloneqq \begin{pmatrix} \partial_{d} \phi_{d-1,0} & -\phi_{d-2,1} \\ 0 & 0 & f_{d-2,1} \\ 0 & 0 & 0 \end{pmatrix} \colon \tilde{C}'_{d} \longrightarrow \tilde{C}'_{d-1}.$$
(3.12)

defines a chain complex.

Lemma 3.7.1. For any number n of parameters, the morphism

$$\alpha_{d,2} \coloneqq (\partial_{d+1}, \phi_{d,0}) \colon C'_{d+1} \oplus F_{d,0} \to Z_d(C'_{\bullet}).$$

$$(3.13)$$

is surjective.

Proof. To see this, note that for $c \in Z_d(C_{\bullet})$, surjectivity of $f_{d,0}$ allows us to choose a $b \in F_{d,0}$ such that $f_{d,0}(b) = q_d(c)$. Then $q_d(c - \phi_{d,0}(b)) = 0$, so by exactness of the lower row, we obtain that $c = \partial_{d+1}(a) + \phi_{d,0}(b) = \alpha_{d,2}(\frac{a}{b})$ for some $a \in C'_{d+1}$.

Lemma 3.7.2. The complex \tilde{C}'_{\bullet} from (3.12) is acyclic.

Proof. Let $(a, b, c) \in \ker \tilde{\partial}_d$. Then c = 0, because $f_{d-2,1}$ is injective. Then $\partial_d(a) + \phi_{d-1,0}(b) = 0$, which gives

$$f_{d-1,0}(b) = q_{d-1}\partial_d(a) = 0$$

By exactness of the upper row, we may choose $b' \in F_{d-1,1}$ such that $b = f_{d-1,1}(b')$. This implies $\partial_d(a + \phi_{d-1,1}(b')) = 0$. By surjectivity of $\alpha_{d,2}$, we may choose $(a', b'') \in C_{d+1} \oplus F_{d,0}$ such that

$$a + \phi_{d-1,1}(b') = \partial_{d+1}(a') + \phi_{d,0}(b'')$$

This shows that $(a, b, c) = \tilde{\partial}_{d+1}(a', b'', b')$, so \tilde{C}'_{\bullet} is acyclic.

Let \tilde{C}_{\bullet} be the chain complex of free modules with

$$\begin{split} \tilde{C}_{d} &\coloneqq C_{d} \oplus E_{k}F_{d-1,0} \oplus E_{k}F_{d-2,1}, \\ \tilde{\partial}_{d} &\coloneqq \begin{pmatrix} \partial_{d} & E_{k}\phi_{d-1,0} & -E_{k}\phi_{d-2,1} \\ 0 & 0 & E_{k}f_{d-2,1} \\ 0 & 0 & 0 \end{pmatrix} : \tilde{C}_{d} \longrightarrow \tilde{C}_{d-1} \end{split}$$

Then $(\tilde{C}_{\bullet})_w = (C_{\bullet})_w$ if $w_k < z_k$ and $(\tilde{C}_{\bullet})_w = (\tilde{C}'_{\bullet})_w$ otherwise. In particular, $\operatorname{colim}_k \tilde{C}_{\bullet} = C'_{\bullet}$ is acyclic. If $F_{d,\bullet}$ is a minimal free resolution for each d, then \tilde{C}_{\bullet} is the smallest chain complex with these properties.

Recall that C'_{\bullet} is a chain complex of finite rank free one-parameter persistence modules, so $B_d(C'_{\bullet})$ and $Z_d(C'_{\bullet})$ are free and $F_{d,\bullet} = (0 \to B_d(C'_{\bullet}) \to Z_d(C'_{\bullet}))$ is a free resolution of $H_d(C'_{\bullet})$. In this case, \tilde{C}_{\bullet} is isomorphic to the cone construction described in Section 3.7.1. Note that since $H_d(C'_{\bullet})$ is a one-parameter persistence module, one can choose a basis of F'_{\bullet} that exhibits the barcode of $H_d(C'_{\bullet})$ as described in Section 2.2, which can be computed efficiently. We give an algorithmic treatment of this approach in Section 5.1.4.

Applying the above construction to \tilde{C}_{\bullet} again, now for k = 2, gives an eventually acyclic complex \tilde{C}_{\bullet} $(\tilde{C}_{\bullet})_w = C_{\bullet}(K_w)$ for all $w \leq z$.

More than two parameters The above construction can be generalized inductively to more than two parameters. Let C_{\bullet} be a chain complex of free *n*-parameter persistence modules. As before, let $C'_{\bullet} \coloneqq \operatorname{colim}_{\{k\}} C'_{\bullet}$

$$0 \to F_{d,n-1} \xrightarrow{f_{d,n-1}} F_{d,n-2} \to \cdots \to F_{d,0} \xrightarrow{f_{d,0}} H_d(C_{\bullet}) \to 0.$$

be a free resolution of $H_{\bullet}(C'_{\bullet})$ of length n-1. For each d, we construct two sequences of morphisms $\phi_{d,0}, \phi_{d,1}, \ldots, \phi_{d,n-1}$ and $\alpha_{d+2,2}, \ldots, \alpha_{d+n-1,n-1}$ that make the diagram

commute and render the lower row exact, where

$$A_{d,0} = Z_d(C'_{\bullet})$$

$$A_{d,1} = C'_{d+1} \qquad \qquad \alpha_{d,1} = \partial_{d+1},$$

$$A_{d,2} = C'_{d+2} \oplus F_{d+1,0} \qquad \qquad \alpha_{d,2} = (\partial_{d+2}, \phi_{d+1,0}),$$

$$A_{d,i} = A_{d+1,i-1} \oplus F_{d+1,i-2} \qquad \qquad \alpha_{d,i} = \begin{pmatrix} \alpha_{d+1,i-1} & (-1)^i \phi_{d+1,i-2} \\ 0 & f_{d+1,i-2} \end{pmatrix}$$
(3.15)

for i > 1. We will then construct a chain complex \tilde{C}_{\bullet} such that $(\tilde{C}_{\bullet})_w = (C_{\bullet})_w$ if $w_k < z_k$ and $(\tilde{C}_{\bullet})_w$ is acyclic otherwise.

Lemma 3.7.3. There exists morphisms $\phi_{d,i}$, such that the diagram (3.14) commutes, and such that the morphisms $\alpha_{d,i}$ make $A_{d,\bullet}$ a free resolution of $H_d(C'_{\bullet})$. that is exact at $A_{d,i}$ for all i > 0.

Proof. We construct the morphisms $\phi_{d,i}$ inductively. As induction base, we note that q_d surjects onto $H_d(C')$ and $F_{d,0}$ is free, so there exists a morphism $\phi_{d,0}$ such that the square (1) commutes. This establishes the construction of $\phi_{d,0}$. We next construct the morphism $\phi_{d,1}$. Commutativity

of (1) gives $q_d \phi_{d,0} f_{d,1} = 0$, so $\phi_{d,0} f_{d,1}$ factors through a unique morphism $F_{d,1} \to \ker q_d$. The lower row is exact at $A_{d,0} = Z_d(C'_{\bullet})$. Therefore, the morphism $\alpha_{d,1}$ surjects onto im $\alpha_{d,1} = \ker q_d$. The universal property of the free (and thus projective) module $F_{d,1}$ gives a morphism $\phi_{d,1}$ such that (1) commutes. We next show that the $A_{d,\bullet}$ form an exact chain complex at $A_{d,1}$. We have

$$\alpha_{d,1}\alpha_{d,2} = \partial_{d+1}(\partial_{d+2}, \phi_{d+1,0}) = (0, \partial_{d+1}\phi_{d+1,0}) = 0,$$

because $\operatorname{im} \phi_{d+1,0} \subseteq A_{d+1,0} = Z_{d+1}(C'_{\bullet})$. This shows that $A_{d,\bullet}$ forms a chain complex at $A_{d,1}$. Exactness at $A_{d,1}$ was already shown in Lemma 3.7.1. With the same argument as for $\phi_{d,1}$, we get the morphism $\phi_{d,2}$ that makes the square (2) commute.

As induction hypothesis, we assume for $i \ge 2$ that

$$\alpha_{d+1,i-1}\phi_{d+1,i-1} = \phi_{d+1,i-2}f_{d+1,i-1} \tag{3.16}$$

$$\alpha_{d,i}\phi_{d,i} = \phi_{d,i-1}f_{d,i} \tag{3.17}$$

and that the chain complex $(A_{d+1,\bullet}, \alpha_{d+1,\bullet})$ is exact at $A_{d+1,i-1}$. For the induction step, we first show that $A_{d,\bullet}$ forms an exact complex at $A_{d,i}$. We have

$$\begin{aligned} \alpha_{d,i}\alpha_{d,i+1} &= \begin{pmatrix} \alpha_{d+1,i-1} & (-1)^i \phi_{d+1,i-1} \\ 0 & f_{d+1,i-1} \end{pmatrix} \begin{pmatrix} \alpha_{d+1,i} & (-1)^{i+1} \phi_{d+1,i} \\ 0 & f_{d+1,i} \end{pmatrix} \\ &= \begin{pmatrix} \alpha_{d+1,i-1}\alpha_{d+1,i} & \pm \alpha_{d+1,i-1}\phi_{d+1,i} \mp \phi_{d+1,i-1}f_{d+1,i} \\ f_{d+1,i-1}f_{d+1,i} \end{pmatrix} \stackrel{(3.16)}{=} 0. \end{aligned}$$

To show acyclicity, let

$$\begin{pmatrix} a \\ b \end{pmatrix} \in \ker \alpha_{d,i} = \ker \begin{pmatrix} \alpha_{d+1,i-1} & (-1)^i \phi_{d+1,i-2} \\ f_{d+1,i-2} \end{pmatrix} \subseteq A_{d+1,i-1} \oplus F_{d+1,i-2} = A_{d,i}.$$
(3.18)

Then $f_{d+1,i-2}(b) = 0$. By exactness of $F_{d+1,\bullet}$, there exists a $c \in F_{d+1,i-1}$ such that $b = f_{d+1,i-1}(c)$. Then

$$\alpha_{d+1,i-1}(a+(-1)^{i}\phi_{d+1,i-1}(c)) \stackrel{(3.16)}{=} \alpha_{d+1,i-1}(a) + (-1)^{i}\phi_{d+1,i-2}f_{d+1,i-1}(c)$$
$$\stackrel{(3.18)}{=} \alpha_{d+1,i-1}(a) + (-1)^{i}\phi_{d+1,i-2}(b) = 0.$$

Because $A_{d+1,\bullet}$ is exact at $A_{d+1,i-1}$ by the induction hypothesis, we get an $m \in A_{d+1,i}$ such that $\alpha_{d+1,i}(m) = a + (-1)^i \phi_{d+1,i-1}(c)$. Now one verifies that

$$\begin{aligned} \alpha_{d,i+1} \begin{pmatrix} m \\ c \end{pmatrix} \stackrel{(3.15)}{=} \begin{pmatrix} \alpha_{d+1,i} & (-1)^{i+1}\phi_{d+1,i-1} \\ f_{d+1,i-1} \end{pmatrix} \begin{pmatrix} m \\ c \end{pmatrix} = \begin{pmatrix} \alpha_{d+1,i}(m) + (-1)^{i+1}\phi_{d+1,i-1}(c) \\ f_{d+1,i-1}(c) \end{pmatrix} \\ &= \begin{pmatrix} a + (-1)^i\phi_{d+1,i-1}(c) + (-1)^{i+1}\phi_{d+1,i-1}(c) \\ f_{d+1,i-1}(c) \end{pmatrix} = \begin{pmatrix} a \\ b \end{pmatrix}, \end{aligned}$$

using the definition of m and c in the last two steps. This shows exactness at $A_{d,i}$.

To construct $\phi_{d,i+1}$, we note that

$$\alpha_{d,i}\phi_{d,i}f_{d,i+1} \stackrel{(3.17)}{=} \phi_{d,i-1}f_{d,i}f_{d,i+1} = 0$$

because $F_{d,\bullet}$ is a chain complex. By the universal property of ker $\alpha_{d,i}$, the morphism $\phi_{d,i}f_{d,i+1}$ factors through a morphism $F_{d,i+1} \to \ker \alpha_{d,i}$. By exactness at $A_{d,i+1}$, the morphism $\alpha_{d,i+1}$ surjects onto this kernel. The universal property of the free (and, in particular, projective) module $F_{d,i+1}$ gives a morphism $\phi_{d,i+1}$ such that

$$\alpha_{d,i+1}\phi_{d,i+1} = \phi_{d,i}f_{d,i+1}.$$

Recall that n is the number of parameters, so C'_{\bullet} is a chain complex of n-1-parameter modules, and $F_{d,\bullet}$ has length n-1. Let \tilde{C}'_{\bullet} be the chain complex defined by

$$\tilde{C}'_d = A_{d-n-1,n+1} = C'_d \oplus F_{d-1,0} \oplus \dots \oplus F_{d-n,n-1}$$

with the boundary morphism

$$\tilde{\partial}'_{d+n+1} \colon \tilde{C}'_{d+n+1} = A_{d,n+1} \xrightarrow{\alpha_{d,n+1}} A_{d,n} \xrightarrow{\begin{pmatrix} 1\\0 \end{pmatrix}} A_{d,n} \oplus F_{d,n-1} = A_{d-1,n+1} = \tilde{C}'_{d+n}.$$

Lemma 3.7.4. The objects \tilde{C}'_{\bullet} and morphisms $\tilde{\partial}'_{\bullet}$ define an acyclic chain complex.

Proof. This is a chain complex indeed, because

$$\tilde{\partial}_{d+n+1}'\tilde{\partial}_{d+n+2}' = \begin{pmatrix} \alpha_{d,n+1} \\ 0 \end{pmatrix} \begin{pmatrix} \alpha_{d+1,n+1} \\ 0 \end{pmatrix} \stackrel{(3.15)}{=} \begin{pmatrix} \alpha_{d+1,n} & \pm \phi_{d+1,n-1} \\ f_{d+1,n-1} \end{pmatrix} \begin{pmatrix} \alpha_{d+1,n+1} \\ 0 \end{pmatrix} = 0$$

because $A_{d+1,\bullet}$ is a chain complex. To show that it is acyclic, let

$$a \in \ker \tilde{\partial}'_{d+n+1} = \ker \begin{pmatrix} \alpha_{d,n+1} \\ 0 \end{pmatrix} \subseteq A_{d,n+1} = \tilde{C}'_{d+n+1}.$$

Because $A_{d,\bullet}$ is acyclic, we get a $b \in A_{d,n+2}$ such that $a = \alpha_{d,n+2}(b)$. There is a commutative diagram
(id 0)

for all $i \ge 2$. For i = n + 1, we have $F_{d+1,n} = 0$ because $F_{d+1,\bullet}$ has length n - 1. Therefore, we get $A_{d,n+2} = A_{d+1,n+1}$ and $\alpha_{d,n+2} = \binom{\alpha_{d+1,n+1}}{0} = \tilde{\partial}'_{d+n+2}$. In particular, $a = \alpha_{d,n+2}(b) = \tilde{\partial}'_{d+n+2}(b)$. This shows acyclicity.

Now,

defines a chain complex, where we use that $E_k C'_{\bullet} \subseteq C_{\bullet}$. The chain complex \tilde{C}_{\bullet} is a complex of free modules such that $(\tilde{C}_{\bullet})_w = (C_{\bullet})_w$ if $w_k < z_k$ and such that $(\tilde{C}_{\bullet})_w$ is acyclic if $w_k \ge z_k$. Repeating this for all $1 \le k \le n$ gives an eventually acyclic chain complex \hat{C}_{\bullet} such that $(\hat{C}_{\bullet})_w = (C_{\bullet})_w$ if w < z. This completes the construction.

3.8 Fringe presentations

Let M be a finitely supported persistence module. The close connection between free and injective resolutions of M from Theorem 3.2.8 can be used to compute a fringe presentation of M, in the sense of [104].

Definition 3.8.1. A fringe presentation of a persistence module M is a morphism $f: F \to I$ such that F is free, I is injective, and $M \cong \text{im } f$.

Recall the Koszul complex functor

$$\Omega_{\bullet}: \quad 0 \to \overbrace{\mathrm{id}}^{\Omega_{n}} \underbrace{\overset{\alpha_{n-1}}{\bigoplus}}_{S \in \binom{[n]}{1}} \operatorname{Colim}_{S} \to \cdots \underbrace{\bigcirc}_{S \in \binom{[n]}{n-1}} \operatorname{Colim}_{S} \underbrace{\overset{\kappa_{1}}{\bigoplus}}_{\Omega_{0}} \to 0$$

from Definition 3.2.10. Recall that $\Omega_{\bullet}F(z)$ is a flat resolution of the injective module $I(z)\langle\epsilon\rangle \cong \nu F(z)\langle\epsilon\rangle$; that is, there is a quasi-isomorphism $\Omega_{\bullet}F(z) \to I(z)\langle\epsilon\rangle$. Since $\Omega_n F(z) = F(z)$, there is also a morphism $\Omega_{\bullet}F(z)[n] \to F(z)$ of chain complexes. Let $(F_{\bullet}, \partial_{\bullet}^F)$ be a chain complex of free modules, and let $I_{\bullet} := \nu F_{\bullet}\langle\epsilon\rangle$. Then $\Omega_{\bullet}F_{\bullet}$ is a double complex, whose differentials we denote by

 $\kappa_{ij} \coloneqq \kappa_i F_j \colon \Omega_i F_j \to \Omega_{i-1} F_j, \qquad \partial_{ij} \coloneqq \Omega_i \partial_j^F \colon \Omega_i F_j \to \Omega_i F_{j-1}.$

Let $\tilde{\Omega}_{\bullet}$ be the total complex of $\Omega_{\bullet}F_{\bullet}$. Explicitly, $\tilde{\Omega}_{\bullet}$ is the chain complex with components and boundary morphisms

$$\tilde{\Omega}_q = \bigoplus_{i+j=q} \Omega_i F_j, \quad \partial_q^{\tilde{\Omega}} = \sum_{i+j=q} ((-1)^q \kappa_{ij} + \partial_{ij}).$$

Seeing F(z) and $I(z)\langle\epsilon\rangle$ as chain complexes concentrated in degree zero, we obtain morphisms

that give rise to the morphisms

$$I_{\bullet}[n] \stackrel{f[n]}{\longleftrightarrow} \tilde{\Omega} \stackrel{g}{\longrightarrow} F_{\bullet}$$

of chain complexes, where f is a chain homotopy equivalence. If F_{\bullet} is eventually acyclic, then according to Theorem 3.2.8, the morphism g is a quasi-isomorphism. We show that g is a chain homotopy equivalence by constructing a homotopy inverse h of g, so that f[n]h will be the desired fringe presentation.

The idea behind the following construction is to use the fact that by the assumption that F_{\bullet} is eventually acyclic, the chain complexes $\operatorname{colim}_{S} F_{\bullet}$ of free (n - |S|)-parameter modules are acyclic and hence contractible for every S with |S| > 0. We may choose chain contractions of $\operatorname{colim}_{S} F_{\bullet}$ for every S, which give rise to chain contractions of $\Omega_{i}F_{\bullet}$ for every $0 \leq i < n$, and use these to construct the desired homotopy inverse h of g.

Let $0 \leq i < n$ and $S \in {[n] \choose i-k}$. Because F_{\bullet} is eventually acyclic, $\operatorname{colim}_{S} F_{\bullet}$ is an acyclic and hence contractible chain complex. Let $(s_{j}^{S})_{j \in \mathbb{Z}}$ be a chain contraction of $\operatorname{colim}_{S} F_{\bullet}$. Then the maps $s_{ij} \coloneqq \bigoplus_{S \in {[n] \choose n-i}} \Delta_{S} s_{j}^{S}$ form a chain contraction $(s_{ij})_{j \in \mathbb{Z}}$ of $\Omega_{i} F_{\bullet}$, where $s_{ij} \colon \Omega_{i} F_{j} \to \Omega_{i} F_{j+1}$. Explicitly, this means that

$$\partial_{i,j+1}s_{ij} + s_{i,j-1}\partial_{ij} = \mathrm{id}_{\Omega_i F_j} \tag{3.20}$$

for all $j \in \mathbf{Z}$ and $0 \le i < n$. For $k \ge 0$, define the morphism

$$t_{ijk} = \begin{cases} \operatorname{id}_{\Omega_i F_j} & \text{if } k = 0, \\ s_{i-k,j+k-1} \kappa_{i-k+1,j+k-1} t_{i,j,k-1} & \text{if } k > 0 \end{cases} \colon \Omega_i F_j \longrightarrow \Omega_{i-k} F_{j+k}. \tag{3.21}$$

Proposition 3.8.2. Let F_{\bullet} be an eventually acyclic chain complex of free modules, and let Ω_{\bullet} be as above.

(i) There is a morphism $h: F_{\bullet} \to \tilde{\Omega}_{\bullet}[n]$ of chain complexes. Its component

$$h_q \colon F_q \longrightarrow \underbrace{\Omega_n F_q \oplus \dots \oplus \Omega_0 F_{q+n}}_{\tilde{\Omega}_{q+n}}$$

is induced by the morphisms $(-1)^{j(q+1)}t_{nqj}$: $F_q \to \Omega_{n-j}F_{q_j}$ for $j = 0, \ldots, n$.

(ii) The morphism h is a chain homotopy inverse of the morphism $g \colon \tilde{\Omega}_{\bullet}[n] \to F_{\bullet}$.

Proof. We first show two statements about the morphisms t_{ijk} . For simplicity of notation, we only write the indices i and j for the rightmost morphism; for the others, these indices then are determined.

Claim. For all i < n, j and k, we have

$$\kappa \partial t_{ijk} = (-1)^k \kappa t_k \partial_{ij}. \tag{3.22}$$

Proof of claim. By induction over k (with the induction step at (*)), we get

$$\kappa \partial t_{ijk} \stackrel{(3.21)}{=} \kappa \partial s \kappa t_{i,j,k-1} \stackrel{(3.20)}{=} \kappa (1-s\partial) \kappa t_{i,j,k-1} \stackrel{(\dagger)}{=} -\kappa s \partial \kappa t_{i,j,k-1} \stackrel{(\dagger)}{=} -\kappa s \kappa \partial t_{i,j,k-1} \stackrel{(\ast)}{=} (-1)^k \kappa s \kappa t_{k-1} \partial_{ij} \stackrel{(3.21)}{=} (-1)^k \kappa t_k \partial_{ij},$$

using that $\partial_{\bullet\bullet}$ and $\kappa_{\bullet\bullet}$ form a double complex in (†).

The claim implies that

$$\partial t_{ijk} - \kappa t_{i,j,k-1} \stackrel{(3.21)}{=} (\partial s - 1) \kappa t_{i,j,k-1}$$

$$\stackrel{(3.20)}{=} -s \partial \kappa t_{i,j,k-1}$$

$$\stackrel{(3.22)}{=} -(-1)^{k-1} s \kappa t_{k-1} \partial_{ij}$$

$$\stackrel{(3.21)}{=} (-1)^k t_k \partial_{ij}.$$
(3.23)

for all i < n, j and k. This allows us to drag symbols ∂ past the symbols t_{ijk} . Let \pm stand for $(-1)^q$ and \mp for $(-1)^{q+1}$. We obtain that

$$\begin{split} \partial_{q}^{\tilde{\Omega}}h_{q} = & \begin{pmatrix} \partial_{nq} & & \\ \pm \kappa_{nq} & \partial_{n-1,q+1} & & \\ \pm \kappa_{n-1,q+1} & \partial_{n-2,q+2} & & \\ & \ddots & \ddots & \ddots & \\ & & \pm \kappa_{1,q+n-1} & \partial_{0,q+n} \end{pmatrix} \begin{pmatrix} \mathrm{id}_{\Omega_{n}F_{q}} \\ \mp t_{nq1} \\ t_{nq2} \\ \vdots \\ (\mp)^{n-1}t_{n,q,n-1} \\ (\mp)^{n}t_{nqn} \end{pmatrix} \\ & = \begin{pmatrix} \partial_{nq} \\ \mp (\partial t_{nq1} - \kappa_{nq}) \\ - (\partial t_{nq2} - \kappa t_{nq1}) \\ \vdots \\ (\mp)^{n}(\partial t_{nqn} - \kappa t_{n,q,n-1}) \end{pmatrix} \begin{pmatrix} \mathrm{id}_{\Omega_{n}F_{q}} \\ \mp t_{nq2} \\ \vdots \\ (\mp)^{n}t_{n}\partial_{nq} \end{pmatrix} = h_{q-1}\partial_{q}^{F}. \end{split}$$
Therefore, h is a chain map. It remains to show that h is a homotopy inverse of g. We clearly have $gh = id_{F_{\bullet}}$. It remains to show that $id_{\tilde{\Omega}_{\bullet}}$ and hg are chain homotopic. For every j, let

$$\sigma_{j} \coloneqq \begin{pmatrix} 0 & & & \\ 0 & s_{n-1,q+1} & & \\ 0 & \mp t_{1}s_{n-1,q+1} & s_{n-2,q+2} & & \\ 0 & t_{2}s_{n-1,q+1} & \mp t_{1}s_{n-2,q+2} & s_{n-3,q+3} & \\ 0 & \mp t_{3}s_{n-1,q+1} & t_{2}s_{n-2,q+2} & \mp t_{1}s_{n-3,q+3} & s_{n-4,q+4} \\ \vdots & & & \ddots \end{pmatrix} : \tilde{\Omega}_{q+n} \to \tilde{\Omega}_{q+n+1}$$

We remark that we have the following two recursive relationship of the symbols t_{ijk} , which both follow directly from the definition:

$$t_{i,j,k+1} \stackrel{(3.21)}{=} s_{i-k-1,j+k} \kappa_{i-k,j+k} t_{ijk} = t_{i-1,j+1,k} s_{i-1,j} \kappa_{i,j}.$$
(3.24)

From this, it follows that for all $k \ge 0$, we have

$$(\kappa t_k s_{ij} - \partial t_k s_{ij}) + (-1)^{k+1} (t_{k+1} s \partial_{ij} - t_k s \kappa_{ij})$$

$$\stackrel{(3.24)}{=} (1 - \partial s) \kappa t_k s_{ij} + (-1)^k t_k s \kappa (s \partial_{ij} - 1)$$

$$\stackrel{(3.20)}{=} s \partial \kappa t_k s_{ij} - (-1)^k t_k s \kappa \partial s_{ij}$$

$$\stackrel{(3.24)}{=} s \partial \kappa t_k s_{ij} - (-1)^k s \kappa t_k \partial s_{ij}$$

$$\stackrel{(3.22)}{=} s \partial \kappa t_k s_{ij} - s \partial \kappa t_k s_{ij}$$

$$= 0. \qquad (3.25)$$

We will use this to show that the σ_j form a chain homotopy between id $\overline{\Omega}_{\bullet}$ and hg. In favor of legibility, we leave out the indices i and j of all morphisms s_{ij} , κ_{ij} , ∂_{ij} and t_{ijk} in the following. We obtain

$$\begin{split} \partial_{j+1}^{\Omega} \sigma_{j} + \sigma_{j-1} \partial_{j}^{\Omega} \\ &= \begin{pmatrix} \stackrel{\partial}{\mp \kappa} & \stackrel{\partial}{\partial} \\ & \stackrel{\pi}{\mp \kappa} & \stackrel{\partial}{\partial} \\ & \stackrel{\pi}{\times} & \stackrel{\pi}{\times} & \stackrel{\pi}{\otimes} \\ & \stackrel{\pi}{\times} & \stackrel{\pi}{\times} \\ & \stackrel{\pi}{\times} \\ & \stackrel{\pi}{\times} & \stackrel{\pi}{\times} \\ & \stackrel{\pi}{$$

This shows that hg and $\mathrm{id}_{\tilde{\Omega}_{\bullet}}$ are chain homotopic.

Corollary 3.8.3. Let $M \in \text{Vec}^{\mathbb{Z}^n}$ be finitely supported, F_{\bullet} be a free resolution of M of length n, let t_{ijk} be as above, and let $a: \Omega_0 F_n \to \nu F_n \langle \epsilon \rangle$ be the augmentation map of the free resolution $\Omega_{\bullet}F_n$ of $\nu F_n \langle \epsilon \rangle$. Then the composite morphism

$$F_0 = \Omega_n F_0 \xrightarrow{t_{n,0,n}} \Omega_0 F_n \xrightarrow{a} \nu F_n \langle \epsilon \rangle$$

is a fringe presentation of M.

Proof. Because F_{\bullet} is a free resolution of M, there is a quasi-isomorphism $F_{\bullet} \to M$. According to Proposition 3.8.2, we obtain the two homotopy equivalences



of chain complexes, where f and h are as above, $I_n = \nu F_n \langle \epsilon \rangle$, and $\tilde{\Omega}_n = \Omega_n F_0 \oplus \cdots \oplus \Omega_0 F_n$. The image of the homotopy equivalence f[n]h is quasi-isomorphic to M. The only non-zero component of f[n]h is

$$f_n h_0 = (0, \dots, 0, a) \begin{pmatrix} {}^{\mathrm{id}_{\Omega_n F_0}} \\ {}^{-t_{n01}} \\ \vdots \\ {}^{(-1)^n t_{n0n}} \end{pmatrix} = (-1)^n a t_{n0n} \colon F_0 \to \Omega_0 F_n \to I_n$$

This shows that $\operatorname{im} at_{n0n} = \operatorname{im} f_n h_0 = M$.

Two-parameter modules Recall that $\Omega_i F_{\bullet} = \bigoplus_{S \in \binom{[n]}{n-k}} \Delta_S \operatorname{colim}_S F_{\bullet}$, where $\operatorname{colim}_S F_{\bullet}$ is a chain complex of (n - |S|)-parameter modules. If n = 2 and F_{\bullet} is a free resolution of a finitely supported module M, then

$$\Omega_1 F_{\bullet} = \Delta_{\{1\}} \operatorname{colim}_{\{1\}} F_{\bullet} \oplus \Delta_{\{2\}} \operatorname{colim}_{\{2\}} F_{\bullet},$$

where $\operatorname{colim}_{\{l\}} F_{\bullet}$ is a contractible chain complex of free one-parameter modules for $l \in \{1, 2\}$. In this case, choosing a chain contraction of $\Omega_1 F_{\bullet}$ can be done as follows.

Let C_{\bullet} be a chain complex of free one-parameter persistence modules, and fix a basis of C_{\bullet} . Applying the Standard Algorithm (Algorithm 1) to the boundary matrices of C_{\bullet} gives a persistence basis of C_{\bullet} ; cf. Example 2.2.7. If colim C_{\bullet} is acyclic, then a persistence basis of C_{\bullet} is a collection $(c_i)_{i \in J} \subseteq C_{\bullet}$ for some index set J such that $(\partial c_i)_{i \in J} \cup (c_i)_{i \in J}$ is a basis of C_{\bullet} . In this case, each pair $(\partial c_i, c_i)$ represents a bar $(g(\partial c_i), g(c_i)) \in \text{barc } H_{\bullet}(C_{\bullet})$. If C_{\bullet} is acyclic, then barc $H_{\bullet}(C_{\bullet})$ contains no bars of non-zero length; which means that $g(\partial c_i) = g(c_i)$ for all $i \in J$. In particular, the assignment

$$C_{\bullet} \to C_{\bullet+1}, \quad \partial c_i \mapsto c_i, \quad c_i \mapsto 0$$

is a well-defined chain contraction of C_{\bullet} . Applying this to the contractible $\operatorname{colim}_{\{l\}} F_{\bullet}$ complexes of one-parameter modules for $l \in \{1, 2\}$ and (by an analogous construction) to the contractible complex $\Omega_0 F_{\bullet}$ of vector spaces gives the desired contractions s_1 of $\Omega_1 F_{\bullet}$ and s_0 of $\Omega_0 F_{\bullet}$. In the light of Corollary 3.8.3, these can be used to construct a fringe presentation of M.

Chapter 4

Persistent cohomology of freely resolved cochain complexes

In this section, we introduce a different approach to computing a minimal free resolution of the two-parameter persistent cohomology of a one-critically \mathbb{Z}^2 -filtered simplicial complex K_* . Namely, we compute minimal free resolutions of $H^{\bullet}(K_*)$ and $H^{\bullet}(K, K_*)$ directly from the cochain complexes $C^{\bullet}(K_*)$ and $C^{\bullet}(K, K_*)$, respectively. Using Corollary 3.2.11, these can be used to obtain minimal free resolutions of $H_{\bullet}(K_*)$ and $H_{\bullet}(K, K_*)$. If $K = \operatorname{colim}_{\mathbb{Z}^2} K_*$ is acyclic, then the long exact sequence (2.3) shows that $H^d(K_*) \cong H^{d+1}(K, K_*)$, which is why we study both absolute and relative cohomology here.

Arguably, computing $H^d(K_*)$ or $H^{d+1}(K, K_*)$ suggests itself more easily than computing $H^{d+2}(N^{\bullet}(K_*))$, given that one is ultimately interested in computing $H_d(K_*)$. The computational challenge lies in the fact that as mentioned earlier, for two and more parameters, $C^{\bullet}(K_*)$ and $C^{\bullet}(K, K_*)$ are no cochain complexes of free modules. Actually, these are not even complexes of finitely generated modules.

To address this, we choose a $z > g(\sigma)$ for all $\sigma \in K_*$, and replace $C^{\bullet}(K_*)$ and $C^{\bullet}(K, K_*)$ by complexes $R_z C^{\bullet}(K_*)$ and $R_z C^{\bullet}(K, K_*)$ of finitely generated modules. We choose a free resolution C^d_{\bullet} of $R_z C^d(K_*)$ for each d, and use these to compute minimal free resolutions of $H^d(R_z C^{\bullet}(K_*))$ and $H^{d+1}(R_z C^{\bullet}(K, K_*))$.

We start in Section 4.1 by deriving explicit formulas for free resolutions of the cocycles, coboundaries and cohomology of an arbitrary cochain complex C^{\bullet} of modules C^{d} , for which a free resolution $F_{\bullet}^{d} \to C^{d}$ are known for each d; see Theorem 4.1.7. In Section 4.2, we apply this to the finitely generated modules $R_{z}C^{\bullet}(K_{*})$ and $R^{z}C^{\bullet}(K, K_{*})$. This gives minimal free resolutions of $H^{\bullet}(K_{*})$ and $H^{\bullet}(K, K_{*})$, see Propositions 4.2.5 and 4.2.7. In Section 4.4, we devise further algorithms to efficiently compute a minimal free resolution of $H^{d+1}(K, K_{*})$. In particular, we develop an optimization scheme analogous to clearing; see Proposition 4.4.6. We also show in Theorem D (page 90) that if K is acyclic, then a minimal free resolution of $H^{d+1}(K, K_{*})$; that is, it is not necessary to know the coboundary morphism $\delta^{d+1} \colon C^{d}(K, K_{*}) \to C^{d+1}(K, K_{*})$; that is, it is not necessary to know the coboundary morphism δ^{d+2} explicitly.

Recall from Corollary 2.2.11 that in one-parameter persistence, $H^{\bullet}(K_*)$ and $H^{\bullet}(K, K_*)$ determine each other uniquely up to isomorphism even if K is not acyclic. We show in Theorem B(b) (page 98) that this is not the case for two or more parameters. This is analogous to the example presented in Section 3.6.

Parts of the results presented in this chapter are joint work with Ulrich Bauer and Michael Lesnick.

4.1 Free resolutions of (co)kernels, images and subquotients

Let $A = k[x_1, x_2]$, Let L, M and N be finitely generated A-modules, let $L \xrightarrow{f} M \xrightarrow{g} N$ be morphisms such that gf = 0. let L_{\bullet}, M_{\bullet} and N_{\bullet} be free resolutions of L, M and N, and let $f_{\bullet}: L_{\bullet} \to M_{\bullet}$ and $g_{\bullet}: M_{\bullet} \to N_{\bullet}$ be lifts of f and g. In this section, we derive closed expressions for free resolutions of ker g, im f, coker f and ker g/ im f. Recall that according to Corollary 2.3.20, kernels of morphisms of free modules are free. In particular, all pullbacks in the following lemmas are free, because they are difference kernels of morphism of free modules.

Lemma 4.1.1 (Kernel resolution). Let $g: M \to N$ be a morphism of finitely generated modules, let M_{\bullet} and N_{\bullet} be free resolutions of M and N of length 2, and let $g_{\bullet}: M_{\bullet} \to N_{\bullet}$ be a lift of g. Then the sequence

$$K_{\bullet}: \qquad 0 \to M_2 \xrightarrow{k_2:=\begin{pmatrix} m_2 \\ -g_2 \end{pmatrix}} M_1 \oplus N_2 \xrightarrow{k_1:=\begin{pmatrix} m_1 & 0 \\ g_1 & m_2 \end{pmatrix}} M_0 \times_{N_0} N_1 \to \ker g \tag{4.1}$$

is a free resolution of ker g, and

is a lift of the canonical inclusion $\kappa \colon \ker g \hookrightarrow M$.

Remark 4.1.2. The matrices in (4.1) have morphisms as entries. This is a convenient way to write down morphisms from and to direct sums of modules. Note that a priori, the matrix $k_1 = \begin{pmatrix} m_1 & 0 \\ g_1 & m_2 \end{pmatrix}$ from the statement of the lemma only defines a morphism $M_1 \oplus N_2 \to M_0 \oplus N_1$. It is part of the lemma that this morphism factors through the submodule $M_0 \times_{N_0} N_1$ of $M_0 \oplus N_1$.

Proof of Lemma 4.1.1. During the proof, we will explain the construction of the first row of the following (supposedly commutative) diagram:

$$K_{\bullet}: \quad 0 \to M_{2} \xrightarrow{k_{2}} M_{1} \oplus N_{2} \xrightarrow{k_{1}} M_{0} \times_{N_{0}} N_{1} \xrightarrow{k_{0}} \ker g \to 0$$

$$\downarrow^{\kappa_{2}} \qquad \downarrow^{\kappa_{1}} \qquad \downarrow^{\kappa_{0}} \qquad \downarrow^{\kappa}$$

$$M_{\bullet}: \quad 0 \to M_{2} \xrightarrow{m_{2}} M_{1} \xrightarrow{m_{1}} M_{0} \xrightarrow{m_{0}} M \longrightarrow 0$$

$$\downarrow^{g_{2}} \qquad \downarrow^{g_{1}} \qquad \downarrow^{g_{0}} \qquad \downarrow^{g}$$

$$N_{\bullet}: \quad 0 \to N_{2} \xrightarrow{m_{2}} N_{1} \xrightarrow{m_{1}} N_{0} \xrightarrow{m_{0}} N \longrightarrow 0.$$

$$(4.3)$$

Its middle and lower row are exact, and the squares between them commute. We construct maps k_{\bullet} and κ_{\bullet} that render the first row exact and the diagram commutative.

Claim. There exists a surjective morphism $M_0 \times_{N_0} N_1 \to \ker g$.

Proof of claim. Let $K_0 \coloneqq M_0 \times_{N_0} N_1 = \ker(g_0, -n_1)$. The morphism $\kappa_0 \coloneqq (\operatorname{id}_{M_1}, 0) \colon K_0 \to M$ satisfies

$$gm_0\kappa_0 = (gm_0, 0)|_{K_0} \stackrel{(\uparrow)}{=} n_0(0, n_1)|_{K_0} = 0,$$

using the definition of K_0 in (†). By the universal property of ker g, there exists a unique morphism $k_0: K_0 \to \ker g$ such that $m_0 \kappa_0 = \kappa k_0$. To show that k_0 is surjective, let $x \in \ker g$. Since m_0 is surjective, there exists an $x' \in M_0$ such that $m_0(x') = \kappa(x)$. We obtain

$$n_0g_0(x') = gm_0(x') = g(x) = 0.$$

Exactness of the lower row implies that there exists a $y \in N_1$ such that $n_1(y) = g_0(x')$. By definition of K_0 , this means that $(x', y) \in K_0$. We obtain $\kappa k_0(x', y) = m_0(x') = \kappa(x)$. Since κ is injective, this implies that $k_0(x', y) = x$, hence k_0 is surjective. This proves the claim.

Now, consider ker k_0 and the pullback

$$K_0 \times_{M_0} M_1 = \ker(\begin{smallmatrix} g_0 & -n_1 & 0\\ \mathrm{id}_{M_0} & 0 & -m_1 \end{smallmatrix}) \subseteq M_0 \oplus N_1 \oplus M_1.$$
(4.4)

By injectivity of κ , we have ker $k_0 = \ker \kappa k_0 = \ker m_0 \kappa_0$. Applying the claim to the morphism $m_0 \kappa_0 \colon K_0 \to M$ (instead of g) shows that the morphism $(\mathrm{id}_{K_0}, 0) \colon K_0 \times_{M_0} M_1 \to K_0$ surjects onto ker k_0 . For $K_1 \coloneqq M_1 \oplus N_2$, consider the morphism

$$\alpha \coloneqq \begin{pmatrix} m_1 & 0\\ g_1 & -n_2\\ \mathrm{id}_{M_1} & 0 \end{pmatrix} \colon K_1 \longrightarrow K_0 \oplus M_1.$$
(4.5)

Because n_2 is injective, so is α . We show that im $\alpha = K_0 \times_{M_0} M_1$. Since

$$\begin{pmatrix} g_0 & -n_1 & 0\\ \mathrm{id}_{M_0} & 0 & -m_1 \end{pmatrix} \begin{pmatrix} m_1 & 0\\ g_1 & -n_2\\ \mathrm{id}_{M_1} & 0 \end{pmatrix} = 0.$$
(4.6)

we have $\operatorname{im} \alpha \subseteq K_0 \times_{M_0} M_1$. To show that $\operatorname{im} \alpha = K_0 \times_{M_0} M_1$, let $(x, y, x') \in K_0 \times_{M_0} M_1$. According to (4.4), we get that $x = m_1(x')$ and

$$n_1(g_1(x') - y) = g_0(m_1(x')) - n_1(y) = g_0(x) - n_1(y) = 0.$$
(4.7)

Exactness of the lower row implies that $g_1(x') - y = n_2(y')$ for some $y' \in N_2$. Therefore,

$$\alpha\begin{pmatrix} x'\\ y' \end{pmatrix} = \begin{pmatrix} m_1 & 0\\ g_1 & -n_2\\ \mathrm{id}_{M_1} & 0 \end{pmatrix} \begin{pmatrix} x'\\ y' \end{pmatrix} = \begin{pmatrix} m(x')\\ g_1(x') - n_2(y')\\ x' \end{pmatrix} = \begin{pmatrix} x\\ y\\ x' \end{pmatrix}.$$
(4.8)

This shows that α maps K_1 isomorphically to $K_0 \times_{M_0} M_1$. We obtain that

$$k_1 = \begin{pmatrix} m_1 & 0\\ g_1 & -n_2 \end{pmatrix} \colon K_1 \xrightarrow{\alpha} K_0 \times_{M_0} M_1 \xrightarrow{(\mathrm{id}_{K_0}, 0)} K_0$$
(4.9)

surjects onto ker k_0 . To obtain K_2 , we calculate

$$\ker k_{1} = \ker \begin{pmatrix} m_{1} & 0 \\ g_{1} & -n_{2} \end{pmatrix} = \ker(m_{1}, 0) \cap \ker(g_{1}, -n_{2}) \stackrel{(i)}{\cong} (M_{2} \oplus N_{2}) \cap \ker(g_{1}, -n_{2})$$

$$\stackrel{(ii)}{\cong} (M_{2} \oplus N_{2}) \cap \ker(g_{1}m_{2}, -n_{2}) = \ker(n_{2}g_{2}, -n_{2}) \stackrel{(iii)}{\cong} \ker(g_{2}, -\operatorname{id}_{N_{2}})$$

$$\cong \{(m, g_{2}(m)) \mid m \in M_{2}\} \cong M_{2}, \quad (4.10)$$

where we use exactness of M_{\bullet} in (i), injectivity of m_2 in (ii), and injectivity of n_2 in (iii). With $K_2 \coloneqq M_2$, we obtain that the morphism

$$k_2 \coloneqq \begin{pmatrix} m_2 \\ -g_2 \end{pmatrix} \colon K_2 \xrightarrow{\begin{pmatrix} \operatorname{id}_{M_2} \\ -g_2 \end{pmatrix}} \ker(g_2, \operatorname{id}_{N_2}) = \ker(g_1 m_2, n_2) \xrightarrow{\begin{pmatrix} m_2 & 0 \\ 0 & \operatorname{id}_{N_2} \end{pmatrix}} K_1$$
(4.11)

is injective and surjects onto ker k_1 . This completes the resolution K_{\bullet} of ker g. One sees easily that $\kappa_1 = (\mathrm{id}_{M_1}, 0)$ and $\kappa_2 = \mathrm{id}_{M_2}$ make the diagram (4.3) commute.

Lemma 4.1.3 (Image resolution). Let $f: L \to M$ be a morphism of finitely generated modules, let L_{\bullet} and M_{\bullet} be free resolutions of length 2 of L and M, and let $f_{\bullet}: L_{\bullet} \to M_{\bullet}$ be a lift of f. Then

$$J_{\bullet}: \qquad 0 \to M_2 \xrightarrow{j_2:=\begin{pmatrix} 0\\m_2 \end{pmatrix}} L_0 \times_{M_0} M_1 \xrightarrow{j_1:=(\mathrm{id}_{L_0},0)} L_0 \xrightarrow{j_0} \mathrm{im}\, f \tag{4.12}$$

is a free resolution of im f, and

is a lift of the canonical map $\pi \colon L \to \operatorname{im} f$.

Proof. The morphism $j_0 \coloneqq \pi l_0 \colon L_0 \to \inf f$ is surjective because it is a composition of two surjective morphisms. Applying Lemma 4.1.1 to j_0 , we get that $j_1 \coloneqq (\operatorname{id}_{L_0}, 0) \colon L_0 \times_{M_0} M_1 \to L_0$ surjects onto ker j_0 . Lastly, ker $j_1 = 0 \times_{M_0} M_1 \cong \ker m_1 \cong M_2$; hence the map $j_2 \coloneqq \binom{0}{m_2} \colon M_2 \to L_0 \times_{M_0} M_1$ maps M_2 isomorphically to ker j_1 , so (4.12) is a free resolution of $\inf f$. Commutativity of (4.13) can be seen immediately.

Lemma 4.1.4 (Image inclusion). Let $L \xrightarrow{f} M \xrightarrow{g} N$ be morphisms of finitely generated modules with gf = 0, let L_{\bullet}, M_{\bullet} and N_{\bullet} be free resolutions of length 2, and let $f_{\bullet} \colon L_{\bullet} \to M_{\bullet}$ and $g_{\bullet} \colon M_{\bullet} \to N_{\bullet}$ be lifts of f and g. Then there are maps $\zeta \colon L_{0} \to N_{1}$ and $\eta \colon L \times_{M_{0}} M_{1} \to N_{2}$ such that the inclusion map $i \colon \text{im } f \hookrightarrow \text{ker } g$ is lifted by the following morphism of free resolutions:

Proof. Consider the diagram

According to Lemmas 4.1.1 and 4.1.3, all rows are exact, and all squares except between rows J_{\bullet} and K_{\bullet} commute. We construct the dashed maps i_{\bullet} that make the diagram commute.

By commutativity, $n_0g_0f_0 = gfl_0 = 0$. Therefore, the morphism $g_0f_0: L_0 \to N_0$ factors uniquely through ker n_0 . By the universal property of the free (and in particular projective) module N_1 , there is a morphism $\zeta: L_0 \to N_1$ such that $g_0f_0 = n_1\zeta$. Then $(g_0, -n_1)(\frac{f_0}{\zeta}) = 0$, so the morphism $i_0 := \begin{pmatrix} f_0 \\ \zeta \end{pmatrix}: L_0 \to M_0 \oplus N_1$ factors through $K_0 = \ker(g_0, -n_1)$. This shows commutativity of (*).

For the construction of i_1 , consider the map $(\zeta, -g_1): L_0 \oplus M_1 \to N_1$. Its restriction to $J_1 = \ker(f_0, -m_1)$ satisfies

$$n_1(\zeta, -g_1)|_{J_1} \stackrel{a)}{=} (g_0 f_0, -n_1 g_1)|_{J_1} \stackrel{b)}{=} g_0(f_0, -m_1)|_{J_1} \stackrel{c)}{=} 0,$$
(4.16)

using the definition of ζ in a), the commutativity of the lower row in b) and the definition of J_1 in c). Bt the universal property of ker n_1 , there is a unique morphism $\eta: J_1 \to \ker n_1 = N_2$ such that $n_2\eta = (\zeta, -g_1)|_{J_1}$. We check that the map $i_1 := \begin{pmatrix} 0 & \mathrm{id}_{M_1} \\ \eta \end{pmatrix} : J_1 \to K_1$ makes the square (**) commute. We calculate

$$k_1 i_1 = \begin{pmatrix} m_1 & 0 \\ g_1 & n_2 \end{pmatrix} \begin{pmatrix} 0 & \mathrm{id} \\ \eta \end{pmatrix} |_{J_1} = \begin{pmatrix} 0 & m_1 \\ \zeta & 0 \end{pmatrix} |_{J_1} \stackrel{(\dagger)}{=} \begin{pmatrix} f_0 & 0 \\ \zeta & 0 \end{pmatrix} |_{J_1} = \begin{pmatrix} f_0 \\ \zeta \end{pmatrix} (\mathrm{id}_{K_0}, 0) |_{J_1} = i_0 j_1,$$

using the definition of J_1 in (†). Therefore, (**) commutes.

Lastly, we check that $i_2 = id_{M_2}$ makes the square (* * *) in commute. By the definition of η , we get

$$n_2\eta j_2 = (\zeta, -g_1)({0 \atop m_2}) = -n_2g_2.$$

As n_2 is injective, this implies $\eta j_2 = -g_2$. Therefore,

$$k_2 i_2 = \begin{pmatrix} m_2 \\ -g_2 \end{pmatrix} = \begin{pmatrix} 0 & \mathrm{id}_{M_1} \\ \eta \end{pmatrix} \begin{pmatrix} 0 \\ m_2 \end{pmatrix} = i_1 j_2$$

so (* * *) commutes.

Lemma 4.1.5 (Cokernel resolution). Let $f: L \to M$ be a morphism of finitely generated modules, let L_{\bullet} and M_{\bullet} be free resolutions of length 2 of L and M, and let $f_{\bullet}: L_{\bullet} \to M_{\bullet}$ be a lift of f. Then

$$K'_{\bullet}: \qquad 0 \longrightarrow L_0 \times_{M_0} M_1 \xrightarrow[k'_2]{} L_0 \oplus M_1 \xrightarrow[k'_1]{} M_0 \xrightarrow[k'_0]{} \operatorname{coker} f \tag{4.17}$$

is a free resolution of $\operatorname{coker} f$, and

is a lift of the canonical morphism $\kappa' \colon M \to \operatorname{coker} f$.

Proof. The morphism $k'_0 \coloneqq \kappa' m_0 \colon M_0 \to \operatorname{coker} f$ is surjective because it is a composition of surjective morphisms. Consider the morphism $k'_1 \coloneqq (-f_0, m_1) \colon L_0 \oplus M_1 \to M_0$. Because

$$k_0'k_1' = k'm_0(-f_0,m_1) = (-k'fl_0,m_0m_1) = 0,$$

the morphism k'_1 factors through ker k'_0 . To show that k'_1 surjects onto ker k'_0 , let $x \in \ker k'_0$. Then $m_0(x) \in \ker \kappa' = \inf f$, so there exists a $y \in L_0$ such that $fl_0(y) = m_0(x)$. Then $m_0(x - f_0(y)) = 0$, so there exist $x' \in M_1$ such that $x = -f_0(y) + m_1(x')$. This shows that k'_1 surjects onto ker $k'm_0$. Then $L_0 \times_{M_0} M_1 = \ker(-f_0, m_1) = \ker k'_1$ is a free module, so (4.17) is a free resolution of coker f. The diagram (4.18) clearly commutes.

Combining the above statements, we obtain the following expression for a free resolution of the subquotient ker $g/\inf f$:

Theorem 4.1.6 (Homology resolution). Let $L \xrightarrow{f} M \xrightarrow{g} N$ be maps of finitely generated modules with gf = 0, let L_{\bullet}, M_{\bullet} and N_{\bullet} be free resolutions of L, M and N of length 2, let $f_{\bullet} : L_{\bullet} \to M_{\bullet}$ and $g_{\bullet} : M_{\bullet} \to N_{\bullet}$ be lifts of f and g, and let ζ and η be as in Lemma 4.1.4. Then the sequence

$$H_{\bullet}: \qquad L_0 \times_{M_0} M_1 \xrightarrow{h_2 \coloneqq \begin{pmatrix} \operatorname{Id}_{L_0} & 0 \\ 0 & \operatorname{id} \end{pmatrix}} L_0 \oplus M_1 \oplus N_2 \xrightarrow{h_1 \coloneqq \begin{pmatrix} -f_0 & m_1 & 0 \\ -\zeta & g_1 & n_2 \end{pmatrix}} M_0 \times_{N_0} N_1 \to \frac{\ker g}{\operatorname{im} f}$$
(4.19)

is a free resolution of ker $g/\operatorname{im} f$, and

is a lift of the canonical map $q \colon \ker g \to \ker g / \operatorname{im} f$.

Proof. Applying Lemma 4.1.5 to the morphism $i_{\bullet}: J_{\bullet} \to K_{\bullet}$ from Lemma 4.1.4 gives that

$$0 \to L_0 \times_{K_0} K_1 \to L_0 \oplus \underbrace{M_1 \oplus N_2}_{K_1} \xrightarrow{h_1} \underbrace{M_0 \times_{N_0} N_1}_{K_0} \to \ker g / \operatorname{im} f \tag{4.21}$$

is a free resolution of ker $g/\operatorname{im} f$, and that

is a lift of q. It remains to show that $L_0 \times_{K_0} K_1 \cong L_0 \times_{M_0} M_1$. Because $K_0 \subseteq M_0 \oplus N_1$, we get that

$$L_0 \times_{K_0} K_1 = L_0 \times_{M_0 \oplus N_1} K_1 = \ker \left(\begin{smallmatrix} f_0 & -m_1 & 0 \\ \eta & -g_1 & -n_2 \end{smallmatrix} \right) \subseteq L_0 \oplus M_1 \oplus N_2.$$

The morphism

$$h_2 \coloneqq \begin{pmatrix} \operatorname{id} & 0\\ 0 & \operatorname{id}\\ \eta \end{pmatrix} \colon L_0 \times_{M_0} M_1 \longrightarrow L_0 \oplus M_1 \oplus N_2.$$

$$(4.23)$$

is obviously injective. We show that im $h_2 = L_0 \times_{K_0} K_1$. Since

$$\begin{pmatrix} f_0 & -m_1 & 0 \\ \zeta & -g_1 & -n_1 \end{pmatrix} \begin{pmatrix} \operatorname{id}_{L_0} & 0 \\ 0 & \operatorname{id}_{M_1} \end{pmatrix} \Big|_{H_2} = \begin{pmatrix} f_0 & -m_1 & 0 \\ 0 & 0 & 0 \end{pmatrix} \Big|_{H_2} = 0,$$
 (4.24)

the morphism h_2 factors through $L_0 \times_{K_0} K_1$, so $\operatorname{im} h_2 \subseteq L_0 \times_{K_0} K_1$. To show equality, let $(x, y, z) \in \ker h_1$. In particular, $(x, y) \in L_0 \times_{M_0} M_1 = \ker(f_0, -m_1)$. By definition of η , we get $n_2\eta(\frac{x}{y}) = \zeta(x) - g_1(y) = n_2(z)$. By injectivity of n_2 , we get $z = \eta(\frac{x}{y})$. This shows that h_2 maps $L_0 \times_{M_0} M_1$ isomorphically to $L_0 \times_{K_0} K_1$.

Remark (Homology as a kernel). Let $L_{\bullet}, M_{\bullet}, N_{\bullet}, f_{\bullet}$ and g_{\bullet} be as in Theorem 4.1.6. From Lemmas 4.1.3 and 4.1.5 we get free resolutions K'_{\bullet} and J'_{\bullet} of coker f and im g, respectively. Analogously to Lemma 4.1.4, the morphisms κ'_{\bullet} and i'_{\bullet} in the commutative diagram

are lifts of the canonical morphisms $\kappa' \colon M \to \operatorname{coker} f$ and $i' \colon \operatorname{coker} f \to \operatorname{im} g$, respectively. Applying Lemma 4.1.1 to i'_{\bullet} gives the same free resolution of $\ker g / \operatorname{im} f = \operatorname{coker} i \cong \ker i'$ as in (4.19).

Recall from Remark 4.1.2 that the matrix used to define, for example, h_1 , a priory defines a morphism $L_0 \oplus M_1 \oplus N_2 \to M_0 \oplus N_1$, and we mean by the notation that it factors uniquely through $M_0 \times_{N_0} N_1$. When it comes to actually computing graded matrices representing the morphisms h_2 , h_1 , we have to be a bit more precise. In particular, we have to distinguish a modules and their submodules more strictly, such as $M_0 \oplus N_1$ and its submodule $M_0 \times_{N_0} N_1$. Therefore, we formulate the following "pedantic" version of Theorem 4.1.6:

Theorem 4.1.7. Let $L \xrightarrow{f} M \xrightarrow{g} N$ be maps of finitely generated modules with gf = 0, let L_{\bullet}, M_{\bullet} and N_{\bullet} be free resolutions of L, M and N of length 2, let $f_{\bullet} \colon L_{\bullet} \to M_{\bullet}$ and $g_{\bullet} \colon M_{\bullet} \to N_{\bullet}$ be lifts of f and g, and let ζ and η be as in Lemma 4.1.4. Then there exist morphisms h_0 and $h_1 = \begin{pmatrix} \eta \\ \eta \end{pmatrix}$ that make the diagram

of free modules commute, where ϕ and ψ are the canonical morphisms. Moreover, h_1 and h_2 form a free resolution of ker $g/\inf f$.

Let C^{\bullet} be a cochain complex of not necessarily free A-modules. Assume we have fixed a free resolution

$$0 \to C_2^d \xrightarrow{c_2^d} C_1^d \xrightarrow{c_1^d} C_0^d$$

of C^d for each q, and a lift $\delta^d_{\bullet}: C^{d-1}_{\bullet} \to C^d_{\bullet}$ of the coboundary morphism δ^d . Then Theorem 4.1.7 states that there is a commutative diagram

$$0 \longrightarrow \underbrace{\ker(\delta_{0}^{d}, -c_{1}^{d})}_{=:H^{d}(C^{\bullet})_{2}} \xrightarrow{-\overset{h_{2}^{d}}{\longrightarrow}} \underbrace{C_{0}^{d-1} \oplus C_{1}^{d} \oplus C_{2}^{d+1}}_{=:H^{d}(C^{\bullet})_{1}} \xrightarrow{-\overset{h_{1}^{d}}{\longrightarrow}} \underbrace{\ker(\delta_{0}^{d+1}, -c_{1}^{d+1})}_{=:H^{d}(C^{\bullet})_{0}}$$

$$(4.26)$$

$$\overset{\kappa^{d-1}}{\longrightarrow} \underbrace{\begin{pmatrix} \operatorname{id}_{c_{2}^{d+1}} \\ \operatorname{id}_{c_{2}^{d+1}} \end{pmatrix}}_{=:H^{d}(C^{\bullet})_{1}} \xrightarrow{\downarrow} \ker(\delta_{0}^{d+1}, -c_{1}^{d+1})}_{\downarrow \kappa^{d}} \xrightarrow{\downarrow} \kappa^{d}$$

$$(\delta_{0}^{d}, -c_{1}^{d}) \underbrace{\begin{pmatrix} \operatorname{id}_{c_{1}^{d}} \\ \operatorname{id}_{c_{-}\delta_{1}^{d+2}} \end{pmatrix}}_{C_{0}^{d}} \xrightarrow{\begin{pmatrix} \operatorname{id}_{c_{1}^{d}} \\ \operatorname{id}_{c_{-}\delta_{1}^{d+2}} \end{pmatrix}} \xrightarrow{\begin{pmatrix} \operatorname{id}_{c_{1}^{d}} \\ \operatorname{id}_{c_{-}\delta_{1}^{d+1}} \\ \operatorname{id}_{c_{0}^{d+1}} \\ \overset{(\delta_{0}^{d+1}, -c_{1}^{d+1})}{\overset{(\delta_{0}^{d+1}, -c_{1}^{d+1})}} \xrightarrow{\downarrow} K^{d}$$

whose first line is a free resolution $H^d(C^{\bullet})_{\bullet}$ of $H^d(C^{\bullet})$. Note that $H^d(C^{\bullet})_2 = H^{d-1}(C^{\bullet})_0$.

Remark (Theorem 4.1.7 as a mapping cone). If the coboundary morphisms $\delta_{\bullet}^d: C_{\bullet}^d \to C_{\bullet}^{d+1}$ satisfy $\delta_i^{d+1}\delta_i^d = 0$ for all *i* and *q*, then then is an easier way to prove that (4.26) is a free resolution of $H^d(C^{\bullet})$. Namely, in this case, $0 \to C_2^{\bullet} \to C_1^{\bullet} \to C_0^{\bullet}$ is an exact sequence of cochain complexes. Recall that for any morphism *f* of cochain complexes, coker *f* is quasi-isomorphic to the mapping cone of *f*, which we denote by cone *f* [127, §1.5.8]. In particular, the iterated mapping cone

$$\hat{C}^{\bullet} \coloneqq \operatorname{cone}(\operatorname{cone}(C_2^{\bullet} \to C_1^{\bullet}) \to C_0^{\bullet}))$$

is a cochain complex of free modules that satisfies $H^d(\hat{C}^{\bullet}) \cong H^d(C^{\bullet})$ for all q, and

$$0 \to Z^{d-1}(\hat{C}^{\bullet}) \to \hat{C}^d \to Z^d(\hat{C}^{\bullet})$$
(4.27)

is a free resolution of $H^d(\hat{C}^{\bullet}) \cong H^d(C^{\bullet})$. Unravelling the definition shows that \hat{C}^{\bullet} is the cochain complex of free modules $\hat{C}^d = C_0^d \oplus C_1^{d+1} \oplus C_2^{d+2}$ with coboundary morphism

$$\hat{\delta}^{d+1} \coloneqq \begin{pmatrix} -\delta_0^{d+1} & c_1^{d+1} & 0\\ 0 & \delta_1^{d+2} & -c_2^{d+2}\\ 0 & 0 & -\delta_2^{d+3} \end{pmatrix} \colon \hat{C}^d \to C'^{d+1}.$$

One may verify that the morphism

$$Z^{d}(\hat{C}^{\bullet}) = \ker \hat{d}^{d+1} \xrightarrow{\begin{pmatrix} \text{id} & 0 & 0 \\ 0 & \text{id} & 0 \end{pmatrix}} \ker(\delta^{d+1}, -c_{1}^{d+1}) = C_{0}^{d} \times_{C_{0}^{d+1}} C_{1}^{d+1}$$

is an isomorphism. Therefore, the free resolution (4.27) coincides with the one from Theorem 4.1.6.

4.2 Simplicial cohomology

In the following, we apply the above construction to the simplicial cochain complex of a finite, one-critically \mathbb{Z}^2 -filtered simplicial complex $K_* \in \operatorname{Simp}^{\subseteq \mathbb{Z}^2}$. As before, let $K = \operatorname{colim} K_*$. Note that Theorem 4.1.6 does not directly apply to the chain complexes $C^{\bullet}(K_*)$ and $C^{\bullet}(K_*,K)$, because the modules $C^d(K_*)$ and $C^d(K_*,K)$ are not finitely generated. Instead, we replace $C^{\bullet}(K_*)$ by an appropriate chain complex of finitely generated modules $R_z C^{\bullet}(K_*)$. Its homology then allows us to obtain $H^{\bullet}(K_*)$; see below.

Definition 4.2.1. For $z \in \mathbb{Z}^2$, let $R_z, E_z : \operatorname{Vec}^{\mathbb{Z}^2} \to \operatorname{Vec}^{\mathbb{Z}^2}$ be the exact functors with

$(R_z M)_w = \{ \begin{smallmatrix} M_w & ext{if } -z \leq w, \\ 0 & ext{otherwise} \end{smallmatrix} \}$	$(R_z M)_{ww'} = \begin{cases} M_{ww'} & \text{if } -z \le w \le w', \\ 0 & \text{otherwise} \end{cases}$
$(E_z M)_w = M_{-z \lor w}$	$(E_z M)_{ww'} = M_{-z \lor w, z \lor w'},$

called *restriction* and *extension* of M with respect to z.

As it will turn out later, using -z rather than z in the definition is for convenience. Example 4.2.2. For z = 0, the functor R_z and E_z maps M to

The idea behind these and the following definition is that we use R_z to "cut off" a non-finitely generated module and replace it by a finitely generated one. Namely if $M \in \text{vec}^{\mathbb{Z}^2}$ is a pointwise finite dimensional module, then $R_z M$ is finitely generated. In particular, if F is a free module of finite rank, then $R_z F^*$ is finitely generated. Furthermore, if $M_w \cong M_{-z \lor w}$ whenever $-z \nleq w$, then $M \cong E_z R_z M$.

The idea behind this is that if M is non-finitely generated, but there is a $z \in Z^2$ such that everything "interesting" in M happens below z, we may replace M by the finitely generated module $R_z M$, do our calculations, and extend the result to \mathbf{Z}^2 using E_z .

For example, recall that the dual $F(w)^*$ of a free module F(w) for some $w \in \mathbb{Z}^2$ has the components $(F(w)^*)_u = \begin{cases} k & \text{if } u \leq -w \\ 0 & \text{otherwise} \end{cases}$. Therefore, it satisfies $(F(v)^*)_w \cong (F(v)^*)_{-z \vee w}$ whenever $w \geq -z$. We obtain:

Lemma 4.2.3. If F is a free module such that z > w for all $w \in \operatorname{rk} F$, then $F^* \cong E_z R_z F^*$.

Fix a $z = (z_1, z_2) \in \mathbb{Z}^2$ as in the lemma. We use the following notation to write down a free resolution of $R_z F^*$.



Figure 4.1: The free resolution (4.28) of the restricted module $R_w F(w)^*$, where w < z. Note that the diagrams are drawn with the structure morphisms pointing downward. The symbols •, • and × indicate the grades of the generators, relations and 2-syzygies of this free resolution.

Definition 4.2.4. For $w = (w_1, w_2) \in \mathbb{Z}^2$ with w < z, define

$$\begin{split} \forall w \coloneqq (-w_1 + 1, -z_2), & w \vDash (-z_1, -z_2), \\ w \coloneqq (-w_1 + 1, -w_2 + 1), & w_{\perp} \coloneqq (-z_1, -w_2 + 1). \end{split}$$

We call coordinates of the form w, w_{\perp} or w^{\neg} infinite coordinates and coordinates of the form w_{\perp} finite coordinates. The infinite coordinates are the ones that contain at least one component from z. The definition is chosen such that the sequence

$$0 \to F(\underline{w}) \xrightarrow{\begin{pmatrix} 1 \\ -1 \end{pmatrix}} F(\underline{w}) \oplus F(\overline{w}) \xrightarrow{(1,1)} F(\overline{w}) \to R_z F(w)^*$$

$$(4.28)$$

is a minimal free resolution of $R_z F(w)^*$; see Figure 4.1. If F is a finite rank free module, then taking a direct sum of (4.28) gives a minimal free resolution of $R_z F^*$. Let C_{\bullet} be a chain complex of finite rank free modules, and let $C^{\bullet} := (C_{\bullet})^*$ be its dual cochain complex. If z > w for all $w \in \operatorname{rk} C_{\bullet}$, then it follows from exactness of the functors R_z and E_z that

$$E_z H^d(R_z C^{\bullet}(K_*)) \cong E_z R_z H^d(C^{\bullet}(K_*)) \cong H^d(C^{\bullet}(K_*)).$$

Now, $R_z C^{\bullet}(K_*)$ is a cochain complex of finitely generated modules. Therefore, we can compute a minimal free resolution of $H^d(C^{\bullet}(K_*))$ by taking the cochain complex $R_z C^{\bullet}(K_*)$ of finitely generated modules, compute a minimal free resolution of its cohomology, and extend this to a free resolution of $H^d(C^{\bullet}(K_*))$, using the functor E_z .

Remark. The last step deserves a more detailed comment. On the level of resolutions, obtaining a minimal free resolution of $E_z H^d(R_z C^{\bullet}(K_*))$ from one of $H^d(R_z C^{\bullet}(K_*))$ involves some calculation. However, no calculation is needed if the ultimate goal is to obtain a minimal free resolution of $H^d(C^{\bullet}(K_*))$. We will explain this now. We assume that $H^{\bullet}(K_*)$ has bounded support. Let $F_{\bullet} \to H^d(R_z C^{\bullet}(K_*))$ be a minimal free resolution. Then, according to Corollary 3.2.11, $F'_{\bullet} \coloneqq (\nu F_{\bullet}[2]\langle \epsilon \rangle)^*$ is a minimal free resolution of $H^d(R_z C^{\bullet}(K_*))^*$. It will contain some relations and 2-syzygies of infinite grades. For $v = (v_1, v_2)$ and $w = (w_1, w_2) \in \mathbb{Z}^2$, we say $v \ll w$ if $v_1 < w_1$ and $v_2 < w_2$. Let G_{\bullet} and G'_{\bullet} be the chain complex with

for i = 0, 1, 2, such that $F'_i = G_i \oplus G'_i$ for all i. One may check that $\partial^F_{\bullet}(G_{\bullet}) \subseteq G_{\bullet}$, which follows because $\operatorname{Hom}(F(v), (w)) = 0$ unless $v \ge w$. Therefore, $(G_{\bullet}, \partial^F_{\bullet}|_{G_{\bullet}})$ defines a chain complex. One may check that G_{\bullet} is a minimal free resolution of $H_{\bullet}(K_*)$. Informally, G_{\bullet} is obtained from F'_{\bullet} by dropping all basis elements of infinite grade.

To compute a minimal free resolution of $H^d(R_z C^{\bullet}(K_*))$, we choose a free resolution C^d_{\bullet} of each module $R_z C^{\bullet}(K_*)$. Then a free resolution of $R_z H^{\bullet}(C^{\bullet}) \cong H^{\bullet}(R_z C^{\bullet})$ is given by (4.26), and (if z > w for all $w \in \operatorname{rk} C_{\bullet}$) a free resolution of $H^{\bullet}(C^{\bullet})$ is obtained by applying E_z to this. One can compute matrices representing this free resolution using Algorithms 6 and 7. In the following, we make this explicit.

4.2.1 Absolute cohomology

Let $K_* \in \operatorname{Simp}^{\subseteq \mathbb{Z}^2}$ be one-critical and $z > g(\sigma)$ for all $\sigma \in K_*$. We construct graded matrices that represent the free resolutions C^d_{\bullet} of each module $C^d(K_*)$ and the lifts δ^d_{\bullet} of the coboundary morphisms δ^d .

For algorithmic reasons, we enumerate the *d*-simplices of K_* reverse colexicographically by grade; that is, for each *d*, we assume that $K^d_* = \{\sigma_{d,1}, \ldots, \sigma_{d,n_d}\}$ with

 $g(\sigma_{d,1}) \succeq_{\operatorname{colex}} \cdots \succeq_{\operatorname{colex}} g(\sigma_{d,n_d});$

see Definition 2.4.2. Then the standard basis of $C_d(K_*) = \bigoplus_{i=1}^{n_d} F(g(\sigma_{di}))$ is colexicographically ordered. We do so because the procedure Ker() (Algorithm 6) requires its input matrix to have its columns colexicographically ordered. For each d, let π_d be the permutation of $\{1, \ldots, n_d\}$ such that

$$g(\sigma_{d,\pi_d(1)}) \succeq_{\text{lex}} \cdots \succeq_{\text{lex}} g(\sigma_{d,\pi(n_d)})$$

It is convenient to assume that the reordering π_d is *stable*; that is, $\pi_d(i) < \pi_d(j)$ for all i < j with $g(\sigma_i) = g(\sigma_j)$.

We define the free modules

$$C_0^d \coloneqq \bigoplus_{i=1}^{n_d} F(\bar{g}(\sigma_{d,i})), \qquad C_1^d \coloneqq C_1^d \oplus C_1^d, \qquad C_2^d \coloneqq \bigoplus_{i=1}^{n_d} F_{\underline{g}}(\sigma_{d,i})), \tag{4.29}$$

where $C_{1_{j}}^{d} \coloneqq \bigoplus_{i=1}^{n_{d}} F(g_{j}(\sigma_{d,i}))$ and $C_{1}^{d} \coloneqq \bigoplus_{i=1}^{n_{d}} F(f(\sigma_{d,\pi_{d}(i)}))$. Note that $C^{d}(K_{*})_{2}$, $C^{d}(K_{*})_{1}$ and $C^{d}(K_{*})_{0}$ have colexicographically ordered bases. Analogously to (4.28), these modules fit into a free resolution

$$0 \to C_2^d \xrightarrow{c_2^d} C_1^d \xrightarrow{c_1^d} C_0^d \to R_z C^d(K_*)$$

$$(4.30)$$

of $R_z C^d(K_*)$. For each d, let E_d be the (ungraded) $n_d \times n_d$ -unit matrix, and Π_d be the permutation matrix with entries $[\Pi_d]_{ij} = \{ \begin{smallmatrix} 1 & \text{if } i = \pi^d(j) \\ 0 & \text{otherwise} \end{smallmatrix}$. With respect to the bases (4.29), the resolution (4.30) is represented by the graded matrices c_2^d and c_1^d with $\mathsf{u}(c_2^d) = \begin{pmatrix} \Pi_d^{-1} \\ -E_d \end{pmatrix}$ and $\mathsf{u}(c_1^d) = (\Pi_d, E_d)$ and the appropriate row grades.

To obtain a graded matrix representing δ^d , let D^d be the (ungraded) matrix representing the coboundary morphism of $C^{\bullet}(K)$ with respect to the standard basis. Then δ^d lifts to the morphism

$$\begin{array}{cccc} C^{d}_{\bullet} \colon & 0 \longrightarrow C^{d}_{2} \xrightarrow{c^{d}_{2}} C^{d}_{1} \oplus C^{d}_{1_{J}} \xrightarrow{c^{d}_{1}} C^{d}_{0} \longrightarrow R_{z}C^{d}(K_{*}) \\ & & \downarrow \\ \delta^{d+1}_{\bullet} & \downarrow \\ C^{d+1}_{\bullet} & \downarrow \\ C^{d+1}_{\bullet} & 0 \longrightarrow C^{d+1}_{2} \xrightarrow{c^{d+1}_{2}} C^{d+1}_{1_{J}} \oplus C^{d+1}_{1_{J}} \xrightarrow{c^{d+1}_{1}} C^{d+1}_{0} \longrightarrow R_{z}C^{d+1}(K_{*}) \end{array}$$

of free resolutions that is represented (with respect to the standard bases (4.29)) by the graded matrices with

$$\mathsf{u}(\delta_0^{d+1}) = D^d, \qquad \mathsf{u}(\delta_1^{d+1}) = \begin{pmatrix} \Pi_{d+1}^{-1} D^{d+1} \Pi_d & 0\\ 0 & D^{d+1} \end{pmatrix}, \qquad \mathsf{u}(\delta_2^{d+1}) = D^d, \tag{4.31}$$

and the appropriate row and column grades.

Proposition 4.2.5 (Absolute cohomology resolution). For each d, there exists the dashed morphisms that make the diagram

commute, such that the first row of the diagram then is a free resolution of $H^d(K_*)$.

Proof. This follows from Theorem 4.1.7. Since $\delta_i^{d+1}\delta_i^d = 0$ for all d, we may choose $\zeta^d = 0$. Actually, $\operatorname{Hom}(C_0^{d-1}, C_1^{d+1}) = 0$ because all basis elements of $C^{d-1}(K_*)_0$ are of grade z, while all basis elements of $C^{d+1}(K_*)_1$ are of grades strictly greater than z. Therefore, $\zeta^d = 0$ is the only possible choice for ζ^d . We obtain that $h_2^d = \binom{\kappa_d^{d-1}}{\eta^d}$, where $\eta: H^d(K_*)_2 \to C_2^{d+1}$ satisfies $c_2^{d+1}\eta^d = (0, -\delta_1^{d+1})\kappa^{d-1}$.

Using the above steps, we obtain the following strategy to compute a minimal free resolution of absolute cohomology:

4.2.6. A pair of graded matrices h_2^d and h_1^d representing a minimal free resolution of $R_z H^d(K_*)$ can be computed in the following steps:

- (i) Compute a graded matrix representing the kernel inclusion κ^d by applying Algorithm 6 to the graded matrix $(\delta_0^{d+1}, -c_1^{d+1})$. Analogously, compute the kernel inclusion κ^{d-1} .
- (ii) Compute h_1^d by solving the linear system $\kappa^d h_1^d = \begin{pmatrix} -\delta_0^d & c_1^d & 0 \\ 0 & \delta_1^{d+1} & c_2^{d+1} \end{pmatrix}$, using Algorithm 7. Proposition 4.2.5 ensures that a solution exists.
- (iii) Compute a graded matrix representing h_2^d . Commutativity of the above diagram implies that h_2^d has to be of the form $h_2^d = \begin{pmatrix} \kappa^{d-1} \\ \eta^{d+1} \end{pmatrix}$, where η^d satisfies $c_2^d \eta^d = (0, \delta_1^{d+1}) \kappa^{d-1}$. Because

$$c_2^{d+1}\eta^d = \begin{pmatrix} \Pi_{d+1}^{-1} \\ -E_{d+1} \end{pmatrix} \eta^d \quad and \quad (0, \delta_1^{d+1})\kappa^{d-1} = \begin{pmatrix} 0 & \Pi_{d+1}^{-1}D^{d+1}\Pi_d & 0 \\ 0 & 0 & D^{d+1} \end{pmatrix} \kappa^{d-1},$$

the graded matrix η^d must have the underlying matrix

$$\mathsf{u}(\eta^d) = -(0, 0, D^{d+1}) \ \mathsf{u}(\kappa^{d-1}) = (0, D^{d+1}\Pi_d, 0) \ \mathsf{u}(\kappa^{d-1}),$$

and the appropriate row and column grades to fit into the diagram. It follows from Theorem 4.1.7 that the thus defined graded matrix is valid.

(iv) To obtain a minimal free resolution of $R_z H^d(K_*)$, apply Algorithm 4.

4.2.2 Relative cohomology

Analogously, we construct a free resolution of $R_z H^{d+1}(K, K_*)$. If $H^{\bullet}(K) = 0$, then we obtain that $H^d(K_*) \cong H^{d+1}(K, K_*)$ for all d, which motivates the study of $R_z H^{d+1}(K, K_*)$, rather than $R_z H^d(K, K_*)$. We have

$$R_z C^d(K, K_*) = \left(\frac{R_z \Delta C_d(K)}{R_z C_d(K_*)}\right)^* = \ker(\rho \colon R_z \Delta C^d(K) \to R_z C^d(K_*)),$$

Figure 4.2: The relative cochain complex $C^{\bullet}(K, K_*)$ is direct sum of the modules $(\Delta k/F(w))^*$, where Δk is the constant module with $(\Delta k)_w = k$ for all w. As before, it is convenient to draw modules with downward pointing coordinate axes.



where ρ is induced by the restriction map $\rho \colon \gamma \mapsto \gamma|_{K_*}$. The module

$$R_z \Delta C^d(K) = \bigoplus_{\sigma \in K} F(\bar{g}(\sigma)) = C_0^d$$

is free, so the restriction ρ lifts to a morphism

$$\begin{array}{ccc} \Delta C^d(K) \colon & 0 \longrightarrow 0 \longrightarrow 0 \longrightarrow R_z \Delta C^d(K) \\ & & \downarrow & \downarrow & \parallel \\ & C^d_{\bullet} \colon & 0 \longrightarrow C^d_2 \xrightarrow[c^d_2]{} C^d_1 \xrightarrow[c^d_1]{} C^d_0 \end{array}$$

of free resolutions, where C^d_{\bullet} is the free resolution from (4.30). Lemma 4.1.1 shows that

$$0 \to C_2^d \xrightarrow{c_2^a} C_1^d \to R_z C^d(K, K_*)$$
(4.32)

is a free resolution of $R_z C^{\bullet}(K, K_*)$, and

is a lift of the coboundary morphism $\delta^{d+1} \colon R_z C^d(K, K_*) \to R_z C^{d+1}(K, K_*)$. The morphisms δ_1^{d+1} and δ_2^{d+1} are the same as in as in (4.31).

Proposition 4.2.7 (Relative cohomology resolution). For each d, there exists morphisms h_1^{d+1} and h_2^{d+1} that make the diagram

commute and form a free resolution of the module $R_z H^{d+1}(K, K_*)$.

Remark. We point to the fact that the morphisms c_i^d and δ_i^d are the same as in the previous section, while the other functions h_i^d and κ^d are not the same as in the previous section.

Proof of Proposition 4.2.7. Applying Theorem 4.1.7 to (4.33) shows that there exist morphisms \tilde{h}_2 , \tilde{h}_1 that make the diagram

commute and form a free resolution of $R_z H^{d+1}(K, K_*)$. To show the statement of the proposition, consider the commutative diagram



One checks that the lower square commutes, so ψ induces the indicated morphism ψ' . Then ψ' is an isomorphism. To see this, we notice that ψ' is the morphism

$$\begin{aligned} &\ker(\delta_{1}^{d+2}, c_{2}^{d+2}) \\ &= \ker\begin{pmatrix} \Pi_{d+2}^{-1} D^{d+2} \Pi_{d+1} & 0 & -\Pi_{d+2}^{-1} \\ 0 & D^{d+2} & E_{d+2} \end{pmatrix} \\ &= \{(\alpha, \beta, \gamma) \in C_{\Gamma_{1}}^{d+1} \oplus C_{1_{j}}^{d+1} \oplus C_{2}^{d+2} \mid -\gamma = D^{d+2} \Pi_{d+1} \alpha = D^{d+2} \beta \} \\ &\stackrel{\psi'}{\to} \{(a, b) \in C_{\Gamma_{1}}^{d+1} \oplus C_{1_{j}}^{d+1} \mid D^{d+2} \Pi_{d+1} \alpha = D^{d+2} \beta \} \\ &= \ker(D^{d+2} \Pi_{d+1}, D^{d+2}) \\ &= \ker c_{1}^{d+2} d_{1}^{d+2}. \end{aligned}$$

An inverse to ψ' is given by $\psi': \begin{pmatrix} \alpha \\ \beta \end{pmatrix} \mapsto \begin{pmatrix} \alpha \\ \beta \\ D^{d+2}\Pi_{d+1}\alpha \end{pmatrix} = \begin{pmatrix} \alpha \\ \beta \\ D^{d+2}\beta \end{pmatrix}$. The statement follows from applying this to (4.35).

Using the above steps, we obtain the following strategy to compute a minimal free resolution of relative cohomology:



Figure 4.4: One-critical filtration K_* of a 2-sphere by two 0-cells x, y, three 1-cells a, b, c with $\partial a = \partial b = \partial c = x + y$ and three 2-cells f, g, h with $\partial f = b + c$, $\partial g = a + c$ and $\partial h = a + b$.

4.2.8. A pair of graded matrices h_2^{d+1} , h_1^{d+1} representing the free resolution (4.34) of $R_z H^{d+1}(K, K_*)$ can be computed in the following steps:

- (i) Compute a graded matrix κ^{d+1} representing a basis of $\ker(c_1^{d+2}\delta_1^{d+2})$ and a graded matrix κ^d representing a basis of $\ker c_1^{d+1}\delta_1^{d+1}$, using Algorithm 6.
- (ii) Compute h_1^{d+1} by solving the linear system $\kappa^{d+1}h_1^d = (d_1^{d+1}, -c_2^{d+1})$, using Algorithm 7.
- (iii) Compute $h_2^{d+1} = {\kappa^d \choose \eta^{d+1}}$, where $c_2^{d+1}\eta^{d+1} = \delta_1^{d+1}\phi^{d+1}$. Because

$$c_{2}^{d+1}\eta^{d+1} = \begin{pmatrix} \Pi_{d+1}^{-1} \\ -E_{d+1} \end{pmatrix} \eta^{d+1} \quad and \quad \delta_{1}^{d+1}\phi^{d} = \begin{pmatrix} \Pi_{d+1}^{-1}D^{d+1}\Pi_{d} & 0 \\ 0 & D^{d+1} \end{pmatrix} \phi^{d},$$

the entries of η^{d+1} satisfy

$$\mathsf{u}(\eta^{d+1}) = -(0, D^{d+1}) \, \mathsf{u}(\kappa^d) = (D^{d+1} \Pi_d, 0) \, \mathsf{u}(\kappa^d).$$

(iv) The obtained resolution can be converted to a minimal free resolution of $R_z H^{d+1}(K, K_*)$ using Algorithm 3.

4.3 Example

In this section, we compute the relative and absolute two-parameter persistent (non-reduced) cohomology of the one-critically \mathbb{Z}^2 -filtered cell complex K_* shown in Figure 4.4. We compute (co)homology with coefficients in \mathbb{F}_2 . The complex K_* has the (absolute) simplicial chain complex



which is a chain complex of free modules. A homogeneous basis is given by the simplices $\sigma \in K_*$. We use the symbol \cdot to denote the grades of the basis elements of C_{\bullet} , which are precisely the grades of the simplices. Analogously, the relative chain complex $C_{\bullet}(K, K_*) = \Delta C_{\bullet}(K)/C(K_*)$

is the chain complex



It is a chain complex of non-finitely generated modules.

4.3.1 Relative cohomology

We start with the easier case of computing a free resolution of $C^{\bullet}(K, K_*)$. The relative cochain complex $C^{\bullet}(K, K_*)$ has the graded components $C^{\bullet}(K, K_*)_z = (C_{\bullet}(K, K_*)_{-z})^*$. Drawing the coordinate axes pointing downwards, we get



We choose a fixed $z \in \mathbb{Z}^2$ with $z \ge g(\sigma)$ for all $\sigma \in K_*$, and compute the persistence module $R_z H^{\bullet}(K, K_*)$. This yields the cochain complex

$$R_{z}C^{\bullet}(K,K_{*}): \qquad 0 \longrightarrow \underbrace{\left(\begin{smallmatrix} X, & Y \\ \downarrow & X, & Y \\ R_{z}C^{0}(K,K_{*}) \end{smallmatrix}\right)}_{R_{z}C^{0}(K,K_{*})} \xrightarrow{f_{a}} \underbrace{\left(\begin{smallmatrix} J \\ J \\ J \\ J \\ R_{z}C^{1}(K,K_{*}) \end{smallmatrix}\right)}_{R_{z}C^{1}(K,K_{*})} \xrightarrow{f_{a}} \underbrace{\left(\begin{smallmatrix} J \\ J \\ J \\ J \\ R_{z}C^{2}(K,K_{*}) \end{smallmatrix}\right)}_{R_{z}C^{2}(K,K_{*})} \xrightarrow{f_{a}} \underbrace{\left(\begin{smallmatrix} J \\ J \\ J \\ J \\ R_{z}C^{2}(K,K_{*}) \end{smallmatrix}\right)}_{R_{z}C^{2}(K,K_{*})} \xrightarrow{f_{a}} \underbrace{\left(\begin{smallmatrix} J \\ J \\ J \\ J \\ R_{z}C^{2}(K,K_{*}) \end{smallmatrix}\right)}_{R_{z}C^{2}(K,K_{*})} \xrightarrow{f_{a}} \underbrace{\left(\begin{smallmatrix} J \\ J \\ J \\ J \\ R_{z}C^{2}(K,K_{*}) \end{smallmatrix}\right)}_{R_{z}C^{2}(K,K_{*})} \xrightarrow{f_{a}} \underbrace{\left(\begin{smallmatrix} J \\ J \\ J \\ J \\ R_{z}C^{2}(K,K_{*}) \end{smallmatrix}\right)}_{R_{z}C^{2}(K,K_{*})} \xrightarrow{f_{a}} \underbrace{\left(\begin{smallmatrix} J \\ J \\ J \\ J \\ R_{z}C^{2}(K,K_{*}) \end{smallmatrix}\right)}_{R_{z}C^{2}(K,K_{*})} \xrightarrow{f_{a}} \underbrace{\left(\begin{smallmatrix} J \\ J \\ J \\ J \\ R_{z}C^{2}(K,K_{*}) \end{smallmatrix}\right)}_{R_{z}C^{2}(K,K_{*})} \xrightarrow{f_{a}} \underbrace{\left(\begin{smallmatrix} J \\ J \\ J \\ J \\ R_{z}C^{2}(K,K_{*}) \end{smallmatrix}\right)}_{R_{z}C^{2}(K,K_{*})} \xrightarrow{f_{a}} \underbrace{\left(\begin{smallmatrix} J \\ J \\ J \\ J \\ R_{z}C^{2}(K,K_{*}) \end{smallmatrix}\right)}_{R_{z}C^{2}(K,K_{*})} \xrightarrow{f_{a}} \underbrace{\left(\begin{smallmatrix} J \\ J \\ J \\ J \\ R_{z}C^{2}(K,K_{*}) \end{smallmatrix}\right)}_{R_{z}C^{2}(K,K_{*})} \xrightarrow{f_{a}} \underbrace{\left(\begin{smallmatrix} J \\ J \\ J \\ J \\ R_{z}C^{2}(K,K_{*}) \end{smallmatrix}\right)}_{R_{z}C^{2}(K,K_{*})} \xrightarrow{f_{z}} \underbrace{\left(\begin{smallmatrix} J \\ J \\ J \\ J \\ R_{z}C^{2}(K,K_{*}) \end{smallmatrix}\right)}_{R_{z}C^{2}(K,K_{*})} \xrightarrow{f_{z}} \underbrace{\left(\begin{smallmatrix} J \\ J \\ J \\ J \\ R_{z}C^{2}(K,K_{*}) \end{smallmatrix}\right)}_{R_{z}C^{2}(K,K_{*})} \xrightarrow{f_{z}} \underbrace{\left(\begin{smallmatrix} J \\ J \\ J \\ J \\ R_{z}C^{2}(K,K_{*}) \end{smallmatrix}\right)}_{R_{z}C^{2}(K,K_{*})} \xrightarrow{f_{z}} \underbrace{\left(\begin{smallmatrix} J \\ J \\ J \\ J \\ R_{z}C^{2}(K,K_{*}) \end{smallmatrix}\right)}_{R_{z}C^{2}(K,K_{*})} \xrightarrow{f_{z}} \underbrace{\left(\begin{smallmatrix} J \\ J \\ J \\ J \\ R_{z}C^{2}(K,K_{*}) \end{smallmatrix}\right)}_{R_{z}C^{2}(K,K_{*})} \xrightarrow{f_{z}} \underbrace{\left(\begin{smallmatrix} J \\ J \\ J \\ R_{z}C^{2}(K,K_{*}) \end{smallmatrix}\right)}_{R_{z}C^{2}(K,K_{*})} \xrightarrow{f_{z}} \underbrace{\left(\begin{smallmatrix} J \\ J \\ J \\ R_{z}C^{2}(K,K_{*}) \end{smallmatrix}\right)}_{R_{z}C^{2}(K,K_{*})} \xrightarrow{f_{z}} \underbrace{\left(\begin{smallmatrix} J \\ J \\ R_{z}C^{2}(K,K_{*}) \right)}_{R_{z}C^{2}(K,K_{*})} \xrightarrow{f_{z}} \underbrace{\left(\begin{smallmatrix} J \\ J \\ R_{z}C^{2}(K,K_{*}) \right)}_{R_{z}C^{2}(K,K_{*})} \xrightarrow{f_{z}} \underbrace{\left(\begin{smallmatrix} J \\ R_{z}C^{2}(K,K_{*}) \right\right)}_{R_{z}C^{2}(K,K_{*})} \xrightarrow{f_{z}} \underbrace{\left(\begin{smallmatrix} J \\ R_{z}C^{2}(K,K_{*}) \right\right)}_{R_{z}C^{2$$

of finitely generated modules. Each simplex $\sigma \in K_*$ corresponds to two generators $\overline{\sigma}$ and σ_{\perp} of $R_z C^{\bullet}(K, K_*)$ of the grades $\overline{g}(\sigma)$ and $g_{\perp}(\sigma)$, indicated by the symbol \bullet , and a relation $\sigma : \overline{\sigma} \equiv \sigma_{\perp}$ of grade $\underline{g}(\sigma)$, indicated by the symbol \bullet . We decorate matrix row and columns by the basis elements of the codomain and domain they correspond to. The matrices

$$D^{1} = {}^{c}_{b} \begin{pmatrix} x & y \\ 1 & 1 \\ 1 & 1 \end{pmatrix}, \qquad D^{2} = {}^{f}_{g} \begin{pmatrix} c & b & a \\ 1 & 1 \\ 1 & 1 \end{pmatrix}$$

represent the coboundary morphisms of the cochain complex $C^{\bullet}(K)$. Let

$$\Pi_{0} = {}^{x}_{y} {\begin{pmatrix} {}^{x}_{1} {}^{y} \\ {}^{1}_{1} {}^{1} \end{pmatrix}}, \qquad \Pi_{1} = {}^{a}_{c} {\begin{pmatrix} {}^{c}_{1} {}^{b}_{1} \\ {}^{1}_{1} {}^{1} \end{pmatrix}}, \qquad \Pi_{2} = {}^{h}_{f} {\begin{pmatrix} {}^{f}_{2} {}^{h}_{1} \\ {}^{1}_{1} {}^{1} \end{pmatrix}},$$

be the permutation matrices that represent the re-ordering of the simplices from reverse lexicographic ordering (columns) to reverse colexicographic ordering (rows).

0-Cohomology The module $R_z H^0(K, K_*)$ is generated by the kernel of the map

$$c_{1}^{d+2}\delta_{1}^{d+2} = C_{2}^{1} (\Pi_{1}^{-1}D^{1} \mid D^{1}) = \vec{b}_{c}^{\vec{a}} (\prod_{1}^{\vec{x}} \vec{y} x_{1}, y_{1}) = C_{1}^{0} C_{1}^{0} (\prod_{1}^{\vec{x}} \prod_{1}^{\vec{a}} \prod_{1}^{\vec{a}} \prod_{1}^{\vec{a}} \prod_{1}^{\vec{a}} \prod_{1}^{\vec{a}} \prod_{1}^{\vec{a}} C_{1}^{0}) = C_{1}^{0} C_{1}^{0} C_{1}^{0} (\prod_{1}^{\vec{a}} \prod_{1}^{\vec{a}} \prod_{1}^$$

The columns of this matrix are in colexicographic order. Applying Algorithm 6 to this matrix, we obtain a homogeneous basis

$$\kappa^{0} = \frac{\int_{y_{j}}^{y_{j}} \left(\frac{1}{1} + \frac{1}{1} + \frac{1}{1}\right)}{\int_{y_{j}}^{y_{j}} \left(\frac{1}{1} + \frac{1}{1} + \frac{1}{1}\right)}$$
(4.36)

of ker $c_1^{d+2}\delta_1^{d+2}$. The grade of each column is the join of the row grades of the non-zero entries in that column; that is

$$g(\gamma_{1}) = g(\underline{x}_{\perp}) \lor g(\underline{y}_{\perp}) = g(\underline{x}_{\perp}) = g(\underline{y}_{\perp}),$$

$$g(\gamma_{2}) = g(\overline{x}) \lor g(\overline{y}) = g(\overline{x}) = g(\overline{y}),$$

$$g(\gamma_{3}) = g(\overline{x}) \lor g(\underline{x}_{\perp}) = g(\underline{x}) = g(\underline{y}).$$
(4.37)

To get the relations of $R_z H^0(K, K_*)$, we have to compute a matrix representing the morphism h_1^0 in (4.34). The matrix (which we also denote by h_1^0) is the unique (because κ^0 is injective) solution

$$h_1^0 = \stackrel{\gamma_1}{\underset{\gamma_3}{\gamma_1}} \begin{pmatrix} \overset{\downarrow^x}{1} & \overset{\downarrow^y}{1} \\ \begin{pmatrix} 1 \\ 1 \end{pmatrix} & \text{of the linear system} & \kappa^0 h_1^0 \stackrel{!}{=} \stackrel{\downarrow^x}{\underset{y_j}{\gamma_1}} \begin{pmatrix} \overset{\downarrow^x}{1} & \overset{\downarrow^y}{1} \\ \frac{1}{1} \\ 1 \end{pmatrix} = c_2^0.$$

Then h_1^0 is a free presentation of $R_z H^0(K, K_*)$. Since it is it injective, it also represents a free resolution (of length one) of $R_z H^0(K, K_*)$. To obtain a minimal free resolution, one has to eliminate all homological 1-balls from h_1^0 . Recall that Algorithm 3 achieves this by identifying a maximal pairwise disjoint set of local pairs of the matrix h_1^0 , and eliminates all non-zero entries in the local rows via left-to-right column additions, and deletes the local rows and columns. Applying Algorithm 3 to h_1^0 identifies the local pair corresponding to the circled entry, eliminates all non-zero entries in the same row by column additions, and deletes the corresponding row and column. We obtain the minimal (in the sense of Definition 2.3.21) graded matrix

$$\tilde{h}_{1}^{0} = \begin{array}{c} \gamma_{1} = x_{j} + y_{j} \\ \gamma_{2} = \bar{x} + \bar{y} \end{array} \begin{pmatrix} r_{1} \\ \vdots \\ 1 \end{pmatrix},$$

which represents a minimal free presentation (and resolution)

$$0 \to \langle r_1 \rangle \xrightarrow{\tilde{h}_1^0} \langle \gamma_1, \gamma_2 \rangle \to R_z H^0(K, K_*)$$
(4.38)

of $R_z H^0(K, K_*)$; see Figure 4.5a.

1-Cohomology Analogously, the module $R_z H^1(K, K_*)$ is generated by the basis

The matrix h_1^1 that represents the relations of $R_z H^1(K, K_*)$ is the unique solution



Figure 4.5: The free resolutions (4.38), (4.42) and (4.43) of $R_z H^{\bullet}(K, K_*)$ for the filtered simplicial complex K_* from Figure 4.4. The symbols \bullet , \bullet and \star denote the grades of the 0-, 1- and 2-syzygies of the modules; the symbols \circ indicate the grades of the relevant generators σ and σ_{\perp} of $C^{\bullet}(K, K_*)$. The module $R_z H^2(K, K_*)$ is indecomposable, because it contains a indecomposable quiver representation.

A matrix h_2^1 that represents a basis of $H^1(K, K_*)_2$ can be obtained by forming the matrix product

$$h_{2}^{1} \coloneqq C_{1}^{0} \begin{pmatrix} C_{1}^{0} & C_{1}^{0} \\ C_{1}^{0} & C_{1}^{0} \\ C_{2}^{1} \end{pmatrix} \cdot \kappa^{0} = \begin{array}{c} \overset{\gamma_{1}}{x_{1} + y_{1}} & \overset{\gamma_{2}}{x_{1} + y_{1}} \\ \overset{\gamma_{2}}{x_{1} + y_{1}} & \overset{\gamma_{3}}{x_{1} + y_{1}} \\ \overset{\gamma_{3}}{x_{1} + y_{1$$

where κ^0 is the matrix representing the basis of $R_z H^1(K, K_*)_2 \cong R_z H^0(K, K_*)_0$ from (4.36). Here, we use the symbols s_i to distinguish the columns of κ^0 from those of κ^1 . Note that the left matrix in (4.41) (specifically, its bottom right block) is not valid in general. Nevertheless, it follows from Theorem 4.1.6 that the resulting matrix h_2^1 is valid. Then h_1^1 and h_2^1 represent a free resolution of $R_z H^1(K, K_*)$.

To obtain a minimal free resolution, we apply Algorithm 3 first to h_2^1 . Note that the rows of h_2^1 (that is, the columns of h_1^1) are not lexicographically or colexicographically ordered. Nevertheless, the block shape of the right side in (4.40) ensures that the rows of h_1^1 are in non-descending order, as required by Algorithm 3. Therefore, reordering the rows of h_1^1 and the columns of h_2^1 is not necessary. Algorithm 3 identifies local pairs of h_2^1 corresponding to the circled entries in (4.41). We get triple $(\tilde{h}_2^1, r, c) = \text{Minimize}(h_2^1)$, where \tilde{h}_2^1 is a minimal graded matrix, and r and c contain the non-local row and column indices of h_2^1 ; that is, the indices that do not belong to a local pair of h_2^1 . We obtain the matrices

$$\tilde{h}_{2}^{1} = \overset{\gamma_{1}}{\underset{\substack{a \\ b \\ c}}{\overset{\gamma_{1}}{1}}} \begin{pmatrix} \gamma_{1} \\ 1 \\ 1 \\ 1 \\ 1 \end{pmatrix}, \qquad \tilde{h}_{1}^{1} \coloneqq ([h_{1}^{1}]_{ij})_{j \in r} = \overset{\gamma_{1}}{\overset{\gamma_{2}}{\gamma_{3}}} \begin{pmatrix} \gamma_{1} \\ 1 \\ 1 \\ 1 \\ 1 \end{pmatrix} \overset{a \ b \ c}{\overset{c}{1}} \begin{pmatrix} \gamma_{1} \\ 1 \\ 1 \\ 1 \\ 1 \end{pmatrix}.$$

We apply Algorithm 3 to \tilde{h}_1^1 , which identifies the circled local entries. With $(\tilde{h}_1'^1, r', c') = \text{Minimize}(\tilde{h}_1^1)$, we get that the matrices

$$\tilde{h}_{2}^{\prime 1} \coloneqq ([\tilde{h}_{2}^{1}]_{ij})_{i \in c'} = \bigcup_{\substack{l \\ l \\ c}}^{a} \begin{pmatrix} \overbrace{r_{1}}^{s_{3}} \\ 1 \\ 1 \\ c \end{pmatrix}, \quad \text{and} \quad \tilde{h}_{1}^{\prime 1} = \gamma_{2} = a_{1} + b_{1} + \lceil c \\ \gamma_{3} = a_{1} + \lceil b + \lceil c \\ 1 \\ 1 \\ 1 \end{pmatrix}$$

represent the minimal free resolution

$$0 \to \langle s_3 \rangle \xrightarrow{\tilde{h}_2'^1} \langle r_2, r_3, r_4 \rangle \xrightarrow{\tilde{h}_1'^1} \langle \gamma_2, \gamma_3 \rangle \to R_z H^1(K, K_*)$$
(4.42)

of $R_z H^1(K, K_*)$, see Figure 4.5b.

2-Cohomology Because K_* has no 3-simplicies, $R_z H^2(K, K_*)$ is generated by the generators [f, [g, [h, f], g] and h_{\perp} of C_1^2 . The relations of $R_z H^2(K, K_*)$ are represented by the matrix

$$h_{1}^{2} = \stackrel{C_{\Gamma_{1}}^{2}}{C_{1_{j}}^{2}} \begin{pmatrix} \frac{C_{\Gamma_{1}}^{1} & C_{1_{j}}^{1} & C_{2}^{2}}{|D^{2}|E_{2}} \end{pmatrix} = \stackrel{f}{\underset{D}{\overset{f}{\overset{J}}}} \begin{pmatrix} \frac{1}{f_{j}} & \frac{1}{f_{j}} \\ \frac{1}{f_{j}} \\ \frac{1}{f_{j}} & \frac{1}{f_{j}} \\ \frac{1}{f$$

Analogously to (4.41), we obtain the 2-syzygies (viz. the kernel of h_1^2) by forming the product

$$h_{2}^{2} \coloneqq C_{1}^{1} \begin{pmatrix} C_{1}^{1} & C_{1}^{1} \\ E_{1} \\ C_{2}^{2} \end{pmatrix} \cdot \kappa^{1} = \frac{a_{j}}{b_{j}} \begin{pmatrix} s_{1} & s_{2} & s_{3} \\ a_{j}+b_{j}+c_{j} & a_{j}+b_{j}+c_{j} \\ a_{j$$

with the matrix κ_0^1 from (4.39). Applying Algorithm 3 to h_2^2 identifies the circled local entries. Analogously to the above, we obtain

Applying Algorithm 3 to \tilde{h}_1^2 identifies the circled local entries. We get that the matrices

represent the minimal free resolution

$$0 \to \langle s_2, s_3 \rangle \xrightarrow{\tilde{h}_2^{\prime 2}} \langle r_1, r_2, r_3 \rangle \xrightarrow{\tilde{h}_1^{\prime 2}} \langle \bar{f}, \underline{h} \rangle \to R_z H^2(K, K_*)$$

$$(4.43)$$

of $R_z H^2(K, K_*)$, see Figure 4.5c.

4.3.2 Absolute cohomology

Analogously, we compute a minimal free resolutions of $R_z H^{\bullet}(K_*)$, using the restricted absolute cochain complex



Each *d*-simplex $\sigma \in K_*$ constributes one generator σ^{\neg} marked by •, two relations $[\sigma, \sigma]: \sigma^{\neg} \equiv 0$ marked by •, and the 2-syzygy $[\sigma: [\sigma \equiv \sigma]]$ marked by ×.

0-Homology From Proposition 4.2.5, we obtain that a generating system of $R_z H^0(K_*)$ is represented by the basis

$$\kappa^{0} = \begin{bmatrix} \vec{x}^{1} & \vec{\gamma}^{2} & \vec{\gamma}^{3} & \vec{\gamma}^{4} & \vec{\gamma}^{5} \\ \vec{y}^{1} & 1 & 1 & 1 \\ \hline 1 & 1 & 1 & 1 \\ \hline \vdots & \vdots \\ \vec{y}^{1} \\ \vec{y}^{1} \\ \vec{y}^{1} \\ \vec{y}^{1} \\ \vec{y}^{1} \\ \vec{z}^{1} \\ \vec{z$$

The relations of $R_z H^0(K_*)$ are represented by the unique solution

The relations of $R_z H^0(K_*)$ (viz. the kernel of h_1^1) are represented by the matrix product

$$h_{1}^{0} = \overset{C_{1}^{0}}{\underset{C_{2}}{C_{1}}} \begin{pmatrix} \overset{C_{1}^{0}}{|} & C_{1}^{1} \\ \overset{E_{0}}{|} & \overset{\vdots}{|} \\ \overset{E_{0}}{|} \\ \overset{\vdots}{|} \\ \overset{z_{1}}{|} \\ \overset{z_{2}}{|} \\ \overset{z_{1}}{|} \\ \overset{z_{2}}{|} \\ \overset{z_{2}}{|} \\ \overset{z_{1}}{|} \\ \overset{z_{2}}{|} & \overset{z_{2}}{|} & \overset{z_{2}}{|} & \overset{z_{2}}{|} & \overset{z_{2}}{|} & \overset{z_$$

The matrix c_2^0 has no local entries. Applying Algorithm 3 identifies the local pairs corresponding to the circled entries in (4.45) We obtain that the matrices

$$\tilde{h}_{2}^{0} = \begin{bmatrix} r_{1} & y \\ r_{2} \\ \vdots \\ c \\ c \\ c \end{bmatrix} \begin{pmatrix} x & y \\ 1 & 1 \\ 1 & 1 \\ \vdots \\ c \\ c \end{bmatrix}, \qquad \tilde{h}_{1}^{0} = \begin{bmatrix} \gamma_{1} = \vec{x}^{1} + \vec{y} \\ \gamma_{3} = \vec{x}^{1} + \vec{a}_{1} + \vec{b}_{2} + \vec{c} \\ \gamma_{4} = \vec{x}^{1} + \vec{a}_{2} + \vec{b}_{3} + \vec{c} \\ \gamma_{4} = \vec{x}^{1} + \vec{a}_{2} + \vec{b}_{3} + \vec{c} \\ 1 & 1 & 1 \\$$

represent the minimal free resolution

$$0 \to \langle \underline{x}, \underline{y} \rangle \xrightarrow{\tilde{h}_{2}^{0}} \langle \overline{y}, \underline{y}, \underline{a}, \underline{b}, \underline{c} \rangle \xrightarrow{\tilde{h}_{1}^{0}} \langle \gamma_{1}, \gamma_{3}, \gamma_{4} \rangle \to R_{z} H^{0}(K_{*})$$

of $R_z H^0(K_*)$. We may perform invertible column operations on \tilde{h}_2^0 without changing the result. Thus, replacing \tilde{h}_2^0 by

$$\tilde{h}_{2}^{\prime 0}\coloneqq \overset{r_{1}}{\underset{\substack{l \\ b \\ c}}{r_{1}}} \left(\overset{\overset{x}{\underbrace{1}} \overset{x+\underline{y}}{\underline{1}}}{\underbrace{1}} \right),$$

we see that

$$aR_z H^0(K_*) \cong \frac{\langle \gamma_1 \rangle}{\langle r_1, r_2 \rangle} \oplus \frac{\langle \gamma_3, \gamma_4 \rangle}{\langle \underline{a}, \underline{b}, \underline{c} \rangle}.$$
(4.46)

see Figure 4.6a.

1-Cohomology Analogously, we obtain the generators or $R_z H^1(K_*)$ by computing the basis

The relations of $R_z H^1(K_*)$ are represented by the unique solution

and the 2-syzygies of $R_z H^1(K_*)$ are represented by the matrix product

where κ^0 is the kernel basis from (4.45). Applying Algorithm 3 yields

and finally

$$\tilde{h}_{2\ r_{3}}^{\prime 1} \begin{pmatrix} s_{3} & s_{4} \\ 1 & 1 \\ r_{2} & r_{3} \end{pmatrix}, \qquad \tilde{h}_{1}^{\prime 1} = \gamma_{4} = \vec{b} + \vec{f} + \vec{h}_{1} \begin{pmatrix} r_{1} \\ ..., h + \vec{b} \\ ..., h + \vec{b} \end{pmatrix}, \qquad \tilde{h}_{1}^{\prime 1} = \gamma_{4} = \vec{b} + \vec{f} + \vec{h}_{1} \begin{pmatrix} r_{1} \\ ..., h + \vec{b} \\ ..., h + \vec{b} \end{pmatrix},$$

which represent the minimal free resolution

$$0 \to \langle s_3, s_4 \rangle \xrightarrow{\tilde{h}_2^{\prime 1}} \langle r_1, r_2, r_3 \rangle \xrightarrow{\tilde{h}_1^{\prime 1}} \langle \gamma_4 \rangle \to R_z H^1(K_*)$$
(4.47)

of $R_z H^1(K_*)$; see Figure 4.6b.

2-Cohomology Lastly, there are no 3-simplices, which implies $C^3_{\bullet} = 0$. According to Proposition 4.2.5, $R_z H^2(K_*)$ is generated by $C^2_0 = \langle \bar{f}^{\uparrow}, \bar{g}^{\uparrow}, \bar{h}^{\uparrow} \rangle$. Without calculation, we obtain that



Figure 4.6: Truncated absolute cohomology $R_z H^{\bullet}(K_*)$ of the filtered simplicial complex K_* from Figure 4.4. The module $R_z H^0(K_*)$ is decomposable into a direct sum of its red and its blue submodule.

Applying Algorithm 3 to h_2^2 yields

$$\tilde{h}_{2}^{2} = \frac{\bar{a}^{\vec{i}}}{\int_{f}} \begin{pmatrix} s_{4} \\ 0 \\ \frac{1}{1} \\ \frac{1}{1} \end{pmatrix}, \qquad \tilde{h}_{1}^{2} = \frac{\bar{f}^{\vec{i}}}{\bar{g}^{\vec{i}}} \begin{pmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{pmatrix}.$$

When we apply Algorithm 3 to \tilde{h}_1^2 , we are in a situation that we have not met before: namely, one of the local pairs of \tilde{h}_1^2 can only be identified after a column addition caused by the removal of the non-zero entries in the local rows. Specifically, Algorithm 3 transforms \tilde{h}_1^2 as follows:

$$\tilde{h}_{1}^{2} = \stackrel{\vec{a}}{\overset{\vec{a}}{|}} \left(\begin{array}{c} \vec{a} & \vec{b} & \vec{f} & h_{\text{j}} \\ 1 & 1 & 1 \\ \vec{h} & \vec{h} \\ \vec{h}$$

where the first arrow happens during the first phase and the second arrow during the second phase of Algorithm 3, with the circled local pairs. Removal of the rows and columns with local pairs gives

$$\tilde{h}_{2}^{\prime 2} = r_{2}^{r_{1}} \begin{pmatrix} s_{4} \\ 1 \end{pmatrix}, \qquad \tilde{h}_{1}^{\prime 2} = \bar{f}^{2} \begin{pmatrix} r_{1} \\ \bar{f}^{2} \end{pmatrix}, \qquad \tilde{h}_{2} + \bar{b}^{2}$$

Then ${\tilde h}_2^{\prime 2}$ and ${\tilde h}_1^{\prime 2}$ represent the minimal free resolution

$$0 \to \langle s_4 \rangle \xrightarrow{\tilde{h}_2'^2} \langle r_1, r_2 \rangle \xrightarrow{\tilde{h}_1'^2} \langle \bar{f} \rangle \to R_z H^2(K_*)$$
(4.48)

of $R_z H^2(K_*)$; see Figure 4.6c.

4.3.2.1 The long exact sequence

Lastly, we give an explicit description of the the long exact sequence

$$0 \longrightarrow R_{z}H^{0}(K, K_{*}) \xrightarrow{p^{0}} R_{z}\Delta H^{0}(K) \xrightarrow{i^{0}} R_{z}H^{0}(K_{*}) \searrow_{\delta}$$

$$(4.49)$$

$$(4.49)$$

$$(4.49)$$

$$(4.49)$$

for the above filtration K_* . Here, p^* and i^* are the morphisms induced by the morphisms $i: K_* \hookrightarrow K$ and $p: K \to K/K_*$. Recall from Propositions 4.2.5 and 4.2.7 that in the free resolutions $R_z H^d(K, K_*)_{\bullet}$ and $R_z H^d(K_*)_{\bullet}$, we had

$$R_{z}H^{d}(K, K_{*})_{0} = \ker(C^{d}(K_{*})_{0} \longrightarrow C^{d+1}(K_{*})_{0}),$$

$$R_{z}H^{d}(K_{*})_{0} = \ker(C^{d}(K_{*})_{0} \oplus C^{d+1}(K_{*})_{1} \longrightarrow C^{d+1}(K_{*})_{0}).$$

Consider the module $R_z \Delta H^d(K)$, which has the components $(R_z \Delta H^d(K))_w = H^d(K)$ if $w \ge -z$, and $(R_z \Delta H^d(K))_w = 0$ otherwise. This has a free resolution

$$0 \to \ker(C_0^{d-1} \to C_0^d) \to C_0^{d-1} \to \ker(C_0^d \to C_0^{d+1}) \to R_z \Delta H^d(K).$$
(4.50)

Note that $R_z \Delta H^d(K) = H^d(C_0^{\bullet})$. The long exact sequence (4.49) is induced by the commutative diagram

$$\cdots \to R_{z}H^{d}(K, K_{*}) \xrightarrow{i^{d}} R_{z}\Delta H^{d}(K) \xrightarrow{p^{d}} R_{z}H^{d}(K_{*}) \xrightarrow{\delta} R_{z}H^{d+1}(K, K_{*}) \to \cdots$$

$$\cdots \to \ker \delta_{0}^{d}c_{1}^{d} \longrightarrow \ker \delta_{0}^{d} \xrightarrow{\gamma} \ker \delta_{0}^{d} \xrightarrow{\gamma} \ker \delta_{0}^{d} \xrightarrow{\gamma} \ker \delta_{0}^{d+1} \longrightarrow \ker \delta_{0}^{d+1} \to \cdots$$

$$\cdots \to C_{1}^{d-1} \oplus C_{2}^{d} \xrightarrow{\left(c_{1}^{d-1}, 0\right)} C_{0}^{d-1} \xrightarrow{\gamma} \left(\begin{pmatrix} 1\\ 0\\ 0\\ 0 \end{pmatrix} \\ C_{0}^{d-1} \oplus C_{1}^{d} \oplus C_{2}^{d+1} \\ \vdots & \vdots \end{pmatrix} \xrightarrow{\gamma} \left(\begin{pmatrix} \delta_{0}^{d-1} c_{1}^{d} & 0\\ 0 & \delta_{1}^{d} c_{2}^{d+1} \end{pmatrix} \\ \vdots & \vdots \end{pmatrix} \xrightarrow{\gamma} \left(\begin{pmatrix} \delta_{0}^{d-1} c_{1}^{d} & 0\\ 0 & \delta_{1}^{d} c_{2}^{d+1} \end{pmatrix} \\ \vdots & \vdots \end{pmatrix} \xrightarrow{\gamma} \left(\begin{pmatrix} \delta_{0}^{d-1} c_{1}^{d} & 0\\ 0 & \delta_{1}^{d} c_{2}^{d+1} \end{pmatrix} \\ \vdots & \vdots \end{pmatrix} \xrightarrow{\gamma} \left(\begin{pmatrix} \delta_{0}^{d-1} c_{1}^{d} & 0\\ 0 & \delta_{1}^{d} c_{2}^{d+1} \end{pmatrix} \\ \vdots & \vdots \end{pmatrix} \xrightarrow{\gamma} \left(\begin{pmatrix} \delta_{0}^{d-1} c_{1}^{d} & 0\\ 0 & \delta_{1}^{d} c_{2}^{d+1} \end{pmatrix} \\ \vdots & \vdots \end{pmatrix} \xrightarrow{\gamma} \left(\begin{pmatrix} \delta_{0}^{d-1} c_{1}^{d} & 0\\ 0 & \delta_{1}^{d} c_{2}^{d+1} \end{pmatrix} \\ \vdots & \vdots \end{pmatrix} \xrightarrow{\gamma} \left(\begin{pmatrix} \delta_{0}^{d-1} c_{1}^{d} & 0\\ 0 & \delta_{1}^{d} c_{2}^{d+1} \end{pmatrix} \\ \vdots & \vdots \end{pmatrix} \xrightarrow{\gamma} \left(\begin{pmatrix} \delta_{0}^{d-1} c_{1}^{d} & 0\\ 0 & \delta_{1}^{d} c_{2}^{d+1} \end{pmatrix} \\ \vdots & \vdots \end{pmatrix} \right) \xrightarrow{\gamma} \left(\begin{pmatrix} \delta_{0}^{d-1} c_{1}^{d} & 0\\ 0 & \delta_{1}^{d} c_{2}^{d+1} \end{pmatrix} \\ \vdots & \vdots \end{pmatrix} \left(\begin{pmatrix} \delta_{0}^{d-1} c_{1}^{d+1} & 0\\ 0 & \delta_{0}^{d+1} c_{1}^{d+1} \end{pmatrix} \\ \vdots & \vdots \end{pmatrix} \right) \xrightarrow{\gamma} \left(\begin{pmatrix} \delta_{0}^{d-1} c_{1}^{d+1} & 0\\ 0 & \delta_{0}^{d+1} c_{1}^{d+1} \end{pmatrix} \\ \vdots & \vdots \end{pmatrix} \right) \xrightarrow{\gamma} \left(\begin{pmatrix} \delta_{0}^{d+1} c_{1}^{d+1} & 0\\ 0 & \delta_{0}^{d+1} & 0\\ \vdots & \delta_{0}^{d} & 0\\ \vdots & \vdots \end{pmatrix} \right) \xrightarrow{\gamma} \left(\begin{pmatrix} \delta_{0}^{d-1} c_{1}^{d+1} & 0\\ 0 & \delta_{0}^{d+1} & 0\\ \vdots & \delta_{0}^{d+1} & 0\\$$

In the diagram, the dashed arrows are the free presentations of $R_z H^{\bullet}(K, K_*)$, $R_z H^{\bullet}(K_*)$ and $R_z \Delta H^{\bullet}(K_*)$ from Propositions 4.2.5 and 4.2.7 and (4.50). The connecting homomorphism $\delta \colon H^d(K_*) \to H^{d+1}(K, K_*)$ is given by projecting a generator $(a, b) \in H^d(K_*)_0 \subseteq C_0^d \oplus C_1^{d+1}$ of $H^d(K_*)$ to its component $b \in C_1^{d+1}$. In practice, this amounts to dropping all summands of the form σ . Recall the generators of $H^{\bullet}(K_*)$ from (4.46) and (4.47) and $H^{\bullet}(K, K_*)$ from (4.42) and (4.43). We obtain that

$$\begin{split} \delta \colon H^0(K_*) &\longrightarrow H^1(K, K_*), & \delta \colon H^1(K_*) \longrightarrow H^2(K, K_*), \\ \gamma_1 &= \vec{x^{\intercal}} + \vec{y^{\intercal}} &\longmapsto 0, & \gamma_4 &= \vec{b^{\intercal}} + \vec{f} + \vec{h_{\tt l}} \longmapsto \vec{f} + \vec{h_{\tt l}}. \\ \gamma_3 &= \vec{x^{\intercal}} + \vec{a_{\tt l}} + \vec{b} + \vec{b} \longmapsto \vec{a_{\tt l}} + \vec{b} + \vec{b} = \gamma_2, \\ \gamma_4 &= \vec{x^{\intercal}} + \vec{a_{\tt l}} + \vec{b} + \vec{b} \longmapsto \vec{a_{\tt l}} + \vec{b} + \vec{b} = \gamma_3, \end{split}$$

Therefore, the long exact sequence (4.49) is the one shown in Figure 4.7.

Remark. This example also exhibits a particularity that cannot happen in one-parameter persistence. Namely, recall that unsplicing the long exact sequence (4.49) gives short exact sequences

$$0 \to \operatorname{coker} p^d \to H^d(K_*) \to \ker p^{d+1} \to 0,$$

$$0 \to \operatorname{coker} i^d \to H^{d+1}(K, K_*) \to \ker i^{d+1} \to 0$$

for all d. In one-parameter persistence, both short exact sequences split for all d (see Proposition 2.2.10). For the above example, Figure 4.7 shows that the short exact sequence

$$0 \to \operatorname{coker} i^1 \to H^2(K, K_*) \to \ker i^2 \to 0 \tag{4.51}$$

need not split for d = 2; see Figure 4.8. This implies that $H^{\bullet}(K_*)$ does not determine $H^{\bullet}(K, K_*)$ uniquely. An explicit example for this is given in Section 4.6.



Figure 4.7: Long exact sequence (4.49) in (truncated) cohomology of the pair (K, K_*) for the complex K_* from Figure 4.4. See Figures 4.5 and 4.6 for details on the modules. Note that the module at the bottom left is indecomposable. This gives rise to the non-split short exact sequence in Figure 4.8.



Figure 4.8: The non-split short exact sequence (4.51) obtained from Figure 4.7. The relations of the middle module $R_z H^2(K, K_*)$ are r_1, r_2, r_3 : $f + f_{\perp} \equiv 0$.

4.4 Relative cohomology revisited

In this section, we will have a closer look at the strategy from Section 4.2.2 to compute a minimal free resolution of $H^{d+1}(K, K_*)$. As before, K_* is a two-parameter filtered complex for which $K = \operatorname{colim}_{\mathbf{Z}^2} K_*$ is acyclic. Let D^{d+1} denote the (ungraded) coboundary matrix of K_* throughout this section. In principle, the free resolution of $H^{d+1}(K, K_*)$ from Proposition 4.2.7 can be computed using the procedures Ker() and Factorize() (Algorithms 6 and 7). This has the following shortcoming:

Recall that our goal is to compute a minimal free resolution of $H_d(K_*)$. As K is acyclic, we have $H^{d+1}(K, K_*) \cong H^d(K_*)$ for all d, so we may compute a minimal free resolution of $H^{d+1}(K, K_*)$, get a minimal free resolution of $H^d(K_*)$ at the same time, and use Corollary 3.2.11 to obtain a minimal free resolution of $H_d(K_*)$. However, computing $H^{d+1}(K, K_*)$ as described in 4.2.8 requires computing ker $c_1^{d+2}\delta_1^{d+2}$. Given that we are mainly interested in Vietoris–Rips complexes, considering the coboundary matrix δ^{d+2} is not feasible because of the sheer size of a matrix representing this morphism.

In one parameter persistence, this problem is solved by observing that it is not necessary in this case to compute $Z^{d+1}(K, K_*)$ from D^{d+2} because $Z^{d+1}(K, K_*)$ can already be computed from D^{d+1} , which may be much smaller for Vietoris–Rips complexes. In two parameter persistence, we already encountered a similar situation in Theorem C: there, we showed that under suitable conditions on K_* , the free resolution

$$0 \to Z^{d+1}(N^{\bullet}(K_*)) \longrightarrow N^{d+1}(K_*) \longrightarrow Z^{d+2}(N^{\bullet}(K_*)) \longrightarrow H^{d+2}(N^{\bullet}(K_*))$$

of $H^{d+2}(N^{\bullet}(K_*))$ can be computed from the coboundary matrix D^{d+1} alone, and not, as one may assume, by reducing the coboundary matrix D^{d+3} of N^{\bullet} .

We present a similar result for $H^{d+1}(K, K_*)$. Recall the functors colim, lim: $\operatorname{Vec}^{\mathbb{Z}^2} \to \operatorname{Vec}$ from Section 2.1.2.

Lemma 4.4.1. If $H^{\bullet}(K) = 0$, then $\operatorname{colim} \operatorname{im}(d_1^{d+1}, c_2^{d+1}) = \operatorname{colim} \ker(c_1^{d+2}\delta_1^{d+2})$ for all d.

Proof. Consider the morphisms h_2^{d+1} and h_1^{d+1} that represent the free resolution $H^{d+1}(K, K_*)$ of $H^{d+1}(K, K_*)$ from (4.34). Because \mathbb{Z}^2 is a direct system, colim is exact. Therefore, the morphisms colim h_2^{d+1} and colim h_1^{d+1} form a free resolution of the vector space colim $H^{d+1}(K, K_*)$. By assumption, colim $H^{d+1}(K, K_*) = \lim H^{d+1}(K_*) = H^{d+1}(K) = 0$. According to Theorem 2.3.18, the free resolution colim $H^{d+1}(K, K_*)_{\bullet}$ is trivial; in particular, it is exact. Therefore, colim h_1^{d+1} is surjective. From Proposition 4.2.7, we obtain a commutative diagram

Recall that (in any abelian category) a morphism f is surjective if and only if im $gf = \operatorname{im} f$ for every morphism g that can be composed with f. Since

$$\operatorname{colim}(d_1^{d+1}, c_2^{d+1}) = (\operatorname{colim} k^{d+1})(\operatorname{colim} h_1^{d+1})$$

the claim follows.

Theorem D. If $H_{\bullet}(K) = 0$, then the free resolution (4.34) of $H^{d+1}(K, K_*)$ can be computed solely from the coboundary matrix D^{d+1} of K_* .

Proof. According to Proposition 4.2.7, the matrix D^{d+1} is not needed to compute h_2^{d+1} . It remains to show the same for h_1^{d+1} . Consider the submodules

$$\begin{aligned} H^{d+1}(K, K_*)_0 &\coloneqq \ker(c_1^{d+2}\delta_1^{d+2}) \\ & \bigcup_{i \in I} \\ N &\coloneqq \operatorname{im}(d_1^{d+1}, c_2^{d+1}) \end{aligned}$$

of C_1^{d+1} . According to Lemma 4.4.1, we have colim $N = \operatorname{colim} H^{d+1}(K, K_*)_0$. Theorem F states that $[\operatorname{colim} N]_{C_1^{d+1}} = H^{d+1}(K, K_*)_0$. A generating system of the vector subspace colim N of colim C_1^{d+1} is given by the underlying matrix $\mathfrak{u}(\tilde{D}^{d+1})$ of $\tilde{D}^{d+1} \coloneqq c_1^{d+1}\delta^{d+1}$. Theorem 3.4.7 states that the matrix $k^{d+1} \coloneqq \operatorname{Bireduce}(\tilde{D}^{d+1})$ represents a basis of $H^{d+1}(K, K_*)_0$. Recall that by Proposition 4.2.7, it remains to compute the solution h_1^{d+1} of the linear system $\tilde{D}^{d+1} = k^{d+1}h_1^{d+1}$, which now does not involve the matrix D^{d+2} anymore. Thus, h_1^{d+1} can be computed without considering D^{d+2} .

Almost clearing An immediate optimization step that we have not mentioned so far is analogous to what we explained in Section 2.4.1. Namely, calling $\text{Bireduce}(\tilde{D}^{d+1})$ will reduce some columns of \tilde{D}^{d+1} to zero that could have been predicted. Since $\tilde{D}^{d+1}h_2^{d+1} = k^{d+1}h_1^{d+1}h_2^{d+1} = 0$, the sequence

$$0 \to H^{d+1}(K, K_*)_2 \xrightarrow{h_2^{d+1}} H^{d+1}(K, K_*)_1 \xrightarrow{\tilde{D}^{d+1}} C_1^{d+1}$$

is a chain complex; namely a free resolution of coker \tilde{D}^{d+1} . Since k^{d+1} is injective and since $H^{d+1}(K, K_*)_{\bullet}$ is exact, this complex can be seen as a free resolution of some quotient module M of C_1^{d+1} . Using KerAndMgsWithKer() (Algorithm 8), we compute a free resolution F_{\bullet} of coker \tilde{D}^{d+1} with $F_0 = C_1^{d+1}$, such that F_{\bullet} does not contain any homological 2-balls. This resolution F_{\bullet} fits into the commutative diagram

$$0 \longrightarrow H^{d+1}(K, K_*)_2 \xrightarrow{h_2^{d+1}} H^{d+1}(K, K_*)_1 \xrightarrow{h_1^{d+1}} H^{d+1}(K, K_*)_0 \longrightarrow H^{d+1}(K, K_*) \longrightarrow 0$$

$$0 \longrightarrow F_2 \xrightarrow{\swarrow} f_2 \xrightarrow{f_2} F_1 \xrightarrow{\swarrow} H^{d+1}(K, K_*)_0 \xrightarrow{\uparrow} H^{d+1}(K, K_*) \longrightarrow 0, \quad (4.52)$$

$$\overbrace{\tilde{D}^{d+1}}_{f_1} \xrightarrow{\tilde{D}^{d+1}} C_1^{d+1}$$

where coker $f_1 \cong \operatorname{coker} \tilde{D}^{d+1}$. Note that f_1 is a semi-minimal presentation of coker \tilde{D}^{d+1} , in the sense of [99]. Thus, we can obtain a matrix representing k^{d+1} by calling $\operatorname{Bireduce}(f_1)$ instead of $\operatorname{Bireduce}(\tilde{D}^{d+1})$. This may yield a different basis of $H^{d+1}(K, K_*)_0$ and thus a different matrix k^{d+1} , which nevertheless represents the same morphism k^{d+1} . The morphism \tilde{f}_1 then is the unique solution of the linear system $k^{d+1}\tilde{f}_1 = f_1$. Now f_2 and \tilde{f}_1 form a free resolution of $H^{d+1}(K, K_*)$ not containing any 2-balls.

Avoiding the factorization of f_1 It seems unnecessary that we first compute k^{d+1} by applying the column reduction algorithm Algorithm 10 to f_1 , and in the second step compute the matrix \tilde{f}_1 by applying Algorithm 7 to f_1 , which is again a column reduction scheme. This can be addressed by keeping track of the column operations during Algorithm 10. Let F and G be free modules, $B: F \to G$ be a valid graded matrix, and consider the free submodule $[\operatorname{colim} \operatorname{im} B]_G$ of G.

Algorithm 12: A variant of Algorithm 10 that keeps track of the column operations.

A valid graded $m \times n$ -matrix $B \colon F \to G'$ Input: **Output**: An injective valid graded $m \times n'$ -matrix $B' : F : \to F$ representing a basis of $F' = [\text{colim im } B]_G$, and a valid graded matrix $W \colon F \to F'$ such that B = B'Wfunction BireduceF(B): $W \leftarrow E \in k^{n \times n}$ $p \leftarrow 0 \in \mathbf{N}^n$ for j = 1, ..., n do while $i \leftarrow \text{c-piv}_{\vec{z}}(B_i) \neq 0$ do if $p_i = 0$ then $p_i = j$; break if $l-piv_{\vec{z}}(B_j) < l-piv_{\vec{z}}(B_{P_i})$ then swap p_i and j(a) (b) $B_j \leftarrow B_j - B_{ij}/B_{ip_i}B_{p_i}$ $W_{p_i,*} \leftarrow W_{p_i,*} + B_{ij}/B_{ip_i}W_{j*}$ (c) $B \leftarrow (B_i)_{B_i \neq 0}$ $W \leftarrow (W_{j*})_{B_i \neq 0}$ $p \leftarrow 0 \in \mathbf{N}^n$ for j' = 1, ..., n do $j \leftarrow j'$ while $i \leftarrow 1$ -piv_{\vec{z}} $(B_i) \neq 0$ do if $p_i = 0$ then $p_i = j$; break if $c-piv_{\vec{z}}(B_j) < c-piv_{\vec{z}}(B_{p_i})$ then swap p_i and j(d) $B_{j} \leftarrow B_{j} - B_{ij}/B_{ip_{i}}B_{p_{i}}$ $W_{p_{i},*} \leftarrow W_{p_{i},*} + B_{ij}/B_{ip_{i}}W_{j*}$ (e) $B' \leftarrow [B]_{rgB}$ $\mathsf{rg}^W \leftarrow \mathsf{cg}^{B'}$ $cg^W \leftarrow cg^B$ return B', W

Proposition 4.4.2. Algorithm 12 computes two valid graded matrix B' and W, such that B' represents a basis of [colim im B]_G and the following diagram commutes:



Proof. Let $F' = [\operatorname{colim} \operatorname{im} B]_G$. Algorithm 12 differs from Algorithm 10 only in maintaining the additional matrix W. Since operations on W do not affect the reduction of B, Theorem 3.4.7 gives that B' is a basis of $[\operatorname{colim} \operatorname{im} B]_G$. It follows from elementary linear algebra that B = B'W as ungraded matrices, and it remains to prove that W is valid. We show this inductively.

Let $B^{(0)} = B, \ldots, B^{(t)} = B'$ and $W^{(0)} = E_{n \times n}, \ldots, W^{(t)} = W$ be the intermediate steps of the matrices B and W during the algorithm. We regard all matrices $B^{(s)}$ and $W^{(s)}$ as graded matrices with

$$\mathrm{rg}^{B^{(s)}}\coloneqq\mathrm{rg}^{B},\qquad \mathrm{rg}^{W^{(s)}}_{j}\coloneqq\mathrm{cg}^{B^{(s)}}_{j}\coloneqqigVert_{B^{(s)}_{ij}
eq 0}\mathrm{rg}^{B}_{i},\qquad \mathrm{cg}^{W^{(s)}}\coloneqq\mathrm{cg}^{B}$$

This renders all matrices $B^{(s)}$ valid and matches the definition of the row and column grades of B' and W at the end of the algorithm. The graded matrix $B^{(0)} = W$ is valid by assumption. The unit matrix $W^{(0)}$ is valid because $cg^{W^{(0)}} = rg^{W^{(0)}} = cg^B$. By induction, we assume that $B^{(s-1)}$ and $W^{(s-1)}$ are valid for some s. Consider a pair of corresponding row and column

Algorithm 13: Relative cohomology algorithm

Coboundary matrices D^1, D^2, \ldots representing $C^{\bullet}(K, K_*)$ with $H^{\bullet}(K) = 0$ Input: **Output**: Matrices representing a minimal free resolution of $H^{d+1}(K, K_*)$ for d > 0. $k \leftarrow \texttt{Ker}((D^1 \Pi_0, D^1) \colon C_1^0 \to C_0^1)$ for $d = 0, 1, ..., d_{max}$ do $\leftarrow \begin{pmatrix} E \\ (0, D^{d+1}) \end{pmatrix} \mathcal{K} \colon \mathcal{H}^{d+1}(\mathcal{K}, \mathcal{K}_{*})_{1} \to \mathcal{C}_{1}^{d} \oplus \mathcal{C}_{2}^{d+1} \\ \leftarrow \begin{pmatrix} \Pi_{d+1}^{-1} D^{d+1} \Pi_{d} & 0 & \Pi_{d+1}^{-1} \\ 0 & D^{d+1} & \mathcal{E}_{d+1} \end{pmatrix} \colon \mathcal{C}_{1}^{d} \oplus \mathcal{C}_{2}^{d+1} \to \mathcal{C}_{1}^{d+1}$ h_2 Ñ $f_2, r, c \leftarrow \texttt{Minimize}(h_2)$ $\leftarrow \texttt{Reduce}(h_2)$ ▷ optional q $\leftarrow (r_j)_{j\notin \{\text{piv } q_k\}} \\ \leftarrow (\tilde{D}_j)_{j\in r}$ f_1 \triangleright colim im $f_1 =$ colim \tilde{D}_r if $d < d_{max}$ then $k, \tilde{f_1} \leftarrow \texttt{BireduceF}(f_1)$ $\triangleright f_1 = k \tilde{f_1}$ else $\tilde{f_1} \leftarrow \texttt{Bireduce}(f_1)$ $\widetilde{f_1}$, r, c \leftarrow Minimize $(\widetilde{f_1})$ $\widetilde{f}_2 \leftarrow ([f_2]_{i*})_{i \in c}$ yield $(\tilde{f}_2, \tilde{f}_1)$ \triangleright resolution of $H^{d+1}(K, K_*)$

operations

(a)

(b)

$$B_{i}^{(s)} \leftarrow B_{i}^{(s-1)} - \lambda B_{p_{i}}^{(s-1)}, \qquad \qquad W_{p_{i},*}^{(s)} \leftarrow W_{p_{i},*}^{(s-1)} + \lambda W_{i,*}^{(s-1)}$$
(4.53)

for some $\lambda \in k$, either in line (b) or line (e). Because $B^{(s-1)}$ is a \mathbb{Z}^2 -graded matrix, two columns grades $\operatorname{cg}_j^{B^{(s)}}$ and $\operatorname{cg}_{p_i}^{B^{(s)}}$ are comparable whenever two columns $B_{p_i}^{(s-1)}$ and $B_j^{(s)}$ have the same lex or colex pivot. By swapping p_i and j if necessary, lines (a) and (d) ensure that whenever the algorithm performs an operation as in (4.53), then $\operatorname{cg}_{p_i}^{B^{(s-1)}} \leq \operatorname{cg}_j^{B^{s-1}}$ and thus $\operatorname{cg}_j^{B^{(s)}} \leq \operatorname{cg}_j^{B^{(s-1)}}$. Using the induction hypothesis that $W^{(s-1)}$ is valid in (*), we obtain

$$\operatorname{rg}_{j}^{W^{(s)}} \stackrel{\text{def}}{=} \operatorname{cg}_{j}^{B^{(s)}} \le \operatorname{cg}_{j}^{B^{(s-1)}} \stackrel{\text{def}}{=} \operatorname{rg}_{j}^{W^{(s-1)}} \stackrel{(*)}{\le} \operatorname{cg}_{k}^{W^{(s-1)}} = \operatorname{cg}_{k}^{W^{(s)}}$$

for all k, so $W^{(s)}$ is valid. By induction, W is valid.

We use Algorithm 12 to compute a free resolution of $H^{\bullet}(K, K_*)$:

Proposition 4.4.3. Algorithm 13 computes graded matrices representing a minimal free resolution of $H^{d+1}(K, K_*)$ for each $0 \le d \le d_{max}$.

Proof. The two gray lines are optional, and we claim that they do not affect the correctness. We first prove correctness without these lines. The algorithm first computes matrices representing the free resolution F_{\bullet} of coker \tilde{D}^{d+1} from (4.52). Recall that F_{\bullet} contains no homological 2-balls. The algorithm computes k and \tilde{f}_1 such that $f_1 = k\tilde{f}_1$ using BireduceF(). Correctness of this step is given by Proposition 4.4.2. Then f_2 and \tilde{f}_1 represent a free resolution of $H^{d+1}(K, K_*)$ that does not contain any homological 2-balls. To obtain a minimal free resolution, it remains to split off all homological 1-balls, which is done by Minimize(). Note that the matrix k is only needed in the next iteration of the main for-loop. Thus, if $d = d_{\max}$, it suffices to compute \tilde{f}_1 but not k, using Bireduce().

Now, consider the algorithm with the two gray lines. Line (a) computes an ungraded, reduced matrix q with im $q = \text{im colim } f_1$. This implies that $u(1)f_1 = 0$. Consider the submatrix $f'_1 := ([f_1]_{r_j})_j$ of f_1 , where j ranges over the indices $j \ge 1$ that do not occur as the pivot of a column of q. Although im $f'_1 \subseteq \text{im } f_1$ may be a proper inclusion, f'_1 still satisfies colim im $f'_1 = \text{colim im } f_1$ and $\operatorname{rg}^{f'_1} = \operatorname{rg}^{f_1}$. This is enough to ensure that BireduceF(f_1) and Bireduce(f'_2) correctly compute matrices k and \tilde{f}_1 that fit into (4.52).

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Algorithm 14: A variant of Algorithm 12 with clearing.

A valid graded $I \times m$ -matrix $B \colon F \to G'$, a reduced valid graded $m \times n$ -matrix A such that Input: BA = 0**Output**: An injective valid graded $m \times n'$ -matrix $B' : F : \to F$ representing a basis of $F' = [\text{colim im } B]_G$, and a valid graded matrix $W \colon F \to F'$ such that B = B'Wfunction BireduceFC(B,A): $V \leftarrow W \leftarrow 1 \in k^{n \times n}$ ▷ (inverse) reduction matrix $r \leftarrow 0 \in \mathbf{N}^n$ \triangleright assignment pivot row \mapsto foreach k = 1, ..., n do column for A if $A_k \neq 0$ then $j \leftarrow \operatorname{piv} A_k$ $r_j \leftarrow k$ (a) $B_i \leftarrow 0$ $B', W \leftarrow \texttt{BireduceF}(B)$ (b) **foreach** nonzero $j \in \{piv A_k\}$ in ascending order **do** ▷ reconstruct W for cleared (c) $W_i \leftarrow 0$ columns of B $\left| \begin{array}{c} W_{j} \leftarrow -\frac{1}{A_{r_{j}r_{j}}}WA_{r_{j}} \\ \mathsf{rg}^{W} \leftarrow \mathsf{cg}^{B'} \end{array} \right.$ $cg^W \leftarrow cg^B$ return B', W

Clearing Algorithm 13 still has the shortcoming that when calling $BireduceF(\tilde{D})$, it spends some time on reducing columns of \tilde{D} to zero that can be predicted. In a last step, we show an algorithm that avoids this. As before, let $B: F \to G$ be a valid graded matrix. Our goal is to compute a basis $B': F' \to G$ of $F' := [colim im B]_G$, and a valid graded matrix $W: F \to F'$ such that B = B'W.

Assume that we know BA = 0 for some reduced matrix A. Let $j_1 < \cdots < j_t$ be the pivot indices of the non-zero columns of A, and let B'' be the matrix obtained from B by setting B_{j_1}, \ldots, B_{j_t} to zero. Then B and B'' generate two submodules im $B'' \subseteq \text{ im } B \subseteq G$, such that colim im $B'' = \text{colim im } B \subseteq \text{colim } G$. Let B', U = BireduceF(B''). Then B' represents a basis of [colim im $B]_G$, and U satisfies B'' = B'U. We obtain the commutative diagram



where $J_j = \begin{cases} 0 & \text{if } j = j_s \text{ for some } s, \\ e_j & \text{otherwise} \end{cases}$. We extend Algorithm 12 such that it computes the matrix W from U and A.

Proposition 4.4.4. Let F, G be free modules, $B: F \to G$ be a valid graded matrix and A be a matrix such that u(B)A = 0. Then Algorithm 14 computes a valid graded matrix $B': F' \to G$ representing a basis of $F' := [\operatorname{colim} \operatorname{im} B]_G$, and a valid graded matrix $W: F \to F'$ such that B = B'W.

The following will be proven after the proposition:

Lemma 4.4.5. Let B, B' be valid graded matrices such that $\operatorname{cg}_{j}^{B'} = \bigvee_{B'_{ij}\neq 0} \operatorname{rg}_{j}^{B'}$ for all j. If W is a graded matrix with row and column grades $\operatorname{cg}^{W} = \operatorname{cg}^{B}$ and $\operatorname{cg}^{W} = \operatorname{rg}^{B'}$ such that u(B) = u(B')u(W), then W is valid. Proof of Proposition 4.4.4. Since BA = 0, setting $B_j \leftarrow 0$ in line (a) if j occurs as the pivot index of some column of A does not change the image of B; so im $B'' = \operatorname{im} B$. It follows from Proposition 4.4.2 that B' represents a basis of [colim im B]_G.

Let W denote the state of W after line (a), and W' be the state of W when the algorithm terminates. It remains to show that W' is valid and satisfies B = B'W'. Let $q = \{j_1 < \cdots < j_t\} := \{\text{piv } A_k \mid A_k \neq 0\}$. If $j \in q$, then BireduceF(B) never performs a column addition from B_j to another column, and thus also no row addition from W_{j*} to any other row of W. This implies that $B_j = B'_j W'_j$ for all $j \notin q$, and that W'_j is valid in this case.

It remains to show the same for $j \in q$. Let $W^{(0)} = W, \ldots, W^{(t)} = W'$ be the intermediate values the matrix W takes during the loop in line (b). Note that $W_j^{(0)} = 0$ for all $j \in q$ because of line (c). We show by induction on $s = 1, \ldots, t$ that for all $j \notin \{j_s, \ldots, j_t\}$, we have $B_j = B'_j W_j^{(s)}$. For s = 1, there is no $j < j_1$ in q, some the statement holds in this case. Assume by induction that for some $s \ge 1$, we have $B_j = BW_j^{(s-1)}$ if $j \notin \{j_s, \ldots, j_t\}$. Note that $W_j^{(s-1)} = 0$ for all $j \in \{j_s, \ldots, j_t\}$. Let $k = r_{j_s}$ be the column index of A with $j_s = \text{piv} A_k$. We obtain that

$$\begin{aligned} 0 &= BA_k = B_{j_s}A_{j_sk} + \sum_{j < j_s} B_j A_{jk} \stackrel{(i)}{=} B_{j_s}A_{j_sk} + \sum_{j < j_s} B' W_j^{(s-1)} A_{jk} \\ \stackrel{(ii)}{=} B_{j_s}A_{j_sk} + \sum_{j \le j_s} B' W_j^{(s-1)} A_{jk} \stackrel{(iii)}{=} B_{j_s}A_{j_sk} + B' W^{(s-1)} A_k, \end{aligned}$$

where (i) holds by the induction hypothesis, (ii) holds because $W_{j_s}^{(s-1)} = 0$, and (iii) holds because j_s is the pivot of A_k . Therefore, we get that

$$B_{j_s} = -\frac{1}{A_{j_s k}} B' W^{(s-1)} A_k$$

so setting

$$W_j^{(s)} = egin{cases} W_j^{(s-1)} & ext{if } j
eq j_s, \ -rac{1}{A_{j_s,k}} W^{(s-1)} A_k & ext{otherwise} \end{cases}$$

satisfies $B_j = B'W_j^{(s)}$ for all $j \notin \{j_{s+1}, \ldots, j_t\}$. Inductively, we obtain that B = B'W'. It follows from Lemma 4.4.5 that W' is valid.

Proof of Lemma 4.4.5. We have

$$\mathrm{cg}_{j}^{W} = \mathrm{cg}_{j}^{B} \overset{(*)}{\geq} \bigvee_{B_{ij} \neq 0} \mathrm{rg}_{i}^{B} \overset{(\dagger)}{\leq} \bigvee_{W_{lj} \neq 0} \bigvee_{B'_{il} \neq 0} \mathrm{rg}_{i}^{B} \overset{(\ddagger)}{\leq} \bigvee_{W_{lj} \neq 0} \mathrm{cg}_{l}^{B'} = \bigvee_{W_{lj} \neq 0} \mathrm{rg}_{l}^{W},$$

where (*) holds because B is valid, (†) holds because u(B) = u(B')u(W) and (‡) holds because B' is valid. Now, (‡) is an equality by assumption, (†) is an equality because B' is bireduced. Then we obtain that $cg_j^W \ge \bigvee_{W_{l_j} \neq 0} rg_l^{W'}$, so W is valid. \Box

Remark. We provide another, more conceptual proof of Lemma 4.4.5. Let F, F' and M be free modules such that $B: F' \to M$ and $B': F' \to M$, and let $\vec{z} = \operatorname{rg}^B = \operatorname{rg}^{B'}$, $N \coloneqq \operatorname{im} B$ and $N' \coloneqq \operatorname{im} B'$. For any module L and any morphism f of modules, let $\bar{L} \coloneqq \operatorname{colim} F$ and $\bar{f} \coloneqq \operatorname{colim} f$. For $\mathfrak{u}(B) = \mathfrak{u}(B')\mathfrak{u}(W)$, it follows that $\bar{N} \subseteq \bar{N}'$. From the definition of $[-]_M$, it follows that there is an injective map $n: N = [\bar{N}]_M \subseteq [\bar{N}']_M = N'$. By Lemma 3.4.6, the matrix

B' represents a basis of $[V']_M$, so im $B \subseteq [V]_M \subseteq [V']_M = \operatorname{im} B'$. We obtain a commutative diagram



Because there is an isomorphism $p': F' \to N'$, there exists a unique morphism $f: F \to F'$ such that np = p'f. Because \bar{p}' is an isomorphism, the colimit \bar{f} is the unique morphism such that $\bar{n}\bar{p} = \bar{p}'\bar{f}$. From the assumption, we obtain that

$$\overline{\imath}'\overline{n}\overline{p} = \overline{\imath}\overline{p} = \mathsf{u}(B) = \mathsf{u}(B')\mathsf{u}(W) = \overline{\imath}'\overline{p}'\mathsf{u}(W).$$

Injectivity of \bar{i}' implies that $\bar{n}\bar{p} = \bar{p}'\mathfrak{u}(W)$. Uniqueness of \bar{f} implies that $\mathfrak{u}(W) = \bar{f}$ and thus W = f. In particular, W is the representative matrix of a morphism of free modules and thus valid.

We can use Algorithm 14 to establish another algorithm that computes a minimal free resolution of $H^{d+1}(K, K_*)$:

Proposition 4.4.6. Let K_* and D^{\bullet} be as above, and let $n_d \coloneqq |K^d|$. Then for each $d \ge 0$, Algorithm 15 computes a minimal free resolution of $H^{d+1}(K, K_*)$.

Algorithm 15: Relative cohomology algorithm II

Input: Coboundary matrices D^1, D^2, \ldots representing $C^{\bullet}(K, K_*)$ with $H^{\bullet}(K) = 0$ **Output**: Matrices representing a minimal free resolution of $H^{d+1}(K, K_*)$ for d > 0.

 $k \leftarrow \operatorname{Ker}((D^1 \Pi_0, D^1): C_1^0 \to C_0^1)$ for d = 0, 1, ... do
$$\begin{split} & a = 0, 1, \dots, \mathbf{u} \\ & h_2 \leftarrow \binom{k}{(0, D^{d+1})k} : H^{d+1}(K, K_*)_1 \to C_1^d \oplus C_2^{d+1} \\ & \tilde{D} \leftarrow \binom{\prod_{d+1}^{-1} D^{d+1} \prod_d & 0 & \prod_{d+1}^{-1}}{0 & D^{d+1} & E_{d+1}} : C_1^d \oplus C_2^{d+1} \to C_1^{d+1} \end{split}$$
Let $R = Vu(h_2)$ be reduced standard reduction (Algorithm 1) of ungraded matrices if d = 0 then $(k, h_1) \leftarrow \texttt{BireduceF}(\tilde{D})$ (a) else if $d < d_{max}$ then $(k, h_1) \leftarrow \texttt{BireduceFC}(\tilde{D}, R)$ (b) else for $j \in \{ piv A_k \mid A_k \neq 0 \}$ do $\tilde{D}_j \leftarrow 0$ (c) $h_1 \leftarrow \texttt{Bireduce}(\tilde{D})$ \triangleright resolution of $H^{d+1}(K, K_*)$ yield MinimizeCpx (h_1, h_2)

Proof. We have to show that the matrices h_2 , h_1 computed in the *d*th iteration of the main for-loop in Algorithm 15 represent the morphisms h_2^{d+1} , h_1^{d+1} from the free resolution (4.52) of $H^{d+1}(K, K_*)$. For d = 0, this follows directly from Proposition 4.4.2. Because $\tilde{D}h_2$ is a free resolution of coker \tilde{D} , we have $\tilde{D}h_2 = 0$. Then also $u(\tilde{D})R = 0$ as ungraded matrices, because Ris obtained from h_2 by (ungraded) standard matrix reduction. Setting the columns \tilde{D}_j to zero for $j = \text{piv } R_k$ thus does not change the image of \tilde{D} . Then Lemma 4.4.5 and Proposition 4.4.4 imply that h_1 and h_2 form a minimal free resolution of $H^{d+1}(K, K_*)$.

4.5 Computational shortcomings

Recall that, to compute a free resolution of $H^d(K_*)$ as described in 4.2.6, we have to compute, i.a., a matrix κ^d representing the inclusion of $\ker(\delta_0^{d+1}, -c_1^{d+1})$ and solve the factorization problem

$$\kappa^{d} h_{1}^{d} = \begin{pmatrix} -\delta_{0}^{d} & c_{1}^{d} & 0\\ 0 & \delta_{1}^{d+1} & c_{2}^{d+1} \end{pmatrix}$$
(4.54)

for h_1^d . Let $n_d := |K_*^d|$, and let D^d represent the coboundary morphism δ^d of the cochain complex $C^d(K)$ of vector spaces. By definition, $(\delta_0^{d+1}, -c_1^{d+1})$ is represented by a graded matrix with the underlying block matrix

$$(D^{d+1} \mid E_{d+1} \mid \Pi_{d+1}) : C_0^d \oplus C_1^{d+1} \to C_0^{d+1}$$
 (4.55)

of size $n_{d+1} \times (n_d + 2n_{d+1})$, and its kernel has rank $n_d + n_{d+1}$. Similarly, the right hand side of (4.54) is represented by the block matrix

$$\begin{pmatrix} \frac{-D^d}{0} | \frac{E_d}{\Pi_{d+1}^{-1} D^{d+1} \Pi_d} \frac{0}{0} \\ 0 & 0 & D^{d+1} | \frac{-E_d}{-E_{d+1}} \end{pmatrix} : C_0^{d-1} \oplus C_1^d \oplus C_2^{d+1} \to C_0^d \oplus C_1^{d+1}$$

of shape $(n_d + 2n_{d+1}) \times (n_{d-1} + 2n_d + n_{d+1})$. In experiments with a prototype implementation not reported here, it has been observed that the mere size of these two matrices causes high running times. Also, it has been observed that computing the kernel of (4.55) using Ker() causes considerable fill-up of matrix columns (see also [17]), which further increases the runtime.

When working with relative cohomology, the situation is slightly better. Namely, as described in 4.2.8, to compute $H^{d+1}(K, K_*)$, one has to compute the kernel of the morphism $(c_1^{d+2}\delta_1^{d+2})$, which is represented by the matrix

$$(D^{d+2}, D^{d+2}\Pi_{d+1}) \colon C_2^{d+1} \to C_0^{d+2}$$

of size $n_{d+2} \times 2n_{d+1}$. As described above, this can be avoided using one of Algorithms 13 and 15. When using Algorithm 13, the expensive step is to run **BireduceF()** (Algorithm 12) on the block matrix

$$\begin{pmatrix} \Pi_{d+1}^{-1} D^{d+1} \Pi_d & 0 \\ 0 & D^{d+1} \\ E_{d+1} \\ \end{pmatrix} : C_1^d \oplus C_2^{d+1} \to C_1^{d+1}$$
(4.56)

of size $2n_{d+1} \times (2n_d + n_{d+1})$. Applying BireduceF() on this matrix has the downside that it needs to reduce many columns to zero, which is why we introduced Algorithm 15 that builds upon BireduceFC() (Algorithm 14). The problem with this approach is the following. Let \tilde{D} denote the matrix from (4.56) as in Algorithm 15. Then by the design of Algorithm 15, it is necessary that BireduceFC(\tilde{D}) outputs both the bireduced reduced matrix $R = \tilde{D}V$ and the reduction matrix V. However, as a clearing strategy, BireduceFC(\tilde{D}) set certain columns of R to zero directly. As described above, if a column R_j has been set to zero by clearing, one has to do some computations to obtain the corresponding column V_j , which is done in Algorithm 14, line (b). This process cannot be parallelized in a simple way. Experiments have shown that the loop in Algorithm 14, line (b) consumes all runtime benefit arising from clearing. For a comparison with Chapter 3, we note that the situation is different in Algorithm 11. Here, the situation is that when $Bireduce(D^{d+1})$ is used to compute a bireduced matrix $R = D^{d+1}V$ for some V, then Algorithm 11 only needs the matrix R for further computations, but not the reduction matrix V. Consequently, it is not necessary to compute the column V_j of V if $R_j = 0$ has been obtained through clearing in this case.

4.6 Absolute and relative cohomology do not determine each other

To finish this section, we present an example that shows that in two-parameter persistence, $H^{\bullet}(K_*)$ and $H^{\bullet}(K, K_*)$ need not determine each other, unless K is acyclic. This is analogous to Section 3.6, where we provided an example that shows the analogous statement for $H^{\bullet}(K_*)$ and $H^{\bullet}(N^{\bullet}(K_*))$. This is different from one-parameter persistence: recall that in this case, $H^{\bullet}(K_*)$ and $H^{\bullet}(K, K_*)$ determine each other uniquely up to isomorphism even if K is not acyclic; see Corollary 2.2.11. Specifically, we show:

Theorem B(b). There exist one-critically two-parameter filtered simplicial complexes K_* , L_* and M_* , such that

- (i) $H^{\bullet}(K, K_*) \cong H^{\bullet}(L, L_*)$, but $H^{\bullet}(K_*) \ncong H^{\bullet}(L_*)$, and
- (ii) $H^{\bullet}(K_*) \cong H^{\bullet}(M_*)$, but $H^{\bullet}(K, K_*) \ncong H^{\bullet}(M, M_*)$.

The colimits K, L and M are not acyclic.

Proof. We use the same complexes as in Section 3.6, but leave out the last 2-cell; see Figure 4.9. Again, we compute non-reduced cohomology with coefficients in \mathbf{F}_2 . An example for reduced cohomology can be constructed analogously.

Let $z > g(\sigma)$ for all $\sigma \in K_*$, L_* and M_* . For K_* , we get

$$R_{z}C^{\bullet}(K_{*}): \qquad 0 \longrightarrow \underbrace{\stackrel{[x]}{\underset{\downarrow}{\times}} \stackrel{[y]}{\underset{\downarrow}{\times}} \stackrel{[x]}{\underset{\downarrow}{\times}} \stackrel{[y]}{\underset{\downarrow}{\times}} \longrightarrow \underbrace{\stackrel{[e]}{\underset{\downarrow}{\times}} \stackrel{[e]}{\underset{f_{\perp}}{\xrightarrow}} \stackrel{[e]}{\underset{f_{\perp}}{\xrightarrow}} \stackrel{[e]}{\underset{f_{\perp}}{\xrightarrow}} \stackrel{[e]}{\underset{K_{z}C^{0}(K_{*})}{\xrightarrow}} 0,$$

$$C^{\bullet}(K,K_{*}): \qquad 0 \longrightarrow \underbrace{\stackrel{[x]}{\underset{\downarrow}{\times}} \stackrel{[y]}{\underset{\downarrow}{\xrightarrow}} \stackrel{[x]}{\underset{\downarrow}{\times}} \stackrel{[e]}{\underset{\downarrow}{\xrightarrow}} \stackrel{[e]}{\underset{\iota}{\xrightarrow}} \stackrel{[e]}{\underset{\iota}{\xrightarrow}} \stackrel{[e]}{\underset{\iota}{\xrightarrow}} \stackrel{[e]}{\underset{\iota}{\xrightarrow}} \stackrel{[e]}{\underset{\iota}{\xrightarrow}} \stackrel{[e]}{\underset{\iota}{\xrightarrow}} \stackrel{[e]}{\underset{\iota}{\xrightarrow}} \stackrel{[e]}{\underset{\iota}{\xrightarrow}} \stackrel{[e]}{\underset{\iota}{\xrightarrow}} \stackrel{[e$$

We compute generators and relations of $H^{\bullet}(K_*)$ and $H^{\bullet}(K, K_*)$ as in Section 4.3 and obtain the minimal free presentations

$$R_{z}H^{0}(K,K_{*}) = \frac{\langle \bar{x} + \bar{y}, x_{\perp} + \bar{y}, x_{\perp} + y_{\perp} \rangle}{\langle \bar{x} + \bar{y} \stackrel{x}{\equiv} x_{\perp} + \bar{y} \stackrel{y}{\equiv} x_{\perp} + y_{\perp} \rangle}, \quad R_{z}H^{1}(K,K_{*}) = \frac{\langle e_{\perp}, \bar{f} \rangle}{\langle e_{\perp} \stackrel{x}{\equiv} \bar{f}, e_{\perp} \stackrel{y}{\equiv} \bar{f} \rangle},$$

$$R_{z}H^{0}(K_{*}) = \frac{\langle x^{\bar{\imath}} + y^{\bar{\imath}}, x^{\bar{\imath}} + e_{\perp} + \bar{f} \rangle}{\langle \bar{x} + \bar{y}, x_{\perp} + y_{\perp} \rangle}, \quad R_{z}H^{1}(K_{*}) = \frac{\langle e^{\bar{\imath}} \equiv \bar{f} \rangle}{\langle f_{\perp}, f_{\perp} \rangle},$$
(4.57)



Figure 4.9: The filtered simplicial complexes from the proof of Theorem B(b).



Figure 4.10: The relative and absolute cohomology modules (4.57) of K_* from Figure 4.9a.



Figure 4.11: The relative and absolute cohomology modules (4.58) of the filtered complex L_* from Figure 4.9b.



Figure 4.12: The relative and absolute cohomology modules (4.59) of the filtered complex M_* from Figure 4.9c.
see Figure 4.10. All four modules are indecomposable, and all higher homology modules are zero. Similarly, we obtain the minimal presentations

$$H^{0}(L,L_{*}) = \frac{\langle \bar{x} + \bar{y} + \bar{z}, x_{\perp} + \bar{y} + \bar{z}, x_{\perp} + y_{\perp} + z_{\perp} \rangle}{\langle \bar{x} + \bar{y} + \bar{z} \stackrel{x}{\equiv} x_{\perp} + \bar{y} + \bar{z} \stackrel{y}{\equiv} x_{\perp} + y_{\perp} + z_{\perp} \rangle}, \qquad H^{1}(L,L_{*}) = \frac{\langle e_{\perp}, \bar{f} \rangle}{\langle e_{\perp} \stackrel{\delta z_{\perp}}{\equiv} \bar{f}, e_{\perp} \stackrel{\delta z_{\perp}}{\equiv} \bar{f} \rangle}, \\H^{0}(L_{*}) = \frac{\langle x^{\bar{z}} + \bar{y} + \bar{z} \rangle}{\langle \bar{z} + \bar{y} + \bar{z}, x_{\perp} + y_{\perp} + z_{\perp} \rangle} \oplus \frac{\langle z^{\bar{z}} + e_{\perp} + \bar{f} \rangle}{\langle \bar{z} + \bar{e}, z_{\perp} + f_{\perp} \rangle}, \qquad H^{1}(L_{*}) = \frac{\langle e^{\bar{z}} \equiv \bar{f} \rangle}{\langle e_{\perp}, \bar{f} \rangle},$$

$$(4.58)$$

and

$$H^{0}(M, M_{*}) = H^{0}(L, L_{*}), \qquad H^{1}(M, M_{*}) = \frac{\langle e_{\downarrow} + \lceil f \rangle}{\langle \lceil y + \lfloor e, x_{\downarrow} + \lceil f \rangle} \oplus \frac{\langle \lceil g, g_{\downarrow} \rangle}{\langle \underline{g} \rangle},$$

$$H^{0}(M_{*}) = \frac{\langle \gamma_{1}, \gamma_{2} \rangle}{\langle \gamma_{1}^{\ r_{1}, r_{3}, r_{5}} 0, \gamma_{1} \stackrel{r_{2}}{\equiv} \gamma_{2}, \gamma_{2} \stackrel{r_{2}}{\equiv} 0 \rangle}, \qquad H^{1}(M_{*}) = \frac{\langle \overline{g} \rangle}{\langle \lceil g, g_{\downarrow} \rangle},$$

$$(4.59)$$

with

$$\begin{array}{ll} \gamma_{1} = x^{\overline{1}} + y^{\overline{1}} + z^{\overline{1}}, & \gamma_{2} = x^{\overline{1}} + e_{\bot} + f + \overline{g}, \\ r_{1} = x + y + z : g_{1} \equiv 0, & r_{2} = y + z + e : g_{1} \equiv g_{2}, \\ r_{3} = x_{\bot} + y + z : g_{1} \equiv 0, & r_{4} = x_{\bot} + f + g : g_{2} \equiv 0, \\ r_{5} = x_{\bot} + y_{\downarrow} + z_{\downarrow} : g_{1} \equiv 0; & \end{array}$$

see Figures 4.11 and 4.12.

These are minimal presentations. All other cohomology modules are zero. Comparing Figures 4.10 to 4.12 shows that $H^d(K, K_*) \cong H^d(L, L_*)$ and $H^d(L_*) \cong H^d(M_*)$ for all d. Nevertheless, $H^0(K_*) \cong H^0(L_*)$ and $H^1(K, K_*) \cong H^1(M, M_*)$, because one is indecomposable while the other is not. This finishes the construction.

To finish this section, we show a two-parameter filtered cell complex Q_* such that the short exact sequences

$$0 \to \operatorname{coker} p^{d} \to R_{z} H^{d}(Q_{*}) \to \ker p^{d+1} \to 0,$$

$$0 \to \operatorname{coker} i^{d} \to R_{z} H^{d+1}(Q, Q_{*}) \to \ker i^{d+1} \to 0$$
(4.60)

_

induced by the long exact sequence

$$0 \longrightarrow R_{z}H^{0}(Q,Q_{*}) \xrightarrow{p^{0}} R_{z}H^{0}(Q) \xrightarrow{i^{0}} R_{z}H^{0}(Q_{*}) \xrightarrow{\delta} \delta$$

$$(4.61)$$

$$(4.61)$$

$$(4.61)$$

never split for any d. Namely, let Q_* be the cell complex with cellular chain complex

$$C_{\bullet}(Q_{*}): \cdots \to \langle \rho_{3}, \sigma_{3}, \tau_{3} \rangle \xrightarrow{\begin{pmatrix} 1 & 1 & 1 \\ 0 & 0 & 0 \\ 1 & 1 & 1 \end{pmatrix}} \langle \rho_{2}, \sigma_{2}, \tau_{2} \rangle \xrightarrow{\begin{pmatrix} 1 & 1 & 1 \\ 0 & 0 & 0 \\ 1 & 1 & 1 \end{pmatrix}} \langle \rho_{1}, \sigma_{1}, \tau_{1} \rangle \xrightarrow{\begin{pmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \end{pmatrix}} \langle \rho_{0}, \tau_{0} \rangle \to 0,$$

one-critically \mathbf{Z}^2 -filtered with

$$g(\rho_d) = (2d, 2d+2),$$
 $g(\sigma_d) = (2d+1, 2d+1),$ $g(\tau_d) = (2d+2, 2d)$

for all d. We obtain the minimal free presentation matrices

$$R_{z}H^{0}(Q,Q_{*}) = \operatorname{coker}\begin{bmatrix} \rho_{0} + \sigma_{0} & 1 & 0\\ \rho_{0} + \sigma_{0} & 1 & 1\\ \rho_{0} + \sigma_{0} & 1 & 1\\ 1 & 1 & 1 \end{bmatrix},$$
(4.62)

$$R_{z}H^{d}(Q,Q_{*}) = \operatorname{coker} \begin{bmatrix} \rho_{d} + \tau_{d-1} & \sigma_{d} & \rho_{d-1} + \tau_{d} \\ \rho_{d} + \tau_{d} & 1 & 1 \\ \sigma_{d} & \sigma_{d} & 1 & 1 \\ \end{pmatrix},$$
(4.63)

$$R_{z}H^{d}(Q_{*}) = \operatorname{coker} \begin{bmatrix} \rho_{d}^{\neg} + \rho_{d+1} + \bar{\sigma}_{d+1} + \bar{\sigma}_{d+1} + \bar{\sigma}_{d+1} & \rho_{d} + \bar{\sigma}_{d+1} + \bar{\sigma}_{d} & \rho_{d} + \bar{\sigma}_{d} \\ \rho_{d}^{\neg} + \rho_{d+1} + \sigma_{d+1} + \bar{\sigma}_{d+1} & 1 & 1 \\ \rho_{d}^{\neg} + \rho_{d+1} + \bar{\sigma}_{d+1} + \bar{\sigma}_{d+1} & 1 & 1 \\ \rho_{d}^{\neg} + \rho_{d+1} + \bar{\sigma}_{d+1} + \bar{\sigma}_{d+1} & \rho_{d}^{\neg} + \bar{\sigma}_{d+1} \\ \rho_{d}^{\neg} + \rho_{d+1} + \bar{\sigma}_{d+1} + \bar{\sigma}_{d+1} & \rho_{d}^{\neg} + \bar{\sigma}_{d+1} \\ R_{z}H^{d}(Q_{*}) = \operatorname{coker} \begin{bmatrix} \rho_{d}^{\neg} + \rho_{d+1} + \bar{\sigma}_{d+1} + \bar{\sigma}_{d+1} & \rho_{d} + \bar{\sigma}_{d+1} \\ \rho_{d}^{\neg} + \rho_{d+1} + \bar{\sigma}_{d+1} & \rho_{d}^{\neg} + \bar{\sigma}_{d+1} \\ \rho_{d}^{\neg} + \rho_{d+1} + \bar{\sigma}_{d+1} + \bar{\sigma}_{d+1} \\ \rho_{d}^{\neg} + \rho_{d+1} + \bar{\sigma}_{d+1} + \bar{\sigma}_{d+1} \\ \rho_{d}^{\neg} + \rho_{d+1} \\ \rho_{d}^{\neg} + \rho_{d} \\ \rho_{d}^{\neg} + \rho_{d+1} \\ \rho_{d}^{\neg} + \rho_{d+1} \\ \rho_{d}^{\neg} +$$

for all d > 0 and $d \ge 0$, respectively. The rows and columns of these matrices are decorated by the representatives in C^{\bullet}_{\bullet} . These are graded matrices, where each row and column has as grade the join of the grades of the summands in the representative attached to it. The long exact sequence (4.61) is illustrated in Figure 4.13. Each of the modules $H^d(Q, Q_*)$, $H^d(Q)$ and $H^d(Q_*)$ is indecomposable. In particular, the short exact sequences (4.60) cannot split for any d.



Figure 4.13: The long exact cohomology sequence (4.61) of the complex Q_* . The diagrams depict the minimal free presentations (4.62) to (4.64) of the respective cohomology modules. The symbols •, • and × denote the grades of the generators, relations and 2-syzygies of the module. The labels stand for the simplex that determines the grade of the generator, relation or syzygy. Where two colors overlap, the components of the respective module are two-dimensional. The small matrices indicate structure maps into and out of these regions. Each of the modules is indecomposable because it contains an indecomposable quiver representation.

Chapter 5

Implementation and experiments

To show that computing two-parameter persistent cohomology is feasible, we provide a C++ implementation of our approach for the computation of two-parameter persistent homology and cohomology from Chapter 3. In this section, we provide some details about and experimental results obtained with this implementation.

5.1 Implementation

Our software 2pac (2-parameter cohomology), which is publicly available at [94], implements the two-parameter cohomology algorithm (Algorithm 11) from Chapter 3 and the homology algorithm (Algorithm 9) from Section 2.4.

We do not include the absolute and relative cohomology algorithms from Chapter 4, because earlier experiments with a prototype implementation did not show any performance benefits over the cohomology algorithm.

The code computes all resolutions with coefficients in \mathbf{F}_2 . The software accepts as input either a sequence of matrices, or a file containing a filtration and a distance matrix of a point cloud. Both kinds of input are read as a binary format described in the file README.md. For the interested reader, we provide an overview over the relevant files and classes and explain technical details in the implementation of algorithms presented in this thesis.

5.1.1 Retrieval, building, using

The software can be retrieved from the git repository at [94]. Building the software requires a reasonably recent version of Boost (including Boost::program_options) and OpenMP to be installed. To build the software, follow the instructions in README.md in the repository. The software has been successfully built on Mac OS using clang++ 15.0.7, and on Ubuntu Linux using g++11.2.0.

To run the software on one of the samples, run the command ./2pac -f samples/[file \leftrightarrow name]. This computes minimal free resolutions of $H^{d+2}(N^{\bullet}(K_*))$ and $H_d(K_*)$, where K_* is the density-Rips complex on the filtered point cloud in [filename]. Run ./2pac -help for an overview of the command line arguments. See README.md for a description of the binary file format.

Jupyter notebook EXAMPLE.ipynb may be convenient to produce the input data, call 2pac, and visualize the output. This notebook can be used to generate point clouds sampled from different spaces, choose a density function and bandwidth parameter interactively, write the filtered point cloud to disk in the required binary format, invoke 2pac and read its output, and visualize the computed minimal free resolutions. The notebook can also be used to read a file in scc2020 format [97] and convert it to the binary format.

5.1.2 Organization of the source code

For the reader who is interested in implementation details, we provide an overview of the source code. The code is grouped in several pairs of corresponding source and header files. A documentation of the classes, functions and files can be generated by running doxygen in the folder. The documentation can then be found in the sub-folder documentation/index.html. We give a short overview of the files and classes of our code. See the documentation in the header files for details.x

- **2pac.cpp** Entry point. The function main reads an input file (which describes a filtered point cloud), sets up its function-Rips complex K_* , applies preprocessing filters (such as chunk preprocessing) according to command line arguments, and computes minimal free resolutions of $H^{d+2}(N^{\bullet}(K_*))$ or $H_d(K_*)$. For each dimension, the computation finishes by printing the Betti numbers. During the computation, the program prints the runtimes for each of the steps of the algorithm.
- computation.hpp Contains classes Cohomology, Homology and RelativeCohomology that implement the computation of minimal free resolutions of (co)homology as described in Algorithms 9, 11 and 13, respectively. To use these, construct an object of the respective type, and call its operator() on the (co)boundary matrices D^1, D^2, \ldots consecutively, which happens in 2pac.cpp.
- matrices.hpp Implements a class GradedMatrix, which represents a Z^2 -graded matrix. The underlying ungraded matrix is implemented by the class SparseMatrix, which represents a matrix as a list of columns. Each column is implemented as an object of type ColumnType, which can be one of HeapColumn and ArrayColumn. See Section 5.1.3 for details. The column type to be used is specified at compile time. The default is heaps; to use vectors, compile with -DARRAY_MATRICES.
- bireduce.hpp implements Algorithm 10 in the function extend_from_colimit. The file defines a struct Params, which can be passed to extend_from_colimit to set some details of the algorithm. This is intended for testing; the defaults should be fine. The function sparsify implements Algorithm 22, which is used to reduce the density the matrix computed by Algorithm 10; see Section 5.1.5.
- complexes.hpp provides the abstract class Complex, which represents either a chain or a cochain complex. A (co)chain complex has a method to yield the next (co)boundary matrix. The file provides the following classes deriving from Complex:
 - **Pointcloud** reads a filtered point cloud from a binary file from disk, and assembles the coboundary matrices of the associated function-Rips complex; for generating the matrices, the class uses the combinatorial number system;
 - MatricesFromFile reads a sequence of \mathbb{Z}^2 -graded matrices from disk, and checks that they represent a cochain complex indeed.

We also provide filters, which take as input a std::unique_ptr to a Complex and represent a new complex, obtained by applying some operation to the original one. For that reason, instances of Complex should always be created through std::make_unique. For example, complexes.hpp provides

- **TransposeComplex** converts a cochain complex to a chain complex and vice versa by taking the graded transpose of all its (co)boundary matrices.
- FlipGrades exchanges the first and second coordinate of all \mathbb{Z}^2 -grades of the rows and columns of the (co)boundary matrices.
- AugmentComplex takes the augmented complex of a given (co)chain complex, used to compute reduced (co)homology.

Other filters, provided in separate files are the following:

- **Cone.hpp** defines the classes **Cone** and **HBasisCone** deriving from **Complex**. The former makes the complex a cone for large filtration values by forming the simplicial cone as described in Section 3.7.1. The latter implements the strategy from Section 3.7.2; see Section 5.1.4 for implementation details.
- chunk.hpp implements a class Chunk deriving from Complex, which implements the chunk preprocessing Algorithm 4 or Algorithm 5, depending on whether it is applied to a chain or cochain complex. Both variants call minimize on the (co)boundary matrices subsequently. It is possible to specify a largest dimension to which Algorithm 3 is supposed to be applied.
- factor.hpp, minimize.hpp and lw.hpp implements Algorithms 3, 7 and 8, respectively.
- indirect.hpp contains classes IndirectColumn and IndirectMatrix, which implement a matrix representation through maintaining a heap of pointers to another, underlying matrix. See Section 5.1.3.3 for details.

The other files contain utilities, helper and wrapper functions and classes, which should be self-explanatory.

5.1.3 Sparse matrices

Most algorithms presented in this thesis are matrix column reduction schemes. It is common in implementations of persistent homology to store matrices in a column sparse way; that is, a matrix is represented by a contiguous list (called a *vector*) of its columns in a way that allows accessing the *n*th column in constant time. In our implementation, we use a std::vector<T> of the C++ standard library, where the type T implements a matrix column. It has to provide efficient access to the pivot of a column, and an efficient implementation of the addition of two columns. Some operations, such as minimization (Algorithm 3), also require the efficient deletion of the pivot of a column and retrieval of the new pivot. In the context of one parameter persistent homology, various matrix formats have been tried in [12]. In our implementation, we provide support for the *vector* and the *heap* format, which we explain now in detail.

5.1.3.1 Matrix columns as vectors

In the vector format, a matrix column M_j is represented by a vector v of pairs (i, M_{ij}) , one for each M_{ij} of M_j . The entries are ordered descendingly by the row index i. Each row index occurs at most once in v. The pivot of M_j is the first element in v. As implementation for v, we define the class **SkipVector**, which is a wrapper around **std::vector** that allows for the efficient removal of the pivot.

For coefficients in \mathbf{F}_2 , it suffices to store the indices *i* of the non-zero entries. In this case, the addition of two columns *v* and *v'* is the symmetric difference of two ordered lists and can be performed in time O(|v| + |v'|). We use the STL-implementation $\mathtt{std}::\mathtt{set_symmetric_} \leftrightarrow \mathtt{difference}$ for this. For other coefficient fields, addition of two columns can be performed by an analogous algorithm.

5.1.3.2 Matrix columns as binary heaps

Let T be a set equipped with a total order. A priority queue Q of objects in T is a data structure containing elements of T that allows to efficiently retrieve the maximal element of Q (Top(Q)), remove it (PopMax(Q)), and insert a new element t into Q (Push(Q,t)). We use the STL class std::priority_queue. This class implements a priority queue as a binary heap, which stores its elements in a contiguous vector on which it maintains a specific indexing scheme. In particular, a descendingly ordered vector is also a binary heap.

Algorithm 16: Consolidation of binary heaps

```
      Algorithm 17: Pivot extraction from a binary heap

      Input:
      A binary heap Q of pairs (i, \lambda) \in \mathbb{N} \times k.

      Output:
      The pivot of the matrix column represented by Q.

      if Q = \emptyset then return 0
      (i, \lambda) \leftarrow PopMax(Q)

      while Q \neq \emptyset do
      (i', \lambda') \leftarrow PopMin(Q)

      if i = i' then \lambda \leftarrow \lambda + \lambda'
      else if \lambda = 0 then (i, \lambda) \leftarrow (i', \lambda')

      else
      Push(Q, (i', \lambda'))

      Push(Q, (i, \lambda)))
      return i

      if \lambda \neq 0 then return 0
      i \in [\lambda \neq 0]
```

In heap format, a matrix column M_j is represented by a binary heap Q, whose entries are pairs (i, λ) for $i \in \mathbb{N}$ and $\lambda \in k$. The elements are ordered by i. The heap Q may contain multiple entries (i, λ) for the same row index i. The heap Q represents the matrix column M_j with entries $M_{ij} = \sum_{(i,\lambda) \in Q} \lambda$ for all i.

To retrieve the pivot index of M_j , one has to extract all entries from Q that have the same maximal value of i using PopMax(Q), and sum up their values λ . If they add up to a non-zero value M_{ij} , then this is the pivot entry of M; see Algorithm 17. To add to columns, one merges the representing heaps and reestablishes the heap condition.

We call the heap Q consolidated if it contains at most one entry (i, λ) for each i. A binary heap can be consolidated using Algorithm 16. Algorithm 16 produces the output Q' with entries in total order. In particular, this routine converts the heap representation of a matrix column into the vector representation.

5.1.3.3 Indirect representations

We explain another way to represent matrices, built around the vector representation. Let M be an $m \times n$ -matrix with entries in \mathbf{F}_2 in vector format, where each column M_j is represented by the vector v_j . To efficiently implement column operations on the matrix M, we propose the following way to represent the intermediate steps of a column reduction algorithm.

Algorithm 18: Consolidation of the pivot of an indirect column

▷ entries cancel
 ▷ v_k still has entries
 ▷ v'_{k'} still has entries
 ▷ pivot established
 ▷ push back the last elements

Algorithm 19: Converting an implicit matrix column to an explicit one				
Input:	A binary heap Q of entries (v_{ki}, i, k) .			
Output:	A descendingly ordered vector v' .			
$v' \leftarrow \emptyset$				
while $Q eq \emptyset$ do				
(<i>v</i> _{ki} ,	$i,k) \leftarrow \mathtt{PopMax}(Q)$			
appe	nd v_{ki} to v' .			

An *indirect matrix* I with underlying matrix M consists of a binary heap Q_j for each $j \leq n$. Each Q_j consists of entries (v_{ki}, i, k) with $k \leq n$ and $i \leq |v_k|$. In Q_j , the entries are ordered lexicographically.

Then I represents a matrix product MV for some square matrix V, where the information about a column V_j is contained in the heap Q_j . Specifically, the heap Q_j represents the linear combination

$$I_j = MV_j = \sum_{\substack{(v_{ki}, i, k) \in Q_j}} M_k \tag{5.1}$$

of columns of M. Since we consider coefficients in \mathbf{F}_2 , the linear combination (5.1) corresponds to the symmetric difference of the vectors v_k for $(v_{ki}, i, k) \in Q_j$.

At any time, component i of an entry of Q is chosen such that

$$I_j = \sum_{\substack{(v_{ki}, i, k) \in Q_j \\ l > i}} e_{v_{kl}}.$$

To explain this, assume that Q_j contains two entries (v_{ki}, i, k) and $(v_{k'i'}, i', k')$, such that $v_{ki} = v_{k'i'}$. Then these two entries represent two entries $M_{v_{ki}k}$ and $M_{v_{k'i'}k'}$ of M that lie in the same row $v_{ki} = v_{k'i'}$. When forming forming the symmetric difference (5.1), they add up to zero. Thus, we may remove these entries from Q_j and replace them by $(v_{ki+1}, i+1, k)$ and $(v_{k'i'+1}, i'+1, k')$, given that v_k and $v_{k'}$ have further entries.

We maintain the property that if $(v_{ki}, i, k) \in Q_j$, then all entries $v_{ki'}$ of v_k with i' < i have been canceled with entries of other vectors $v_{k'}$ occurring in Q_j . We also maintain the property that if $Q_j \neq \emptyset$, then there is a unique entry (v_{ki}, i, k) for which v_{ki} becomes maximal. Then this entry represents the pivot of (5.1). We implement the following operations on the columns I_j :

Initialization We initialize I such that I_j represents the column M_j . To do so, we let $Q_j = \{(v_{j1}, 1, j)\}$ for each j; that is, the unique entry in Q_j points to the entry v_{j1} that represents the pivot of M_j .

- **Pivot retrieval** The pivot index piv I_j of a non-zero column I_j is the unique element $(v_{ji}, i, j) \in Q_j$ for which v_{ji} is maximal.
- **Pivot consolidation** When Q_j has been manipulated, Algorithm 18 reestablishes the property that Q_j has a unique entry (v_{ki}, i, k) for which v_{ki} becomes maximal. The algorithm keeps taking elements (v_{ki}, i, k) from Q_j (in descending order of the row indices v_{ki}) as long as there are two entries with the same row index v_{ki} . If entries (v_{ki}, i, k) and $(v_{k'i'}, i', k')$ satisfy $v_{ki} = v_{k'i'}$, they correspond to two entries $M_{v_{ki},k}$ and $M_{v_{k'i'},k'}$ that add to zero in (5.1). If $i < |v_k|$, then the vector v_k representing M_k has further entries $v_{ki'} < v_{ki}$ for i' > i that have not been processed yet. Therefore, we add an entry $(v_{k,i+1}, i+1, k)$ pointing to the next entry of v_k back to Q_j . Analogously for i'.
- **Pivot deletion** To set the pivot entry of I_j to zero, we delete the maximal entry of Q_j and consolidate the pivot (Algorithm 18).
- **Column addition** A column operation $I_j \leftarrow I_j + M_{j'}$ is implemented by adding $(v_{j'1}, 1, j')$ to Q_j . A column operation $I_j \leftarrow I_j + I_{j'}$ is implemented by setting $Q_j \leftarrow Q_j \cup Q_{j'}$. In either case, it is necessary to call Algorithm 18 afterwards to consolidate the pivot.

If there are many column operations of the form $I_j \leftarrow I_j + I_{j'}$, setting $Q_j \leftarrow Q_j \cup Q_{j'}$ may cause Q_j to grow prohibitively large. As a remedy, we propose the following approach inspired by the *accelerated* matrix representations proposed in [12]. During a column reduction scheme, as long as we perform column operations $M_j \leftarrow M_j + M_k$ to the same column M_j , we implement these operations by working on I_j instead. After the last column operation on M_j has happened, we convert the manipulated matrix column I_j to an explicit vector v'_j of the non-zero entries in I_j using Algorithm 19, and set $v_j \leftarrow v'_j$. In other words, only the currently manipulated column of M is represented indirectly, while all others are stored explicitly.

5.1.4 Coning off K_*

Recall that Algorithm 11 requires that the input chain complex C_{\bullet} is eventually acyclic. If K_* does not have this property, we replace $C_{\bullet}(K_*)$ by a suitable filtered complex C_{\bullet} as described in Section 3.7. Implementing the simplicial cone construction from Section 3.7.1 is straight forward, see the class **Cone** in file **Cone.hpp**. In the following, we describe the implementation of the strategy from Section 3.7.2 that can be found in the class **HBasisCone** in the same file. This strategy first computes a persistence bases of the one-parameter persistent homologies $H_{\bullet}(\bigcup_{y \in \mathbf{Z}} K_{(*,y)})$ and $H_{\bullet}(\bigcup_{x \in \mathbf{Z}} K_{(x,*)})$, and then uses these persistence bases to construct the desired chain eventually acyclic chain complex \hat{C}_{\bullet} .

Computing a persistence basis Before we explain how we build \hat{K}_* , we describe an efficient algorithm to compute persistence basis [8, 53].

Let $K_* \in \text{Simp}^{\subseteq \mathbb{Z}}$. For each d, let D_d represent ∂_d with respect to the standard basis. Let V_d be upper triangular such that $R_d = D_d V_d$ is reduced. Assume that $\text{barc } H_d(C_{\bullet}) = \{(b_i, d_i) \mid i \in I\}$ for some index set I. Recall from Example 2.2.7 that

$$\{[R_{d+1}]_j \neq 0\} \cup \{[V_d]_i \mid [R_d]_i = 0 \text{ and } \nexists j : i = \text{piv}[R_{d+1}]_j\}$$

is a persistence basis of $H_d(K_*)$. Recall from Section 2.2.1 that for Vietoris–Rips complexes, computing the matrices R_d and V_d cannot be efficiently combined with clearing, because clearing cannot be applied to the first (viz. highest dimensional) boundary matrix $D_{d_{\text{max}}}$. If the barcode was known, however, clearing can be realized efficiently also with boundary matrices.

The barcode, in turn, can be efficiently computed using cohomology. The computation of a persistence basis of $H_d(K_*)$ thus proceeds in two steps: first, one applies the clearing algorithm (Algorithm 2) to the coboundary matrices $D^d := (D_d)^{\perp}$. From the resulting matrices R^d and V^d one obtains barc $H^d(K, K_*)$ and thus barc $H_d(K_*)$. Second, one uses this barcode to apply

the clearing algorithm to the boundary matrices D_d . This results in the matrices R_d and V_d , from which one obtains the desired persistence basis of the homology modules $H_d(K_*)$. In the following, we explain this algorithm in detail.

Recall the following terminology. If $[R_d]_i = 0$, we call *i* a birth index of $H_d(K_*)$. We call *i* a non-essential birth index if $i = \operatorname{piv}[R_{d+1}]_j$ for some *j*. In this case, we call *j* a death index of $H_d(K_*)$ and (i, j) a persistence pair of $H_d(K_*)$. If there is no such *j*, we call *i* an essential birth index of $H_d(K_*)$. Analogously, if $j = \operatorname{piv}[R^{d+1}]_i$, then (i, j) is called a persistence pair, *j* a non-essential birth index and *i* a death index $H^{d+1}(K, K_*)$. If $[R^{d+1}]_i = 0$ and *i* is not a non-essential birth index, then *i* is an essential birth index of $H^d(K, K_*)$ (sic).

There is a correspondence

$$\left\{ \begin{array}{c} \text{ persistence pairs } (j,i) \text{ of } \\ H^{d+1}(K,K_*) \end{array} \right\} \xrightarrow{1:1} \left\{ \begin{array}{c} \text{ persistence pairs } (i',j') \text{ of } \\ H_d(K_*) \end{array} \right\}$$

$$\left\{ \begin{array}{c} \text{essential birth indices } i \text{ of } \\ H^d(K,K_*) \end{array} \right\} \xrightarrow{1:1} \left\{ \begin{array}{c} \text{essential birth indices } i' \text{ of } \\ H_d(K_*) \end{array} \right\}$$

$$(5.2)$$

where $i' = n_d + 1 - i$ and $j' = n_{d+1} + 1 - j$. This follows from Corollary 2.2.11. However, this is only a correspondence of matrix row and column indices and thus of the intervals in the respective barcodes. There is no correspondence of representative (co)chains.

Example 5.1.1. Let K_* be the simplicial complex that has the (reduced) (co)boundary matrices

$$D_0 = (), \quad D_1 = \begin{pmatrix} 1 & 1 & 0 & 0 & 0 \\ 1 & 0 & 1 & 0 & 1 \\ 0 & 0 & 0 & 1 & 1 \\ 0 & 1 & 1 & 1 & 0 \end{pmatrix}, \quad D_2 = \begin{pmatrix} 1 \\ 1 \\ 0 \\ 1 \\ 1 \end{pmatrix}, \quad D^0 = (), \quad D^1 = \begin{pmatrix} 0 & 1 & 1 & 0 \\ 1 & 1 & 0 & 0 \\ 1 & 0 & 0 & 1 \\ 0 & 0 & 1 & 1 \end{pmatrix}, \quad D^2 = (1 \ 1 \ 0 \ 1 \ 1)$$

with coefficients in \mathbf{F}_2 and $n_0 = 1$, $n_1 = 5$ and $n_2 = 1$. Left-to-right reduction yields

$$R_{0} = (), \quad R_{1} = \begin{pmatrix} 1 & 1 & 0 & 1 & 0 \\ 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 & 0 \end{pmatrix}, \quad R_{2} = \begin{pmatrix} 1 \\ 0 \\ 1 \\ 1 \end{pmatrix}, \quad R^{0} = (), \quad R^{1} = \begin{pmatrix} 0 & 1 & 1 & 0 \\ 1 & 1 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{pmatrix}, \quad R^{2} = (1 \ 0 \ 0 \ 0 \ 0).$$

We obtain the persistence pairs that we list in the following schematic:

$$\begin{array}{c} H_{\bullet}(K_{*}) & H^{\bullet}(K,K_{*}) \\ R_{0} \left\{ \left(\begin{array}{c} (1,\infty) \\ (2,1) \\ (3,4) \\ (4,2) \end{array} \right\} H_{0}(K_{*}) \stackrel{(1:1)}{\leftarrow} \left\{ \begin{array}{c} H^{0}(K,K_{*}) \left\{ \begin{array}{c} (4,\infty) \\ (2,2) \\ (4,1) \\ (3,\infty) \\ (5,1) \end{array} \right\} H_{1}(K_{*}) \stackrel{(1:1)}{\leftarrow} \left\{ \begin{array}{c} (5,3) \\ (2,2) \\ (4,1) \\ (3,\infty) \\ H^{1}(K,K_{*}) \left\{ \begin{array}{c} (5,3) \\ (2,1) \\ (3,\infty) \end{array} \right\} R^{2} \end{array} \right\} R^{2} \end{array} \right\}$$

Here, the essential birth indices i are written as pairs (i, ∞) . The indices of $H_{\bullet}(K_*)$ on the left can be obtained from the reduced matrices R_0 , R_1 on R_2 . Persistence pairs of $H_d(K_*)$ are obtained from R_{d+1} . For the essential birth indices of $H_d(K_*)$, one needs both R_d and R_{d+1} . Red, green and blue numbers are column indices of R_0 , R_1 and R_2 on the left and correspond to the respective row indices of R^0 , R^1 and R^2 on the right.

To compute a persistence basis of $H_d(C_{\bullet})$, we need to compute the reduced columns $[R_{d+1}]_j$, where j ranges over all death indices of $H_d(C_{\bullet})$, and the reduction matrix columns $[V_d]_i$ with $[R_d]_i = 0$, where i ranges over the essential birth indices of $H_d(C_{\bullet})$. This is the essential idea of the following algorithm.

Proposition 5.1.2 (Efficient persistent homology computation). Let the **Z**-graded matrices D_1, D_2, \ldots represent a chain complex $C(K_*)$ of free modules for some $K_* \in \text{Simp}^{\subseteq \mathbf{Z}}$. Then Algorithm 20 computes a persistence basis of the reduced homology $H_d(K_*)$ for each d.

Algorithm 20: One-parameter persistence basis algorithm. Efficient computation of a persistence basis of $H_{\bullet}(K_*)$

Input: **Z**-graded matrices $(D_d \in k^{n_{d-1} \times n_d})_{d \ge 0}$ with row and column grades in ascending order that represent a chain complex C_{\bullet} of free modules.

Output: Persistence bases for each $H_d(C_{\bullet})$.

 $R_0 \leftarrow () \in k^{0 \times n_0}; V_0 \leftarrow E \in k^{n_0 \times n_0}$ $b \leftarrow \{1,\ldots,n_0\}$ \triangleright birth indices of $H_{d+1}(K_*)$ $p \leftarrow () \in \mathbb{N}^0$ ▷ pivot row to column (a) for d = 0, 1... do assignment of R₀ let $R^{d+1} = D_{d+1}^{\perp} V^{d+1}$ be reduced for V^{d+1} invertible upper triangular \triangleright use Algorithm 2 $t \leftarrow \{n_{d+1} + 1 - \mathsf{piv}[R^{d+1}]_i \mid [R^{d+1}]_i \neq 0\}$ \triangleright death indices of $H_d(K_*)$ $b_{\mathsf{e}} \leftarrow \{i \in b \mid [R^{d+1}]_{n_d+1-i} = 0\}$ \triangleright ess. birth indices of $H_d(K_*)$ $b \leftarrow \{j \leq n_{d+1} \mid j \notin t\}$ \triangleright birth indices of $H_{d+1}(K_*)$ for $i \in b_e$ with do \triangleright compute representatives $[V_d]_i$ (b) while $[R_d]_i \neq 0$ do of essential classes in $H_d(K_*)$ $\leftarrow \operatorname{piv}[R_d]_i$ h $[R_d]_i \leftarrow [R_d]_i - [R_d]_{hi}/[R_d]_{hp_h}[R_d]_{p_h}$ $[V_d]_i \leftarrow [V_d]_i - [R_d]_{hi}/[R_d]_{hp_h}[V_d]_{p_h}$ \triangleright *i* is a birth index, so $p_h \neq 0$ $R_{d+1} \leftarrow D_{d+1}$ $V_{d+1} \leftarrow E \in k^{n_{d+1} \times n_{d+1}}$ $\leftarrow 0 \in \mathbf{N}^{n_d}$ for $j \in t$ in asc. order do compute representatives (c) while $p_i \neq 0$ for $j = \text{piv } L_k$ do $[R_{d+1}]_j = D[V_{d+1}]_j$ of $[R_{d+1}]_j \leftarrow [R_{d+1}]_j - [R_{d+1}]_{ij}/[R_{d+1}]_{ip_i}[R_{d+1}]_{p_i}$ persistence pairs of $H_d(K_*)$ $[V_{d+1}]_j \leftarrow [V_{d+1}]_j - [R_{d+1}]_{ij}/[R_{d+1}]_{ip_i}[V_{d+1}]_{p_i}$ $p_i \leftarrow i$ **yield** {($[R_{d+1}]_j, [V_{d+1}]_j$) | $j \in t$ } \cup {[V_d]_i | $i \in b_e$ } \triangleright persistence basis of H_d (d)

Proof. Assume inductively that at the beginning of the loop in line (a), the set b contains the birth indices of $H_d(K_*)$. Assume further that $R_d = D_d V_d$ such that the submatrix $([R_d]_j)_{j \notin b}$ is reduced, and assume that $p_i = j$ for all non-zero columns $[R_d]_j$ with $i = \text{piv}[R_d]_j$. We show that the algorithm correctly computes a persistence basis of $H_d(K_*)$, and establishes the above conditions for the next iteration.

Let $R^{d+1} = D_{d+1}^{\perp} V^{d+1}$ be reduced for V^{d+1} invertible upper triangular. If $[R^{d+1}]_i \neq 0$ with $j = \operatorname{piv}[R^{d+1}]_i$ then (j,i) is a persistence pair of $H^{d+1}(C^{\bullet})$, so $(n_d+1-i, n_{d+1}+1-j)$ is persistence pair of $H_d(K_*)$. In particular, n_d+1-i is a non-essential birth index and $n_{d+1}+1-j$ is a death index of $H_d(K_*)$. Therefore,

$$t = \{n_{d+1} + 1 - \operatorname{piv}[R^{d+1}]_i \mid [R^{d+1}]_i \neq 0\}$$

is the set of death indices of $H_d(K_*)$. On the other hand, if $[R^{d+1}]_i = 0$, then *i* is a birth index of $H^d(C^{\bullet})$. Then *i* is either an essential birth index of $H^d(C^{\bullet})$ (equivalently, $n_d + 1 - i$ is a birth index of $H_d(K_*)$), or a non-essential birth index of $H^d(C^{\bullet})$ (equivalently, $n_d + 1 - i$ is a death index of $H_{d-1}(K_*)$). Since by assumption, *b* contains precisely the birth indices of $H_d(K_*)$, the set

$$b_{e} = \{n_{d} + 1 - i \in b \mid [R^{d+1}]_{i} = 0\}$$

contains precisely the essential birth indices of $H_d(K_*)$. Because an index $j \leq n_{d+1}$ is either a death index of $H_d(K_*)$ or a birth index of $H_{d+1}(K_*)$, setting

$$b = \{j \le n_{d+1} \mid j \notin t\}$$

establishes that b contains precisely the birth indices of $H^{d+1}(K_*)$. To compute a persistence basis of $H_d(K_*)$, we need to compute the reduced matrix column $[R_{d+1}]_j$ if $j \in t$ and the reduction matrix columns $[V_d]_i$ if $i \in b_e$.

The latter happens in line (b), which is the usual reduction scheme. In order to compute the columns $[R_d]_i$ and $[V_d]_i$, it suffices to know the columns $[R_d]_k$ and $[V_d]_k$, where k ranges over the death indices of $H_{d-1}(K_*)$, simply because a reduced matrix R_d would satisfy $[R_d]_k = 0$ if k not a death index. A column index k of R_d is either a death index of $H_{d-1}(K_*)$ or a birth index of $H_d(K_*)$. By assumption, $([R_d]_j)_{j\notin b}$ is reduced, and p contains the corresponding row-to-pivot assignment. Therefore, the loop in line (b) correctly computes $[V_d]_i$ for all $i \in b_e$. Note that for each $i \in b_e$, the while loop inside line (b) terminates with $[R_d]_i = 0$. Analogously, line (c) computes the reduced matrix columns $[R_{d+1}]_j$ and $[V_{d+1}]$ where $j \in t$ ranges over the death indices of $H_d(K_*)$, and thus establishes the induction hypothesis that $([R_{d+1}]_j)_{j\notin b} = D_{d+1}([V_{d+1}]_j)_{j\notin b}$ be reduced and p contains the corresponding pivot assignment.

Thus, when the main loop reaches line (d), the hypotheses for the next iteration are satisfied, $\{[R_{d+1}]_j, [V_{d+1}]_j \mid j \in t\}$ is a system of representatives of the finite bars in barc $H_d(K_*)$, and $\{[V_d]_i \mid i \in b_e\}$ is a system of representatives of the infinite bars in barc $H_d(K_*)$.

To obtain matrices R^{d+1} and V^{d+1} such that $R^{d+1} = D^{d+1}V^{d+1}$ is reduced, one may use the Standard Algorithm with clearing (Algorithm 2). As remarked earlier, most birth columns of the boundary matrices of a Vietoris–Rips complex are non-essential and account for the largest fraction of the runtime of Algorithm 1. Because Algorithm 20 skips reduction of these columns, it computes a persistence basis much faster than Algorithm 1 in practice.

Remark. The construction generalizes immediately to the situation of computing a persistence basis of $H_{\bullet}(C_{\bullet})$, where C_{\bullet} is a chain complex of finite rank free **Z**-persistence modules. In this case, one replaces $H^{\bullet}(K, K_*)$ by $H^{\bullet}(\nu C_{\bullet})$, where ν is the Nakayama functor from Definition 3.2.2. This makes sense, because if $K_* \in \text{Simp}^{\subseteq \mathbf{Z}}$, then $C^{\bullet}(K, K_*) = \nu C_{\bullet}(K_*)$.

Constructing the eventually acyclic complex \hat{C}_{\bullet} Let $C_{\bullet} = \bigoplus_{l=1}^{N} F(z_l)$ be a chain complex of finite rank free \mathbb{Z}^2 -modules as above, and let $z_0 > z_l$ for all l. We use Algorithm 20 to build a chain complex \hat{C}_{\bullet} with $(\hat{C}_{\bullet})_z = C_z$ if z < z and $H_{\bullet}(\hat{C}_{\bullet})_z = 0$ otherwise.

Let $z_l = (x_l, y_l)$ for all l and $z_0 = (x_0, y_0)$, and consider the chain complex $C'_{\bullet} = \operatorname{colim}_y C_{\bullet} = \bigoplus_{i=1}^N F(x_i)$ of free one-parameter persistence modules. For each $d \ge 0$, let $\operatorname{barc} H_d(C_{\bullet}) = \{(b_{d,i}, d_{q_i}) \mid i \in I_d\}$ for some indexing set I_d . Let $I_d^{\mathrm{f}} \coloneqq \{i \in I \mid d_{d,i} < \infty\}$ and $I_d^{\mathrm{e}} = I_d \setminus I_d^{\mathrm{f}}$. We use Algorithm 20 to compute a persistence basis

$$\{(z_{d,i} = \partial c_{d,i}, c_{d,i}) \mid i \in I_d^{\rm f}\} \cup \{z_{d,i} \mid i \in I_d^{\rm e}\}$$

of $H_d(C_{\bullet})$ as in Definition 2.2.6. For each $i \in I_d^{\mathrm{f}}$, define the chain complex

$$\Delta_{d,i}: \qquad \cdots \longrightarrow 0 \longrightarrow F(d_{d,i}, y_0) \stackrel{1}{\longrightarrow} (b_{d,i}, y_0) \longrightarrow 0 \longrightarrow \cdots$$

concentrated in degrees d, d + 1. There is morphism of complexes

$$\begin{array}{ccc} \Delta_{d,i} \colon & \cdots \longrightarrow 0 \longrightarrow F(d_{d,i},y_0) \xrightarrow{1} F(b_{d,i},y_0) \longrightarrow 0 \longrightarrow \cdots \\ & & & & \\ C_{\bullet} \colon & \cdots \longrightarrow C_{d+2} \longrightarrow C_{d+1} \longrightarrow C_d \longrightarrow C_{d-1} \longrightarrow \cdots \end{array}$$

Analogously, if $i \in I_d^e$, define the chain complex

 $\Delta_{d,i}: \qquad \cdots \longrightarrow 0 \longrightarrow 0 \longrightarrow (b_{d,i}, y_0) \longrightarrow 0 \longrightarrow \cdots$

and the morphism

$$\begin{array}{cccc} \Delta_{d,i} \colon & \cdots \longrightarrow 0 \longrightarrow 0 \longrightarrow F(b_{d,i},y_0) \longrightarrow 0 \longrightarrow \cdots \\ & & \downarrow & \downarrow^{z_{d,i}} \\ C_{\bullet} \colon & \cdots \rightarrow C_{d+2} \rightarrow C_{d+1} \longrightarrow C_{d} \longrightarrow C_{d-1} \rightarrow \cdots . \end{array}$$

Then we get a morphism

$$e \colon \bigoplus_{d \ge 0, i \in I_d} \Delta_{d,i} \to C_{\bullet}$$

and we let $\hat{C}^x_{\bullet} := \operatorname{cone} e$. To make this explicit, define for each $d \ge 0$ the free module

$$G_d \coloneqq \langle \hat{z}_{d-1,i} \mid i \in I_{d-1} \rangle \oplus \langle \hat{c}_{d-2,i} \mid i \in I_{d-2}^{\mathrm{f}} \rangle$$

spanned by symbols $\hat{z}_{d,i}$ and $\hat{c}_{d-1,i}$ of grades $g(\hat{z}_{d,i}) = (b_{d,i}, y_0)$ and $g(\hat{c}_{d-1,i}) = (d_{d-1,i}, y_0)$. Then

$$\hat{C}_{\bullet} = \begin{pmatrix} \cdots \to C_{d+2} \oplus G_{d+2} \longrightarrow C_{d+1} \oplus G_{d+1} \longrightarrow C_d \oplus G_d \to \cdots \\ \hat{c}_{d-2,i} \longmapsto c_{d-2,i} - \hat{z}_{d-2,i} \\ \hat{z}_{d-2,i} \longmapsto z_{d-2,i}, \end{pmatrix}.$$

Repeating this process with the roles of x and y exchanged yields a chain complex \hat{C}_{\bullet} with $\hat{C}_z = C_z$ if $z < z_0$ and $H_{\bullet}(\hat{C}_{\bullet})_z = 0$ otherwise.

5.1.5 Sparsification

We observe that in Algorithm 10, the second for-loop takes orders of magnitude longer than the first loop. Furthermore, it has the tendency to increase the sparsity of the matrix considerably, which may incur high costs on all further operations on these matrices. In this section, we give a heuristic explanation, and devise a partial remedy.

Explanation Let K_* be a \mathbb{Z}^2 -filtered simplicial complex and D^{d+1} represent its coboundary matrices with respect to the standard basis. We let the simplices of K_* be enumerated colexicographically, such that

$$K_* = \{\sigma_1, \ldots, \sigma_n\}$$
 with $g(\sigma_1) \preceq_{\text{colex}} \cdots \preceq_{\text{colex}} g(\sigma_n).$

We obtain a **Z**-filtered complex K'_* with $K'_z = \{\sigma_i \in K \mid i \leq z\}$. It has the same coboundary matrices as K_* , including row and column order. The first for-loop in Algorithm 10 can be seen as a usual one-parameter persistence left-to-right reduction scheme, preceded by re-ordering the columns of D^{d+1} by their colex grades. Recall that Algorithm 10 is invoked after performing a clearing step; see Algorithm 11.

Recall that in one-parameter persistence, if applied to a coboundary matrix D, the standard algorithm spends most of its runtime on the columns it reduces to zero (i.e., birth columns). Consequently, clearing remedies this by skipping the reduction of most of the birth columns. However, this is only true if the matrix being reduced comes from a filtration, as this imposes a certain block upper triangular shape on the matrix. Therefore, apart from the birth columns, the matrix is "almost" reduced already at the beginning.

This also holds for coboundary matrices arising from two-parameter filtrations. To illustrate this, consider the following example. Let the different grades $g_1 \leq_{\text{colex}} \cdots \leq_{\text{colex}} g_9$ that occur as row or column grades of a valid \mathbb{Z}^2 -graded matrix D be arranged as in the following schematic.

Then validity of D imposes the following block structure on D, where empty blocks are zero:

$\begin{bmatrix} g_7 & g_8 & g_9 \end{bmatrix}$	$\left. \begin{array}{c} g_1 \ g_2 \ g_3 \ g_4 \ g_2 \ g_2 \ & * \ & & * \ & * \ & * \ & * \ & * \ & * \ & * \ & * \ & * \ & * \ & * \ $	^{g5} * \
$egin{array}{c c} g_4 & g_5 & g_6 \end{array}$	$D = \frac{g_4}{g_5}$ * *	*
$egin{array}{c c} g_1 & g_2 & g_3 \end{array}$: (·)

Of course, block rows or columns may be empty, in case D has no rows or columns of that grade. Nevertheless, validity of D imposes that D has the above block upper triangle shape. Ordering the rows of D by their lex-pivots does not change this. To see this, note that in the above example, we obtain

	g_1	g_4	g_7	g_2	g_5	g_8	g_3	g_6	g_9	
g_1	/*	*	*	*	*	*	*	*	* \	
\overline{g}_2	1			*	*	*	*	*	*	۱.
\bar{q}_3	1							*	*	۱.
q_{Δ}		*	*		*	*		*	*	
q_5					*	*		*	*	L
q_6								*	*	1
a7	1		*			*			*	L
a _s	1					*			*	1
g_9	1								*/	<u></u>

where all empty entries are zero. Although the matrix is not in block upper triangular shape anymore, we see that to the left of the column column g_j , which has block pivot g_j , no other column has an entry in that block row. This ensures that despite the re-ordering, reduction is fast.

For the second for-loop, which can be seen as a left-to-right reduction scheme, preceded by reordering the rows lexicographically by grade and the columns by colex-pivot, the performance drastically changes. We assume that this reordering renders D the coboundary matrix of a randomly ordered simplicial complex. It is known reducing the (co)boundary matrix of a simplicial complex with simplices in random order takes much longer in practice than reducing the (co)boundary matrix of a **Z**-filtered complex [17, 78].

To make a concrete example, consider the following numbers. We have run our algorithm on the coboundary matrices of the full density-Rips complex of 150 points sampled from the orthogonal group O(5), embedded in \mathbb{R}^{25} . To compute $H^3(N^{\bullet}(K_*))$, we have to apply Algorithm 10 to the coboundary matrix D^{d+2} of N^{\bullet} , which is a graded 551, 300 × 11, 175. After the first for loop of Algorithm 10, which takes 5 ms, the densest columns have up to 1,456 entries in the heaps representing the columns. After the second, which takes 10,350 ms, the densest columns have up to 78,875,970 entries in the heaps.¹ This is a somewhat extreme example, but illustrates the problem. This implies that all further matrix computations with this matrix (namely, minimization and cokernel computation) will be expensive.

Sparsification We provide a partial remedy for this. Recall that our goal is to compute the submodule $[V]_F$ of a free module $F = \bigoplus_{i=1}^n F(z_i)$, where V is a subspace of colim F. The space V is given by a matrix B representing a generating system of V with respect to the standard basis of colim F. Then Algorithm 10 computes a graded matrix B' = Bireduce(B) that represents a basis of the free module $[V]_F$ with respect to the standard basis of F.

However, in Algorithm 11, which is where Algorithm 10 is invoked, we do not require that B' represents a basis of $[V]_M$ with respect to the standard basis of F. Instead, it suffices if B' represents a basis of $[V]_F$ with respect to any basis of F; see also Corollary 3.5.1. We may, therefore, choose the basis of F that suits our task best; i.e., for which B' is particularly sparse.

Definition 5.1.3. A non-zero entry of a graded matrix M is *dominant* if it is the only non-zero entry in its row that is not dominated. A non-zero entry M_{ij} is *dominated* if there is a dominant entry M_{kj} in the same column, such that $\operatorname{rg}_i^M \leq \operatorname{rg}_k^M$.

¹Note that this shows a particular shortcoming of the heap representation: namely, using 78,875,970 to represent a matrix column with 551,300 entries is certainly not sparse. With the vector representation, we obtain up to 1,644 non-zero entries in one column after the first for-loop (3 ms) and up to 141,026 non-zero entries after the second (971 ms). Compare also the results from [17].

Remark (Base changes of \mathbb{Z}^n -graded modules). Let $M: F \to E$ be morphism of finite rank free modules. Let M be the valid graded matrix representing M with respect to the homogeneous basis f_1, \ldots, f_m of F and e_1, \ldots, e_n of E.

- If $g(f_i) \leq g(f_j)$, then replacing f_j by $f'_j = f_j + \lambda F_{g(f_j),g(f_i)}(f_i)$ defines another basis of F. The matrix M' representing M with respect to this basis is obtained by replacing the column M_j by $M'_j \coloneqq M_j + \lambda M_{*i}$.
- If g(e_i) ≤ g(e_j), then replacing e_j by e'_j = e_j + λF_{g(e_j),g(e_i)}(e_i) defines another basis of E. The matrix M' representing M with respect to this basis is obtained by replacing the row M_{i*} by M'_{i*} = M_{i*} λM_{j*}.

If M_{ij} is dominated by M_{kj} , then M_{ij} can be erased by the row operation $M_{i*} \leftarrow M_{i*} - \frac{M_{ij}}{M_{kj}}M_{k*}$, and this row operation only changes the entry M_{ij} . Such eliminations of dominated entries using dominating ones are therefore compatible with column sparse matrix formats. *Example.* Consider the graded matrix

The entry $M_{6,1}$ is dominant. The row operation $M_{5*} \leftarrow M_{5,*} - M_{6,*}$ is valid because $b \leq a$, and results in setting the entry $M_{5,1}$ to zero. After the operation, also $M_{5,2}$ becomes dominant, and we perform row operation $M_{4,*} \leftarrow M_{4,*} - M_{5,*}$, which sets the entry $M_{4,2}$ to zero. No entry of $M_{3,*}$ can be removed, so the two non-zero entries in that row are not dominant. In particular, they cannot be used to set the row $M_{2,*}$ to zero without a proper row operation. Instead, the row operation $M_{2,*} \leftarrow M_{2,*} - M_{5,*}$ sets $M_{2,5}$ to zero and leaves the dominant entry $M_{2,1}$. This can be used for the last operation $M_{1,*} \leftarrow M_{1,*} - M_{2,*}$. We obtain the following sequence of row operations from bottom to top:

The struck through entries are set to zero in the course of the operation. Of the remaining entries, the ones printed in boldface are dominant.

We obtain the following algorithm to eliminate dominated matrix entries. We first describe the algorithm on a high level (Algorithm 21) before we describe a more implementation-oriented formulation (Algorithm 22).

Proposition 5.1.4. Let M be a valid graded matrix such that $rg_i^M \not\geq rg_j^M$ if $i \leq j$. Then Algorithm 21 computes a graded matrix M' with the same row and column grades as M, such that there is an invertible valid graded matrix V with M' = MV by eliminating all possible dominated entries of M.

Proof. During the main for-loop, the algorithm maintains the property that d_j contains the indices i such that $M_{ij} \neq 0$ is dominant. During each iteration, it maintains the property that u contains the column indices j of the entries in row M_{i*} that are not dominated. Assume that $M_{ij} \neq 0$ and that there exists a $h \in d_j$ such that $\operatorname{rg}_i^M \leq \operatorname{rg}_h^M$. Then the row operation $M_{i*} \leftarrow M_{i*} - \frac{M_{ij}}{M_{h*}} M_{h*}$ is a valid base change and sets the entry M_{ij} to zero. Since M_{hj} is a

Algorithm 21: Elimination of	of do	ominated	entries
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Input: A graded $m \times n$ -matrix M such that $rg_i^M \ge rg_j^M$ if $i \le j$. **Output:** A graded matrix M' = VM for some valid invertible upper triangular matrix V, such that M' contains not more non-zero entries than M

$d_j \leftarrow \emptyset$ for all j for $i = m, \dots, 1$ do	row indices of dominant entries in column j
$\begin{vmatrix} u \leftarrow \emptyset \\ \text{for } j = 1, \dots, n \text{ with } M_{ij} \neq 0 \text{ do} \\ \text{ if } \exists h \in d_j : \operatorname{rg}_i^M \leq \operatorname{rg}_h^M \text{ then} \\ M_{ij} \leftarrow 0 \end{vmatrix}$	$\triangleright \exists$ a dominant entry M_{hj}
$ \begin{array}{c} else \\ M'_{ij} \leftarrow M_{ij} \\ u \leftarrow u \cup \{j\} \\ if u = \{j\} \text{ for some } j \text{ then} \\ d_j \leftarrow d_j \cup \{i\} \\ \end{array} $	⊳ <i>M_{ij} is dominant</i>

dominant entry, no other row entry of M_{i*} is affected by this row operation. Therefore, the row operation can be realized by just setting $M_{ij} \leftarrow 0$. If there is no such h, then M_{ij} is not dominated. If $u = \{j\}$ and the end of the *i*th iteration, then M_{ij} is the only non-zero entry in M_{i*} that is not dominated, it is dominant by definition.

To describe a possible implementation of this algorithm, we assume that we are using the vector representation of matrices; see Section 5.1.3.1. In case we are using heap representations, one can use consolidation (Algorithm 16) to convert a heap-representation to a vector-representation of the same matrix. For simplicity, we describe the following refinement of Algorithm 21 only for matrices with coefficients in \mathbf{F}_2 ; the approach generalizes to arbitrary coefficient fields. Let the rows of M be in colexicographic² order with respect to their grades, and let (v_1, \ldots, v_n) be the vectors representing M, where each v_j contains the row indices of the non-zero entries in M_j in descending order.

Proposition 5.1.5. In the above situation, Algorithm 22 computes the same valid graded matrix M' as Algorithm 21.

Proof. We show that Algorithm 22 performs the same row operations as Algorithm 21. The central data structure is the priority queue Q. Its entries are triples (λ, k, j) with $j \leq n$ and $k \leq |v_j|$ and $\lambda = v_{j,k}$ is the row grade of the last but kth non-zero entry in M_j . Thus, an entry in Q represents the non-zero entry $M_{v_{k,j}j}$. The priority queue Q is ordered lexicographically.

Claim 2. Repeatedly calling PopPush(Q) retrieves all indices (i, j) of non-zero entries of M, in order of descending row indices i.

Proof of claim. The function PopPush(Q) retrieves the maximal (i, k, j) from Q. Because Q is ordered by the row indices $i = v_{jk}$ first, popping the maximal elements from Q thus retrieves the elements in reverse order of the row grades. To see that the function retrieves all indices of non-zero entries of M, we argue as follows. Recall that v_j is descendingly sorted. Thus, if $k = |v_j|$, then i is the smallest row index of a non-zero entry in M_j . Therefore, it is not necessary to process any further entries of v_k . Otherwise, v_j contains more entries i' corresponding nonzero entries $M_{i'j}$ with i' < i. Pushing $((v_j)_{k+1}, k+1, j)$ to Q ensures that the non-zero entry $M_{i',j}$ with the largest row index $i' = (v_j)_{k+1}$ less than i is processed as the next entry from M_j . Therefore, PopPush(Q) eventually yields all non-zero entries of M.

Claim 3. If an entry M_{ij} is dominated, then it is dominated by the dominant entry M_{lj} with the least row index l > i.

²or lexicographic; adapt algorithm and proof accordingly.

Algorithm 22: Possible implementation of Algorithm 21. Input: Descendingly ordered vectors v_1, \ldots, v_n representing a reduced valid graded $m \times n$ -matrix Mwith entries in \mathbf{F}_2 , with rg^M ordered colexicographically. **Output**: Descendingly ordered vectors v'_1, \ldots, v'_n representing M' = VM as in Algorithm 21. for $j = 1, \ldots, n$ do $v_j \leftarrow \emptyset$ \triangleright vectors representing M'_i for $j = 1, \ldots, n$ do $d_j \leftarrow -1$ ▷ row index of last dominating $Q \leftarrow \{((v_j)_0, 0, j) \mid M_{ij} \neq 0\}$ as lex. ordered priority queue entry in Ma (a) function PopPush(): \triangleright retrieves next entry from Q $i, k, j \leftarrow \texttt{PopMax}(Q)$ if $k < |v_j|$ then Push $(Q, ((v_j)_{k+1}, k+1, j))$ return i, j $i, k, j \leftarrow \texttt{PopPush}(Q)$ ▷ largest pivot of M append *i* to v'_i $r \leftarrow 1$ number of entries not while $Q \neq do$ dominated in ith row. $i', j' \leftarrow \texttt{PopPush}(Q)$ (b) if $i \neq i'$ then if r = 1 then $d_i \leftarrow i$ \triangleright M_{ij} is dominating $i \leftarrow i'$ $j \leftarrow j'$ $r \leftarrow 0$ if $d_j = -1$ or $rg_{i'}^M \nleq rg_{d_{i'}}^M$ then $\triangleright M_{i'i'}$ is not dominated (c) append i' to $v_{i'}$ $i \leftarrow i'$ $j \leftarrow j'$ $r \leftarrow r + 1$

Proof of claim. Assume that M_{ij} is dominated by some non-zero entry M_{kj} . This implies that i < k and $\operatorname{rg}_i^M \leq \operatorname{rg}_k^M$. Recall that $z \leq z'$ in \mathbb{Z}^2 if and only if $z \preceq_{\operatorname{lex}} z'$ and $z \preceq_{\operatorname{colex}} z'$. In particular, $\operatorname{rg}_i^M \preceq_{\operatorname{colex}} \operatorname{rg}_k^M$ and $\operatorname{rg}_i^M \preceq_{\operatorname{lex}} \operatorname{rg}_k^M$. Let l be the least row index l > i of a dominant entry in column M_j . Then $\operatorname{rg}_i^M \preceq_{\operatorname{colex}} \operatorname{rg}_l^M \preceq_{\operatorname{colex}} \operatorname{rg}_k^M$ since by assumption, the rows of M are ordered colexicographically by grade. Two distinct dominant entries necessarily have incomparable row grades, we get $\operatorname{rg}_i^M \preceq_{\operatorname{lex}} \operatorname{rg}_k^M \preceq_{\operatorname{lex}} \operatorname{rg}_l^M$. Together, this shows that $\operatorname{rg}_i^M \leq \operatorname{rg}_l^M$.

The variables i and j are initialized such that M_{ij} is the pivot of M with the largest row index. Assume by induction that at the beginning of an iteration of the main while-loop, i and j are the row and column indices of the last processed non-zero entry M_{ij} , that r is the number of non-zero entries in row M_{i*} not eliminated, and that d_j contains the row index of the last processed dominating entry in column M_j (or -1, if there is none). Let i' and j' be the next entry from Q as in line (b).

If $i \neq i'$, then all entries in row M_{i*} have been processed. If r = 1, there has been a single non-zero entry M_{ij} in row *i*. It is therefore dominant, which justifies setting $d_j \leftarrow i$. Since there have not been any previously processed entries in row *i'*, the algorithm sets $r \leftarrow 0$, which reestablishes the induction hypothesis.

We have to check if the new entry $M_{i'j'}$ is dominated. By Claim 3, this is the case if and only if it is dominated by the last dominating entry $M_{d_jj'}$; i.e., if $\operatorname{rg}_i^M \leq \operatorname{rg}_{d_j}^M$, given that there exists any previous dominating entry in $M_{j'}$. In this case, the algorithm simply discards the entry, which corresponds to setting ot to zero. Otherwise (line (c)), the entry is kept unchanged, which corresponds to appending its row index to $v'_{j'}$. Setting $i \leftarrow i', j \leftarrow j'$ and $r \leftarrow r + 1$ reestablishes the induction hypothesis. Note that the entries of v'_j are constructed in descending order, as required by the vector representation.

5.1.6 Chunk(*) preprocessing

As mentioned before, chunk preprocessing is an efficient preprocessing scheme that reduces the size of the input chain complex without changing its homology isomorphism type. More concretely, given chain complex C_{\bullet} of finite rank free two-parameter persistence modules concentrated in degrees ≥ 0 , chunk preprocessing computes a minimal chain complex C'_{\bullet} homotopy equivalent to C_{\bullet} . Assuming that C_{\bullet} is given as a sequence D_{\bullet} of boundary matrices, chunk preprocessing applies minimization (Algorithm 3) to all matrices D_{\bullet} ; see Algorithm 4.

For chain complexes C_{\bullet} arising from filtered simplicial complexes in practice, C'_{\bullet} typically has much smaller rank than C_{\bullet} . It has been observed that when computing a minimal free resolution of $H_d(C_{\bullet})$ using the LW-algorithm, applying chunk preprocessing drastically improves the efficiency of the entire pipeline [73, table 3]. A particular reason for this is that, while there is no efficient way known to parallelize the LW-algorithm (Algorithm 6), Algorithm 3 can be implemented in an embarrassingly parallel way; see Section 2.4.2.

For full function Rips-complexes (such as density-Rips), it has been observed that the boundary matrices D'_{\bullet} of the the chain complex C'_{\bullet} computed by chunk preprocessing computes, although having fewer rows and columns than D_{\bullet} , have more non-zero entries; see [73, table 2] Nevertheless, applying chunk preprocessing before running the LW-algorithm reduces the runtime.

Chunk* preprocessing Algorithm 3, which underlies chunk preprocessing, minimizes each boundary matrix D_d through left-to-right column additions. Alternatively, we propose to minimize D_d through row operations. Equivalently, this corresponds to applying Algorithm 3 to the transpose D^d of D_d . We thus propose the following preprocessing scheme, which we refer to as chunk* preprocessing: we apply Algorithm 3 to all coboundary matrices D^d , starting with D^1 . To compute a minimal free resolution of homology, we transpose the result; see Algorithm 5. We expect that for boundary matrices with much more rows than columns, such as the boundary matrices of Vietoris–Rips complexes, the performance may differ.

5.2 Experiments

In this section, we evaluate the performance of the cohomology algorithm Algorithm 11 using our implementation 2pac [94]. To do so, we compare the performance of our implementation of the cohomology algorithm (Algorithm 11) and the homology algorithm (Algorithm 9). Although the latter is implemented in the state-of-the-art software mpfree [85], we use our own implementation of it. We do this in order to be able to explore the influence of implementation details, such as the matrix representation or the specific preprocessing. This is justified by Table 1, which shows that mpfree and the version of 2pac that uses the vector representation have similar runtimes.

5.2.1 Samples

Although our software can be used to compute minimal free resolutions of the (co)homology of any chain complex of finite rank free two-parameter persistence modules, our main motivation for computing cohomology are two-parameter Rips complexes. This is because in oneparameter persistence, cohomology (with clearing) has increased the practically tractable size of Vietoris–Rips complexes drastically. Therefore, we focus on function-Rips complexes in the experimental evaluation. We have run our experiments on (subsamples of) the following datasets, which have been generated using the Jupyter notebook EXAMPLE.ipynb in the repository. The data sets, which consists of the filtration values for the vertices and the distance matrix, are

outliers	density function	bandwidth
$[-2, 2]^{\times 2}$	Gaussian	0.25
$[-1.5, 1.5]^{\times 3}$	Gaussian	0.30
$[-1.25, 1.25]^{\times 4}$	Gaussian	0.35
$[-1, 1]^{\times 5}$	Gaussian	0.40
$[-1.25, 1.25]^{\times 9}$	Gaussian	0.60
$[-1.25, 1.25]^{ imes 16}$	Gaussian	0.85
$[-1.25, 1.25]^{\times 25}$	Gaussian	0.85
$[-1.5, 1.5]^{\times 4}$	Gaussian	0.35
$[-1.5, 1.5]^{\times 6}$	Gaussian	0.50
[-1.25, 1.25] ^{×8}	Gaussian	0.70
-	-	
	$\begin{array}{c} \text{outliers} \\ [-2,2]^{\times 2} \\ [-1.5,1.5]^{\times 3} \\ [-1.25,1.25]^{\times 4} \\ [-1,1]^{\times 5} \\ [-1.25,1.25]^{\times 9} \\ [-1.25,1.25]^{\times 16} \\ [-1.25,1.25]^{\times 25} \\ [-1.5,1.5]^{\times 4} \\ [-1.5,1.5]^{\times 6} \\ [-1.25,1.25]^{\times 8} \end{array}$	outliersdensity function $[-2,2]^{\times 2}$ Gaussian $[-1.5, 1.5]^{\times 3}$ Gaussian $[-1.25, 1.25]^{\times 4}$ Gaussian $[-1,1]^{\times 5}$ Gaussian $[-1.25, 1.25]^{\times 9}$ Gaussian $[-1.25, 1.25]^{\times 16}$ Gaussian $[-1.25, 1.5]^{\times 25}$ Gaussian $[-1.5, 1.5]^{\times 4}$ Gaussian $[-1.5, 1.5]^{\times 6}$ Gaussian $[-1.25, 1.25]^{\times 8}$ Gaussian

Table 5.1: Data sets used in the experiments. The first four samples (spheres) consist of 200 points sampled from the actual space, and 100 points sampled from the space described in column "outliers". For the orthogonal groups and tori, we sampled 400 points from the space and 200 outliers. For details about the last one, see the text.

generated prior to to the experiments, written to disk, and then read by 2pac during the experiments to generate the function-Rips complexes. The precise code to reproduce the data files can be found in Generate Samples.ipynb.

Density-Rips complexes We have sampled points from spheres $S^n \subseteq \mathbf{R}^{n+1}$, tori $T^n := (S^1)^n \subseteq \mathbf{R}^{2n}$ and orthogonal groups; see Table 5.1. Arguably, the most common density functions $\rho: S \to \mathbf{R}$ used in persistent homology are

• the Gaussian density function

$$p\mapsto \sum_{q\in S\backslash\{p\}}\exp(\frac{-d(p,q)^2}{2\sigma^2}),$$

• the ball density function

$$p \mapsto |\{q \in S \setminus p \mid d(p,q) \le \sigma\}|,$$

• and the kth nearest neighbor density function $p \mapsto d(p,q)$, where $q \in S$ satisfies

 $|\{r \in S \setminus \{p\} \mid d(p,r) < d(p,q)\}| < k.$

Each of these relies on the choice of a bandwidth parameter $\sigma \in \mathbf{R}$ or $k \in \mathbf{N}$. We observed that usually, the Gaussian density function achieves the best distinction between sample and outliers. Using the notebook **EXAMPLE.ipynb**, we manually chose the bandwidth parameter σ for each of the data sets in Table 5.1. A projection onto the first two coordinates of \mathbf{R}^n of each of the function-Rips data sets is shown in Figure 5.1.

Cyclooctane (C_8H_{16}) The cyclooctane data set, which can be obtained from [101], is sampled from the configuration space of the cyclooctane molecule. More accurately, the sample is taken from the configuration space of a linkage of eight rigid rods of equal length, with fixed equal angles of approximately 118° at the eight joints. Because of this angle, the linkage forms no planar octagon but has to "pucker" in space. In particular, three consecutive edges need not lie in the same plane.

Each of the approx. 6000 entries in the data set consists of the 24 coordinates of the eight vertices in \mathbf{R}^3 . The data set has previously been studied in persistence [101, 102, 103]. It has been suggested that the described configuration space forms a two-dimensional subspace of \mathbf{R}^{24} ;



Figure 5.1: The plots show projections of the datasets on the first two coordinates of the ambient Euclidean space. The colors encode the values of the density function. All datasets use a Gaussian density function, with the bandwidth parameters from Table 5.1.



Figure 5.2: Isomap projection [121] of the cyclooctane data set from $\mathbb{R}^2 4$ to \mathbb{R}^3 with respect to the Euclidean distance. The color encodes the metric distortion of the projection (red=high distortion). The component appearing as an hourglass is a projection of the Klein Bottle.



Figure 5.3: Cyclooctane data set. The plot shows the first two torsion angles of the data points. Points in the diagonal of the plot correspond to points from the spherical component of the configuration space. Points from the actual space are printed with circles, outliers with diamonds.

namely, a union of a 2-sphere and a Klein bottle that intersect in two disjoint 1-spheres. For an illustration, see the projection to \mathbf{R}^3 in Figure 5.2, which has been obtained with the Isomap algorithm [121].

We equip the sample with the following distance. To each consecutive four vertices v_1 , v_2 , v_3 and v_4 , we assign the signed angle enclosed by the affine plane spanned by v_1 , v_2 and v_3 and the affine plane spanned by v_2 , v_3 and v_4 . This angle is also called the *torsion angle* of the edge v_2v_3 . This gives a vector of eight oriented angles. As the distance between two configurations, we use the Euclidean distance between their torsion angle vectors.

We equip the vertices with the function $(\alpha_1, \ldots, \alpha_8) \mapsto |(\alpha_1 - \alpha_5, \ldots, \alpha_4 - \alpha_8)|$. The reason for this choice is that as mentioned above, the configuration space of C_8H_{16} has been described as the union of a sphere and a Klein Bottle, intersecting in two circles [102]. We have observed [95] that configurations from the spherical component have the property $\alpha_i = \alpha_{i+4}$ for all $i \leq 4$. This is not the case in general for configurations from the Klein bottle component. Therefore, the above function measures the deviation from this symmetry.

5.2.2 Parameters

Our main interest lies in comparing the following three approaches:

• the cohomology algorithm (Algorithm 11) called by

./2pac -c -cone=2 -f[dataset],

• the homology algorithm with chunk preprocessing (Algorithms 4 and 9), called by

./2pac -nn -hDC(n+1) -f[dataset]),

• the homology algorithm with chunk* preprocessing (Algorithms 5 and 9), called

./2pac -hC(n+1) -f[dataset].

Here *n* denotes the dimension up to which one wishes to compute a free resolution of H_n . These program calls generate the (co)boundary matrices of the full function-Rips complex K_* of the respective data set, and compute a minimal free resolution of $H_d(K_*)$ (resp., $H^{d+2}(N^{\bullet}(K_*))$) for $d = 0, \ldots, n$. In the case of homology, passing -DC(n + 1) (resp. -C(n + 1)) applies chunk (resp. chunk*) preprocessing to all boundary matrices D_1, \ldots, D_{n+1} .

Remark (mpfree). The second approach listed above is the same that is implemented in mpfree. As mentioned at the beginning of this section, use our own implementation to be able to vary implementation details such as the underlying matrix type; see Table 1 for a runtime comparison. A conceptual difference is that mpfree is designed for computing a minimal free resolution of homology for a single degree, while 2pac is designed to produce a sequence of of resolutions of homology in ascending dimensions. This has the effect that 2pac reuses some data used in the resolution computation in lower homology dimensions (e.g., chunk preprocessing).

We found that this is the best choice for chunk preprocessing: applying minimization only to the first n or less boundary matrices (-Cn) does not reduce the runtime of the following homology computation considerably, while minimizing the (n + 2)nd boundary matrix, despite reducing the runtime of the homology algorithm drastically, is too expensive.

For the cohomology algorithm, one has to replace the input complex by an eventually acyclic one, because the function-Rips filtration is not eventually acyclic with regard to the function-parameter. To do this, we pass -cone=2 when computing cohomology. This builds the eventually acyclic chain complex \hat{C}_{\bullet} as described in Section 5.1.4. Alternatively, one can pass -cone=1 to construct \hat{C}_{\bullet} as the cone of $C_{\bullet}(K_*)$; see Section 3.7.1. If not stated otherwise, we use -cone=2, as we have observed this usually yields better results than -cone=1; see Section 5.3.2 below.

Remark. If the input complex is not eventually acyclic also for the second filtration parameter, one has to also pass -Cone=1 or -Cone=2. This is the case, for example, for truncated function-Rips complexes.

In our experiments, we investigate the effect of

- the size of a random subsample taken from the input. The subsample size to take can be specified by passing -s; this chooses a random subsample with a fixed seed to ensure reproducibility;
- varying the dimension up to which we compute (co)homology (using -n);
- the matrix representation (matrix or heap representation; see Section 5.1.3); to change this, run either ./2pac (for heaps) or ./2pac.vectors;
- the strategy to cone off K_{*}, either using the simplicial cone (-cone=1), or using a homology basis (-cone=2); and
- combining the cohomology algorithm with chunk(*) preprocessing (using -C for chunk* and -DC for chunk).

Machine and setup The code version used for the experiments is revision 1853b6f in the git repository [94]. The repository also contains the shell script (run-experiments.sh) to run the experiments, together with the Jupyter notebooks Generate Samples.ipynb to generate the samples used in these experiments (see above) and the notebook Plots and tables.ipynb used to evaluate these.

The experiments were run on a Ubuntu 22.04.2 LTS machine with 48 3.0 GHz Intel(R) Xeon(R) CPU E5-2687W v4 CPUs and 504 GiB RAM, with the software compiled using g++11.3.0. Some experiments are also run on a MacBook Pro 2017 with a 2.3 GHz Dual-Core Intel Core i5 CPU and 16GiB RAM, with the software compiled using clang++ 15.0.7. We refer to this as machine "M" in the captions.

Each instance was killed when exceeding five minutes runtime. If not stated otherwise, the software uses the heap representation of matrices (see Section 5.1.3.2). Algorithms 3 and 7 can be implemented in an embarrassingly parallel way, which is done in our implementation. Algorithms 6 and 10 The software was run with a thread limit OMP_THREAD_LIMIT=2.

5.3 Results

An overview of the runtime of the cohomology and the homology algorithm for different data sets, subsample sizes and homology dimensions is given in Figures 1 to 3. These figures show the runtime necessary to compute H_1 , H_2 and H_3 (resp. $H^3(N^{\bullet})$, $H^4(N^{\bullet})$ and $H^5(N^{\bullet})$ for the cohomology algorithm). In the case of the homology algorithm, when computing H_d , the numbers also include the time necessary for running Minimize (D_{d+1}) (for the chunk algorithm) or Minimize (D^{d+1}) (for the chunk* algorithm). The runtime does not include time for I/O, time for generating the (co)boundary matrices of the function-Rips complexes, and, in the case of the cohomology algorithm, time for coning off the input complex. We observed that in general, coning off the input complex is a rather cheap preprocessing step, compared to the total runtime.

Note that the plots use logarithmic scale both for the runtime and the number of n + 1simplices. Missing points in the data series occur if program has terminated within five minutes. We see that in most instances, the cohomology algorithm is the fastest method for computing H_2 . For H_3 , both the homology with chunk^{*} preprocessing and the cohomology algorithm are faster than the homology algorithm with chunk preprocessing. Which one is fastest depends on the instances. In Tables 2 to 6, we have listed the runtimes for the three algorithms for a fixed size of the subsample for two different computers. Missing entries indicate that the experiment did not terminate within five minutes. The table also includes the the quotient of the runtime of the homology algorithm by the two compared methods (homology algorithm with chunk* preprocessing, and the cohomology algorithm). We see that for H_2 , we consistently get the best speedup with the cohomology algorithm. While for H_3 , the cohomology algorithm is still consistently faster than the homology algorithm, it is outperformed in some cases by the homology algorithm with chunk* preprocessing.

5.3.1 Matrix representations

The performance of the algorithms depends on the chosen sparse matrix representation. This was already known from one-parameter persistence [12]. It is also known that the impact of the choice of the matrix representation is stronger for cohomological algorithms than for homological ones [12, Table 1, 17, Table 3].

In Tables 7 to 11, we collected runtime data for both versions of the homology algorithm (with chunk(*) preprocessing) and the cohomology algorithm, taken from an implementation with the matrix representation and from an implementation with the vector representation. The columns "speedup" show by which factor the former is faster.

We see that the homology algorithm with chunk preprocessing generally runs slightly slower when using heap matrices. In contrast, chunk^{*} preprocessing can benefit from the heap representation considerably in higher homology dimensions. The cohomology algorithm apparently does not benefit from using heap matrices to the same extend. Comparing Table 8 with 9 and Table 10 with 11 suggests that the impact of the choice of the matrix representation also depends on the machine.

A possible explanation for faster runtimes of the chunk^{*} preprocessing and the cohomology algorithm is the following. The heap representation is better for tall and narrow matrices, which is typically the shape of the coboundary matrices of a Vietoris–Rips complex. Namely, to add two columns in vector representation, one has to iterate over all entries of both summands, and write the symmetric difference of the two vectors to another vector. In contrast, the heap representation allows an efficient in-place addition by copying the entries of one column into the heap representing the other. Typically, most of the runtime goes into reducing a few columns that end up with a high number of non-zero entries. In this case, the heap representation avoids iterating over these entries for every addition to such a column.

Our software also contains code for the implicit matrix representation described in Section 5.1.3.3. However, experiments carried out at an earlier step showed no performance benefit from this matrix representation. Therefore, we did not include this approach in the experiments in this section.

5.3.2 Variants of the cohomology algorithm

We have claimed above that the cohomology algorithm works best without chunk(*) preprocessing, and with a sparsification step (Algorithm 22) added between the two for-loops of Algorithm 10. To justify this, we have run the cohomology algorithm on the same datasets. Apart from the "normal" variant, which includes these steps and which is also used for the other experiments, we

- leave out sparsification,
- leave out clearing,
- use the simplicial cone to make the complex eventually acyclic, instead of Section 5.1.4,

- do not make the input eventually acyclic at all,³
- apply chunk or chunk* preprocessing before we run the cohomology algorithm.

The results are listed in Tables 16 to 19. We see that sparsification may introduce an additional cost; however, it can also reduce the total runtime considerably. The last is particularly true for the vector representation; see Table 18 In general (for both matrix representations), the potential runtime savings are much higher than the additional runtime, which is why we enable sparsification as the default option. We see, however, that the benefit from sparsification is considerably smaller for $H^5(N^{\bullet})$ than for $H^4(N^{\bullet})$

Clearing, on the other hand, has no significant impact at all. This is because the runtime for the first for-loop in Algorithm 10, which is the step where, without clearing, reduction of columns to zero would happen, is negligible in comparison with all other steps of the cohomology algorithm; see Section 5.3.4 below.

With regard to coning off the input complex, we see that the homology-basis-based approach is typically more efficient than taking the simplicial cone. Possibly, this is because the former method constructs the smallest possible chain complex with the desired properties, while the latter method may construct an unnecessarily large complex. This is particularly true for clique complexes. Comparing these numbers to the column "no cone", which contains runtimes of the algorithm without coning off the input complex at all, we see that this step does not introduce a significant overhead.

Lastly, we see that the algorithm does not profit from chunk(*) preprocessing. Indeed, the total runtimes of chunk(*) preprocessing and a following run of the cohomology algorithm maybe much longer than just running the cohomology algorithm. This is particularly true in higher homology dimensions. See Section 5.3.3 for details

5.3.3 Chunk preprocessing

In Tables 12 to 15, we have listed the runtimes for the chunk and the chunk* preprocessing alone, together with the runtimes of running the homology and the cohomology algorithm without this preprocessing, and in combination with the chunk(*) preprocessing; both for the heap and the vector representation. The tables confirm that the homology algorithm profits from chunk and even more from chunk* preprocessing, and that the cohomology algorithm does not run faster in combination with the chunk(*) preprocessing. In fact, we see that chunk(*) preprocessing already takes longer than the cohomology algorithm would take without any preprocessing.

A possible explanation for this is that the chunk algorithm is known, despite decreasing the number of rows and columns of the input boundary matrices, to potentially increase the total number of non-zero entries. In other words, the chunk preprocessing may increase the density of the matrix considerably. This is particularly true for full function–Rips complexes [73, Table 3]. We hypothesize the same is true for chunk* preprocessing.

5.3.4 Steps of the cohomology algorithm

In Tables 20 to 27, we have listed the runtimes for the steps in the cohomology algorithm Algorithm 11. The runtime for the call to **Bireduce**() is spread among the first three columns, listing the runtime of the first and second reduction loop of Algorithm 10, and the cost of sparsifying the matrix between the two loops (Algorithm 22). The last column lists the total runtime for computing the minimal free resolution, which may comprise smaller values not part of the preceding steps.

³In this case, the computed resolution of $H^{\bullet}(N^{\bullet}(K_*))$ may not be correct, and $H^{\bullet}(N^{\bullet}(K_*))$ may not be isomorphic to $H^{\bullet-2}(K_*)$. We included this case nevertheless in order to demonstrate that making the input eventually acyclic does not increase runtime too much.

From these tables, we see that the runtime is dominated by the calls to Minimize() and to KerAndMgsWithKer(). Although sparsification is called before the second reduction loop, it does not improve the performance of this step in most cases (except for O(3)) when using heap matrices, but only the runtime of the following calls to Minimize() and KerAndMgsWithKer(). Interestingly, the benefit arising from sparsification seems to be much bigger for $H^4(N^{\bullet})$ than for $H^5(N^{\bullet})$.

Summary and concluding remarks

In this thesis, we have shown that it is possible to extend several aspects of the duality between persistent homology and cohomology from the context of one-parameter persistence to twoand in parts also to multi-parameter persistence. Specifically, for any $n \ge$ and for a onecritically *n*-parameter filtered simplicial complex K_* we defined a chain complex $N^{\bullet}(K_*)$ of free modules functorial in K_* (Definition 3.1.2 and Lemma 3.2.4). In the special case n = 1, we have that $N^{\bullet}(K_*) = K^{\bullet}(K, K_*)$. The well-known correspondence between absolute and relative one-parameter persistent cohomology, which is induced by the long exact sequence of relative cohomology, can be generalized for any number *n* of parameters to a natural isomorphism

$$H^d(K_*) \cong H^{d+n}(N^{\bullet}(K_*)),$$

assuming that K_* is eventually acyclic, i.e., $H_{\bullet}(K_z) \neq 0$ for only finitely many z. This follows from the Calabi–Yau property of persistence modules (Theorem 3.2.8). The same property also allowed us to find a correspondence between minimal free resolutions of $H^d(K_*)$ and $H_d(K_*)$ (Theorem E). This is a generalization of the correspondence of barcodes of $H^d(K_*)$ and $H_d(K_*)$ in one-parameter persistence. Lastly, we used the Calabi–Yau property to devise an algorithm that efficiently computes $H^{d+n}(N^{\bullet}(K_*))$ for n = 2 (Algorithm 11).

This algorithm computes a minimal free resolution of $H^{d+2}(N^{\bullet}(K_*))$ from the (d+1)st coboundary matrix $\delta^{d+1} \colon N^d(K_*) \to N^{d+1}(K_*)$ and not, as a naive algorithm would do, from δ^{d+2} and δ^{d+3} . This generalizes a corresponding statement in one-parameter persistence, which states that if $K \coloneqq \operatorname{colim} K_*$ is acyclic, then a barcode of $H^{d+1}(K, K_*)$ can be computed just from $\delta^{d+1} \colon C^d(K, K_*) \to C^{d+1}(K, K_*)$, i.e., without considering δ^{d+2} .

Our motivation for this was to compute the persistent homology of two-parameter filtrations of Vietoris–Rips complexes. In one-parameter persistence, Vietoris–Rips complexes have become feasible in practice only through optimization strategies such as clearing, which rely on computing persistent cohomology. The fact that $H^{d+2}(N^{\bullet}(K_*))$ can be computed from the (d+1)st coboundary matrix of $N^{\bullet}(K_*)$ is vital for the practicality of our algorithm, because for Vietoris–Rips complexes and other clique complexes, the higher dimensional coboundary matrices would be prohibitively large. Furthermore, our algorithm offers the possibility to certain optimizations similar to clearing.

We evaluated the practicality of our algorithm in experiments, using a software implementation that we made publicly available [94]. In these experiments, we compared the runtime of our algorithm with the runtime of the state of the art algorithm for minimal free resolutions of two-parameter persistent homology, which is the Lesnick–Wright (LW) algorithm in conjunction with chunk preprocessing. These experiments show that our algorithm is practical and efficient indeed, and, when computing $H_d(K_*)$ for d > 1, outperforms the LW-algorithm by a factor of often more than 10. In these experiments, we also evaluated the impact of different technical details of the implementation, such as sparse matrix representations and others.

Nevertheless, the margin by which our algorithm outperforms the LW algorithm is much smaller than the difference between the Standard Algorithm and the Clearing Algorithm in one-parameter persistent cohomology. We observe that in many cases, the performance of our algorithm suffers from matrix fill-up. This is not a particularity of our algorithm, but also affects the LW algorithm. In fact, it comes from \mathbf{Z}^2 not being totally ordered that in practice, a (co)boundary matrix that is compatible with a \mathbf{Z}^2 -indexed filtration may be much harder to

reduce than a (co)boundary matrix coming from a **Z**-indexed filtration. We assume it is vital for future improvements in multi-parameter persistent (co)homology to overcome the performance bounds posed by matrix fill-up.

Memory-efficient strategies, such as not storing the matrix being reduced in memory, or exploiting frequent special cases that can be handled efficiently (such as apparent pairs) may come in handy. Because our software was mainly designed as a versatile framework for twoparameter persistent cohomology, and in order to investigate the practicality of the algorithms presented in this thesis, these ideas are not implemented in our software, notwithstanding their conceptual simplicity.

We also remark that there is no known algorithm that computes minimal free resolutions of *n*-parameter persistent homology or cohomology in polynomial time for n > 2. In particular, the LW algorithm does not generalize to more than two parameters. The same is true for our algorithm, although most of the underlying algebraic properties (such as the *n*-Calabi–Yau property of *n*-parameter persistence modules, which underlies the isomorphism $H^d(K_*) \cong H^{d+n}(N^{\bullet}(K_*))$, and the correspondence between minimal free resolutions of $H_d(K_*)$ and $H^d(K_*)$ hold for any number of parameters. We thus assume that closing this gap may increase the feasibility of multi-parameter persistence significantly.

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Runtime data tables



Figure 1: Runtimes (ms) of the homology and cohomology algorithm for H_1 , plottet over the number of 2-simplices (bottom) and vertices (top) of the subsample. Command line parameters were -hn1 -cone=2 for the cohomology algorithm, -cn1 -DC2 for the homology algorithm with chunk and -cn1 -C2 with chunk* preprocessing.


Figure 2: Runtimes (ms) of the homology and cohomology algorithm for H_2 , plottet over the number of 3-simplices (bottom) and vertices (top) of the subsample. Command line parameters were $-\ln 2$ -cone=2 for the cohomology algorithm -cn2 -DC3 for the homology algorithm with chunk and -cn2 -C3 with chunk* preprocessing.



Figure 3: Runtimes (ms) of the homology and cohomology algorithm for H_3 , plottet over the number of 4-simplices (bottom) and vertices (top) of the subsample. Command line parameters were -hn3 -cone=2 for the cohomology algorithm -cn3 -DC4 for the homology algorithm with chunk and -cn3 -C4 with chunk* preprocessing.

Table 1: Runtime comparison between mpfree and our reimplementation of the homology algorithm Algorithm 9 in 2pac. The table shows the run time to compute a minimal free resolution of H_2 of the function-Rips complex of a subsample of size 80 of the datasets from Table 5.1. Here, we compare mpfree with the version of 2pac that uses the vector representation. Although mpfree often is slightly faster, the runtimes do not differ much. In contrast to the other experiments, these were run on a MacBook Pro 2017 with a 2.3 GHz Dual-Core Intel Core i5 CPU and 16 GiB RAM.Runtime comparison between mpfree and 2pac.

	mp	free	2pac			
	chunk	H_2	chunk	H_2		
sample						
1-sphere	7 348	471	9 0 3 8	1067		
2-sphere	11095	1870	11978	2 564		
3-sphere	10861	3114	11 333	3817		
4-sphere	13016	5 884	15843	7 672		
O3	12740	8 692	12 529	9633		
O4	11561	6923	11187	8 359		
O5	13 195	7 351	13 466	11 287		
2-torus	12085	2137	12 470	2 567		
3-torus	14 437	6707	14 080	8111		
4-torus	16675	8 575	18060	9 699		
cyclooctane	78816	11 533	58 328	7 183		

Table 2: Comparison of the runtimes of the homology algorithm (with chunk and chunk* preprocessing) and the cohomology algorithm. Each row contains the runtime (ms) for computing a minimal free resolution of $H_1(-)$ of the density-Rips complex of subsamples of size 200. The 6th and 8th column contain the ratio of the respective runtimes.

	chunk	H_1	chunk*	H_1	$\frac{chunk + H_1}{chunk^* + H_1}$	H^3	$\frac{chunk + H_1}{H^3}$
S^1	4 804	7 580	5 042	8 035	0.95	3017	4.10
S^2	3 155	10 526	3 283	10048	1.03	7 230	1.89
S^3	2796	11750	2 840	11 269	1.03	14 463	1.01
S^4	3138	14872	2 926	13474	1.10	26061	0.69
<i>O</i> (3)	2 316	21 932	1804	22 100	1.01	90 836	0.27
O(4)	1828	25 762	1 543	27 091	0.96	-	-
O(5)	2170	15 398	1815	14 541	1.07	15 355	1.14
T^2	3 344	11028	2 638	10 172	1.12	15 672	0.92
T^3	2930	18 197	2 343	17 074	1.09	23 625	0.89
T^4	2932	17 193	2137	16017	1.11	15711	1.28
C_8H_{16}	5 456	19 263	4 009	19910	1.03	12 459	1.98

	chunk	H ₂	chunk*	H ₂	$\frac{chunk + H_2}{chunk^* + H_2}$	H^4	$\frac{chunk + H_2}{H^4}$
S^1	27 414	2 578	10034	2 422	2.41	7 569	3.96
S^2	42 190	9 524	15 576	8 983	2.11	9733	5.31
S^3	47 248	19 103	21 650	18 087	1.67	13564	4.89
S^4	45 497	19 590	19696	-	-	8 206	7.93
<i>O</i> (3)	45 057	61 030	31 625	58 126	1.18	86 347	1.23
O(4)	39831	66 573	31 397	62109	1.14	17 077	6.23
O(5)	45 235	55 750	28783	53817	1.22	10183	9.92
T^2	44 220	15 815	18794	14733	1.79	9 304	6.45
T^3	43711	25 648	21 984	25 279	1.47	9615	7.21
T^4	53 143	51 937	28762	49 007	1.35	14 444	7.27
C_8H_{16}	-	-	28 193	37 339	_	21910	_

Table 3: Same as Table 2, but for H_2 , and with subsamples of size 100.

Table 4: Same as Table 3, but run on the machine M.

	chunk	H_2	chunk*	H_2	$\frac{chunk + H_2}{chunk^* + H_2}$	H^4	$\frac{chunk + H_2}{H^4}$
S^1	37 289	5 200	12 155	4 638	2.53	8744	4.86
S^2	52 294	14 242	18764	13 568	2.06	7 982	8.34
S^3	50 261	23 272	19211	25 089	1.66	15 998	4.60
S^4	50 206	25 615	21 198	29 454	1.50	8964	8.46
<i>O</i> (3)	55 566	67 668	24 786	66 158	1.36	21742	5.67
O(4)	49 257	50 104	22 577	52 270	1.33	8 980	11.06
O(5)	49861	59 751	25 093	55 153	1.37	11 150	9.83
T^2	44 758	13 240	16 377	13137	1.97	7 921	7.32
T^3	51814	44 898	22 597	39 410	1.56	17 216	5.62
T^4	54 126	60 019	25 565	54 947	1.42	16546	6.90
C_8H_{16}	209 094	47 583	29 800	47 482	3.32	32 365	7.93

Table 5: Same as Table 2, but for H_3 , and with subsamples of size 60.

	chunk	H ₃	chunk*	H_3	$\frac{chunk + H_3}{chunk^* + H_3}$	H^5	$\frac{chunk + H_3}{H^5}$
S^1	42 826	1 993	1 401	2169	12.55	13759	3.26
S^2	86 650	2771	6 365	2846	9.71	13 373	6.69
S^3	128 907	5 184	9707	4 272	9.59	12753	10.51
S^4	123 745	5 981	10916	5815	7.75	14015	9.26
<i>O</i> (3)	164 526	20 048	28 1 19	17735	4.03	14 266	12.94
O(4)	143 490	19 329	28 167	18 563	3.48	13632	11.94
O(5)	106 272	20 293	24 470	19070	2.91	13895	9.11
T^2	124 135	4 578	9 806	4 005	9.32	13 209	9.74
T^3	123 169	5 940	11 526	4 886	7.87	13 301	9.71
T^4	171 873	16 470	21 924	15 203	5.07	13 154	14.32
C_8H_{16}	-	-	71 454	11535	-	14 958	-

				5.0 0, 540			
	chunk	H_3	chunk*	H_3	$\frac{chunk + H_3}{chunk^* + H_3}$	H^5	$\frac{chunk + H_3}{H^5}$
S^1	52 476	2731	2 989	2702	9.70	14 428	3.83
S^2	98841	3188	7 815	3 1 2 1	9.33	14703	6.94
S^3	112 448	6 480	9 329	6241	7.64	11943	9.96
S^4	121 720	9062	14 025	9 456	5.57	14838	8.81
<i>O</i> (3)	116 542	13710	24 206	14 847	3.34	14621	8.91
O(4)	130 011	14 386	26 825	13728	3.56	14843	9.73
O(5)	118761	24 942	25 274	21 144	3.10	14736	9.75
T^2	108 809	5216	9 573	5188	7.72	13102	8.70
T^3	139 316	14 464	22 173	12603	4.42	13040	11.79
T^4	166 287	13722	20 985	14 603	5.06	12897	13.96
C ₈ H ₁₆	_	-	116 980	30 0 16	-	15 421	_

Table 6: Same as Table 5, but run on the machine M.

Table 7: Runtime comparison of heap and vector representation. The table lists runtimes of the homology algorithm with chunk, chunk^{*}, and the cohomology algorithm, each with the heap and the vector representation, The column "speedup" lists the runtime with the vector representation, divided by the runtime of the heap representation. The runtimes are taken for computing H_1 for subsamples of size 200 of the respective samples.

		$chunk+H_1$			chunk*+ <i>H</i>	1		H^3		
	heaps	vectors	speedup	heaps	vectors	speedup	heaps	vectors	speedup	
S^1	12 384	10912	0.88	13077	12 551	0.96	3017	3746	1.24	
S^2	13681	12029	0.88	13 331	11 124	0.83	7 2 3 0	13028	1.80	
S^3	14 546	11784	0.81	14 109	11711	0.83	14 463	30 507	2.11	
S^4	18010	13720	0.76	16 400	12 636	0.77	26061	22 234	0.85	
<i>O</i> (3)	24 248	18 193	0.75	23 904	17 924	0.75	90 836	39 414	0.43	
O(4)	27 590	24 795	0.90	28 634	22 854	0.80	-	53 629	-	
O(5)	17 568	14 153	0.81	16 356	12843	0.79	15 355	14 342	0.93	
T^2	14 372	11 298	0.79	12810	10 001	0.78	15672	9839	0.63	
T^3	21 127	15751	0.75	19417	15 442	0.80	23 625	31 1 16	1.32	
T^4	20 125	15 790	0.78	18154	15 081	0.83	15711	13794	0.88	
C_8H_{16}	24719	14 119	0.57	23919	15 194	0.64	12 459	16 058	1.29	

Table 8: Same as Table 7, but for computing H_2 , and with subsamples of size 100.

		$chunk+H_2$!		$chunk^* + H_2$			H^4		
	heaps	vectors	speedup	heaps	vectors	speedup	heaps	vectors	speedup	
S^1	29 992	30 115	1.00	12 456	132 855	10.67	7 569	7 566	1.00	
S^2	51714	42 938	0.83	24 559	118 823	4.84	9733	20 546	2.11	
S^3	66 351	46 422	0.70	39737	115 088	2.90	13564	53 948	3.98	
S^4	65 087	51741	0.79	-	114 482	-	8 206	7 274	0.89	
<i>O</i> (3)	106 087	76740	0.72	89751	148 436	1.65	86 347	185 573	2.15	
O(4)	106 404	64 582	0.61	93 506	156 135	1.67	17077	28919	1.69	
O(5)	100 985	58 988	0.58	82 600	161 102	1.95	10 183	13966	1.37	
T^2	60 035	51770	0.86	33 527	104 008	3.10	9 304	17 221	1.85	
T^3	69 359	50702	0.73	47 263	135 179	2.86	9615	11014	1.15	
T^4	105 080	65 117	0.62	77 769	161 908	2.08	14 444	30770	2.13	
C_8H_{16}	-	171 381	-	65 532	-	-	21910	170 360	7.78	

					,				
		$chunk+H_2$	2		chunk*+ <i>H</i>	2		H^4	
	heaps	vectors	speedup	heaps	vectors	speedup	heaps	vectors	speedup
S^1	42 489	51 606	1.21	16793	183 232	10.91	8744	26 296	3.01
S^2	66 536	80 296	1.21	32 332	205 232	6.35	7 982	8 2 3 2	1.03
S^3	73 533	78 732	1.07	44 300	173794	3.92	15 998	105 595	6.60
S^4	75 821	81 801	1.08	50 652	193 228	3.81	8964	12 446	1.39
<i>O</i> (3)	123 234	121 360	0.98	90 944	257 593	2.83	21742	80 548	3.70
O(4)	99 361	93 838	0.94	74 847	217 728	2.91	8 980	10 427	1.16
O(5)	109 612	107 905	0.98	80 246	248 209	3.09	11150	12 253	1.10
T^2	57 998	72 185	1.24	29 514	192961	6.54	7 921	7 036	0.89
T^3	96712	99 274	1.03	62 007	196 956	3.18	17 216	91 363	5.31
T^4	114 145	121 173	1.06	80 512	240 337	2.99	16546	37 658	2.28
C_8H_{16}	256 677	-	-	77 282	-	-	32 365	-	-

Table 9: Same as Table 8, but run on machine M.

Table 10: Same as Table 7, but for computing H_3 , and with subsamples of size 50.

		chunk+H3	3	$chunk^* + H_3$				H^5		
	heaps	vectors	speedup	heaps	vectors	speedup	heaps	vectors	speedup	
S^1	18728	12 189	0.65	1 476	5 223	3.54	4 821	4 346	0.90	
S^2	48 394	42 502	0.88	2896	20 528	7.09	4769	4 273	0.90	
S^3	45 226	35 858	0.79	3765	18 240	4.84	4 643	4 277	0.92	
S^4	44 188	41 964	0.95	4 940	21 607	4.37	5 062	3 924	0.78	
<i>O</i> (3)	91 427	82 780	0.91	12176	58 628	4.82	4 708	4 079	0.87	
O(4)	58 547	53 700	0.92	10 541	54 100	5.13	4 6 9 1	4 150	0.88	
O(5)	32781	28 006	0.85	8 644	32 805	3.80	5 208	4 662	0.90	
T^2	36 027	29 737	0.83	4 000	16 295	4.07	4 655	4 353	0.94	
T^3	37 427	33 144	0.89	6 347	34 430	5.42	5 063	3 932	0.78	
T^4	54739	50 096	0.92	11 006	44 921	4.08	4 946	4 289	0.87	
C ₈ H ₁₆	281 220	-	-	24 105	164 893	6.84	6 6 2 6	4 560	0.69	

Table 11: Same as Table 10, but run on machine M.

		chunk+H ₃		$chunk^* + H_3$				H^5		
	heaps	vectors	speedup	heaps	vectors	speedup	heaps	vectors	speedup	
S^1	23 202	29 425	1.27	1 955	10 586	5.41	4 920	3 798	0.77	
S^2	29830	45 988	1.54	3 0 0 3	24 803	8.26	4 998	4 268	0.85	
S^3	64 478	97 334	1.51	5726	39 121	6.83	5 066	4 447	0.88	
S^4	58 205	78 114	1.34	7 152	39748	5.56	5147	3 762	0.73	
O(3)	60 287	110 063	1.83	12881	104 865	8.14	5142	4 064	0.79	
O(4)	44 307	63 782	1.44	11668	94 886	8.13	5072	3814	0.75	
O(5)	43 197	65 849	1.52	8 908	81 485	9.15	4967	3 923	0.79	
T^2	44 114	70 518	1.60	3 3 3 1	23 829	7.15	4137	3 796	0.92	
T^3	37 744	61 395	1.63	9031	66 894	7.41	4 4 4 9	3 955	0.89	
T^4	62 315	101 003	1.62	12963	68 146	5.26	4133	3971	0.96	
C_8H_{16}	-	-	-	32 654	166 399	5.10	5 223	4 572	0.88	

Table 12: Runtime and effect of chunk and chunk* preprocessing. The reported runtime of chunk (resp.
chunk*) preprocessing is the time needed to minimize (Algorithm 3) the 3th boundary (resp. coboundary)
matrix. The two other groups show the runtime of the homology and the cohomology algorithm, each without
preprocessing, after chunk and after chunk* preprocessing. The run times were measured for subsamples of
size 100.

				H_2			H^4	
	chunk	chunk*	w/o	w. chunk	w. chunk*	w/o	w. chunk	w. chunk*
S^1	27 414	10034	118 209	2 578	2 422	7 569	3 636	3 371
S^2	42190	15576	178 587	9 524	8 983	9 7 3 3	8 4 1 9	7 306
S^3	47 248	21 650	-	19 103	18 087	13 564	28 563	-
S^4	45 497	19696	-	19 590	-	8 206	7 781	7 534
<i>O</i> (3)	45 057	31 625	-	61 0 30	58 126	86 347	-	-
O(4)	39 831	31 397	-	66 573	62 109	17 077	127 683	126 768
O(5)	45 235	28783	-	55 750	53 817	10 183	13 905	12904
T^2	44 220	18794	262 684	15815	14733	9 304	10 394	9 965
T^3	43711	21 984	-	25 648	25 279	9 6 1 5	12625	-
T^4	53143	28762	-	51 937	49 007	14 444	25 265	-
C_8H_{16}	-	28 193	-	-	37 339	21 910	44 551	42 301

 Table 13: Same as Table 12, but for the vector representation.

				H_2			H^4	
	chunk	chunk*	w/o	w. chunk	w. chunk*	w/o	w. chunk	w. chunk*
S^1	27 779	130 466	209 212	2 336	2 389	7 566	3 2 3 1	3 2 4 7
S^2	36 583	112 898	-	6 355	5 925	20 546	4 944	4 433
S^3	35 817	104 276	_	10 605	10812	53 948	8 4 3 3	7 668
S^4	40 265	102 946	_	11 476	11 536	7 274	4418	4 423
<i>O</i> (3)	38 574	116 482	_	38 166	31 954	185 573	26 989	26 027
O(4)	31 466	121 588	-	33116	34 547	28 919	10519	10123
O(5)	31 092	132 402	-	27 896	28700	13966	6836	7 533
T^2	42 311	95 259	-	9 459	8749	17 221	5 341	4727
T^3	35 576	117 914	_	15 126	17 265	11014	5 866	6 087
T^4	38 868	137 073	-	26 249	24 835	30 770	10828	10 055
C_8H_{16}	155 365	-	-	16016	-	170 360	13 254	-

Table 14: Same as Table 12, but for H_3 , and with a subsample size of 60.

				H_3			H^5	
	chunk	chunk*	w/o	w. chunk	w. chunk*	w/o	w. chunk	w. chunk*
S^1	42 826	1 401	-	1 993	2169	13759	4 485	4 359
S^2	86 650	6 365	-	2771	2846	13 373	4 4 1 8	5 026
S^3	128 907	9 707	-	5 184	4 272	12753	5 170	5 446
S^4	123745	10916	-	5981	5815	14015	5 645	5 417
<i>O</i> (3)	164 526	28 1 19	-	20 048	17 735	14 266	8918	8 308
O(4)	143 490	28 167	-	19 329	18 563	13632	8778	7 747
O(5)	106 272	24 470	-	20 293	19070	13895	8 862	7 602
T^2	124 135	9 806	-	4 578	4 005	13209	5 978	5 265
T^3	123 169	11 526	-	5 940	4 886	13 301	6 045	6 007
T^4	171 873	21 924	-	16 470	15 203	13154	7 692	7 053
C_8H_{16}	-	71 454	-	-	11 535	14 958	-	6 564

				H ₃			H^5	
	chunk	chunk*	w/o	w. chunk	w. chunk*	w/o	w. chunk	w. chunk*
S^1	41 923	4 406	-	2067	2 389	13672	4 462	4743
S^2	78721	108 905	-	2573	2 401	11646	4614	4 548
S^3	104 420	98 1 18	-	3 635	3727	11515	4 363	4 405
S^4	125 403	147 877	-	4 471	4218	12 496	4 472	4 930
<i>O</i> (3)	150 888	-	-	10088	-	13 287	5 429	-
O(4)	130 892	-	-	9847	-	13169	4 988	-
O(5)	97 518	-	-	11 479	-	13625	5 656	-
T^2	120 352	130 758	-	3 361	3 371	12234	4751	4 822
T^3	112 293	170 990	-	3 990	4 2 2 9	12795	5 023	4 595
T^4	161 083	261 555	-	9196	-	11792	5 249	4 890
C_8H_{16}	-	-	-	-	-	14 347	-	-

Table 15: Same as Table 14, but for the vector representation.

Table 16: Run times of different variants of the cohomology algorithm. Each row contains the runtime for computing $H^4(N^{\bullet}(-))$ ($= H_2(K_*)$) of the density-Rips complex of a subsample of size 100. The first column contains the runtime of the cohomology algorithm as used in all other tables, including sparsification (Algorithm 22) between the two loops of Algorithm 10, and coning off the input complex (Section 5.1.4 with respect to the density parameter of the input. The column "no sparsification" contains runtimes without this sparsification step. The column "no clearing" contains runtimes without the clearing step in line (a) of Algorithm 11. The column "simplicial cone" contain runtimes when using the simplicial cone to make the input eventually acyclic (Section 3.7.1). The column "no cone" contains run times without any preprocessing that makes the input eventually acyclic. Note that in this case, one cannot infer H_2 from $H^4(N^{\bullet}$. The last two columns show the runtime of the cohomology algorithm when combined with chunk and chunk* preprocessing.

	normal	no sparsification	no clearing	simp cone	no cone	chunk	chunk*
S^1	7 569	5 461	7 548	8 162	8174	33 870	13 521
S^2	9733	6 909	8104	10 290	8 058	49 117	22 266
S^3	13 564	17 332	12212	14630	12173	73 401	-
S^4	8 206	14 786	6 845	8 358	7732	55 021	27 297
O(3)	86 347	137 011	79 025	59 528	39 186	-	-
O(4)	17077	69 926	16916	24738	17 097	169 454	155 382
O(5)	10 183	60 882	9 080	11 394	12076	56 908	38 4 18
T^2	9 304	10 797	8 891	9 807	9 275	59 798	31 761
T^3	9615	22 271	9148	9789	9730	54 777	-
T^4	14 444	37 775	15 921	15 041	12647	80 866	-
C_8H_{16}	21 910	28 445	24 618	23 587	18 000	237 735	73 353

	normal	no sparsification	no clearing	simp cone	no cone	chunk	chunk*
S^1	13759	7 043	6 432	14 885	12717	47 747	5 704
S^2	13373	7 897	6 6 9 7	14 597	13219	92410	10999
S^3	12753	8067	8 4 4 3	15 179	12 529	135 566	13571
S^4	14015	8744	8 492	14 440	14 263	135 491	16 280
<i>O</i> (3)	14 266	20 644	21 233	15 004	14 865	166 358	37 476
O(4)	13632	15970	18958	14 940	13 220	146 703	34 616
O(5)	13895	22 545	24 174	15 365	13725	113 428	31 973
T^2	13209	8757	8 524	13754	13531	136 668	15 008
T^3	13 301	9081	8 7 9 0	14 310	13 357	122 560	17718
T^4	13 154	14965	15977	16 297	13212	182112	27 347
C_8H_{16}	14 958	13784	13958	17 776	15 317	-	81 506

Table 17: Same as Table 16, but for $H^5(N^{\bullet})$ ($\doteq H_3(K_*)$), and with a subsample size of 60.

Table 18: Same as Table 16, but with the vector representation.

	normal	no sparsification	no clearing	simp cone	no cone	chunk	chunk*
S^1	7 566	5 538	7 286	8 897	11 102	32 189	139 049
S^2	20 546	34 070	19 168	23 135	18 363	41 672	109 123
S^3	53 948	136 226	52 552	57 022	56 479	47 938	110 360
S^4	7 274	89 539	7 462	8776	7 981	43 085	108 275
<i>O</i> (3)	185 573	-	181 390	242 204	203 544	61 565	135 672
O(4)	28 919	-	28 862	36 612	29 574	44 218	128 983
O(5)	13966	-	13988	15 431	15614	43 037	147 286
T^2	17 221	52 698	16 407	18 989	16 095	45 278	93917
T^3	11014	140 227	11 185	11 984	11 950	40 512	123 177
T^4	30770	191 980	27 653	27 534	25 933	55 042	138 676
C_8H_{16}	170 360	262 912	156 842	194 073	133 231	185 177	-

 Table 19: Same as Table 17, but with the vector representation.

	normal	no sparsification	no clearing	simp cone	no cone	chunk	chunk*
S^1	4 346	2 193	2 0 3 8	5 0 5 4	3 599	12840	5 694
S^2	4273	3 1 3 3	2973	5 655	3 968	42 198	19179
S^3	4 277	2 5 3 4	2 4 9 0	5 380	4864	37 125	17 032
S^4	3924	4 2 2 5	3 381	5732	4194	37 132	20 253
<i>O</i> (3)	4 0 7 9	12985	12 336	5013	4 508	76 542	52 926
O(4)	4 150	9 209	9164	5 312	3942	46 173	56 870
O(5)	4 662	9261	9214	5 267	4830	25 404	32764
T^2	4 353	6916	7 221	4 850	4 307	28 604	16 321
T^3	3932	6196	5 534	4972	4 554	30 047	35 352
T^4	4 289	8042	7 636	5 262	4 362	45 320	39 200
C_8H_{16}	4 560	20 191	19873	5 5 17	4 985	-	169 944

Table 20: Run times for the steps of the cohomology algorithm (Algorithm 11). The columns mean, from left to right, the first and second reduction for-loop in Algorithm 10, the sparsification step (Algorithm 22) between the two loops, and the call to Minimize() and and KerAndMgsWithKer() in Algorithm 11. The numbers are from computing H^4 with a subsample size of 100.

	first	second	sparsify	minimize	kernel	res
S^1	45	122	3071	724	1716	7 569
S^2	62	424	3710	1 155	2189	9733
S^3	73	503	3 4 3 1	2991	3 2 3 0	13 564
S^4	64	92	3513	786	1793	8 206
<i>O</i> (3)	48	17971	3661	41758	14611	86 347
O(4)	79	1 374	3162	3 800	5 207	17077
O(5)	45	404	3 265	1779	2514	10 183
T^2	63	271	3 624	1149	2032	9 304
T^3	63	209	3679	1 298	2083	9615
T^4	67	370	3 968	2767	4048	14 444
C ₈ H ₁₆	99	1 227	4 694	6 233	5 319	21 910

Table 21: Same as Table 20, but for the vector representation.

	first	second	sparsify	minimize	kernel	res
S ¹	54	489	3 0 2 0	668	1 564	7 566
S^2	53	8 351	3 1 3 1	4784	2 0 9 7	20 546
S^3	53	16069	3 1 5 3	28 474	2986	53 948
S^4	51	173	2801	853	1 597	7 274
<i>O</i> (3)	73	34 128	3 552	129 835	10 389	185 573
O(4)	61	4 745	3 4 2 4	12840	4 445	28919
O(5)	60	1896	3 265	4 374	2163	13966
T^2	54	5 282	3 2 2 9	4 360	2049	17 221
T^3	56	1 195	3076	2612	1954	11014
T^4	57	4 535	3 301	15 659	3 989	30770
C_8H_{16}	130	36 454	4 069	120 199	5142	170 360

 Table 22: Same as Table 20, but without the sparsification step.

	first	second	sparsify	minimize	kernel	res
S^1	103	156	-	595	2 294	5 461
S^2	60	471	-	1 547	2 396	6 909
S^3	60	689	-	6 895	4 6 2 6	17 332
S^4	57	161	-	4 954	4 966	14786
<i>O</i> (3)	62	27 688	-	72 867	23 456	137011
O(4)	51	4 5 3 3	-	27 675	25 453	69 926
O(5)	61	763	-	18851	29 556	60 882
T^2	54	297	-	3 297	3 543	10797
T^3	61	252	-	7 984	7 408	22 271
T^4	73	581	-	12601	15 574	37 775
C_8H_{16}	90	1 554	-	12 229	8 2 3 1	28 445

	first	second	sparsify	minimize	kernel	res
S^1	64	1 400	-	571	1623	5 538
S^2	52	17 181	-	12 120	2 273	34 070
S^3	62	19 626	-	106 965	4 686	136 226
S^4	61	1 212	_	77 378	5616	89 539
<i>O</i> (3)	70	46 602	-	-	-	-
O(4)	69	11 093	-	-	-	-
O(5)	72	3 848	_	255 013	_	_
T^2	54	10 602	-	34 891	3 404	52 698
T^3	72	2 321	-	124 181	6764	140 227
T^4	73	11 325	_	158 941	12718	191 980
C ₈ H ₁₆	138	47 227	-	202 075	7 261	262 912

Table 23: Same as Table 22, but without the sparsification step.

Table 24: Same as Table 20, but for H^5 and a subsample size of 60.

	first	second	sparsify	minimize	kernel	res
S^1	362	254	6 669	1 154	2 295	13759
S^2	345	209	7011	1018	2 095	13 373
S^3	326	203	6 582	1019	2 088	12753
S^4	342	228	6 855	1 183	2 319	14015
O(3)	425	300	6 992	1 354	2 206	14 266
O(4)	382	292	6 859	1 214	2 106	13632
O(5)	387	247	6 867	1 2 3 1	2 359	13 895
T^2	334	240	6 599	1 101	2 213	13 209
T^3	372	248	6 669	1 223	2018	13 301
T^4	348	211	6 4 9 6	1 040	2 235	13154
C_8H_{16}	678	232	8 400	1 004	1 936	14 958

 $\label{eq:table 25: Same as Table 24, but for the vector representation.$

	first	second	sparsify	minimize	kernel	res
S^1	217	155	6810	914	2 363	13672
S^2	179	132	5748	846	2 101	11 646
S^3	186	156	5811	878	1961	11515
S^4	314	144	5 939	929	2 227	12 496
<i>O</i> (3)	197	233	6 305	1 249	2 271	13 287
O(4)	221	173	6632	1 055	2 279	13169
O(5)	218	181	6 888	1073	2 441	13625
T^2	191	160	6116	885	2 1 3 9	12 234
T^3	194	144	6410	1 062	2 211	12795
T^4	192	152	5842	874	2078	11792
C_8H_{16}	509	160	7 7 4 4	794	2 135	14 347

	first	second	sparsify	minimize	kernel	res
S^1	312	192	_	1 1 2 9	2 495	7 043
S^2	325	218	-	1 275	2716	7 897
S^3	314	200	-	1837	2 405	8067
S^4	326	254	-	1818	2802	8744
<i>O</i> (3)	375	335	-	8 0 3 6	5 387	20 644
<i>O</i> (4)	329	326	-	4 975	4789	15 970
O(5)	353	383	-	8071	6 535	22 545
T^2	317	229	-	2 0 9 8	2 569	8757
T^3	430	293	-	1937	2 6 9 4	9081
T^4	325	261	-	4 603	4 471	14 965
C_8H_{16}	688	275	-	5 995	2725	13784

Table 26: Same as Table 24, but without the sparsification step.

first second sparsify minimize kernel res S^1 177 149 882 2113 6154 _ S^2 161 163 _ 766 2138 5986 *S*³ 202 203 2414 _ 8 376 14 307 S^4 221 231 _ 8982 2 508 $15\,218$ *O*(3) 205 667 _ $203\,171$ 4 549 214 265 O(4)206 251 _ 54 535 3 989 63785 O(5) T^2 _ 232 320 193 038 5 5 5 4 205 314 254 371 _ 31 209 2839 38 3 30 T^3 194 212 _ 13717 2 4 4 0 19907 T^4 204 272 _ 47 274 3940 56 157 $\mathsf{C}_8\mathsf{H}_{16}$ _ 2768 525 151 102 046 109 022

Table 27: Same as Table 26, but for the vector representation.