



Analysis and Optimization of Networks for Flexibility

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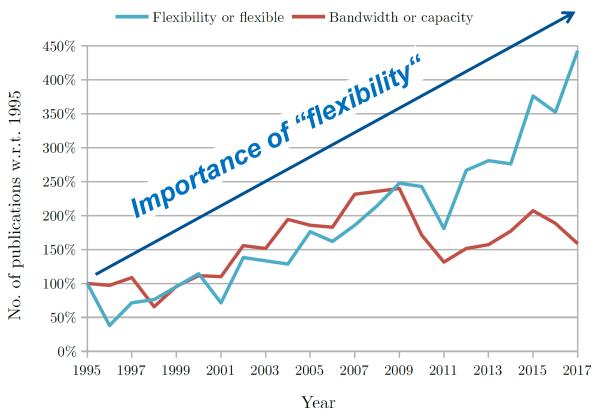




The rise of flexibility

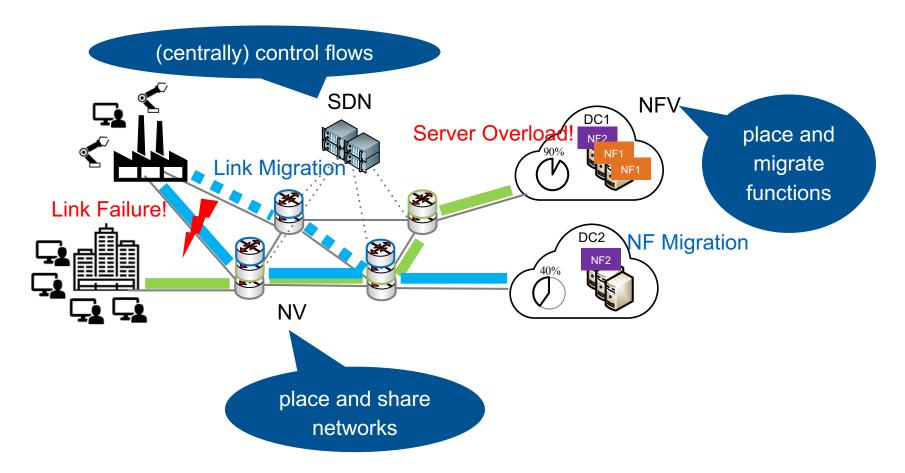


Flexibility is gaining increasing attention and importance



Evolution of the number of publications containing the words "flexible" or "flexibility" in contrast with those containing "bandwidth" or "capacity" in four major IEEE journals and magazines on communication, with respect to the number of publications in 1995.

Fueling this flexibility trend: Softwarized Networks



Network Virtualization (NV), Network Function Virtualization (NFV), and Software Defined Networking (SDN)

...promise to create and adapt networks and functions on demand in software

Why is flexibility so important?



Evolution tells us that the more flexible species can better survive

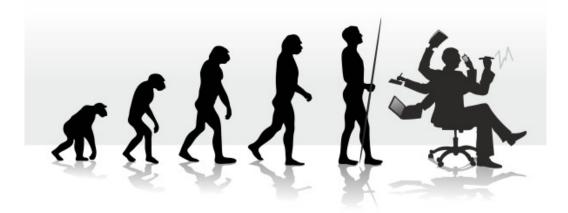
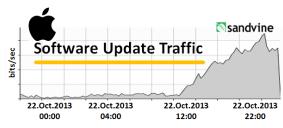


Image source: http://www.paleoplan.com

- What about networks? Will they survive?
- So far less <u>explicitly</u> addressed: <u>flexibility</u> to adapt to future demands
- Considering the Future is <u>very</u> important for survival
 - enables operators to cover the future
 - key decision factor between network designs
 - optimize networks for flexibility

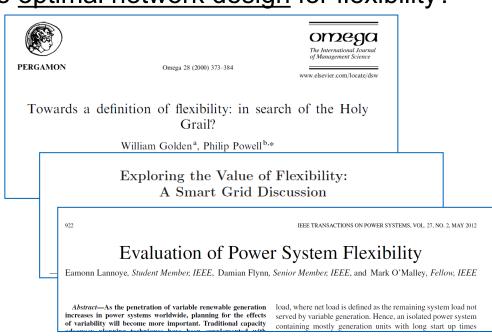


Are we there already?



- Are we <u>100% flexible</u> already (e.g. with Softwarized Networks)?
- How <u>far</u> can we go? What is the <u>optimal network design</u> for flexibility?

What is network flexibility?



We need

- a fundamental understanding of how to provide flexibility
- a quantitative measure for flexibility pro and contra certain designs

An exercise on measuring flexibility





VS.



Fixed-set tool

Re-configurable tool box

- Which tool is more flexible?
 - re-configuration shows more potential to be more flexible
- When can both exibit the same flexibility?
 - maybe there is no need to change -> probability of requests make a difference
 - maybe both cannot satsify my requests → infeasible
- When can the re-configurable tool be less flexible?
 - adaptation time

 re-configurable object might not be handy
 - cost → inefficient



Our approach for Network Flexibility



Network **flexibility** = ability to support *adaptation requests (challenges)* (e.g., new requirements or traffic patterns) in a *timely* and *efficient* manner

W. Kellerer, et al., "How to measure network flexibility? A proposal for evaluating softwarized networks," *IEEE Communications Magazine*, 2018.

www.networkflexibility.org

We provide

- a definition of network flexibility
- a quantitative measure for flexibility pro and contra certain designs
- Optimization for flexibility
- Empower networks for flexibility to cover the future

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Measuring Network Flexibility (our proposal)



(comparing network designs)

adaptation time threshold (T) and cost budget (C)

Input: Constraints T, C

- 1. Design sequence $\mathbb{C} = \{s_{i_1,j_1}, s_{i_2,j_2}, \dots\}$ with $\nu(s_{i,j}) = V$
- 2. Initialize $\Sigma = 0$
- 3. For k = 1:K
 - a. Challenge state switch $S_{i_k} \mapsto S_{j_k}$
 - b. Observe τ_X and c_X
 - c. If $\tau_X \leq T$ and $c_X \leq C$: $\Sigma \coloneqq \Sigma + 1$
- 4. END
- 5. $\varphi(T,C) \coloneqq \Sigma/K$

count

successes

challenges: request sequence

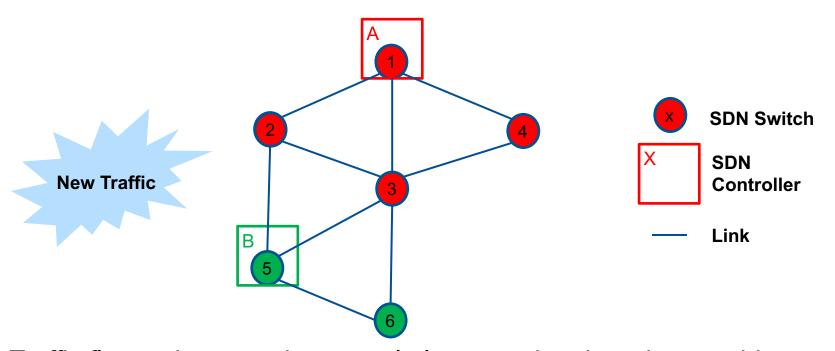
check if system can adapt and record time and cost

Flexibility

$$\varphi(T,C) = \frac{|\text{supported requests within constraints } (T,C)|}{|\text{Number of requests}|}$$

Case study 1: Dynamic Controller Placement



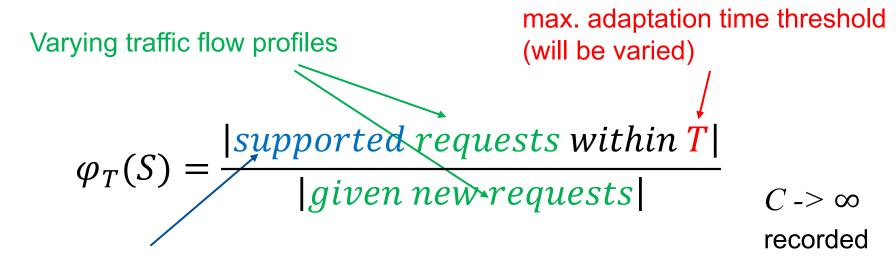


- Traffic fluctuations require control plane to adapt in order to achieve better control performance → Dynamic Control Plane
 - SDN controller migration & SDN switch reassignment

Flexibility Aspect	New Request	Flexibility Measure	System Objective	Cost in focus
C .: 1	α ' 1	C 4' C C 1	, 1 C	(ODEV)
function placement	new flow arrival	fraction of successful	control performance:	operation latency (OPEX):
	(from distribution)	controller placements	(min. avg. flow setup time)	avg. flow setup time

Case study 1: Dynamic Controller Placement



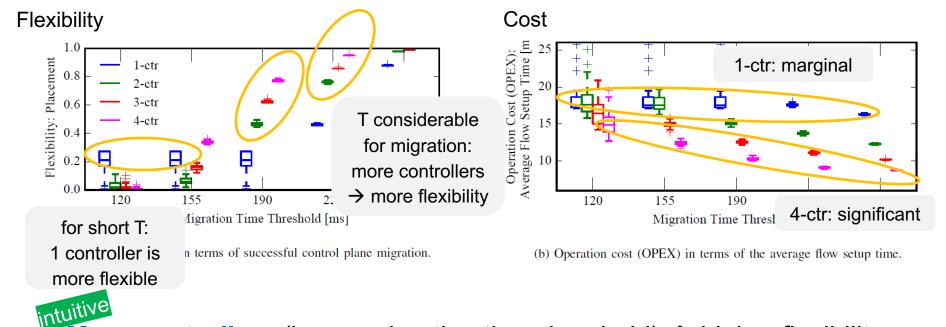


SDN controller migration and switch reassignment can be done within T

- Flexibility → Migration Success Ratio
 - Calculate controller migration and switch reassignment time T_migration
 - If T_migration smaller than T → count as a supported request

Case study 1: Dynamic Controller Placement





- More controllers (larger migration time threshold) → higher flexibility
- Single controller case: more flexible for tight time threshold as unexpected! In the single controller stays in optimal location is high
- 1 controller → marginal performance improvement vs. adaptation T
- 4 controllers → significant performance improvement vs. adaptation T
- However, if we consider all cost factors, we can reach a trade-off!

Case study 2: SDN Resilience

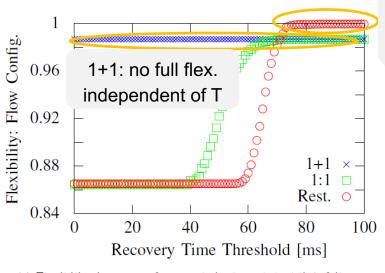


- Flexibility aspect of flow configuration for a resilience scenario in an SDN network under a given recovery time threshold T.
- Objective: system recovery
- Compare 3 systems: 1:1 protection vs 1+1 protection vs restoration
- Flexibility measure: fraction of recoverable failures
- New requests: all possible single and dual link failures

New Request	Flexibility Measure	System Objective	Cost in focus
all possible single	fraction of	system recovery:	resources overhead (CAPEX):
and dual failures	recoverable failures	(single and dual failures)	node and link reservation

Case study 2: SDN Resilience





restoration:
full flex.
needs enough T

	Resources Cost (CAPEX)		
	Node reservation:	Link reservation:	
	Avg. number of flow table entries	Number of required links	
1 + 1	11.78	13038	
1:1	11.78	13038	
Rest.	5.05	5400	

(b) System resources cost (CAPEX) in terms of nodes and links used for reservation.

1+1 can not reach full flexibility

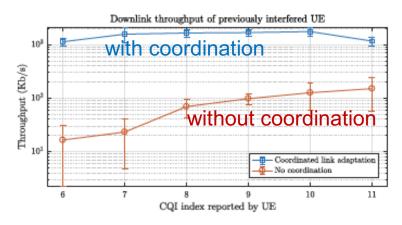
- However, 1+1 is obviously independent of recovery time
- Restoration can cover all failures if given enough recovery time
- Protection imposes more than 2x capex overhead than restoration
- Again, if we consider all cost factors, we can reach a trade-off!

⁽a) Flexibility in terms of covered single and dual link failures.

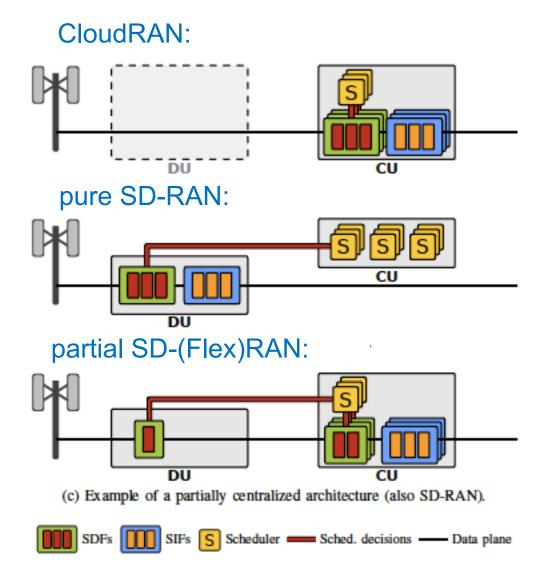
Case study 3: FlexRAN (ongoing work)



- Radio Access Network plus SDN/NFV
 - → unexplored flexibility
- our use case: coordinated scheduling
- initial results: PoC



 next: quantify flexibility flexibility: ratio of successful handling of request





- We can measure flexibility
 so far relatively between multiple systems
- Results can be less intuitive than one might think
- Measure can be used to design for flexibility

Optimize for Flexibility φ





Design Phase

- Optimize for performance metric (e.g. latency and throughput)
- quantify flexibility value (success ratio)

 Optimize for flexibility measure, decide system design parameters (e.g., bandwidth, # base stations, etc.)

Use Case example: Dynamic Controller Placement Problem

- Requests: traffic profiles with target average flow setup time
- Objective: max. flexibility (success: # accomodated traffic profiles)
- Design parameters: # data centers and their locations

Optimize for Flexibility



- Design of methods to optimize for flexibility
 - offline optimization
 - online optimization
 - adaptation time speedup through machine learning
 - empower a network to optimize for the future
 - runtime reconfigurability of HW

Speedup adaptation time

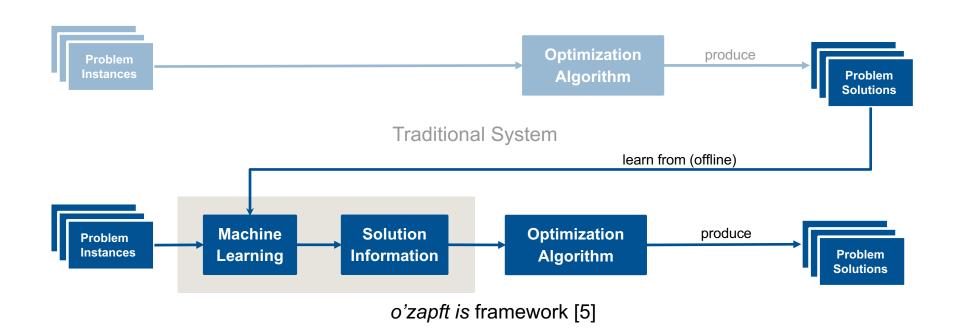


- Adaptation time is very important for flexible networks
- Adaptation examples:
 - function migration, e.g., SDN controller
 - (re-)embedding of virtual networks/flows, e.g. for resilience
 - shift of Radio Access Network functions to a central node
- How can we speedup?
- Yet another heuristic for a specific case study?

We propose:

- Keep your favourite optimization algorithms and
- Boost your network algorithm with ML preprocessing

How can we boost the solving of the related optimization problems (leaving you algs. untouched)?



State-of-the-art: Neglects produced data!

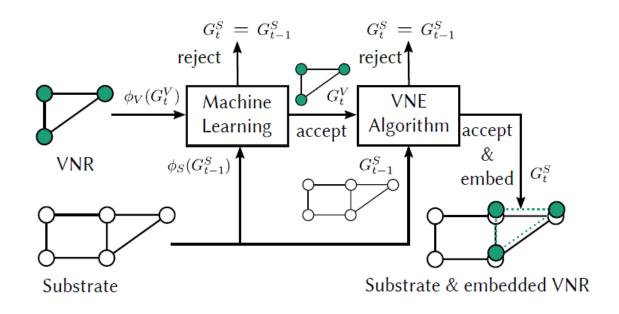
Idea: Use problem/solution data generated by algorithms regularly solving problems

A. Blenk, P. Kalmbach, S. Schmid, W. Kellerer: *o'zapft is: Tap Your Network Algorithm's Big Data!*ACM SIGCOMM 2017 Wrksp. on Big Data Analytics and Machine Learning for Data Communication Networks (Big-DAMA), 2017.

<u>Data Available:</u> P. Kalmbach, J. Zerwas, M. Manhart, A. Blenk, S. Schmid, W. Kellerer. Data on "o'zapft is Tap Your Network Algorithm's Big Data!",2017 https://doi.org/10.14459/2017md1361589

Case Study: Predicting Acceptance Probabilities of VNE Requests

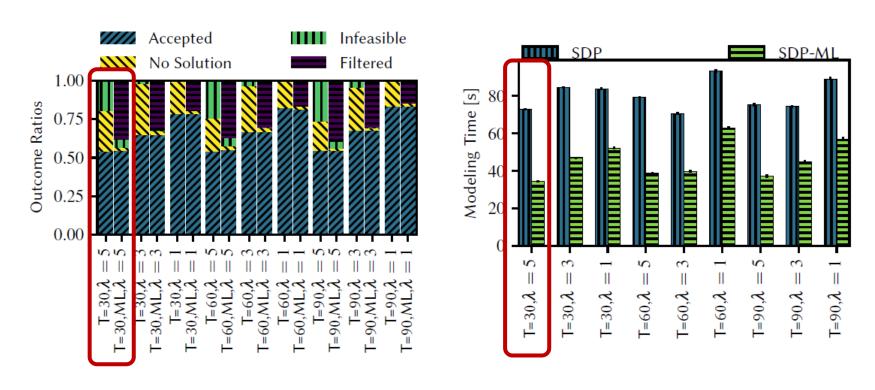




- Supervised learning: use data with accepted and rejected requests! Offline training!
- Recurrent neural network (RNN) for classification
- Filter infeasible and requests with unacceptable algorithm runtime ("no solution")

Can we speed-up optimal algorithms using admission control?





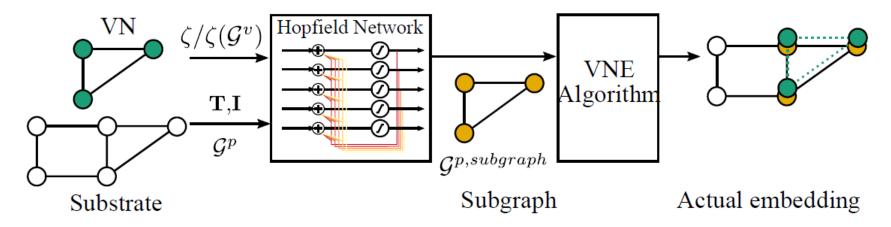
Efficient Filtering of infeasible and unacceptable requests Efficient saving of model creation time

Latest Results: Neurovine



Hopfield neural network to preprocess (subgraph extraction) VNE algorithms

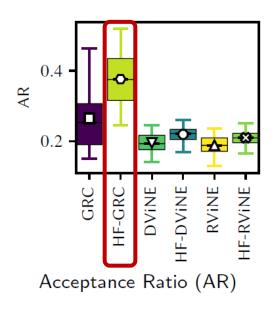
– tailored filtering

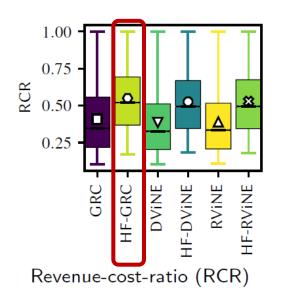


 Idea: Extract subgraph with physical nodes close to each other and high available capacities

Neurovine: Efficiency on Real Network Topologies







- VNE algorithms (GRC, DViNE, RViNE) vs. Hopfield variants (HF-GRC, HF-DViNE, HF-RViNE)
- NeuroViNE accepts more networks with less costs

Optimize for Flexibility



- Design of methods to optimize for flexibility
 - offline optimization
 - online optimization
 - adaptation time speedup through machine learning
 - > we still have a clear objective here to optimize for
 - empower a network to optimize for the future

Empower your network



optimize for the (unknown) future:

prepare for possibly unexpected events → flexibility

we need:

(online) self-optimization

self-driving networks (Rexford, Feamster): networks which measure, analyze and control themselves in an automated manner, reacting to changes in the environment

prepare for the unknown

We propose:

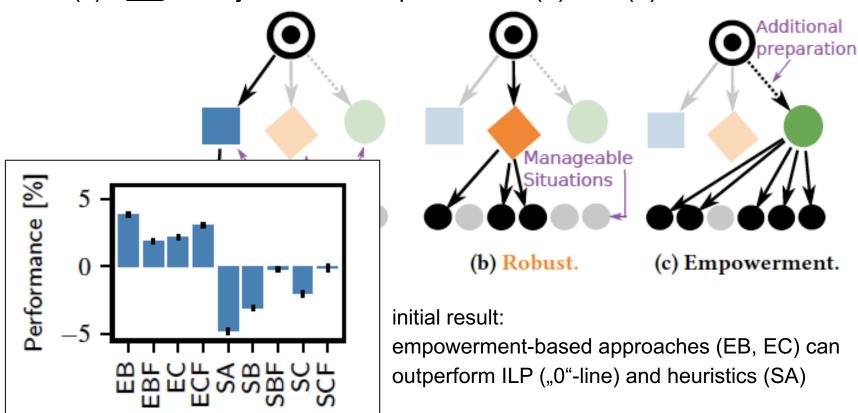
use empowerment for preparedness

P. Kalmbach, J. Zerwas, P. Babarczi, A. Blenk, W. Kellerer, S. Schmid: *Empowering Self Driving Networks*, ACM SIGCOMM 2018 workshop on self-driving networks August 2018.

Empowering Networks



empowerment: quantify the influence of an agent on its environment: agent (several actuators, 1 sensor) restructures networks to maximize options (c) - not an objective as in optimization (a) and (b)



P. Kalmbach, J. Zerwas, P. Babarczi, A. Blenk, W. Kellerer, S. Schmid: *Empowering Self Driving Networks*, ACM SIGCOMM 2018 workshop on self-driving networks August 2018.

Key Takeaways & outlook



- We propose a definition and measure for flexibility
 - to compare flexible systems
 - to explicitly design for flexibility
- (online) optimization for flexibility is supported by
 - Speedup of opt. algorithms through ML-preprocessing
 - Empowerment to optimize for flexibility to cover the future
 - Runtime reconfigurability of HW with P4
 - → Mu He et al.: *P4NFV: An NFV Architecture with Flexible Data Plane Reconfiguration* in today's afternoon session

join us on

networkflexibility.org

References



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