

Flexible and Data-Driven Softwarized Networks

Wolfgang Kellerer

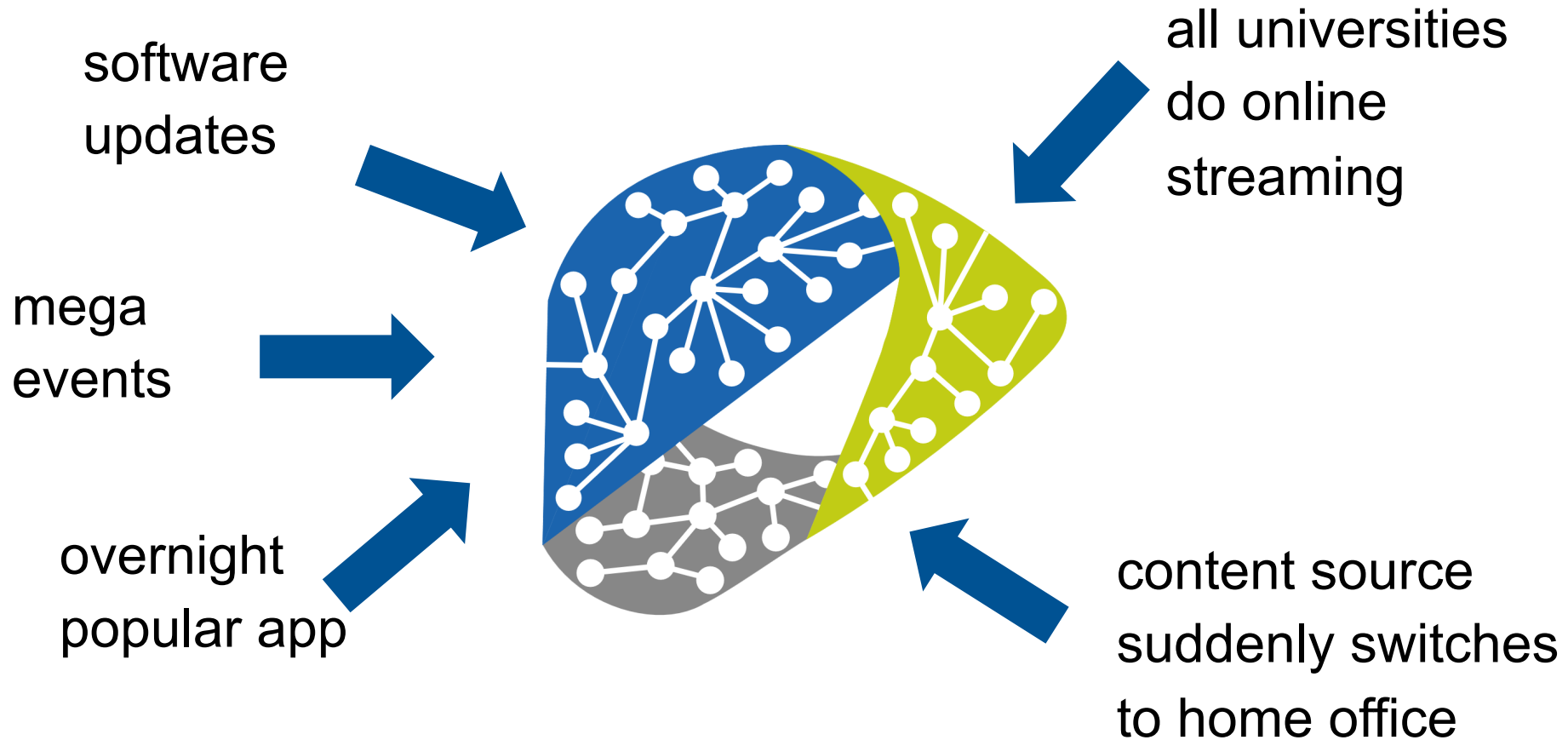
Technical University of Munich (TUM), Germany

with Prof. Dr. Peter Babarzci, Dr. Andreas Blenk, Dr. Arsany Basta, Mu He, Patrick Kalmbach, Dr. Markus Klügel, Alberto Martinez Alba, Prof. Dr. Martin Reisslein, Prof. Dr. Stefan Schmid, Johannes Zerwas

NetSoft 2020, Ghent, Belgium, July 2, 2020

Based on work published in W. Kellerer, et al.
*Adaptable and Data-Driven Softwarized Networks:
Review, Opportunities and Challenges.*
In Proc. of the IEEE, Vol 107, No 4. 2019.

Need for Flexible Network Adaptation



to react to dynamic events network as a critical infrastructure has to adapt to new contexts

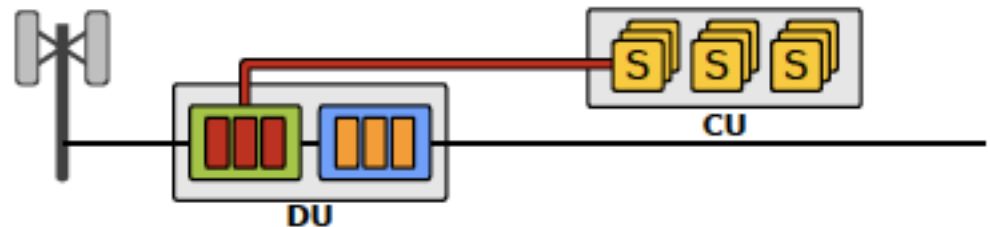
Flexible network adaptation example

- Radio Access Network plus SDN/NFV
→ SD-(Flex) RAN
- Dynamic allocation of RAN functions across fixed function split boundaries addressing
 - user densities
 - resource availability

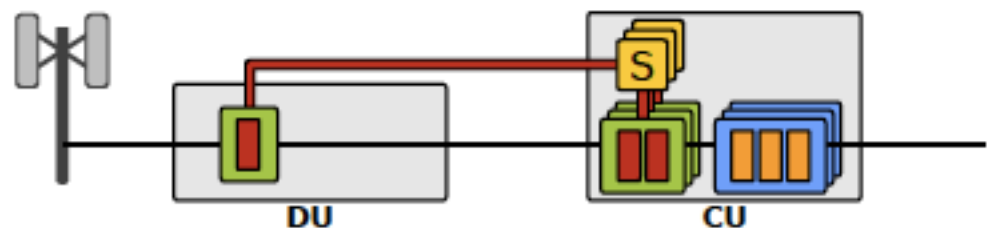
CloudRAN:



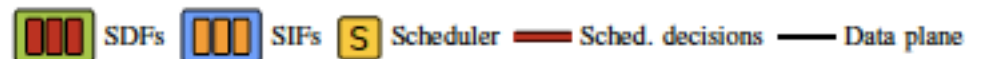
pure SD-RAN:



partial SD-(Flex)RAN:

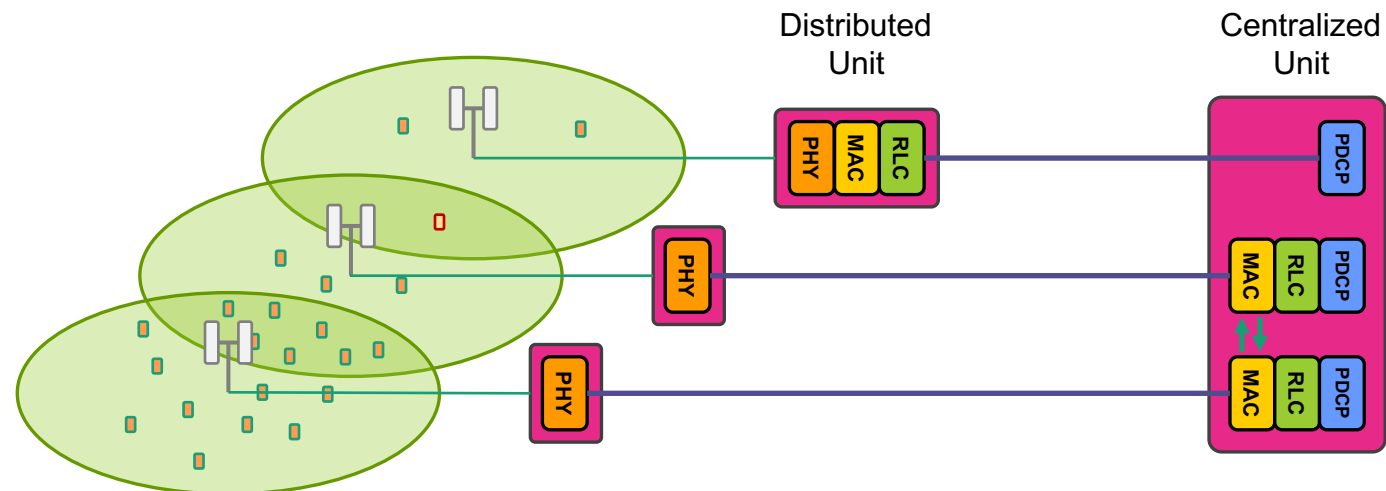


(c) Example of a partially centralized architecture (also SD-RAN).



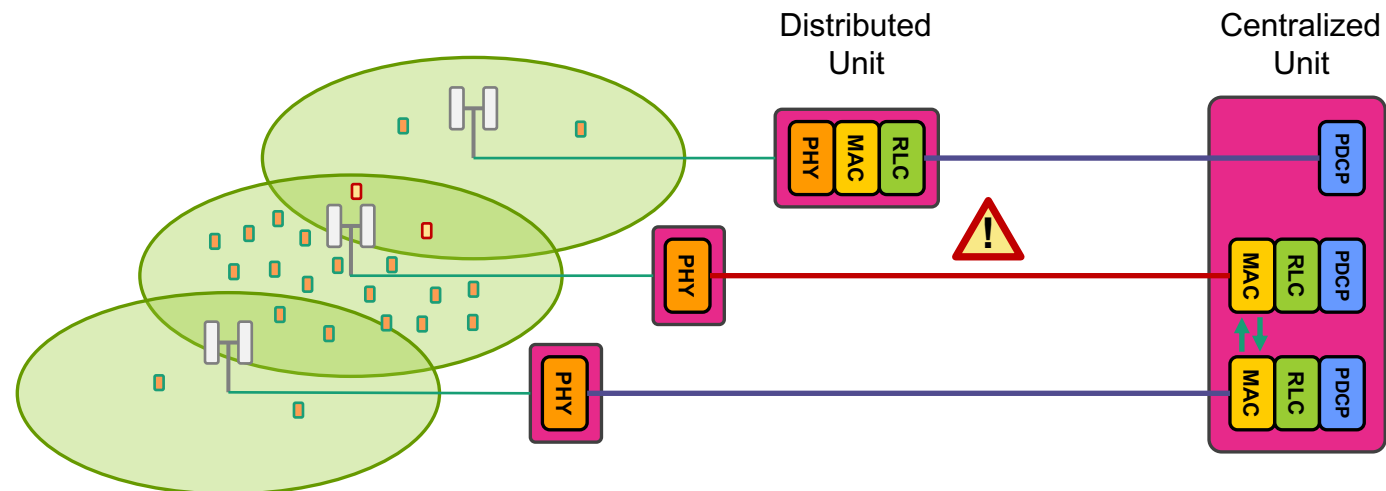
Fixed 5G Function Split

- Function split implemented on dedicated hardware
- Difficult to update
- Deviations from expected distribution of users lead to



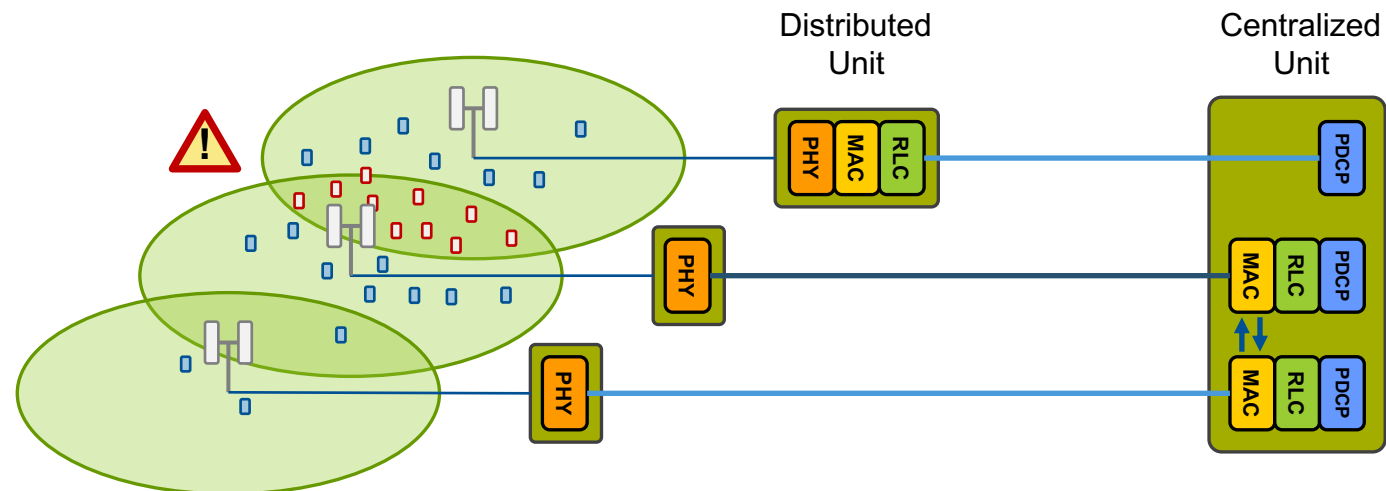
Fixed 5G Function Split

- Function split implemented on dedicated hardware
- Difficult to update
- Deviations from expected distribution of users lead to
 - Network congestion



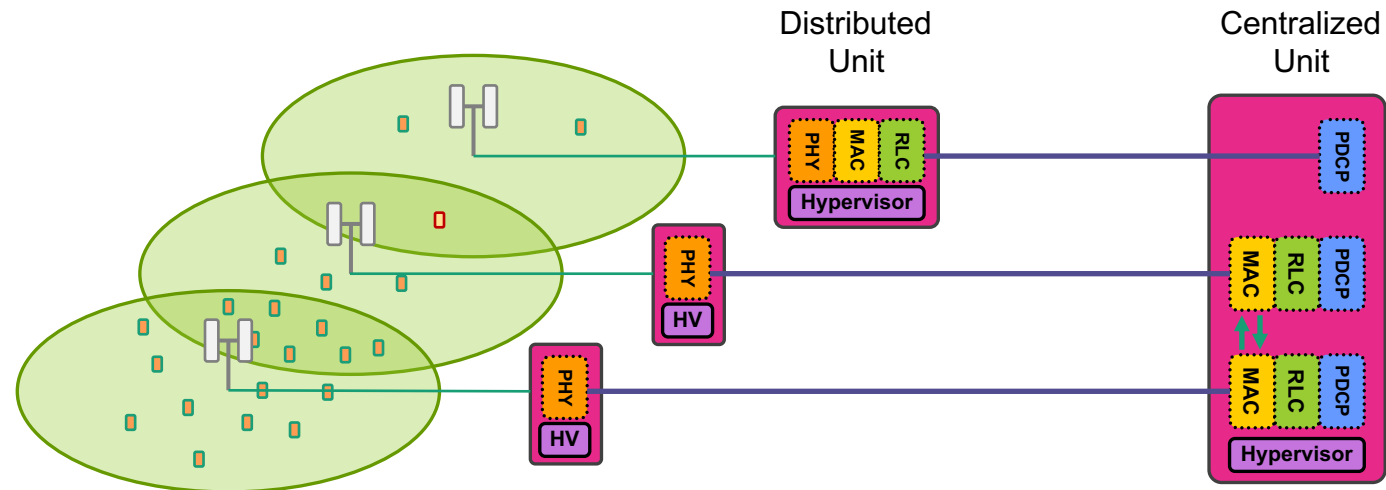
Fixed 5G Function Split

- Function split implemented on dedicated hardware
- Difficult to update
- Deviations from expected distribution of users lead to
 - Network congestion
 - Unmanaged interference



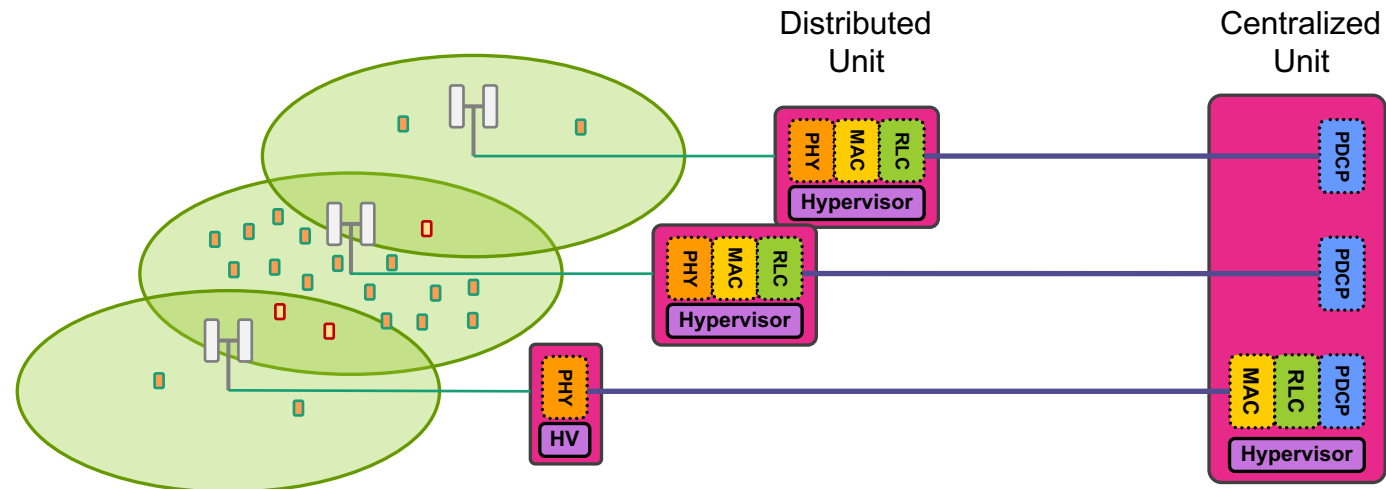
NFV-based 5G+ Function Split

- Functions are softwarized and implemented on off-the-shelf hardware
- Simple to deploy and update



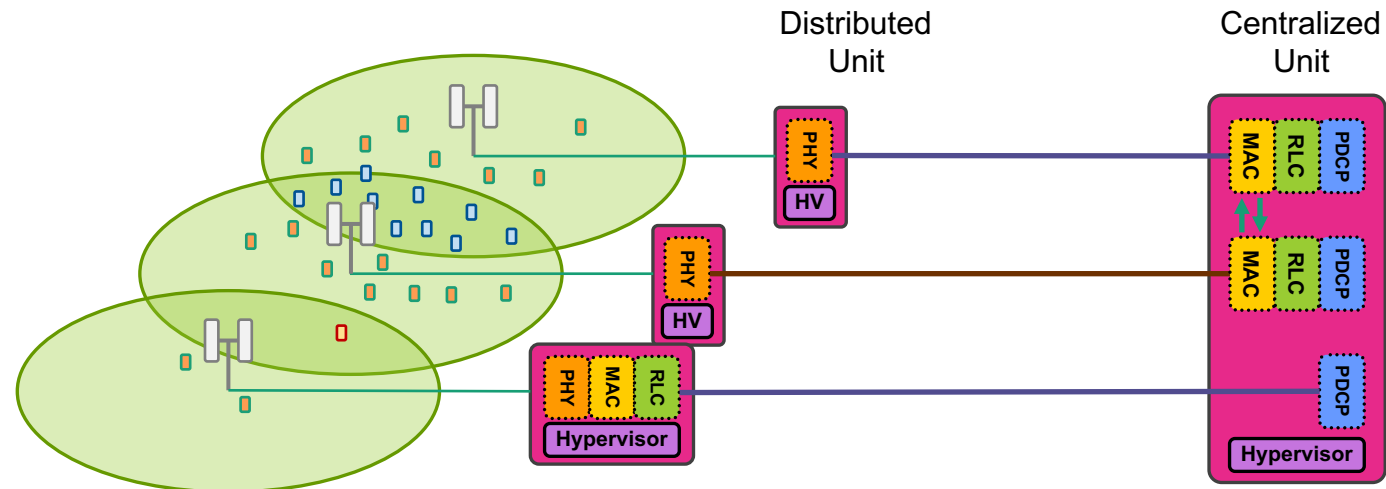
NFV-based 5G+ Function Split

- Functions are softwarized and implemented on off-the-shelf hardware
- Simple to deploy and update
- Functions can be migrated to adapt to network changes



NFV-based 5G+ Function Split

- Functions are softwarized and implemented on off-the-shelf hardware
- Simple to deploy and update
- Functions can be migrated to adapt to network changes



Based on a full Proof-of-Concept implementation at

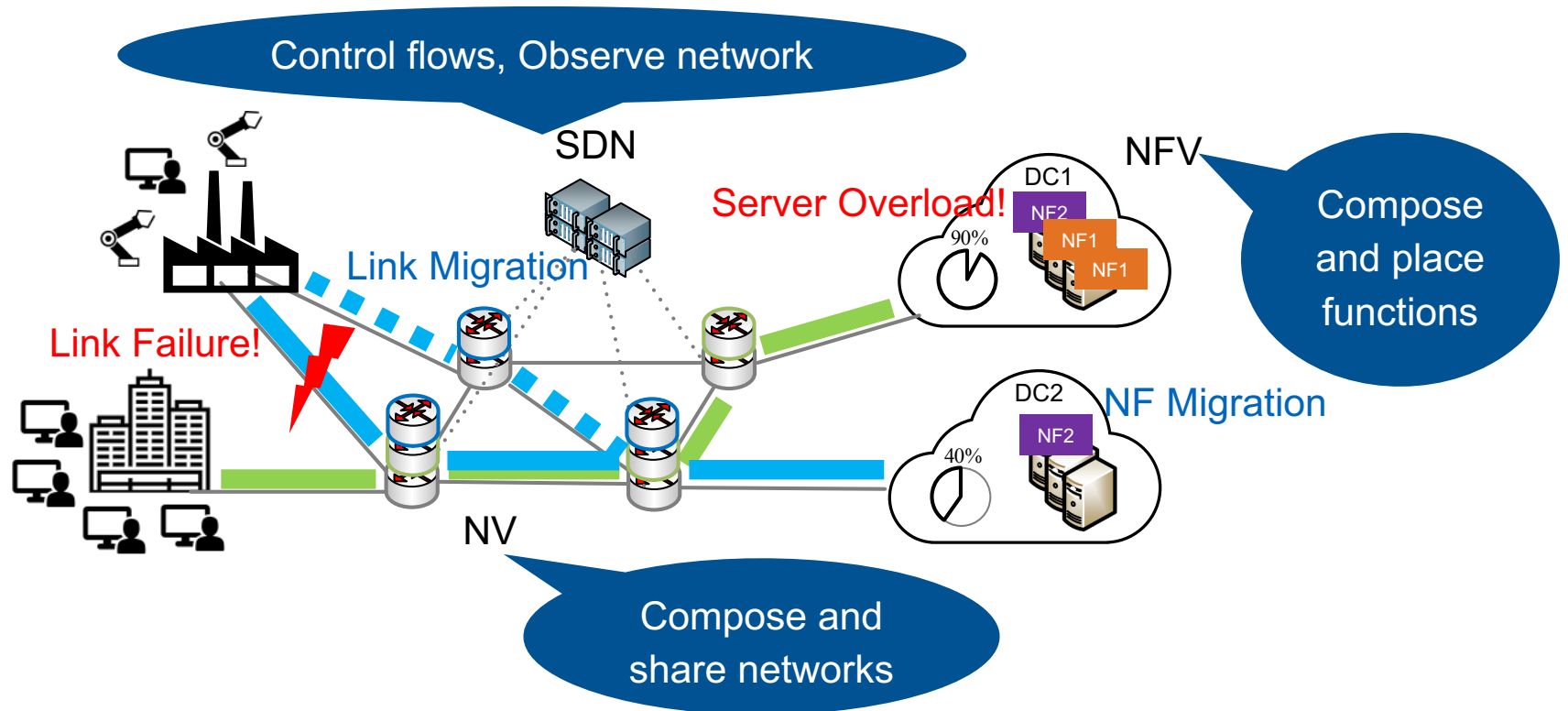


Martínez Alba, Alberto; Gómez Velásquez, Jorge Humberto; Kellerer, Wolfgang:

An adaptive functional split in 5G networks.

2019 IEEE INFOCOM WKSHPS - 3rd Workshop on Flexible and Agile Networks: 5G and Beyond

Flexible network adaptation: Softwarized Networks



Network Virtualization (NV)

Network Function Virtualization (NFV)

Software Defined Networking (SDN)

enablers for flexible adaptation

- Network systems today: max. throughput or min. latency
- Flexible systems keep future options open
 - flexibility as an objective
- Flexibility can **measure** how good different implementations can **adapt** to future challenges?
 - **time** and **cost** are significant constraints

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

www.networkflexibility.org

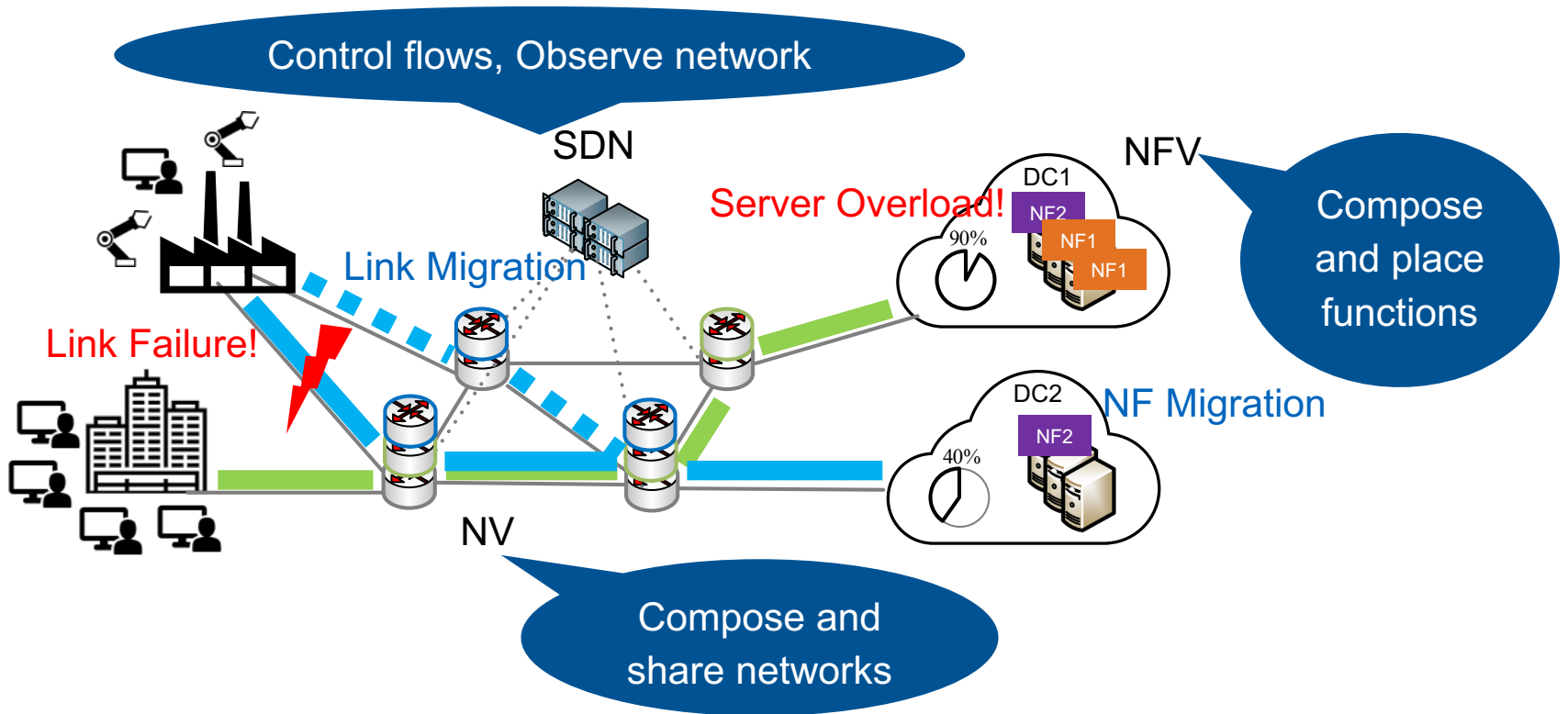
www.networkflexibility.org



M. Klügel, M. He, W. Kellerer, P. Babarczy:

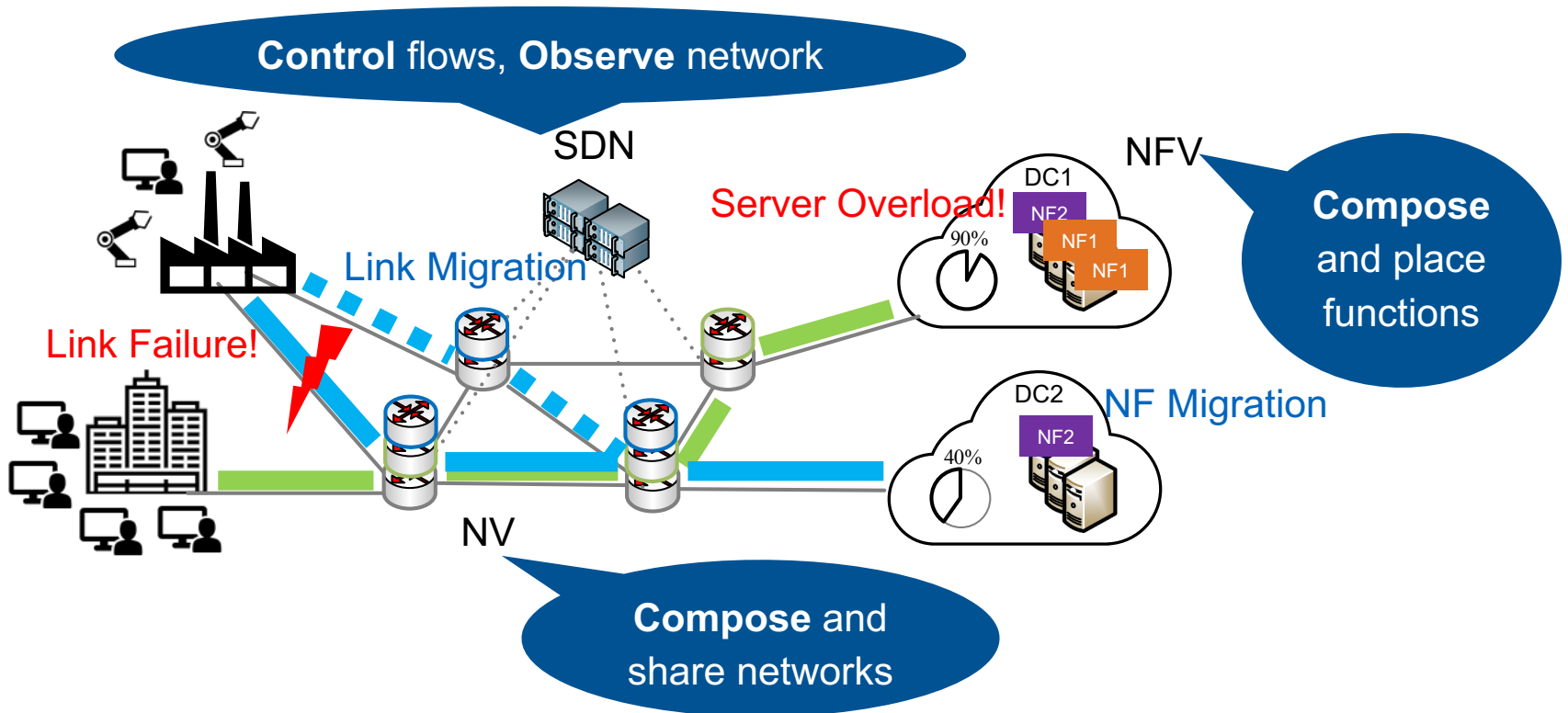
A Mathematical Measure for Flexibility in Communication Networks. **IFIP NETWORKING 2019.**

3 phases of network adaptation



Adaptation: (1) Detection – (2) Decision – (3) Execution

3 functional primitives by softwarized networks



Adaptation: (1) Detection – (2) Decision – (3) Execution

Network Observation
(SDN, NFV)

Composition (NFV, NV)

Control (SDN, NV)

Enablers contribute differently to primitives

	Enabler(s)		Primitive(s)		
	SDN	NFV	Obs.	Comp.	Ctl.
Adaptation					
Monitoring	•	•	•		
Event detection	•	•	•		
Adaptive measurements	•	•	•		
Function configuration		•		•	
Fct. cfg. (push to SDN node)	•			•	•
Fct. res. dimen. (scale)		•		•	
Add/remove f.ct./ctl. (scale)	•	•		•	
Function placement	•			•	
Controller placement	•			•	
Function (de-)composition					
Function chaining		•	•	•	•
HW / SW configuration	•	•	•	•	•
Flow steering (direct rout.)	•				•
Traffic engineering	•				•
Rule/policy adaptation	•	•	•		•
Consistent network update	•		•		•
Admit and embed services and virtual networks	•		•		•

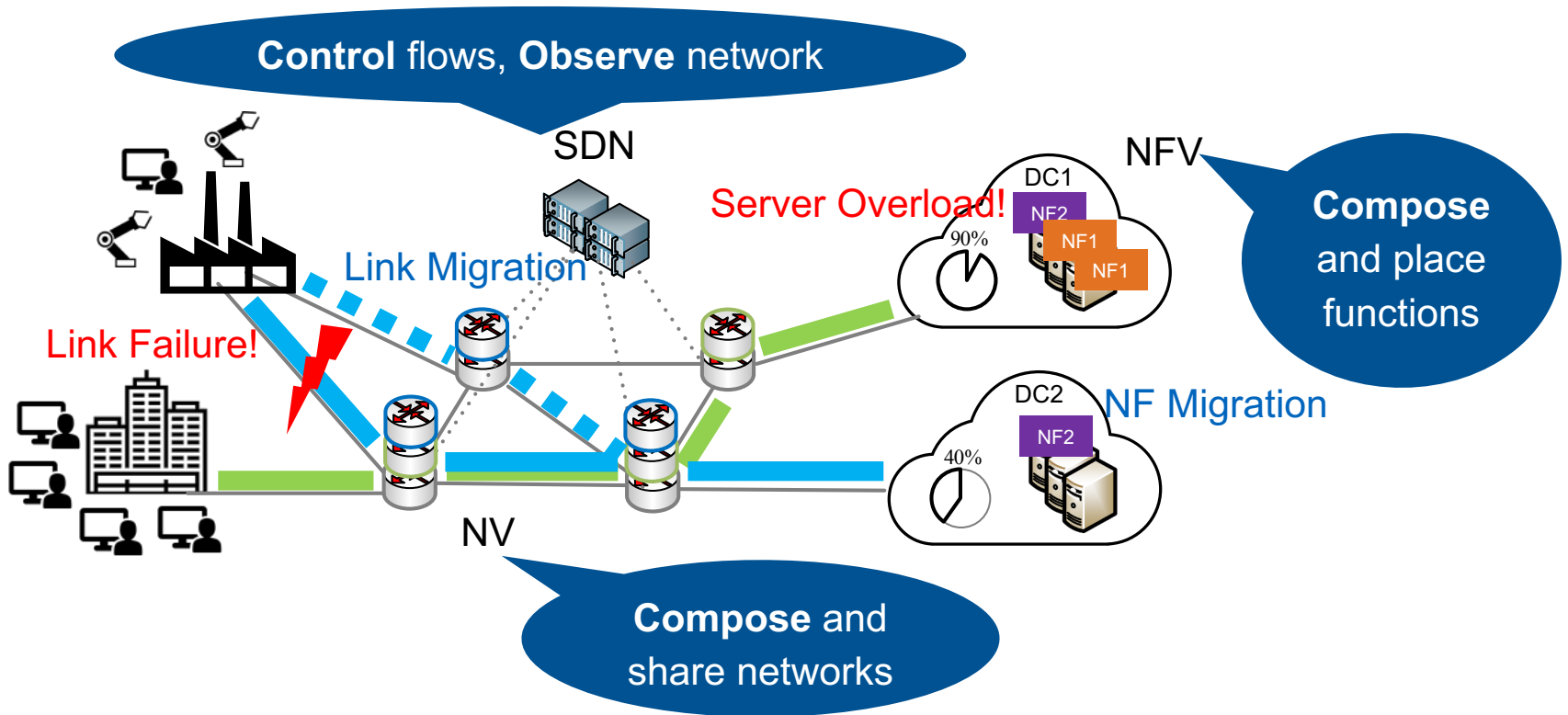
data collection

compose, configure,
Place, migrate, chain

flow steering,

traffic engineering

Adaptation phases meet functional primitives



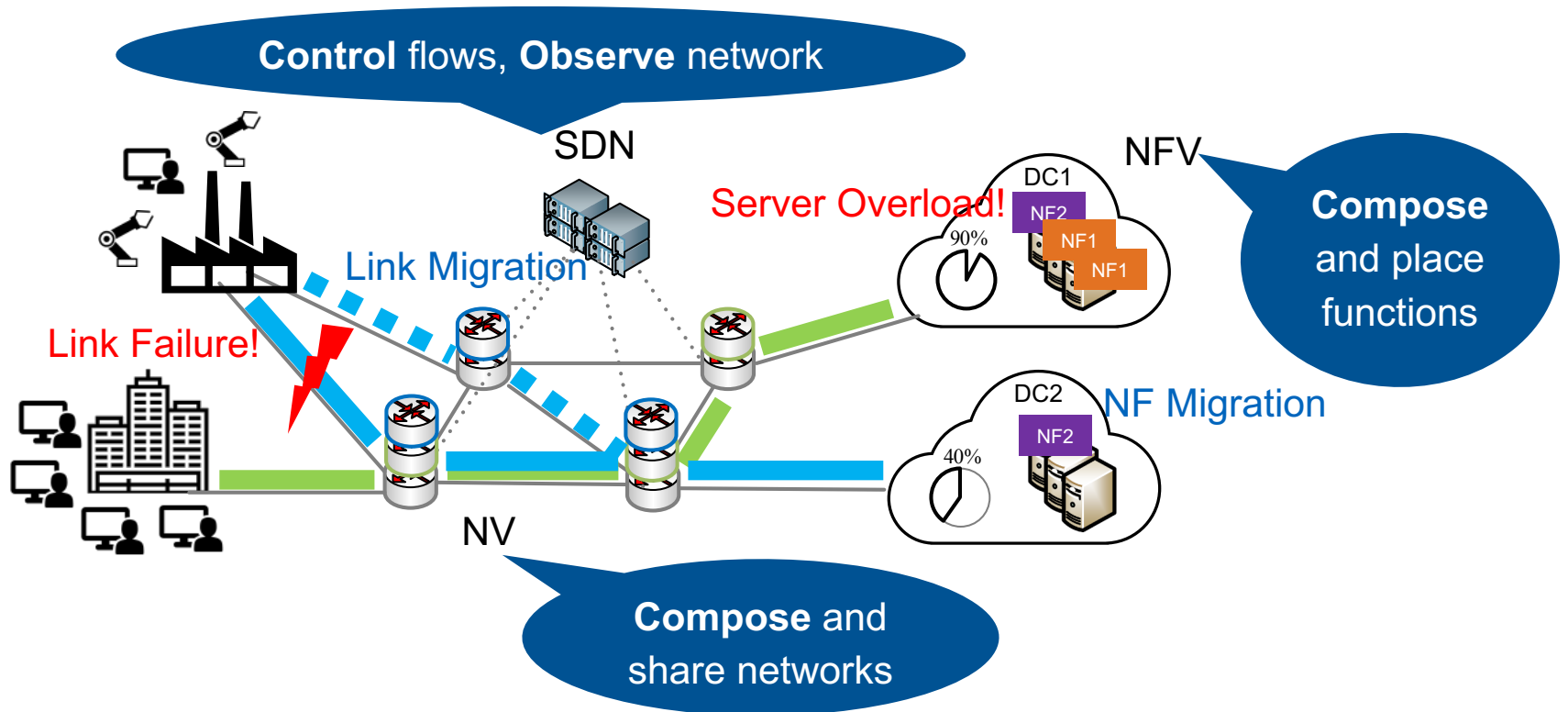
Adaptation: (1) Detection – (2) Decision – (3) Execution

Network Observation
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Composition (NFV, NV)

Control (SDN, NV)

Adaptation phases meet functional primitives



Adaptation: (1) Detection – (2) Decision – (3) Execution

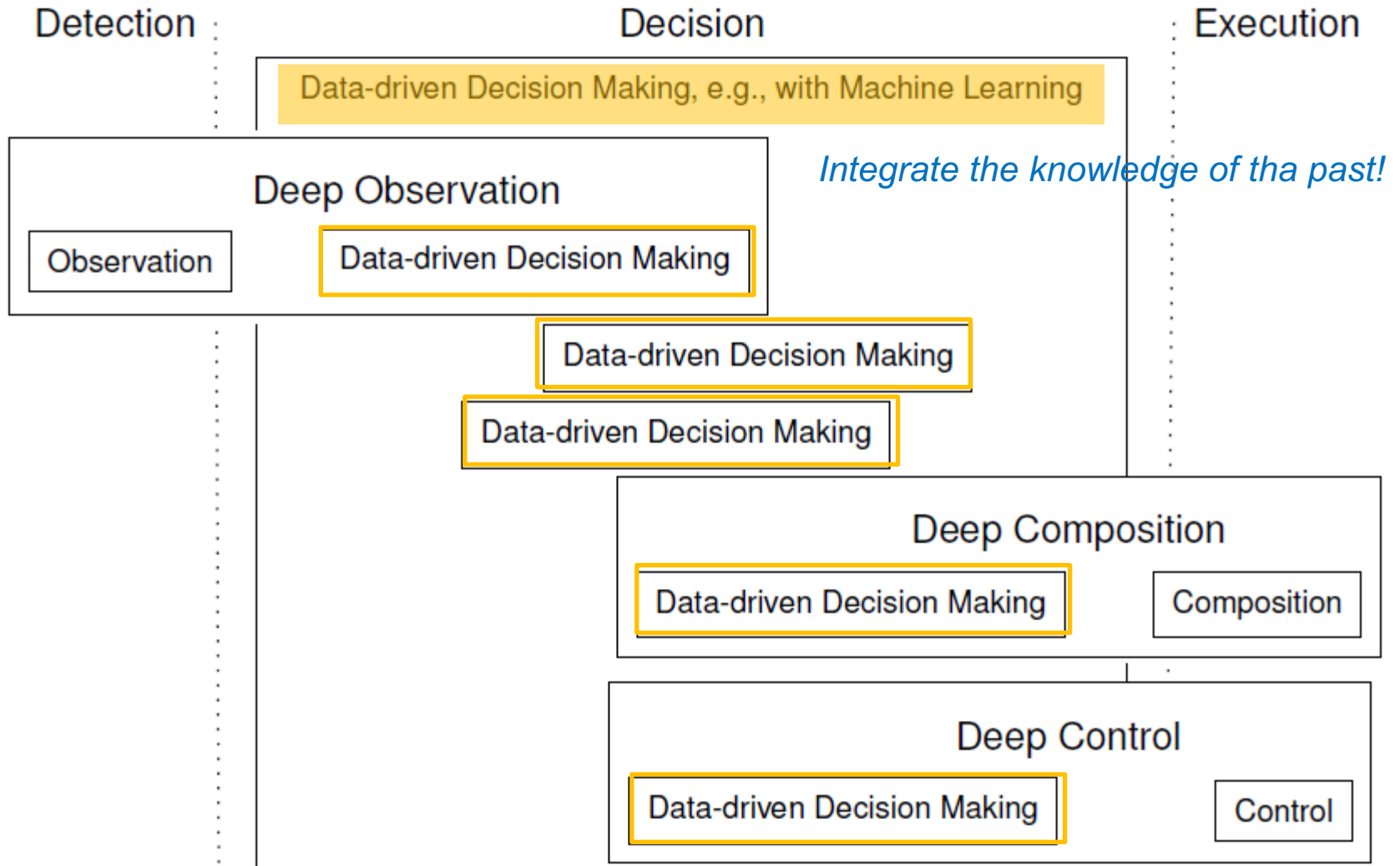
Network Observation
(SDN, NFV)

Composition (NFV, NV)

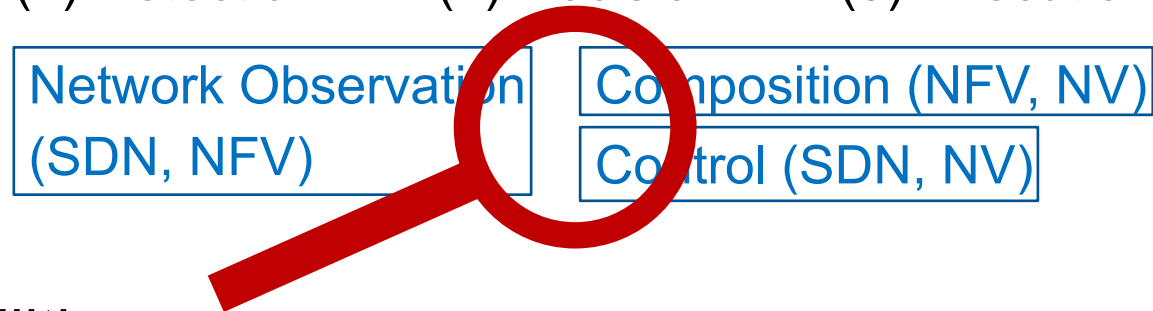
Control (SDN, NV)

challenge:
fill the gap!

Our Proposal: Deep Observation, *Composition* and *Control*



Adaptation: (1) Detection – (2) Decision – (3) Execution



Increased flexibilities

→ *more options*

→ *optimization potential*

→ *optimization problem*

computational complexity

accuracy of models

e.g. constraint-based routing, waypoint routing,...

We need new ways for improving existing network management algorithms!

Data-driven approaches

- *facilitate fast heuristics*
- *adapt and optimize for the actual state of the network (rather than for worst case network state ...)*
- *can consider knowledge of the past*

→ **Machine Learning in softwarized networks**

Machine Learning for Improving SDN/NFV-Related Network Optimization

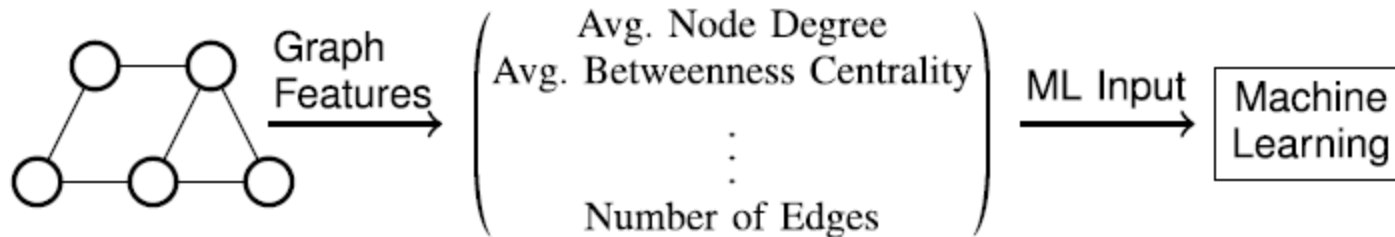
Deep Observation

- Data Acquisition
- Data Representation
 - Dimensionality reduction
 - Representing relational network data
 - Network graph and node transformation

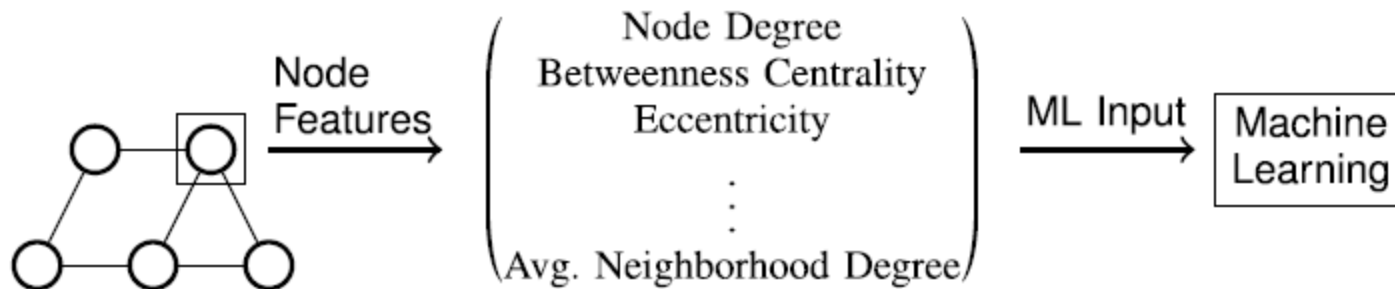
Deep Composition and Deep Control

Network Graph and Node Transformations

From graph to feature vector (topological attributes)



Node feature vector



Machine Learning for Improving SDN/NFV-Related Network Optimization

Deep Observation

- Data Acquisition
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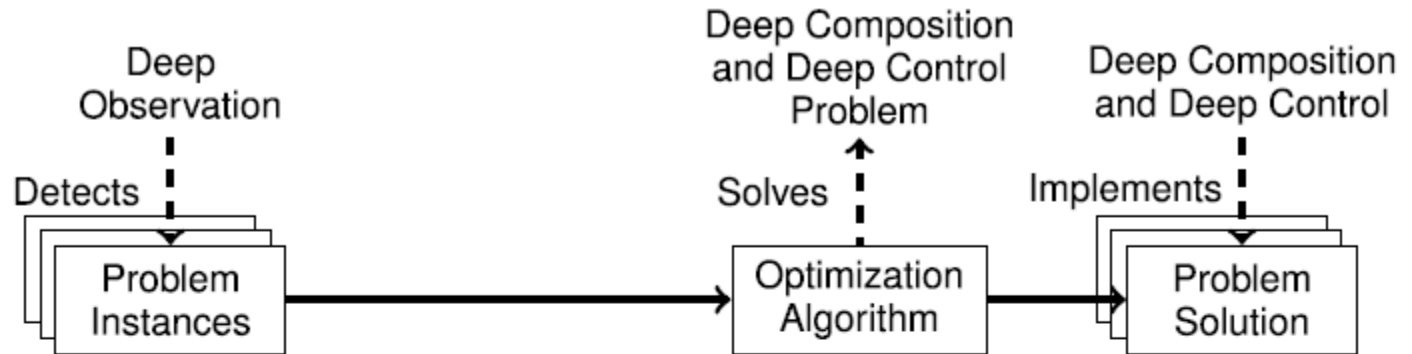
Deep Composition and Deep Control

The boosting approach: find better solutions faster

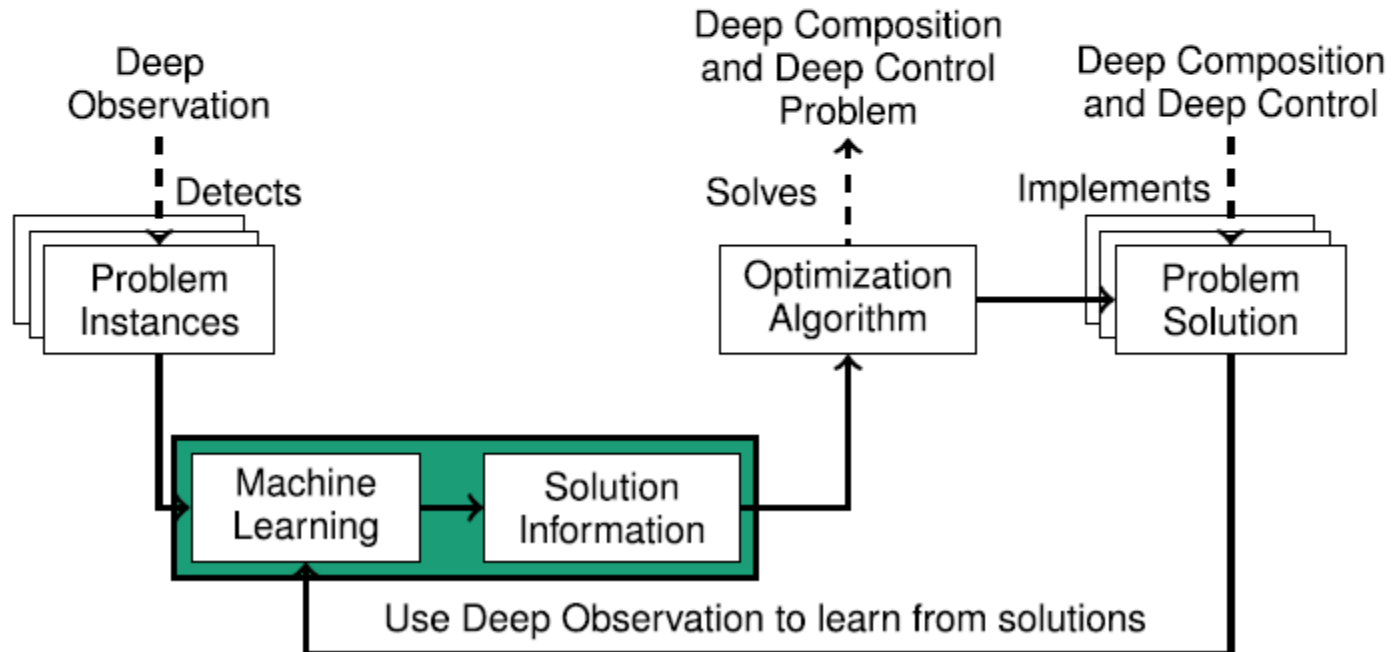
(keep your original optimization algorithm)

Machine Learning-enhanced optimization

Traditional problem optimization

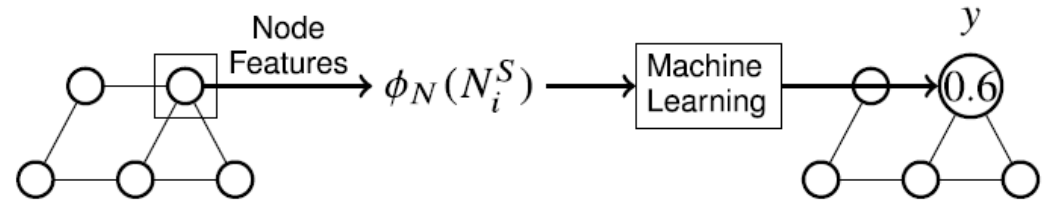


ML-enhanced problem optimization

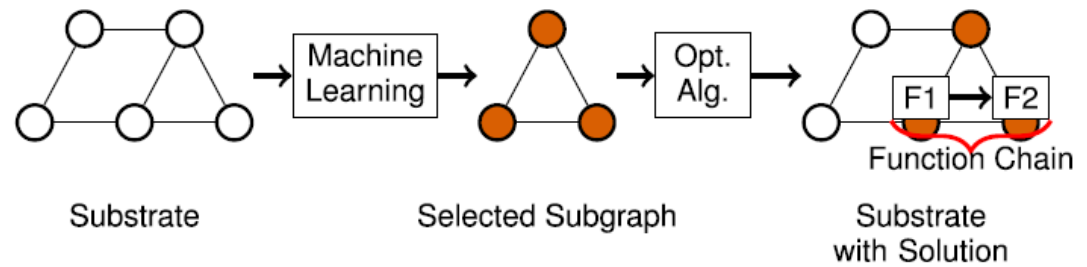


(Problem) Search Space Reduction

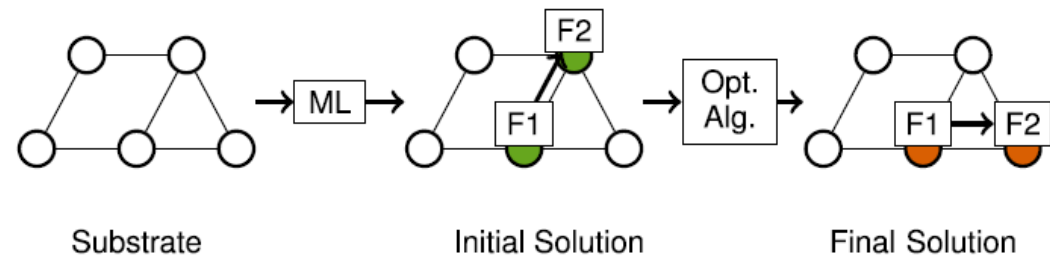
Basis for the
boosting approach



(a)

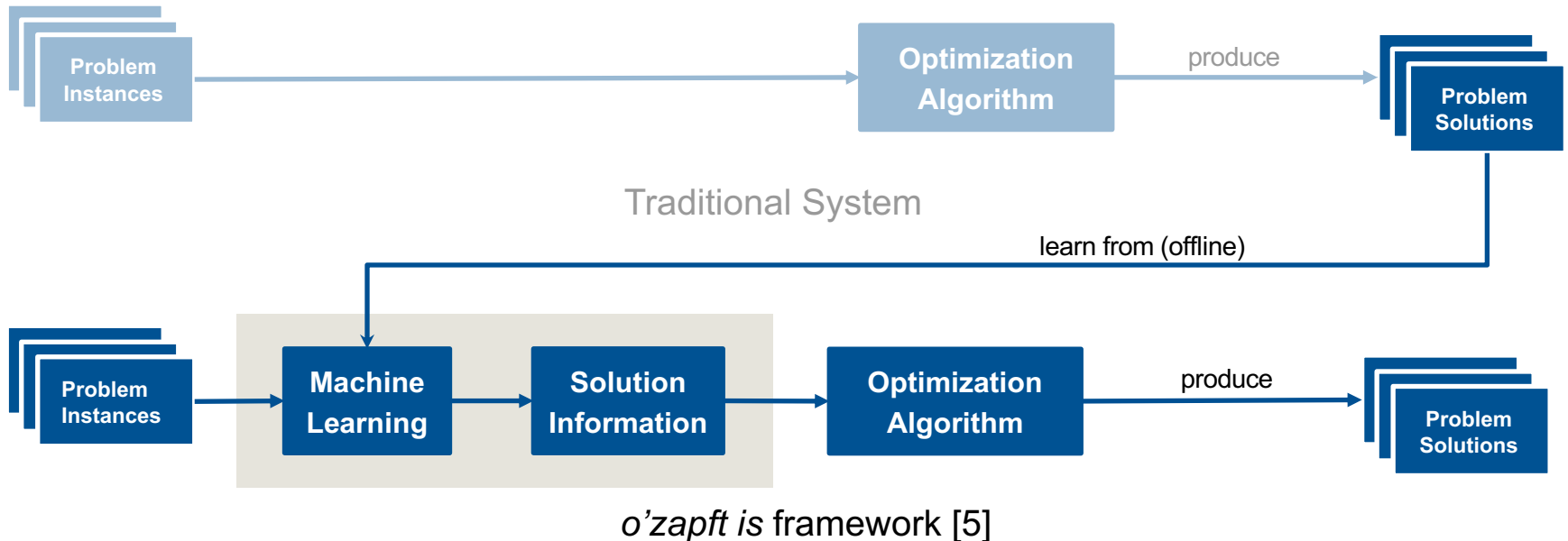


(b)



(c)

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



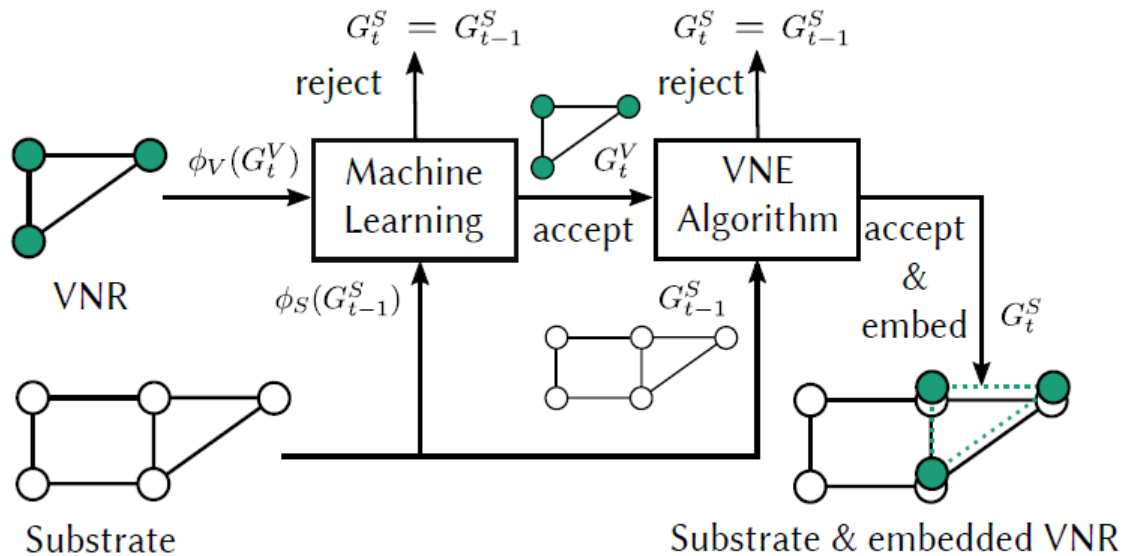
State-of-the-art: Neglects produced data!

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.

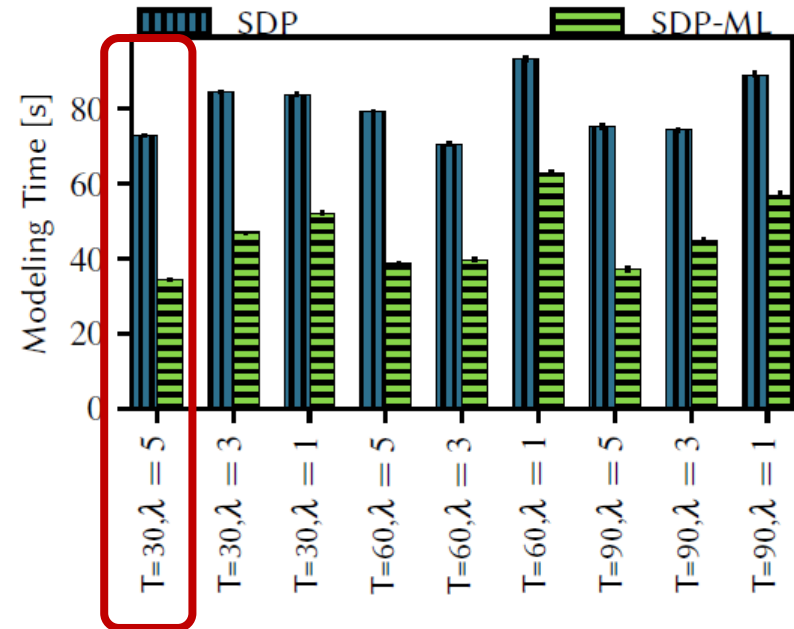
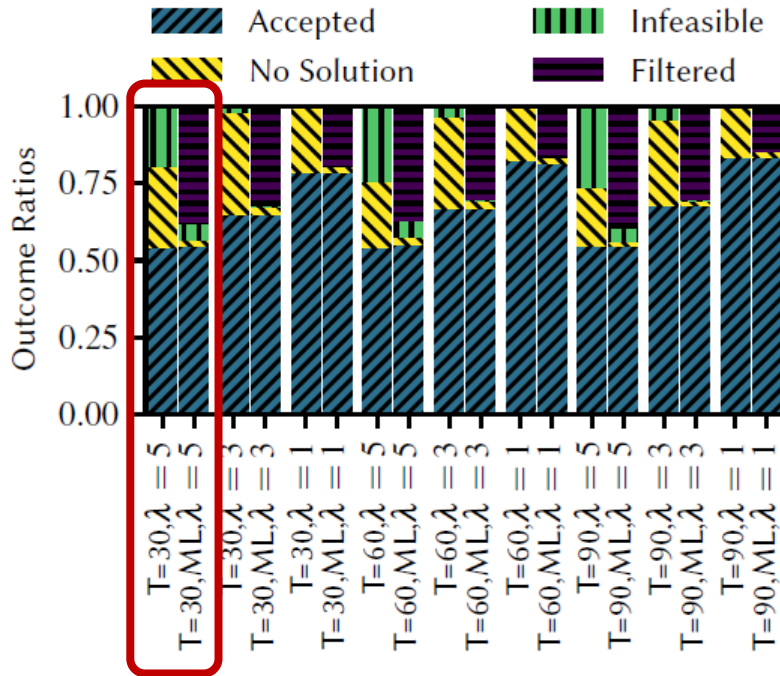
Data Available: P. Kalmbach, J. Zerwas, M. Manhart, A. Blenk, S. Schmid, W. Kellerer. Data on "o'zapft is Tap Your Network Algorithm's Big Data!", 2017 <https://doi.org/10.14459/2017md1361589>

Case Study: Predicting Acceptance Probabilities of VNE Requests



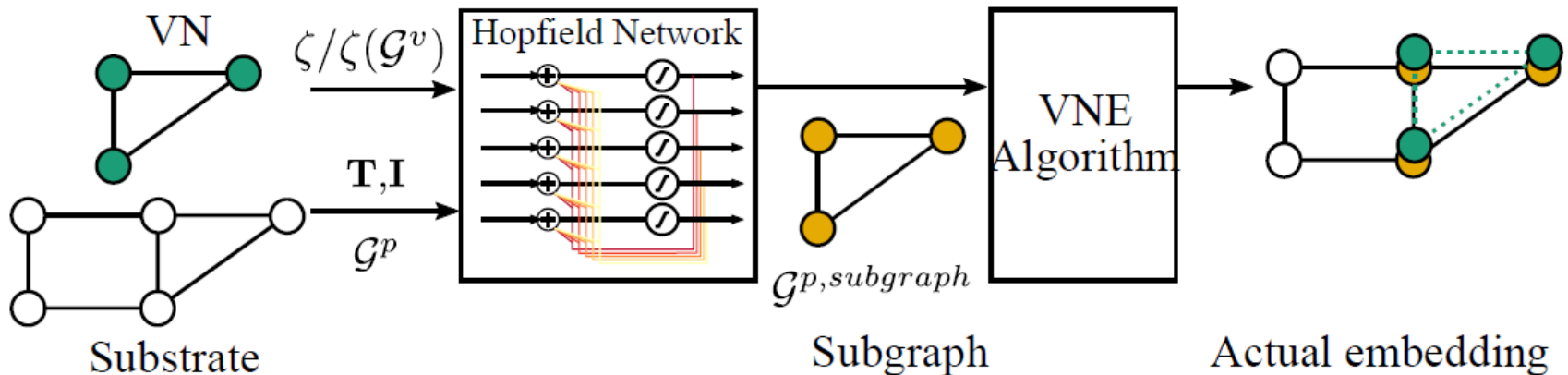
- Supervised learning: use data with accepted and rejected requests! Offline training!
- Recurrent neural network (RNN) for classification
- Filter infeasible and requests with unacceptable algorithm runtime (“no solution“)

Can we speed-up optimal algorithms using admission control?



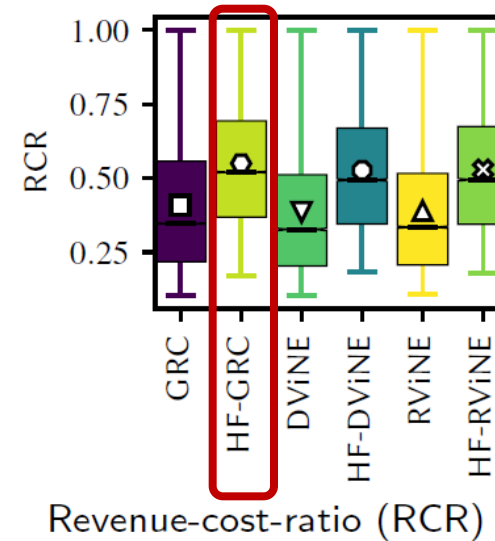
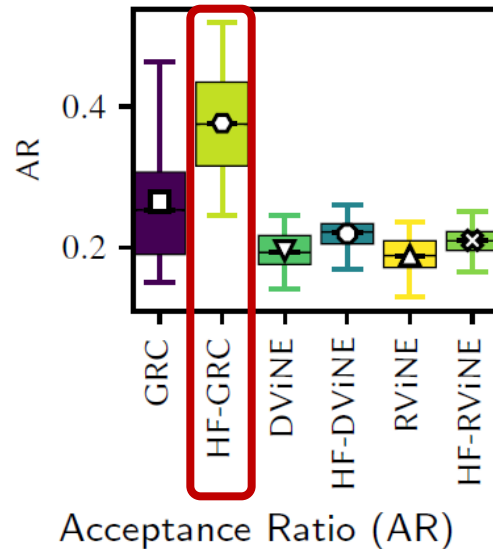
Efficient Filtering of infeasible and unacceptable requests
 Efficient saving of model creation time

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**

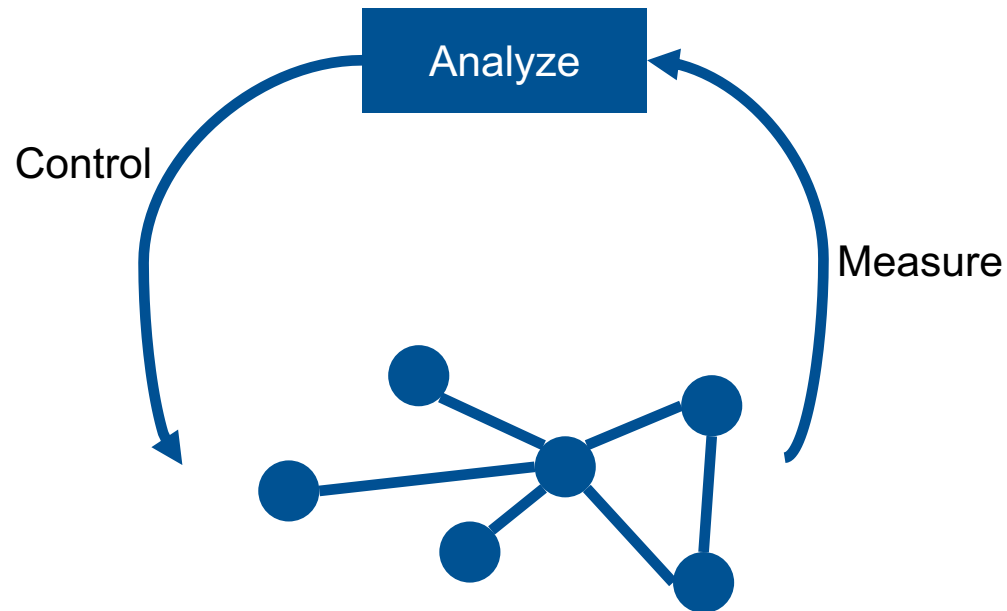
Neurovine: Efficiency on Real Network Topologies



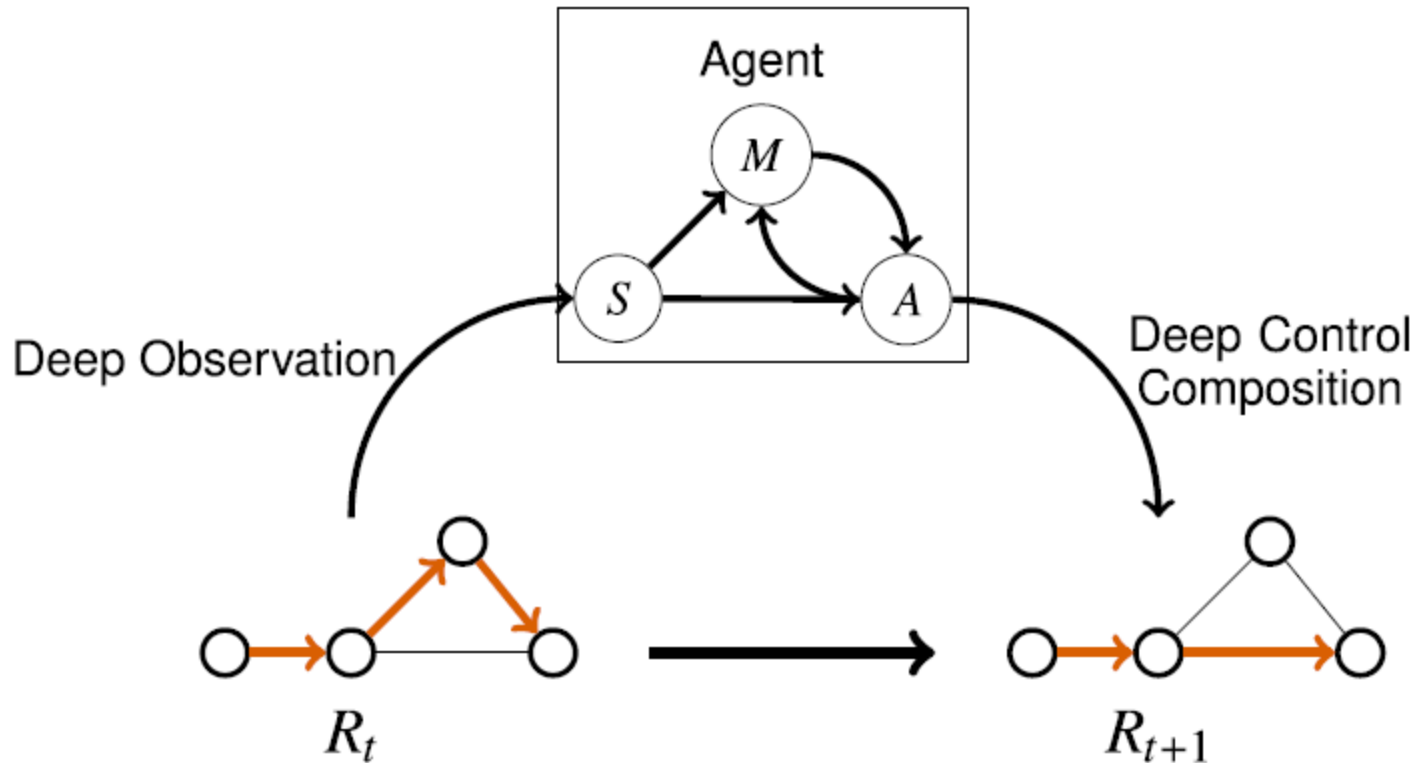
- VNE algorithms (GRC, DViNE, RViNE) vs. Hopfield variants (HF-GRC, HF-DViNE, HF-RViNE)
- NeuroViNE accepts more networks with less costs

Towards Autonomous Networks

Network Managers' all-time Dream:
lean back and watch!

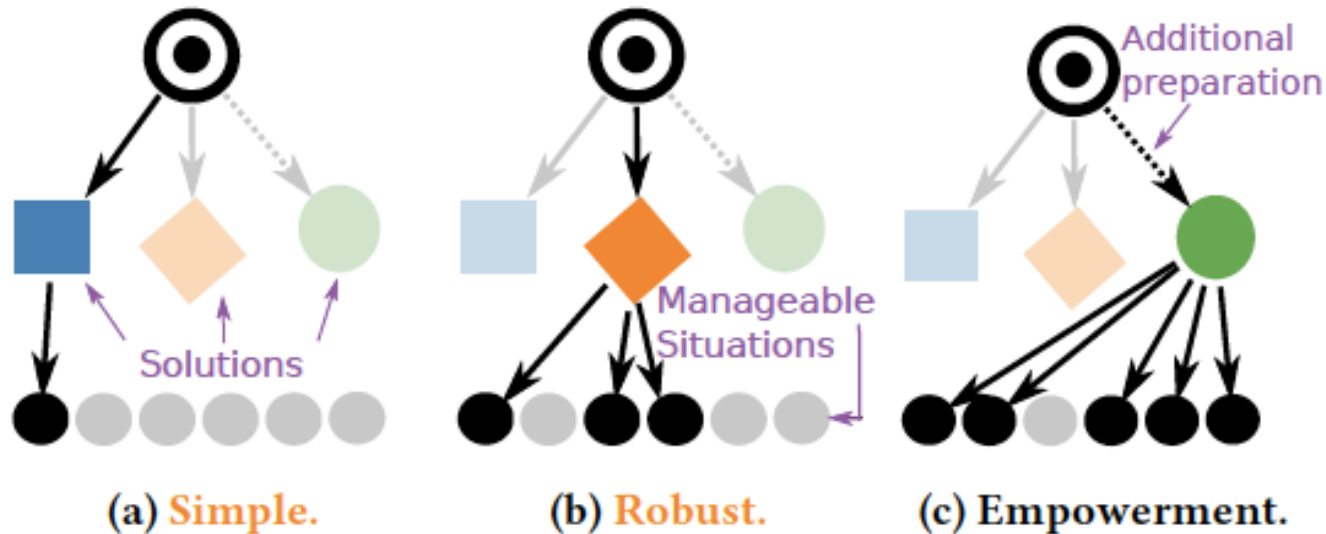


Proposed concept: Empowerment



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)



- Softwarized Networks provide the adaptability required for future dependable and flexible communication networks
- *Adaptation(netsoft) = observation, composition and control*
- Challenges: more data and more options → algorithmic nature
- Benefits from data-driven decision making based on ML
- First step towards self-driving networks

8 Research Challenges to take home

- 1) Adaptation is a process over time. How to consider time constraints?
- 2) At which scale of network sizes do the new concepts provide improvements for adaptation?
- 3) Adaptation functionalities provided by the control plane or by the data plane?
- 4) Can adaptation be supported by hardware (P4?) or purely by software?
- 5) How can system performance be guaranteed based on adaptation?
- 6) How to measure the adaptation potential in relation to the provided flexibility?
- 7) How can Machine Learning be supported by hardware?
- 8) How far can we go with empowerment?



FlexNets

join us on **networkflexibility.org**

find our research videos on YouTube Channel “LKN TUM”

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Some References



W. Kellerer, P. Kalmbach, A. Blank, A. Basta, S. Schmid, M. Reisslein: *Adaptable and Data-Driven Softwarized Networks: Review, Opportunities and Challenges*. **Proc. of the IEEE**, 2019 (open access).

M. Klügel, M. He, W. Kellerer, P. Babarczi: *A Mathematical Measure for Flexibility in Communication Networks*. **IFIP NETWORKING 2019** (to appear).

M. He, A. Martinez Alba, A. Basta, A. Blenk, W. Kellerer. *Flexibility in Softwarized Networks: Classifications and Research Challenges*. **IEEE Communication Surveys & Tutorials**, 2019.

P. Kalmbach, J. Zerwas, P. Babarczi, A. Blenk, W. Kellerer, S. Schmid, *Empowering Self-Driving Networks*. **ACM SIGCOMM 2018 Workshop** on Self-Driving Networks - SelfDN 2018

A. Blenk, P. Kalmbach, J. Zerwas, M. Jarschel, S. Schmid, W. Kellerer: *NeuroViNE: A Neural Preprocessor for Your Virtual Network Embedding Algorithm*. **IEEE INFOCOM 2018** (main conference), Honolulu, HI, USA, April 15-19, 2018.

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

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

M. He, A. Basta, A. Blenk, W. Kellerer, *How Flexible is Dynamic SDN Control Plane?*, **IEEE INFOCOM Workshop**, SWFAN'17, Atlanta, USA, May 2017.

W. Kellerer, A. Basta, A. Blenk, *Using a Flexibility Measure for Network Design Space Analysis of SDN and NFV*, **IEEE INFOCOM Workshop**, SWFAN'16, SF, USA, April 2016.