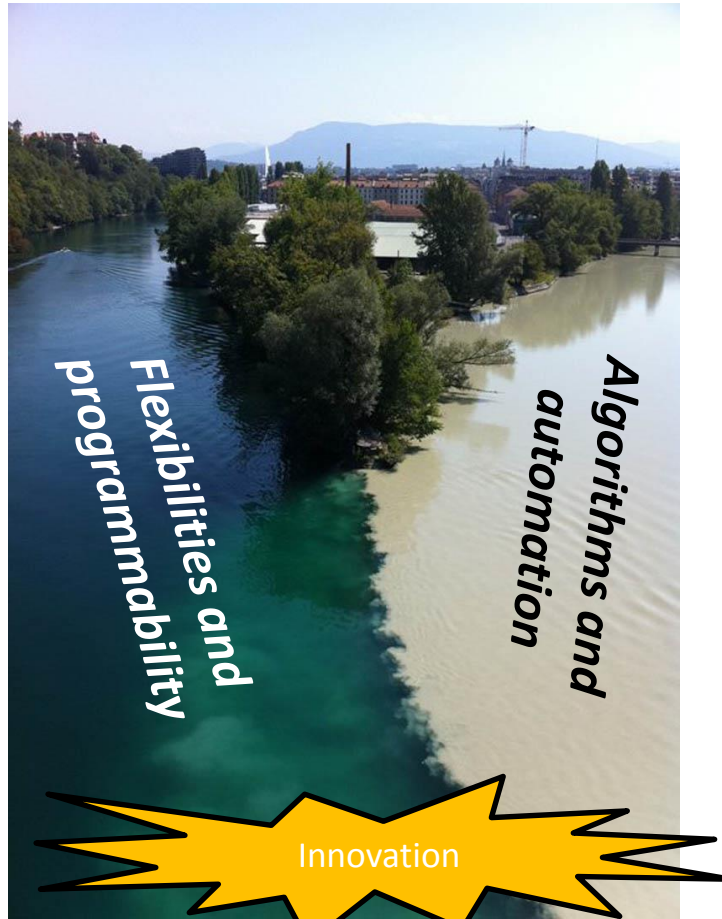


Algorithms for Flexible Networks: Opportunities and Challenges

Stefan Schmid (Uni Vienna)



A Great Time to Be a Networking Researcher!



Rhone and Arve Rivers,
Switzerland

Credits: George Varghese.

Flexibilities: Along 3 Dimensions



Passau, Germany

Inn, Donau, Ilz

Flexibilities: Along 3 Dimensions



Passau, Germany

Inn, Donau, Ilz

Flexibilities: Along 3 Dimensions

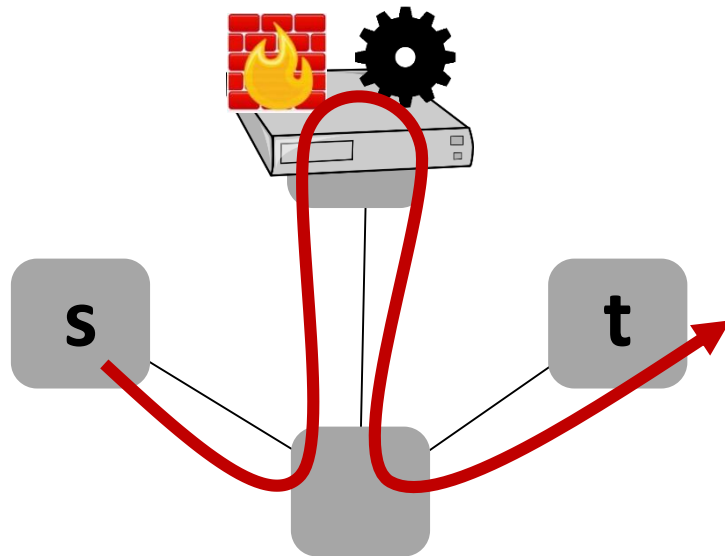


Regensburg, Germany

Donau, Inn, Ilz

Opportunity: Flexible Routing

- **Direct control** over paths
- **Generalized** match-action
- Composing **innovative services**
- Not even simple paths: **walks**!

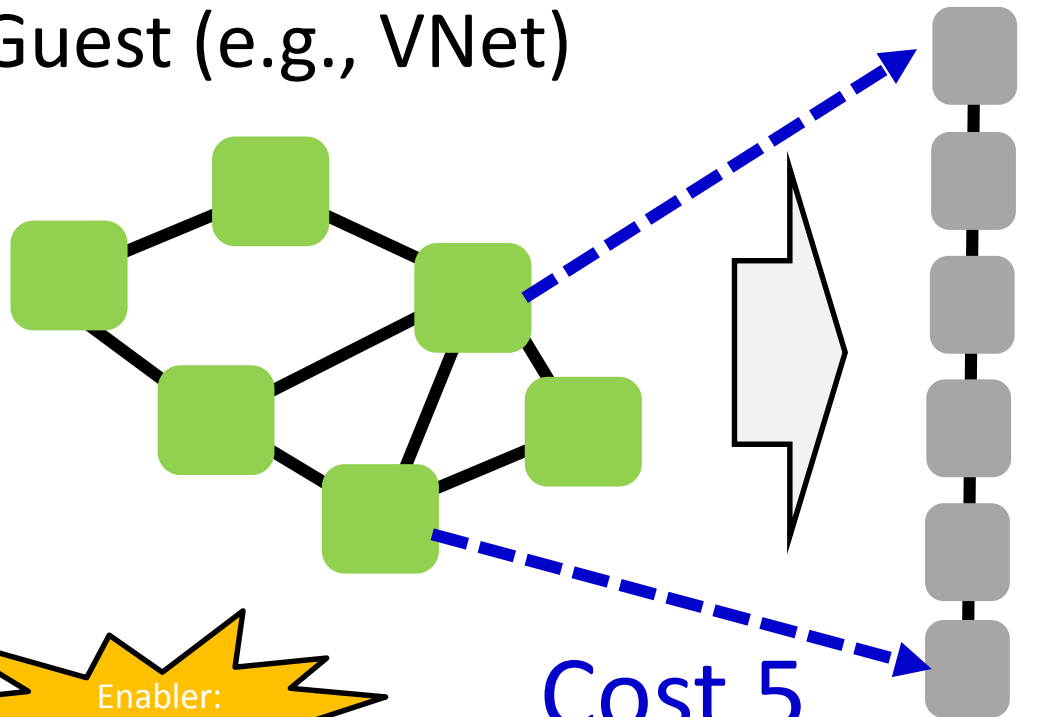


Enabler:
SDN

Charting the Algorithmic Complexity of Waypoint Routing. Amiri et al. ACM SIGCOMM CCR, 2018.

Opportunity: Flexible Embedding

Guest (e.g., VNet)



- Improved **resource allocation**
- Minimize communication paths: lower **latency**, load, etc.

Enabler:
Virtualization

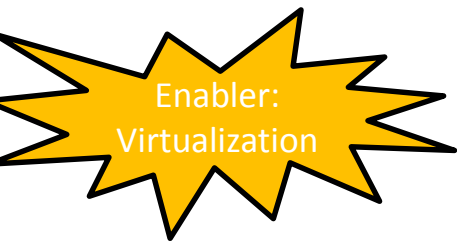
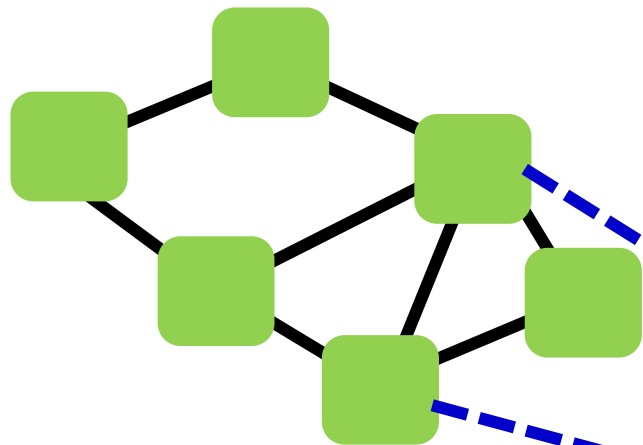
Cost 5

Host

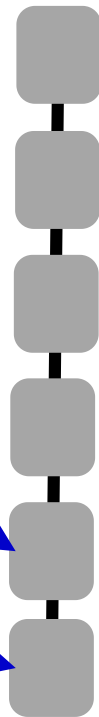
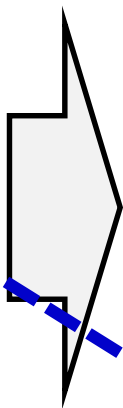
Charting the Complexity Landscape of Virtual Network Embeddings. Rost et al. IFIP Networking, 2018.

Opportunity: Flexible Embedding

Guest (e.g., VNet)



Cost 1

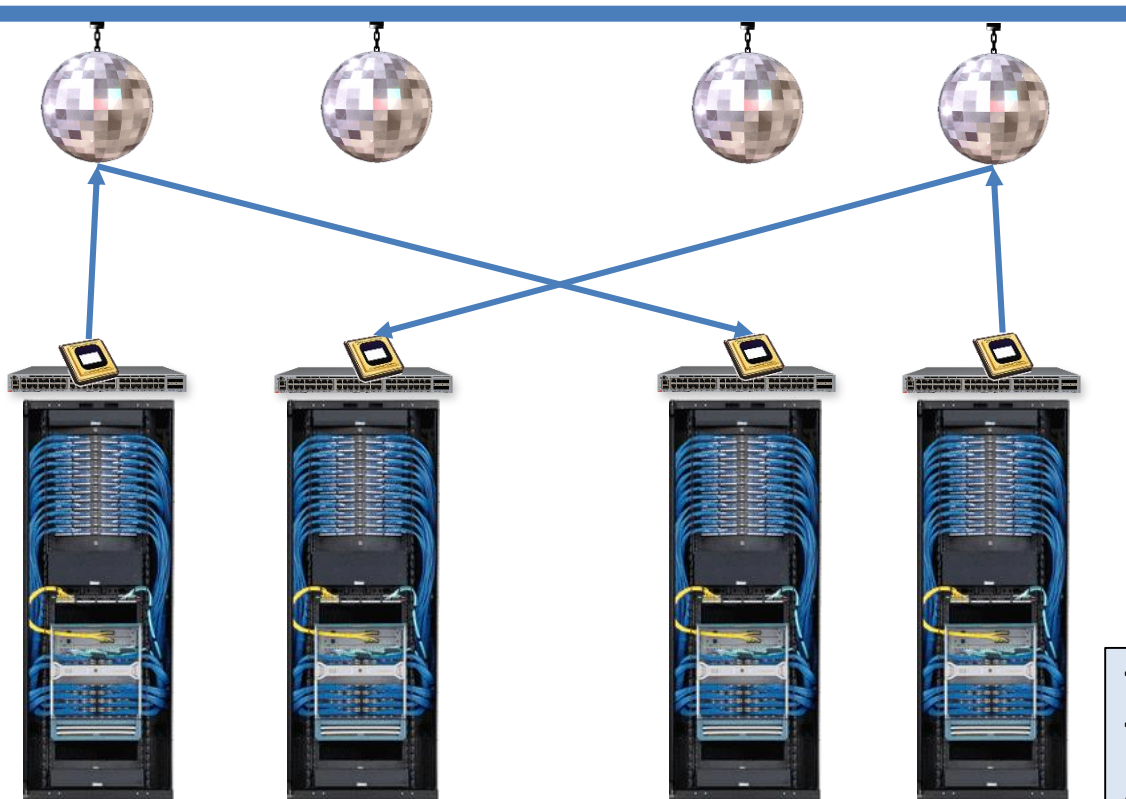


Host

- Improved **resource allocation**
- Minimize communication paths: lower **latency**, load, etc.

Charting the Complexity Landscape of Virtual Network Embeddings. Rost et al. IFIP Networking, 2018.

Opportunity: Flexible Topology Programming

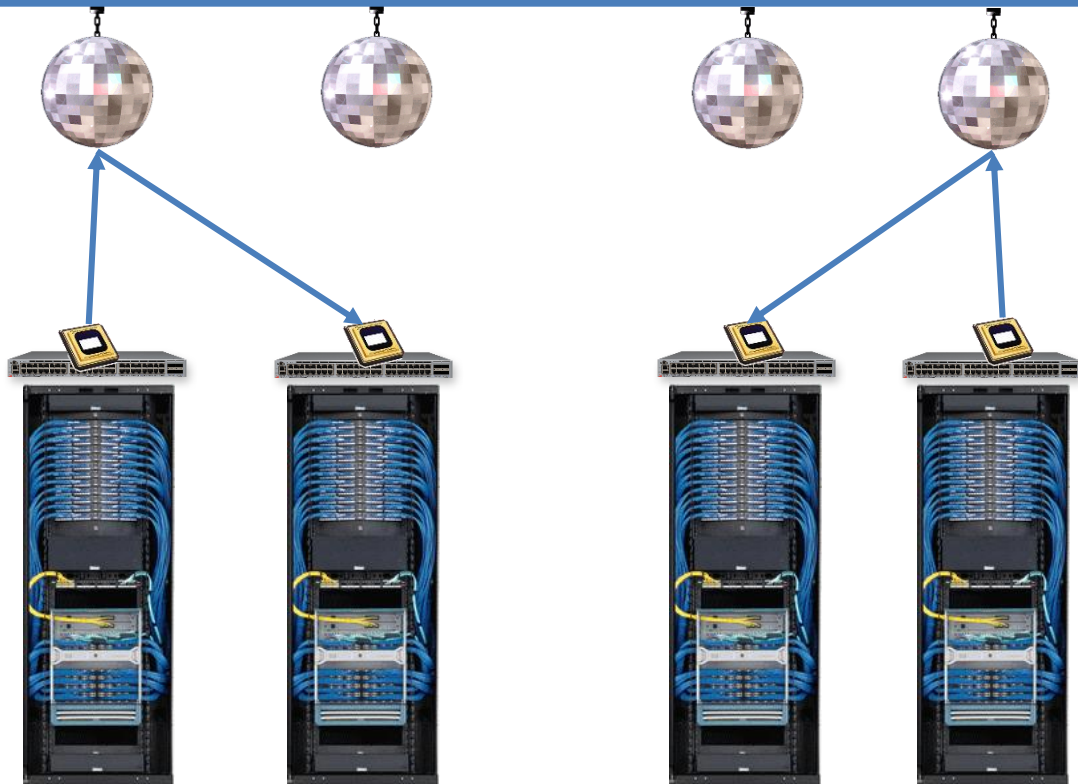


- **Reconfigure** networks towards needs

Enabler: Free-space optics

Toward Demand-Aware Networking: A Theory for Self-Adjusting Networks.
Avin et al. ACM SIGCOMM CCR, 2018.

Opportunity: Flexible Topology Programming



- **Reconfigure** networks towards needs

Enabler: Free-space optics

Toward Demand-Aware Networking: A Theory for Self-Adjusting Networks.
Avin et al. ACM SIGCOMM CCR, 2018.

Opportunity



Additional **dimensions for optimization**: can be exploited to improve performance, utilization, ...



New network **services** (e.g., service chaining)

Challenge



But: optimizations become **harder** and are sometimes not yet well-understood (e.g., embedding, topology programming)

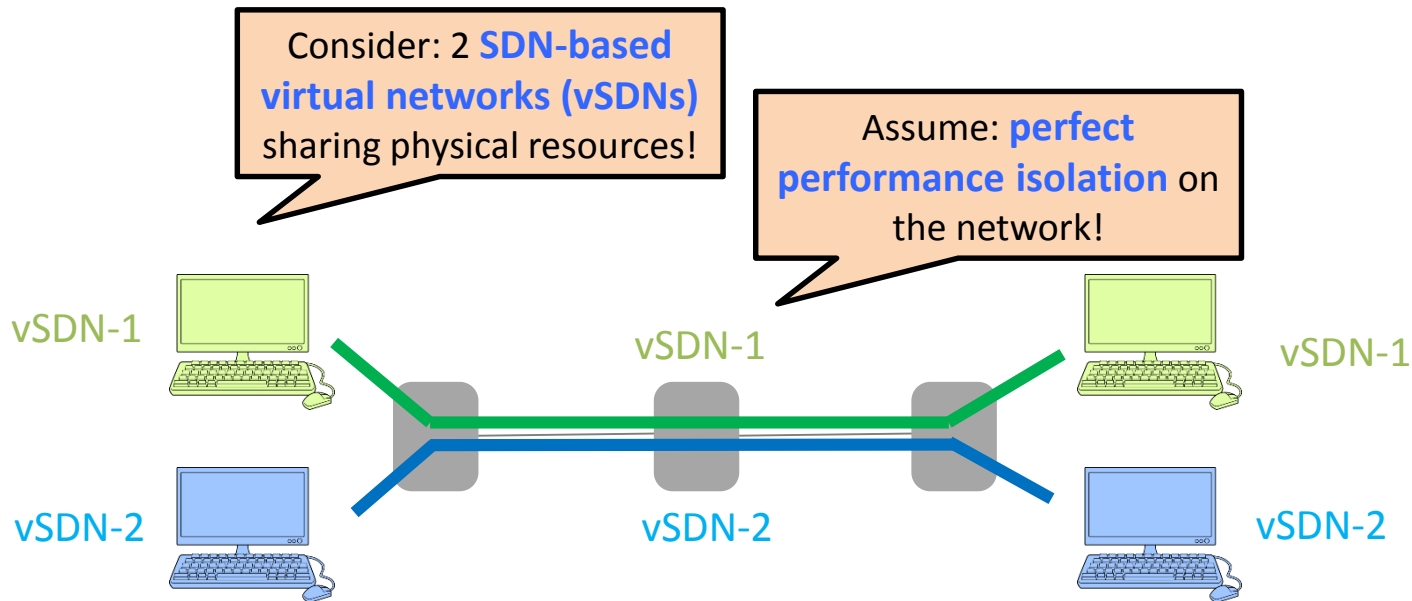
Another Challenge: Model vs Reality 😊

You: I invented a great new algorithm to route and embed service chains at low resource cost and providing minimal bandwidth guarantees!

Boss: So can I promise our customers a predictable performance?

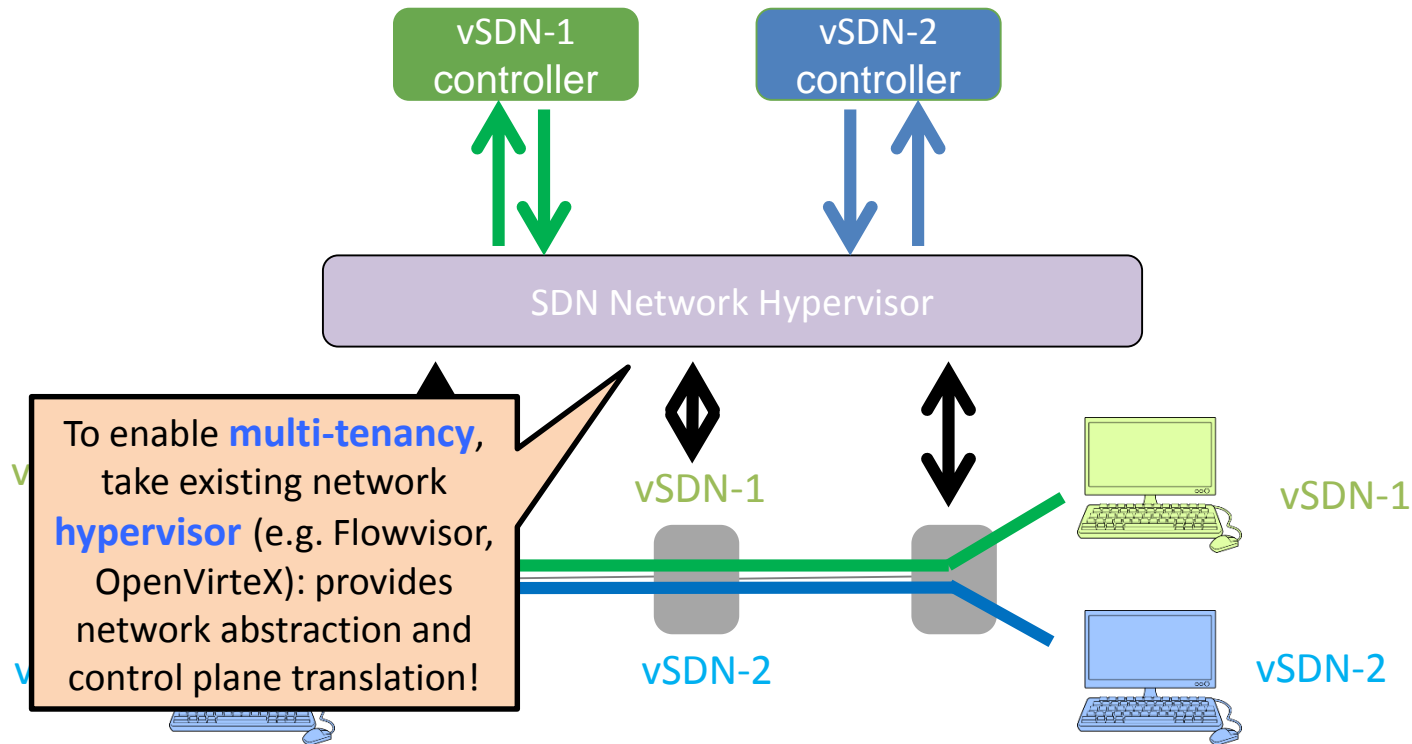
You: hmm...

Predictable Performance in SDNs?



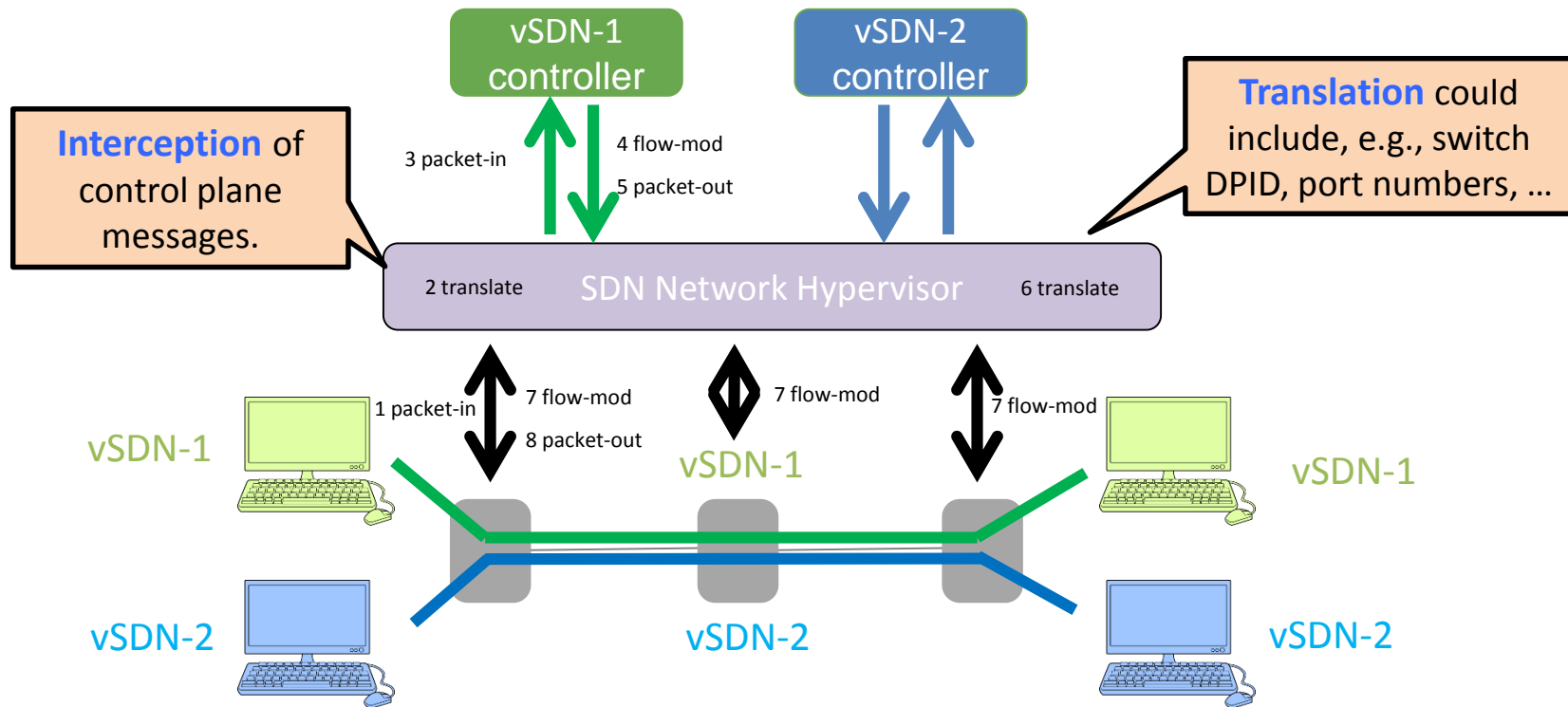
An Experiment: 2 vSDNs with bw guarantee!

Predictable Performance in SDNs?



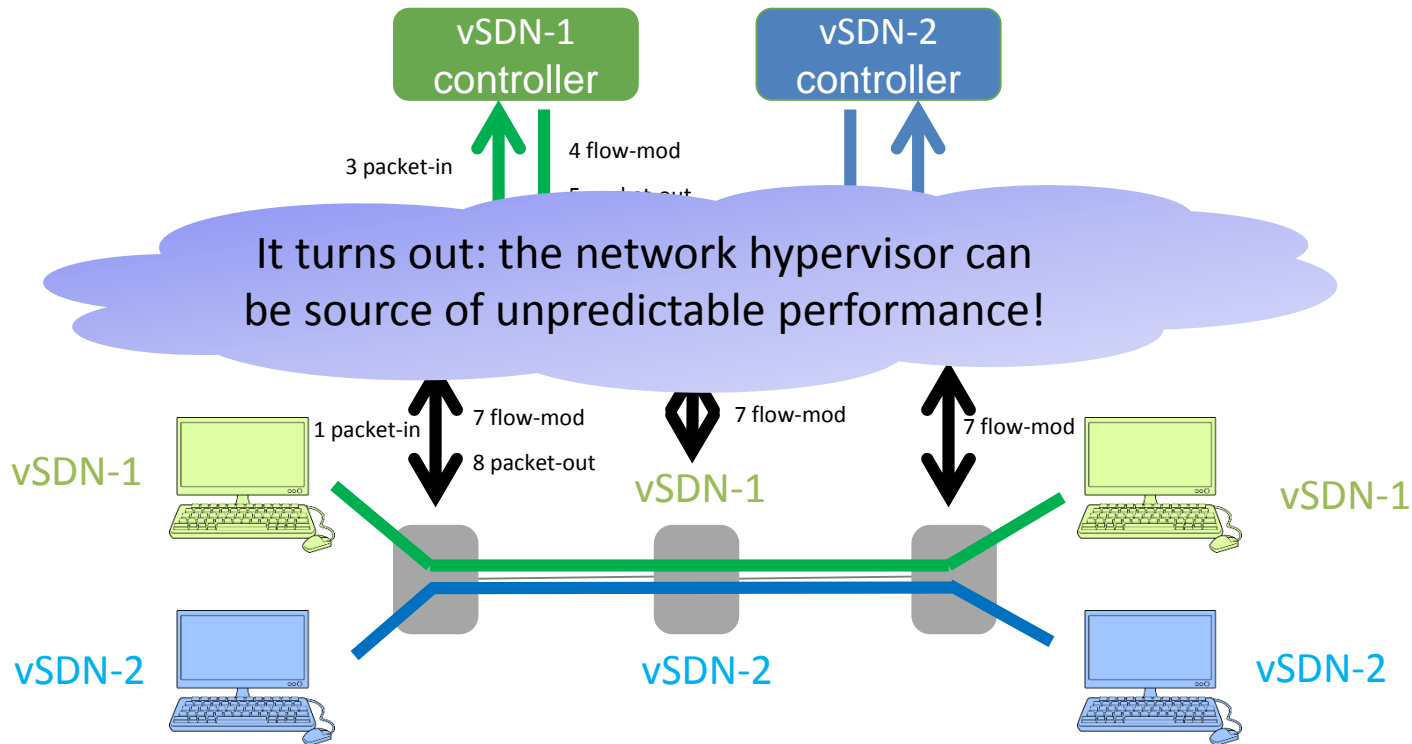
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Predictable Performance in SDNs?



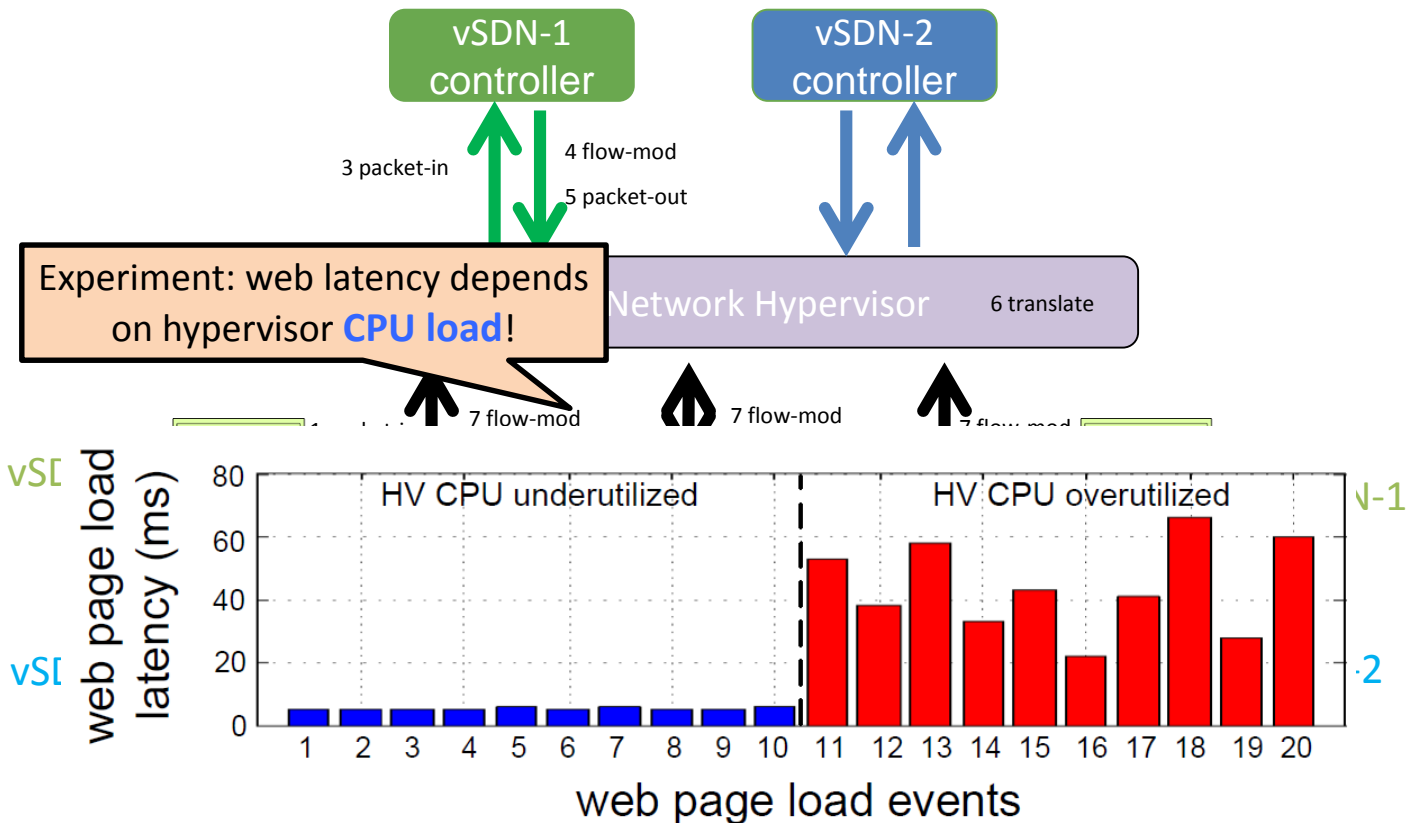
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Predictable Performance in SDNs?

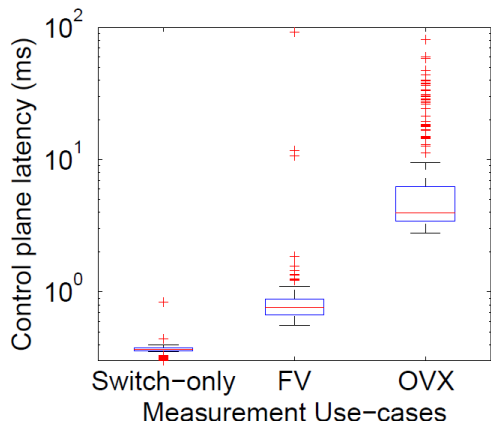


An Experiment: 2 vSDNs with bw guarantee!

Predictable Performance in SDNs?

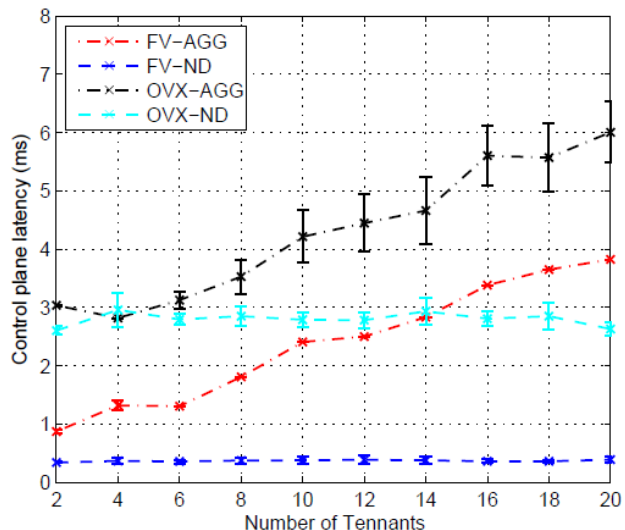


Predictable Performance in SDNs?



**Performance also depends
on hypervisor **type**...**
(*multithreaded or not, which version
of Nagle's algorithm, etc.*)

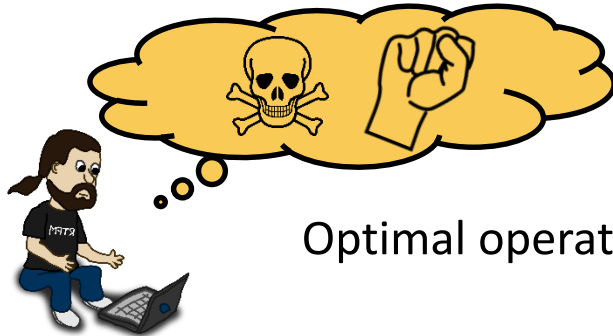
... **number** of tenants...



**On The Impact of the Network Hypervisor on
Virtual Network Performance.** Blenk et al. *IFIP
Networking*, 2019.

First Conclusions

- Exploiting network flexibilities is non-trivial, especially if **fine-grained** and **fast** reactions are desired
- Also modelling such networked systems is challenging: details of **interference**, **demand**, etc. will only be available at runtime



Optimal operation of flexible networks ***too complex for humans.***



Let's give up control: self-* networks!

Self-observing, self-adjusting, self-repairing, self-driving, ...

It's about
automation!

Roadmap

- Opportunities of self-* networks
 - Example 1: Demand-aware, self-adjusting networks
 - Example 2: Self-repairing networks
- Challenges of designing self-* networks



Roadmap

- Opportunities of self-* networks
 - **Example 1: Demand-aware, self-adjusting networks**
 - Example 2: Self-repairing networks
- Challenges of designing self-* networks



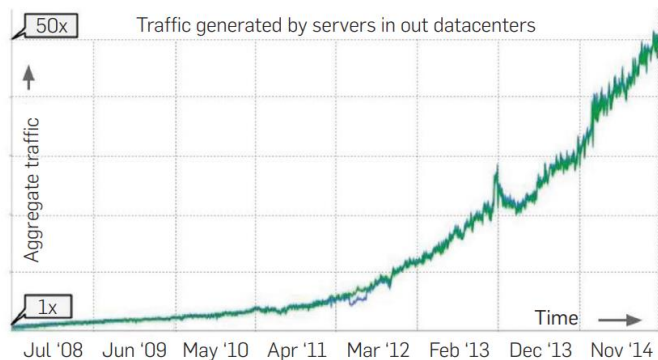
Why Demand-Aware...?

Case study: data-center networks

Explosive Growth of Demand...

... But Much Structure!

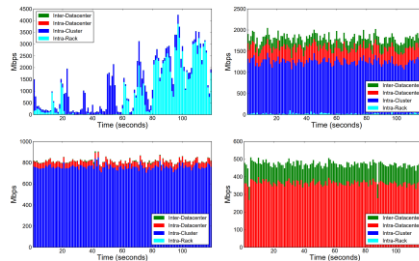
Batch processing, web services, **distributed ML**, ...: **data-centric applications** are distributed and interconnecting network is **critical**



Source: Jupiter Rising. SIGCOMM 2015.

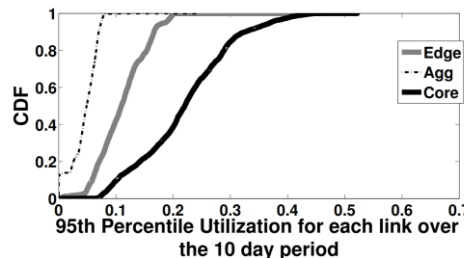
Aggregate server traffic in
Google's datacenter fleet

Facebook



Inside the Social Network's
(Datacenter) Network @
SIGCOMM 2015

Benson et al.

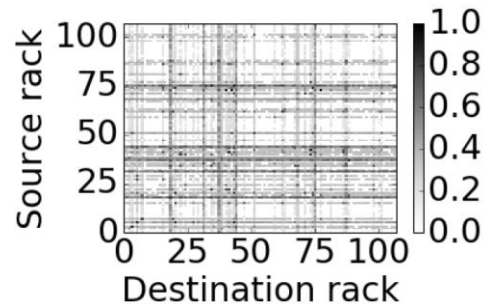


Understanding Data Center Traffic
Characteristics @ WREN 2009



Spatial (**sparse!**) and
temporal **locality**

Microsoft



ProjecToR @ SIGCOMM 2016

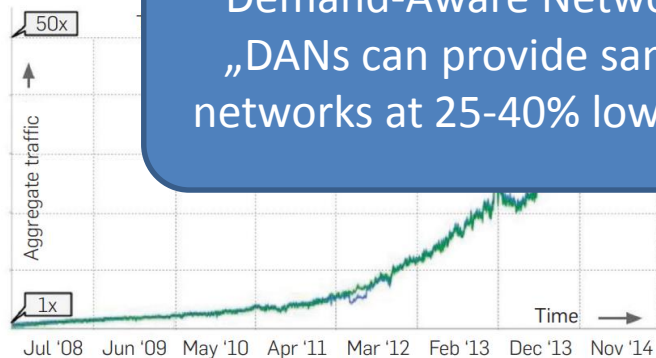
Explosive Growth of Demand...

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Batch processing, web services, **distributed ML**, ...: **data-centric applications** are distributed and interconnecting network is **critical**



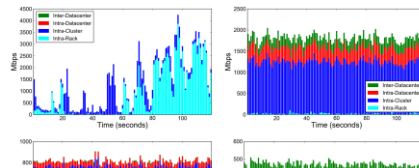
Demand-Aware Networks (DANs) can exploit this structure:
„DANs can provide same performance as demand-oblivious networks at 25-40% lower costs.“ Firefly, SIGCOMM CCR, 2014.



Source: Jupiter Rising. SIGCOMM 2015.

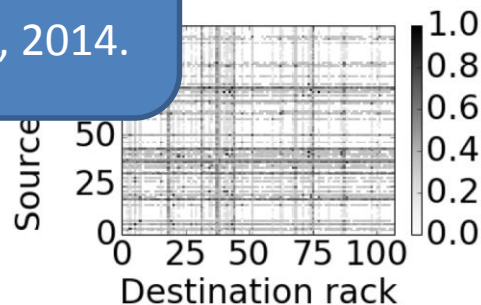
Aggregate server traffic in
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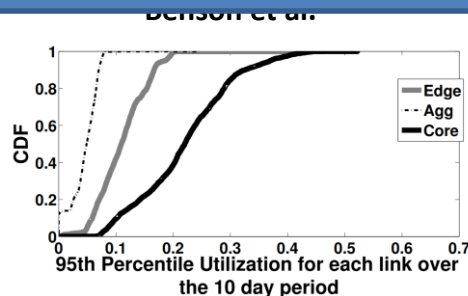


... Spatial (**sparse!**) and
temporal **locality**

Microsoft



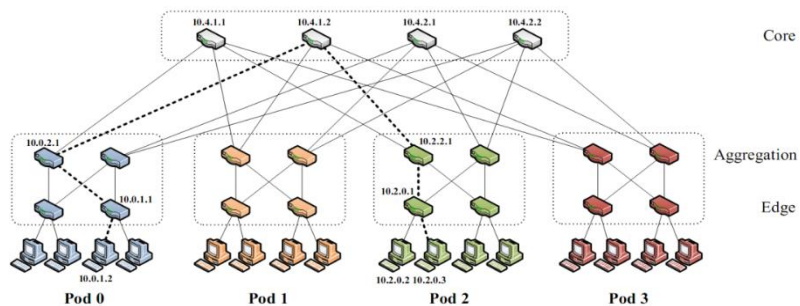
ProjecToR @ SIGCOMM 2016



Understanding Data Center Traffic
Characteristics @ WREN 2009

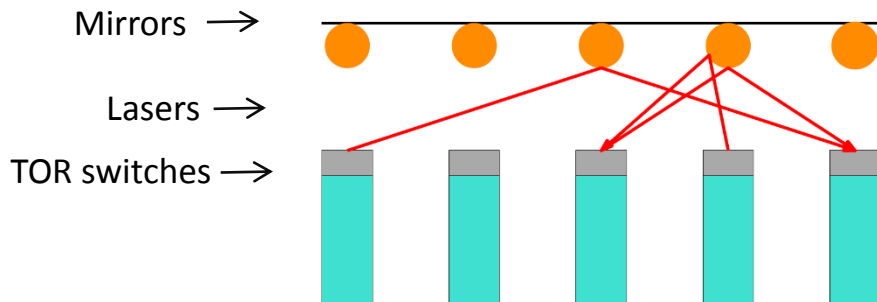
Example: Demand-Aware Topology

Traditional datacenter network



- Usually optimized *for the “worst-case”* (all-to-all communication)
- Example, fat-tree topologies: provide full bisection bandwidth

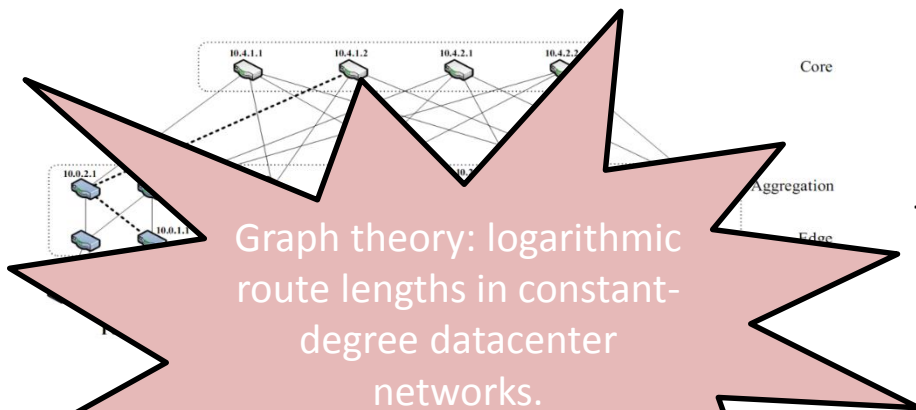
Reconfigurable datacenter network



- Optimized *toward the workload* it serves (e.g., route length)
- Statically or *even dynamically*

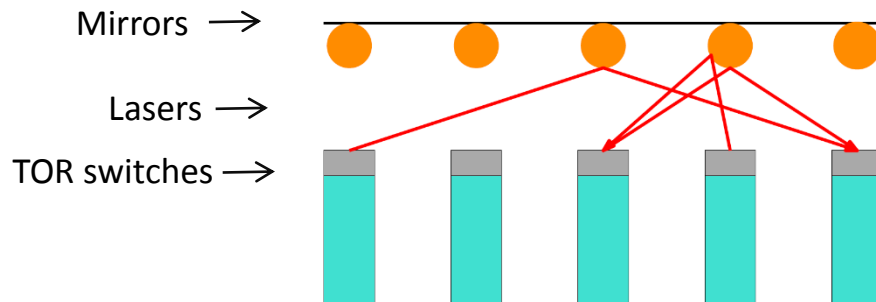
Example: Demand-Aware Topology

Traditional datacenter network



- Usual **case** (all-to-all communication)
- Example, fat tree topologies: provide **full bisection bandwidth**

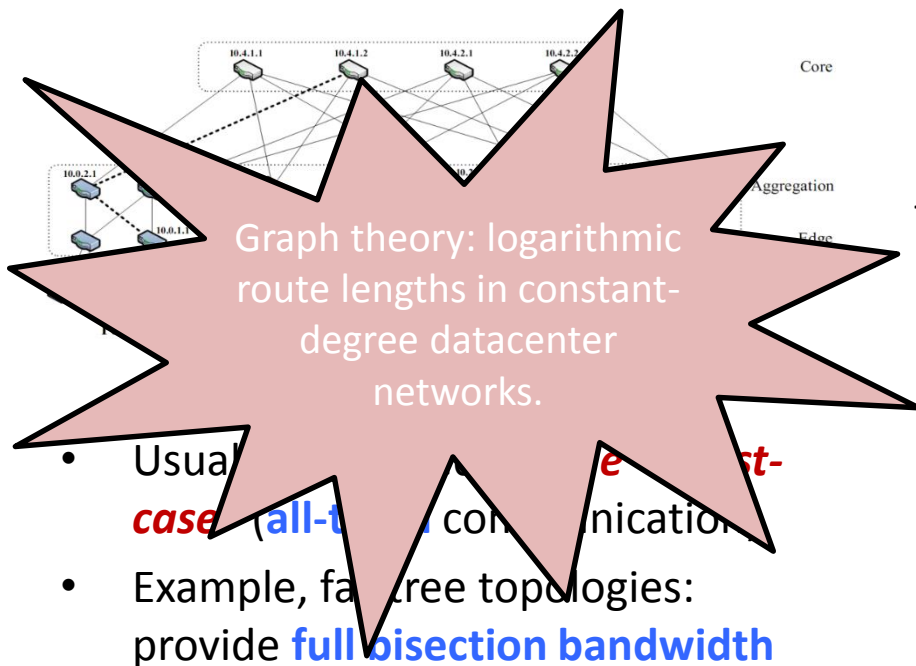
Reconfigurable datacenter network



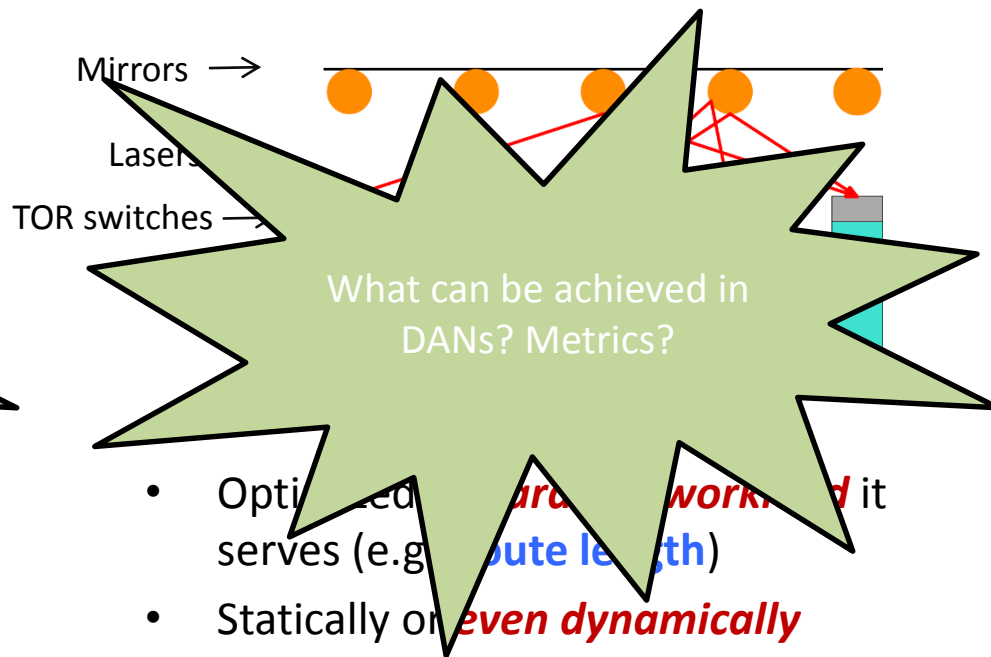
- Optimized **toward the workload** it serves (e.g., **route length**)
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Reconfigurable datacenter network



DAN Design: New Types of Problems

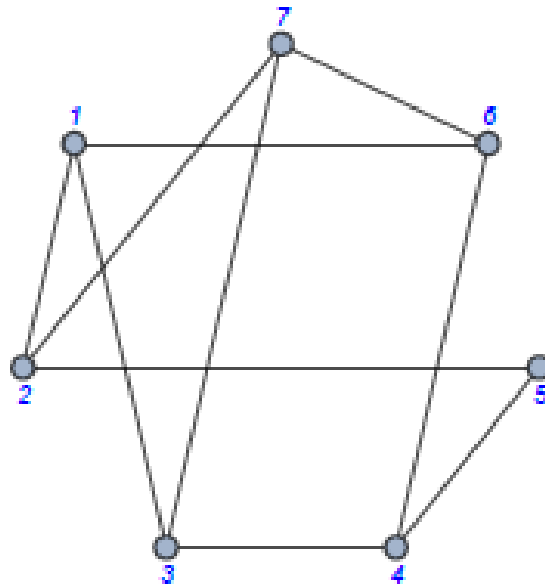
Input: Workload

Destinations

Sources	Destinations						
	1	2	3	4	5	6	7
	1	0	$\frac{2}{65}$	$\frac{1}{13}$	$\frac{1}{65}$	$\frac{1}{65}$	$\frac{2}{65}$
	2	$\frac{2}{65}$	0	$\frac{1}{65}$	0	0	$\frac{2}{65}$
	3	$\frac{1}{13}$	$\frac{1}{65}$	0	$\frac{2}{65}$	0	$\frac{1}{13}$
	4	$\frac{1}{65}$	0	$\frac{2}{65}$	0	$\frac{4}{65}$	0
	5	$\frac{1}{65}$	0	$\frac{3}{65}$	$\frac{4}{65}$	0	0
	6	$\frac{2}{65}$	0	0	0	0	$\frac{3}{65}$
7	$\frac{3}{65}$	$\frac{2}{65}$	$\frac{1}{13}$	0	0	$\frac{3}{65}$	0

design

Output: DAN



Demand matrix: joint distribution

... of **constant degree** (scalability)

DAN Design: New Types of Problems

Input: Workload

Destinations

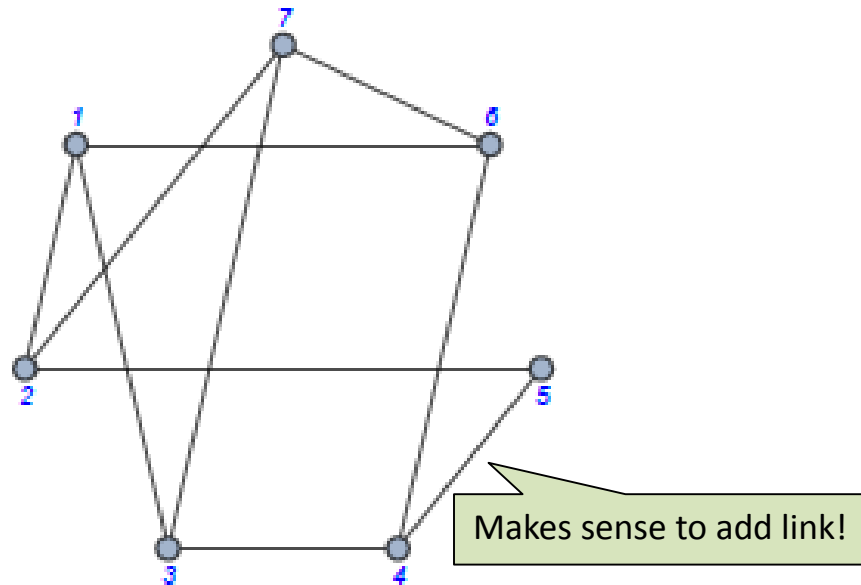
Sources

	1	2	3	4	5	6	7
1	0	$\frac{2}{65}$	$\frac{1}{13}$	$\frac{1}{65}$	$\frac{1}{65}$	$\frac{2}{65}$	$\frac{3}{65}$
2	$\frac{2}{65}$	0	$\frac{1}{65}$	0	0	0	$\frac{2}{65}$
3	$\frac{1}{13}$	$\frac{1}{65}$	0	$\frac{2}{65}$			
4	$\frac{1}{65}$	0	$\frac{2}{65}$	0	$\frac{4}{65}$	0	0
5	$\frac{1}{65}$	0	$\frac{3}{65}$	$\frac{4}{65}$	0	0	0
6	$\frac{2}{65}$	0	0	0	0	0	$\frac{3}{65}$
7	$\frac{3}{65}$	$\frac{2}{65}$	$\frac{1}{13}$	0	0	$\frac{3}{65}$	0

Much from 4 to 5.

design

Output: DAN



Demand matrix: joint distribution

... of **constant degree** (scalability)

DAN Design: New Types of Problems

Input: Workload

Destinations

1 communicates to many.

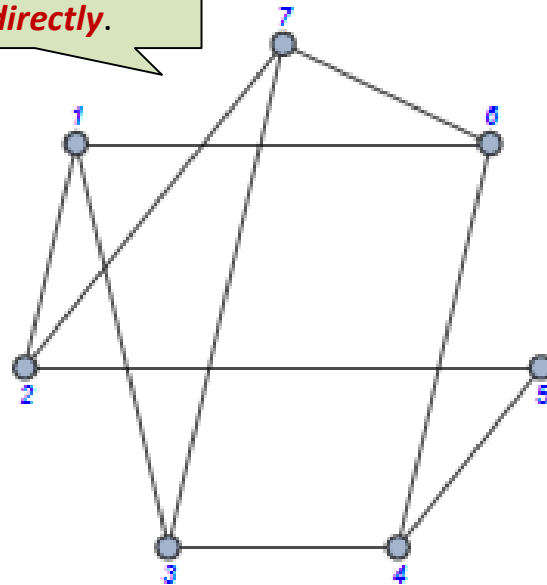
1 communicates to many.

		6	7				
1	0	$\frac{2}{65}$	$\frac{1}{13}$	$\frac{1}{65}$	$\frac{1}{65}$	$\frac{2}{65}$	$\frac{3}{65}$
2	$\frac{2}{65}$	0	$\frac{1}{65}$	0	0	0	$\frac{2}{65}$
3	$\frac{1}{13}$	$\frac{1}{65}$	0	$\frac{2}{65}$	0	0	$\frac{1}{13}$
4	$\frac{1}{65}$	0	$\frac{2}{65}$	0	$\frac{4}{65}$	0	0
5	$\frac{1}{65}$	0	$\frac{3}{65}$	$\frac{4}{65}$	0	0	0
6	$\frac{2}{65}$	0	0	0	0	0	$\frac{3}{65}$
7	$\frac{3}{65}$	$\frac{2}{65}$	$\frac{1}{13}$	0	0	$\frac{3}{65}$	0

Output: DAN

Bounded degree: route to 7 *indirectly*.

design



Demand matrix: joint distribution

... of *constant degree* (scalability)

DAN Design: New Types of Problems

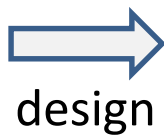
Input: Workload

Destinations

Sources

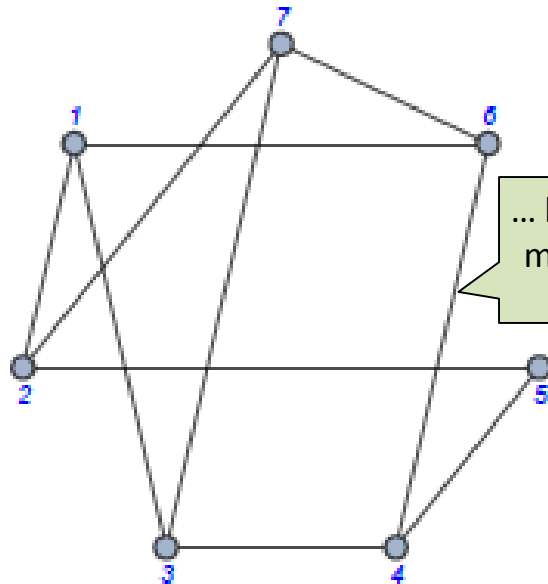
	1	2	3	4	5	6	7
1	0	$\frac{2}{65}$	$\frac{1}{13}$	$\frac{1}{65}$	$\frac{1}{65}$	$\frac{2}{65}$	$\frac{3}{65}$
2	$\frac{2}{65}$	0	$\frac{1}{65}$	0	0	0	$\frac{2}{65}$
3	$\frac{1}{13}$	$\frac{1}{65}$	0	$\frac{2}{65}$	0	0	$\frac{1}{13}$
4	$\frac{1}{65}$	0	$\frac{2}{65}$	0	$\frac{4}{65}$	0	0
5	$\frac{1}{65}$	0	$\frac{3}{65}$	$\frac{4}{65}$	0	0	0
6	$\frac{2}{65}$	0	0	0	0	0	0
7	$\frac{3}{65}$	$\frac{2}{65}$	$\frac{1}{13}$	0	0	$\frac{3}{65}$	0

4 and 6 don't communicate...



design

Output: DAN



... but “extra” link still makes sense: *not a subgraph*.

Demand matrix: joint distribution

... of *constant degree* (scalability)

More Formally: DAN Design Problem

Input:

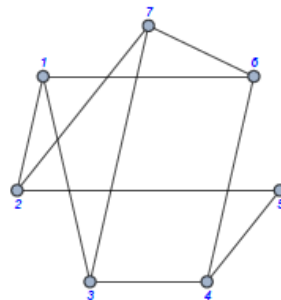
$\mathcal{D}[\mathbf{p}(\mathbf{i}, \mathbf{j})]$: joint **distribution**, Δ

		Y						
		1	2	3	4	5	6	7
X	1	0	$\frac{2}{65}$	$\frac{1}{13}$	$\frac{1}{65}$	$\frac{1}{65}$	$\frac{2}{65}$	$\frac{3}{65}$
	2	$\frac{2}{65}$	0	$\frac{1}{65}$	0	0	0	$\frac{2}{65}$
	3	$\frac{1}{13}$	$\frac{1}{65}$	0	$\frac{2}{65}$	0	0	$\frac{1}{13}$
	4	$\frac{1}{65}$	0	$\frac{2}{65}$	0	$\frac{4}{65}$	0	0
	5	$\frac{1}{65}$	0	$\frac{3}{65}$	$\frac{4}{65}$	0	0	0
	6	$\frac{2}{65}$	0	0	0	0	0	$\frac{3}{65}$
	7	$\frac{3}{65}$	$\frac{2}{65}$	$\frac{1}{13}$	0	0	$\frac{3}{65}$	0



Output:

N: DAN



Bounded degree
 $\Delta=3$

Objective:

Expected Path Length (EPL):
Demand-weighted route length

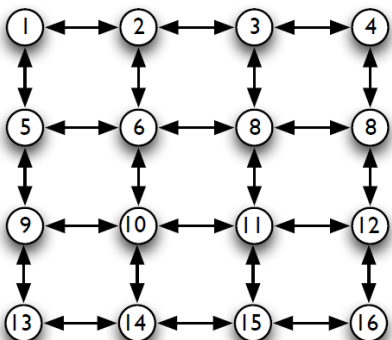
$$\text{EPL}(\mathcal{D}, N) = \sum_{(u,v) \in \mathcal{D}} p(u,v) \cdot d_N(u,v)$$

Path length **on DAN N.**

Frequency

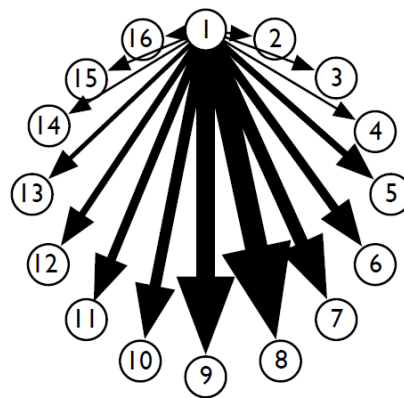
Sometimes, DANs can be much better!

Example 1: low-degree demand



- Already low degree: degree-4 DAN can serve this **at cost 1**.

Example 2: high-degree but skewed demand



- If sufficiently skewed: constant-degree DAN can serve it at cost **$O(1)$**

So on what does it depend?

So on what does it depend?



We argue (but still don't know!): on the
“entropy” of the demand!



„Coming to Wroclaw?“

An Analogy to Coding



00110101...



if demand **arbitrary** and **unknown**

worst case network:
Full BW

log diameter

worst case coding:
00, 01, 10, 11

log # bits / symbol

An Analogy to Coding



01011...



if demand **arbitrary** and **unknown**

worst case network: Full BW
worst case coding: 00, 01, 10, 11

log diameter

log # bits / symbol



DAN!



if demand **known** and **fixed**

entropy?

entropy / symbol

static Demand-Aware Nets
static Huffman: 1, 01, 001, 000

„Coming to Wroclaw?“

An Analogy to Coding



011...



if demand **arbitrary** and **unknown**

worst case network: Full BW
worst case coding: 00, 01, 10, 11

$\log \text{diameter}$

$\log \# \text{ bits / symbol}$



Dynamic DANs:
Aka. **Self-Adjusting Networks (SANs)!**



DAN!



SAN!

if demand **known** and **fixed**

if demand **unknown** but **reconfigurable**

entropy?

entropy / symbol

static Demand-Aware Nets
static Huffman: 1, 01, 001, 000

dynamic Demand-Aware Nets
dynamic Huffman codes

„Coming to Wroclaw?“



011...



An Analogy to Coding

if demand **arbitrary** and **unknown**

worst case network: Full BW
worst case coding: 00, 01, 10, 11

log diameter

log # bits / symbol

Dynamic DANs:
Aka. **Self-Adjusting Networks (SANs)!**



DAN!



SAN!

if demand **known** and **fixed**

if demand **unknown** but **reconfigurable**

static Demand-Aware Nets
static Huffman: 1, 01, 001, 000

dynamic Demand-Aware Nets
dynamic Huffman codes



Can exploit
spatial locality!



Additionally exploit
temporal locality!

„Coming to Wroclaw?“

An Analogy to Coding



011...



if demand **arbitrary** and **unknown**

worst case network: Full BW
worst case coding: 00, 01, 10, 11

log diameter

log # bits / symbol

Dynamic DANs:
Aka. **Self-Adjusting Networks (SANs)!**



DAN!



SAN!

if demand **known**

if demand **unknown** but **repeating**



Can exploit **spatial locality**

Aware Nets
static Huffman: 1, 01, 001, 000



Additionally exploit **temporal locality!**

dynamic Demand
Huffman codes

„Cheating“: need to know demand!

Need online algorithms!

Analogous to *Datastructures*: Oblivious...

- Traditional, **fixed** BSTs do not rely on any assumptions on the demand

- Optimize for the **worst-case**

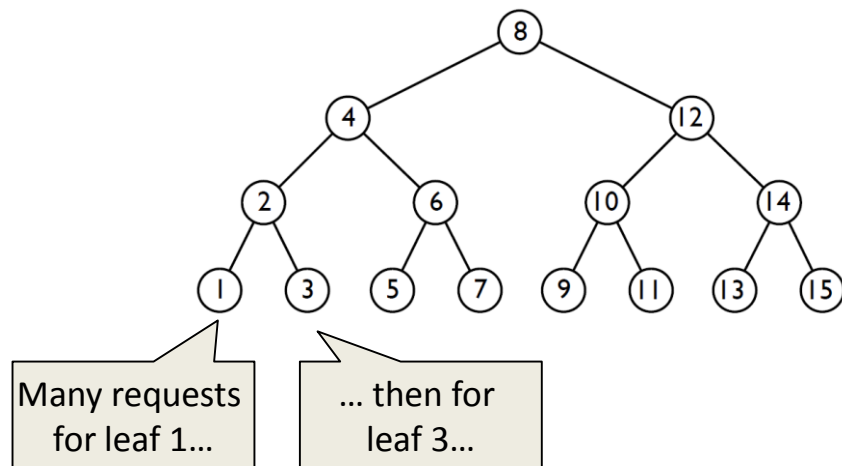
- Example **demand**:

1,...,1,3,...,3,5,...,5,7,...,7,...,log(n),...,log(n)
↔ ↔ ↔ ↔ ↔
many many many many many

- Items stored at **$O(\log n)$** from the root, **uniformly** and **independently** of their frequency

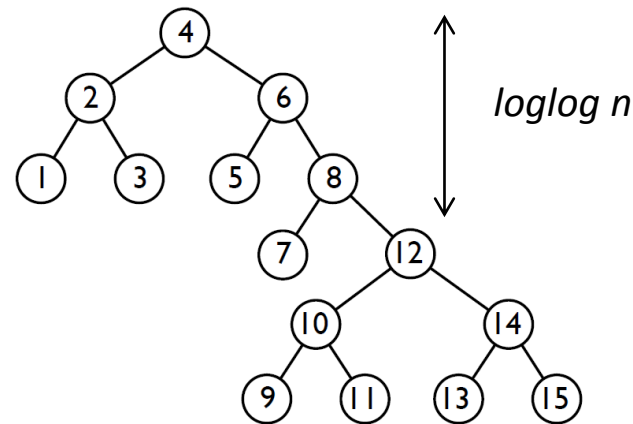


Corresponds to
max possible demand!



... Demand-Aware ...

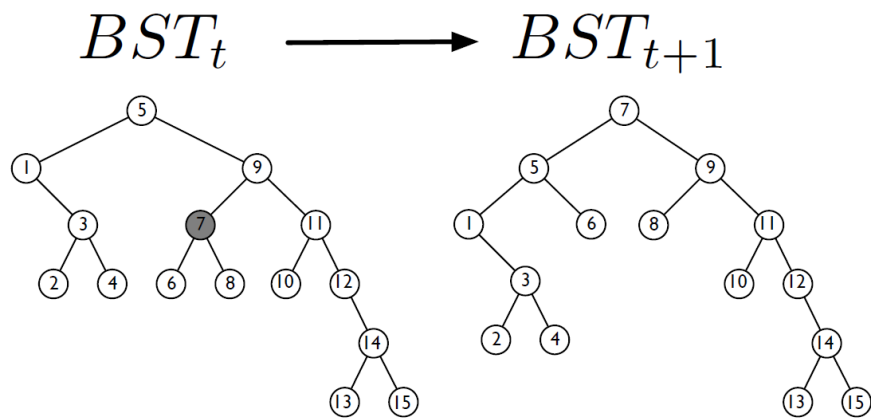
- **Demand-aware fixed** BSTs can take advantage of *spatial locality* of the demand
- E.g.: place frequently accessed elements close to the root
- E.g., **Knuth/Mehlhorn/Tarjan** trees
- Recall example **demand**:
1,...,1,3,...,3,5,...,5,7,...,7,...,log(n),...,log(n)
 - Amortized cost **$O(\log \log n)$**



Amortized cost corresponds
to **empirical entropy of demand!**

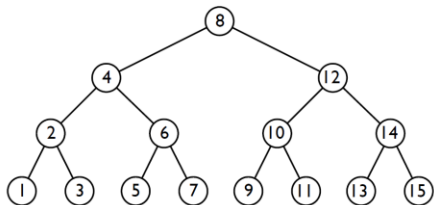
... Self-Adjusting!

- **Demand-aware reconfigurable** BSTs can additionally take advantage of **temporal locality**
- By moving accessed element to the root: amortized cost is **constant**, i.e., $O(1)$
 - Recall example **demand**:
 $1, \dots, 1, 3, \dots, 3, 5, \dots, 5, 7, \dots, 7, \dots, \log(n), \dots, \log(n)$



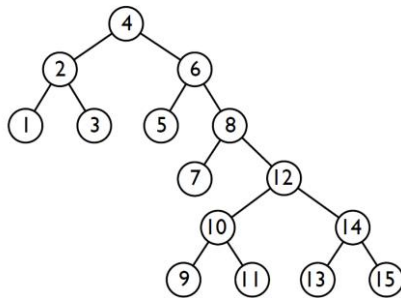
Datastructures

Oblivious



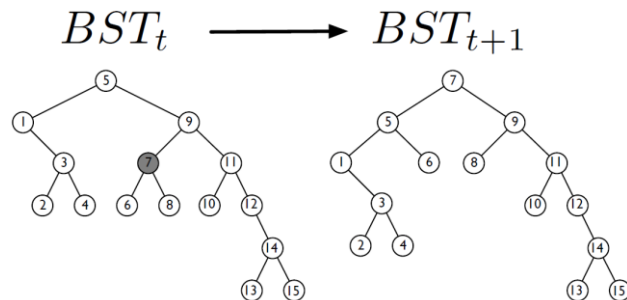
Lookup
 $O(\log n)$

Demand-Aware



Exploit **spatial locality**:
empirical entropy $O(\log \log n)$

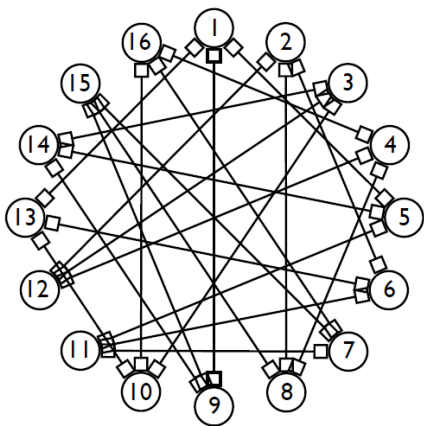
Self-Adjusting



Exploit **temporal locality** as well:
 $O(1)$

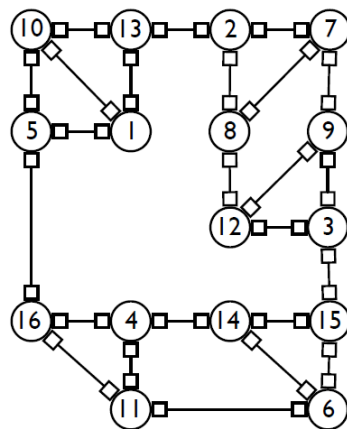
Analogously for Networks

Oblivious



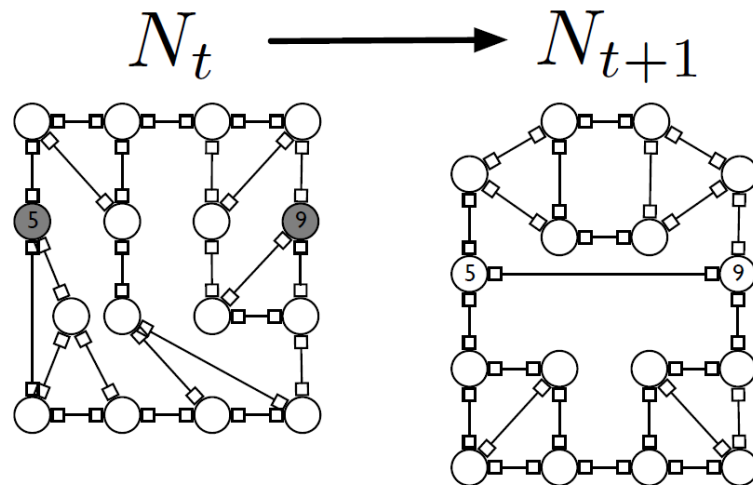
Const degree
(e.g., **expander**):
route lengths $O(\log n)$

DAN



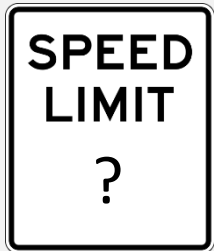
Exploit **spatial locality**

SAN



Exploit **temporal locality** as well

Avin, S.: Toward Demand-Aware Networking: A Theory for Self-Adjusting Networks. **SIGCOMM CCR** 2018.



Intuition: Entropy Lower Bound



Lower Bound Idea: Leverage Coding or Datastructure

		Destinations						
		1	2	3	4	5	6	7
Sources	1	0	$\frac{2}{65}$	$\frac{1}{13}$	$\frac{1}{65}$	$\frac{1}{65}$	$\frac{2}{65}$	$\frac{3}{65}$
	2	$\frac{2}{65}$	0	$\frac{1}{65}$	0	0	0	$\frac{2}{65}$
	3	$\frac{1}{13}$	$\frac{1}{65}$	0	$\frac{2}{65}$	0	0	$\frac{1}{13}$
	4	$\frac{1}{65}$	0	$\frac{2}{65}$	0	$\frac{4}{65}$	0	0
	5	$\frac{1}{65}$	0	$\frac{3}{65}$	$\frac{4}{65}$	0	0	0
	6	$\frac{2}{65}$	0	0	0	0	0	$\frac{3}{65}$
	7	$\frac{3}{65}$	$\frac{2}{65}$	$\frac{1}{13}$	0	0	$\frac{3}{65}$	0

- DAN just for a *single (source) node 1*: cannot do better than Δ -ary **Huffman tree** for its destinations

- How good can this tree be?



Entropy lower bound on EPL known for binary trees, e.g. *Mehlhorn* 1975 for BST

Lower Bound Idea: Leverage Coding or Datastructure

An optimal “**ego-tree**”
for this source!

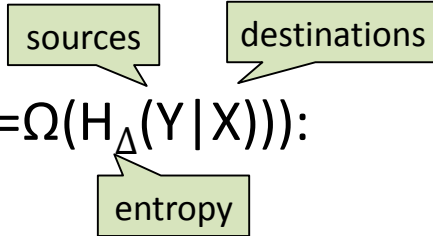
		Destinations						
		1	2	3	4	5	6	7
Sources	1	0	$\frac{2}{65}$	$\frac{1}{13}$	$\frac{1}{65}$	$\frac{1}{65}$	$\frac{2}{65}$	$\frac{3}{65}$
	2	$\frac{2}{65}$	0	$\frac{1}{65}$	0	0	0	$\frac{2}{65}$
	3	$\frac{1}{13}$	$\frac{1}{65}$	0	$\frac{2}{65}$	0	0	$\frac{1}{13}$
	4	$\frac{1}{65}$	0	$\frac{2}{65}$	0	$\frac{4}{65}$	0	0
	5	$\frac{1}{65}$	0	$\frac{3}{65}$	$\frac{4}{65}$	0	0	0
	6	$\frac{2}{65}$	0	0	0	0	0	$\frac{3}{65}$
	7	$\frac{3}{65}$	$\frac{2}{65}$	$\frac{1}{13}$	0	0	$\frac{3}{65}$	0

- DAN just for a **single (source) node 1**: cannot do better than Δ -ary **Huffman tree** for its destinations
- How good can this tree be?

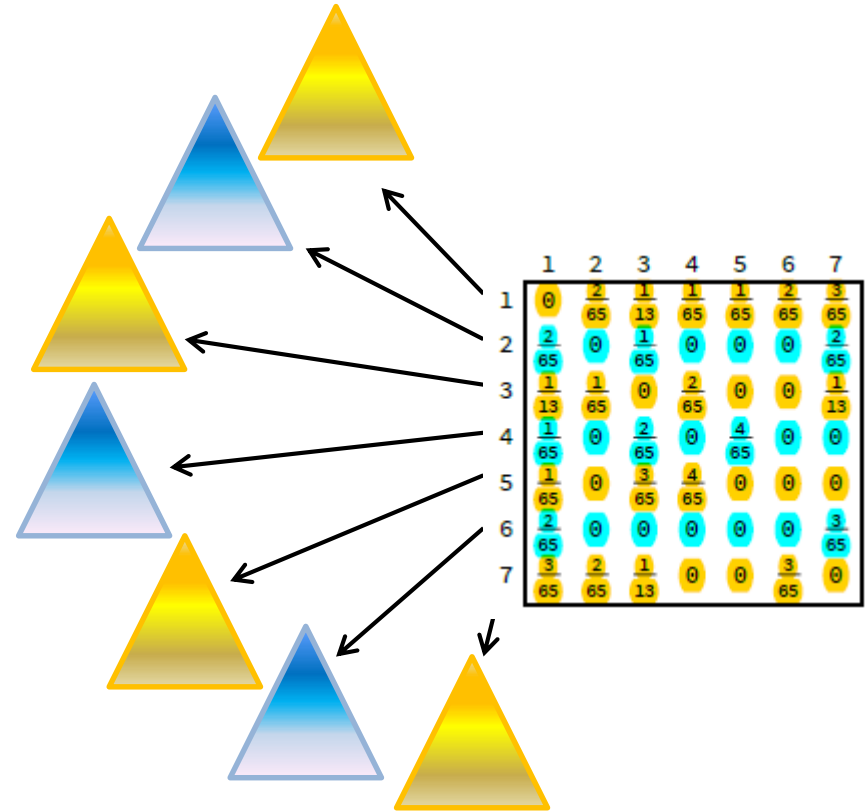


Entropy lower bound on EPL known for binary trees, e.g. **Mehlhorn** 1975 for BST

So: Entropy of the *Entire* Demand



- **Proof idea** ($EPL = \Omega(H(Y|X))$):
 - Compute **ego-tree** for each source node
 - Take **union** of all **ego-trees**
 - Violates **degree restriction** but valid lower bound



Entropy of the *Entire* Demand: Sources *and* Destinations

Do this in **both dimensions**:

$$\text{EPL} \geq \Omega(\max\{H_{\Delta}(Y|X), H_{\Delta}(X|Y)\})$$

$\Omega(H_{\Delta}(X|Y))$

	1	2	3	4	5	6	7	
1	0	$\frac{2}{65}$	$\frac{1}{13}$	$\frac{1}{65}$	$\frac{1}{65}$	$\frac{2}{65}$	$\frac{3}{65}$	
2	$\frac{2}{65}$	0	$\frac{1}{65}$	0	0	0	$\frac{2}{65}$	
3	$\frac{1}{13}$	$\frac{1}{65}$	0	$\frac{2}{65}$	0	0	$\frac{1}{13}$	
4	$\frac{1}{65}$	0	$\frac{2}{65}$	0	$\frac{4}{65}$	0	0	
5	$\frac{1}{65}$	0	$\frac{3}{65}$	$\frac{4}{65}$	0	0	0	
6	$\frac{2}{65}$	0	0	0	0	0	$\frac{3}{65}$	
7	$\frac{3}{65}$	$\frac{2}{65}$	$\frac{1}{13}$	0	0	$\frac{3}{65}$	0	

$\Omega(H_{\Delta}(Y|X))$

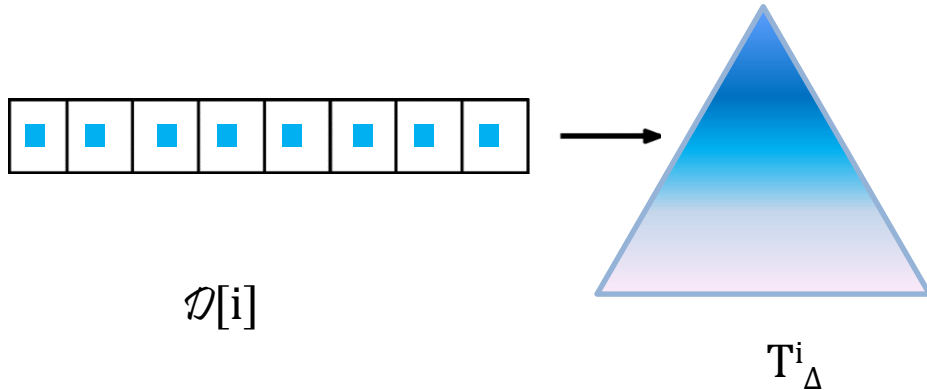
\mathcal{D}

Intuition: Reaching Entropy Limit in Datacenters



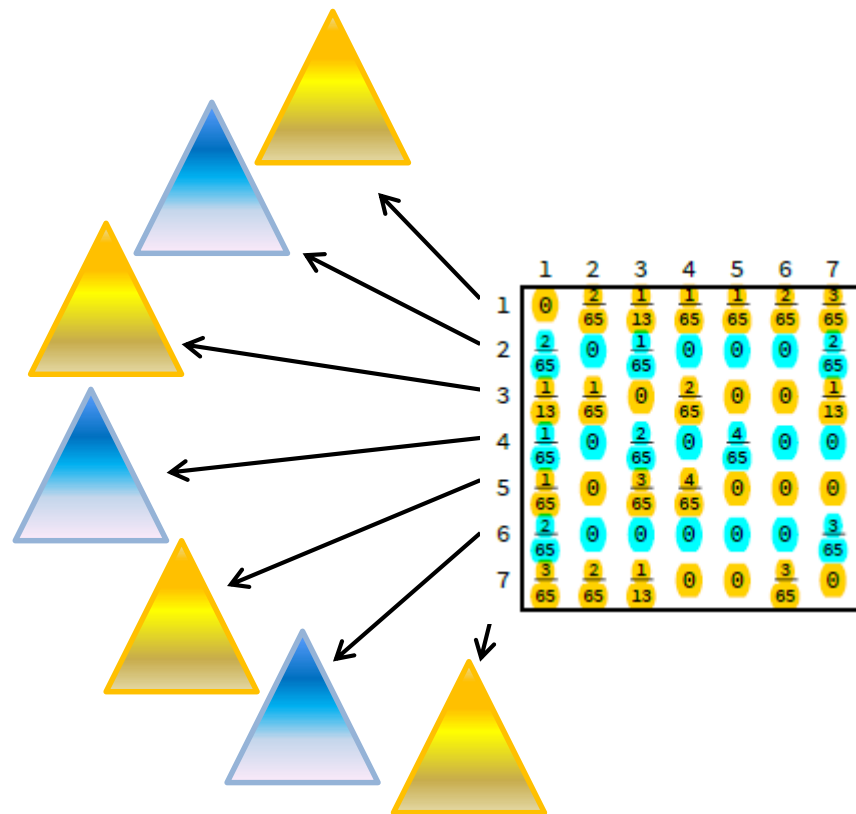
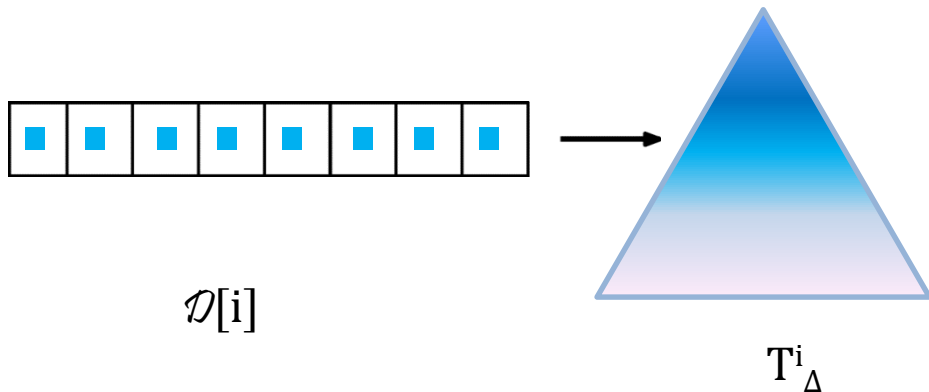
Ego-Trees Revisited

- ego-tree: optimal tree for a row (= given source)



Ego-Trees Revisited

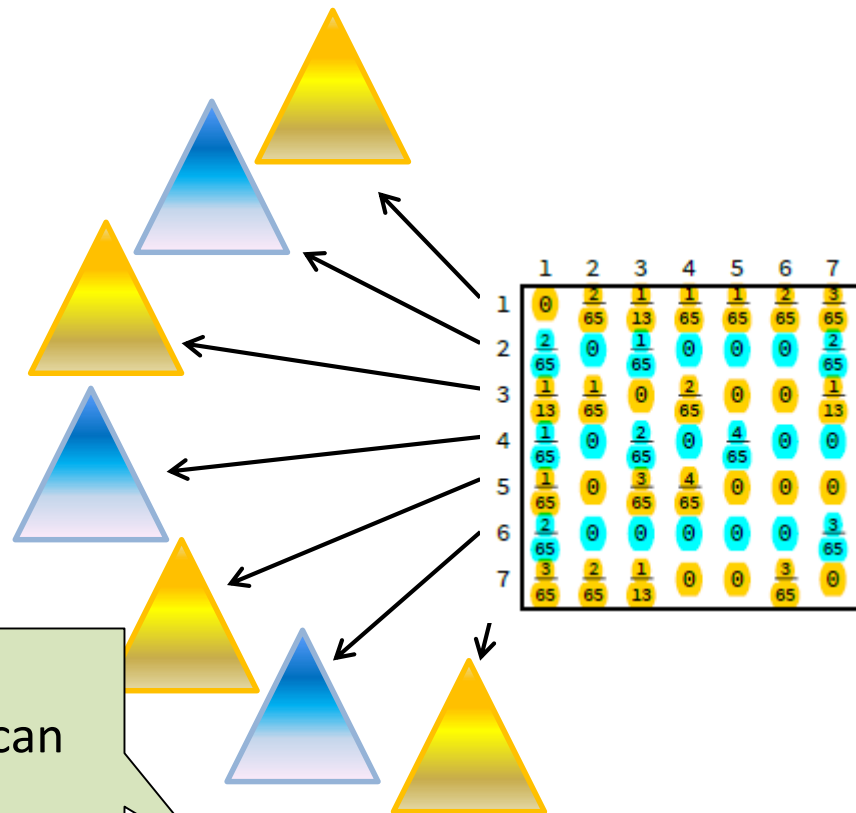
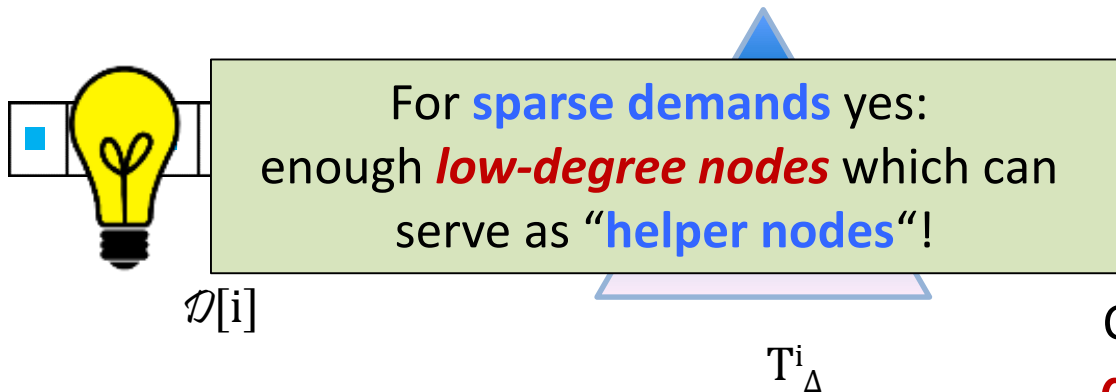
- ego-tree: optimal tree for a row (= given source)



Can we merge the trees **without distortion** and **keep degree low**?

Ego-Trees Revisited

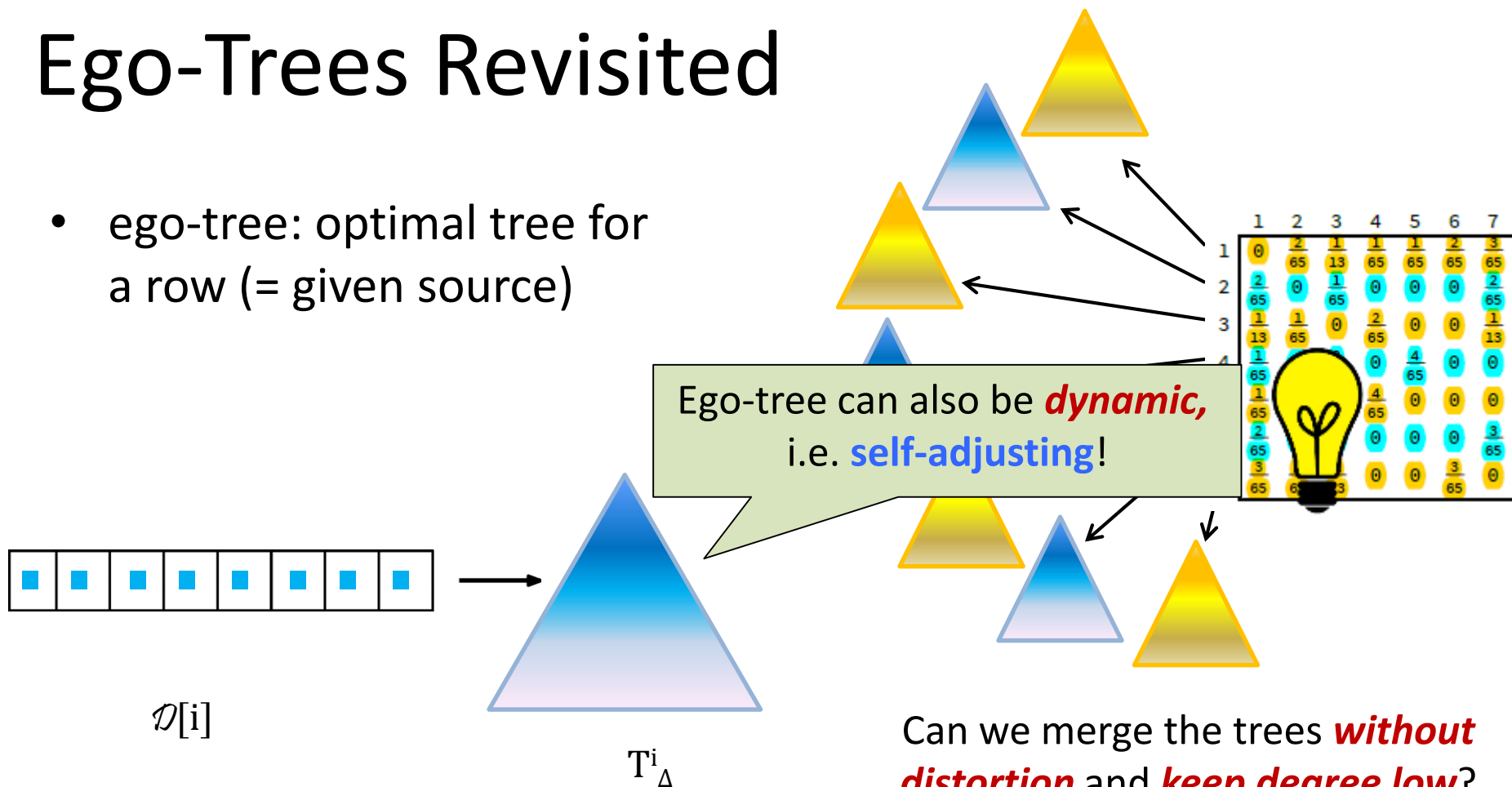
- ego-tree: optimal tree for a row (= given source)



Can we merge the trees **without distortion** and **keep degree low**?

Ego-Trees Revisited

- ego-tree: optimal tree for a row (= given source)



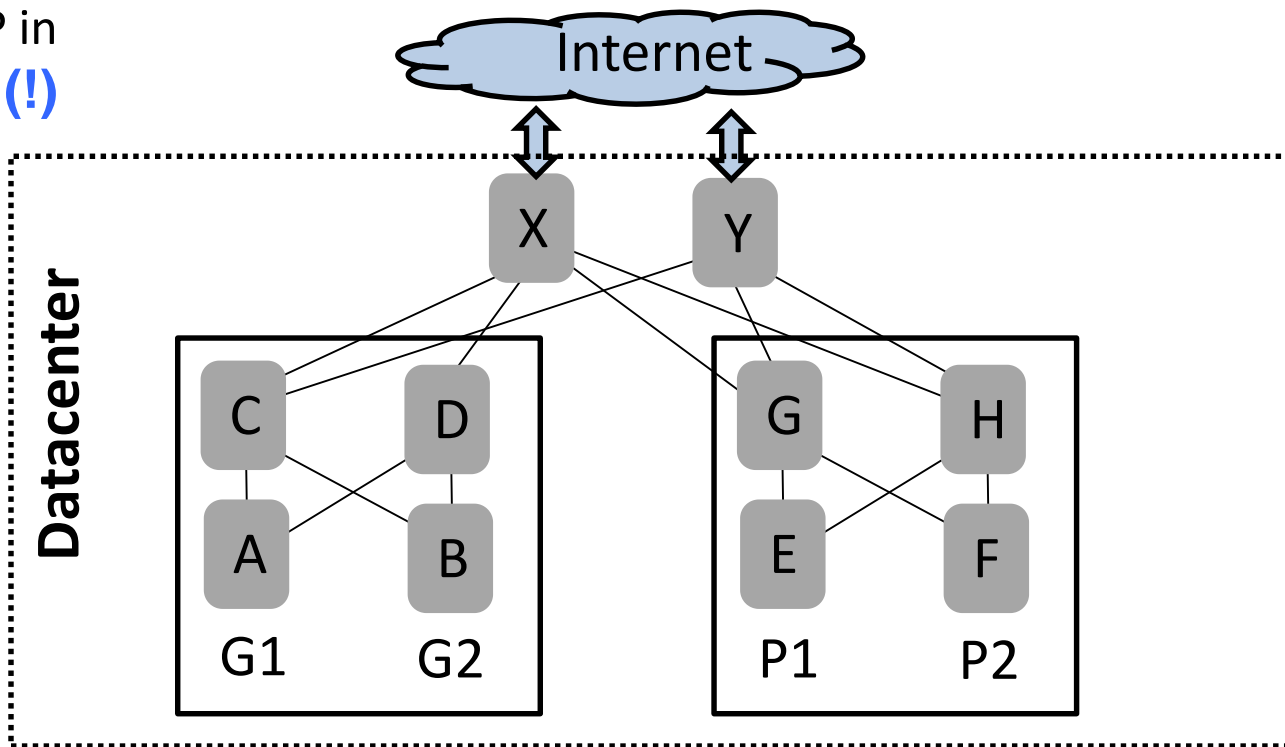
Roadmap

- Opportunities of self-* networks
 - Example 1: Demand-aware, self-adjusting networks
 - **Example 2: Self-repairing networks**
- Challenges of designing self-* networks



Reasoning About Failures is Hard

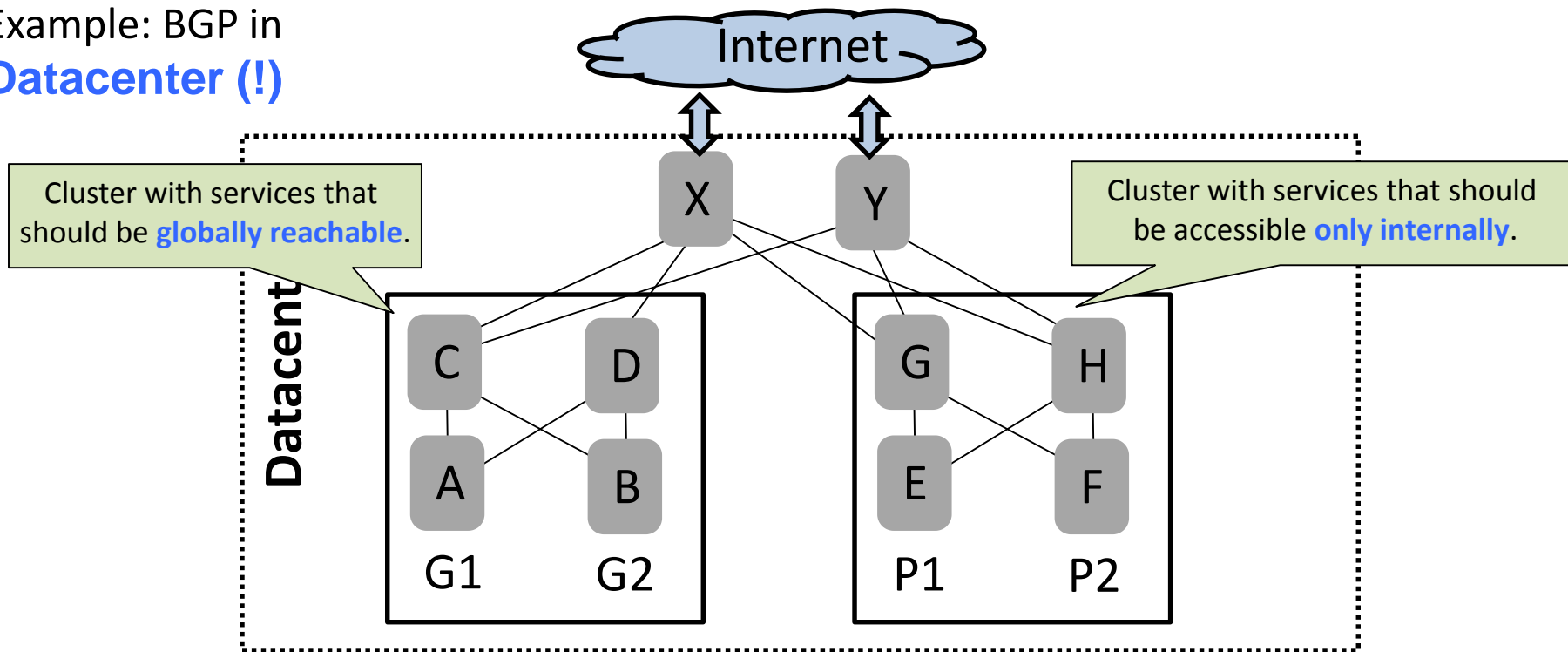
Example: BGP in
Datacenter (!)



Credits: Beckett et al. (SIGCOMM 2016): Bridging Network-wide Objectives and Device-level Configurations.

Reasoning About Failures is Hard

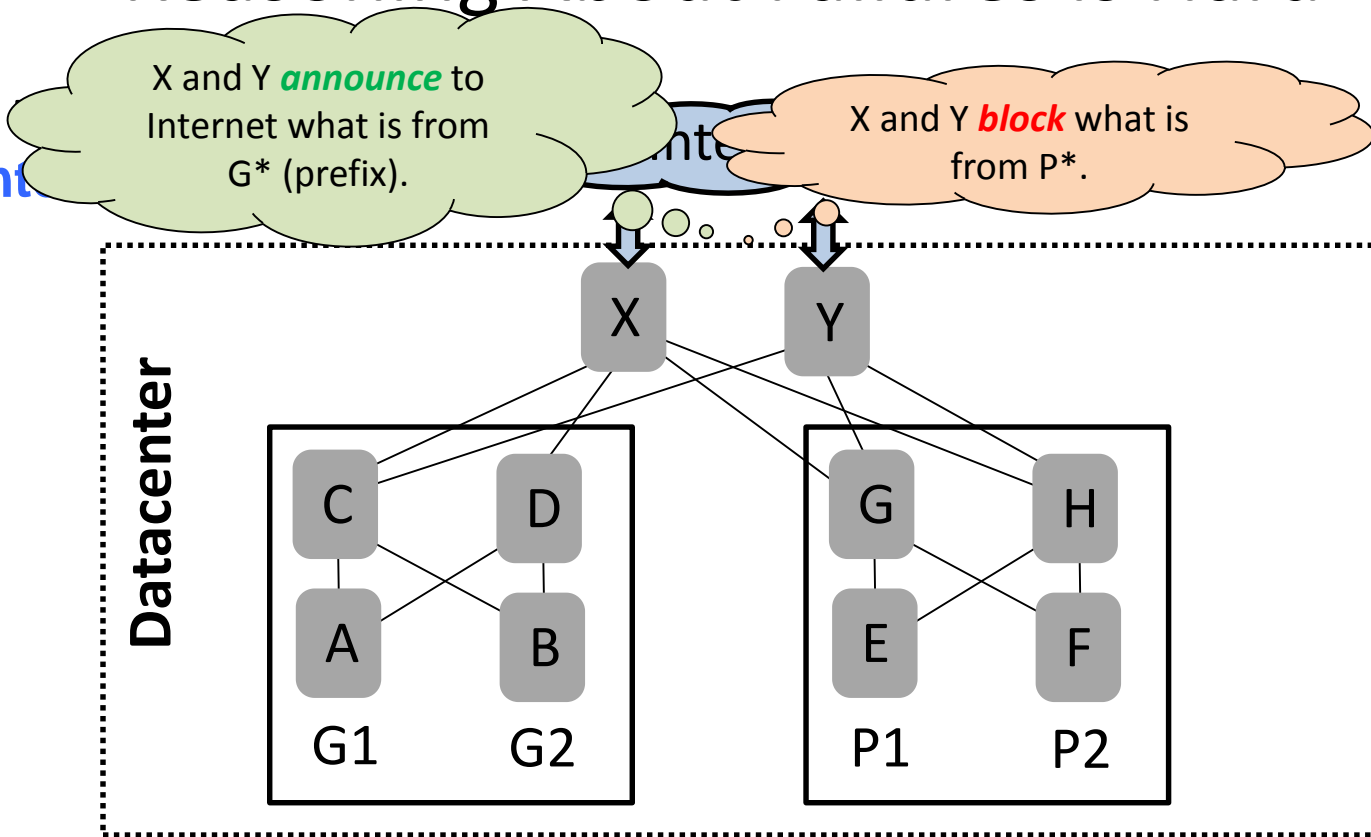
Example: BGP in
Datacenter (!)



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Reasoning About Failures is Hard

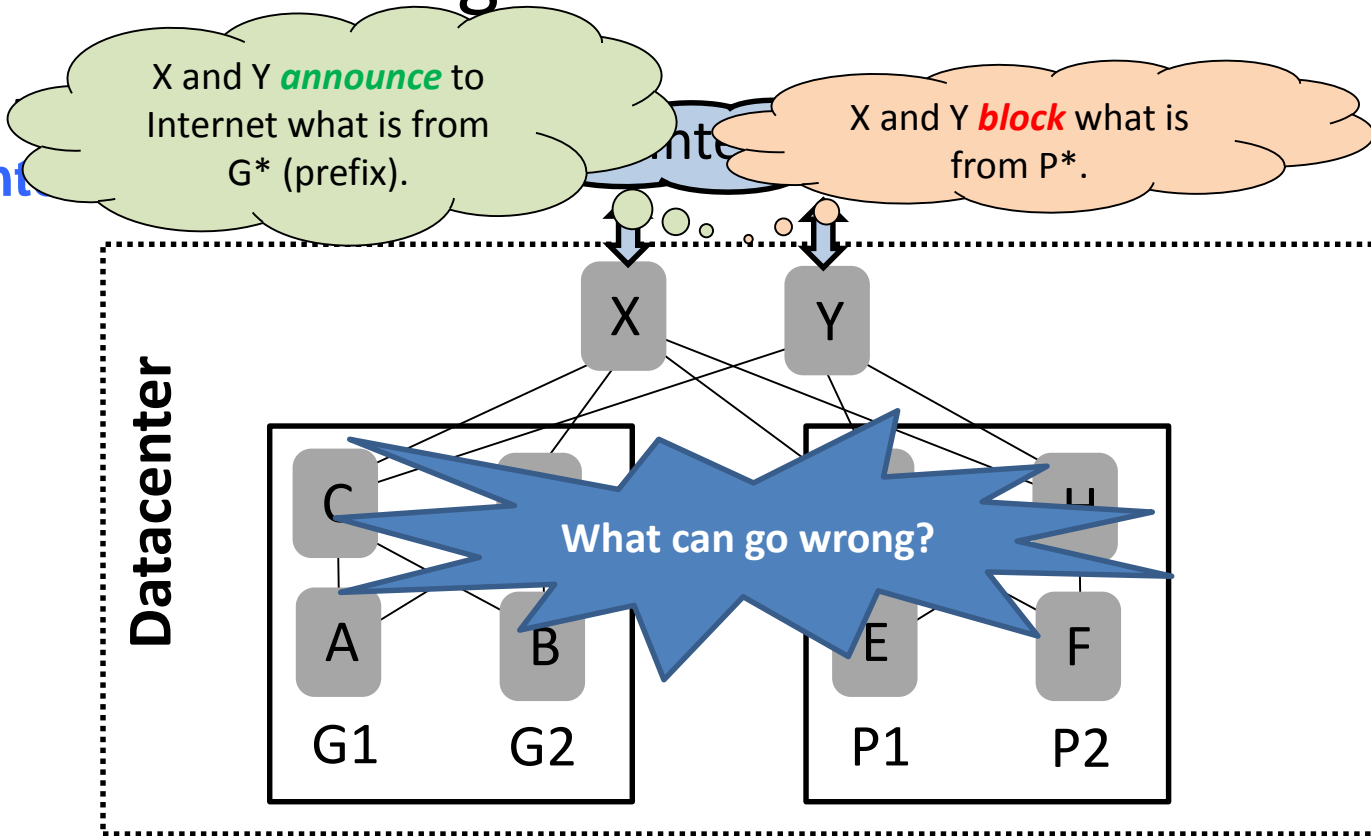
Example:
Datacenter



Credits: Beckett et al. (SIGCOMM 2016): Bridging Network-wide Objectives and Device-level Configurations.

Reasoning About Failures is Hard

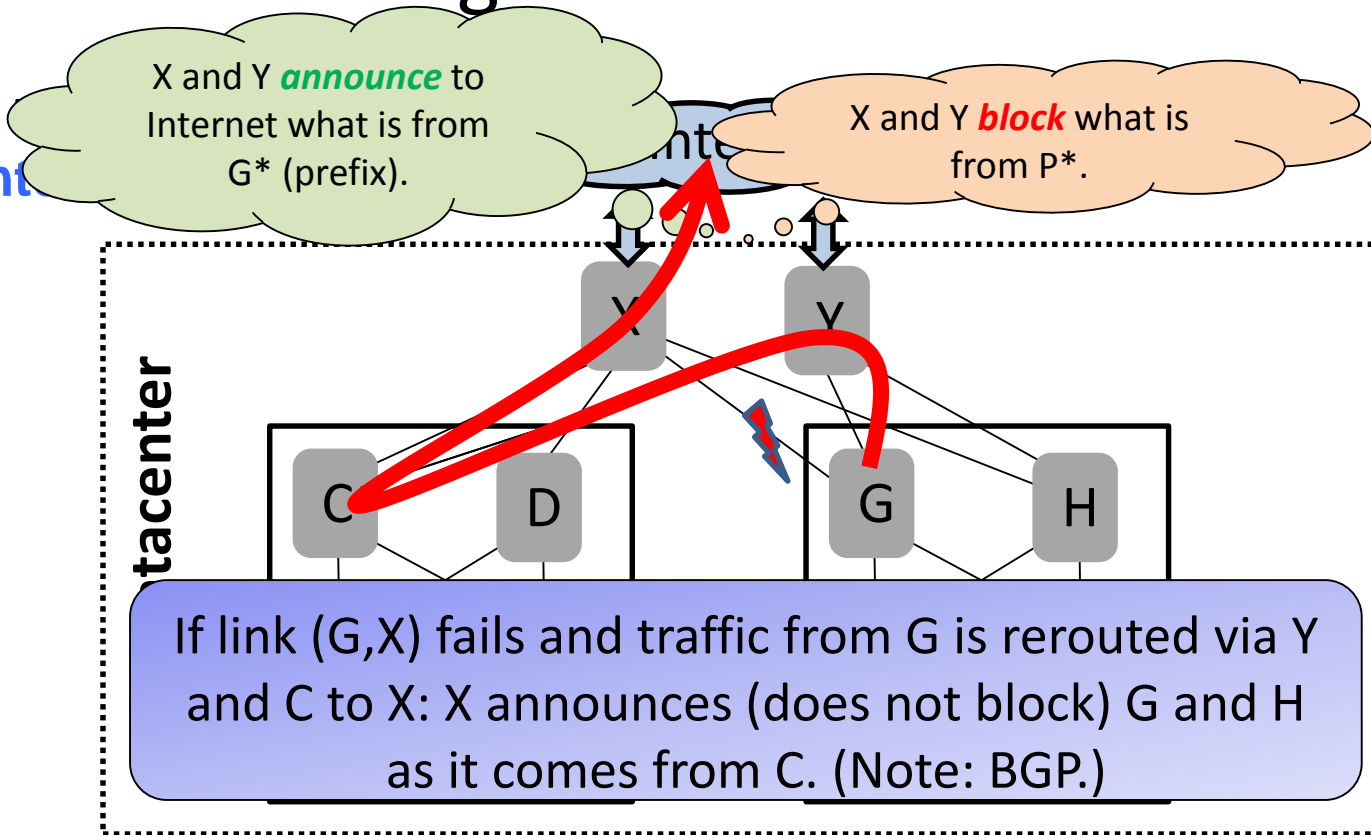
Example:
Datacenter



Credits: Beckett et al. (SIGCOMM 2016): Bridging Network-wide Objectives and Device-level Configurations.

Reasoning About Failures is Hard

Example:
Datacenter



Credits: Beckett et al. (SIGCOMM 2016): Bridging Network-wide Objectives and Device-level Configurations.

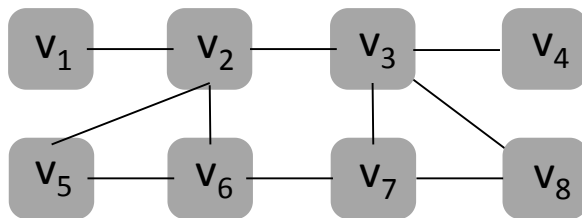
Managing Complex Networks is Hard for Humans



Another Case for Automation!

Example: Self-Repairing MPLS Networks

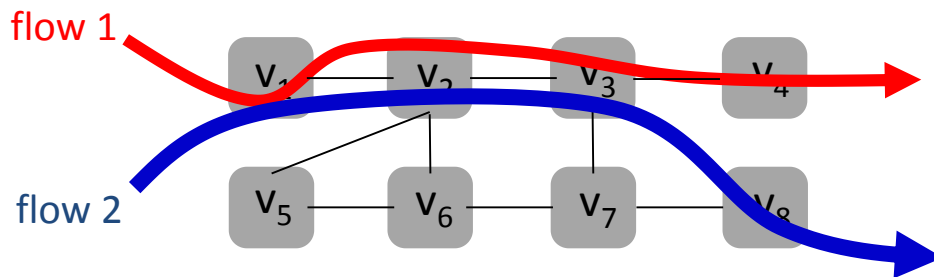
- MPLS: forwarding based on **top label** of label **stack**



Default routing of
two flows

Example: Self-Repairing MPLS Networks

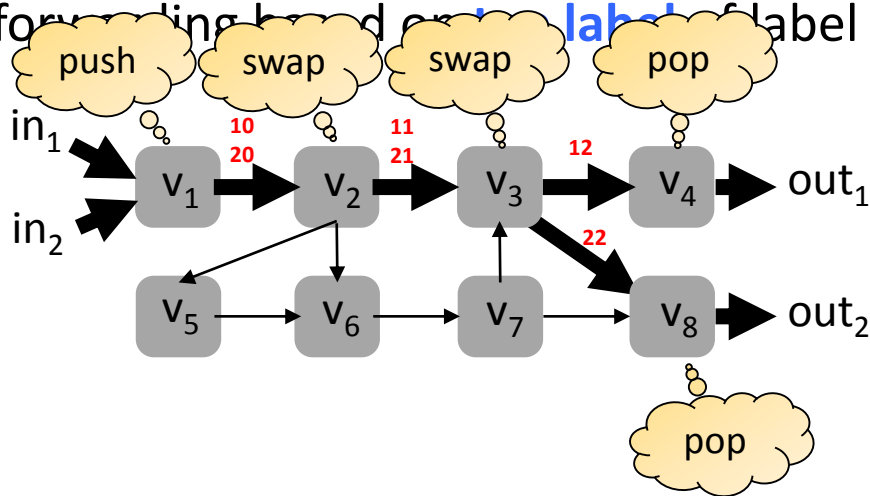
- MPLS: forwarding based on **top label** of label **stack**



Default routing of
two flows

Example: Self-Repairing MPLS Networks

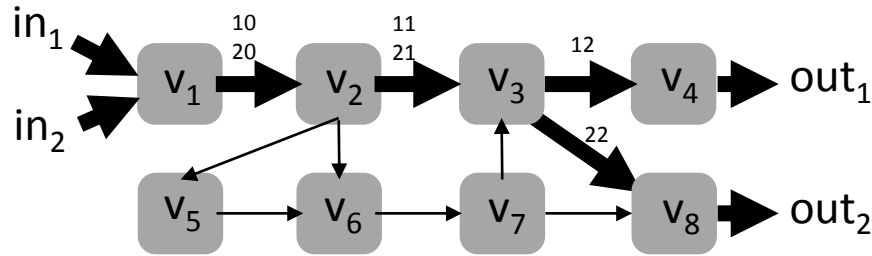
- MPLS: forwarding based on **label** of label **stack**



Default routing of
two flows

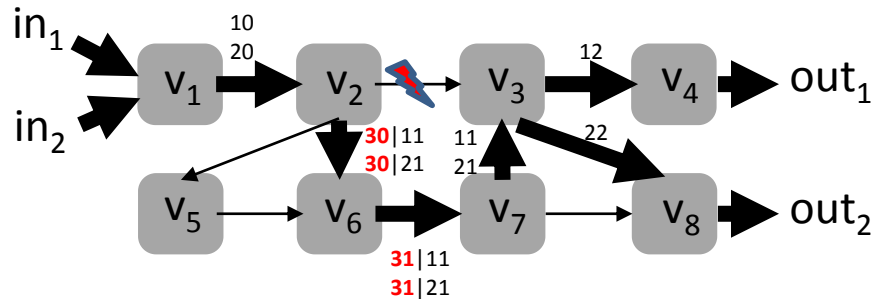
Fast Reroute Around *1 Failure*

- MPLS: forwarding based on **top label** of label **stack**



Default routing of two flows

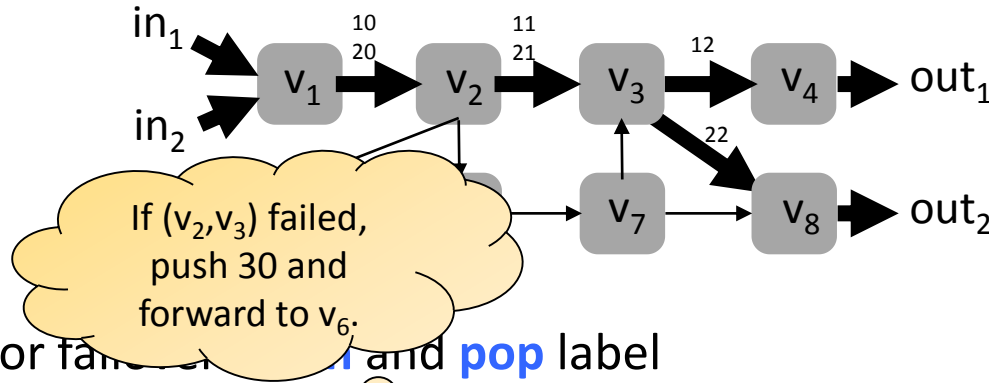
- For failover: **push** and **pop** label



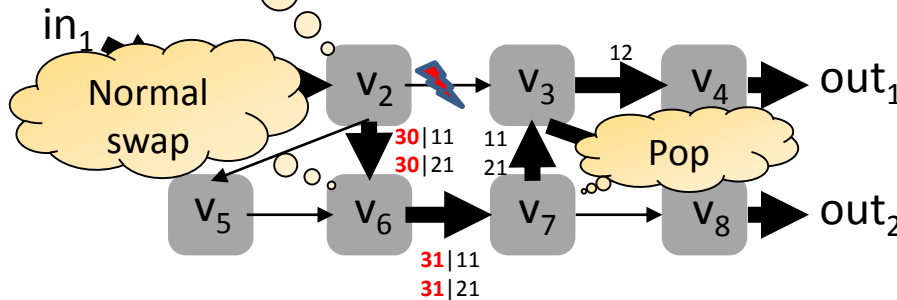
One failure: **push 30**:
route around (v2,v3)

Fast Reroute Around *1 Failure*

- MPLS: forwarding based on **top label** of label **stack**

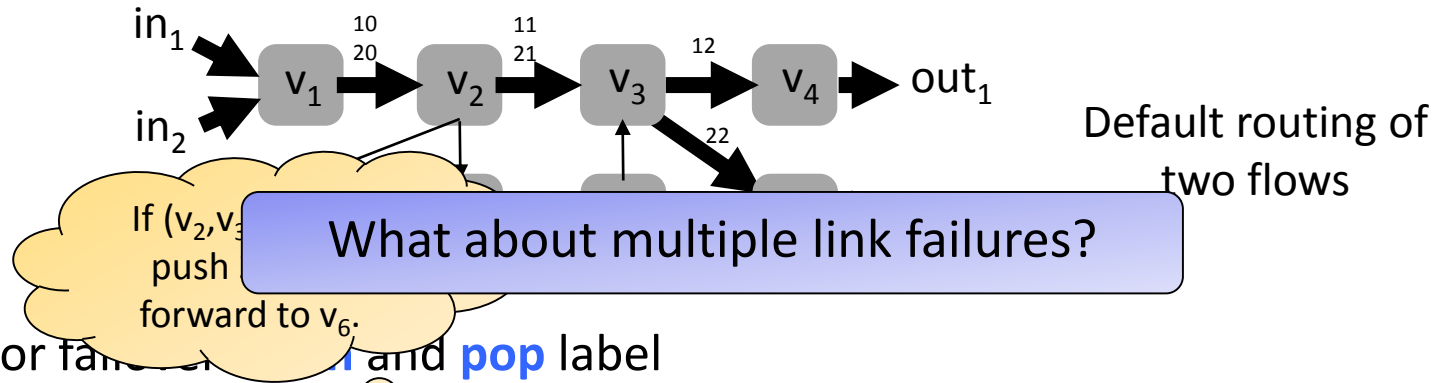


- For failure, push **pop** label

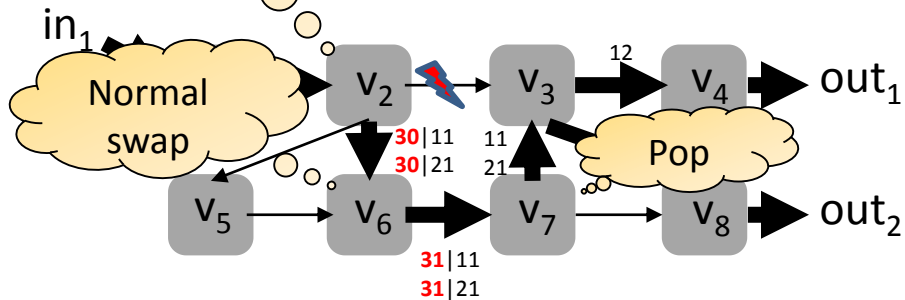


Fast Reroute Around *1 Failure*

- MPLS: forwarding based on **top label** of label **stack**

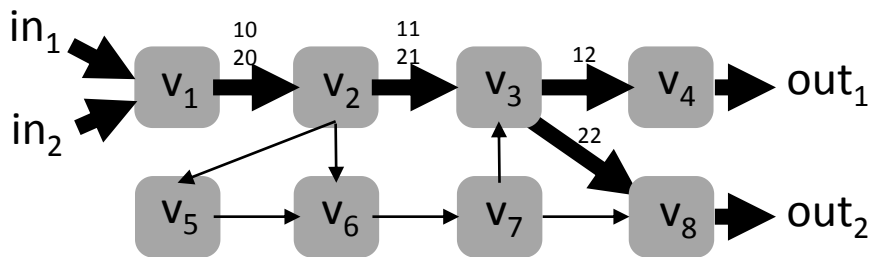


- For fast reroute, use **push** and **pop** label

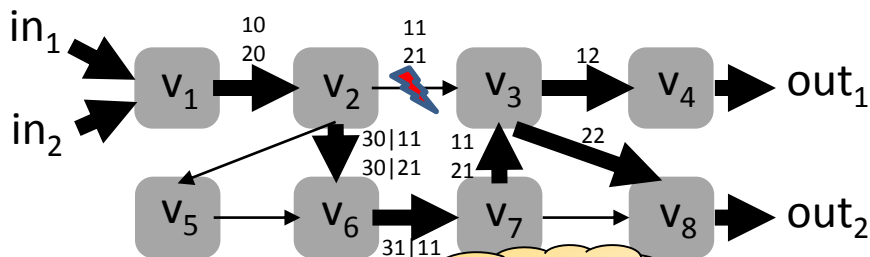


One failure: **push 30**:
route around (v_2, v_3)

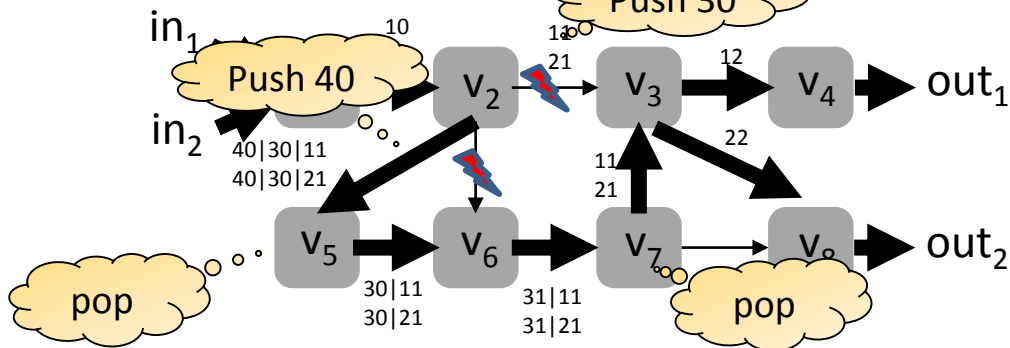
2 Failures: Push *Recursively*



Original Routing



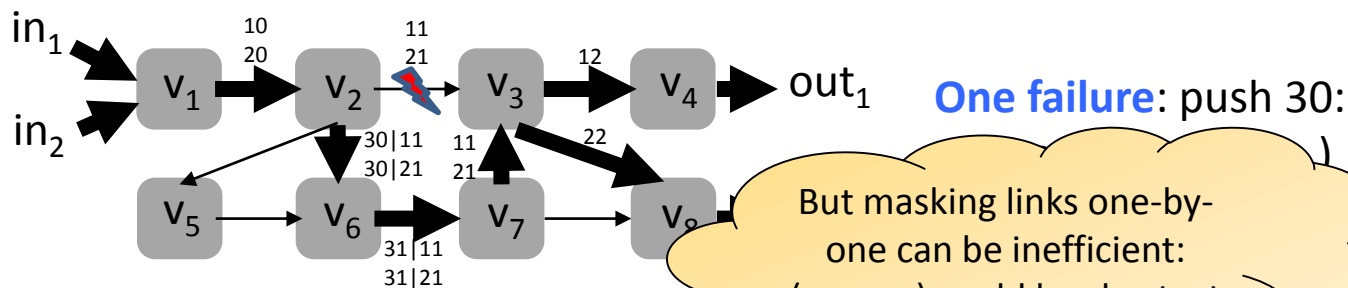
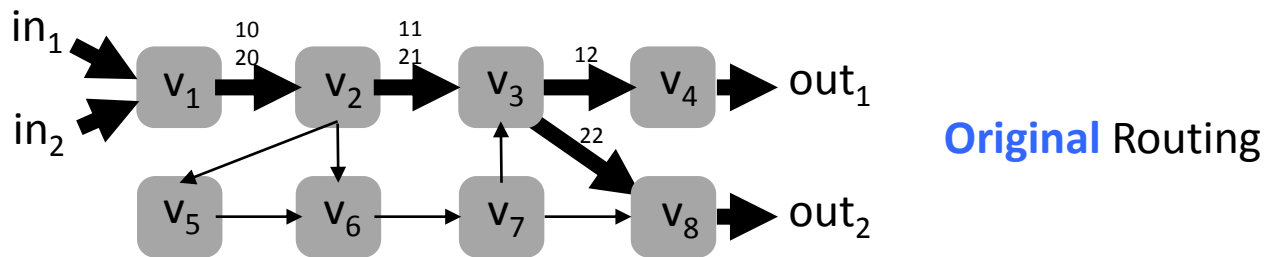
One failure: push 30:
route around (v_2, v_3)



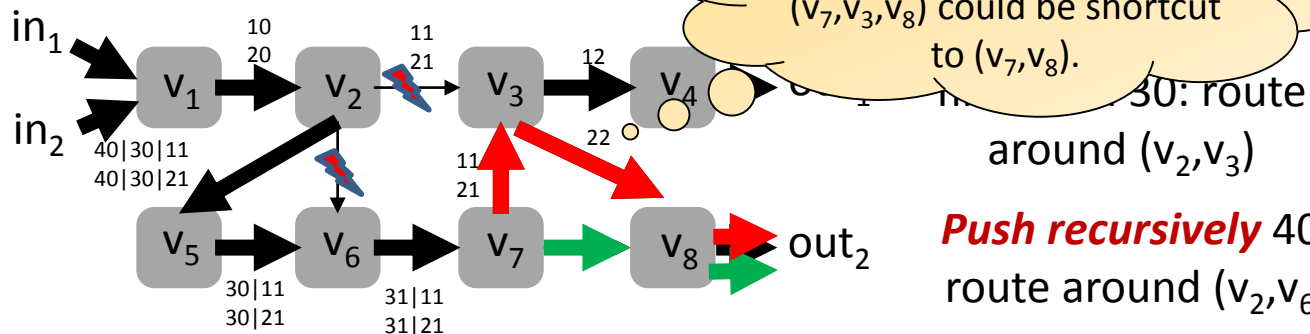
Two failures:
first push 30: route
around (v_2, v_3)

Push recursively 40:
route around (v_2, v_6)

2 Failures: Push *Recursively*

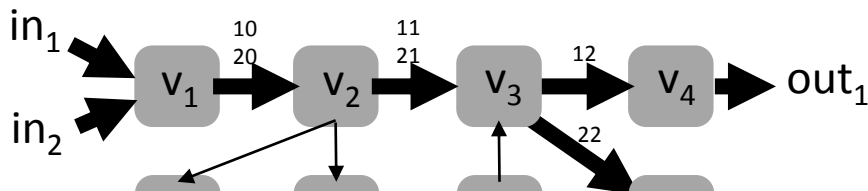


But masking links one-by-one can be inefficient:
 (v_7, v_3, v_8) could be shortcut
 to (v_7, v_8) .



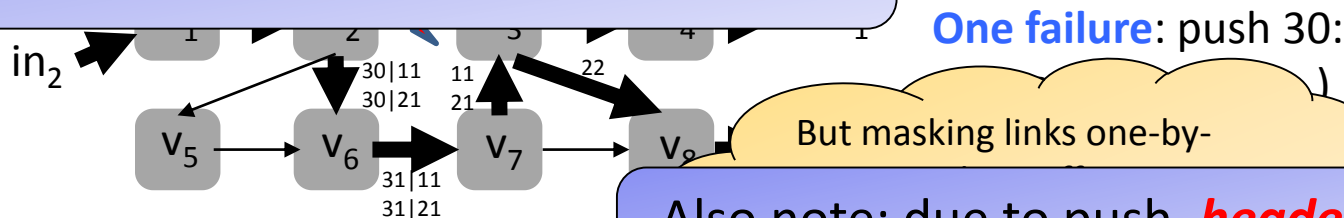
Push recursively 40:
 route around (v_2, v_6)

2 Failures: Push *Recursively*



Original Routing

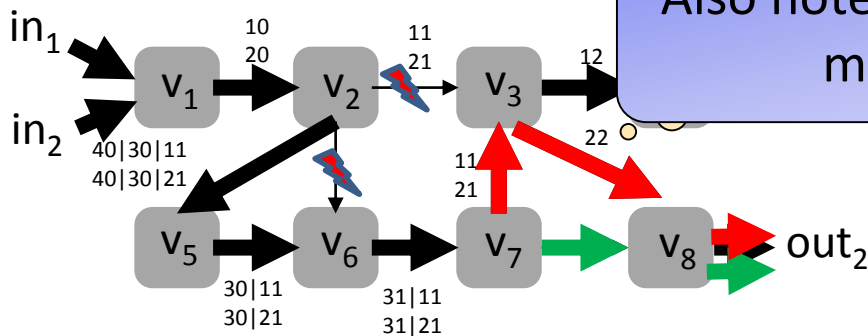
More efficient but also more complex:
Cisco does *not recommend* using this option!



One failure: push 30:

But masking links one-by-

Also note: due to push, *header size*
may grow arbitrarily!



around (v_2, v_3)

Push recursively 40:
route around (v_2, v_6)

Forwarding Tables for Our Example

FT	In-I	In-Label	Out-I	op
τ_{v_1}	in_1	\perp	(v_1, v_2)	$push(1)$
	in_2	\perp	(v_1, v_2)	$push(2)$
τ_{v_2}	(v_1, v_2)	10	(v_2, v_3)	$swap(21)$
	(v_1, v_2)	20	(v_2, v_3)	$swap(21)$
τ_{v_3}	(v_2, v_3)	11	(v_3, v_4)	$swap(12)$
	(v_2, v_3)	21	(v_3, v_8)	$swap(22)$
	(v_7, v_3)	11	(v_3, v_4)	$swap(12)$
	(v_7, v_3)	21	(v_3, v_8)	$swap(22)$
τ_{v_4}	(v_3, v_4)	12	out_1	pop
τ_{v_5}	(v_2, v_6)	40	(v_2, v_5)	pop
τ_{v_6}	(v_2, v_3)	11	(v_2, v_6)	$push(30)$
	(v_2, v_3)	21	(v_2, v_6)	$push(30)$
	(v_2, v_6)	30	(v_2, v_5)	$push(40)$
	(v_5, v_6)	71	(v_6, v_7)	$swap(72)$
	(v_6, v_7)	31	(v_7, v_3)	pop
τ_{v_7}	(v_6, v_7)	62	(v_7, v_3)	$swap(11)$
	(v_6, v_7)	72	(v_7, v_8)	$swap(22)$
	(v_3, v_8)	22	out_2	pop
τ_{v_8}	(v_7, v_8)	22	out_2	pop

Protected link

Alternative link

Label

Version which does not mask links individually!

local FFT	Out-I	In-Label	Out-I	op
τ_{v_2}	(v_2, v_3)	11	(v_2, v_6)	$push(30)$
	(v_2, v_3)	21	(v_2, v_6)	$push(30)$
	(v_2, v_6)	30	(v_2, v_5)	$push(40)$
global FFT	Out-I	In-Label	Out-I	op
τ'_{v_2}	(v_2, v_3)	11	(v_2, v_6)	$swap(61)$
	(v_2, v_3)	21	(v_2, v_6)	$swap(71)$
	(v_2, v_6)	61	(v_2, v_5)	$push(40)$
	(v_2, v_6)	71	(v_2, v_5)	$push(40)$

Failover Tables

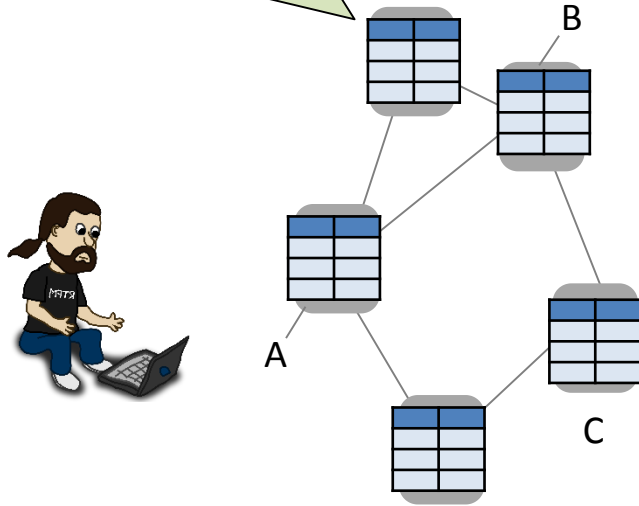
Flow Table

MPLS Tunnels in Today's ISP Networks

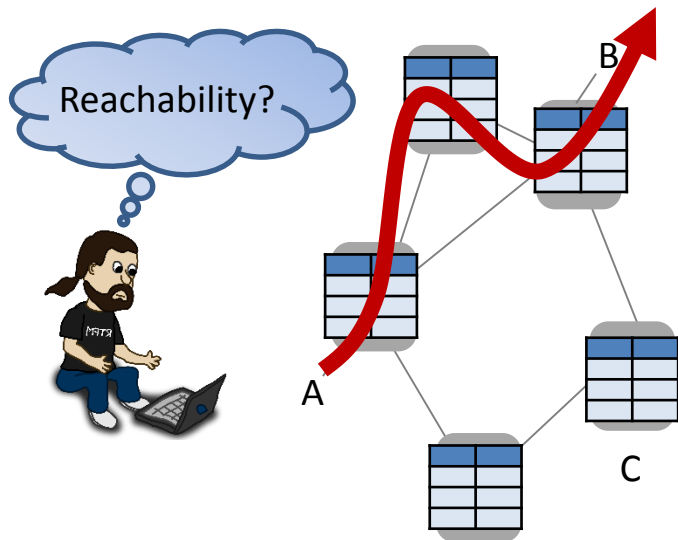


Responsibilities of a Sysadmin

Routers and switches store list of **forwarding rules**, and conditional **failover rules**.



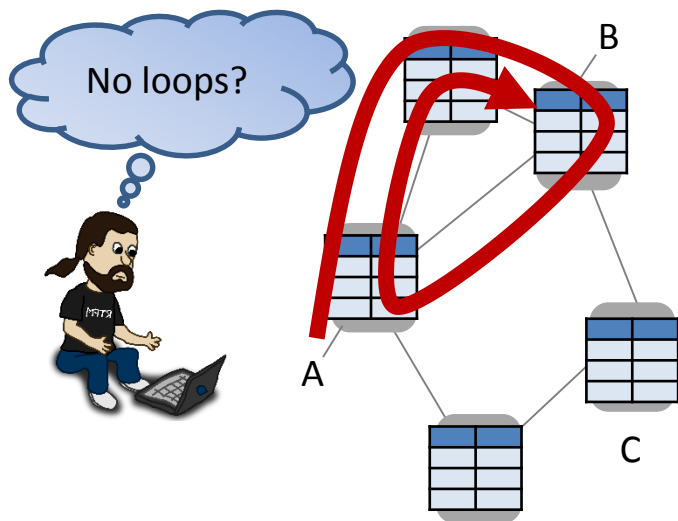
Responsibilities of a Sysadmin



Sysadmin responsible for:

- **Reachability:** Can traffic from ingress port A reach egress port B?

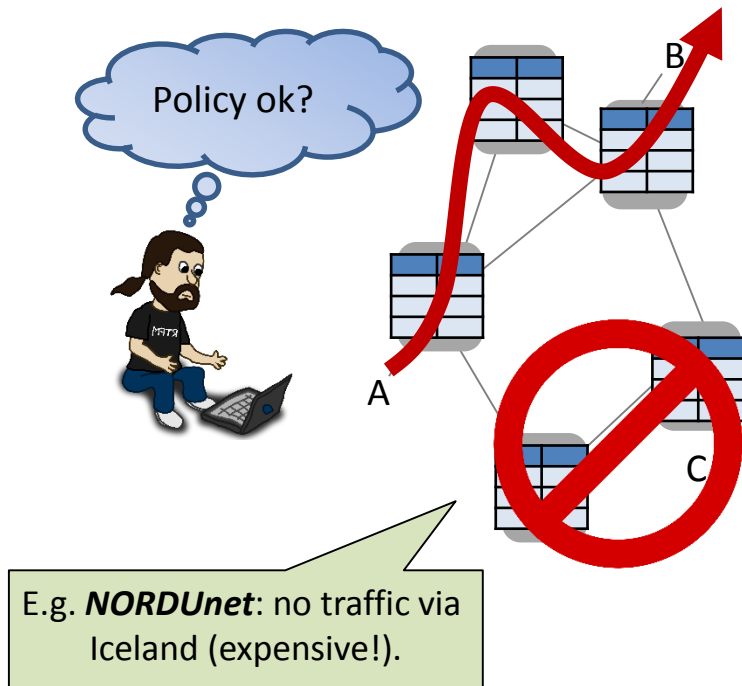
Responsibilities of a Sysadmin



Sysadmin responsible for:

- **Reachability:** Can traffic from ingress port A reach egress port B?
- **Loop-freedom:** Are the routes implied by the forwarding rules loop-free?

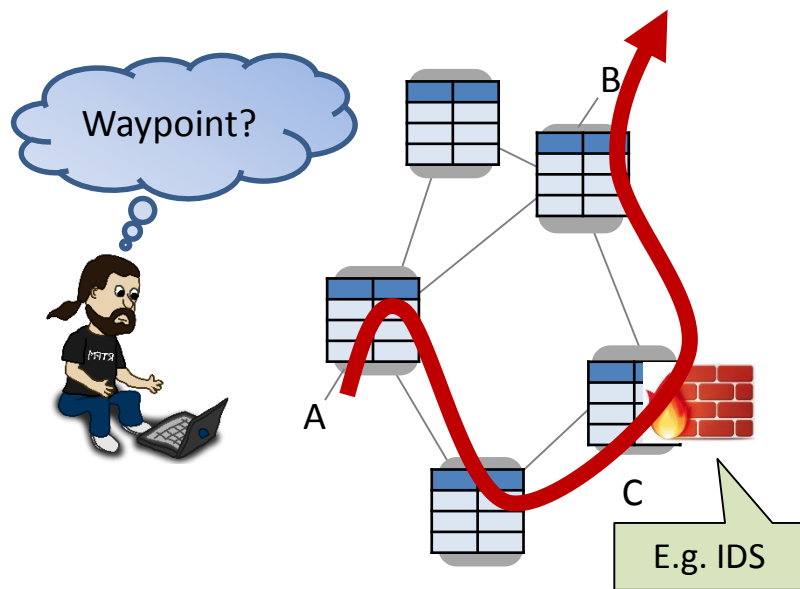
Responsibilities of a Sysadmin



Sysadmin responsible for:

- **Reachability:** Can traffic from ingress port A reach egress port B?
- **Loop-freedom:** Are the routes implied by the forwarding rules loop-free?
- **Policy:** Is it ensured that traffic from A to B never goes via C?

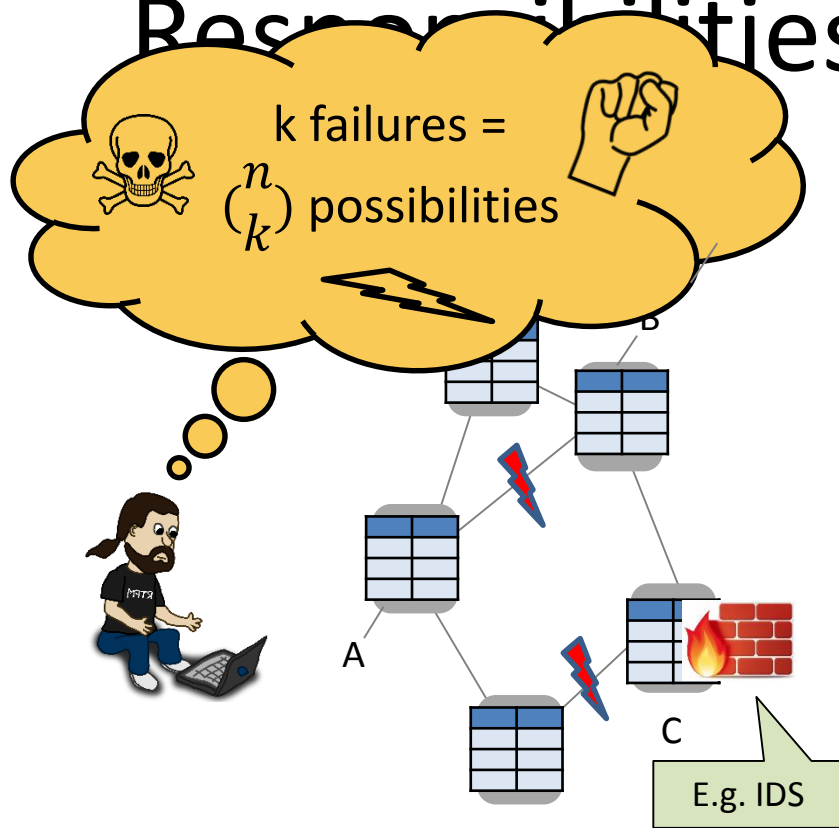
Responsibilities of a Sysadmin



Sysadmin responsible for:

- **Reachability:** Can traffic from ingress port A reach egress port B?
- **Loop-freedom:** Are the routes implied by the forwarding rules loop-free?
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- **Waypoint enforcement:** Is it ensured that traffic from A to B is always routed via a node C (e.g., intrusion detection system or a firewall)?

Responsibilities of a Sysadmin



Sysadmin responsible for:

- **Reachability:** Can traffic from ingress port A reach egress port B?
- **Loop-freedom:** Are the routes implied by the forwarding rules loop-free?
- **Policy:** Is it ensured that traffic from A to B never goes via C?
- **Waypoint enforcement:** Is it ensured that traffic from A to B is always routed via a node C (e.g., intrusion detection system or a firewall)?

... and everything even under multiple failures?!

Can we automate such tests
or even self-repair?

Can we automate such tests or even self-repair?

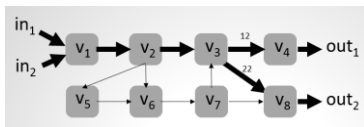


Yes! Automated **What-if Analysis Tool** for
MPLS and SR in ***polynomial time***.
(INFOCOM 2018, CoNEXT 2018, IFIP Networking 2019)

Leveraging Automata-Theoretic Approach

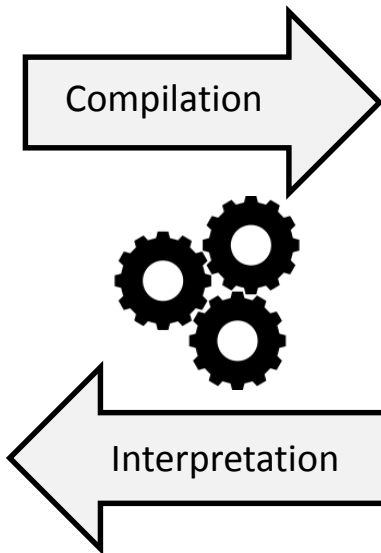


FT	In-I	In-Label	Out-I	op
τ_{v_1}	m_1	\perp	(v_1, v_2)	<i>push</i> (10)
	m_2	\perp	(v_1, v_2)	<i>push</i> (20)
τ_{v_2}	(v_1, v_2)	10	(v_2, v_3)	<i>swap</i> (11)
	(v_1, v_2)	20	(v_2, v_3)	<i>swap</i> (21)
τ_{v_3}	(v_2, v_3)	11	(v_3, v_4)	<i>swap</i> (12)
	(v_2, v_3)	21	(v_3, v_4)	<i>swap</i> (22)
	(v_7, v_3)	11	(v_3, v_4)	<i>swap</i> (12)
	(v_7, v_3)	21	(v_3, v_4)	<i>swap</i> (22)
τ_{v_4}	(v_3, v_4)	12	out_1	<i>pop</i>
τ_{v_5}	(v_3, v_4)	40	(v_5, v_6)	<i>pop</i>
τ_{v_6}	(v_2, v_6)	30	(v_6, v_7)	<i>swap</i> (31)
	(v_5, v_6)	30	(v_6, v_7)	<i>swap</i> (31)
τ_{v_7}	(v_5, v_6)	61	(v_6, v_7)	<i>swap</i> (62)
	(v_5, v_6)	71	(v_6, v_7)	<i>swap</i> (72)
	(v_6, v_7)	31	(v_7, v_2)	<i>pop</i>
	(v_6, v_7)	62	(v_7, v_2)	<i>swap</i> (11)
τ_{v_8}	(v_6, v_7)	72	(v_7, v_8)	<i>swap</i> (22)
	(v_7, v_8)	22	out_2	<i>pop</i>



local FFT	Out-I	In-Label	Out-I	op
τ_{v_2}	(v_2, v_3)	11	(v_2, v_6)	<i>push</i> (30)
	(v_2, v_3)	21	(v_2, v_6)	<i>push</i> (30)
	(v_2, v_6)	30	(v_2, v_5)	<i>push</i> (40)
global FFT	Out-I	In-Label	Out-I	op
τ'_{v_2}	(v_2, v_3)	11	(v_2, v_6)	<i>swap</i> (61)
	(v_2, v_3)	21	(v_2, v_6)	<i>swap</i> (71)
	(v_2, v_6)	61	(v_2, v_5)	<i>push</i> (40)
	(v_2, v_6)	71	(v_2, v_5)	<i>push</i> (40)

MPLS **configurations**,
Segment Routing etc.



$$pX \Rightarrow qXX$$

$$pX \Rightarrow qYX$$

$$qY \Rightarrow rYY$$

$$rY \Rightarrow r$$

$$rX \Rightarrow pX$$

Pushdown Automaton
and **Prefix Rewriting**
Systems Theory

Leveraging Automata

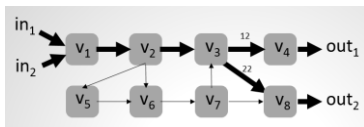
Use cases: Sysadmin *issues queries* to test certain properties, or do it on a *regular basis* automatically!

Approach

What if...?!



FT	In-I	In-Label	Out-I	op
τ_{v_1}	m_1	\perp	(v_1, v_2)	<i>push</i> (10)
	m_2	\perp	(v_1, v_2)	<i>push</i> (20)
τ_{v_2}	(v_1, v_2)	10	(v_2, v_3)	<i>swap</i> (11)
	(v_1, v_2)	20	(v_2, v_3)	<i>swap</i> (21)
τ_{v_3}	(v_2, v_3)	11	(v_3, v_4)	<i>swap</i> (12)
	(v_2, v_3)	21	(v_3, v_4)	<i>swap</i> (22)
τ_{v_4}	(v_2, v_3)	11	(v_3, v_4)	<i>swap</i> (12)
	(v_2, v_3)	21	(v_3, v_4)	<i>swap</i> (22)
τ_{v_5}	(v_3, v_4)	12	out_1	<i>pop</i>
	(v_3, v_4)	40	(v_5, v_6)	<i>pop</i>
τ_{v_6}	(v_2, v_6)	30	(v_6, v_7)	<i>swap</i> (31)
	(v_5, v_6)	30	(v_6, v_7)	<i>swap</i> (31)
τ_{v_7}	(v_5, v_6)	61	(v_6, v_7)	<i>swap</i> (62)
	(v_5, v_6)	71	(v_6, v_7)	<i>swap</i> (72)
τ_{v_8}	(v_6, v_7)	31	(v_7, v_8)	<i>pop</i>
	(v_6, v_7)	62	(v_7, v_8)	<i>swap</i> (11)
τ_{v_9}	(v_6, v_7)	72	(v_7, v_8)	<i>swap</i> (22)
	(v_7, v_8)	22	out_2	<i>pop</i>



local FFT	Out-I	In-Label	Out-I	op
τ_{v_2}	(v_2, v_3)	11	(v_2, v_6)	<i>push</i> (30)
	(v_2, v_3)	21	(v_2, v_6)	<i>push</i> (30)
	(v_2, v_6)	30	(v_2, v_5)	<i>push</i> (40)
global FFT	Out-I	In-Label	Out-I	op
τ'_{v_2}	(v_2, v_3)	11	(v_2, v_6)	<i>swap</i> (61)
	(v_2, v_3)	21	(v_2, v_6)	<i>swap</i> (71)
	(v_2, v_6)	61	(v_2, v_5)	<i>push</i> (40)
	(v_2, v_6)	71	(v_2, v_5)	<i>push</i> (40)

Compilation



Interpretation

$$pX \Rightarrow qXX$$

$$pX \Rightarrow qYX$$

$$qY \Rightarrow rYY$$

$$rY \Rightarrow r$$

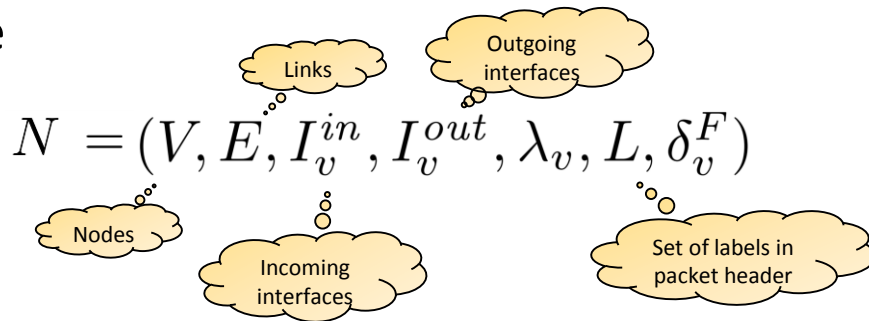
$$rX \Rightarrow pX$$

MPLS *configurations*,
Segment Routing etc.

Pushdown Automaton
and *Prefix Rewriting*
Systems Theory

Mini-Tutorial: A Network Model

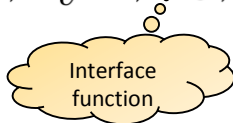
- Network: a 7-tuple



Mini-Tutorial: A Network Model

- Network: a 7-tuple

$$N = (V, E, I_v^{in}, I_v^{out}, \lambda_v, L, \delta_v^F)$$



Interface function: maps outgoing interface to next hop node and incoming interface to previous hop node

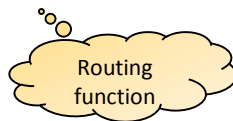
$$\lambda_v : I_v^{in} \cup I_v^{out} \rightarrow V$$

That is: $(\lambda_v(in), v) \in E$ and $(v, \lambda_v(out)) \in E$

Mini-Tutorial: A Network Model

- Network: a 7-tuple

$$N = (V, E, I_v^{in}, I_v^{out}, \lambda_v, L, \delta_v^F)$$



Routing function: for each set of **failed links** $F \subseteq E$, the routing function

$$\delta_v^F : I_v^{in} \times L^* \rightarrow 2^{(I_v^{out} \times L^*)}$$

defines, for all **incoming interfaces** and packet **headers**, **outgoing interfaces** together with **modified headers**.

Routing in Network

Packet routing sequence can be represented using **sequence of tuples**:

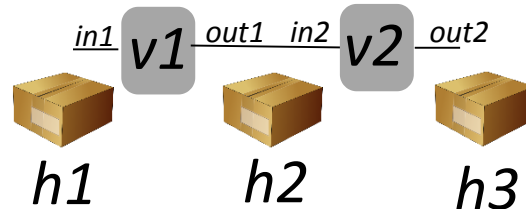


- Example: **routing** (in)finite sequence of tuples

$(v_1, in_1, h_1, out_1, h_2, F_1),$

$(v_2, in_2, h_2, out_2, h_3, F_2),$

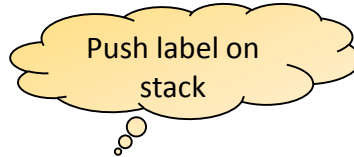
...



Example Rules:

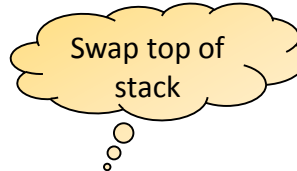
Regular Forwarding on Top-Most Label

Push:



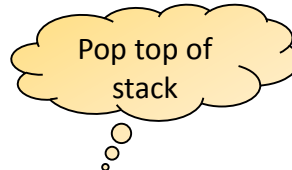
$$(v, in)\ell \rightarrow (v, out, 0)\ell'\ell \text{ if } \tau_v(in, \ell) = (out, push(\ell'))$$

Swap:



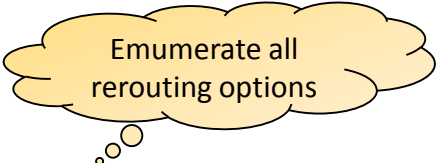
$$(v, in)\ell \rightarrow (v, out, 0)\ell' \text{ if } \tau_v(in, \ell) = (out, swap(\ell'))$$

Pop:



$$(v, in)\ell \rightarrow (v, out, 0) \text{ if } \tau_v(in, \ell) = (out, pop)$$

Example *Failover* Rules



Emumerate all rerouting options

Failover-Push:

$(v, out, i)\ell \rightarrow (v, out', i + 1)\ell'\ell$ for every i , $0 \leq i < k$,
where $\pi_v(out, \ell) = (out', push(\ell'))$

Failover-Swap:

$(v, out, i)\ell \rightarrow (v, out', i + 1)\ell'$ for every i , $0 \leq i < k$,
where $\pi_v(out, \ell) = (out', swap(\ell'))$,

Failover-Pop:

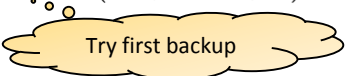
$(v, out, i)\ell \rightarrow (v, out', i + 1)$ for every i , $0 \leq i < k$,
where $\pi_v(out, \ell) = (out', pop)$.

Example rewriting sequence:

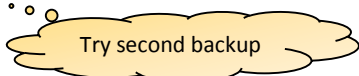
$(v_1, in_1)h_1\perp \rightarrow (v_1, out, 0)h\perp \rightarrow (v_1, out', 1)h'\perp \rightarrow (v_1, out'', 2)h''\perp \rightarrow \dots \rightarrow (v_1, out_1, i)h_2\perp$



Try default



Try first backup



Try second backup

A Complex and Big Formal Language!

Why Polynomial Time?!



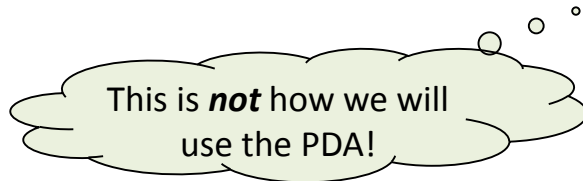
- Arbitrary number k of failures: How can I avoid checking all $\binom{n}{k}$ many options?!
- Even if we reduce to **push-down automaton**: simple operations such as **emptiness testing** or **intersection on Push-Down Automata (PDA)** is computationally non-trivial and sometimes even **undecidable**!

A Complex and Big Formal Language!

Why Polynomial Time?!



- Arbitrary number k of failures: How can I avoid checking all $\binom{n}{k}$ many options?!
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A Complex and Big Formal Language!

Why Polynomial Time?!



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The words in our language are sequences of pushdown stack symbols, not the labels of transitions.

Time for Automata Theory (from Switzerland)!

- Classic result by **Büchi** 1964: the set of all reachable configurations of a pushdown automaton is a **regular set**
- Hence, we can operate only on **Nondeterministic Finite Automata (NFAs)** when reasoning about the pushdown automata
- The resulting **regular operations** are all **polynomial time**
 - Important result of **model checking**



Julius Richard Büchi

1924-1984

Swiss logician

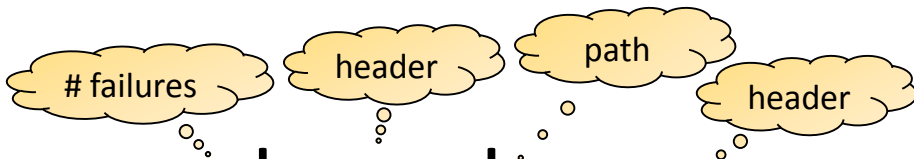
Tool and Query Language

Part 1: Parses query and constructs Push-Down System (PDS)

- In Python 3

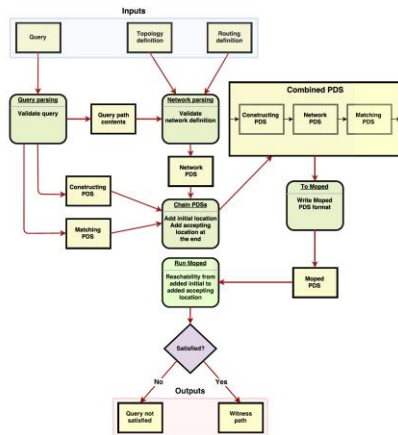
Part 2: Reachability analysis of constructed PDS

- Using *Moped* tool



$k \langle a \rangle \quad b \langle c \rangle$

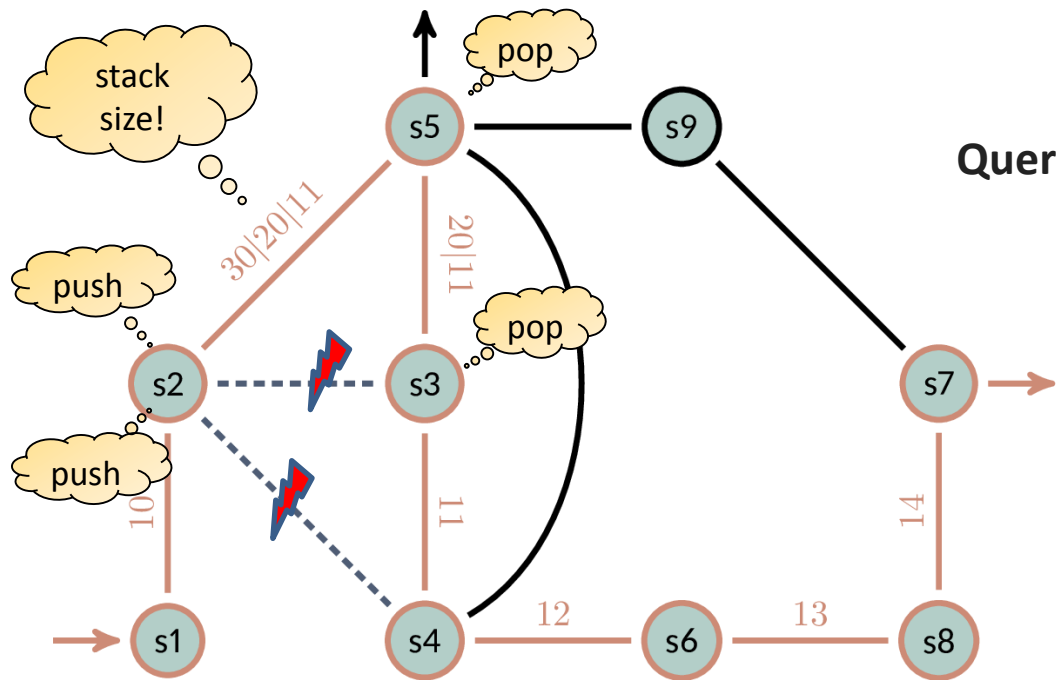
Regular query language



query processing flow

Example: Traversal Testing With 2 Failures

Traversal test with $k=2$: Can traffic starting with `[]` go **through s5**, under up to **$k=2$ failures**?



Query: $k=2$ `[] s1 >> s5 >> s7 []`

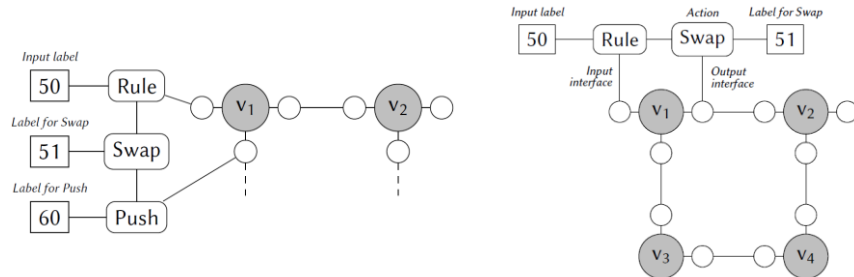
2 failures

YES!
(Gives witness!)

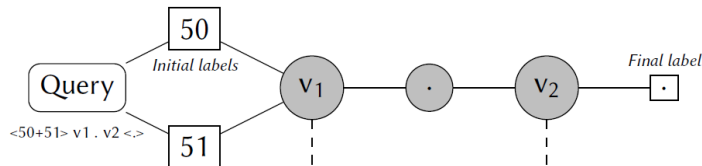
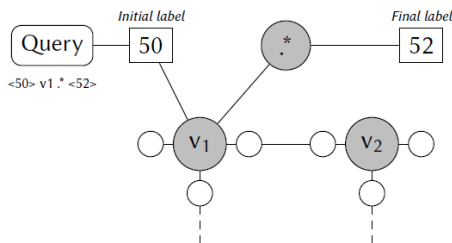
Formal methods are nice (give guarantees!)... But what about ML...?!

Speed Up Further and Synthesize: Deep Learning

- Yes sometimes **without losing guarantees**
- Extend **graph-based neural networks**
- **Predict** counter-examples and **fixes**



Network topologies and MPLS rules



Network topologies and query

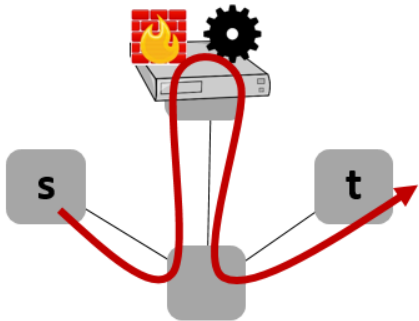
Roadmap

- Opportunities of self-* networks
 - Example 1: Demand-aware, self-adjusting networks
 - Example 2: Self-repairing networks
- Challenges of designing self-* networks

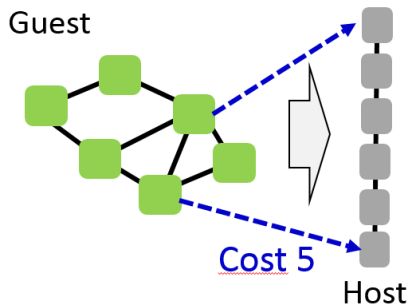


Challenge 1: Hard Problems

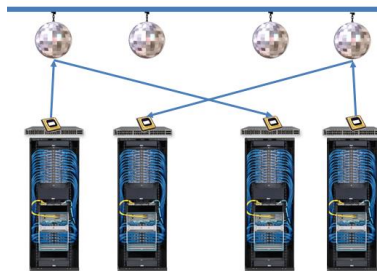
- Optimization problems are often **NP-hard**: hard *even for computers*!



Waypoint routing:
disjoint paths

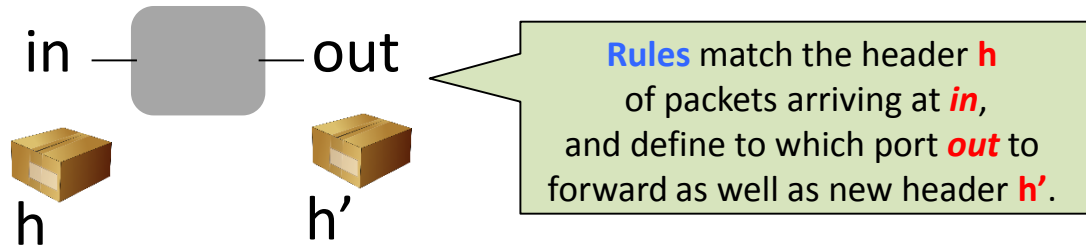


Embedding:
Minimum Lin. Arrangement



Topology design:
Graph spanners

It can get worse...: intractable!



prefix rewriting

$in \times L \rightarrow out \times OP$

where $OP = \{swap, push, pop\}$

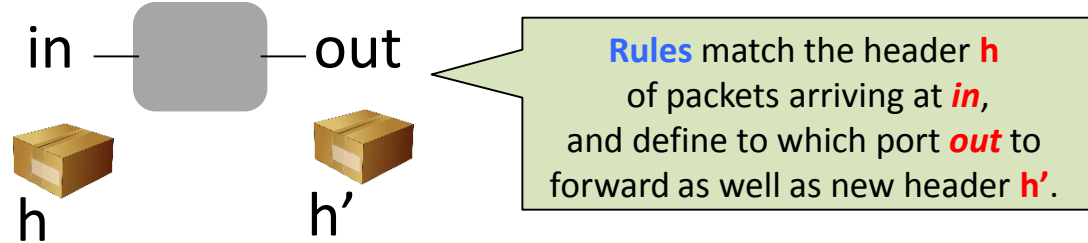
VS

Rules of general networks (e.g., SDN):

arbitrary header rewriting

$in \times L^* \rightarrow out \times L^*$

It can get worse...: intractable!



(Simplified) MPLS rules:

prefix rewriting

Polynomial time

$in \times OP$

where $OP = \{swap, push, pop\}$

VS

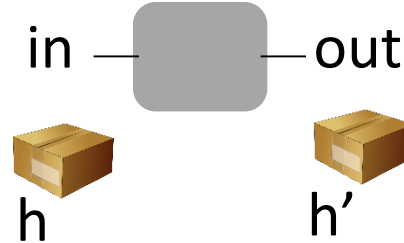
Rules of general networks (e.g., SDN):

arbitrary header rewriting

Undecidable!

$in \rightarrow out \times L^*$

It can get worse...: intractable!



What is a good tradeoff between generality and performance?

(Simple
pre

Polynomial $n \times O$

where $OP = \{swap, push, pop\}$

$in \rightarrow out \times L^*$

Challenge 2: Realizing Limits?

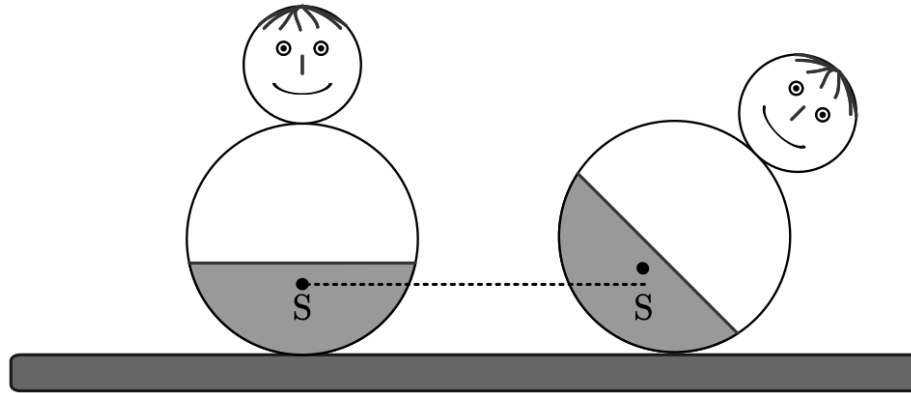
- Can a self-* network realize its **limits**?
- E.g., when quality of **input data** is not good enough?
- When to hand over to human? Or **fall back** to „safe/oblivious mode“?
- Can we learn from self-driving **cars**?



Challenge 3: Self-Stabilization

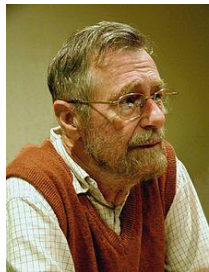
- Could be an attractive property of self-* network!

A **self-stabilizing** system guarantees that it *reconverges to a desirable configuration or state, from any initial state*.



„Stehaufmännchen“

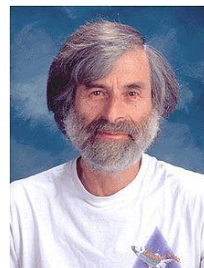
Self-Stabilization



Self-stabilizing algorithms pioneered by **Dijkstra** (1973): for example **self-stabilizing mutual exclusion**.

“I regard this as Dijkstra’s most brilliant work. Self-stabilization is a very important concept in **fault tolerance**.”

Leslie **Lamport** (PODC 1983)

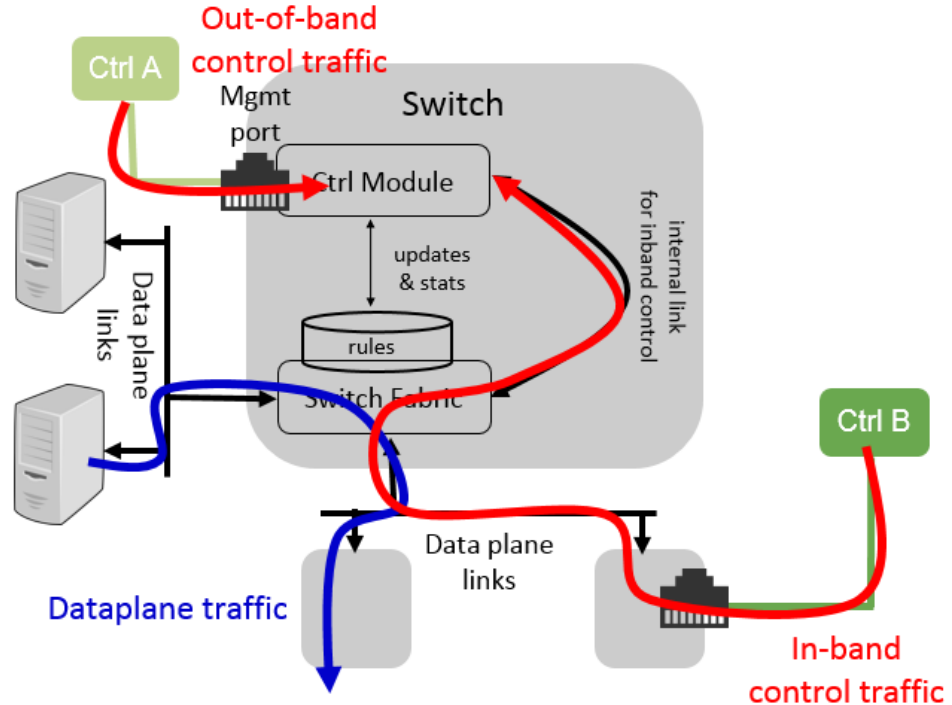


Some notable works by **Perlman** toward self-stabilizing Internet, e.g., **self-stabilizing spanning trees**.

Yet, many protocols in the Internet are *not* self-stabilizing. Much need for future work.

E.g., Self-Stabilizing SDN Control?

- Distributed SDN control plane which **self-organizes management** of switches?
- Especially challenging: **inband control** (how to distinguish traffic?)



Challenge 4: Uncertainties

- How to deal with **uncertainties**?
- How to maintain flexibilities?
- Use of principles from robotics? E.g., **empowerment**?

Conclusion

- **Flexibilities** in networks: great opportunities for **optimization** and **automation**
- **Demand-aware** and **self-adjusting** networks: beating the routing lower bounds of oblivious networks, *reaching entropy bounds*
- Potential of **self-repairing** networks, self-stabilizing networks, etc.
- Much work ahead: *tradeoff* generality vs efficiency? How to self-monitor and **fall-back** if needed? Use of **formal methods** and ML?

Flexibilities and Complexity

[On The Impact of the Network Hypervisor on Virtual Network Performance](#)

Andreas Blenk, Arsany Basta, Wolfgang Kellerer, and Stefan Schmid.

IFIP Networking, Warsaw, Poland, May 2019.

[Adaptable and Data-Driven Softwarized Networks: Review, Opportunities, and Challenges](#) (Invited Paper)

Wolfgang Kellerer, Patrick Kalmbach, Andreas Blenk, Arsany Basta, Martin Reisslein, and Stefan Schmid.

Proceedings of the IEEE (**PIEEE**), 2019.

[Efficient Distributed Workload \(Re-\)Embedding](#)

Monika Henzinger, Stefan Neumann, and Stefan Schmid.

ACM/IFIP **SIGMETRICS/PERFORMANCE**, Phoenix, Arizona, USA, June 201

[Parametrized Complexity of Virtual Network Embeddings: Dynamic & Linear Programming Approximations](#)

Matthias Rost, Elias Döhne, and Stefan Schmid.

ACM SIGCOMM Computer Communication Review (**CCR**), January 2019.

[Charting the Complexity Landscape of Virtual Network Embeddings](#) (Best Paper Award)

Matthias Rost and Stefan Schmid.

IFIP Networking, Zurich, Switzerland, May 2018.

[Tomographic Node Placement Strategies and the Impact of the Routing Model](#)

Yvonne Anne Pignolet, Stefan Schmid, and Gilles Tredan.

ACM **SIGMETRICS**, Irvine, California, USA, June 2018. hmid.

ACM/IEEE Symposium on Architectures for Networking and Communications Systems (**ANCS**), Ithaca, New York, USA, July 2018.

Demand-Aware and Self-Adjusting Networks

[Toward Demand-Aware Networking: A Theory for Self-Adjusting Networks](#) (Editorial)

Chen Avin and Stefan Schmid.

ACM SIGCOMM Computer Communication Review (**CCR**), October 2018.

[Demand-Aware Network Design with Minimal Congestion and Route Lengths](#)

Chen Avin, Kaushik Mondal, and Stefan Schmid.

38th IEEE Conference on Computer Communications (**INFOCOM**), Paris, France, April 2019.

Documents: paper [pdf](#), bibtex [bib](#)

[Distributed Self-Adjusting Tree Networks](#)

Bruna Peres, Otavio Augusto de Oliveira Souza, Olga Goussevskaia, Chen Avin, and Stefan Schmid.

38th IEEE Conference on Computer Communications (**INFOCOM**), Paris, France, April 2019.

[Efficient Non-Segregated Routing for Reconfigurable Demand-Aware Networks](#)

Thomas Fenz, Klaus-Tycho Foerster, Stefan Schmid, and Anaïs Villedieu.

IFIP Networking, Warsaw, Poland, May 2019.

[Demand-Aware Network Designs of Bounded Degree](#)

Chen Avin, Kaushik Mondal, and Stefan Schmid.

31st International Symposium on Distributed Computing (**DISC**), Vienna, Austria, October 2017.

[SplayNet: Towards Locally Self-Adjusting Networks](#)

Stefan Schmid, Chen Avin, Christian Scheideler, Michael Borokhovich, Bernhard Haeupler, and Zvi Lotker.

IEEE/ACM Transactions on Networking (**TON**), Volume 24, Issue 3, 2016. Early version: IEEE **IPDPS** 2013.

[Characterizing the Algorithmic Complexity of Reconfigurable Data Center Architectures](#)

Klaus-Tycho Foerster, Monia Ghobadi, and Stefan Schmid.

ACM/IEEE Symposium on Architectures for Networking and Communications Systems (**ANCS**), Ithaca, New York, USA, July 2018.

Self-Repairing Networks

[P-Rex: Fast Verification of MPLS Networks with Multiple Link Failures](#)

Jesper Stenbjerg Jensen, Troels Beck Krogh, Jonas Sand Madsen, Stefan Schmid, Jiri Srba, and Marc Tom Thorgersen.
14th International Conference on emerging Networking EXperiments and Technologies (**CoNEXT**), Heraklion, Greece, December 2018.

[Polynomial-Time What-If Analysis for Prefix-Manipulating MPLS Networks](#)

Stefan Schmid and Jiri Srba.

37th IEEE Conference on Computer Communications (**INFOCOM**), Honolulu, Hawaii, USA, April 2018.

[Renaissance: A Self-Stabilizing Distributed SDN Control Plane](#)

Marco Canini, Iosif Salem, Liron Schiff, Elad Michael Schiller, and Stefan Schmid.

38th IEEE International Conference on Distributed Computing Systems (**ICDCS**), Vienna, Austria, July 2018.

[Empowering Self-Driving Networks](#)

Patrick Kalmbach, Johannes Zerwas, Peter Babarczi, Andreas Blenk, Wolfgang Kellerer, and Stefan Schmid.

ACM SIGCOMM 2018 Workshop on Self-Driving Networks (**SDN**), Budapest, Hungary, August 2018.

[DeepMPLS: Fast Analysis of MPLS Configurations using Deep Learning](#)

Fabien Geyer and Stefan Schmid.

IFIP Networking, Warsaw, Poland, May 2019.