

Self-driving/monitoring networks

in the age of deep network programmability



Laurent Vanbever

nsg.ee.ethz.ch

TMA

June 19 2019

1962



Paul Baran (1926–2011)

American electrical engineer

created the notion of a distributed network
which could maintain communications
in the face of a thermonuclear attack

inventor of packet switching
(with Donald Davies)

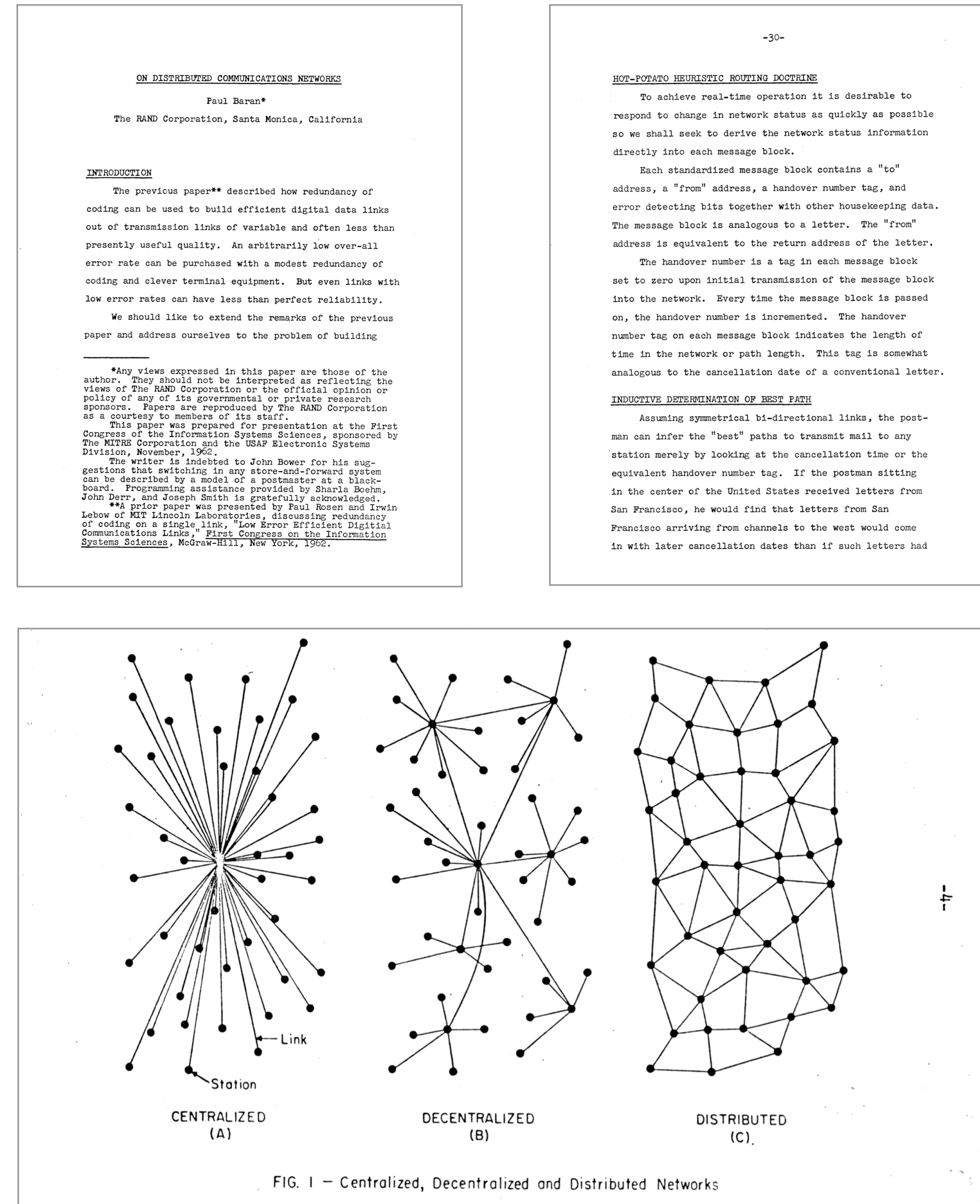
On distributed communication networks

Paul Baran—Sept. 1962

Introduces the concept of

- packet switching
- distributed routing
- reliable transport

among many others...



Thanks to Paul Barlan et al.

the Internet infrastructure is extremely resilient

ability to route packets around failures
(not necessarily fast though)

Thanks to Paul Barlan et al.

the Internet infrastructure is **extremely resilient**

... at least when it comes to **thermonuclear wars**...

Thanks to Paul Barlan et al.

the Internet infrastructure is **extremely resilient**

... at least when it comes to thermonuclear wars...

but how does it fare against, say...

Thanks to Paul Barlan et al.

the Internet infrastructure is **extremely resilient**

... at least when it comes to thermonuclear wars...

but how does it fare against, say... **us humans?**

engadget

Gear

Gaming

Entertainment

Tomorrow

The Buyer's Guide

Video

Reviews

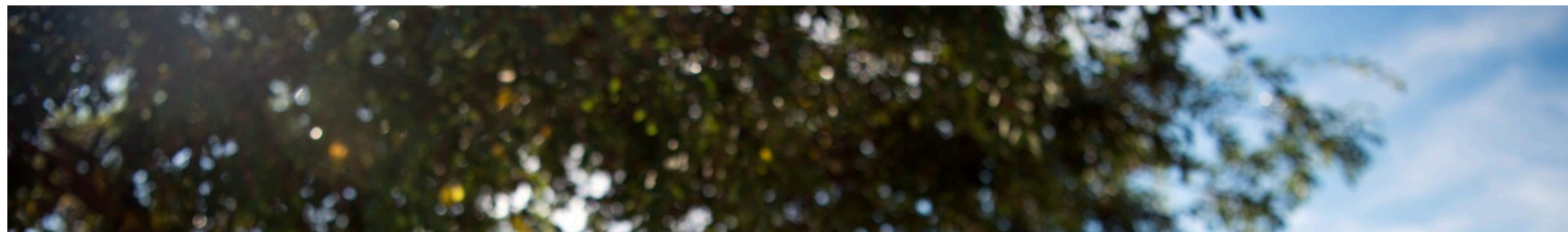
US Edition

Google accidentally broke the internet throughout Japan

A mistake led to internet outages for about half of the country.



Mallory Locklear, @mallorylocklear
08.28.17 in [Internet](#)



27 August 2017

Google made a configuration mistake which caused their Chicago point-of-presence to wrongly advertise 160k IP prefixes to its neighbors.

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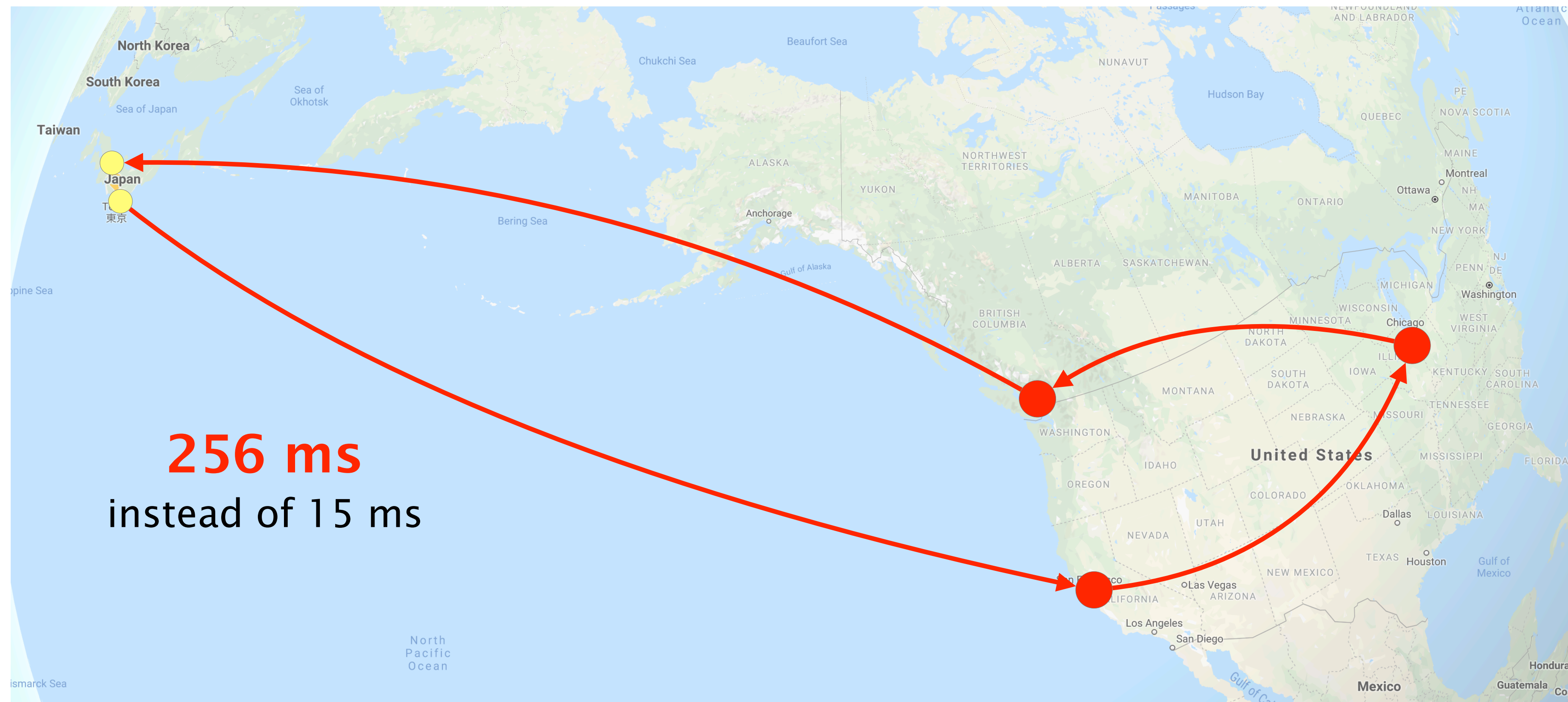
These advertisements propagated in the Internet and got picked up by Japanese giants (IIJ and KDDI) which started to direct local Japanese traffic to Google.

Tokyo to Nagoya (normally)



Tokyo to Nagoya via... Chicago?!

A **17x** increased latency!



Google made a configuration mistake which caused their Chicago point-of-presence to wrongly advertise 160k IP prefixes to its neighbors.

These advertisements propagated in the Internet and got picked up by Japanese giants (NTT and KDDI) which started to direct local, Japanese traffic to Google.

The outage in Japan *only* lasted a couple of hours but was so severe that the country's ministries wanted carriers to report on what went wrong.

This is **far** from being an isolated event...

JUL 8, 2015 @ 03:36 PM 11,261 VIEWS

United Airlines Blames Router for Grounded Flights

'Configuration Error' Blamed for AWS Outage

By David Ramel ■ 08/12/2015



The summer of network misconfigurations

CONNECTIVITY MANAGEMENT FIREWALL CHANGE MANAGEMENT
SECURITY RISK MANAGEMENT AND VULNERABILITIES
ICY MANAGEMENT



Amazon's massive AWS outage was caused by human error

One incorrect command and the whole internet suffers.

By Jason Del Rey | @DelRey | Mar 2, 2017, 2:20pm EST

Data Centre ► Networks

Level3 switch config blunder for US-wide VoIP blackout

CenturyLink: 750 calls to 911 missed during Aug. 1 outage caused by human error in Minnesota, North Dakota

By [Barry Amundson](#) on Aug 15, 2018 at 4:43 p.m.

affected Comcast, Spectrum, Verizon and AT&T customers

BY: CNN
POSTED: 3:45 PM Nov 6, 2017

Data Centre ► Networks

CloudFlare apologizes for Telia screwing you over

Unhappy about

By [Kieren McCarthy](#) in

Facebook struggles to deal with epic outage



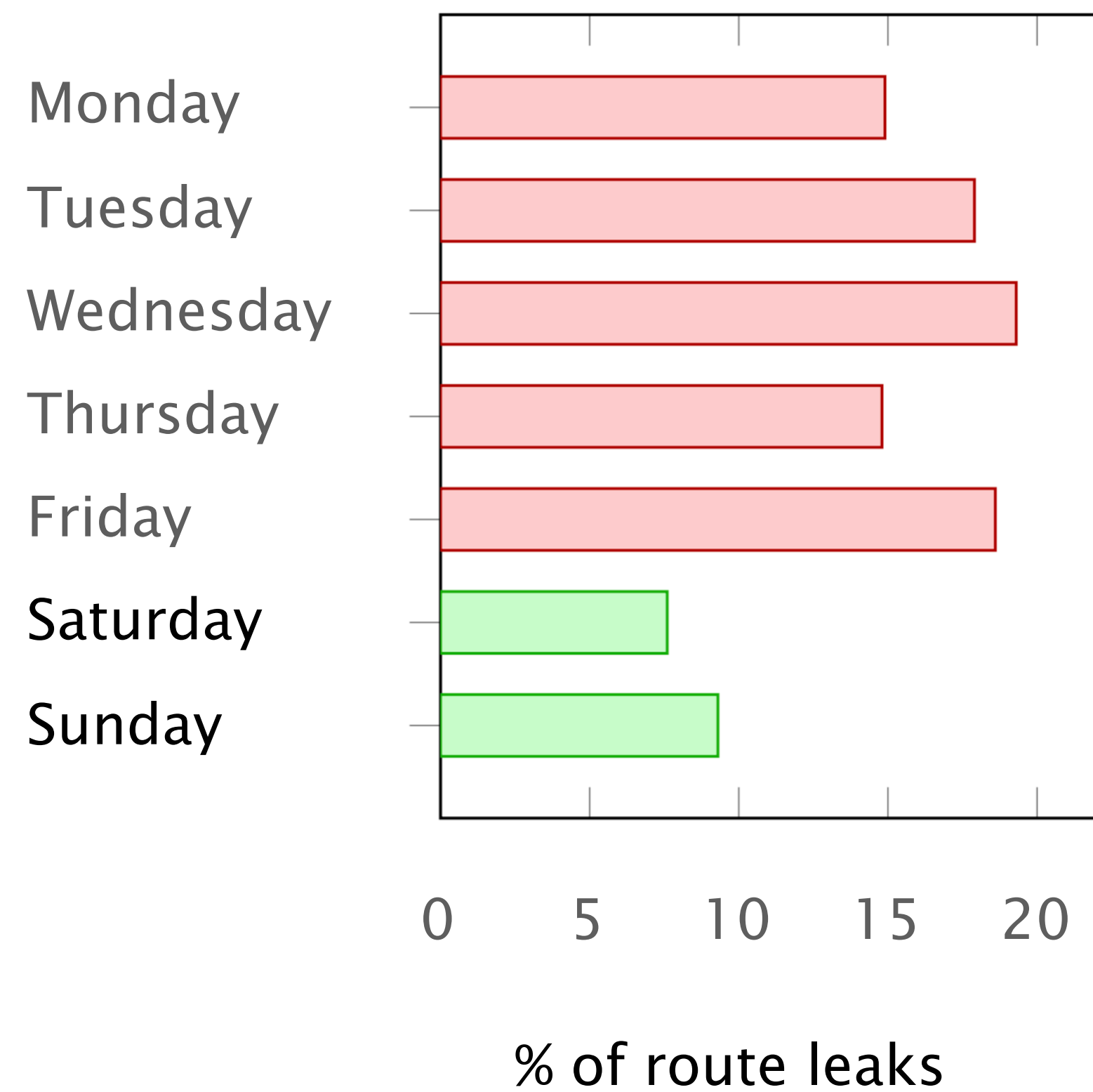
By [Donie O'Sullivan](#) and [Heather Kelly](#), [CNN Business](#)

Updated 0654 GMT (1454 HKT) March 14, 2019

“Human factors are responsible
for 50% to 80% of network outages”

Juniper Networks, *What's Behind Network Downtime?*, 2008

A perhaps ironic consequence is that the Internet works better during the week-ends



source: Job Snijders, 2008–2016

“Human factors are responsible
for 50% to 80% of network outages”

Juniper Networks, *What's Behind Network Downtime?*, 2008

“Human factors are responsible
for [REDACTED] of [REDACTED]”

“Human factors are responsible
for **>90%** of **car accidents**”

NHTSA, National Motor Vehicle Crash Causation Survey, February 2015

Enters...

Self-Driving Car

Enters...

Self-Driving Car

Wikipedia

A vehicle that is capable of
monitoring its environment and
moving with little or no human input

Enters...

Self-Driving Car

Wikipedia

A vehicle that is capable of
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Promise

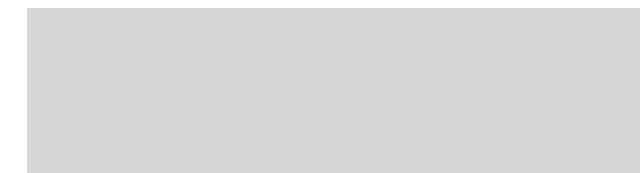
A drastic reduction in the number of (fatal) accidents

Enters...

Self-Driving Car

Enters...

Self-Driving



Enters...

Self-Driving Network

Enters...

Self-Driving Network

Definition

A network that is capable of
monitoring its environment and
adapting its behavior accordingly
with little or no human input

Enters...

Self-Driving Network

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A network that is capable of
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Questions

How do we build and deploy such networks?

Enters...

Self-Driving Network

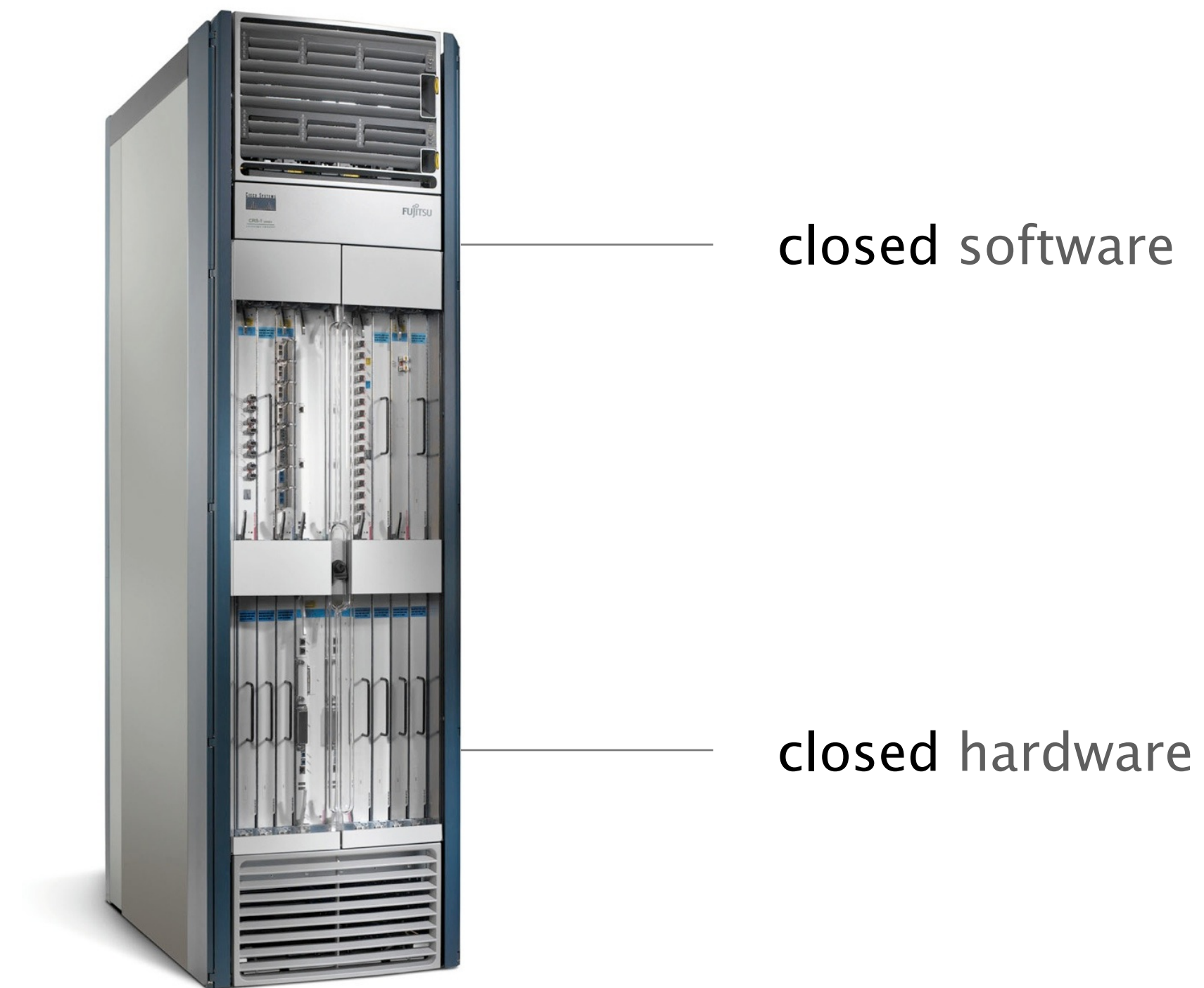
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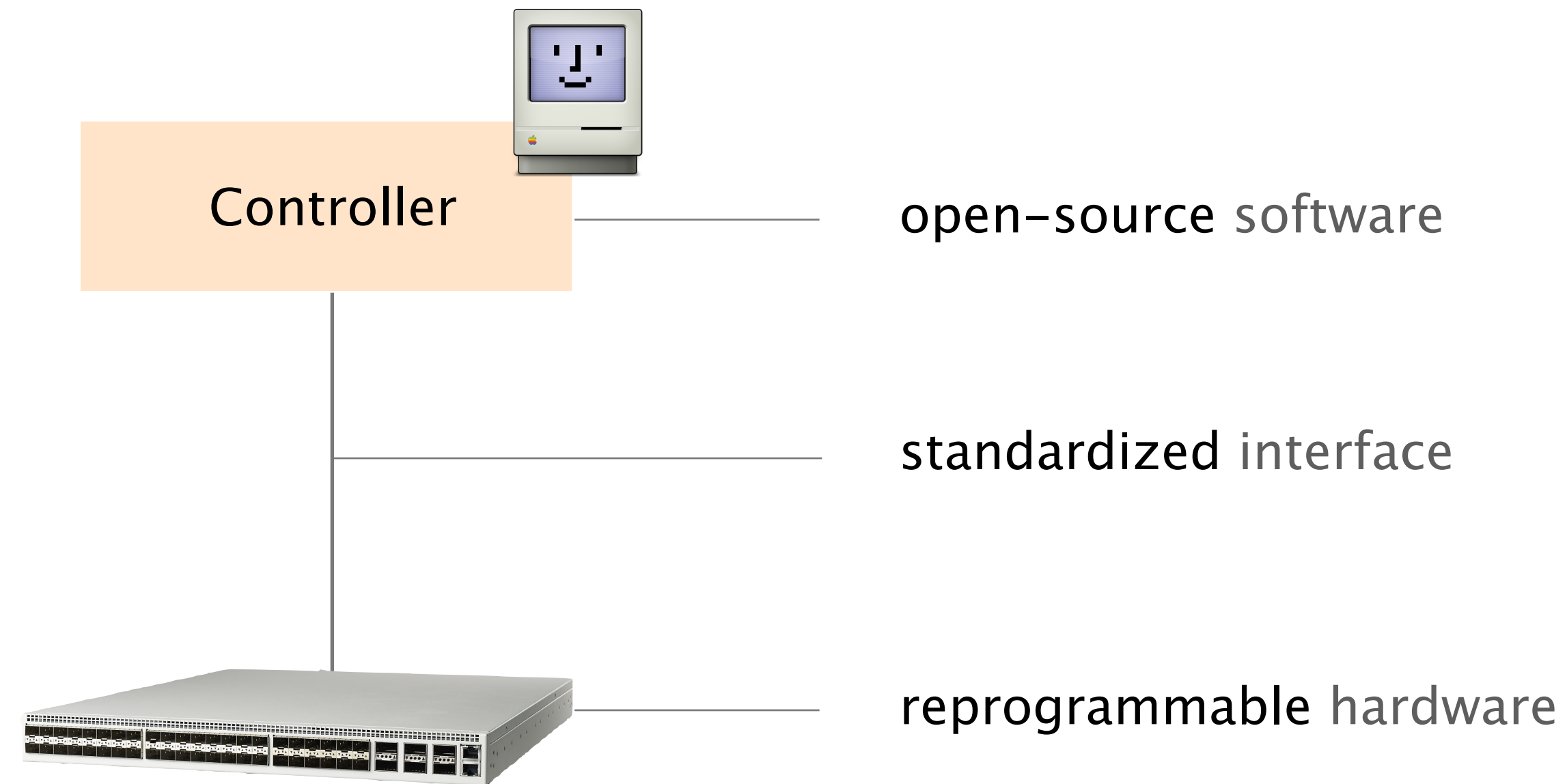
Until recently, innovating in networks was hard
because devices were completely locked down

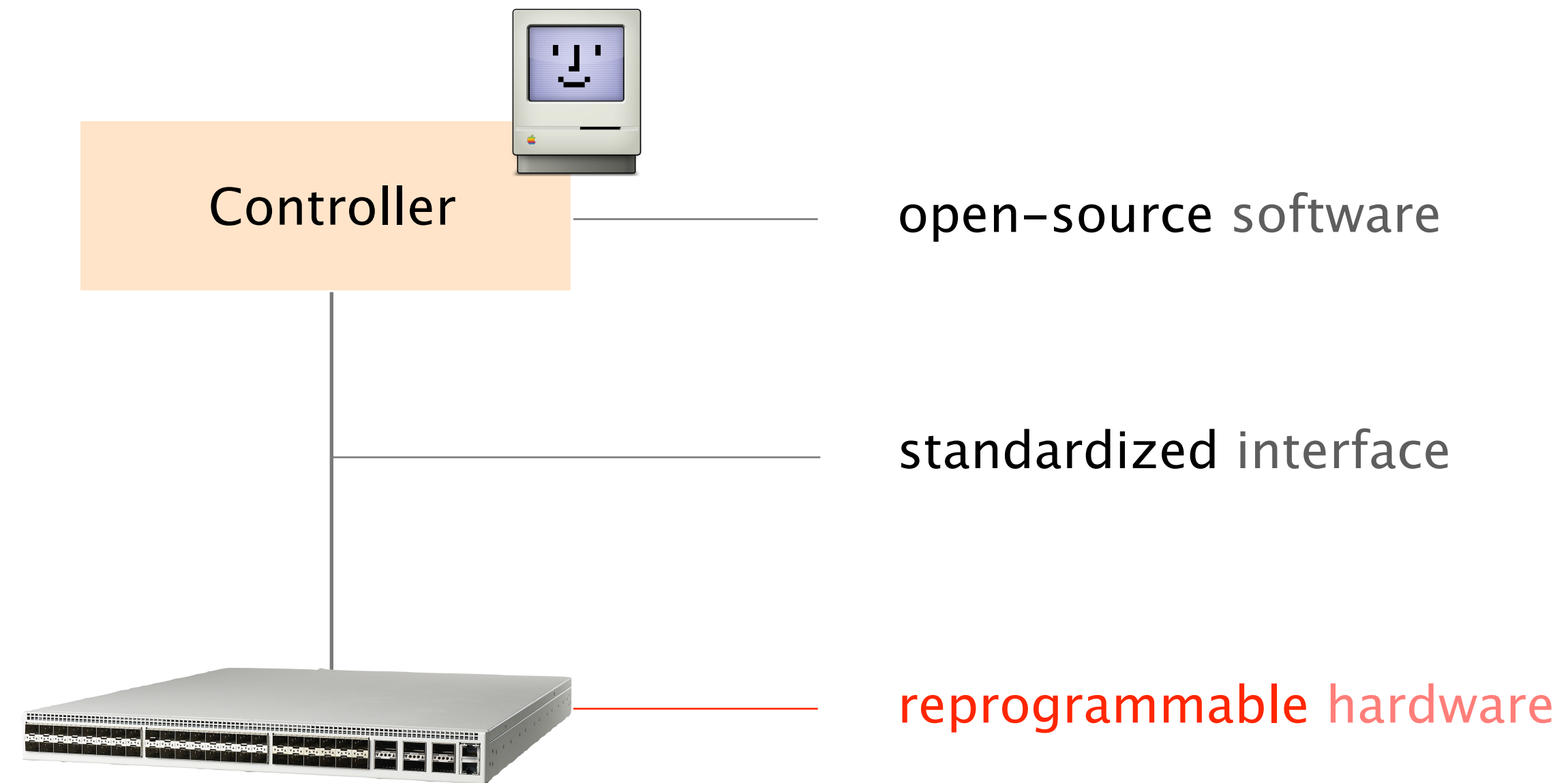


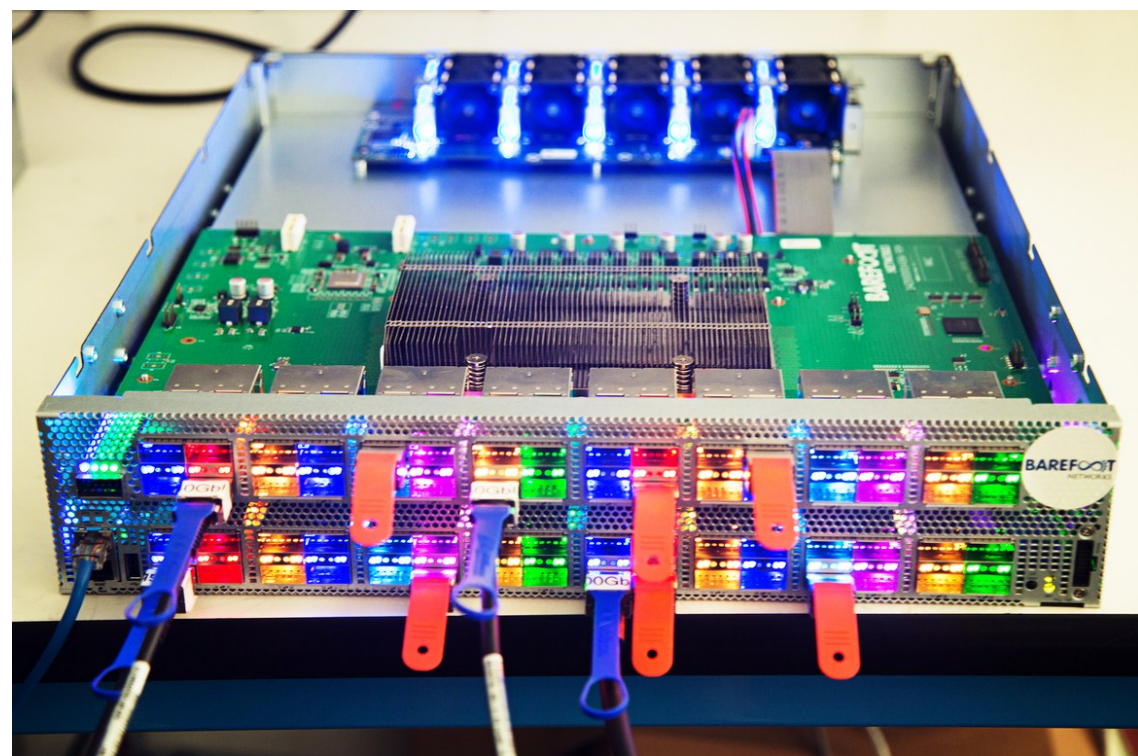
Cisco™ device

Things are changing though!

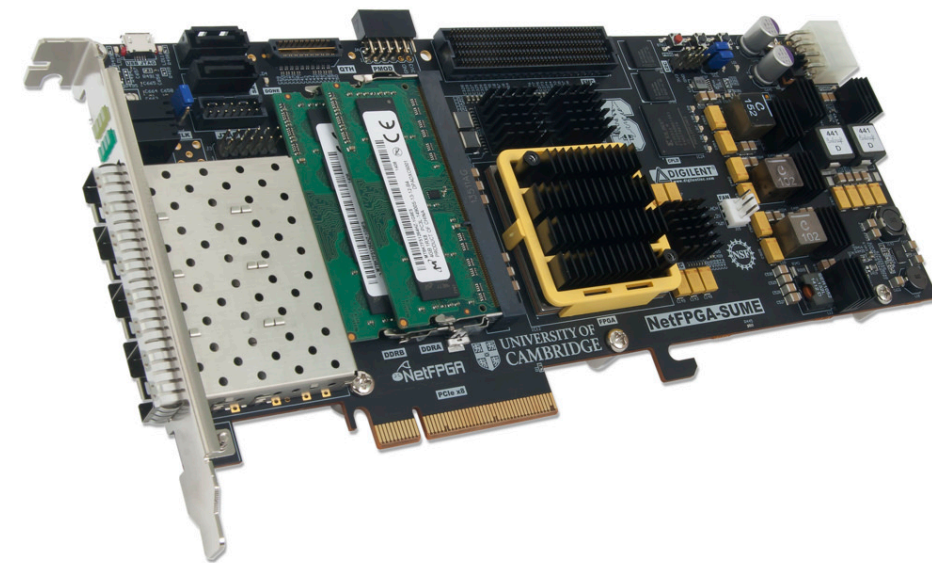
Networks are on the verge of a paradigm shift
towards **deep programmability**





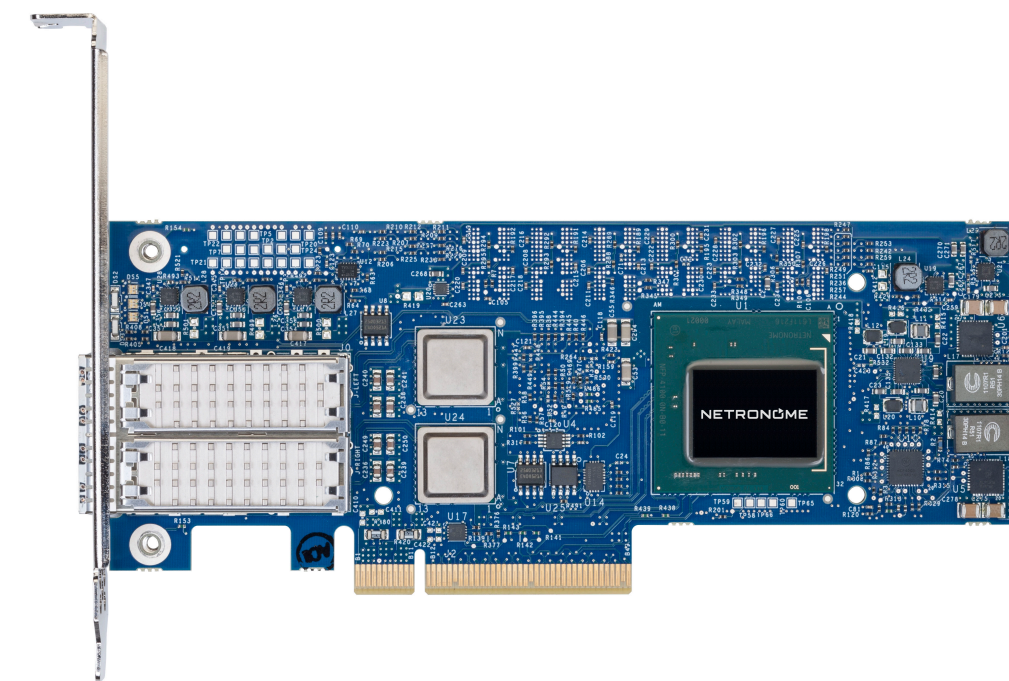


Barefoot Tofino 12.8 Tbps
programmable network switch



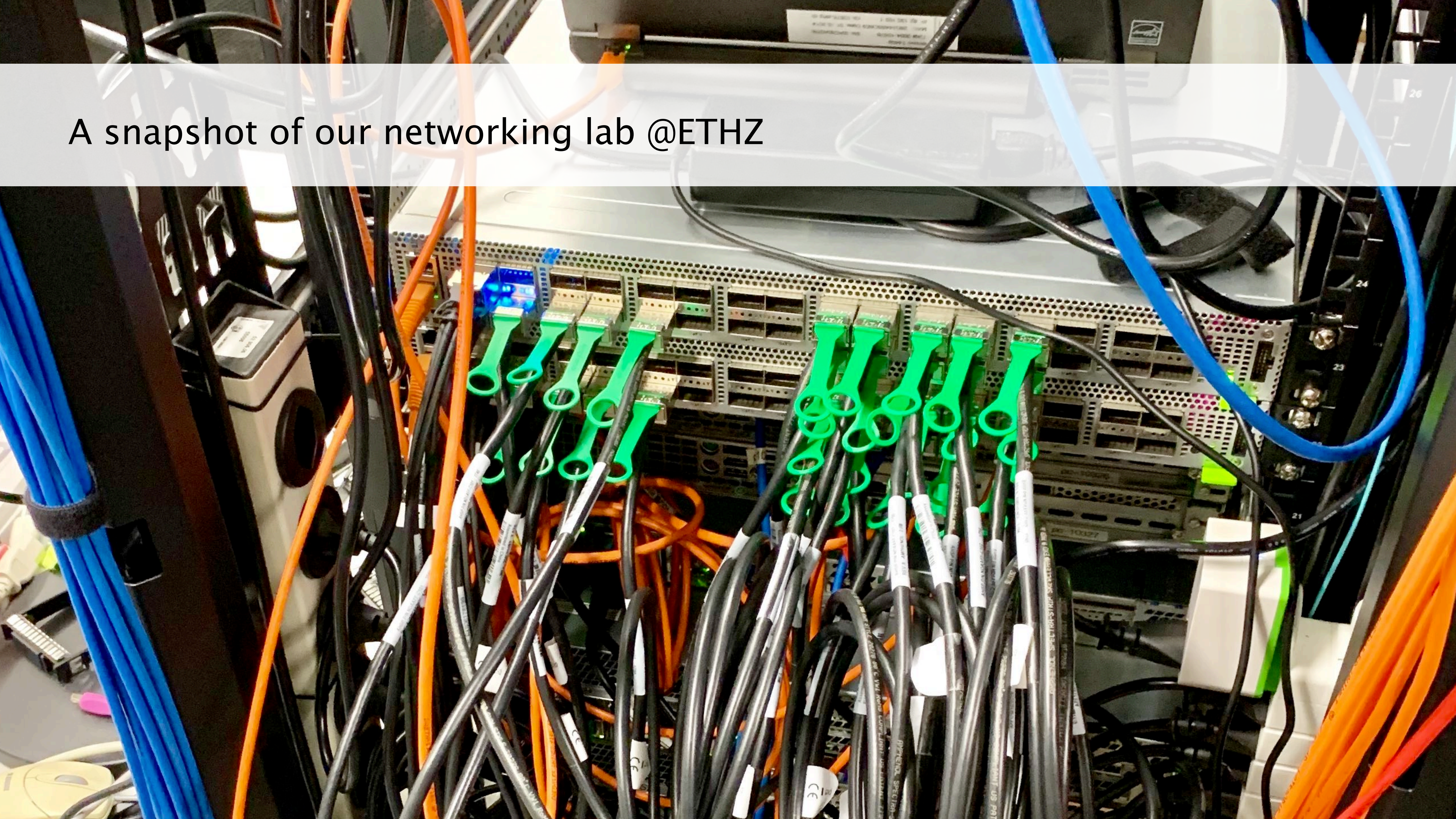
NetFPGA SUME

100 Gbps programmable network card

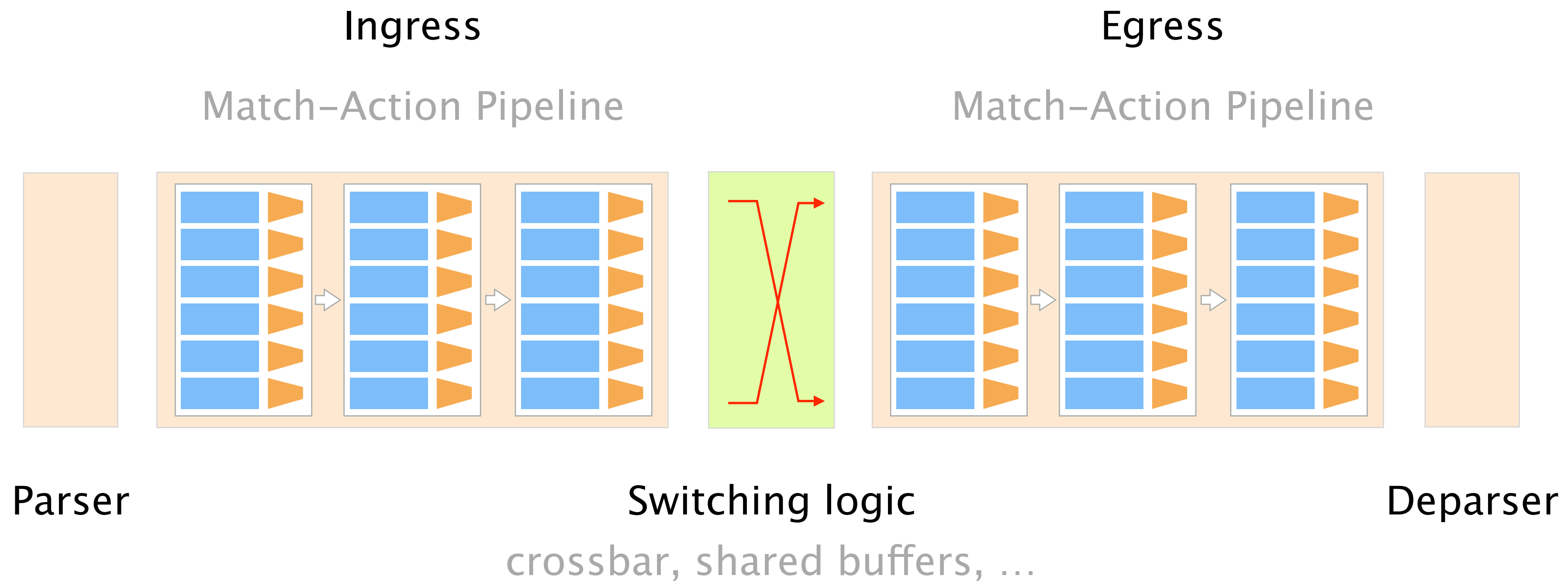


Netronome Agilio CX

A snapshot of our networking lab @ETHZ



Reprogrammable network hardware allows
to completely redefine the forwarding logic



Programmable networks can be made "self-driving"

Programmable devices can

- measure
- perform statistical inference
- adapt their forwarding decisions

at line rate, on a per-packet basis

Enters...

Self-Driving Network

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A network that is capable of
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Questions

How do we build and **deploy** such networks?

Levels of autonomy in self-driving cars

level 0 no automation

1 driver assistance

2 partial automation

3 conditional automation

4 high automation

5 full automation

human monitors
the environment

system monitors
human as fallback

no more human

Self-driving/monitoring networks

in the age of deep network programmability

1 operator assistance

2 partial automation

3 conditional automation

4 high automation

5 full automation

Part I

assisting operators

Part II

taking control

too futuristic? 😊

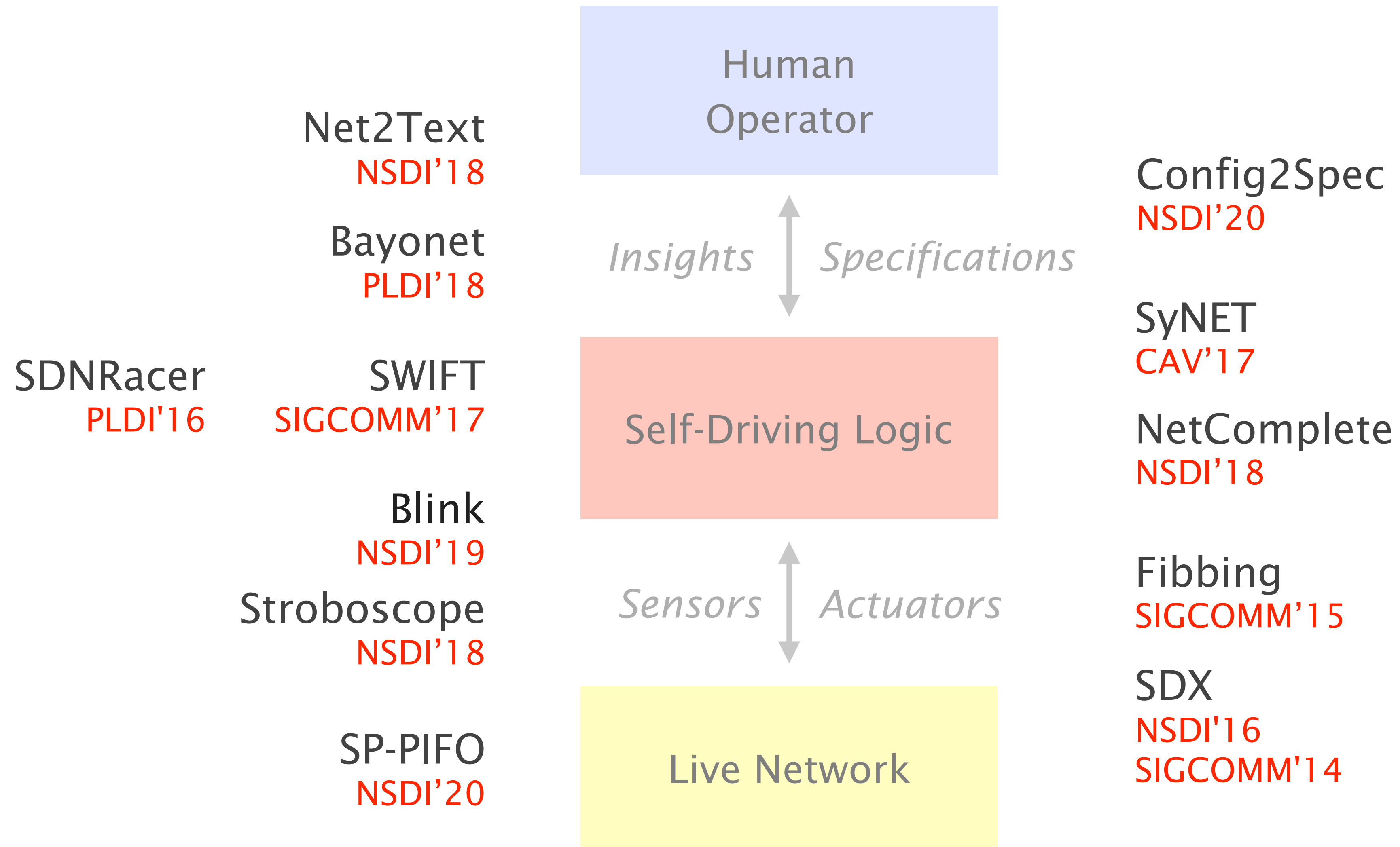
Human
Operator

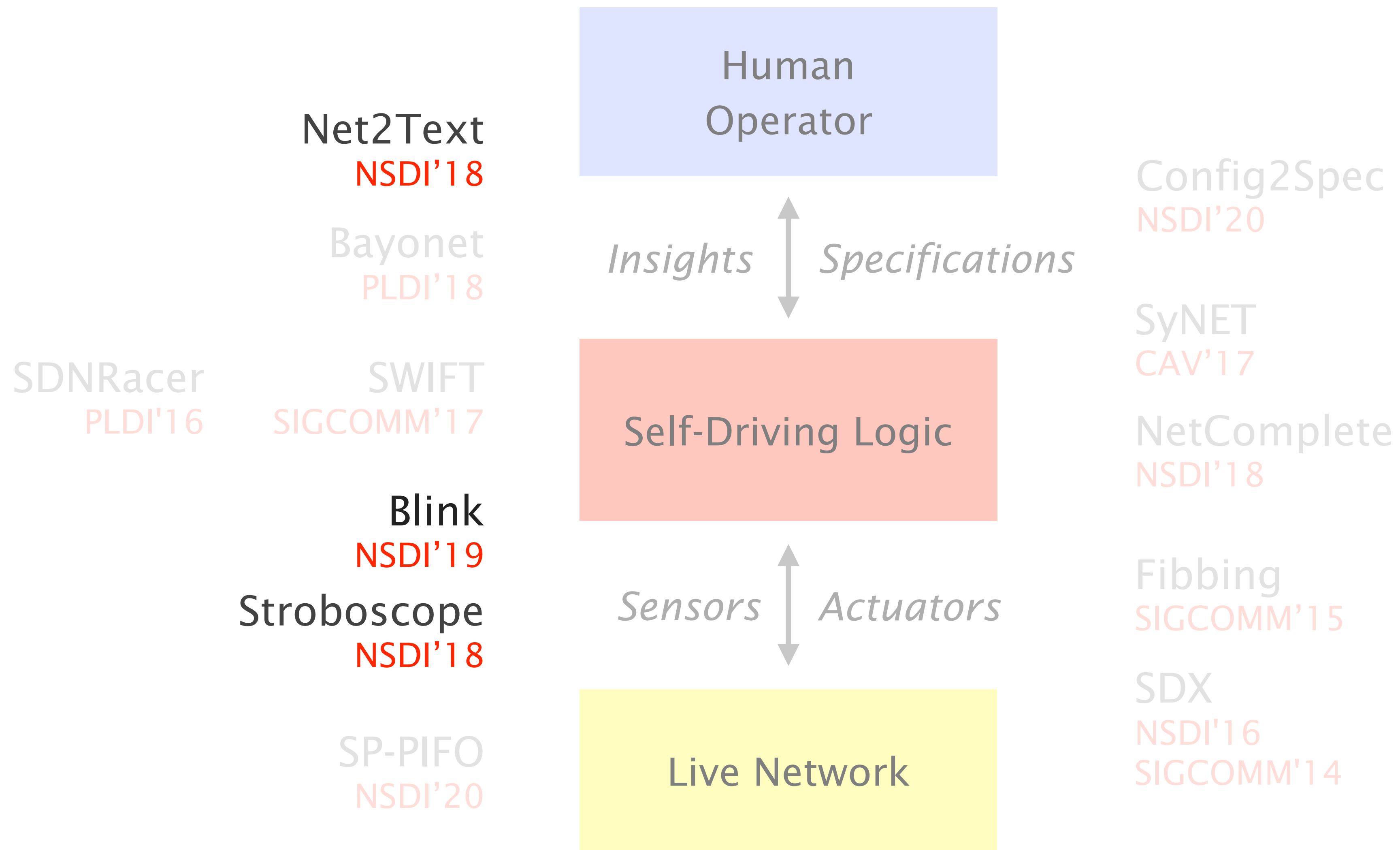
Insights ↔ *Specifications*

Self-Driving Logic

Sensors ↔ *Actuators*

Live Network





Self-driving/monitoring networks

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Stroboscope: Declarative Network Monitoring on a Budget



Olivier Tilmans



Tobias Bühler



Ingmar Poesse



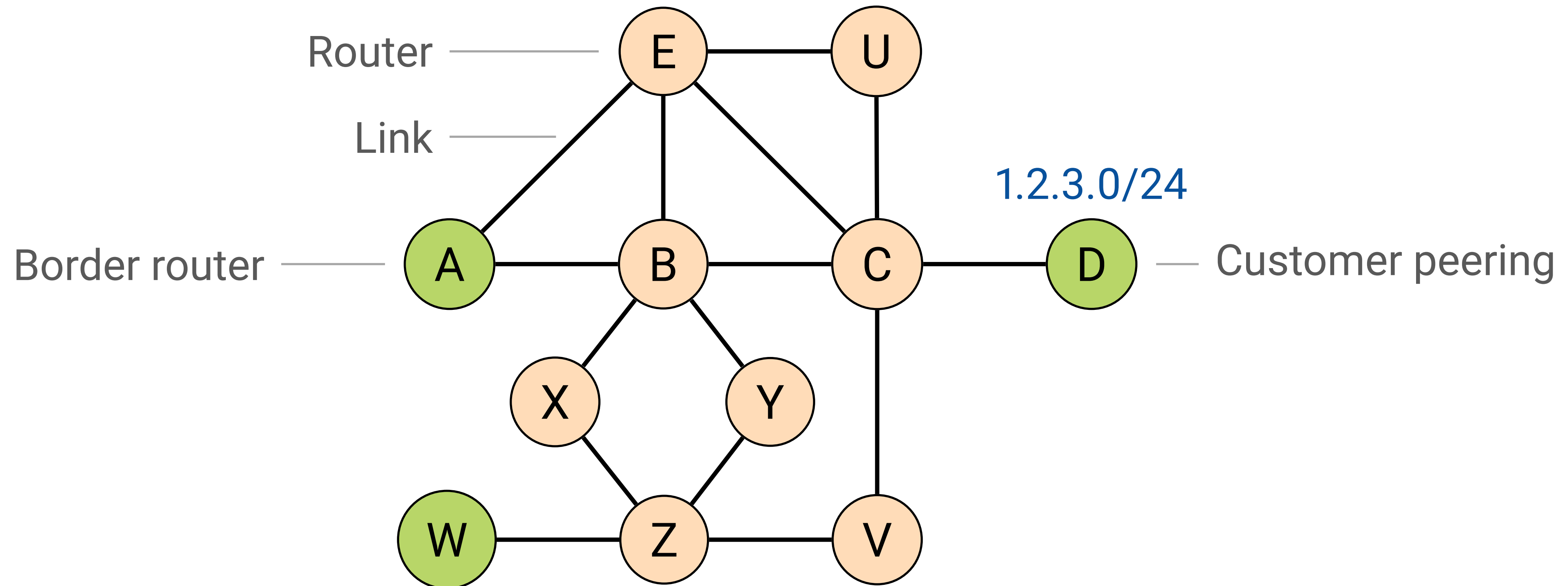
Stefano Vissicchio



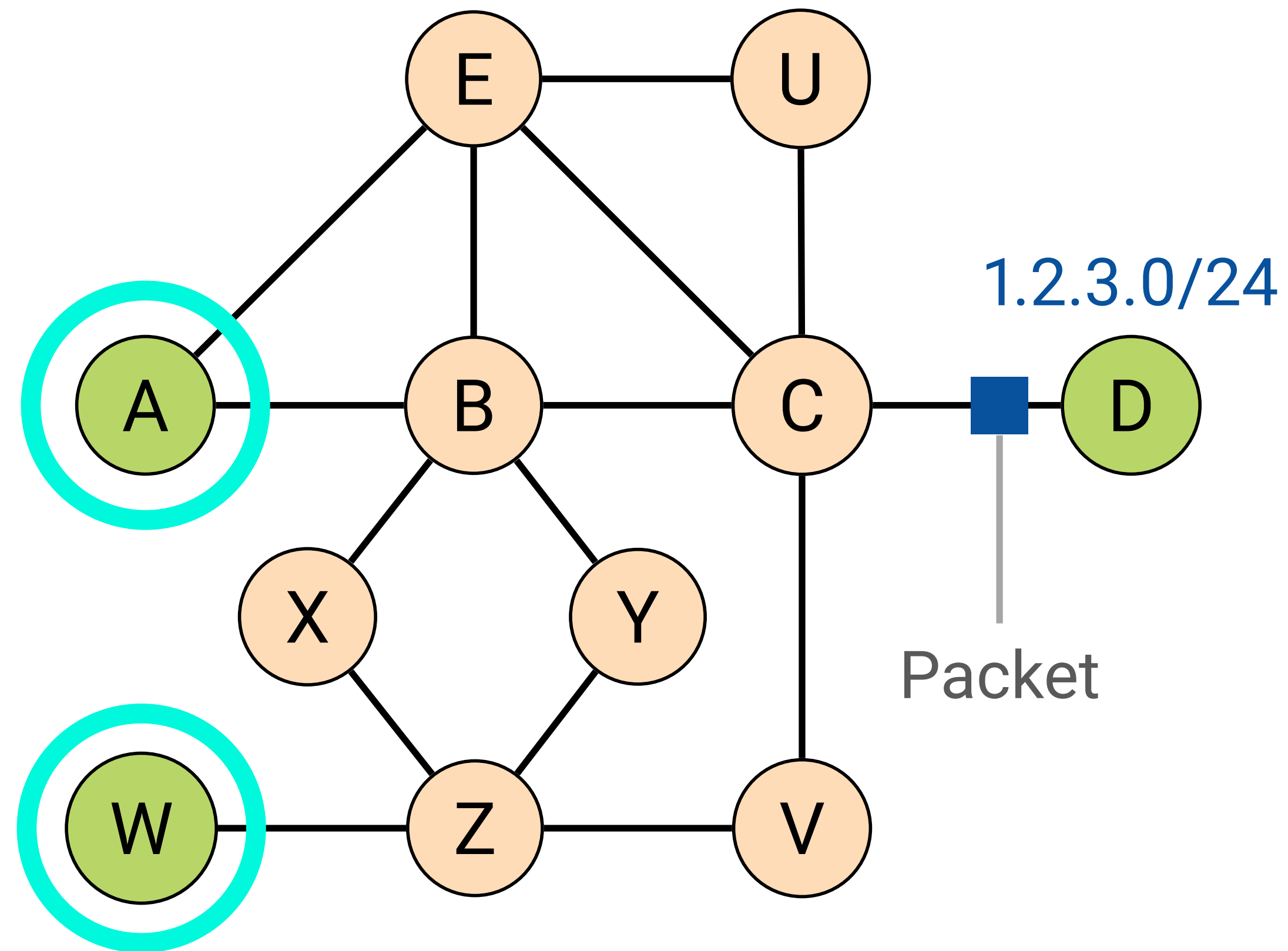
Laurent Vanbever

USENIX Symposium on Networked Systems Design and Implementation. April 2018.

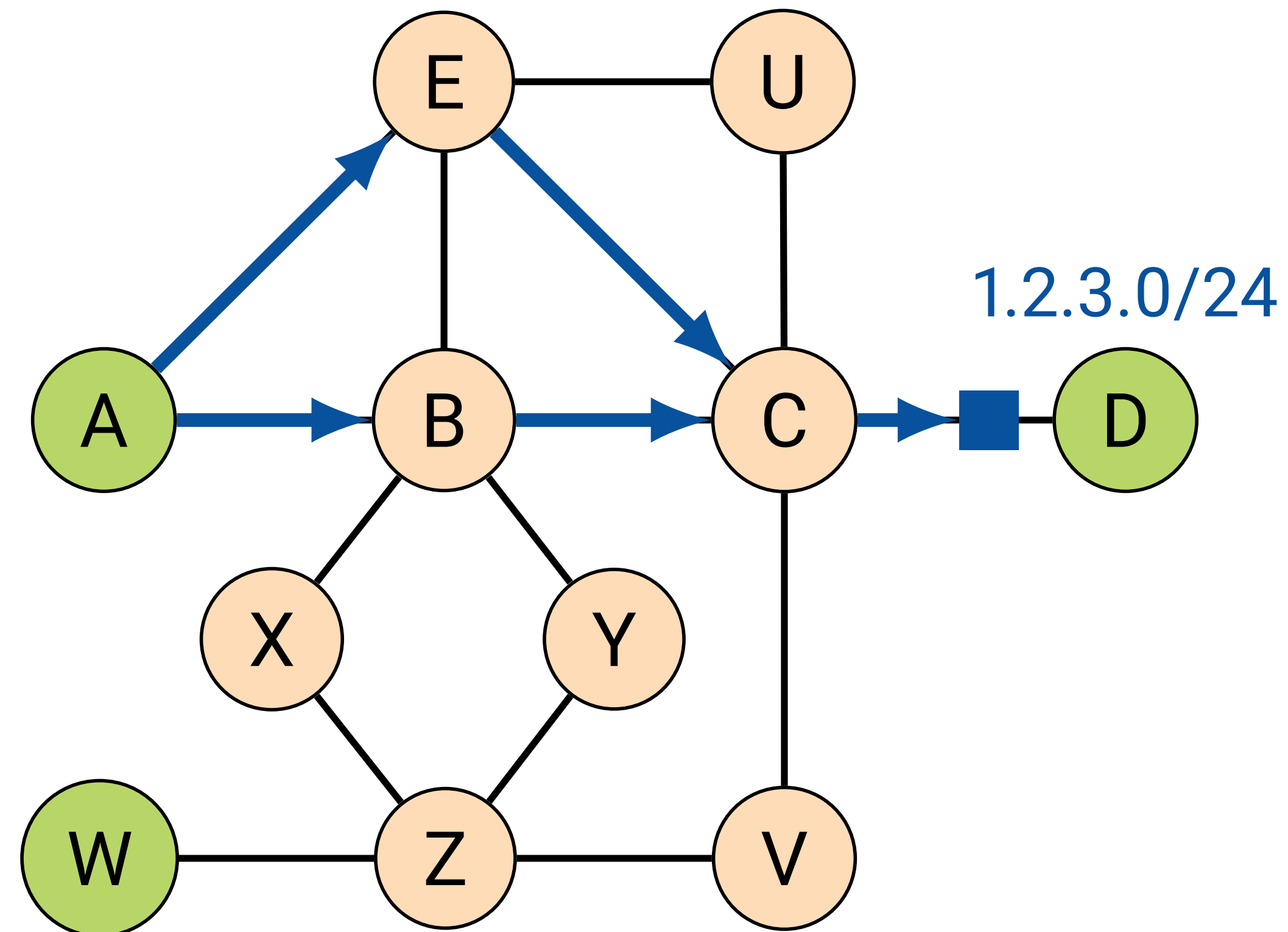
Consider this example ISP network topology



What is the ingress router for this packet arriving at router D?

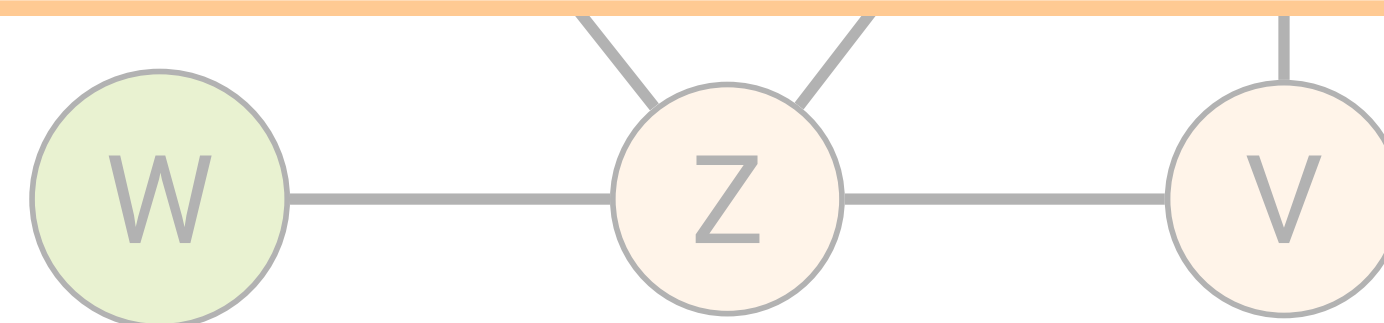


Which paths does the traffic follow?

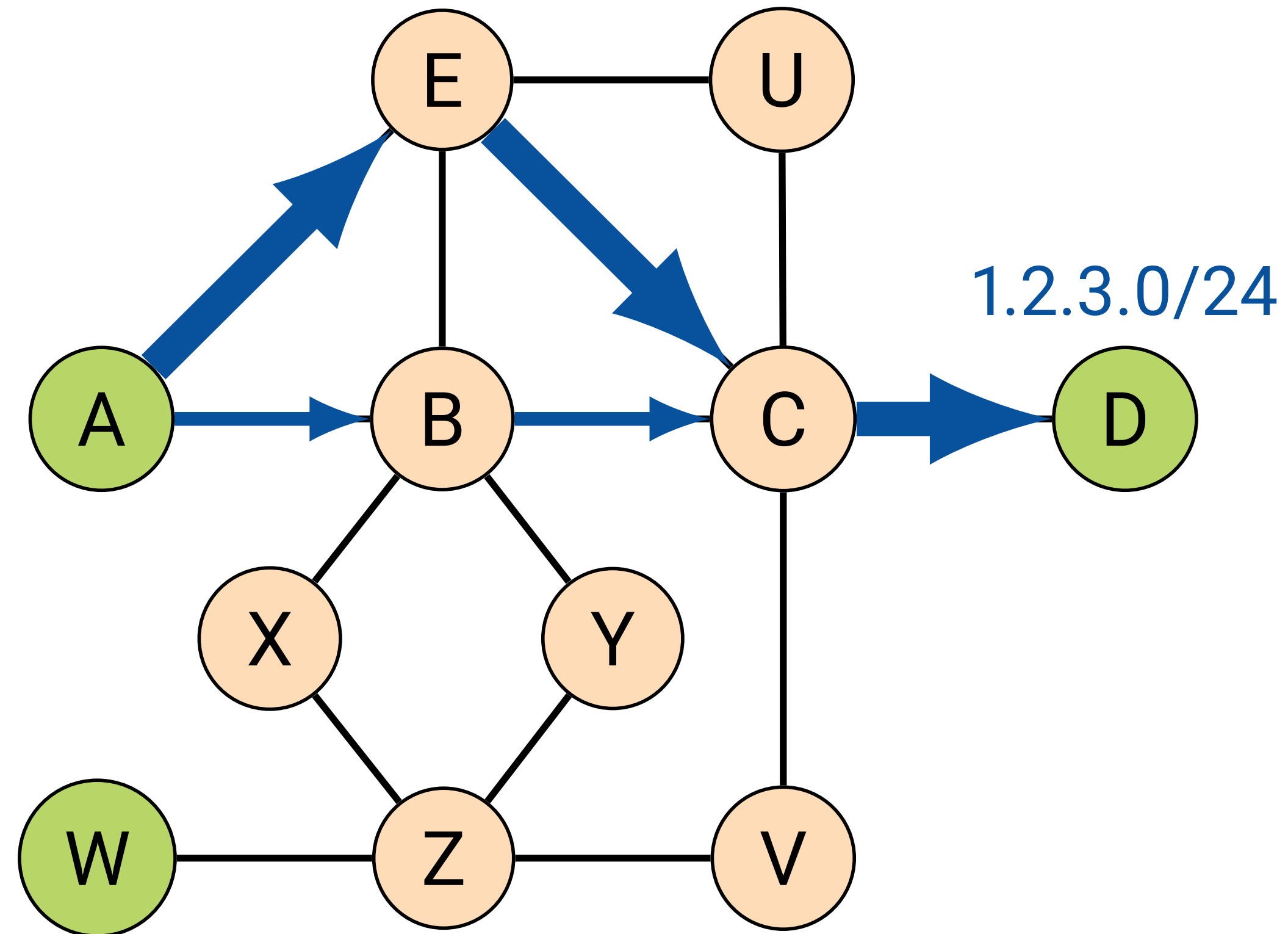


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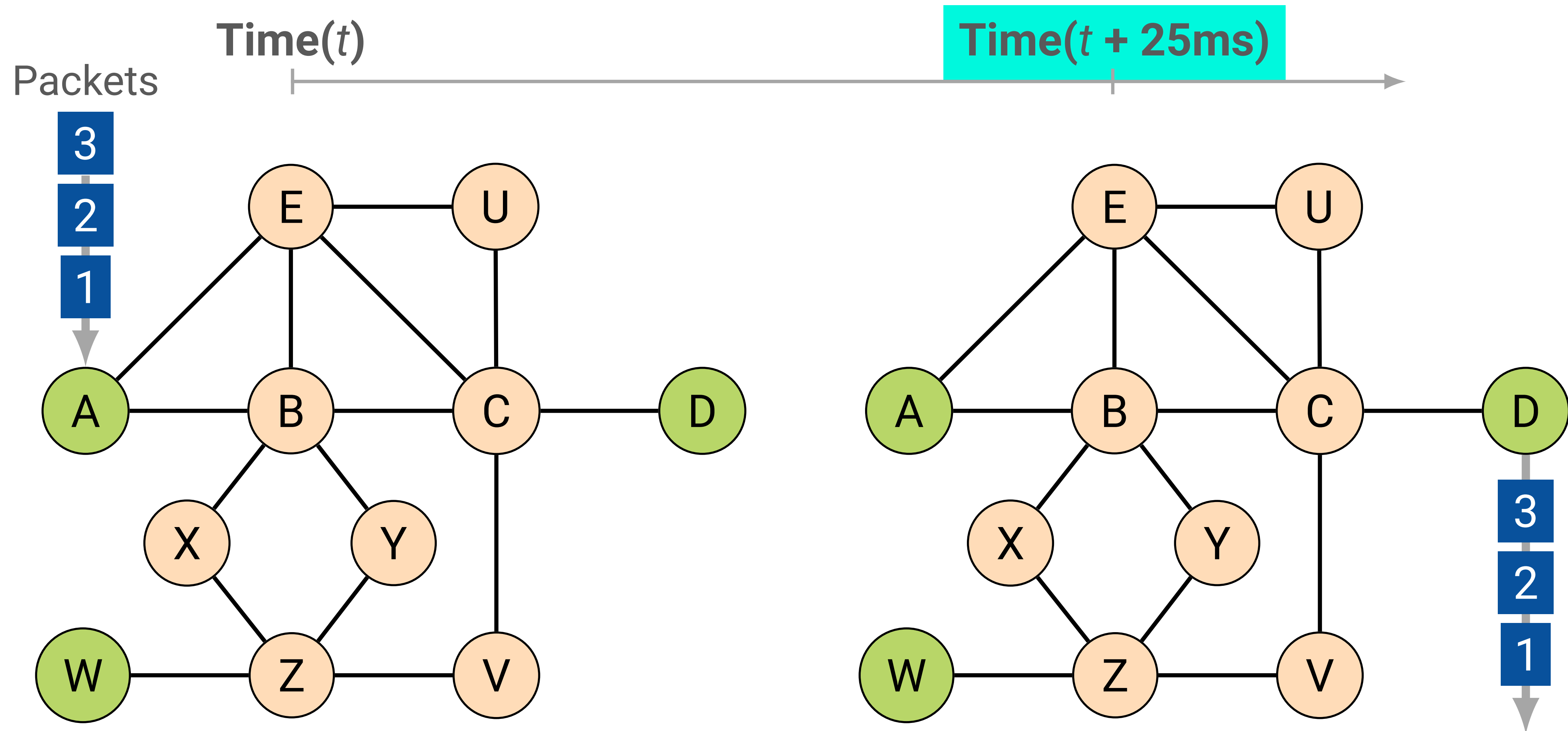
Tracking flows network-wide requires to
match measurements across multiple vantage points



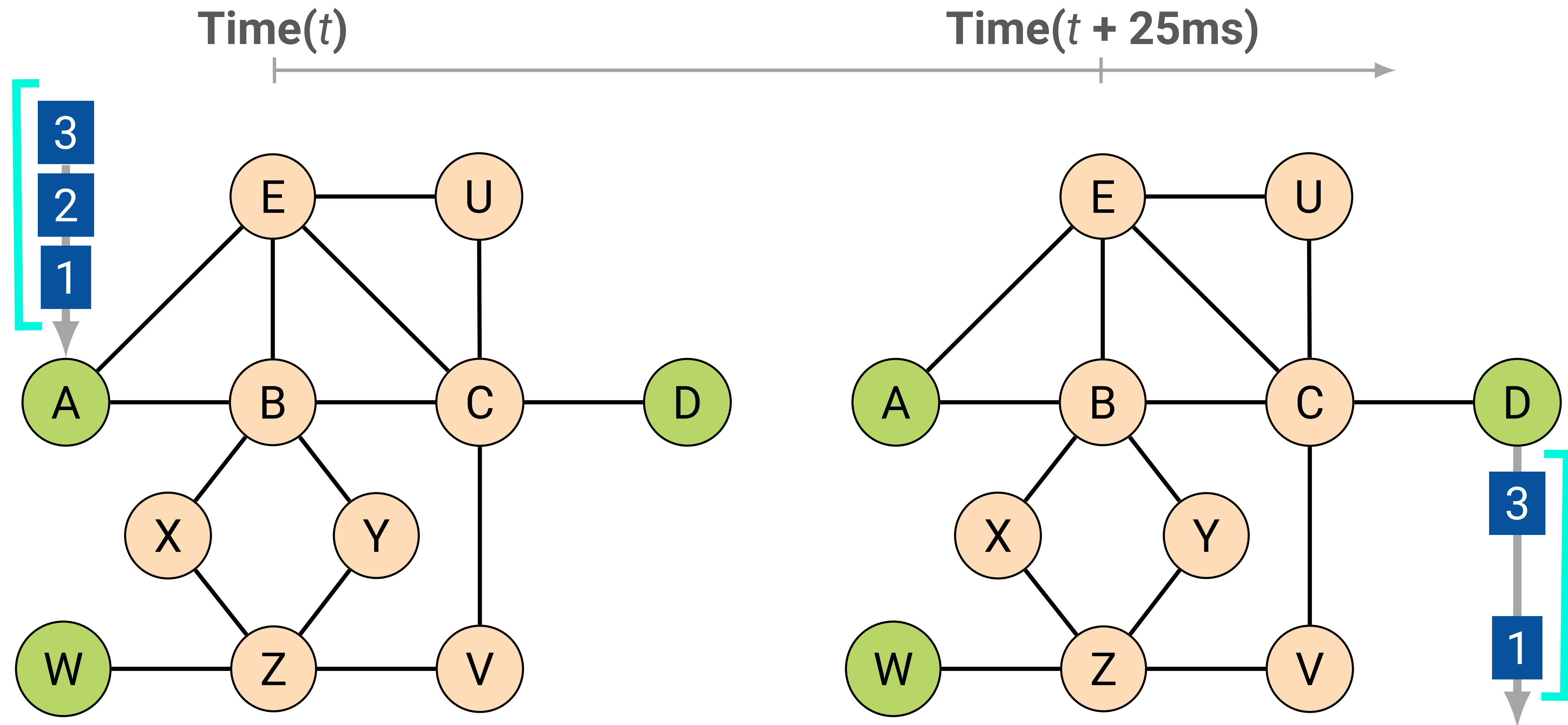
Is traffic load-balanced as expected?



Is the latency acceptable?



Are there losses?

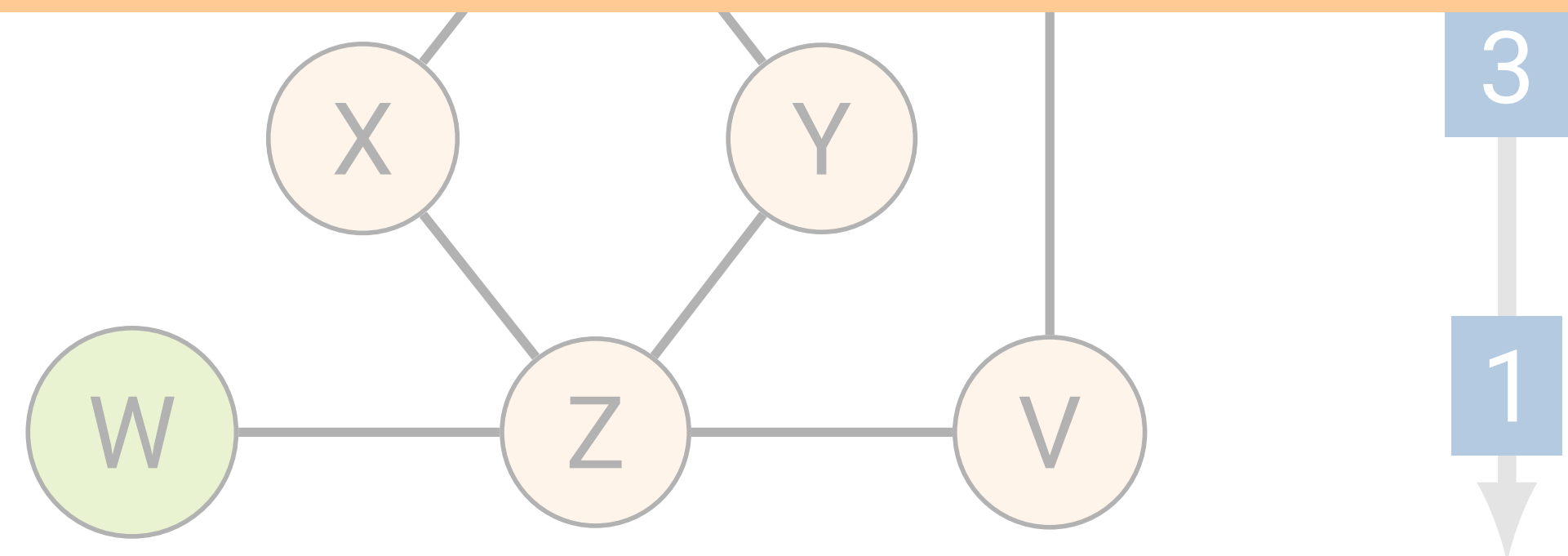
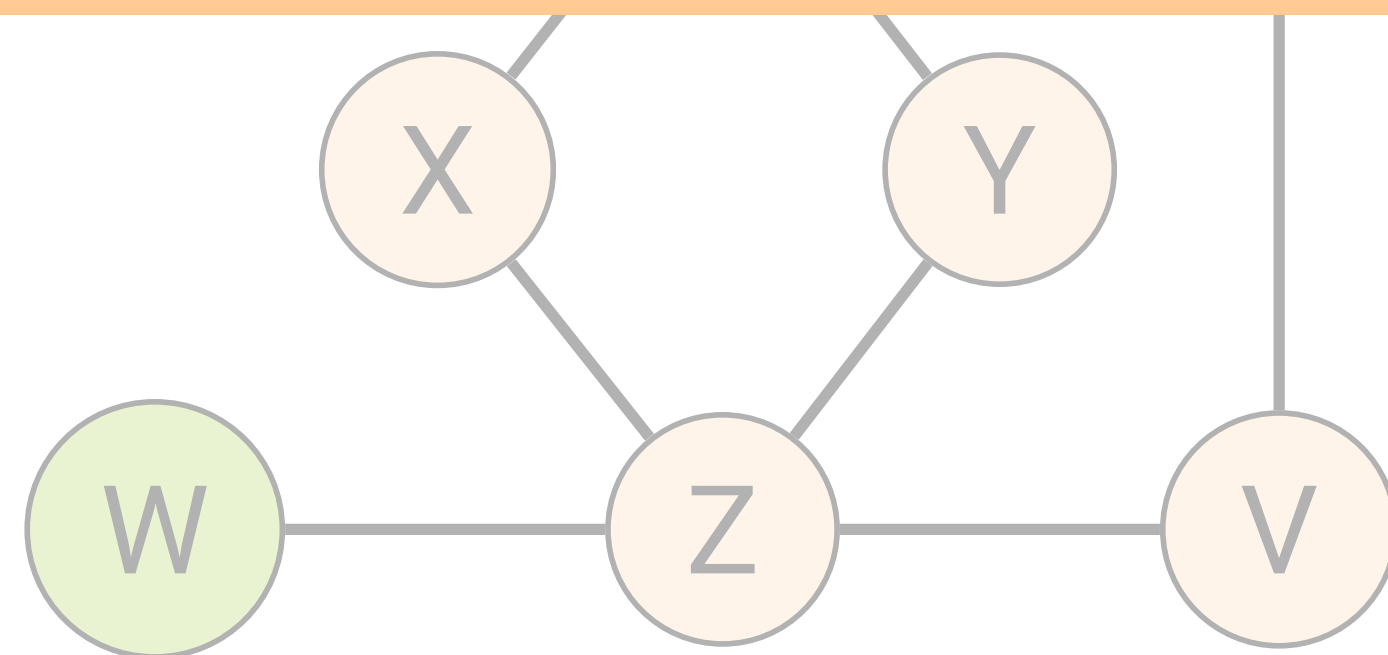


Are there losses?

Time(t)

Time($t + 25\text{ms}$)

Fine-grained data-plane performance metrics require
packet-level visibility over individual flows



Fined-grained network monitoring is widely researched

Gigascop3e [SIGMOD'03]

Planck [SIGCOMM'14]

Everflow [SIGCOMM'15]

Compiling Path Queries [NSDI'16]

Trumpet [SIGCOMM'16]

Marple [SIGCOMM'17]

Fined-grained **ISP** network monitoring poses unique and unmet challenges

- No control over end hosts

Gigascop3e [SIGMOD'03]

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Fined-grained **ISP** network monitoring poses unique and unmet challenges

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Fined-grained **ISP** network monitoring poses unique and unmet challenges

- No control over end hosts
- Limited data-plane flexibility
- Limited monitoring bandwidth

~~Gigascope [SIGMOD'03]~~

~~Planck [SIGCOMM'14]~~

~~Everflow [SIGCOMM'15]~~

~~Compiling Path Queries [NSDI'16]~~

~~Trumpet [SIGCOMM'16]~~

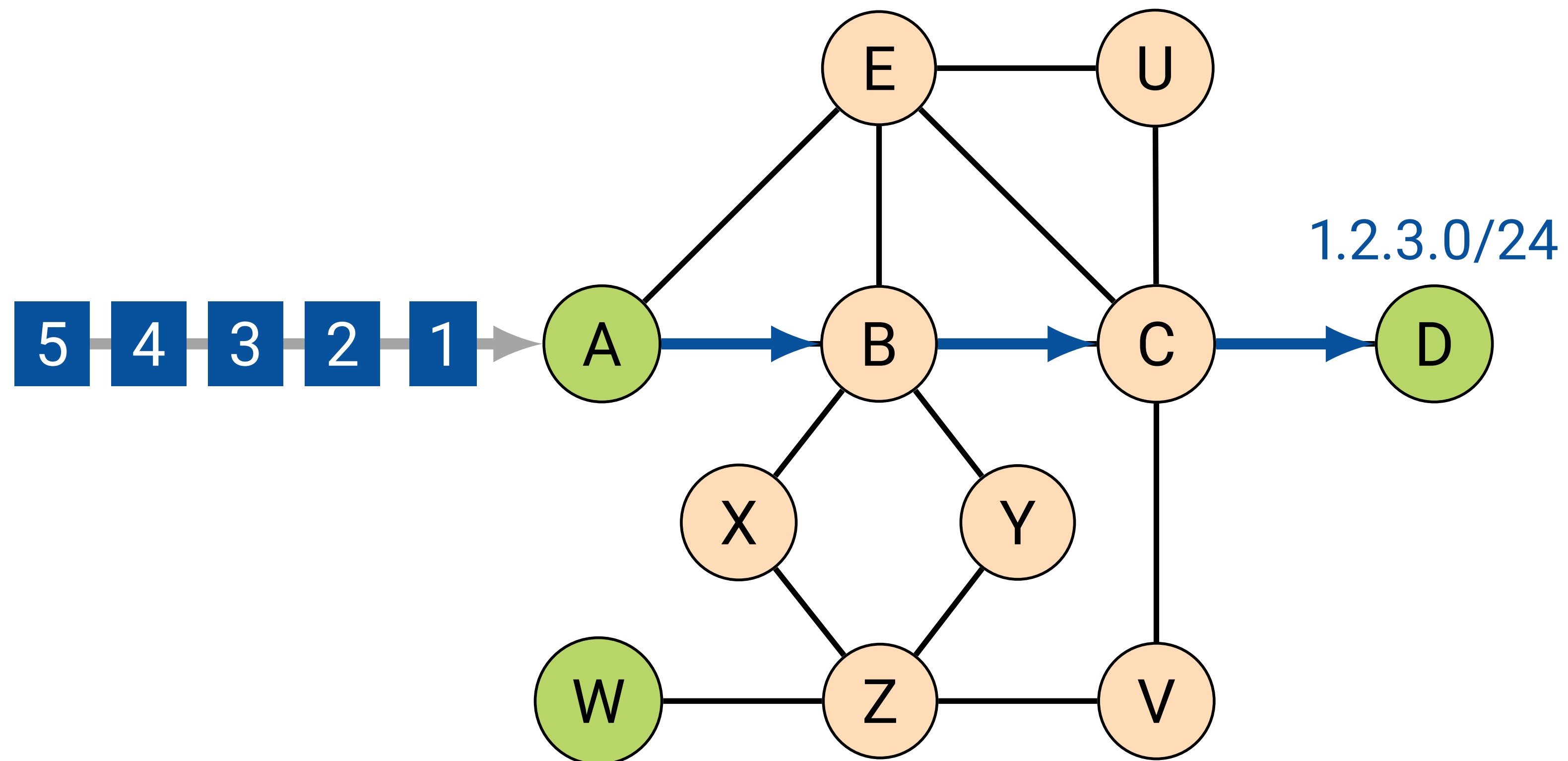
~~Marple [SIGCOMM'17]~~

Stroboscope: Declarative Network Monitoring on a Budget

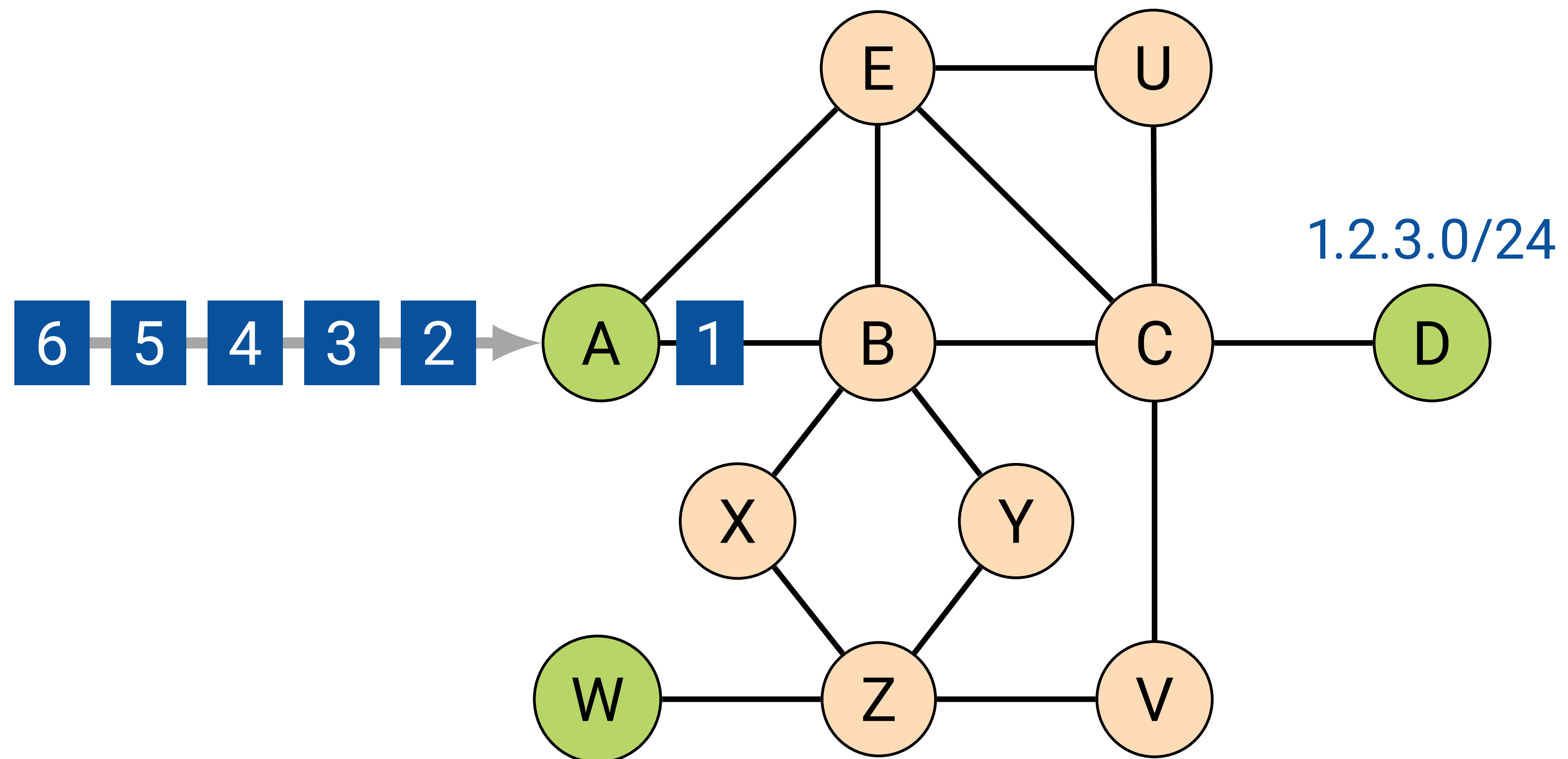


- Collecting traffic slices to monitor networks
- Adhering to a monitoring budget
- Using Stroboscope today

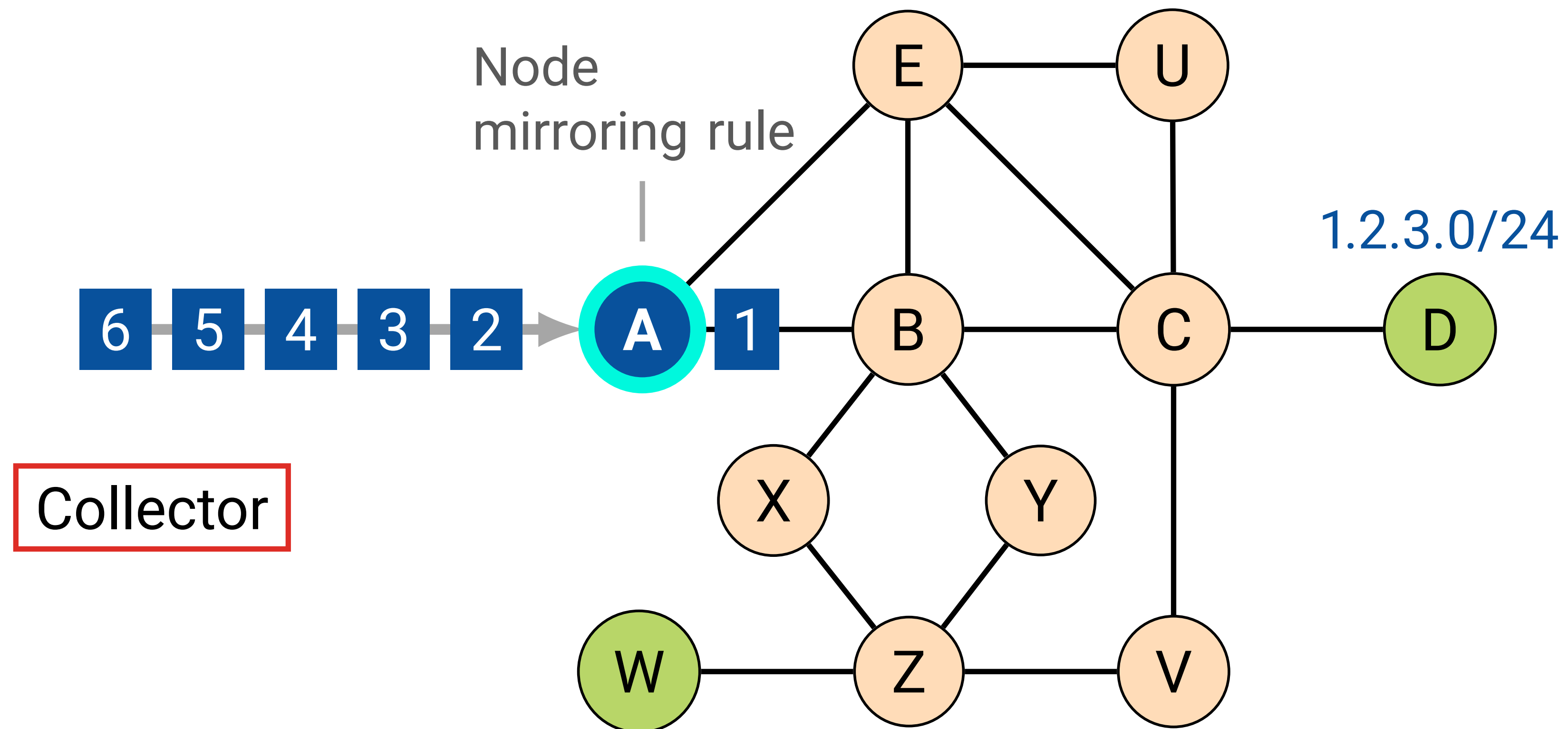
Consider the following flow of packets



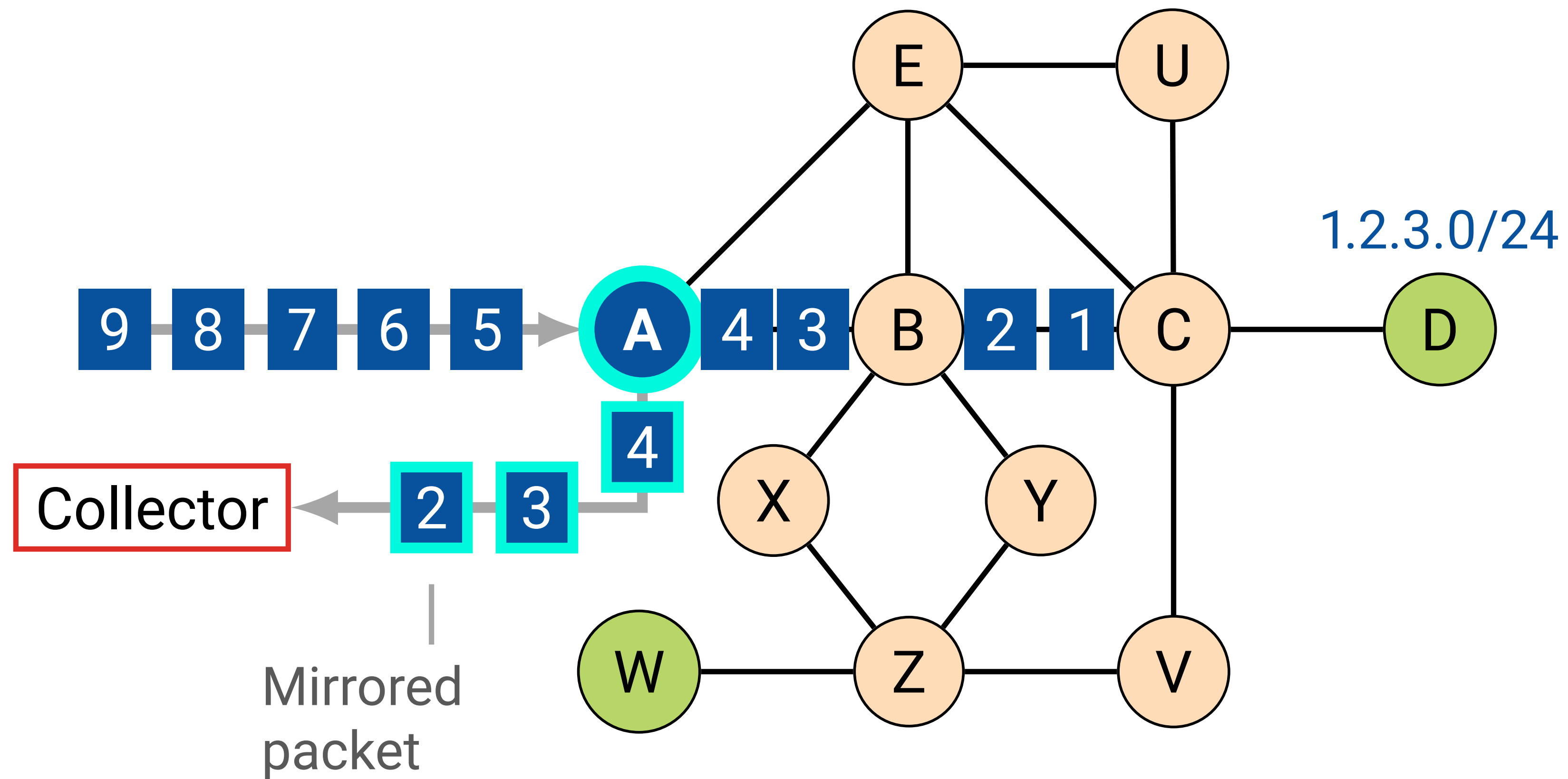
Consider the following flow of packets



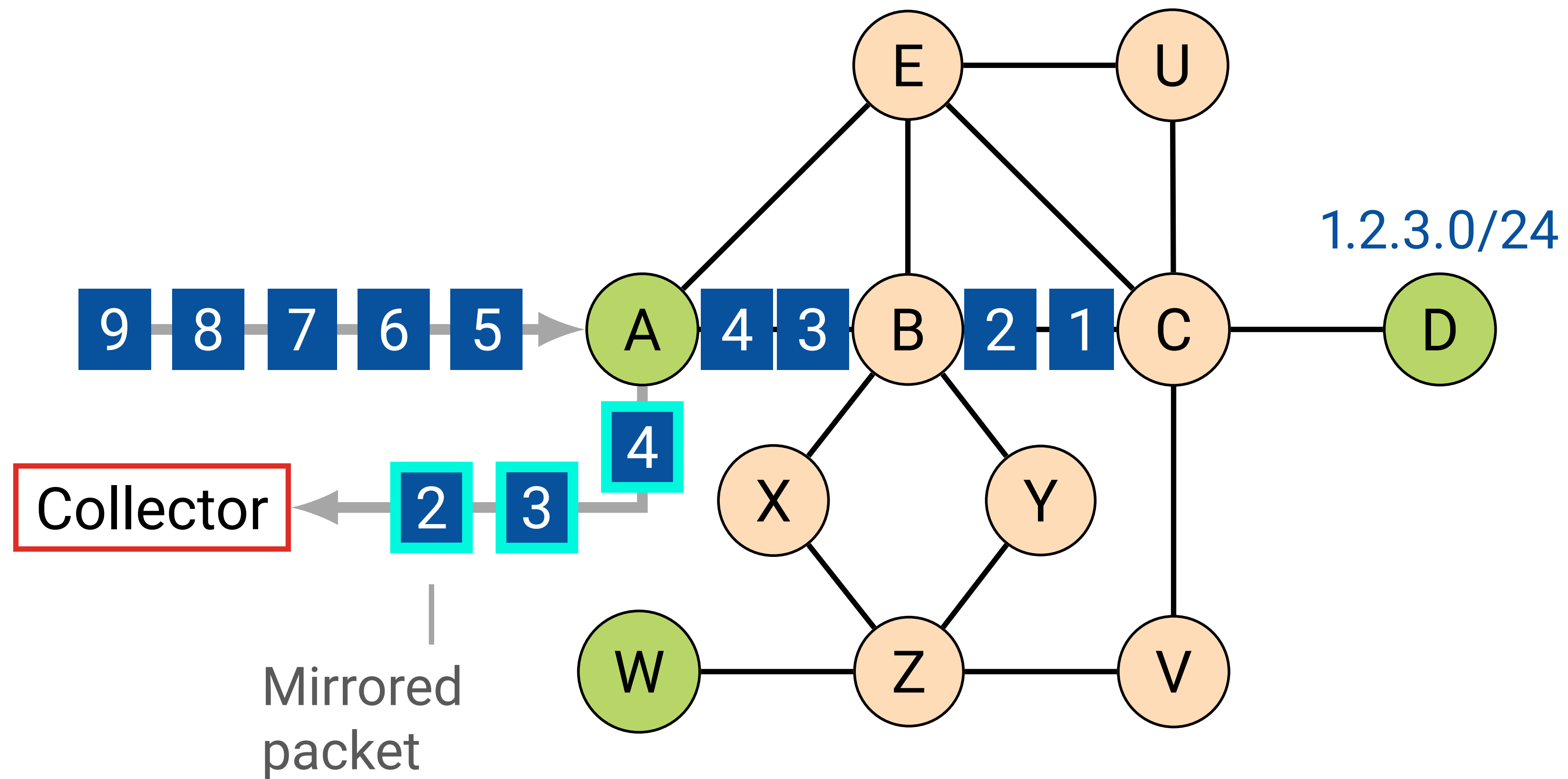
Stroboscope activates mirroring for the flow



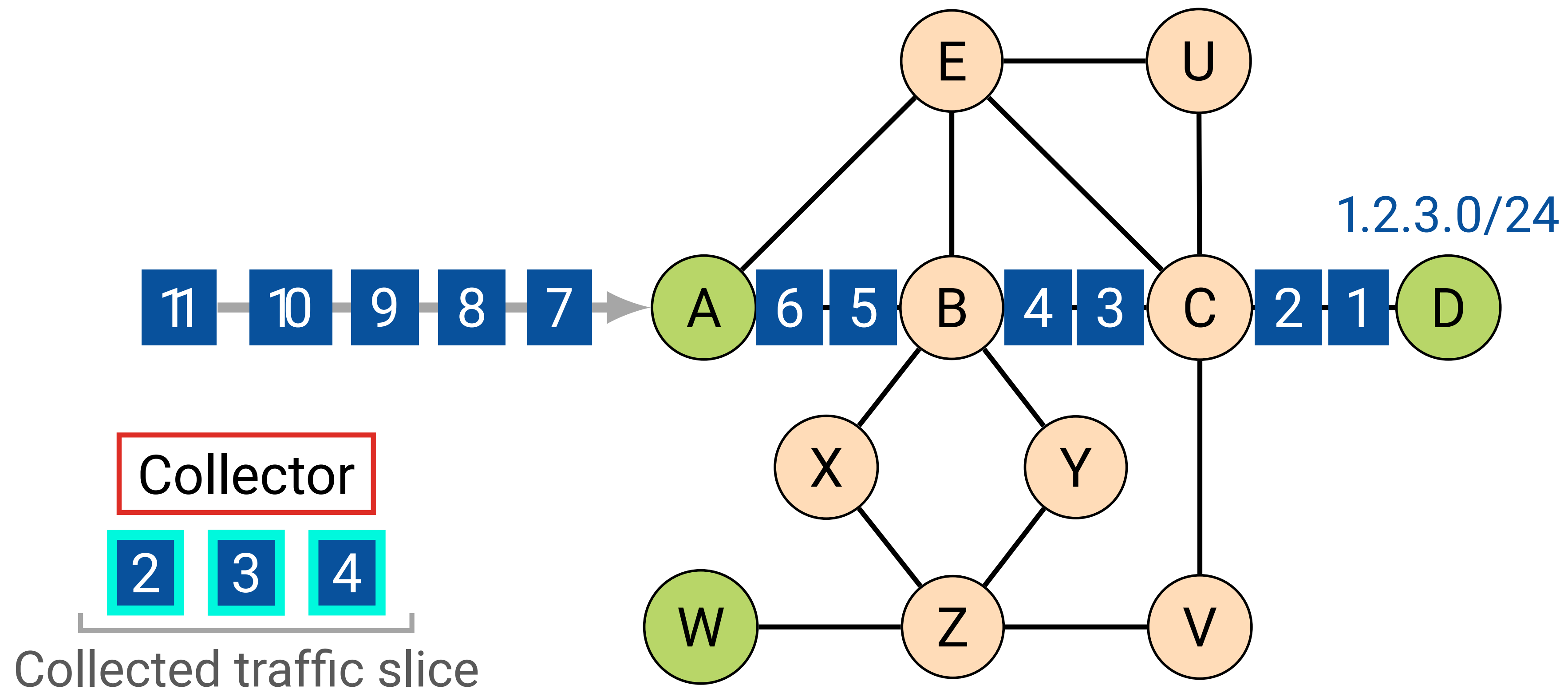
Packets are copied and encapsulated towards the collector



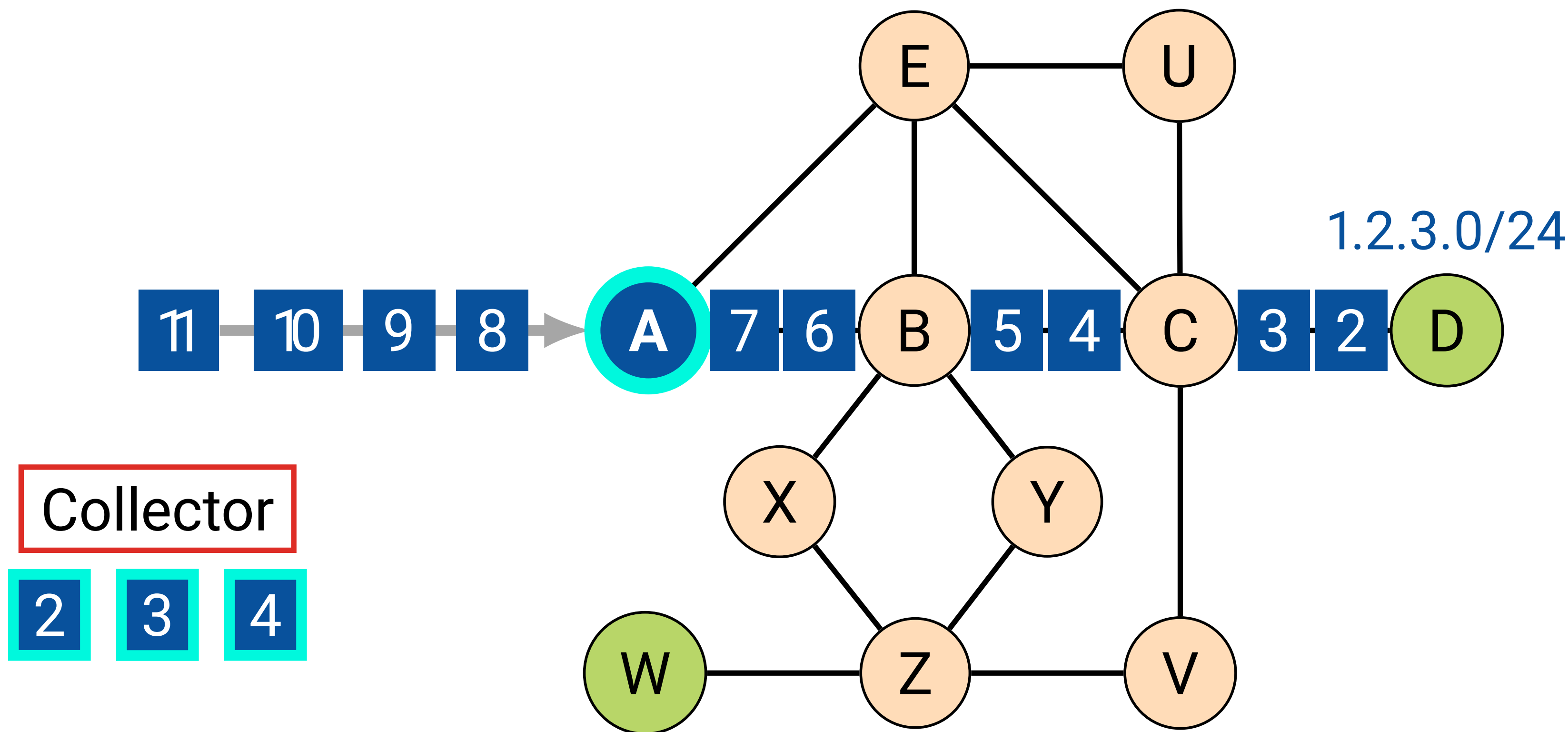
The mirroring rule is deactivated after a preset delay



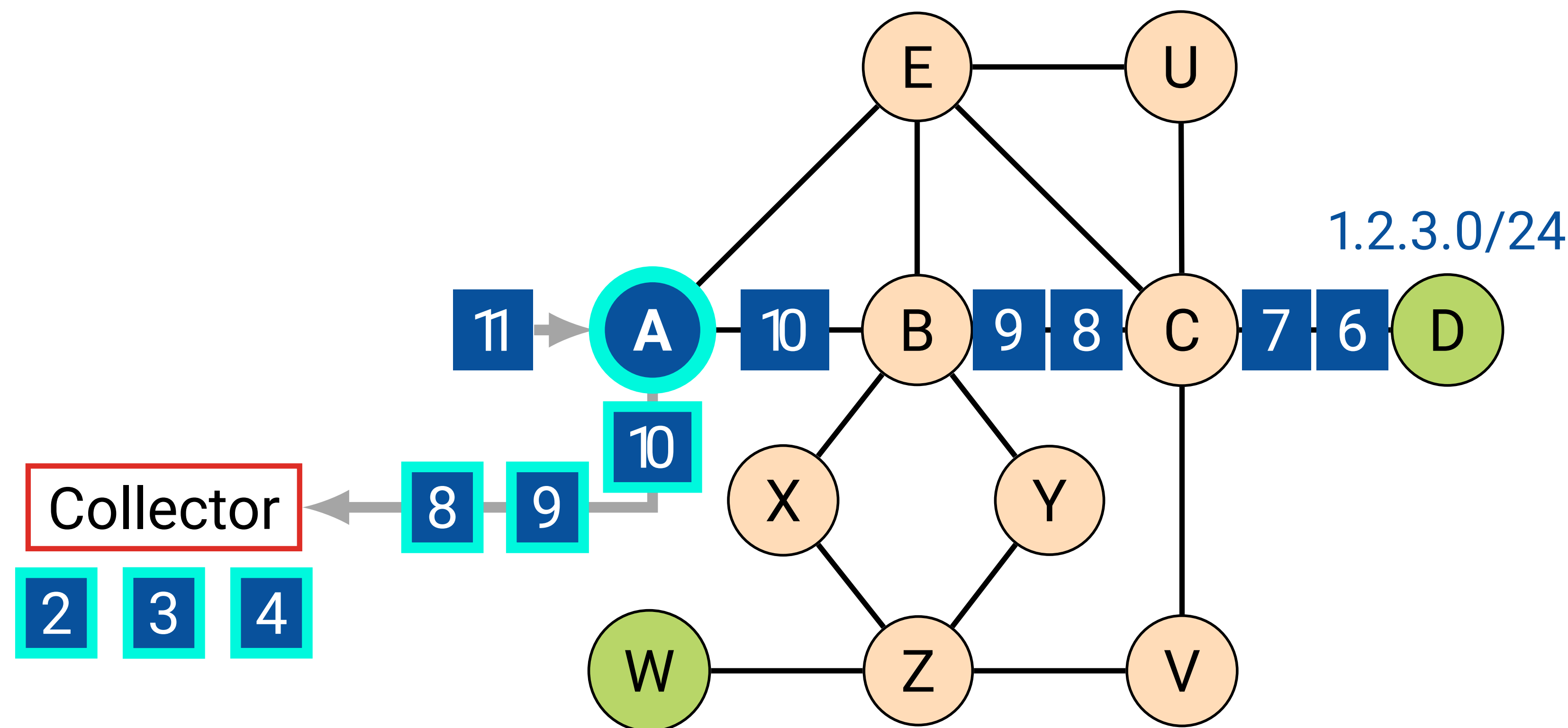
Stroboscope stores the traffic slice for analysis



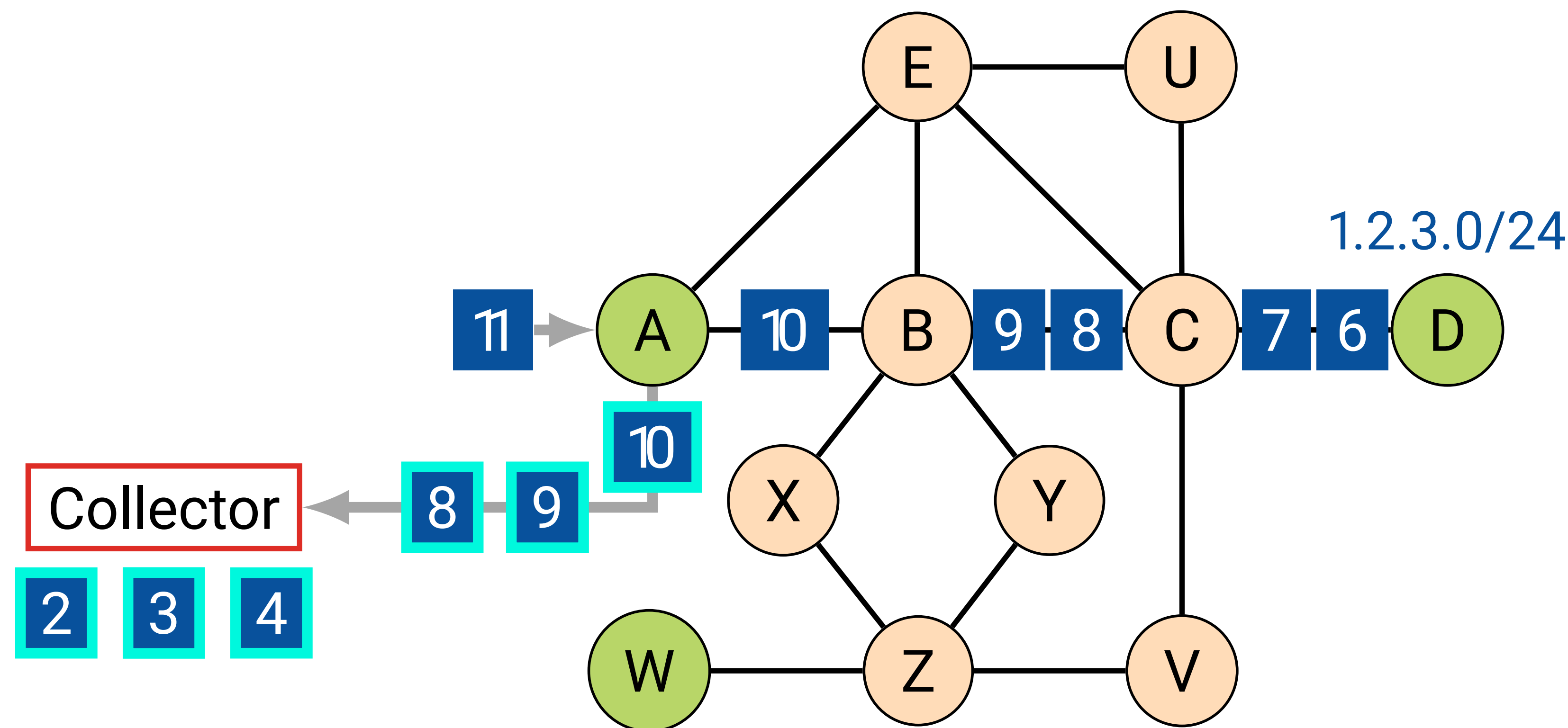
Stroboscope periodically toggles the mirroring rule



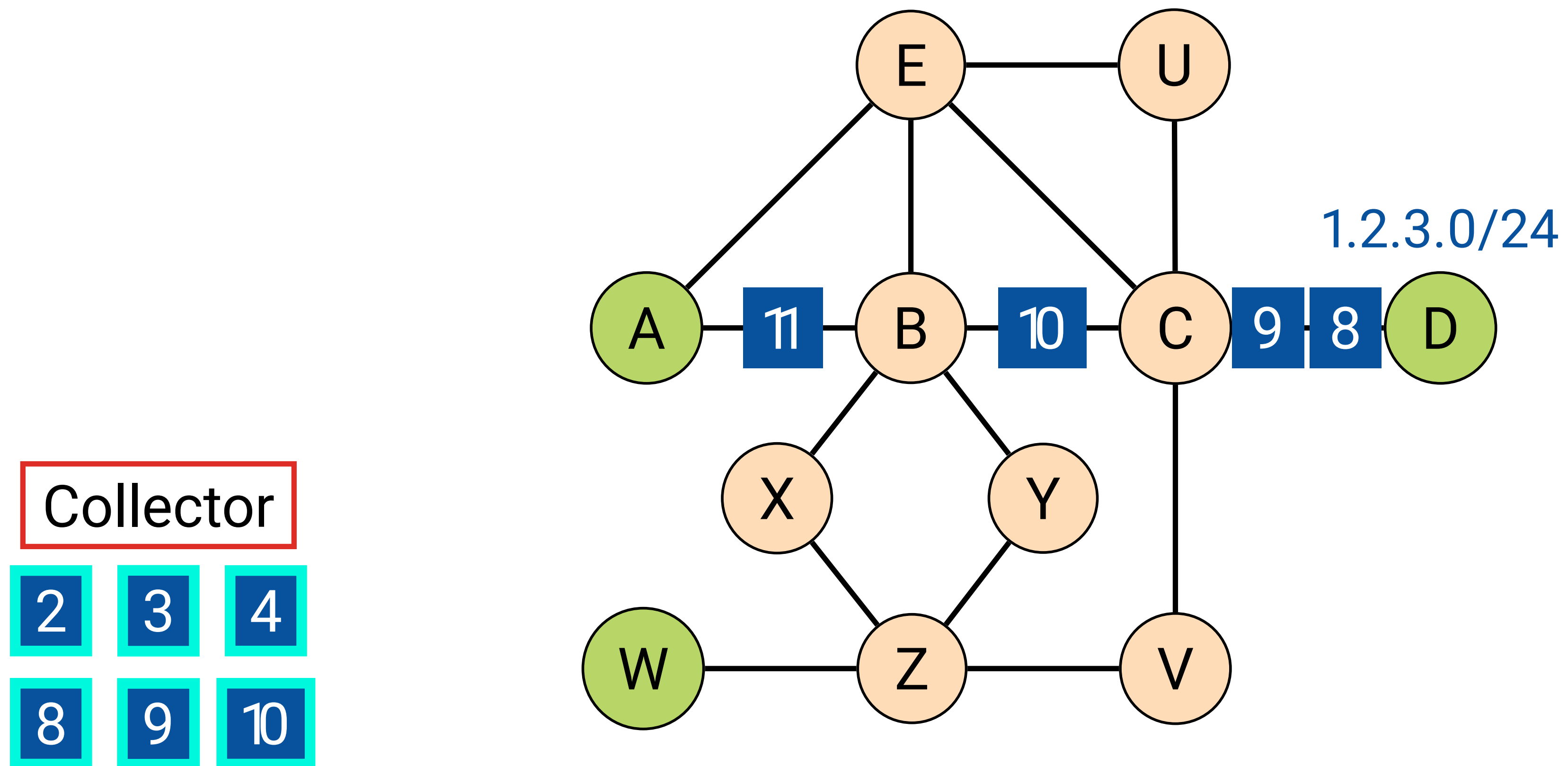
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Stroboscope periodically toggles the mirroring rule



Stroboscope collects multiples traffic slices over time



Analyzing matching packets across traffic slices
enables fine-grained measurements at scale

Analyzing matching packets across traffic slices
enables fine-grained measurements at scale

Forwarding paths discovery, timestamp reconstruction, payload inspection, ...

Stroboscope defines a declarative requirement language

MIRROR 1.2.3.0/24 ON [A B C D], [A E C D]

MIRROR 1.2.3.0/24 ON [A -> D]

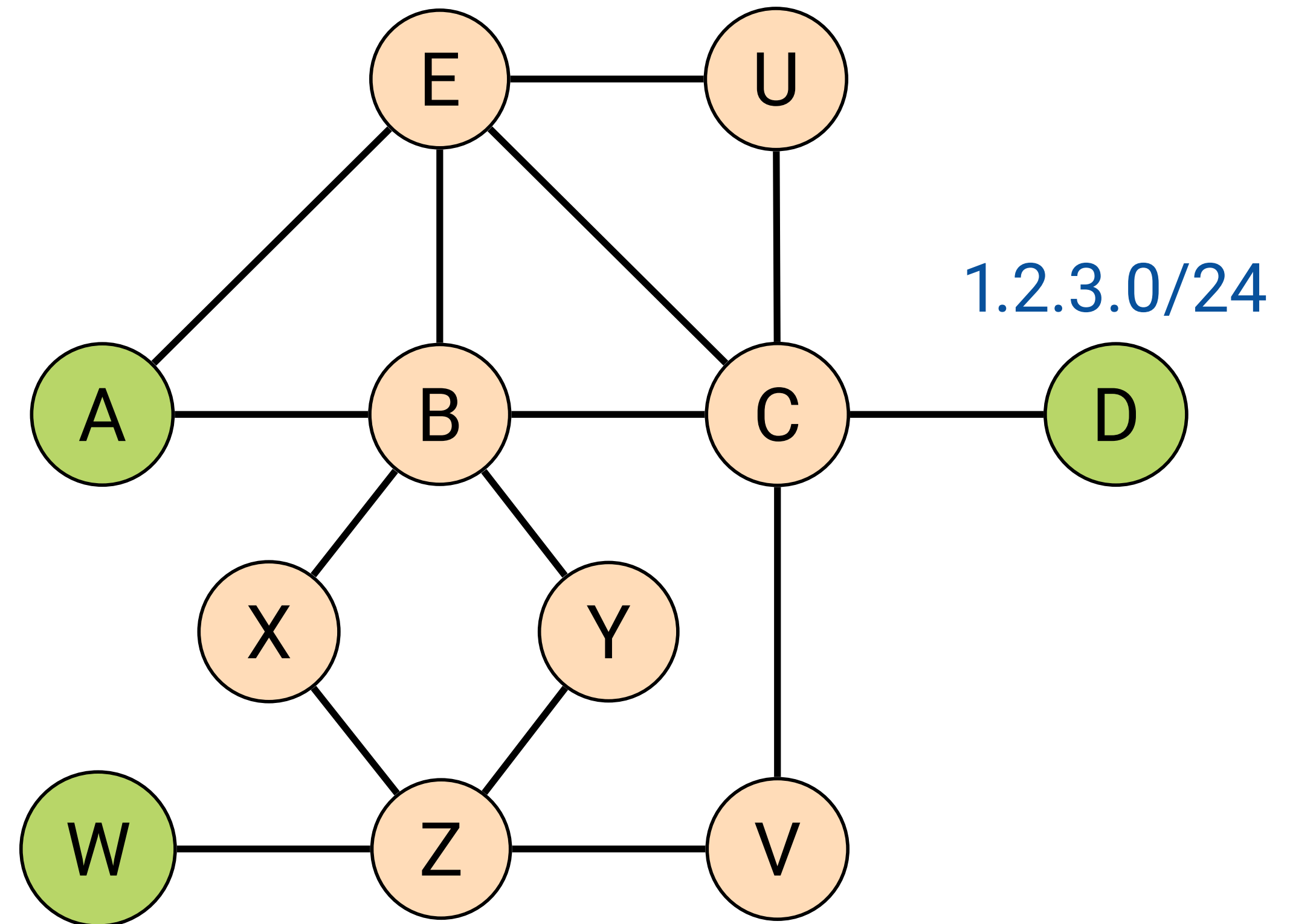
CONFINE 1.2.3.0/24 ON [A B E C D]

CONFINE 1.2.3.0/24 [A -> D]

MIRROR 1.2.3.0/24 ON [-> D]

CONFINE 1.2.3.0/24 ON [-> D]

USING 15 Mbps DURING 500 ms EVERY 5 s



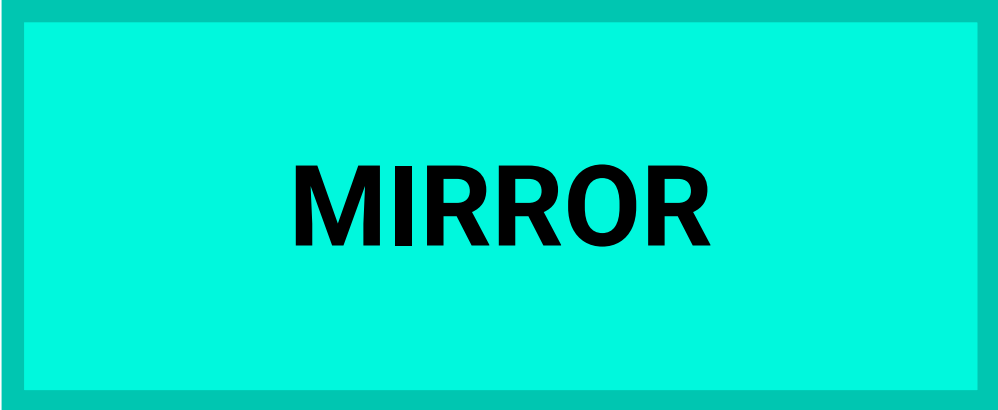
Stroboscope defines two types of queries



MIRROR

CONFINE

Stroboscope defines two types of queries



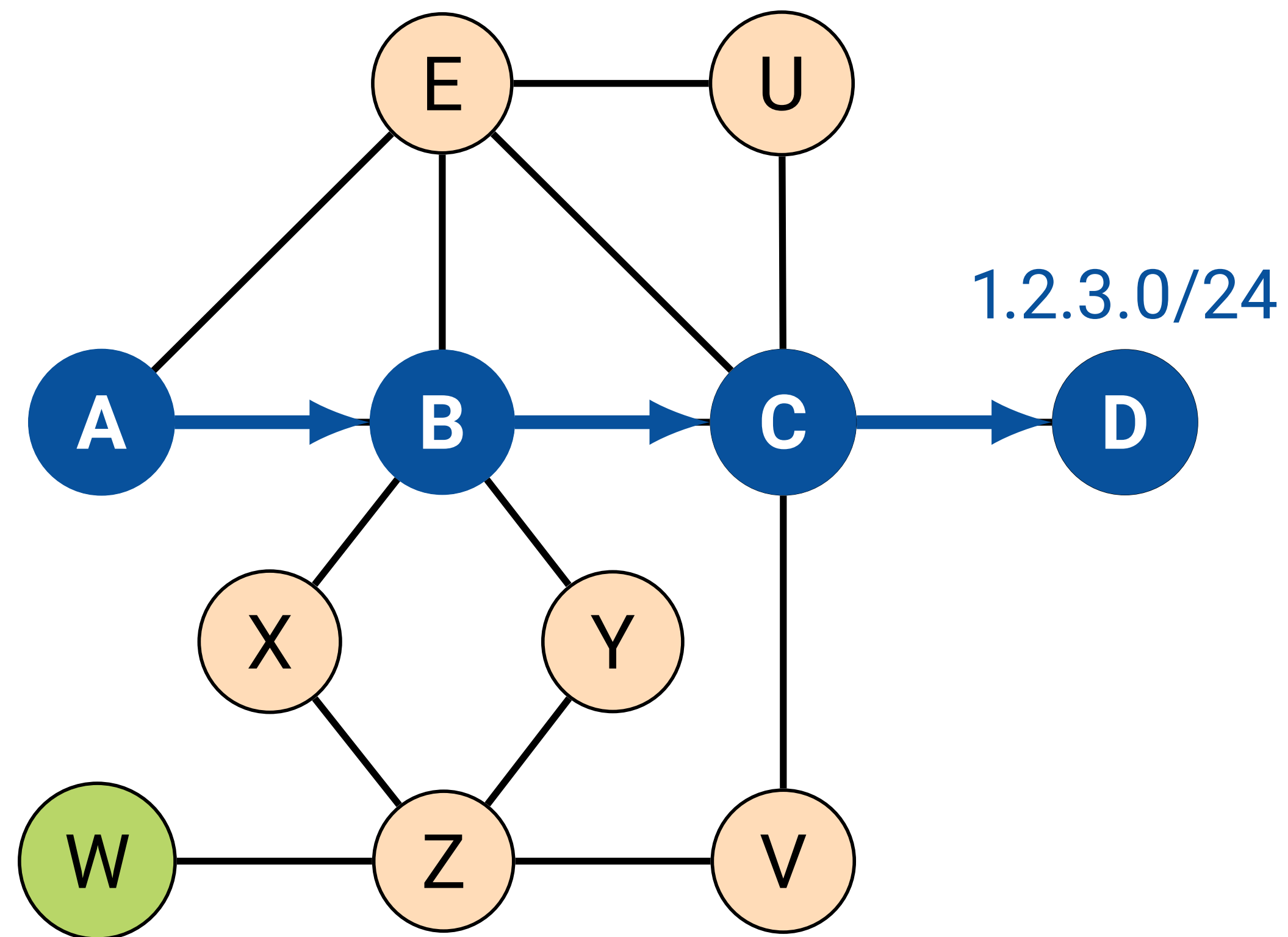
MIRROR



CONFINE

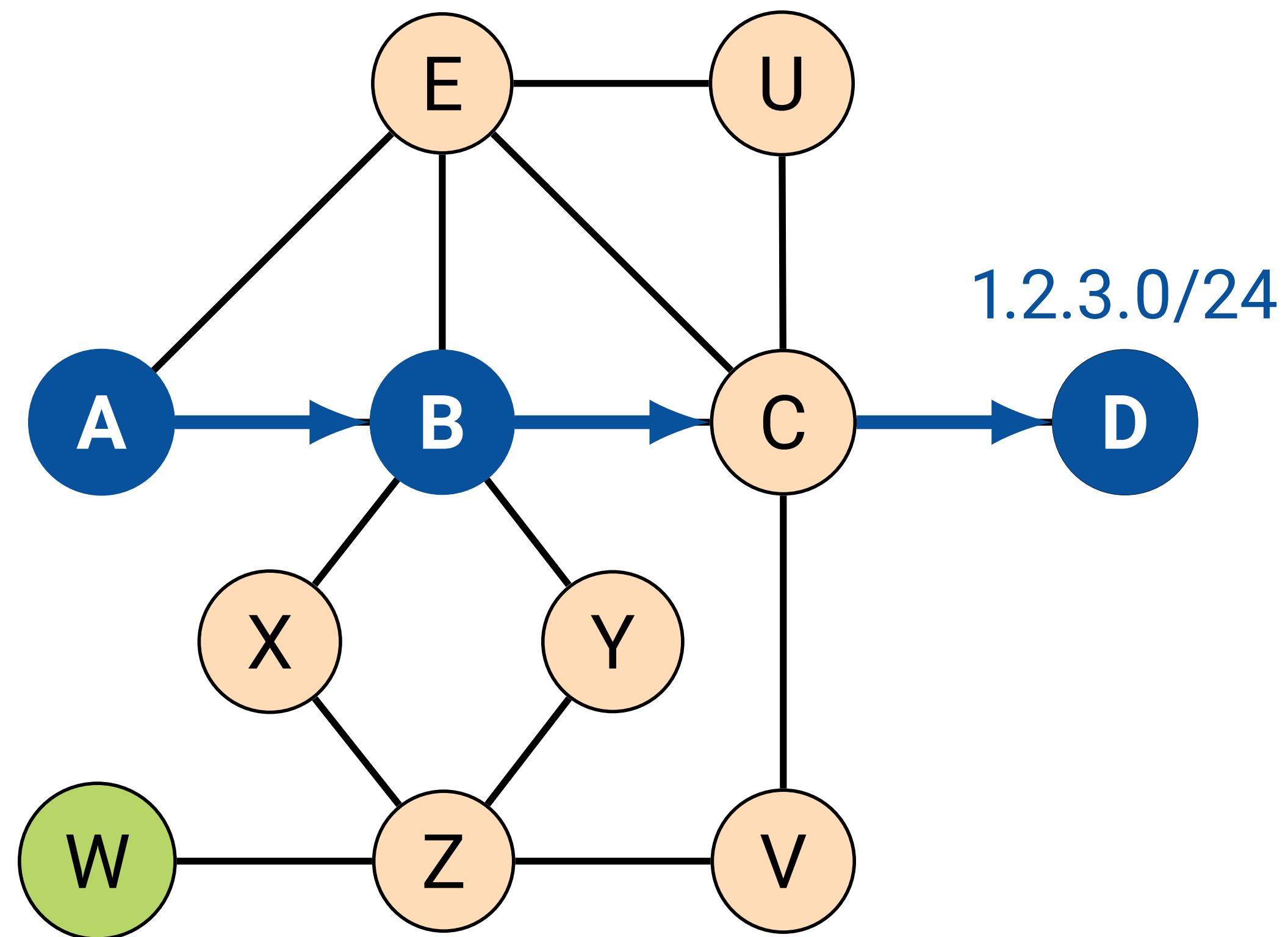
MIRROR queries reconstruct the path taken by packets

MIRROR 1.2.3.0/24 ON [A B C D]



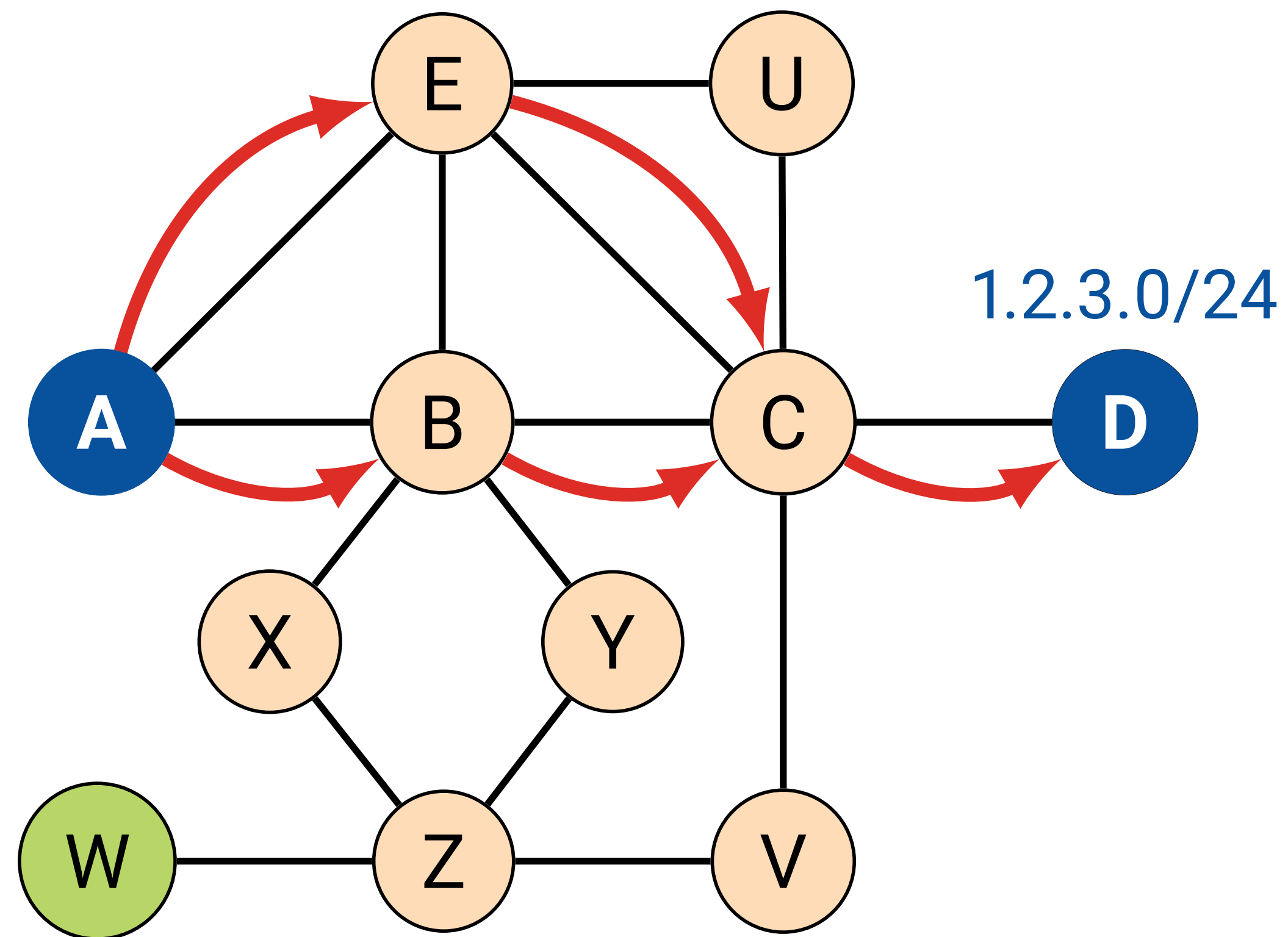
Fewer mirroring rules reduces bandwidth usage

MIRROR 1.2.3.0/24 ON [A B C D]



Too few mirroring rules creates **ambiguity**

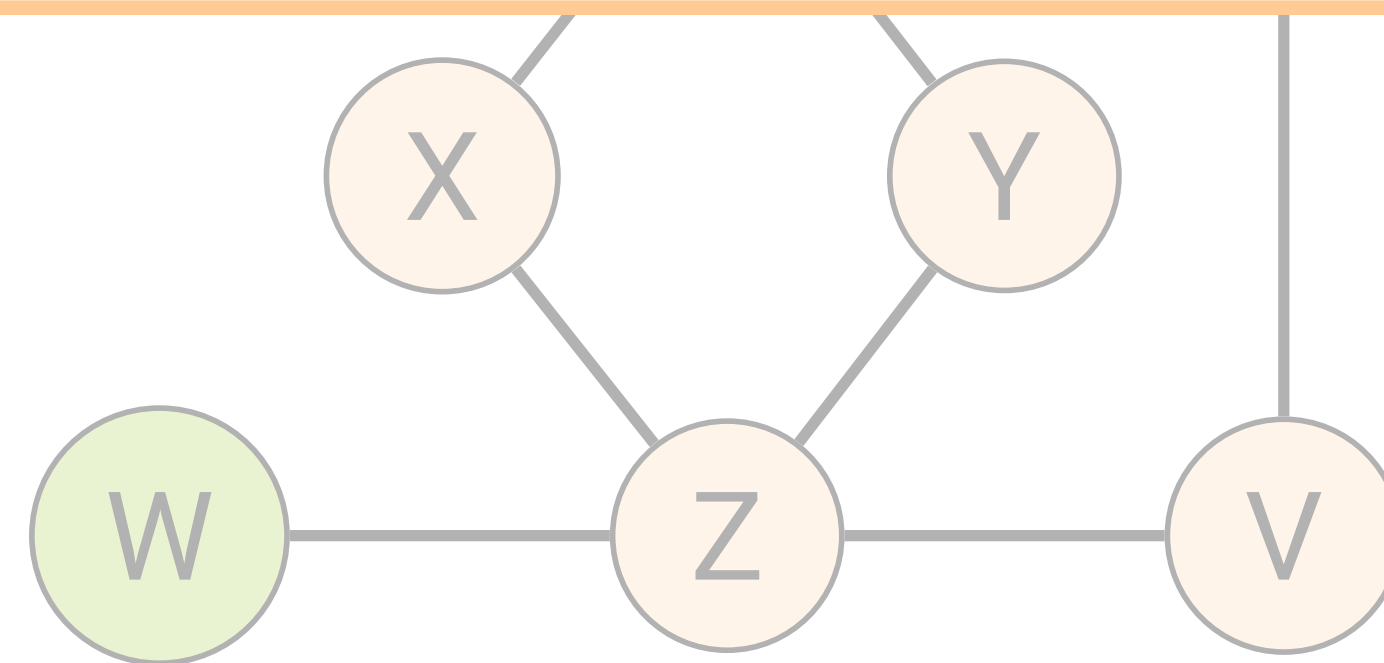
MIRROR 1.2.3.0/24 ON [A B C D]



Too few mirroring rules creates **ambiguity**

MIRROR 1.2.3.0/24 ON [A B C D]

The **Key-Points Sampling** algorithm minimizes mirroring rules and guarantees non-ambiguous reconstructed paths



Stroboscope defines two types of queries



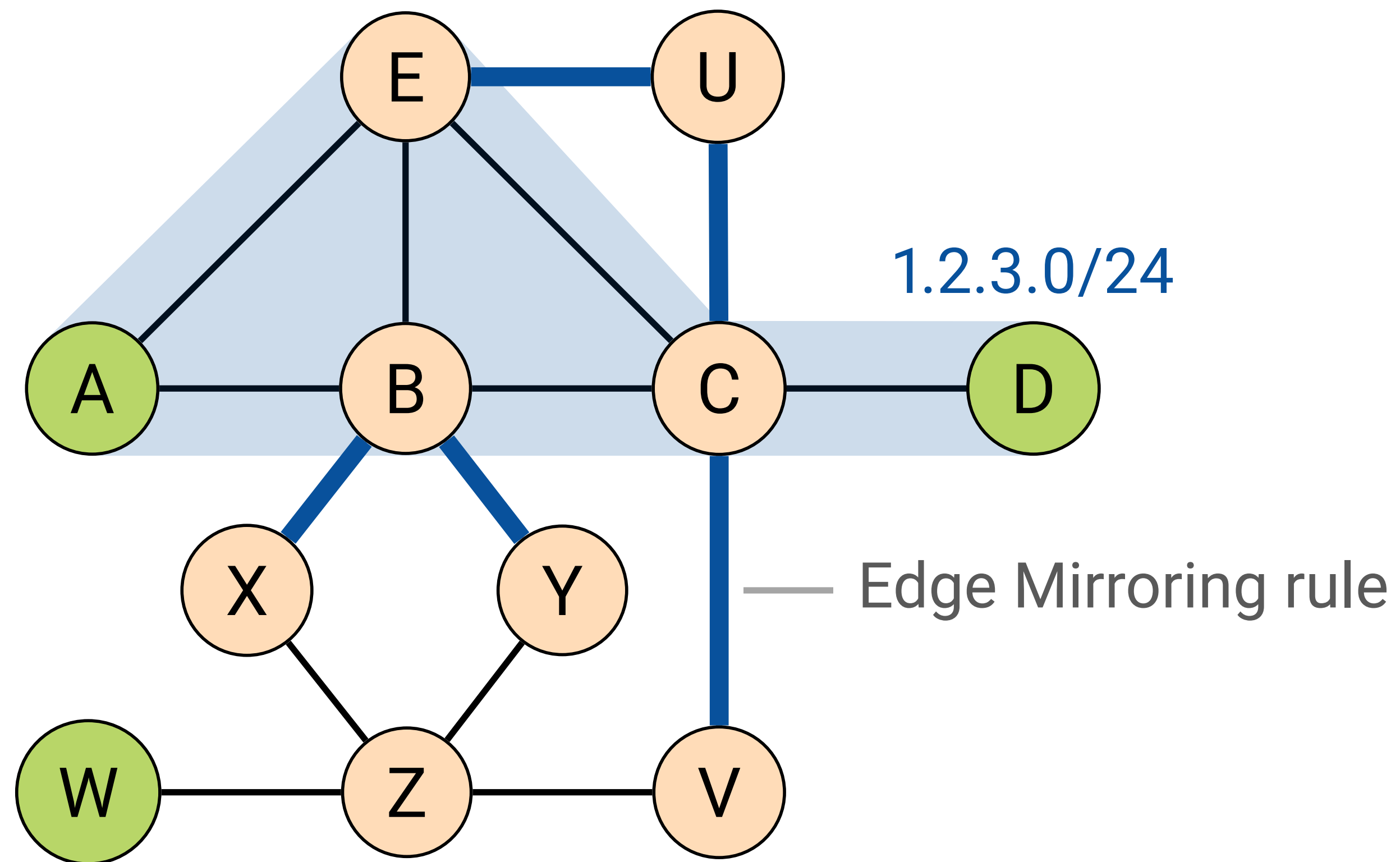
MIRROR



CONFINE

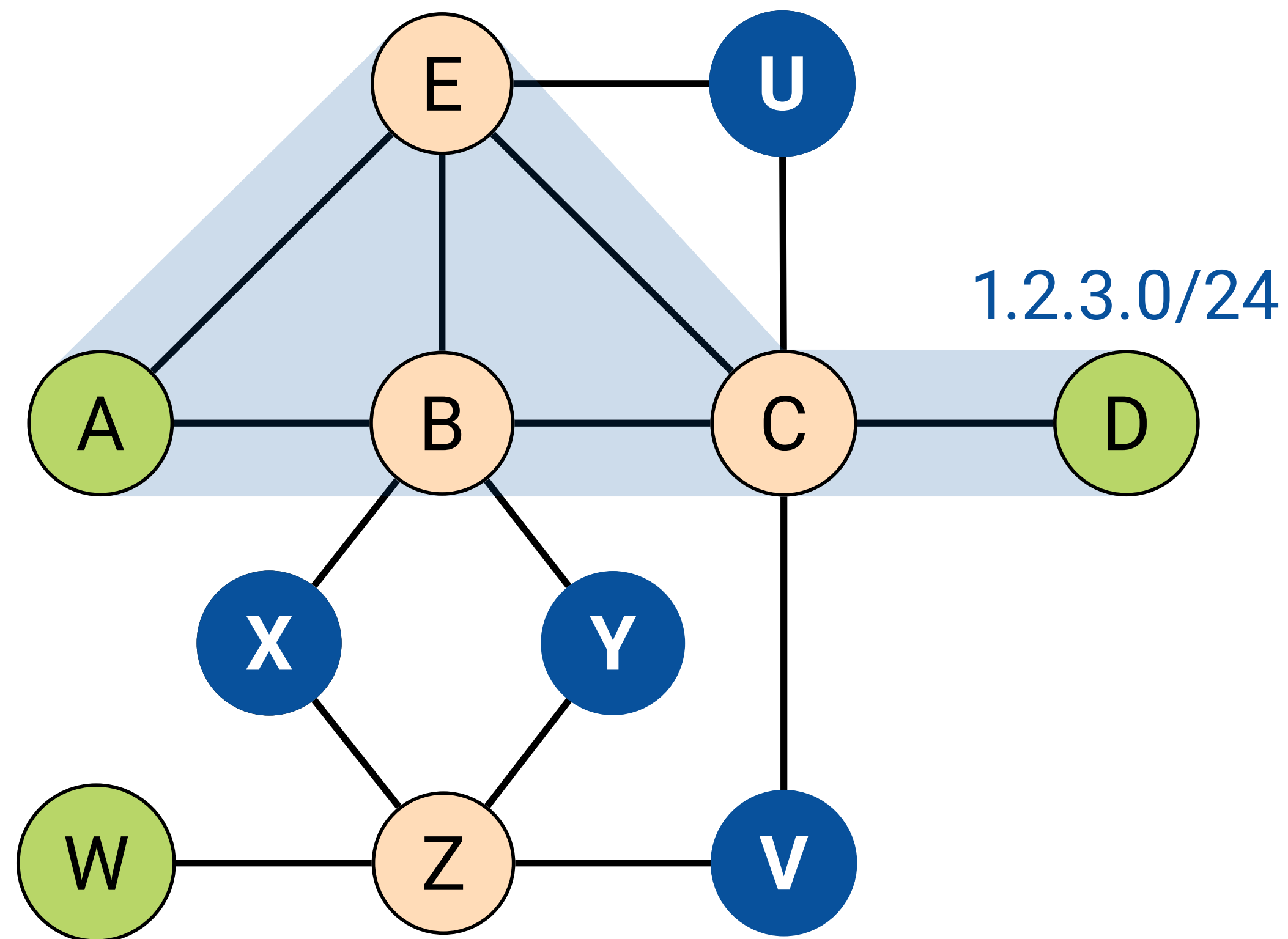
CONFINE queries mirror packets leaving a confinement region

CONFINE 1.2.3.0/24 ON [A B E C D]



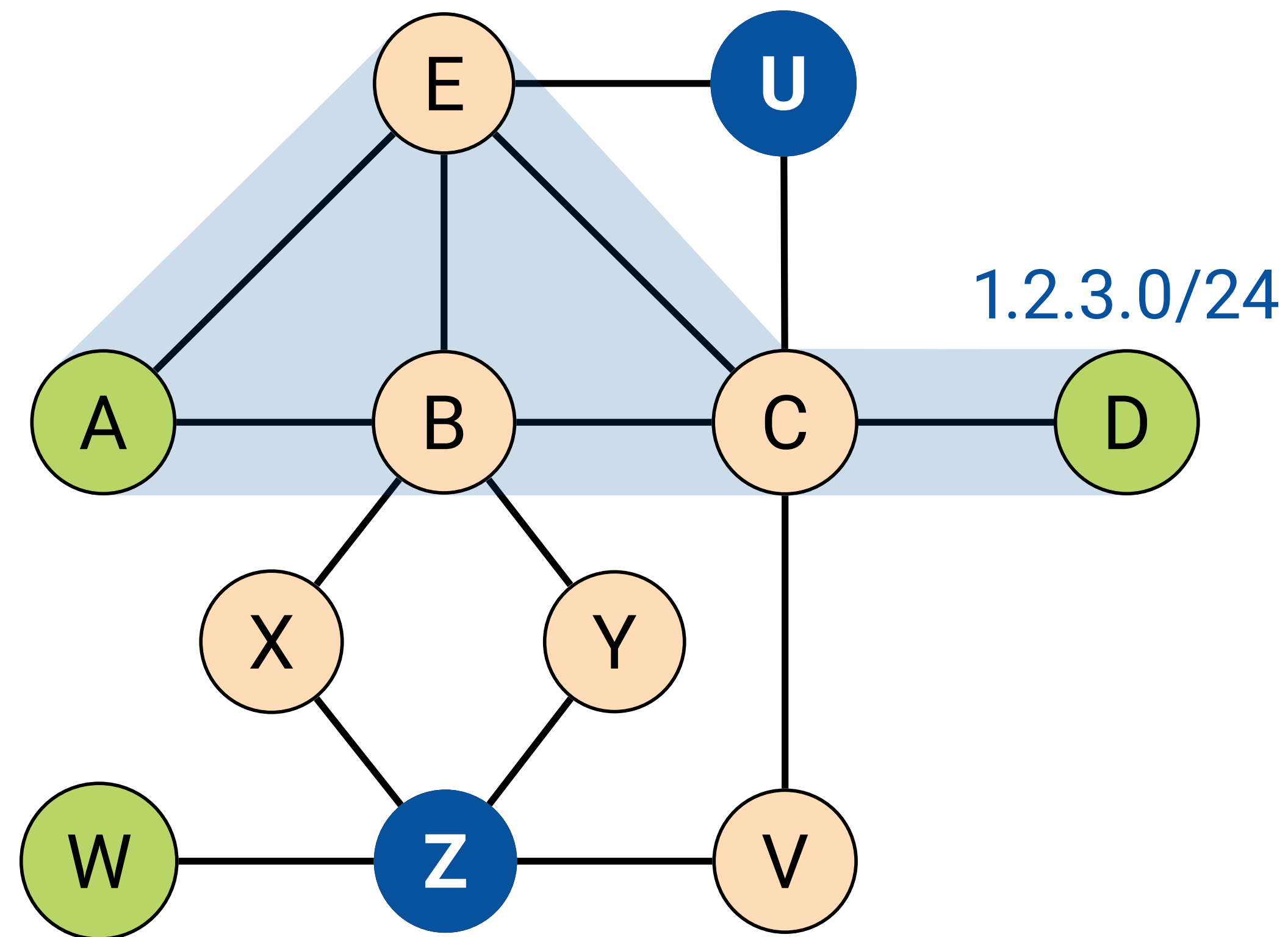
Fewer mirroring rules minimizes control-plane overhead

CONFINE 1.2.3.0/24 ON [A B E C D]



The lower bound is a multi-terminal node cut

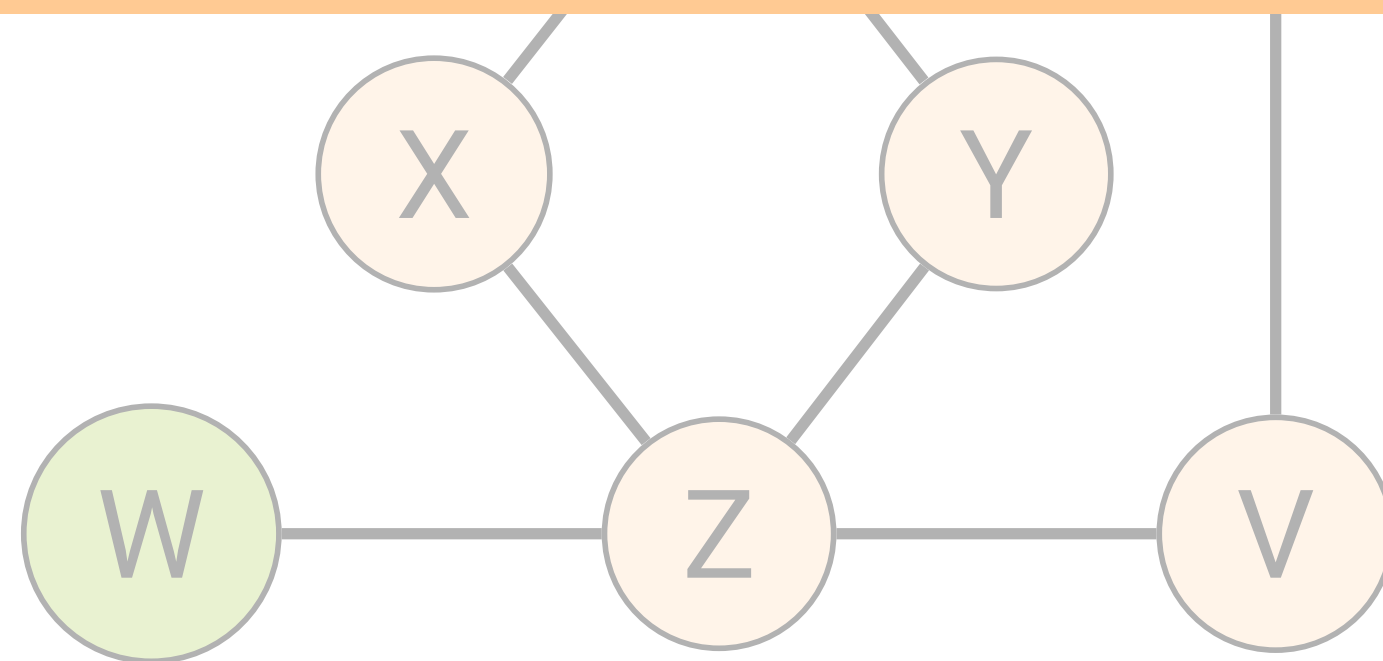
CONFINE 1.2.3.0/24 ON [A B E C D]



The lower bound is a multi-terminal node cut

CONFINE 1.2.3.0/24 ON [A B E C D]

The **Surrounding** algorithm minimizes mirroring rules and guarantees to mirror any packet leaving the confinement region



Stroboscope: Declarative Network Monitoring on a Budget

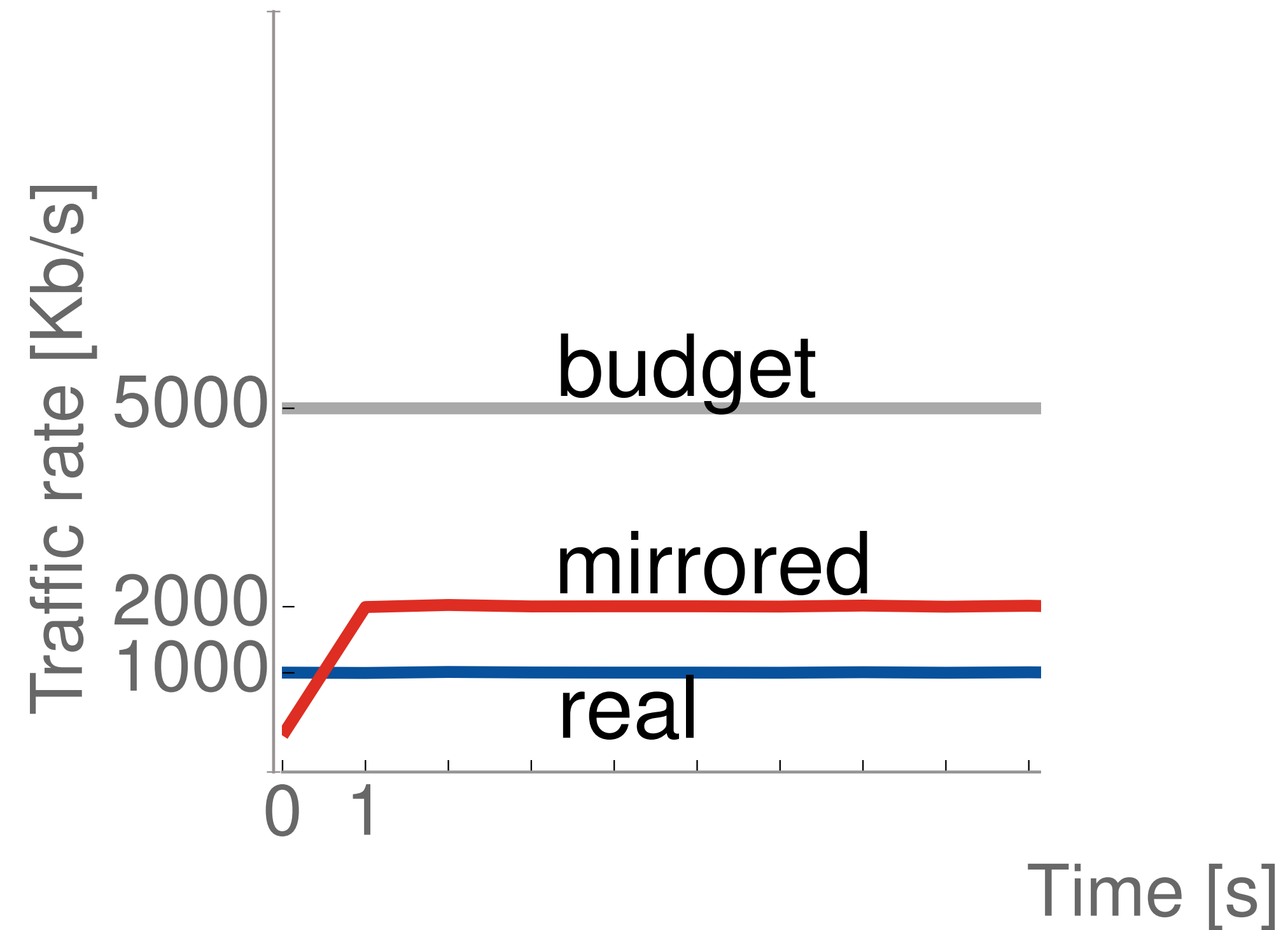
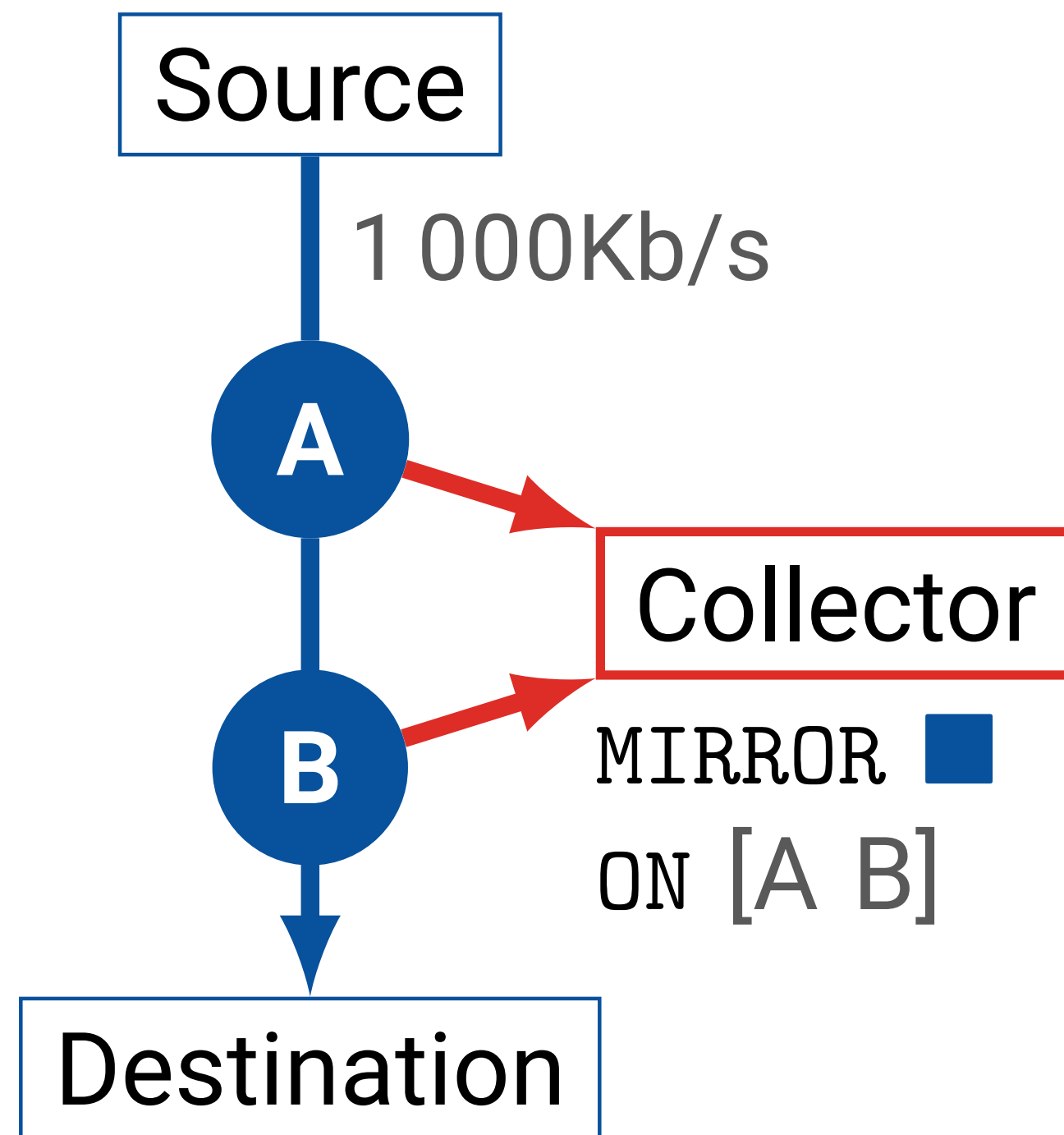


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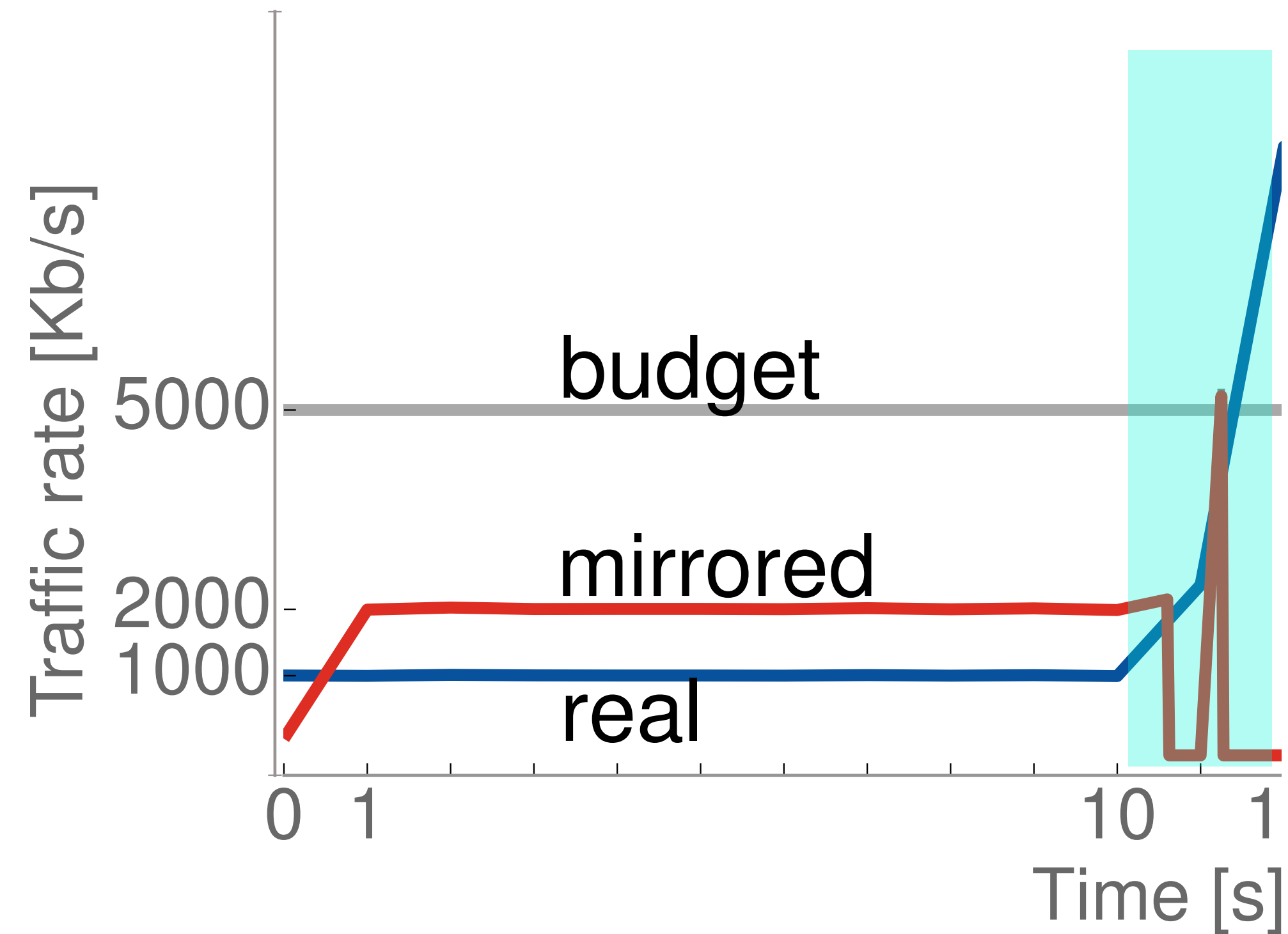
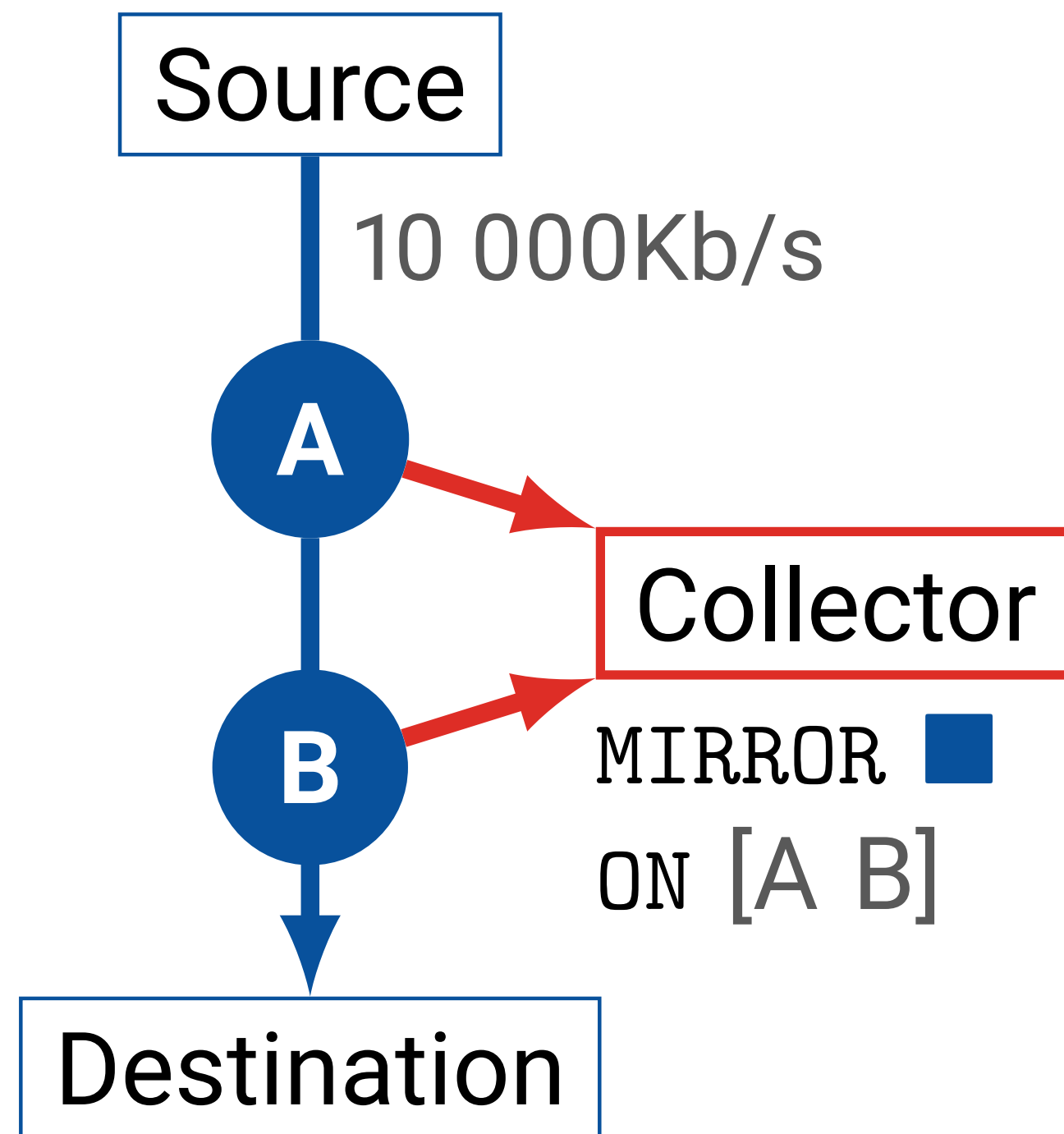
Stroboscope works with currently deployed routers

- Most vendors provide traffic mirroring and encapsulation primitives
- The collector activates mirroring for a flow by updating one ACL
- Routers autonomously deactivate mirroring rules using timers
- Traffic slices can be as small as **23 ms** on our routers (Cisco C7018)

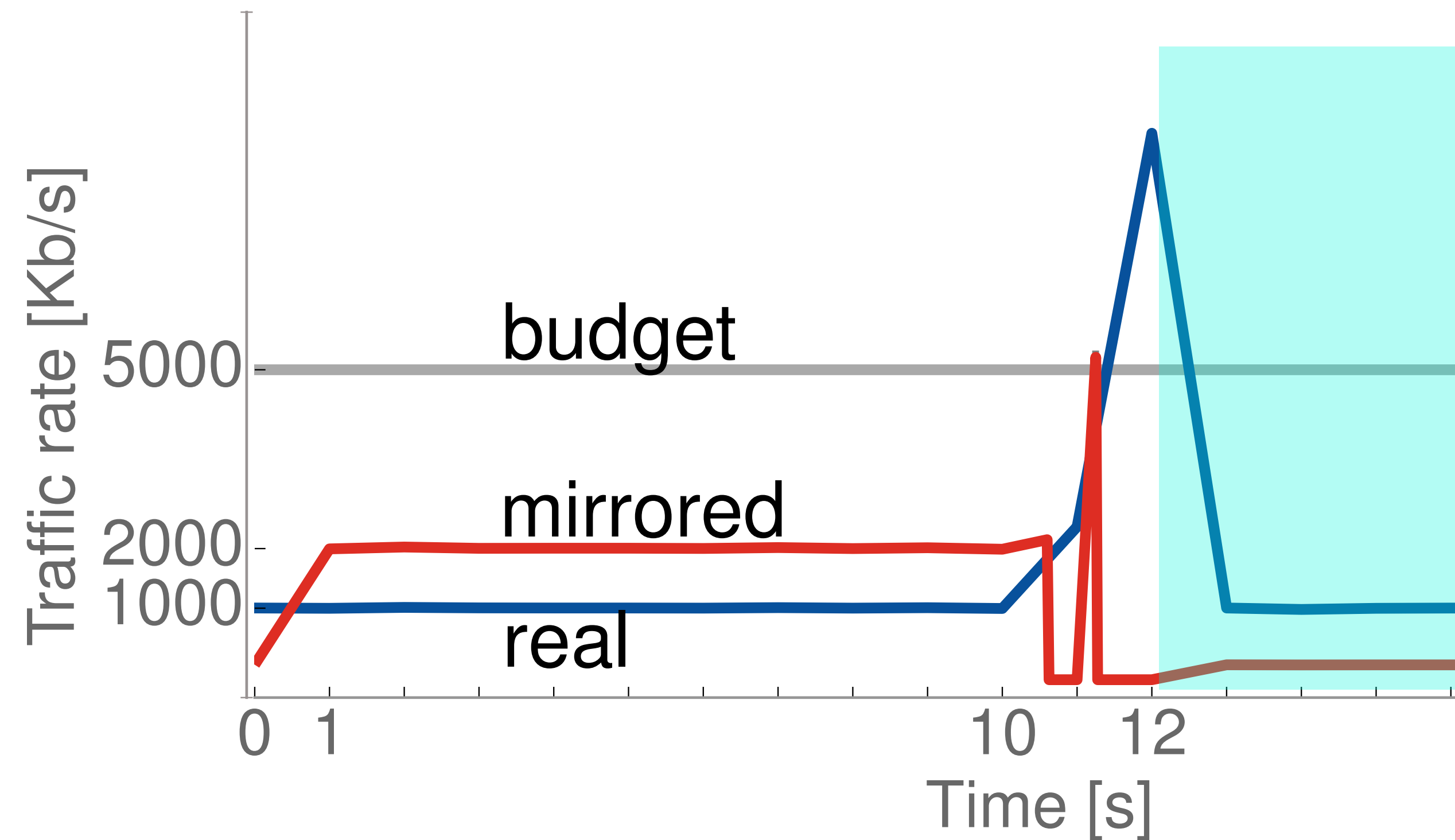
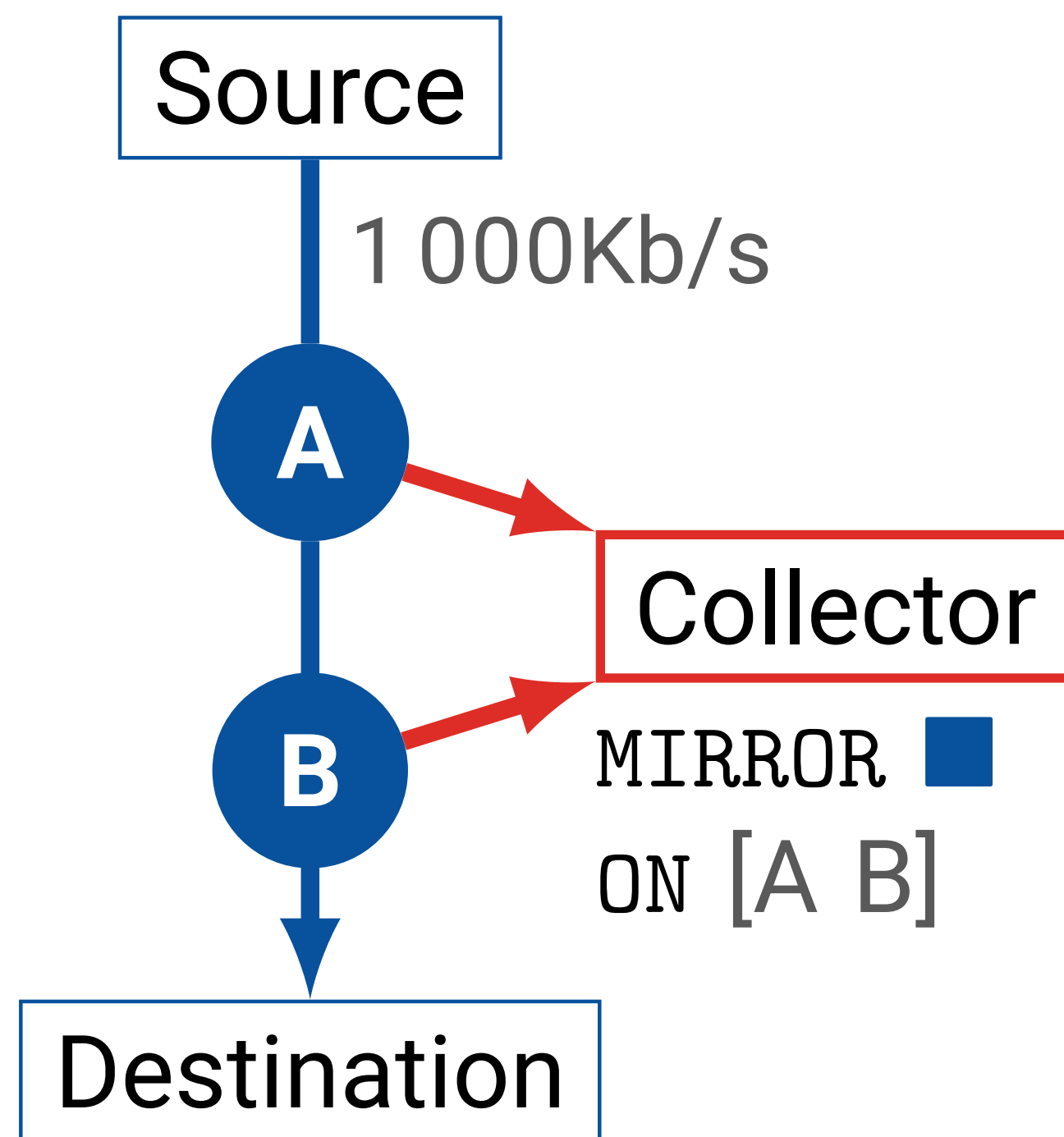
Stroboscope tracks the rate of mirrored traffic in real time



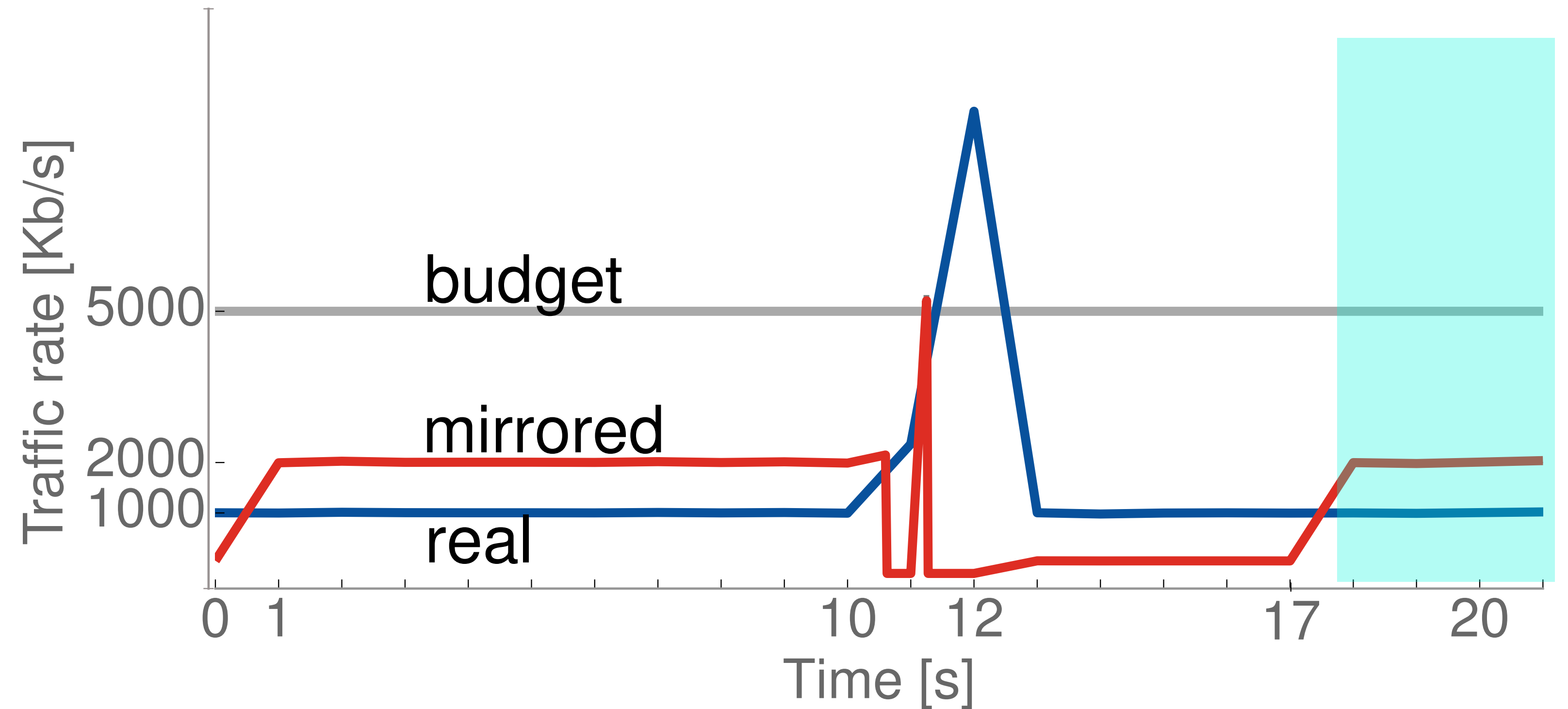
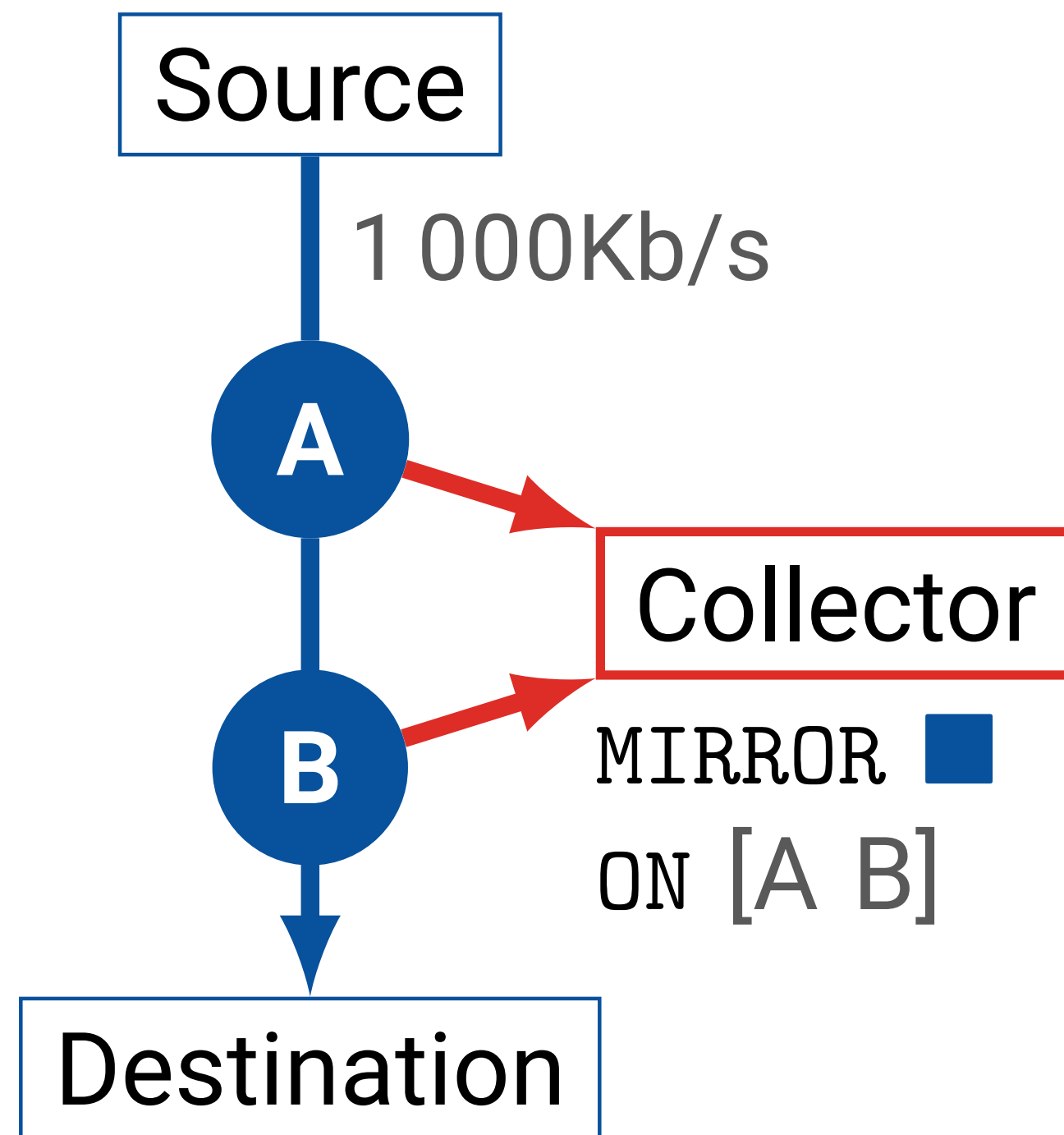
Measurement campaigns are stopped early if the estimated demand are exceeded



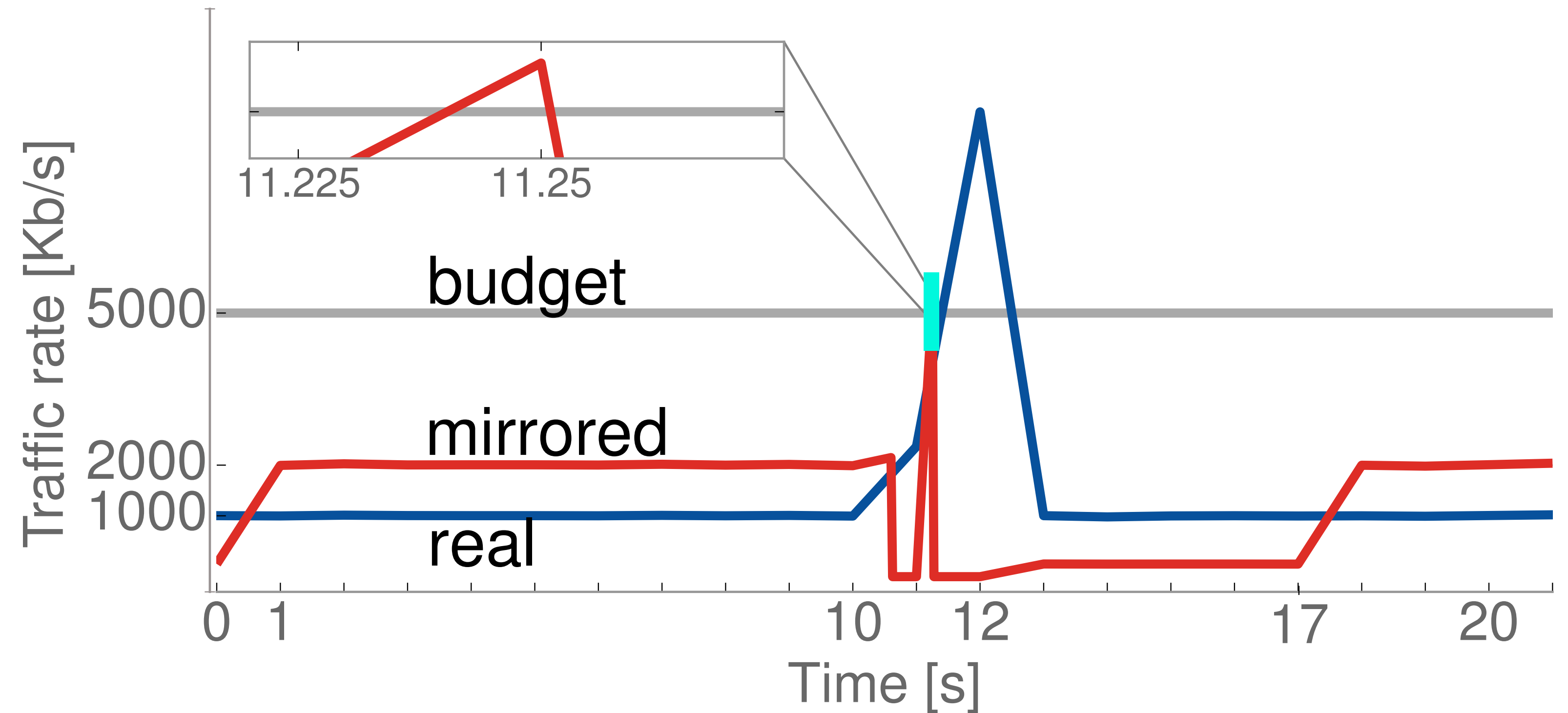
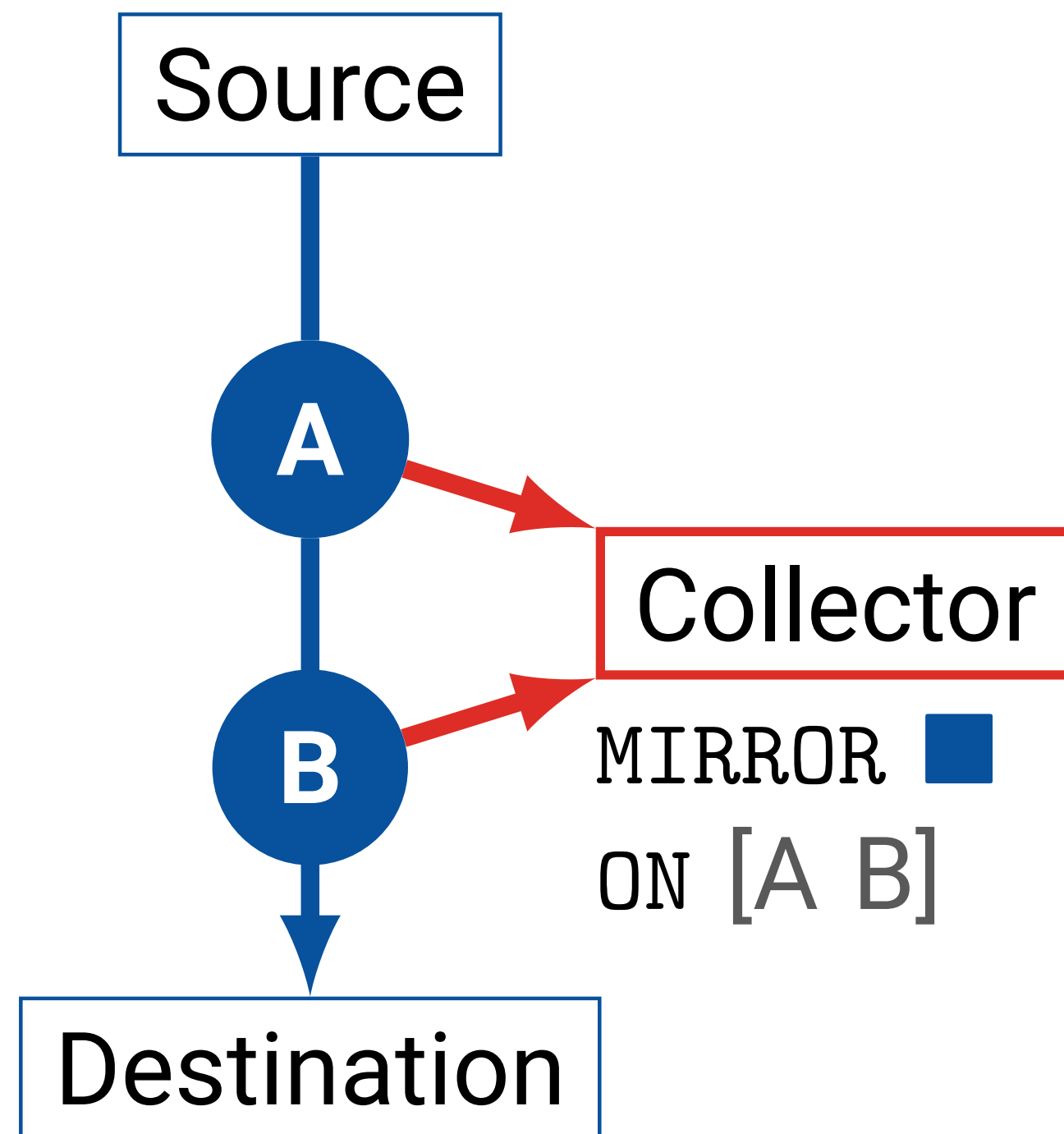
Exceeding the total budget schedules the query once per measurement campaign



Stable recorded traffic rates are used for future estimations



Stroboscope exceeds the monitoring budget for at most one timeslot



Stroboscope: Declarative Network Monitoring on a Budget



- Traffic slicing as a first-class data-plane primitive
- Strong guarantees on budget compliance and measurement accuracy
- Measurement analysis decoupled from measurement collection

Net2Text: Query-guided Network Captioning



Rüdiger Birkner



Dana Drachsler-Cohen



Martin Vechev



Laurent Vanbever

USENIX Symposium on Networked Systems Design and Implementation. April 2018.

Net2Text



Net2Text



|Type a message...

Net2Text



Where is the traffic...

Net2Text



question *in*
natural language

Where is the traffic leaving
in NEWY coming from?



Type a message...

Net2Text



Where is the traffic leaving
in NEWY coming from?



Type a message...

Net2Text



Where is the traffic leaving
in NEWY coming from?



The traffic enters mostly in PHIL
and goes to Youtube and Netflix.



Type a message...

summary *in*
natural language

Net2Text



question *in*
natural language

Where is the traffic leaving
in NEWY coming from?



The traffic enters mostly in PHIL
and goes to Youtube and Netflix.



summary *in*
natural language

Type a message...

Finding *a* summary of the
network-wide forwarding state is simple

Finding *a* summary of the
network-wide forwarding state is simple

Traffic is forwarded.

Finding *a* summary of the
network-wide forwarding state is simple

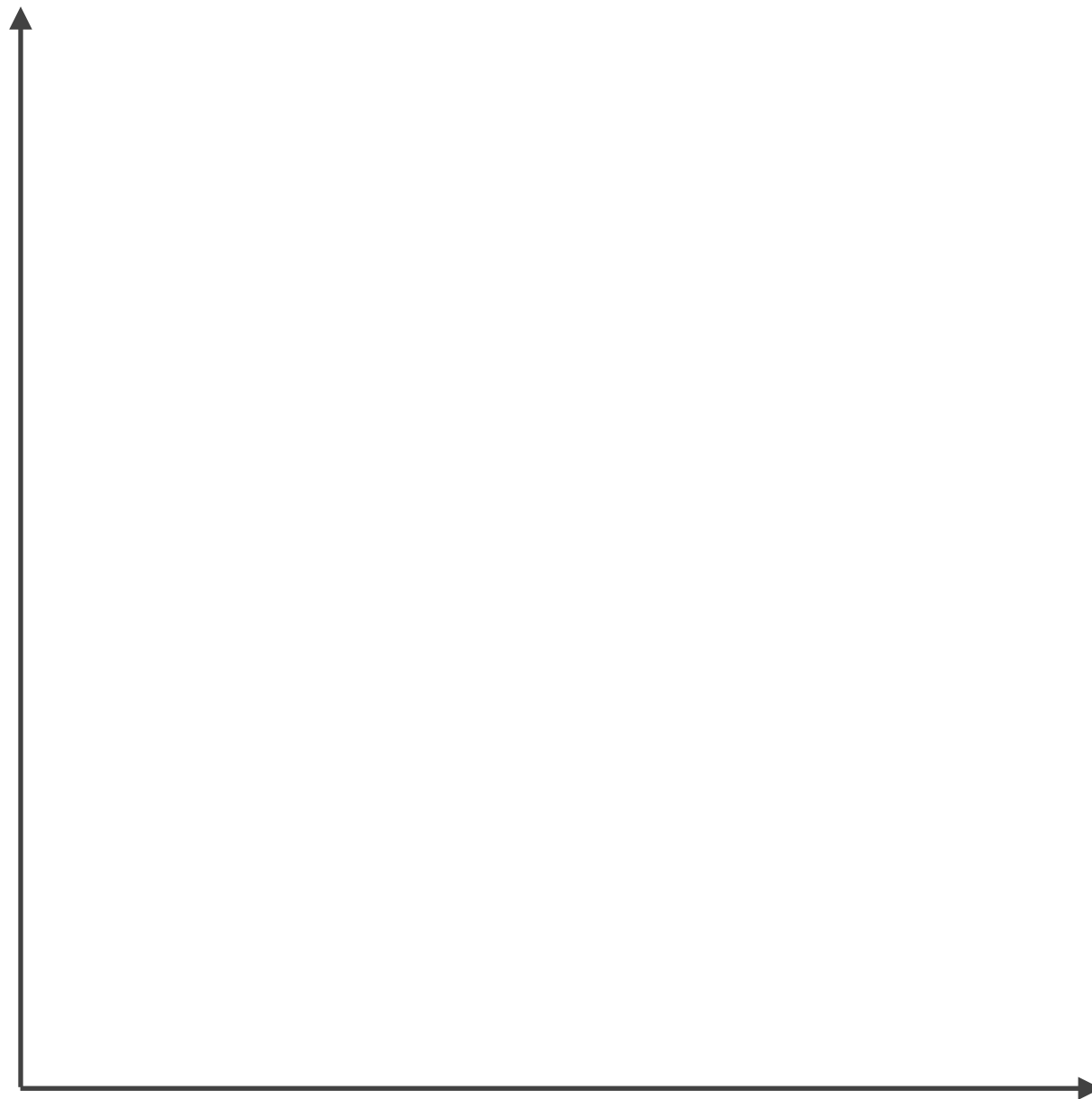
Traffic from LOSA to 35.184.0.0/19,
which is owned by Google,
is leaving the network in CHIC
and takes the path
SUNV, DENV, KSCY, INDI to CHIC.

Finding **a good** summary of the
network-wide forwarding state is **hard**

Summarization is a multi-objective optimization problem

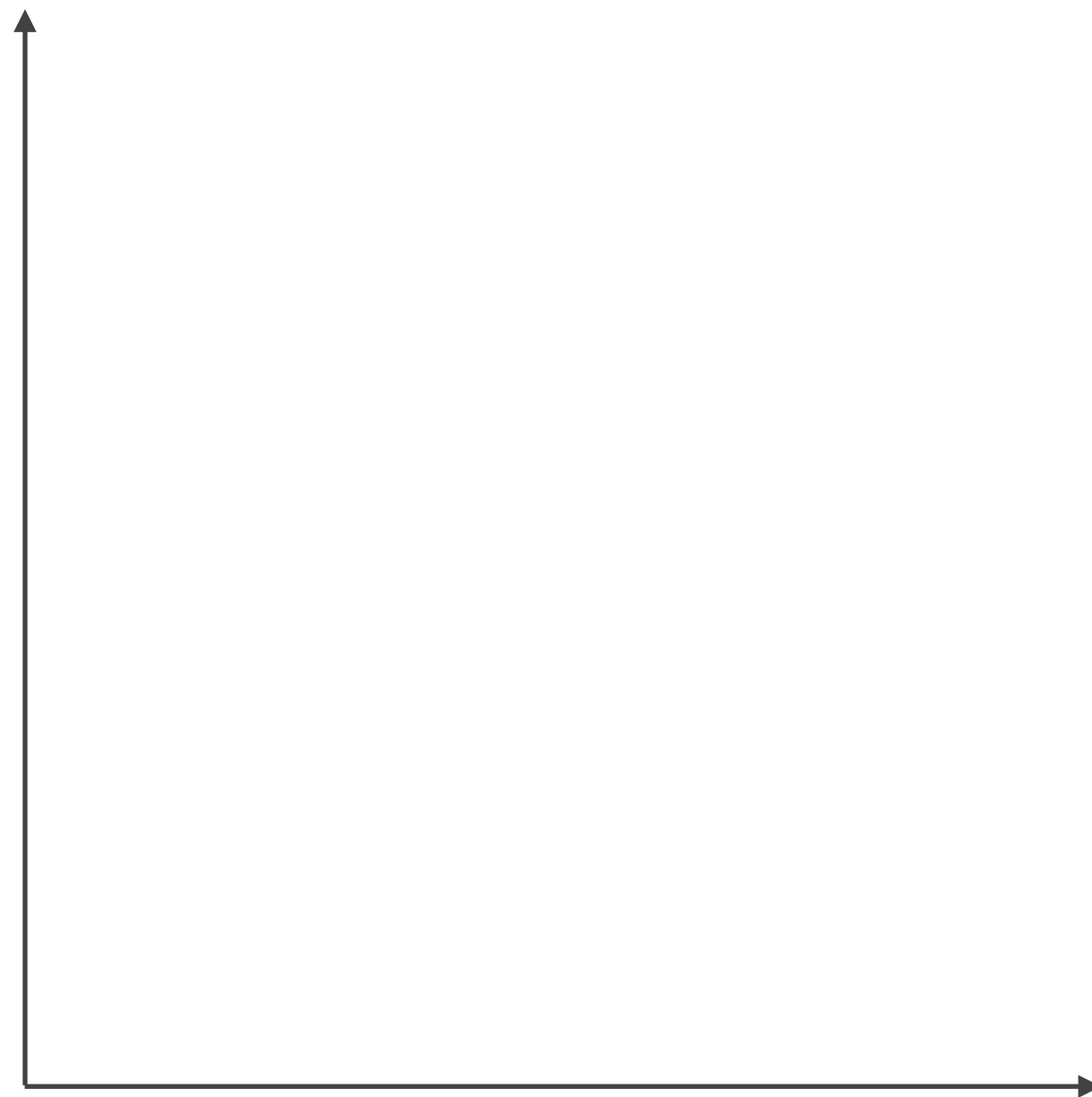
Coverage

amount of data
described by the summary



Coverage

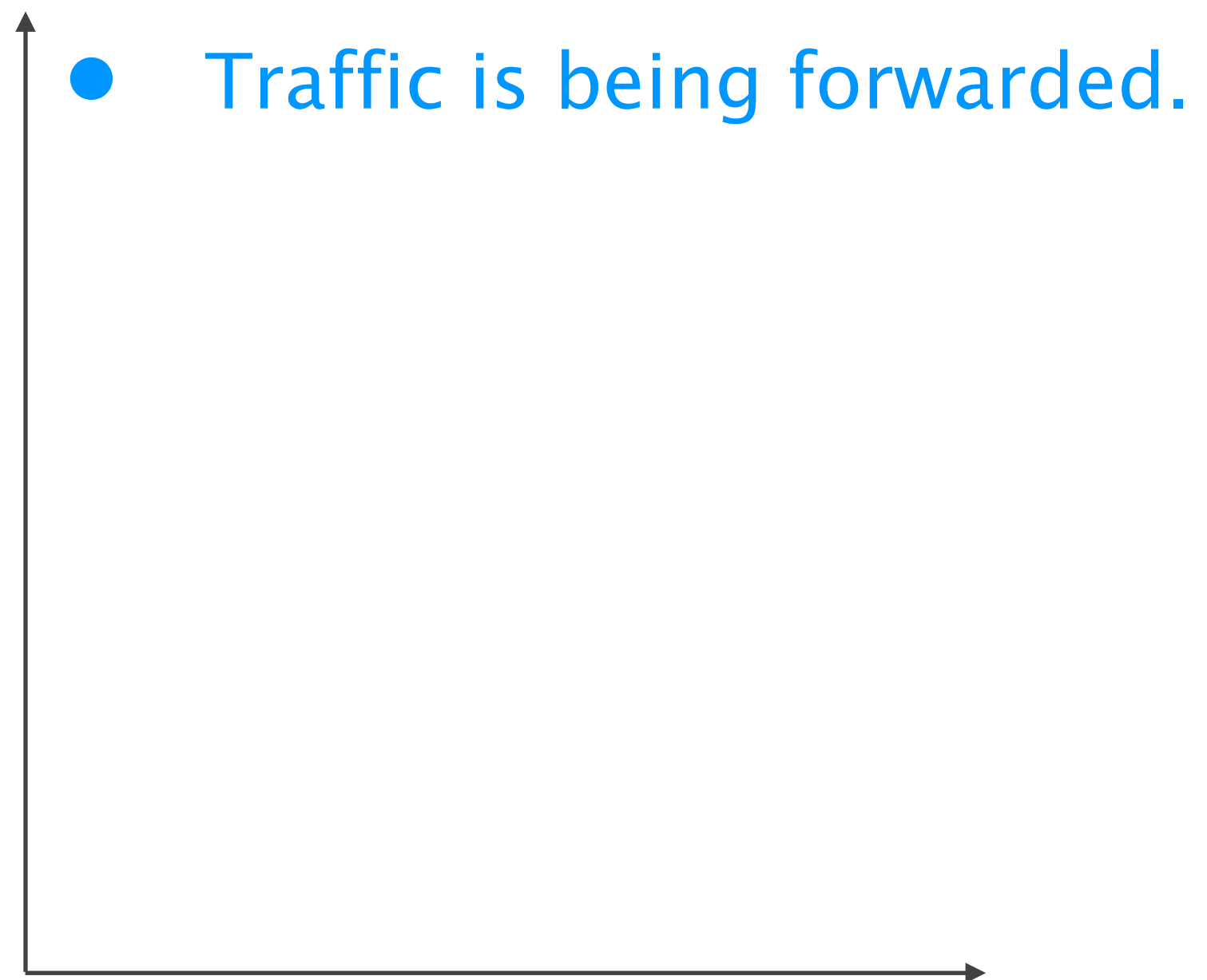
amount of data
described by the summary



Explainability

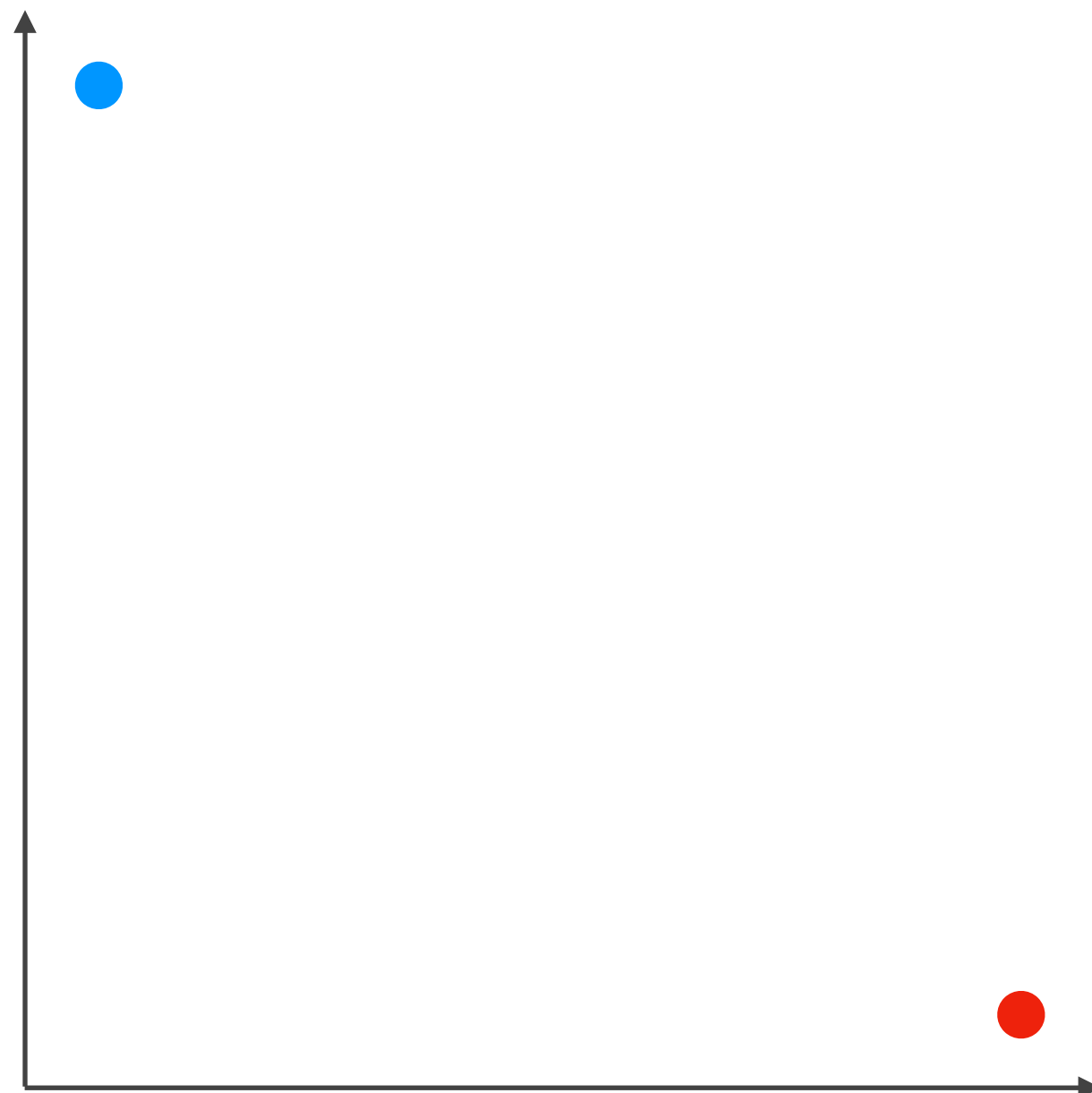
amount of detail
provided by the summary

Coverage



Explainability

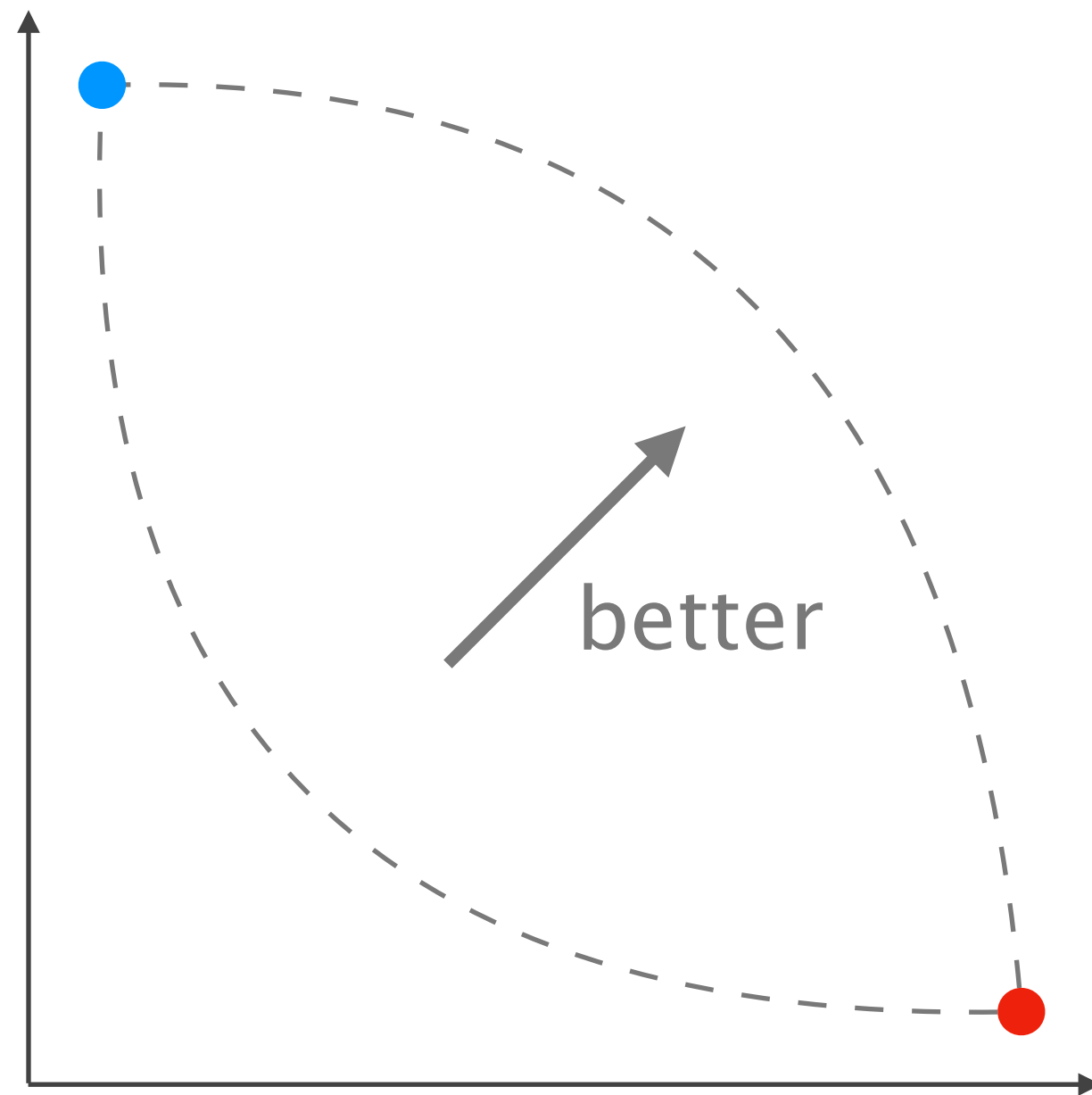
Coverage



Traffic from LOSA to 35.184.0.0/19,
which is owned by Google, ...

Explainability

Coverage



Explainability

Summarization is a multi-objective optimization problem

Score

Weighted sum of the amount of traffic covered by each path specification in the summary.

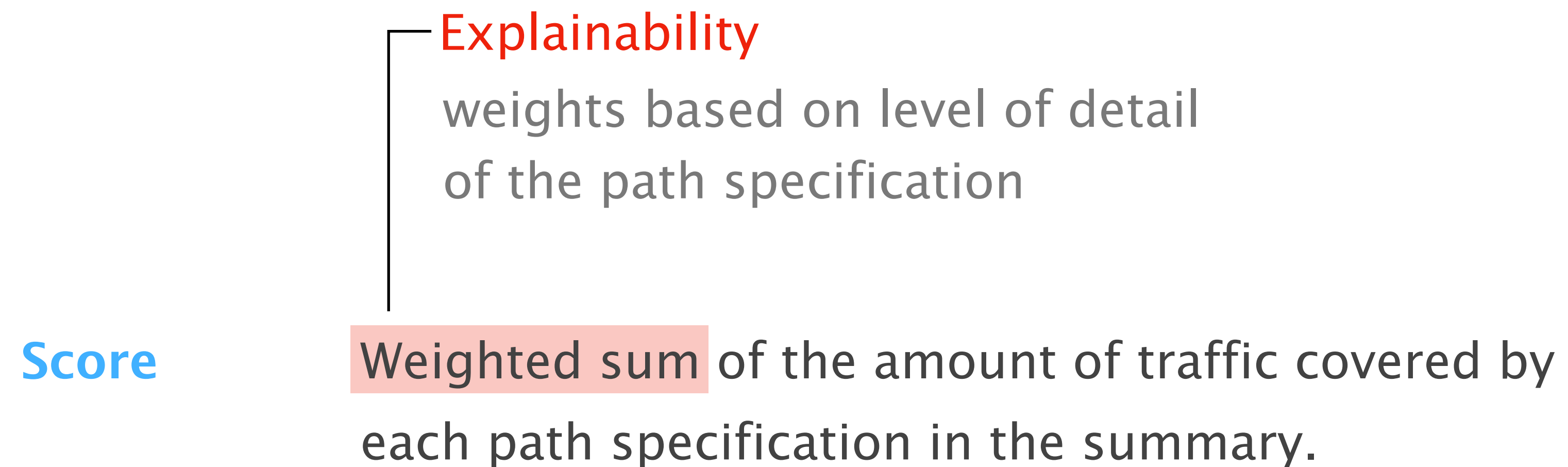
Summarization is a multi-objective optimization problem

Score

Weighted sum of the amount of traffic covered by each path specification in the summary.

Coverage

Summarization is a multi-objective optimization problem



Summarization is a multi-objective optimization problem

Score

Weighted sum of the amount of traffic covered by each path specification in the summary.

Goal

Find path specifications that maximize the score.

Summarization is a multi-objective optimization problem

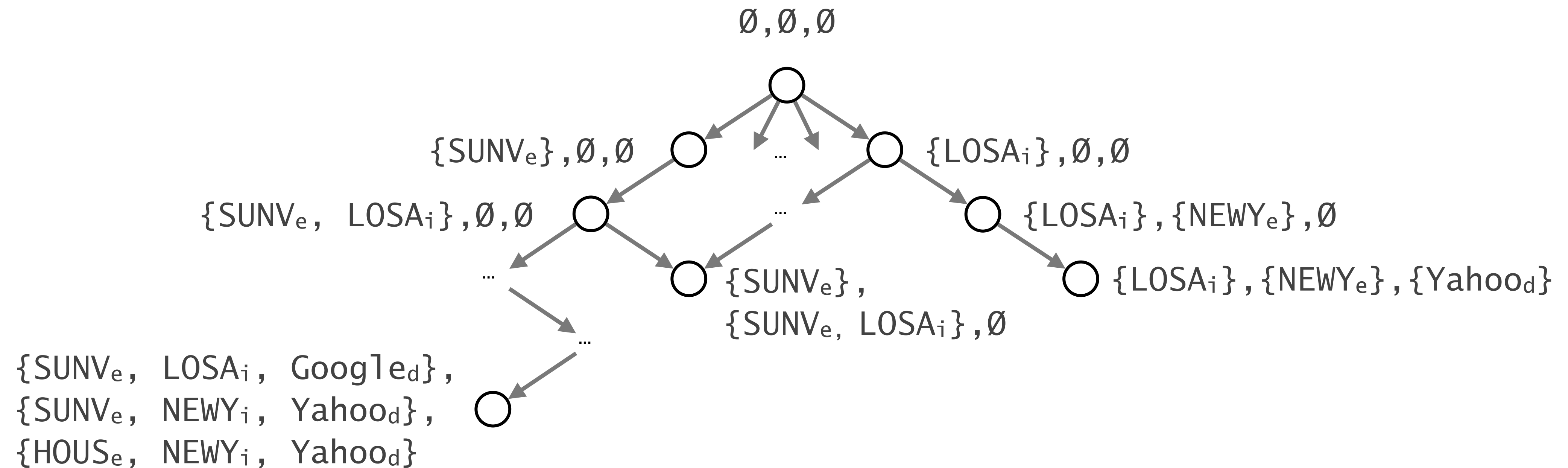
Score

Weighted sum of the amount of traffic covered by each path specification in the summary.

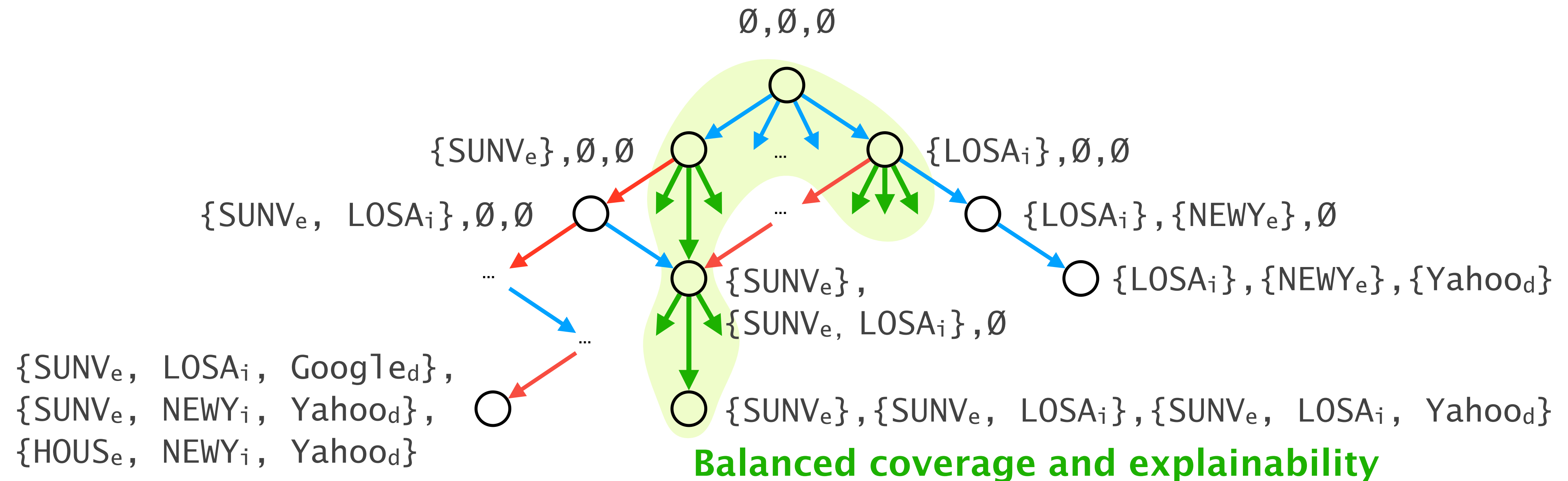
Goal

Find k path specifications each of size at most t that maximize the score.

The search space is exponential
in the number of path specifications and feature values



Net2Text reduces the search space to solutions that balance coverage and explainability



Self-driving/monitoring networks

in the age of deep network programmability

1 operator assistance

2 partial automation

3 conditional automation

4 high automation

5 full automation

Part I

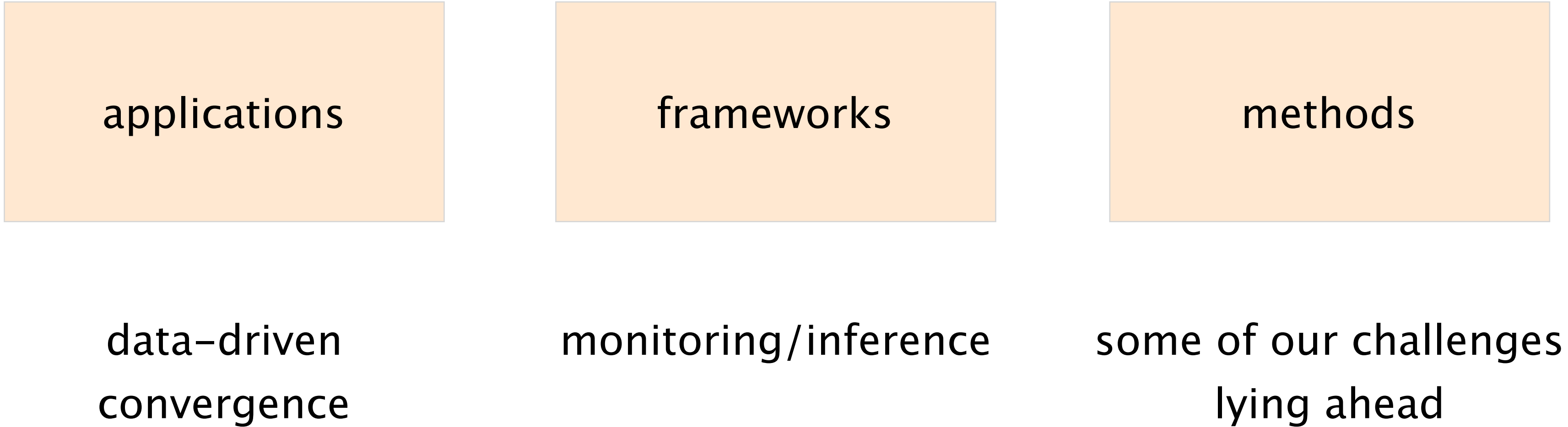
assisting operators

Part II

taking control

too futuristic? 😊

Part II: Taking control



applications

data-driven
convergence

frameworks

monitoring/inference

methods

some of our challenges
lying ahead

Part II: Taking control

applications

frameworks

methods

data-driven
convergence

Blink

Fast Connectivity Recovery Entirely in the Data Plane



Thomas Holterbach



Edgar
Costa Molero



Maria Apostolaki



Alberto Dainotti



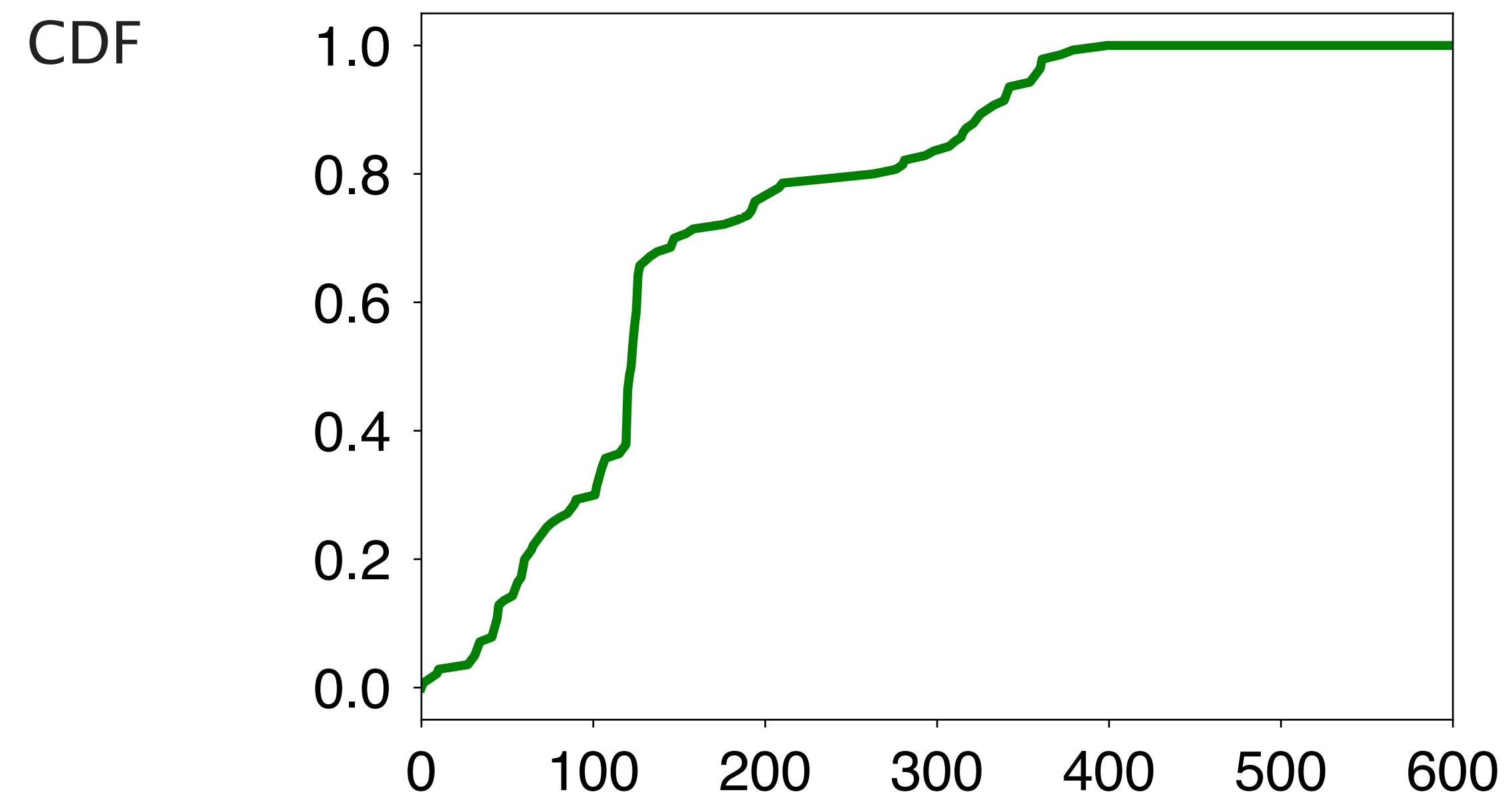
Stefano Vissicchio



Laurent Vanbever

USENIX Symposium on Networked Systems Design and Implementation. February 2019.

Internet control plane can take **minutes to converge**
upon remote failures

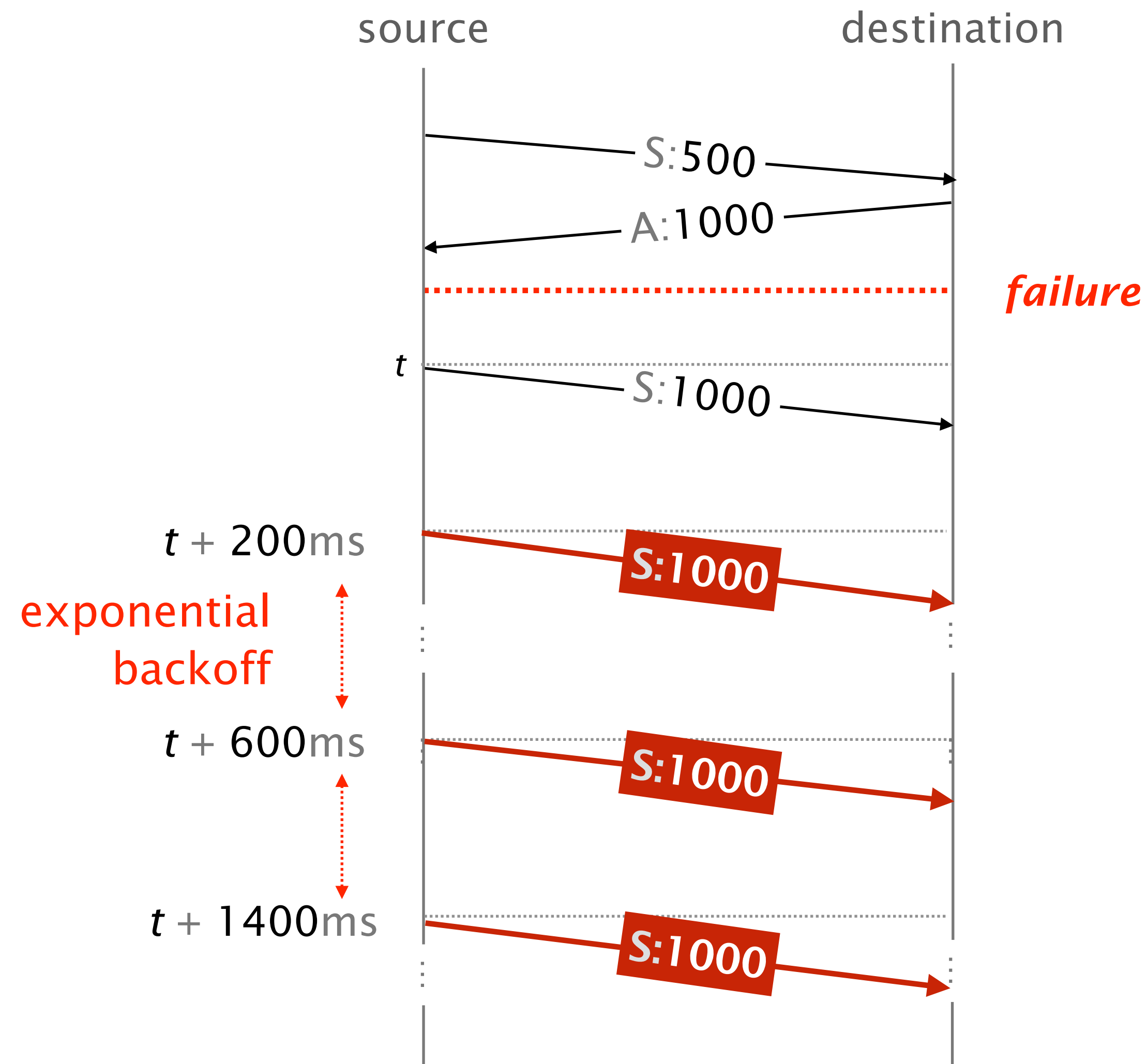


Second elapsed between the data plane failure
and the control plane learning about it

While the control plane can take minutes to learn about a failure,
the actual Internet traffic is affected almost instantaneously

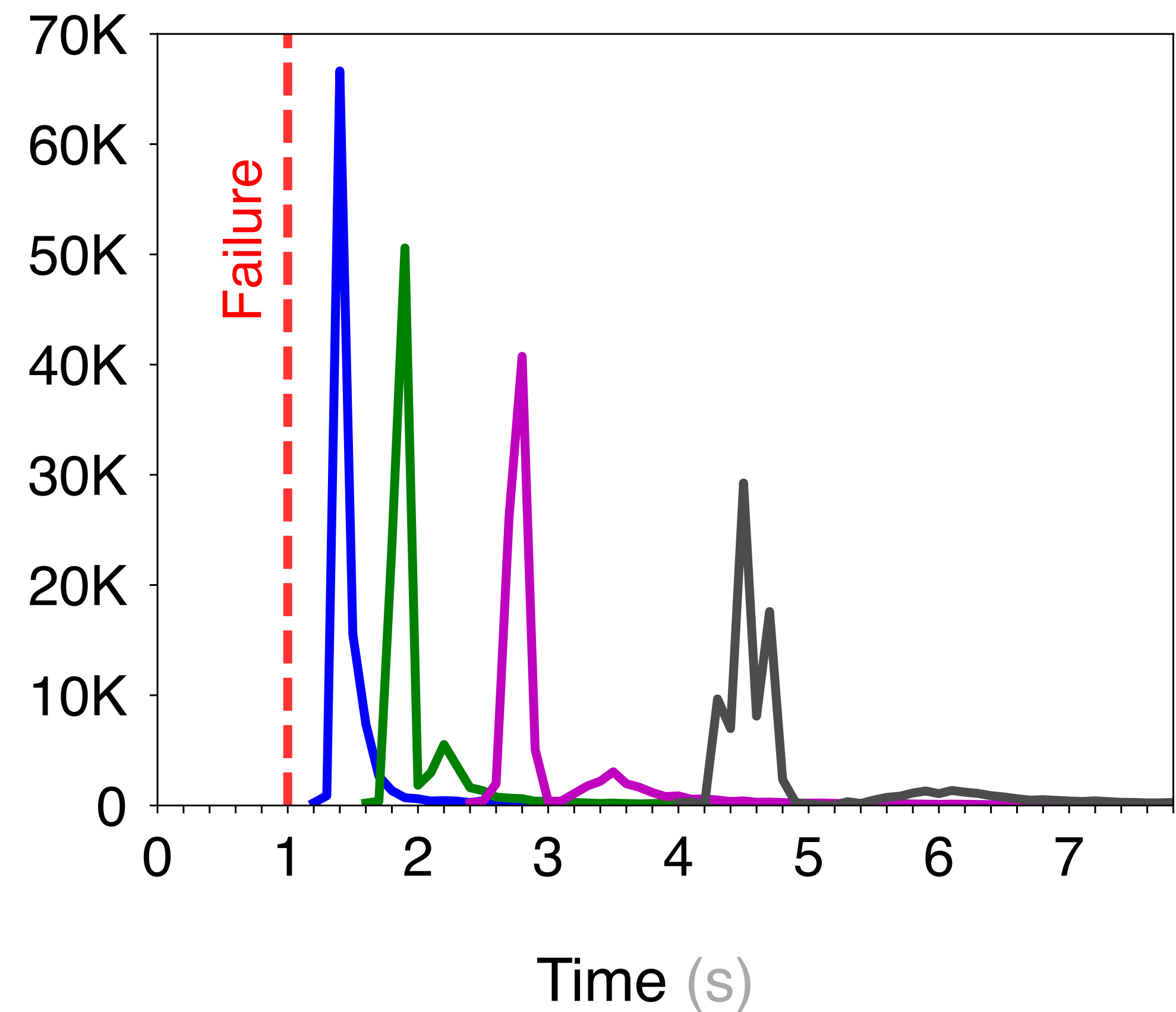
How about we track this signal instead?

Internet traffic end-points retransmit packets upon experiencing packet drops



Many end-points retransmitting leads to
waves of retransmission

Number of
retransmissions



Tracking this signal in the data plane is challenging

Challenges

Signal is noisy

packets loss are routinely observed

Signal fades away quickly

due to the exponential backoff

Signal is compounded over many small ones

requires per-connection tracking, which is hard to scale

We solve these challenges by considering a subset of the signal that we carefully craft for maximal signal-to-noise ratio

Solutions

Signal is noisy

Focus on retransmissions caused by bursty losses

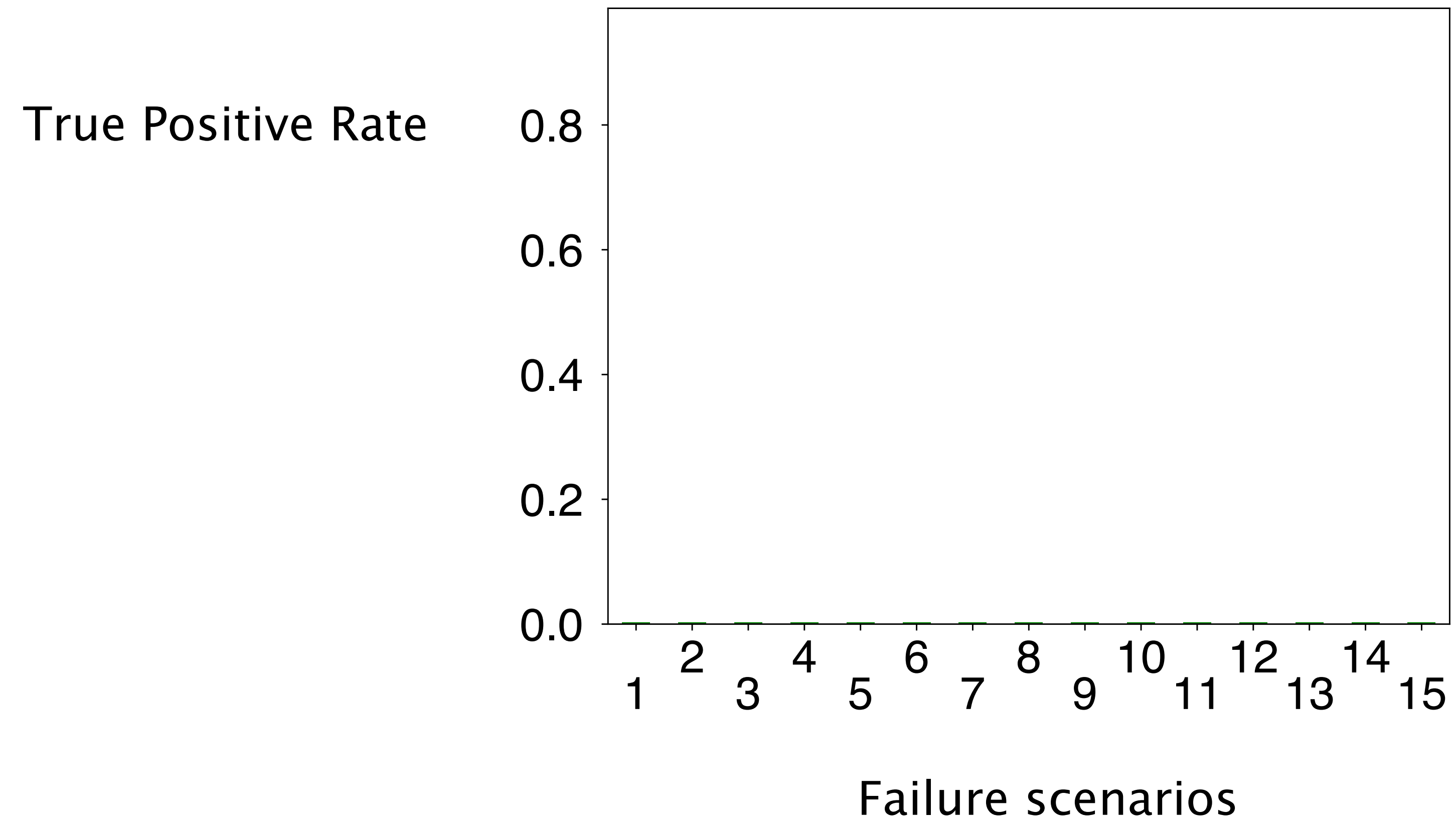
Signal fades away quickly

Focus on active flows (fast to retransmit after a failure)

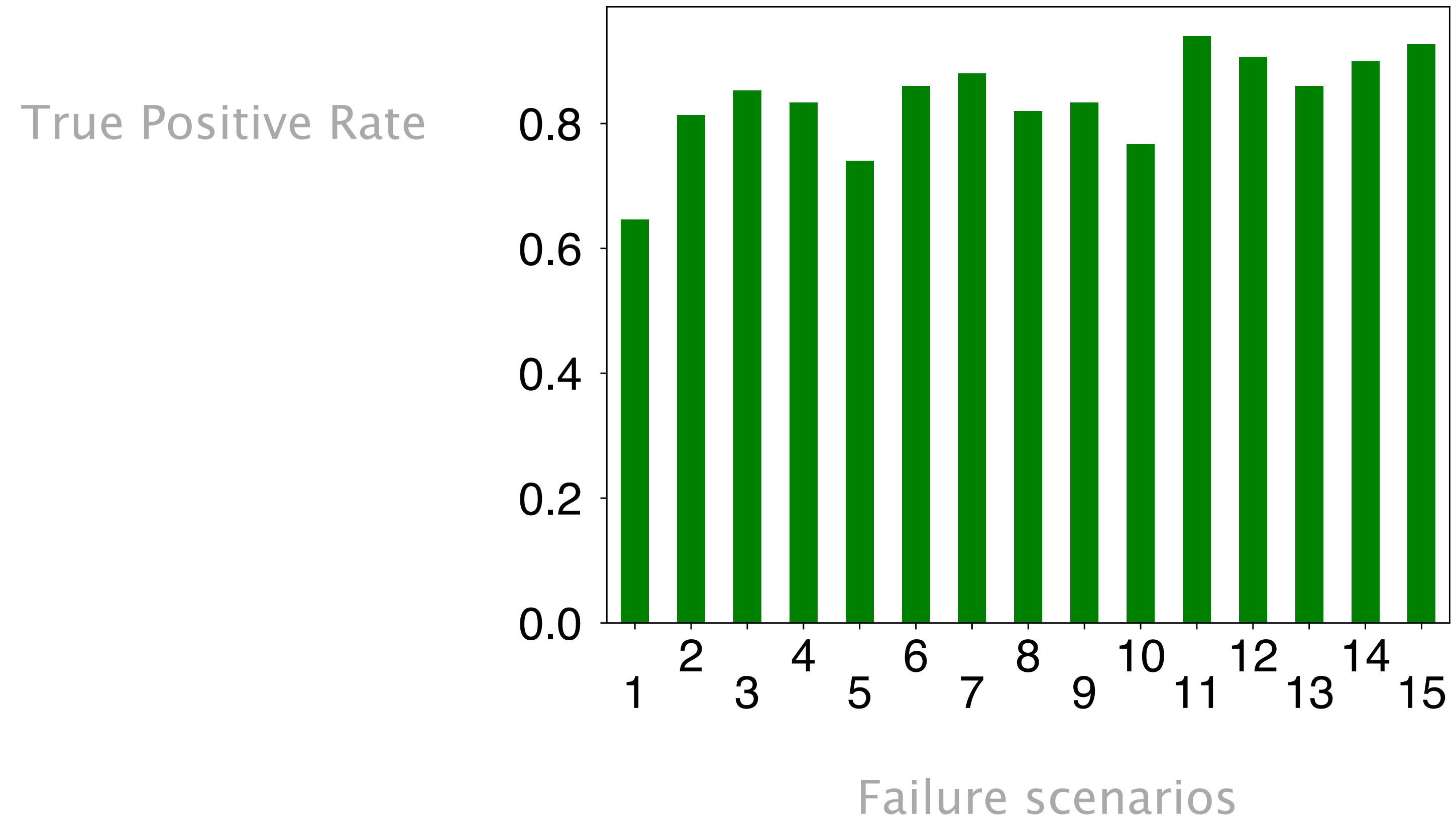
Signal is compounded over many small ones

Rely on scalable data structures and sampling

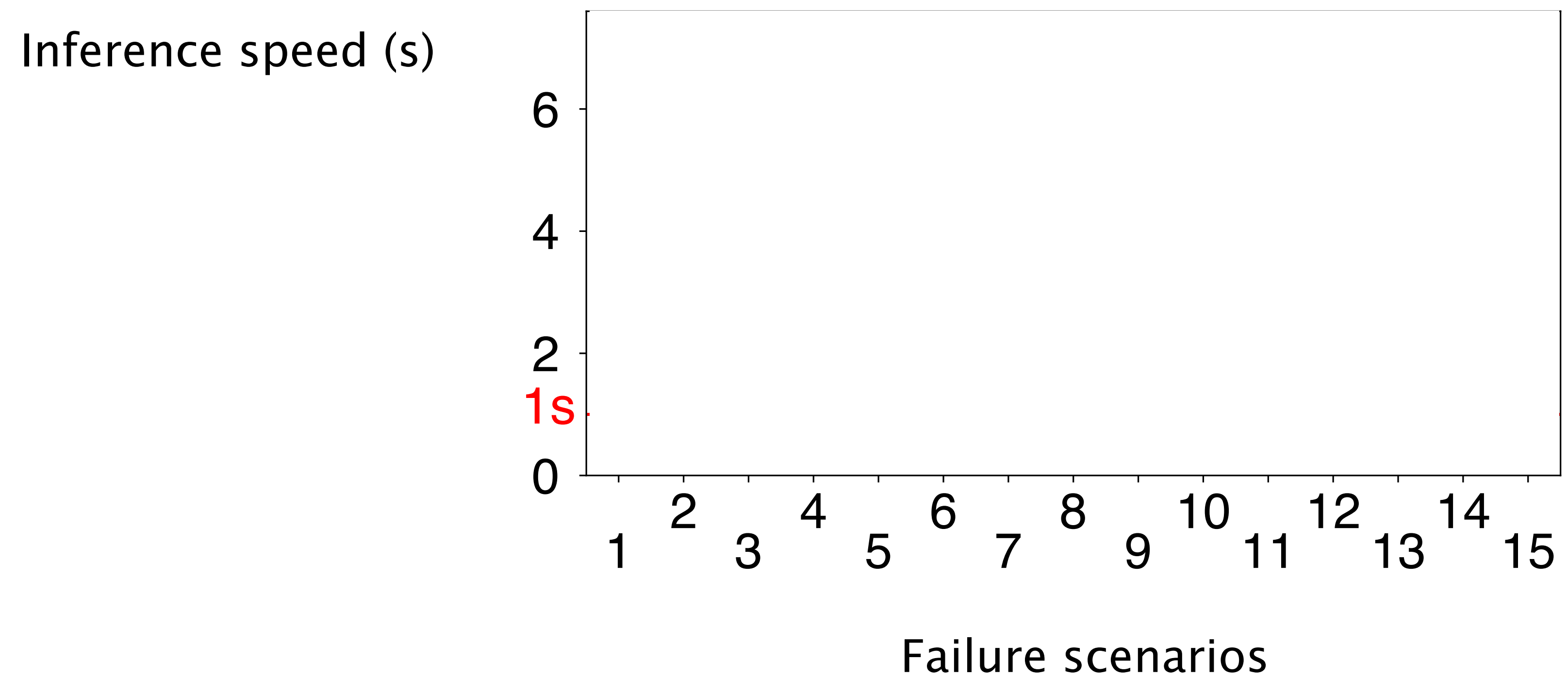
Blink works well in practice,
with an inference accuracy above 80% in most cases



Blink works well in practice,
with an inference accuracy above 80% in most cases

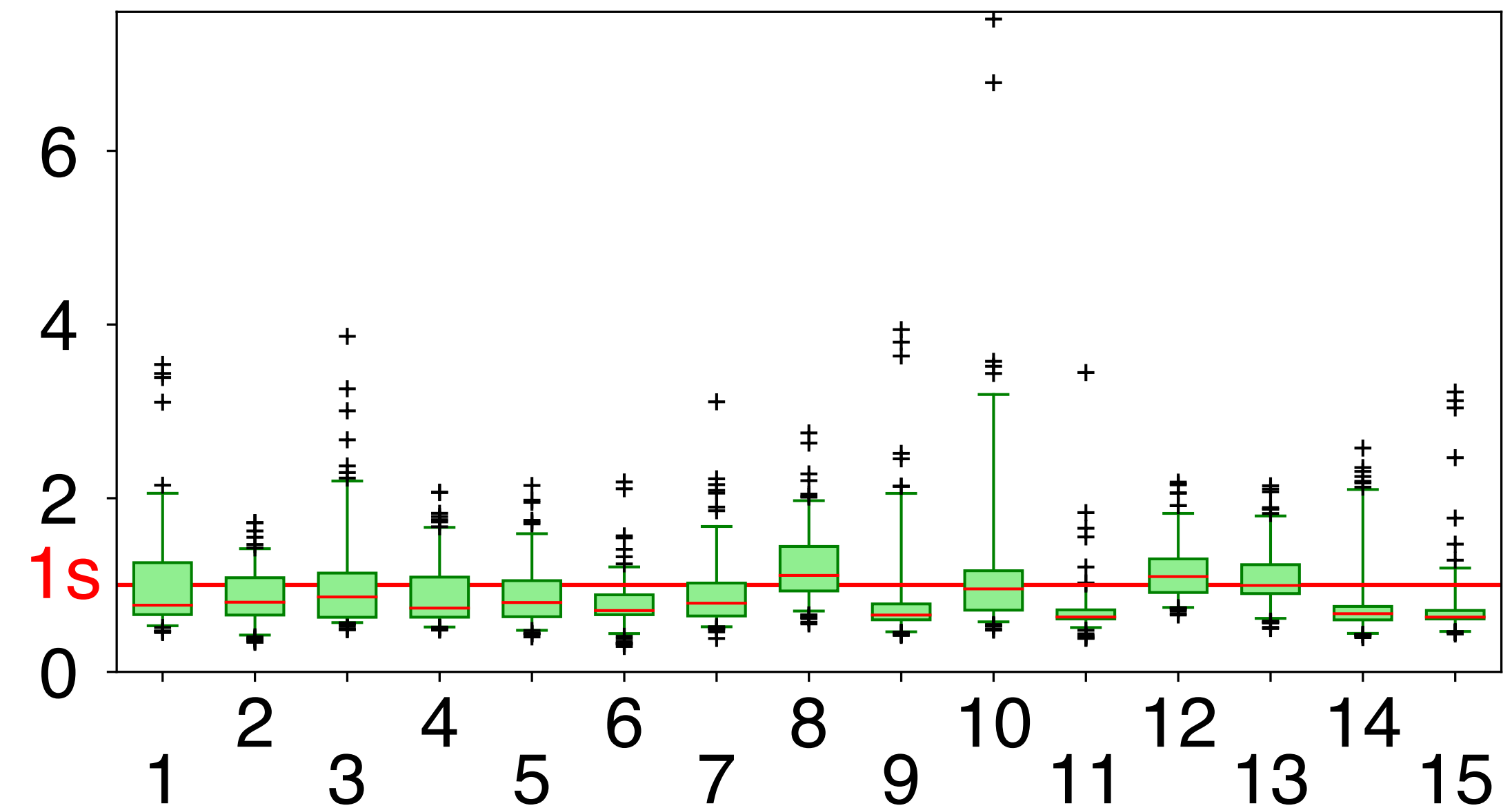


Blink works well in practice,
with an inference speed below 1 second in most cases



Blink works well in practice,
with an inference speed below 1 second in most cases

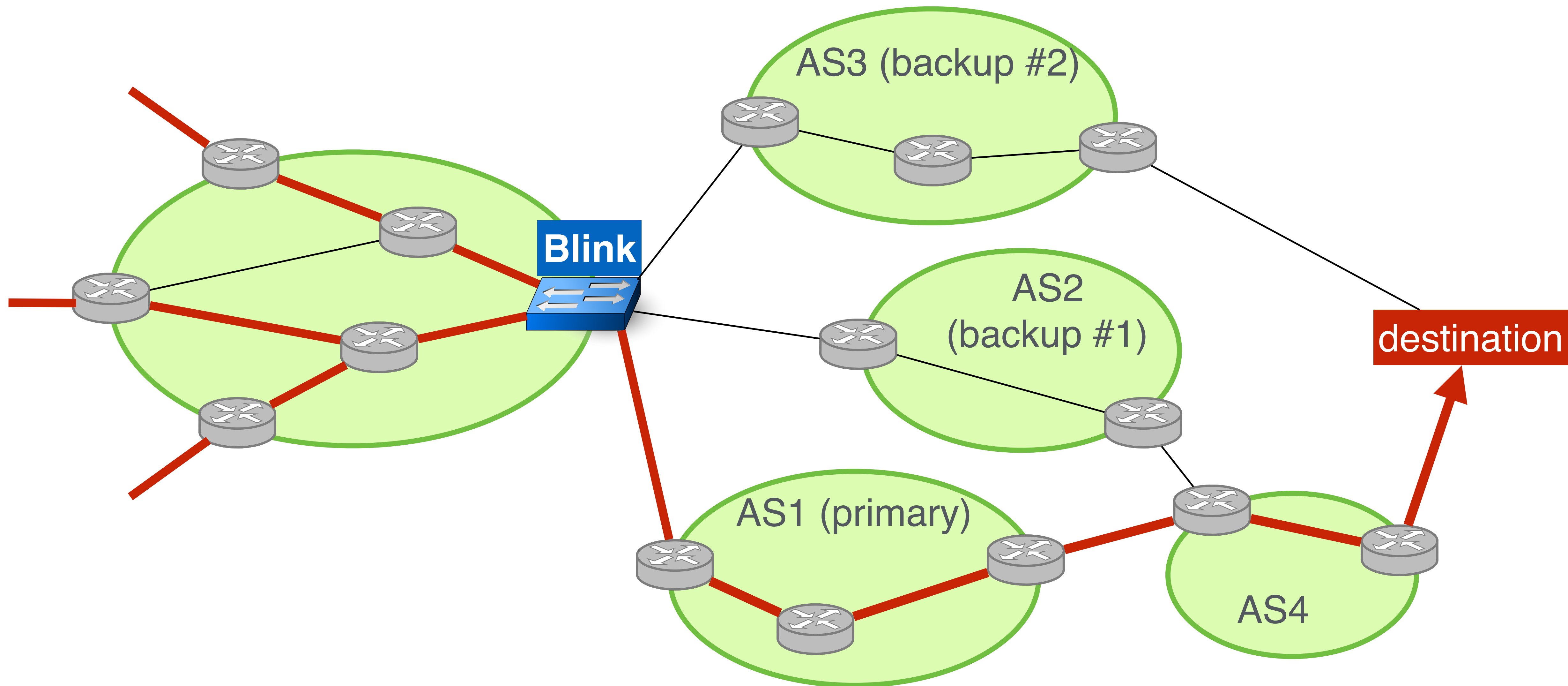
Inference speed (s)

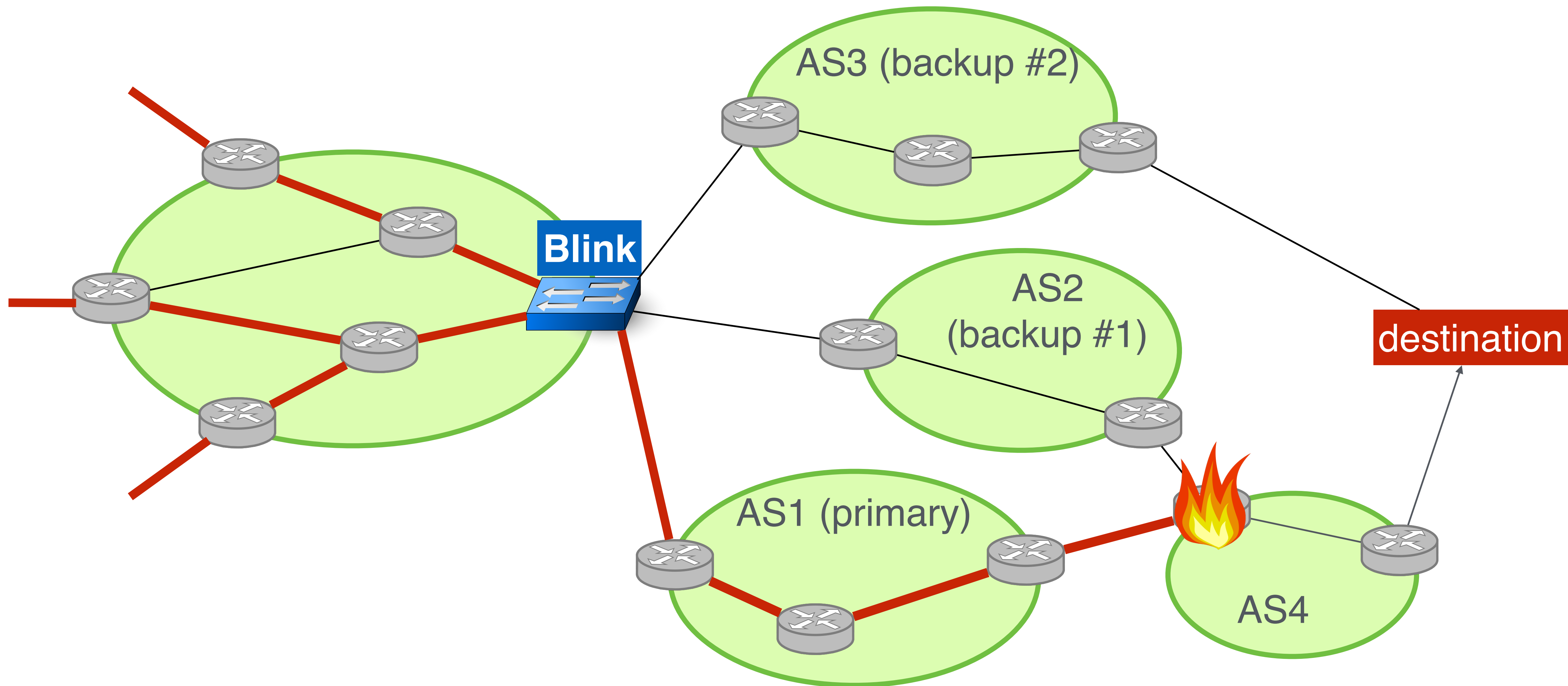


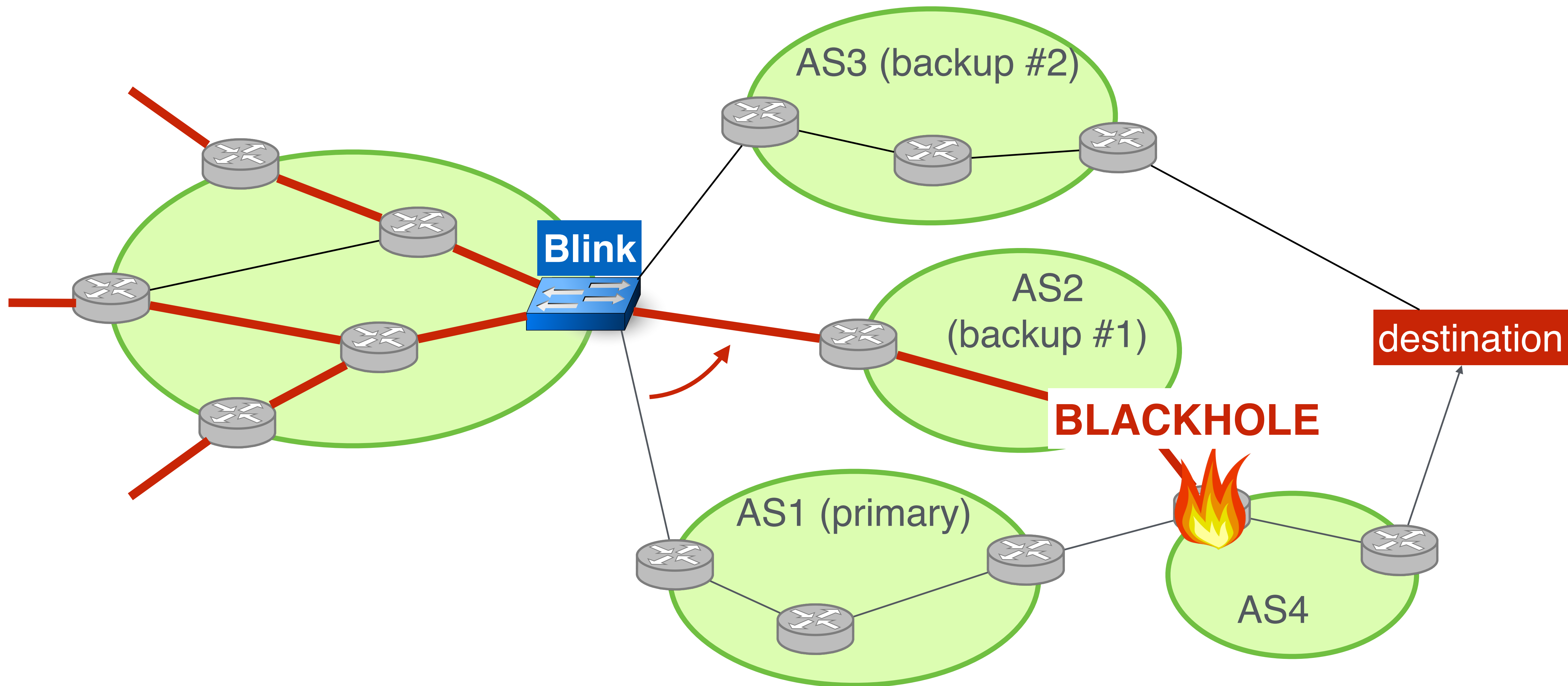
Failure scenarios

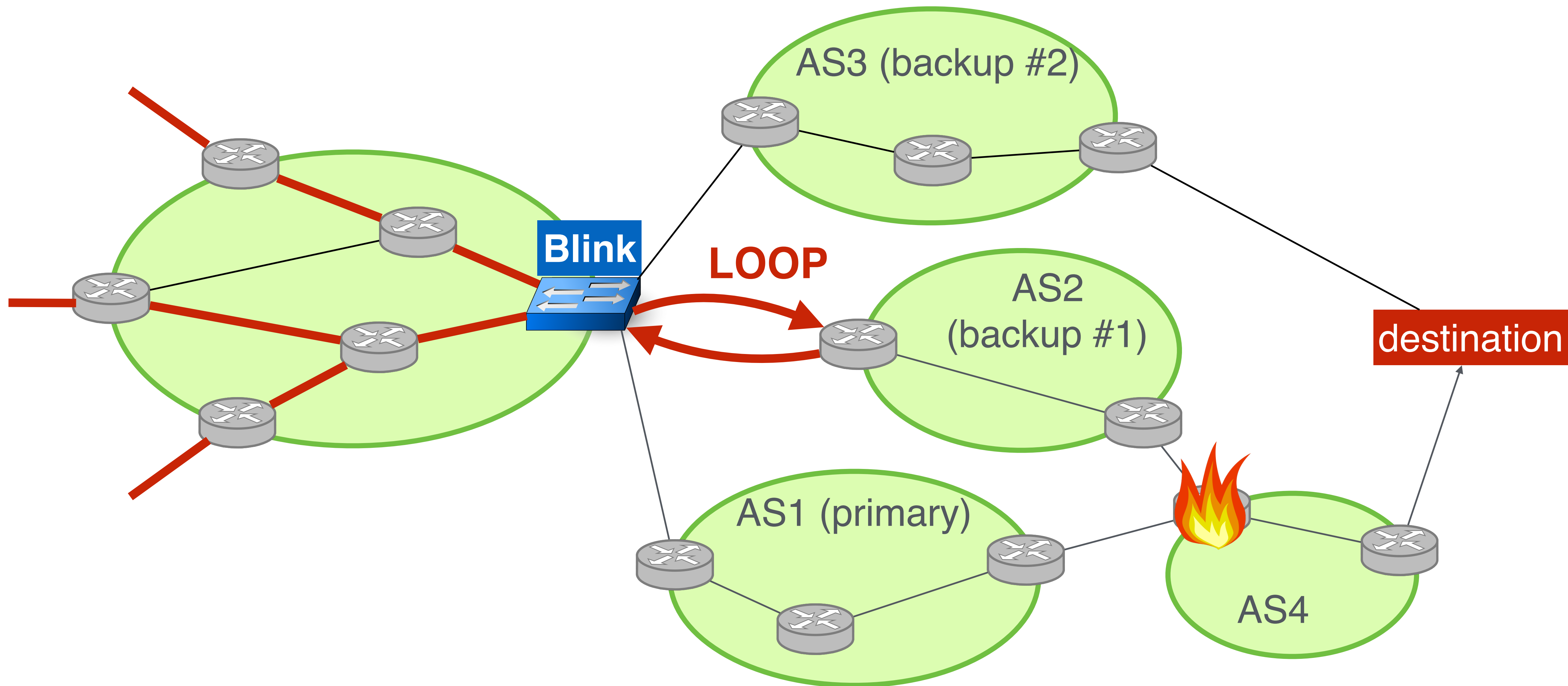
What about actuation?

Where do we send the traffic upon detecting a failure?

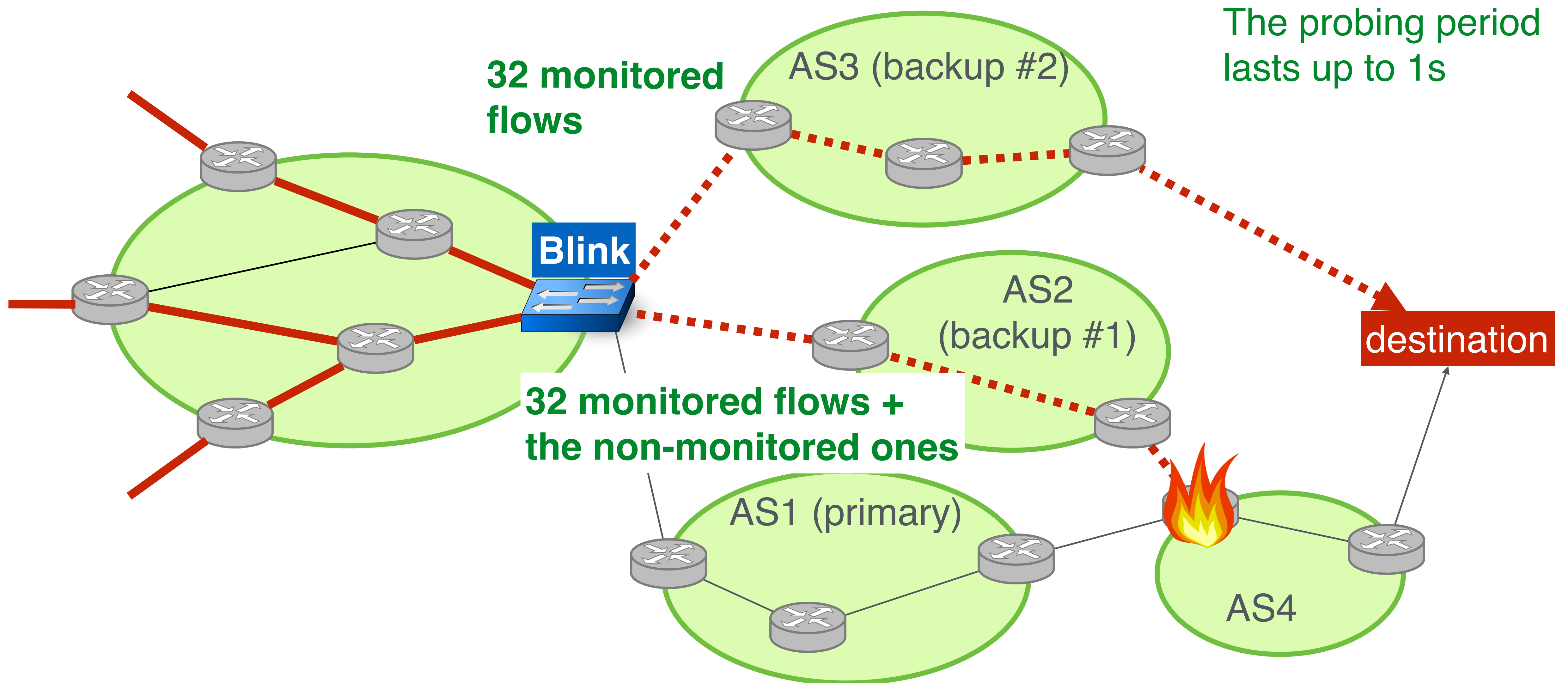


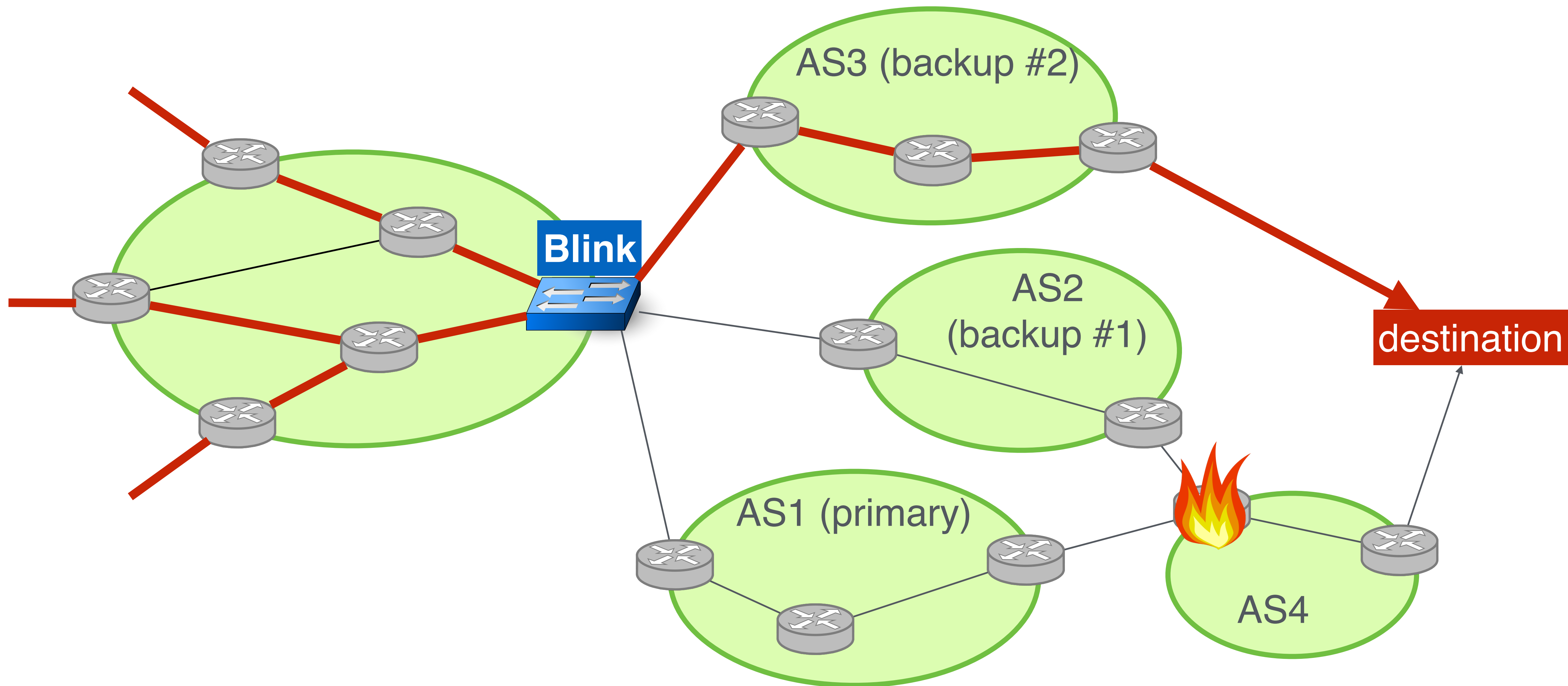






As for failures, Blink relies on data plane signals to pick a working backup path





We intend to generalize the signals we track,
for a wider variety of applications

reliability

reroute around fine-grained failures

performance

steer traffic along the most performant path

security

detect unwanted traffic redirection

Part II: Taking control

applications

frameworks

methods

monitoring/inference

Adaptive Network Monitoring



Alexander Dietmüller



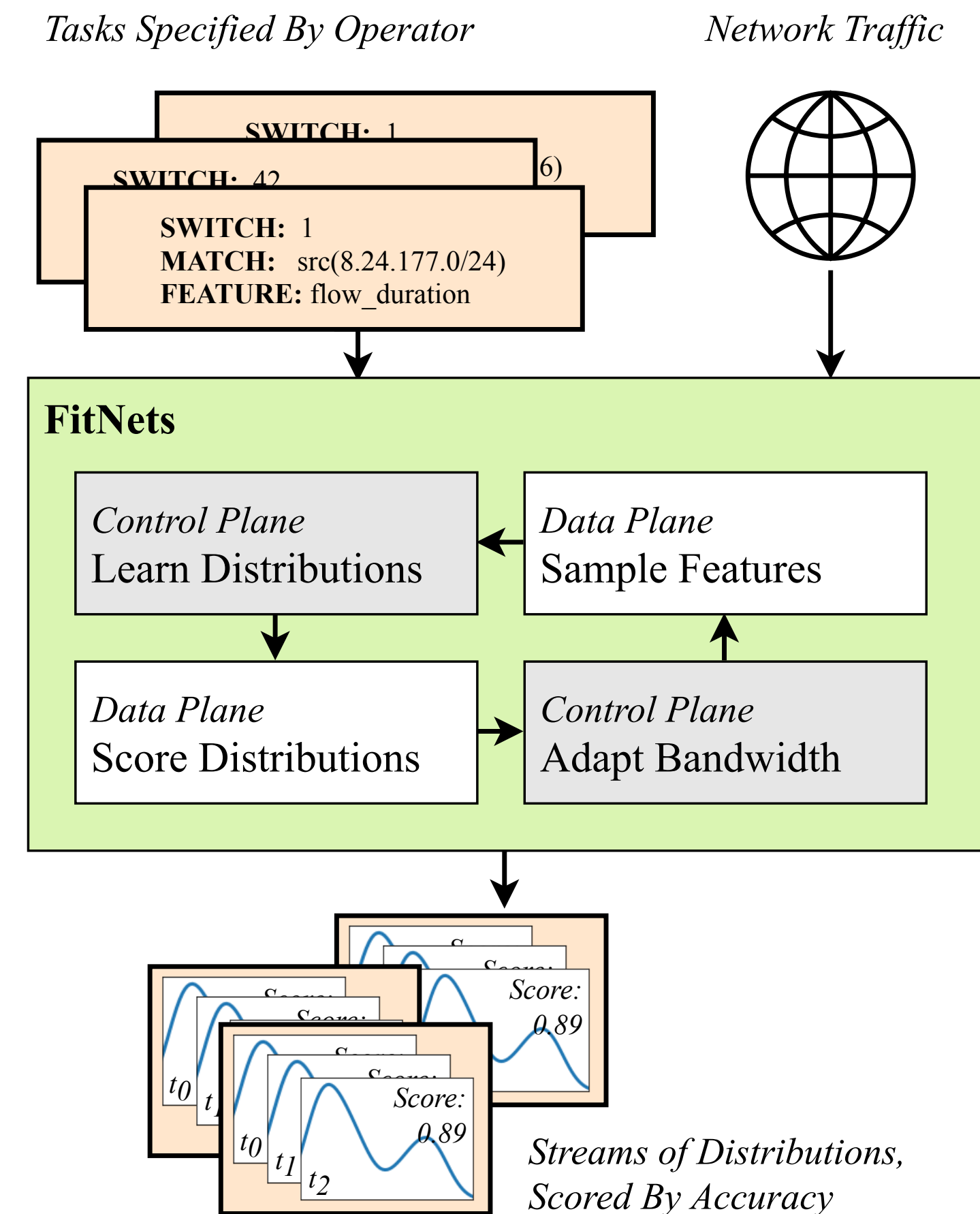
Albert
Gran Alcoz



Laurent Vanbever

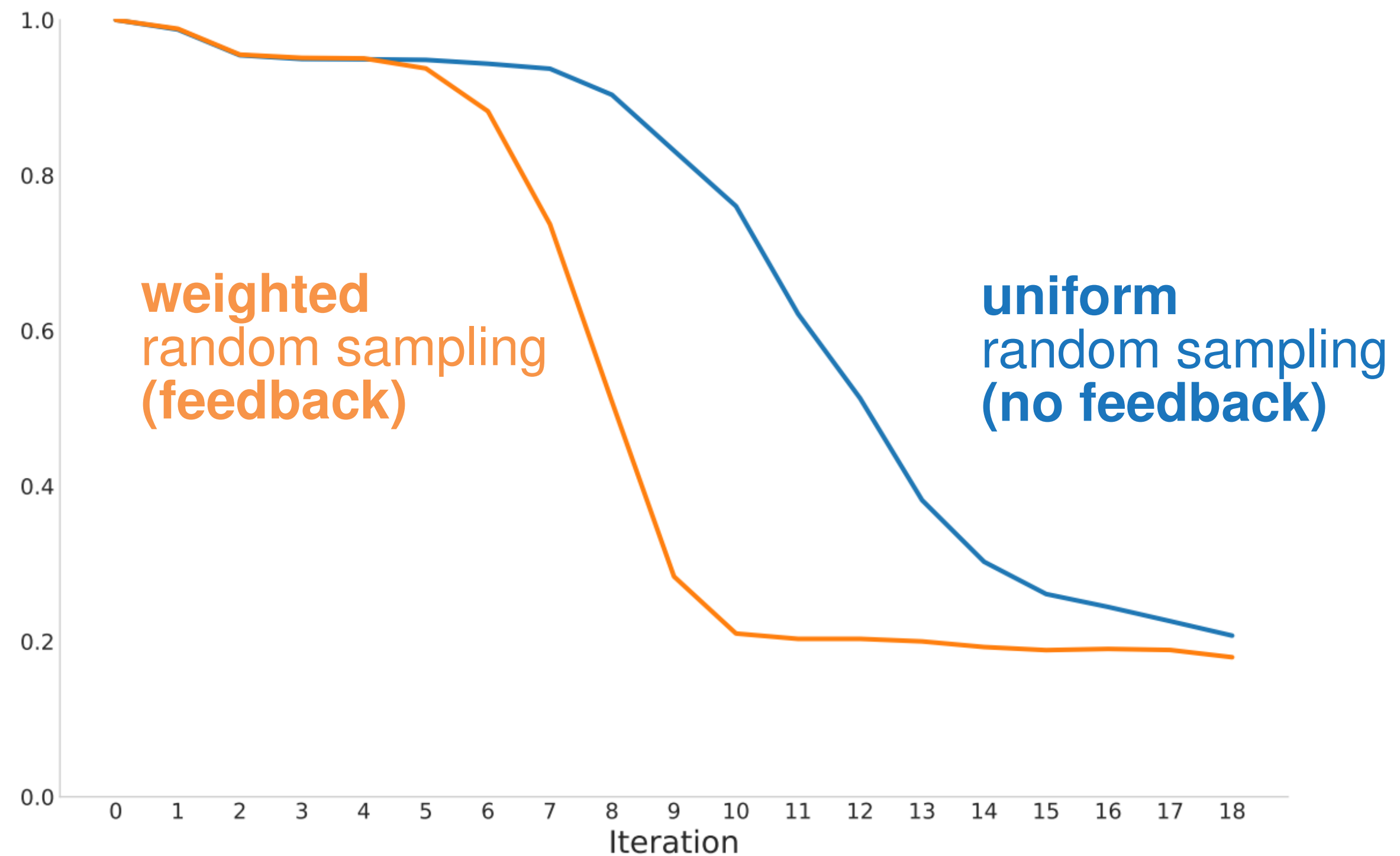
We're building an adaptive monitoring framework that can learn high-quality traffic distributions

- computes traffic distribution in the control plane, on traffic samples
- scores these distributions in the data plane, on *all* traffic
- adapts the sampling rate according to the score



Our framework enables to learn distribution better and faster than simply randomly sampling traffic

Distance from
true distribution
(Jensen-Shannon)



Part II: Taking control

applications

frameworks

methods

monitoring/**inference**

In-network inference



Coralie Busse-Grawitz



Roland Meier



Alexander Dietmüller



Tobias Bühler



Laurent Vanbever

Why (and How) Networks Should Run Themselves

Nick Feamster and Jennifer Rexford
Princeton University

Abstract

The proliferation of networked devices, systems, and applications that we depend on every day makes managing networks more important than ever. The increasing security, availability, and performance demands of these applications suggest that these increasingly difficult network management

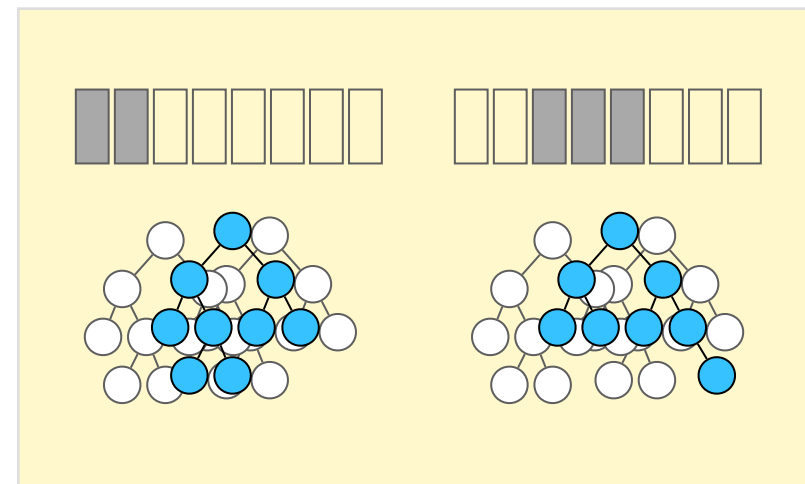
relationships between them and user quality of experience become increasingly complex. Twenty years ago, we had some hope of (and success in) creating clean, closed-form models of individual protocols, applications, and systems [4, 24]; today, many of these are too complicated for closed-form analysis. Prediction problems such as determining how search query response time would vary in response to the placement

Performing automated, real-time inference: The past ten years has demonstrated significant promise in using machine learning to both detect and predict network attacks; we must build on the increasing amount of work in automated infer-

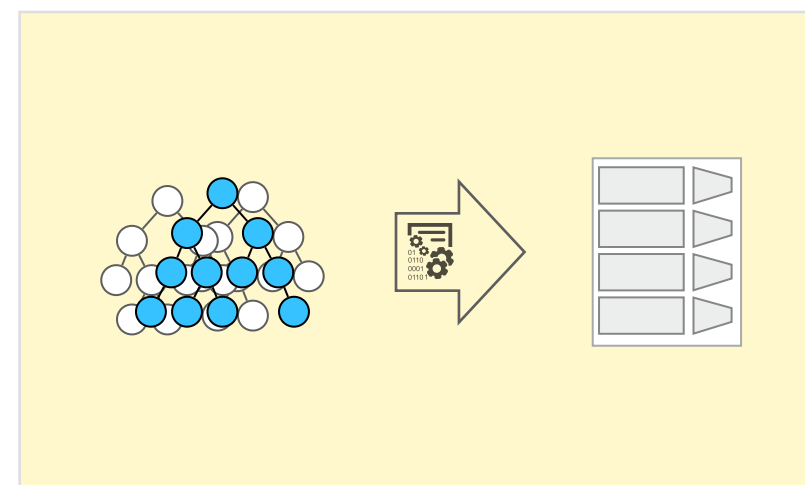
Are programmable network devices powerful enough for machine learning inference?

Yep!

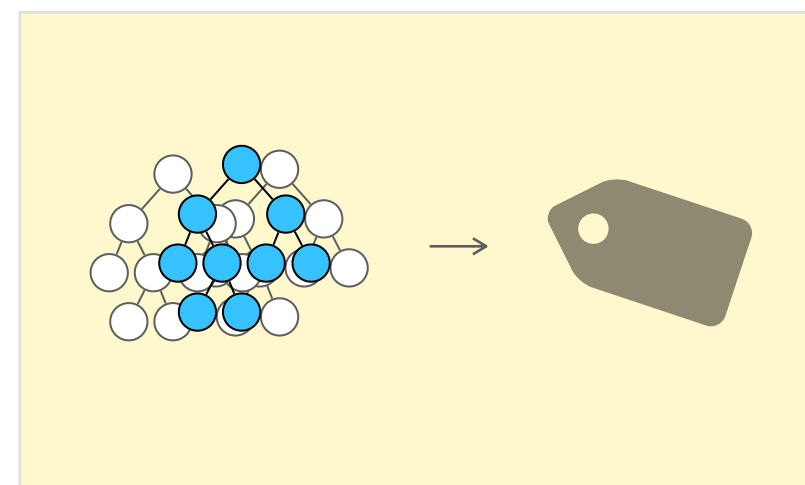
We're building a real-time, in-network inference framework



Optimizing random forest models
for programmable network devices



Compiling random forest models
to programmable network devices



Performing runtime classification
at Tbps

Early, accurate & efficient classification of an ongoing flow
as **an optimization problem**

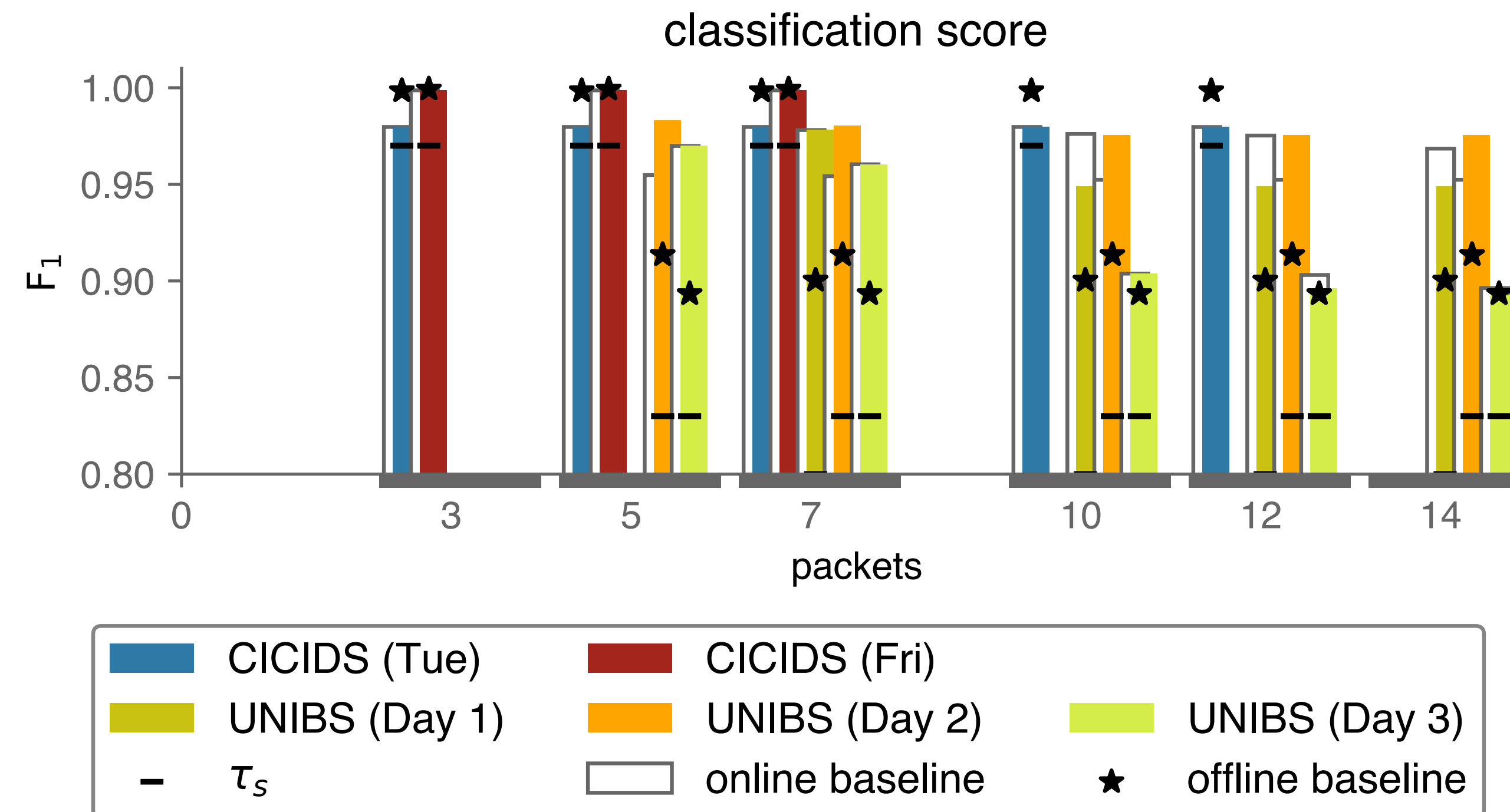
Given a labeled dataset $\mathcal{F}$ and a threshold τ
find a classifier C such that

- $\text{accuracy}(C) \geq \tau$
- C fits in programmable network devices

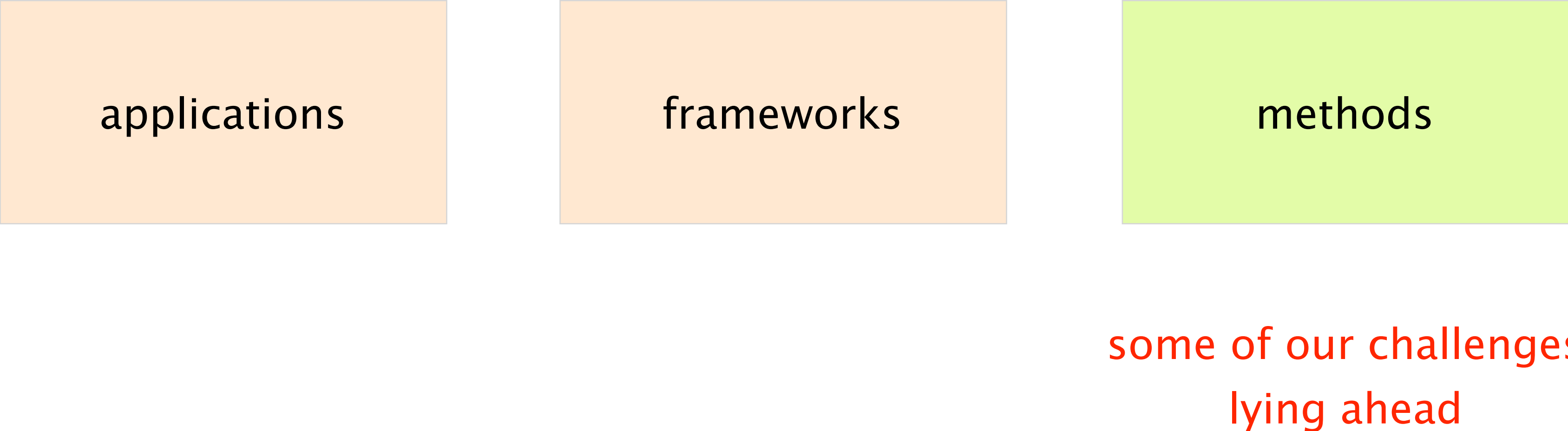
while

- minimizing memory usage
- maximizing classification speed

Our prototype is already able to classify traffic after few packets, with an accuracy which is on-par with software-based solutions



Part II: Taking control



applications

frameworks

methods

some of our challenges
lying ahead

Building and operating truly self-driving networks
require us to overcome fundamental challenges

data

correctness

interpretability

data

correctness

interpretability



data is network-dependent

poor generalization capabilities

interesting events are rare

few interesting examples (if any)

data can easily be polluted

by anyone with a Internet connection...

data

correctness

interpretability

how do we...

ensure correctness?

network should be reachable

guarantee stability?

agents can clash with each other

reason about optimality?

data-driven > non-data-driven (?)

data

correctness

interpretability

how can operators...

reason about self-driving networks?

especially in partial deployment

debug their self-driving networks?

manual override

Self-driving/monitoring networks

in the age of deep network programmability

1 operator assistance

2 partial automation

3 conditional automation

4 high automation

5 full automation

Part I

assisting operators

Part II

taking control

too futuristic? 😊

Self-driving/monitoring networks

in the age of deep network programmability



Laurent Vanbever

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TMA

June 19 2019