



#### **Network Security and Measurement**

## - Network Tomography -

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#### Agenda

Introduction to Network Tomography

Binary Network Tomography

Pinpointing Router Behavior in the Wild



# Looking into the inner Core **TOMOGRAPHY**



#### Problem

Most measurements are performed from endpoints at the network edge

- Which observations are caused by the network core?
- What are underlying characteristics of the network core?

In many cases, it is impossible to measure the Internet core directly



#### Problem

Can we study the internal characteristics of a network using only information visible at its edge? Most measurements are performed from endpoints at the network edge

- Which observations are caused by the network core?
- What are underlying characteristics of the network core?

In many cases, it is impossible to measure the Internet core directly



#### The Concept of Network Tomography

Coined by Vardi in 1996

- Model the properties of your network
- Take appropriate measurements at many endpoints
- Correlate the measurements
- Invert the problem with the help of your model
- Infer properties of the internal network

Y. Vardi: *Network Tomography: Estimating Source-Destination Traffic Intensities from Link Data* Journal of the American Statistical Association, Vol. 91, No. 433 (March, 1996), pp. 365-377



#### **Internet Tomography**

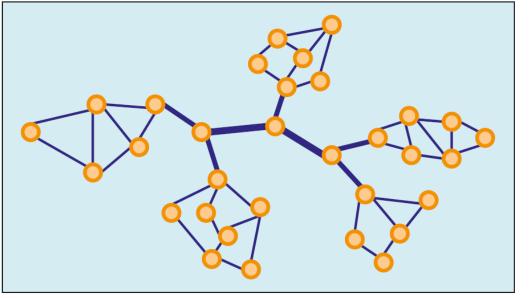
Many paths share links, why measurements correlate

Tomography problem:

 $y = A \cdot \theta + \epsilon$ , with

A the routing matrix,  $\theta$  network parameters, and  $\epsilon$  random noise.

*y* are the measured observables. This linear stochastic model needs inversion to infer inner link properties.



M. Coates, et al: Internet Tomography, IEEE Signal Processing Magazine. **19** (3): 47–65, 2002



#### **Examples**

**Congested Link** – is reduced available bandwidth due to congested link?  $\rightarrow \theta$  is the vector of traffic intensity per link

**Lossy Link** – is packet loss due to a link?  $\rightarrow \theta$  is the vector of link success probabilities

**Packet delays** – which links produce large delays?  $\rightarrow \theta$  is the vector of link delays



# Reducing the Problem Space BINARY NETWORK TOMOGRAPHY



#### Limitations of General Network Tomography

Modeling space can be very large. Network conditions vary – models often need time-dependence.

Interpretation and Inversion often require complex mathematical models, e.g., about queuing.

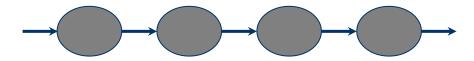
Measurements often too sparse or too inconsistent to grant full insights.

A simplified approach promises more success.



#### The basics of Binary Network Tomography

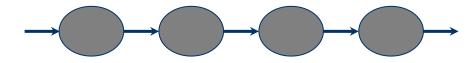
If a path does not have property A, then **none** of the nodes have property A.



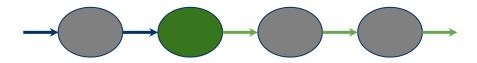


#### The basics of Binary Network Tomography

If a path does not have property A, then **none** of the nodes has property A.



If a path has property A, then at least one of the nodes has property A.





#### The Benefit of Binary

Observation of property A on a path is binary.

• We see only two types of paths

For each transition node *i*, we can model this with a variable

 $x_i = \begin{cases} 0 \text{ if node } i \text{ has property A,} \\ 1 \text{ if node } i \text{ does not have property A.} \end{cases}$ 

Then the property of a path *j* of nodes  $N_J$ can be expressed as  $y_j = \prod_{i \in N_i} x_i$ 



#### **Resolving Binary Network Tomography**

With just one path of property A, we cannot decide which node is responsible for the observation.

$$\rightarrow ? \rightarrow ? \rightarrow ? \rightarrow ? \rightarrow$$

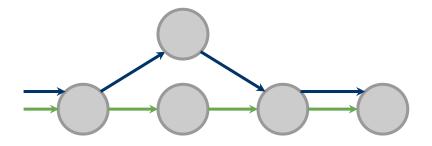


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But with multiple paths we can solve a simple question.



Which node is only on the green path?

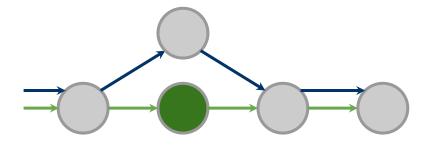


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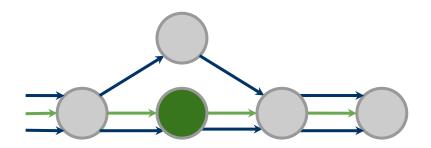


Which node is only on the green path?



#### **Measurements in the wild. More challenges.**

#### Measurements introduce noise.

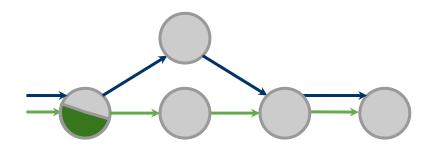




#### Measurements in the wild. More challenges.

Measurements introduce noise.

Nodes behave inconsistently.

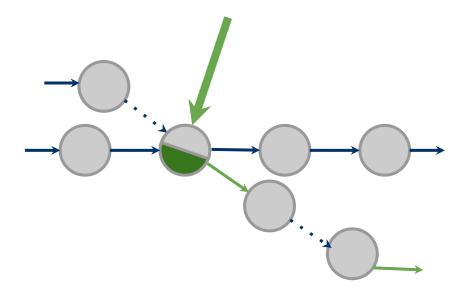




#### BeCAUSe –

#### **Bayesian Computation for AUtonomous System**

## **Pinpoint ASes based on path information.**





### Approach: BeCAUSe

**Bayesian Computation for AUtonomous System** 

Instead of framing pinpointing as a binary network tomography problem we consider a **probabilistic setting**.



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Each AS *i* has a **probability value** (*pi*) of implementing A.



### Approach: BeCAUSe Bayesian Computation for AUtonomous System

Instead of framing pinpointing as a binary network tomography problem we consider a **probabilistic setting**.

Each AS *i* has a **probability value** ( $p_i$ ) of implementing A.

Likelihood of a single path:

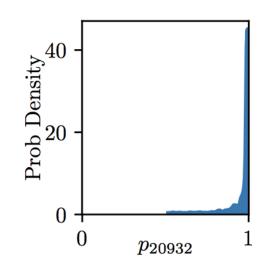
$$\mathbf{P}(\text{path}|\mathbf{p}) = \begin{cases} \prod_{i \in \text{path}} (1 - p_i), & \text{if path does not show A,} \\ 1 - \prod_{i \in \text{path}} (1 - p_i), & \text{if path shows A.} \end{cases}$$



#### **Bayesian inference of all paths**

## **Posterior distribution** of A given the observed data set *D*

# $$\begin{split} P(p|D) \propto P(D|p) \cdot P(p) \\ \text{Likelihood Prior} \end{split}$$



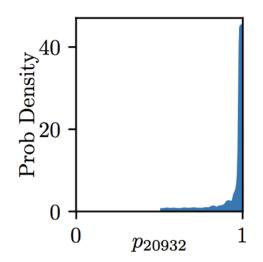


#### **Bayesian inference of all paths**

**Posterior distribution** of A given the observed data set *D* 

## $$\begin{split} \mathbf{P}(\mathbf{p}|D) \propto \mathbf{P}(D|\mathbf{p}) \cdot \mathbf{P}(\mathbf{p}) \\ \text{Likelihood Prior} \end{split}$$

We take samples of the LHS using **Metropolis-Hastings** and **Hamiltonian Monte Carlo** to get a **probability distribution** for each AS.



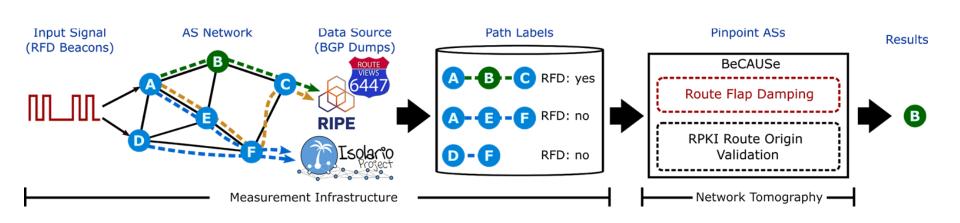
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## Applying BeCAUSe **PINPOINTING ROUTER BEHAVIOR IN THE WILD**



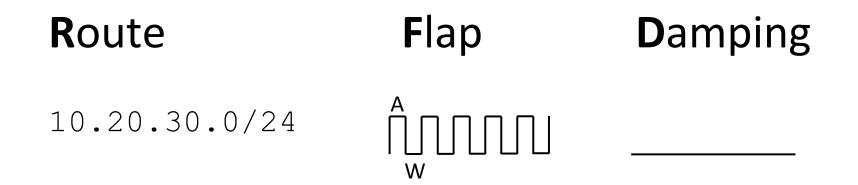


#### **Application Examples**





#### What is BGP Route Flap Damping?





#### Why should you care about RFD deployment?

RFD **impacts** passive and active BGP **measurements**. Current results on BGP Update noise may be an underestimation.

Deprecated default **parameters affect Internet reachability.** Especially in today's rich topology.

Many **different recommendations** in the past two decades. Different configurations may lead to conflicting goals.



#### Why should you care about RFD deployment?

#### RFD impacts passive and active BGP measurements.

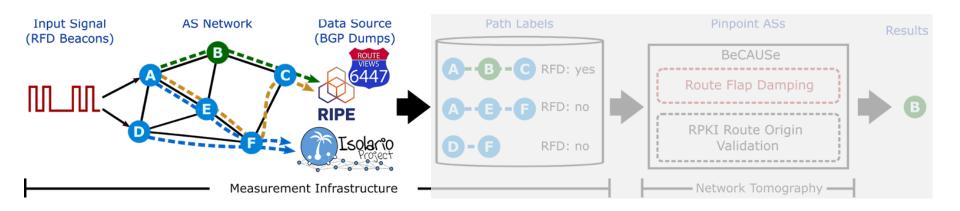
Current reculte on RCD Undate noise may be an

# No measurements of BGP Route Flap Damping deployment.

Different configurations may lead to conflicting goals.

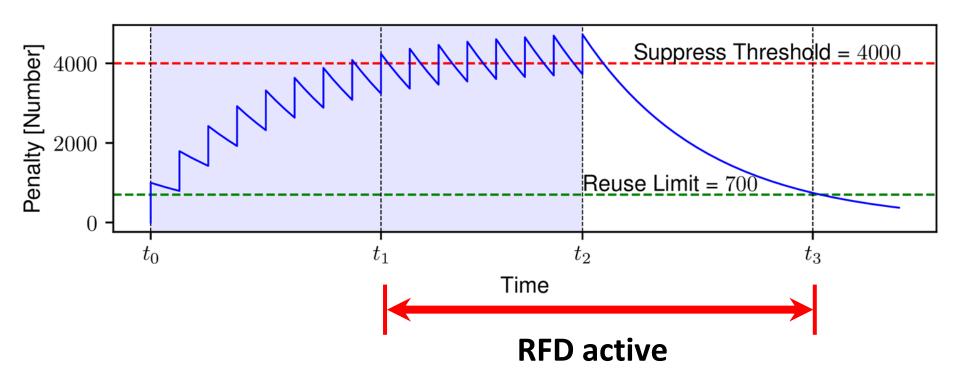


#### Let us measure deployment of RFD. First we need path data.

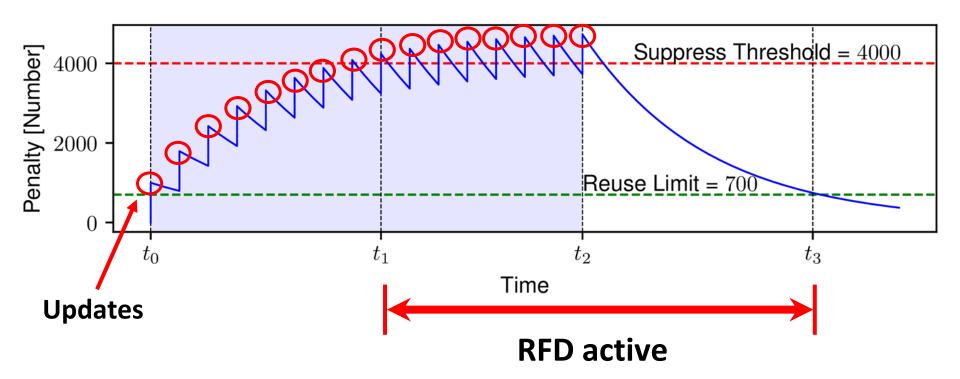




#### How does Route Flap Damping work?

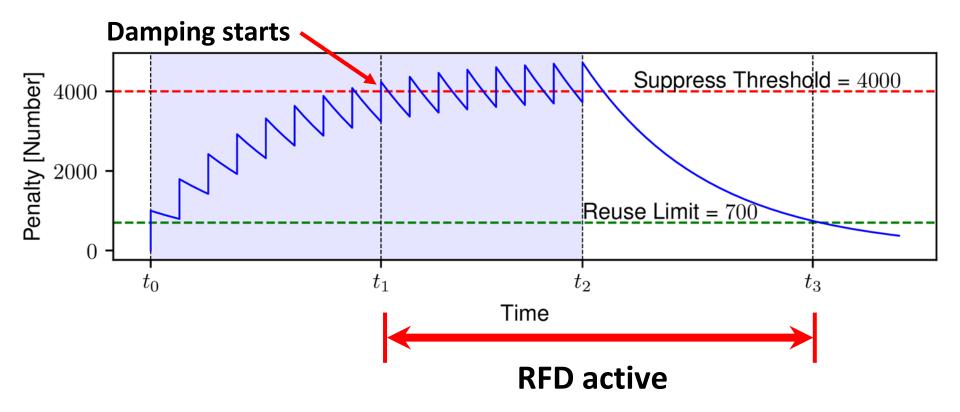






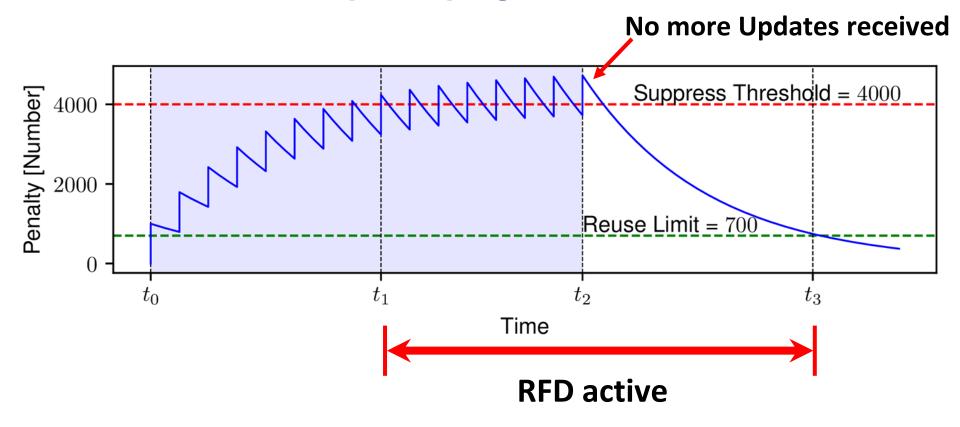


#### How does Route Flap Damping work? Start.





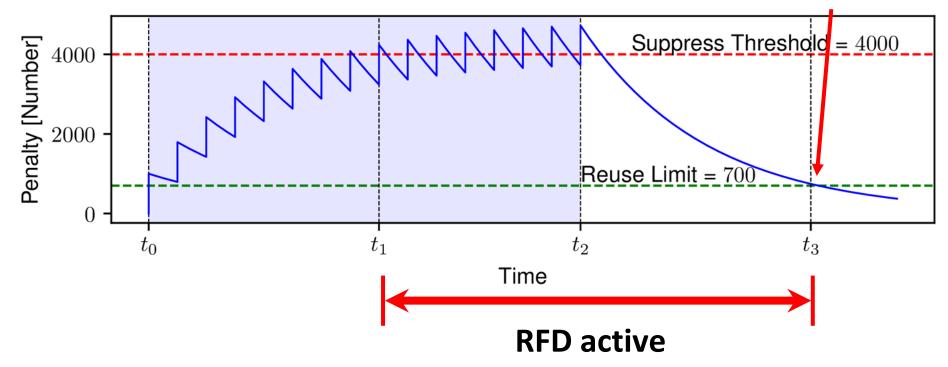
#### How does Route Flap Damping work? Wait.





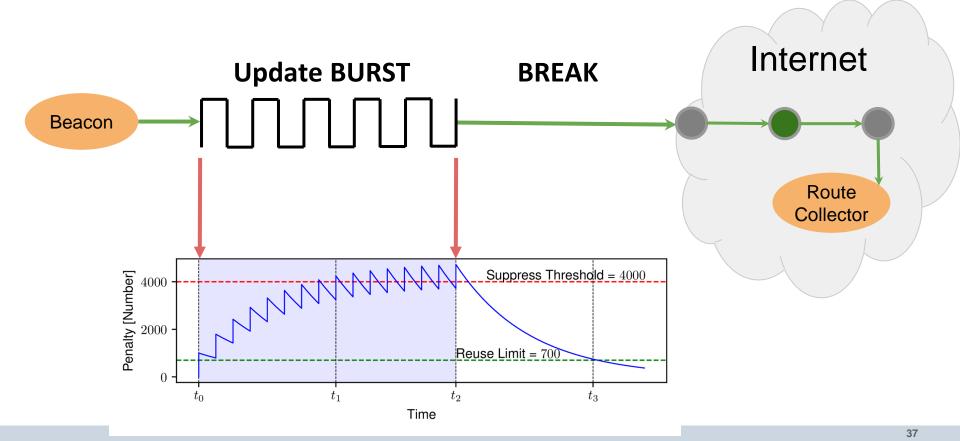
#### How does Route Flap Damping work? Release.

#### **Damping stops**



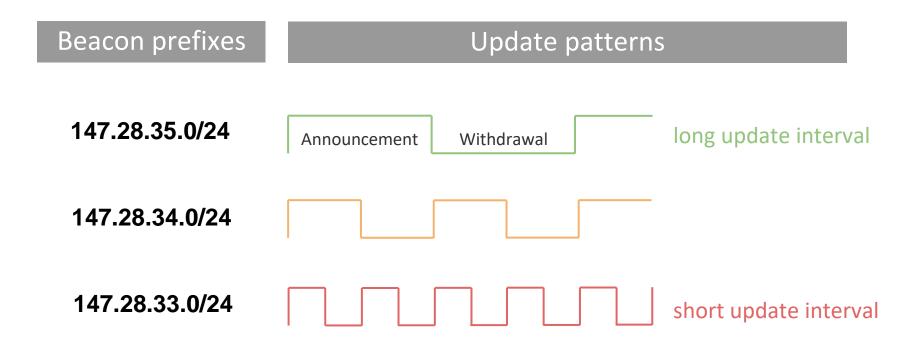


#### **Generating the RFD update signature**





#### **BGP Beacons with different frequencies**





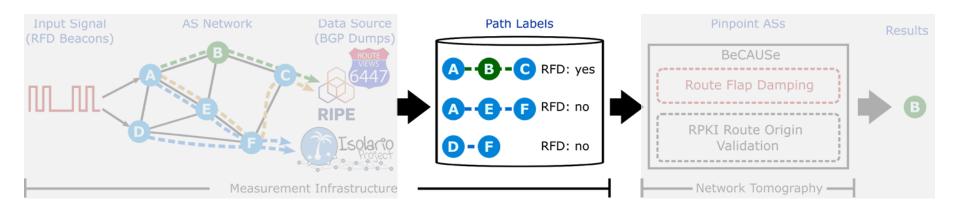
#### **Locations of our BGP Beacons**

Bangkok, TH Johannesburg, ZA København, DK München, DE São Paulo, BR Seattle, US Tokyo, JP

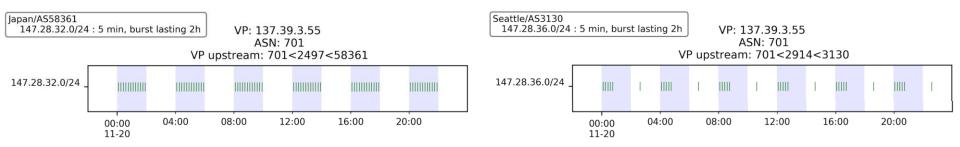




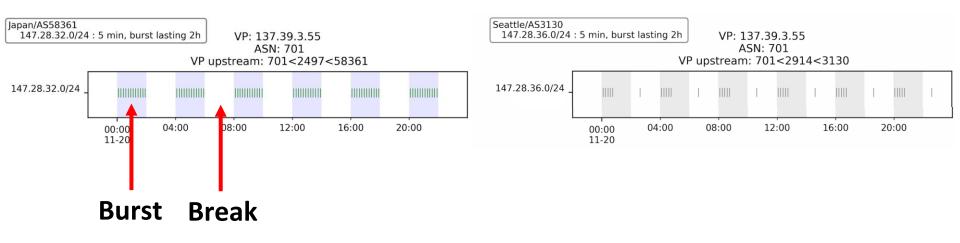
# RFD causes a very recognizable pattern.



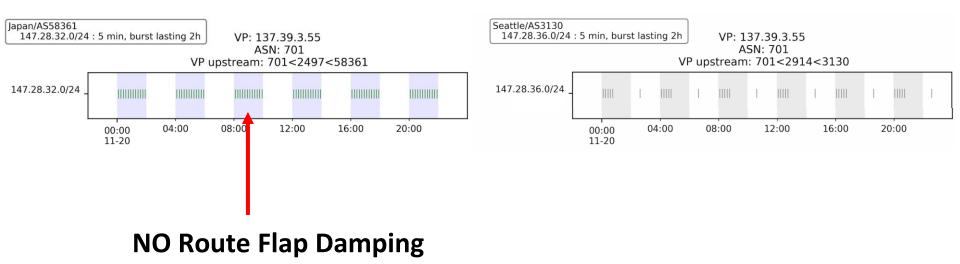




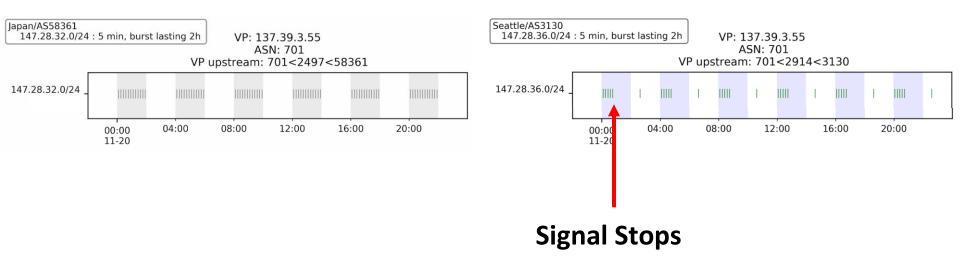




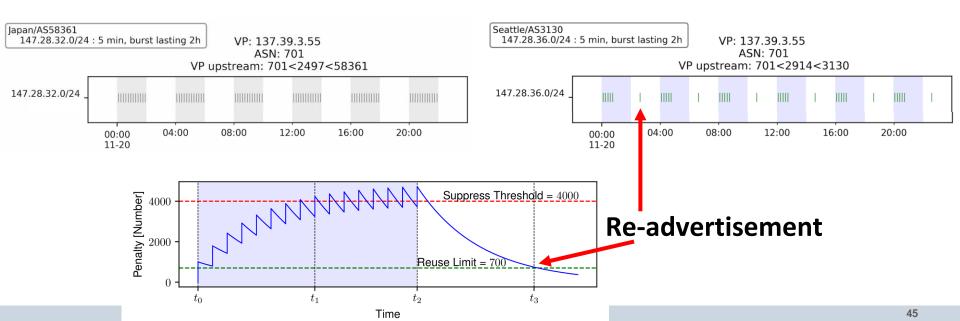






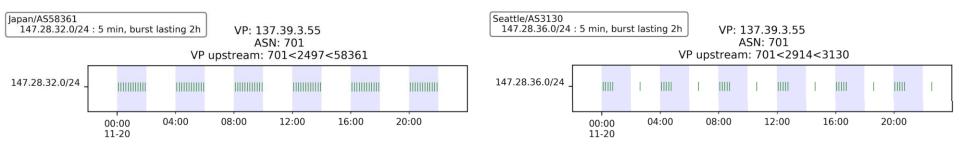








#### Our prefixes are damped during the Burst (Blue) and re-advertised during the Break (White).

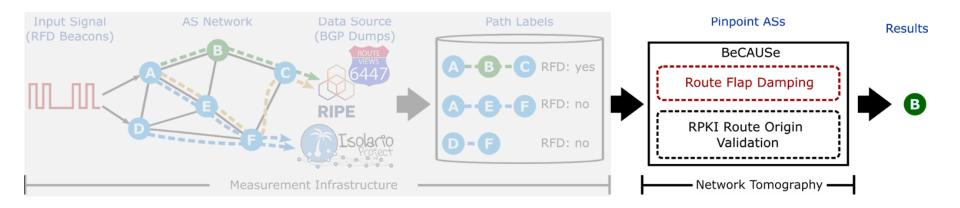


NO Route Flap Damping

**Route Flap Damping** 



# Pinpointing ASs based on path information.





#### **Challenges when pinpointing RFD ASs**

1) If we find the Route Flap Damping pattern at some vantage points, **the damper could be anywhere on the AS path**.

- 2) Some ASs use **Route Flap Damping selectively** on a subset of neighbors.
- **3) Noise** inherent in the measurement cannot result in a binary solution.



### **Applying BeCAUSe**

Measure RFD on all path available via Vantage Points.

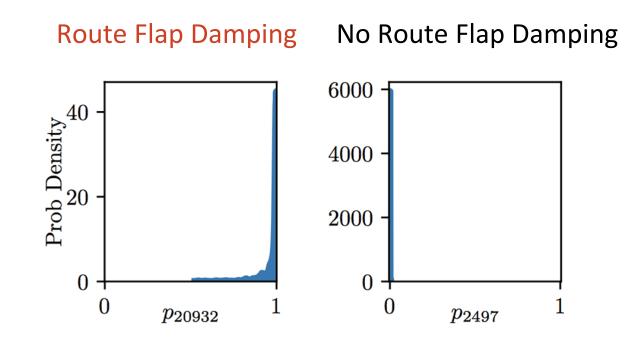
Compute RFD probabilities for paths.

Perform Monte Carlo sampling to get obtain a distribution for each AS.

Evaluate and classify results.

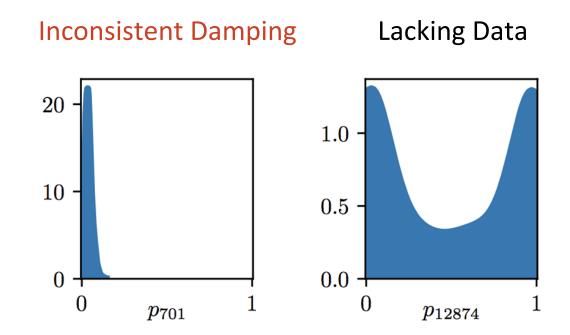


### Output distribution types Low Variance / High Certainty



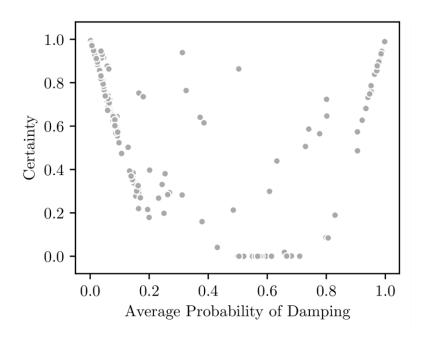


#### Output distribution types Special cases





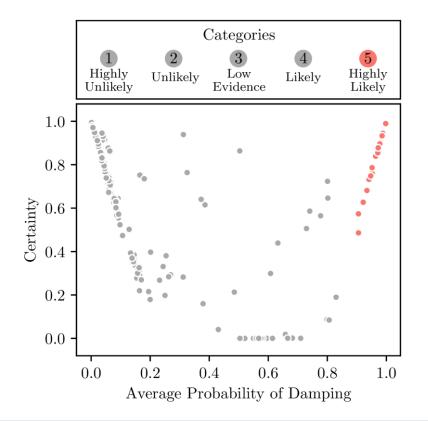
#### **Distribution summary**



Goal: **Hard decisions** on RFD deployment.



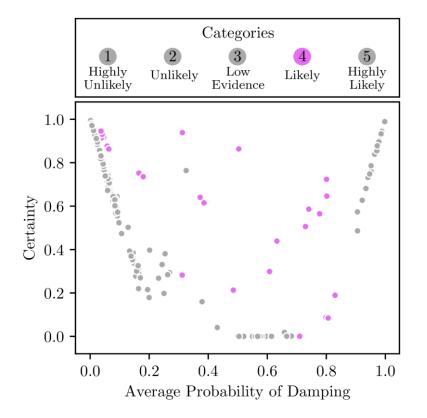
#### **Distribution summary - Damping ASs**



High damping probability combined with high certainty are labeled damping ASs.



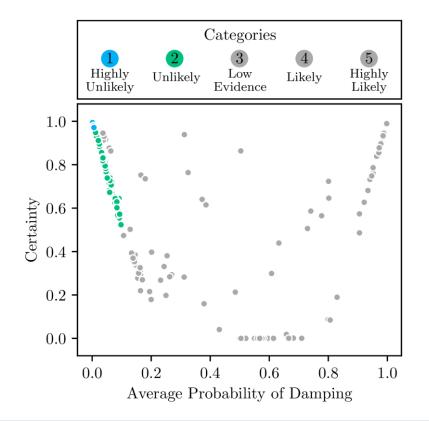
#### **Distribution summary - Inconsistently damping ASs**



For **RFD labeled paths** without nodes labeled as damping, we find and label the node that is **most likely** causing RFD.



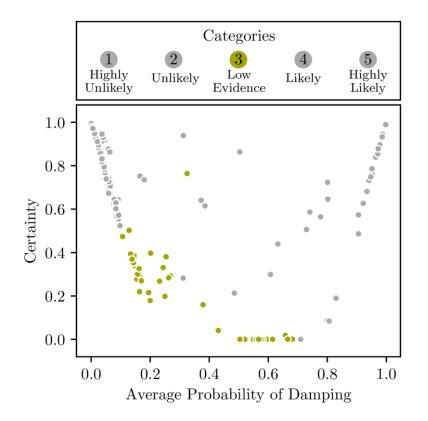
#### **Distribution summary - Non-damping ASs**



Low damping probability combined with high certainty are labeled non-damping ASs.



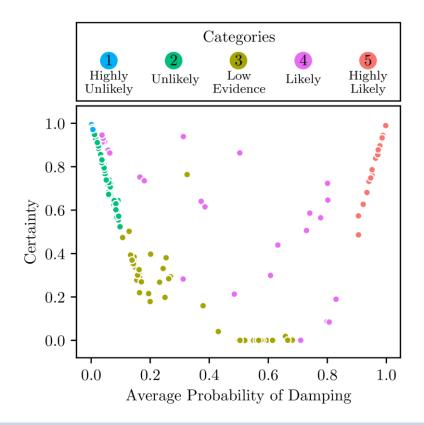
#### **Distribution summary - Lacking evidence**



# For the **remaining ASs** we cannot draw conclusions about RFD deployment.



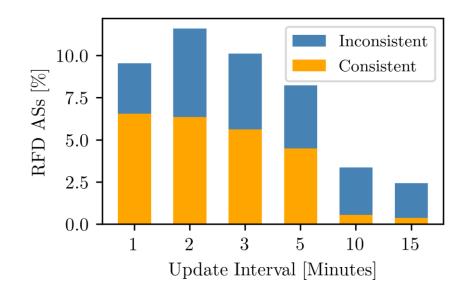
#### **Summarised distributions**



				RFD	
Category	1	2	3	4	5
Total	166	283	72	25	28
Share	28.9%	49.3%	12.5%	4.3%	4.8%



#### **Route Flap Damping deployment**

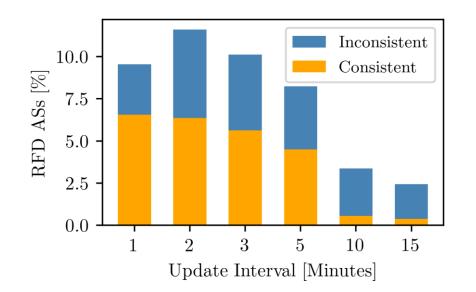


**9%** is the **lower bound** of RFD deployment due to lacking visibility.

Verified by 75 ASs BeCAUSe has **100% precision** and **87% recall.** 



#### Which RFD parameters are deployed?



Routers with **RFD default** params start damping at the 5 minute update interval and lower.

Most ASs use **deprecated vendor default configurations**.



#### **Using BeCAUSe for ROV detection**

### **Route Origin Validation:** RPKI-based filtering of invalid prefix origins, as increasingly deployed on the Internet.



#### **Using BeCAUSe for ROV detection**

**Route Origin Validation:** RPKI-based filtering of invalid prefix origins, as increasingly deployed on the Internet.

**ROV** is a binary property that can be measured in similar controlled experiments. Ground truth exists from a previous operator study.



#### **Using BeCAUSe for ROV detection**

**Specific to research question:** summarisation techniques of probability and certainty results.

**ROV** is a binary property that can be measured in similar controlled experiments. Ground truth exists from a previous operator study.

In a separate simulation, we used **BeCAUSe to pinpoint ROV** deployment. We achieved identical precision, but lower recall due to high ROV usage.



#### Literature

C.Gray, C. Mosig, R. Bush, C. Pelsser, M. Roughan, T.C. Schmidt, M. Wählisch (2020).

BGP Beacons, Network Tomography, and Bayesian Computation to Locate Route Flap Damping. ACM Internet Measurement Conference

(IMC), p. 492–505, ACM : New York, 2020.

#### BGP Beacons, Network Tomography, and Bayesian Computation to Locate Route Flap Damping

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ABSTRACT

Pinpointing autonomous systems which deploy specific inter-domain techniques such as Route Flap Damping (RFD) or Route Origin Validation (ROV) remains a challenge today. Previous approaches to detect per-AS behavior often relied on heuristics derived from passive and active measurements. Those heuristics, however, often lacked accuracy or imposed tight restrictions on the measurement methods.

We introduce an algorithmic framework for network tomography, BeCAUSe, which implements Bayesian Computation for Autonomous Systems. Using our original combination of active probing and stochastic simulation, we present the first study to expose the deployment of RFD. In contrast to the expectation of the Internet community, we find that at least 9% of measured ASs enable RFD, most using deprecated vendor default configuration parameters. To illustrate the power of computational Bayesian methods we compare BeCAUSe with three RFD heuristics. Thereafter we successfully apply a generalization of the Bayesian method to a second challenge, measuring deployment of ROV.

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#### **1 INTRODUCTION**

In the mid '90s, many global backbone BGP-speaking routers were under-powered and began to experience damaging CPU load in the presence of BGP churn, frequent announcements and withdrawals of the same prefix. Some core operators met with vendors to design Route Flap Damping (RFD) and codified it in RFC 2439 [43]. With RFD, routers maintain a penalty value per prefix per session. Prefixes with a penalty above a given threshold are damped, e.g., newly received announcements are suppressed and not considered as suitable alternatives to reach a destination.

In 2002-2003, it was shown by Mao et al. [24] that RFD was too aggressive and had a negative affect on Internet routing. Routers in 2006 were more powerful so it was presumed that operators followed best practice and removed RFD from their configurations [5] In 2011 Pelsser et al. [30] showed that more considered settings.