### Blink: Fast Connectivity Recovery Entirely in the Data Plane



#### Joint work with

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NSDI 26th February 2019

https://blink.ethz.ch

# Fire at AT&T facility causes widespread outage in North Texas

# Time Warner Cable comes back from nationwide Internet outage by Brian Stelter @brianstelter



November 6, 2017 | Emerging Threats





## Your network



## Your network





### Remote

# Remote

Remote

### Your network



Remote

Remote

Local



### Upon local failures, connectivity can be quickly restored

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Fast failure detection using *e.g.*, hardware-generated signals

Fast traffic rerouting using *e.g.*, Prefix Independent Convergence or MPLS Fast Reroute

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... and the Internet converges very slowly\*

#### \*Holterbach et al. SWIFT: Predictive Fast Reroute ACM SIGCOMM, 2017

# Fire at AT&T facility causes widespread outage in North Texas

# Time Warner Cable comes back from nationwide Internet outage by Brian Stelter @brianstelter

(L) August 27, 2014: 11:07 PM ET

November 6, 2017 | Emerging Threats





#### BGP took minutes to converge upon the Time Warner Cable outage in 2014 1.0 8.0 0.6 CDF over the BGP peers 0.4 0.2 0.0 100 200 300 400 500 600 0 Time difference between the outage and the BGP withdrawals (s)



# Control-plane (*e.g.*, BGP) based techniques typically converge slowly upon remote outages

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What about using data-plane signals for fast rerouting?

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Outline

- 1. Why and how to use data-plane signals for fast rerouting
- 2. *Blink* infers more than 80% of the failures, often within 1s
- 3. Blink quickly reroutes traffic to working backup paths
- 4. *Blink* works in practice, on existing devices

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### TCP flows exhibit the same behavior upon failures

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#### destination

### TCP flows exhibit the same behavior upon failures destination source S:500 A:1000 failure





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#### TCP flows exhibit the same behavior upon failures











We simulated a failure affecting 100k flows with NS3

Same RTT distribution than in a real trace\*

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Number of retransmissions



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70K 60K 50K L 40K Number of retransmissions 30K 20K 10K 0 5 3 2 6 7 4 0 Time (s)

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### To detect failures, *Blink* looks at TCP retransmissions
#### number of retransmissions



#### number of retransmissions



#### number of retransmissions



### number of retransmissions



### Solution #1: *Blink* looks at consecutive packets with the same sequence number

#### Solution #1: Blink looks at consecutive packets with the same sequence number destination source S:500 A:1000 $= SRTT + 4 * RTT_VAR$ failure **RTO: 200ms** cwnd:4 pkts(=congestion window) S:1000 S:2100-5:3100 5:4100 *t* + 200ms cwnd:1 S:1000 exponential backoff cwnd:1 *t* + 600ms S:1000 cwnd:1 *t* + 1400ms S:1000



### Solution #2: *Blink* monitors the number of flows experiencing retransmissions over time using a sliding window

number of retransmissions



number of flows experiencing retransmissions



number of retransmissions





number of flows experiencing retransmissions

number of retransmissions





number of flows experiencing retransmissions

number of retransmissions



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#### Blink is intended to run in programmable switches

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### Solution #1: *Blink* focuses on the popular prefixes, *i.e.*, the ones that attract data traffic

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#### As Internet traffic follows a Zipf-like distribution\* (1k pref. account for >50%), **Blink** covers the vast majority of the Internet traffic

\*Sarra et al. Leveraging Zipf's Law for Traffic offloading ACM CCR, 2012



#### Solution #2: Blink monitors a sample of the flows for each monitored prefix

TCP flows



#### Traffic to a destination prefix

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#### TCP flows

#### default 64 flows monitored

#### Traffic to a destination prefix

To monitor active flows, *Blink* evicts a flow from the sample if it does not send a packet for a given time (default 2s)

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and selects a new one in a *first-seen, first-selected* manner

### **Blink** infers a failure for a prefix when the majority of the monitored flows experience retransmissions

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number of flows experiencing retransmissions



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#### We evaluated *Blink* failure inference using 15 real traces, 13 from CAIDA, 2 from MAWI, covering a total of 15.8 hours

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We are interested in:





**Speed**: How long does Blink take to infer failures

#### **Accuracy:** True Positive Rate vs False Positive Rate

### As we do not have ground truth, we generated synthetic traces following the traffic characteristics extracted from the real traces

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**Step #1 -** We extracted the RTT, Packet rate, Flow duration from the real traces

**Step #2 -** We used NS3 to replay these flows and simulate a failure

**Step #3 -** We ran a Python-based version of **Blink** on the resulting traces

#### Blink failure inference accuracy is above 80% for 13 real traces out of 15



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#### 8.0 0.6 **True Positive Rate** 0.4 0.2 0.0 6 4 2 3 5



#### **Real traces ID**

#### **Blink** avoids incorrectly inferring failures when packet loss is below 4%





#### Blink avoids incorrectly inferring failures when packet loss is below 4%



2	3	4	5	 8	9	
	0	0.67	0.67	 1.3	2.7	

#### **Blink** infers a failure within 1s for the majority of the cases



# 10 12 14 9 11 13 15 15 Real traces ID
## **Blink** infers a failure within 1s for the majority of the cases



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# Upon detection of a failure, *Blink* immediately activates backup paths pre-populated by the control-plane











# Solution: As for failures, *Blink* uses data-plane signals to pick a working backup path

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# As for failures, **Blink** compares the sequence number of consecutive packets to detect blackholes or loops\*



\*See the paper for an evaluation of the rerouting

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# We ran *Blink* on the 15 real traces (15.8 hours)

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#### RTTs in [10ms; 300ms]





#### RTTs in [10ms; 300ms]





#### RTTs in [10ms; 300ms]

# **Blink:** Fast Connectivity Recovery Entirely in the Data Plane

Infers failures from data-plane signals with more than 80% accuracy, and often within 1s

Fast reroutes traffic at line rate to working backup paths

Works on real traffic traces and on existing devices









# https://blink.ethz.ch

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When multiple flows experience the same failure the signal is a wave of retransmissions

We simulated a failure affecting 100k flows with NS3

Same RTT distribution than in a real trace\*

Number of retransmissions

\*CAIDA equinix-chicago direction A, 2015



# When multiple flows experience the same failure the signal is a wave of retransmissions





# Blink failure inference accuracy is close to a best case scenario, and is above 80% for 13 real traces out of 15





# **Blink** infers a failure within 1s for the majority of the cases



Real traces ID

"best case", *i.e.*, no sampling but threshold still 32







**Blink** avoids incorrectly inferring failures when packet loss is below 4%

	3	4	5		8	9
False Positive Rate						
	0	0.67	0.67		1.3	2.7
5	93	94	95		97	98

# Blink quickly infers and avoids forwarding loops

