Tempest: Temporal Dynamics in Anonymity Systems

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Introduction

Motivation

- Tor (USENIX 2004)
- DeNASA (PETS 2016)
- Counter-RAPTOR (S&P 2017)
- TAPS (NDSS 2017)
- LAP (S&P 2012)
- HORNET (CCS 2015)
- Dovetail (PETS 2014)
- PHI (PETS 2017)

Prior work: **static** security analyses

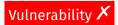
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- 2. **User Behavior**: Users make many connections over time

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- 2. **User Behavior**: Users make many connections over time
- 3. Routing Changes: Internet routes change over time

		Client Mobility	User Behavior	Routing Changes
Tor	Tor DeNASA			
	Counter-RAPTOR			
	TAPS			

Legend:



		Client Mobility	User Behavior	Routing Changes
Tor	Tor			
	DeNASA			
	Counter-RAPTOR			
	TAPS			
	HORNET			
Net-Layer	LAP			
	Dovetail			
	PHI			

Legend:

Vulnerability 🗡

		Client Mobility	User Behavior	Routing Changes
Tor	Tor DeNASA		Known X	Known X
	Counter-RAPTOR			Known 🗸
	TAPS		Known ✓	
Net-Layer	HORNET			
	LAP			
	Dovetail			
	PHI			

Legend:

Vulnerability 🗡

		Client Mobility	User Behavior	Routing Changes
	Tor	Novel X	Known X	Known X
Tor	DeNASA	Novel X	Novel X	
=	Counter-RAPTOR	Novel X	Novel ✓	Known 🗸
	TAPS		Known ✓	Novel X
Net-Layer	HORNET	Novel X		Novel X
	LAP	Novel X		Novel X
	Dovetail		Novel X	
	PHI		Novel X	

Legend:

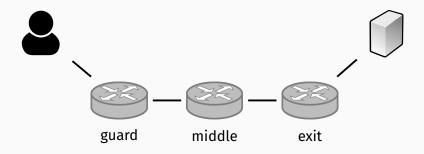
Vulnerability 🗡

Outline

- 1. Client Mobility & Tor
- 2. User Behavior & DeNASA

Brief Tor Background

Tor Background



- 1. Clients use only one guard for a few months
- 2. Relays are selected with probability prop. to bandwidth

Client Mobility & Tor

Ryan as an example...

Connected to Tor from

1) Home

Which networks (ASes) saw my ingress Tor traffic?

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Verizon Tata LeaseWeb

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Which networks (ASes) saw my ingress Tor traffic?

Verizon Tata LeaseWeb Comcast Telia AT&T Zayo

Connected to Tor from

- 1) Home 2) Coffee Shop 3) Mobile Hotspot 4) Airport
- 5) Hotel

Which networks (ASes) saw my ingress Tor traffic?

Verizon Tata LeaseWeb Comcast Telia AT&T Zayo

Connected to Tor from

- 1) Home 2) Coffee Shop 3) Mobile Hotspot 4) Airport
- 5) Hotel

Which networks (ASes) saw my ingress Tor traffic?

Verizon Tata LeaseWeb Comcast Telia AT&T Zayo TelefonicaEspaña Telxius

Connected to Tor from

- 1) Home 2) Coffee Shop 3) Mobile Hotspot 4) Airport
- 5) Hotel 6) CCCB

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Which networks (ASes) saw my ingress Tor traffic?

Verizon Tata LeaseWeb Comcast Telia AT&T Zayo TelefonicaEspaña Telxius OrangeEspaña

During travel from United States to Spain:

- Connected to Tor from 6 different locations
- Exposed my traffic to additional 7 ASes (3.3× increase)

Adversary Model

- Adversary compromises a single AS
- Passive
- · Goal: observe client-guard traffic

Mobility Analysis

- 1. How mobile are some clients?
- 2. Does mobility weaken system security?

Mobility Datasets

1. Foursquare (F)

- 270,000 users
- 18 months (Apr 2012 Sep 2013)

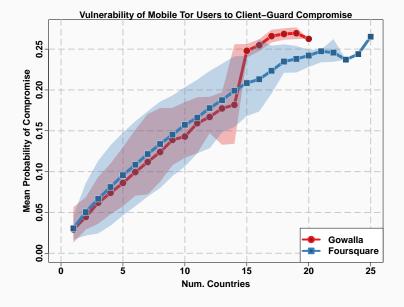
2. Gowalla (G)

- 100,000 users
- 20 months (Feb 2009 Oct 2010)

# Countries		2	3	4	5	6	≥7
Users	F	40145	13179	5649	2708	1490	2574
Users	G	17884	4557	1694	705	305	299
Q₁ Days	F	48	120	195	228	248	245
Q₁ Days	G	7	31	56	77	103	125

Mobility & Tor

- Assume each user connects from most popular Tor AS in each country.
- Compute average probability that largest 50 ASes compromise client-guard path.



Points denote median user, shade shows range

Takeaways

- 1. Many clients are mobile!
- 2. Mobility can reduce system security.

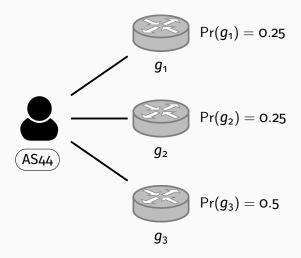
User Behavior &

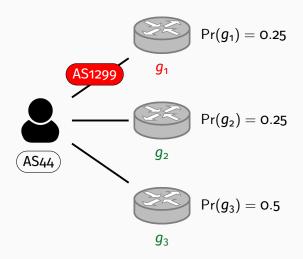
DeNASA (PETS 2016)

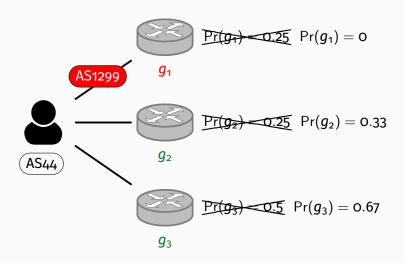
DeNASA

The DeNASA "g-select" algorithm:

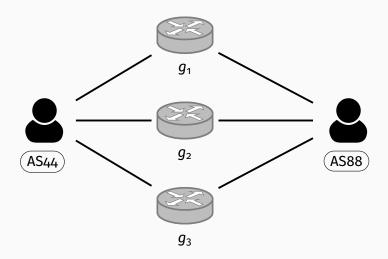
Do not select guards where *suspects* AS1299 (Telia) or AS3356 (Level 3) are on the client-guard link.

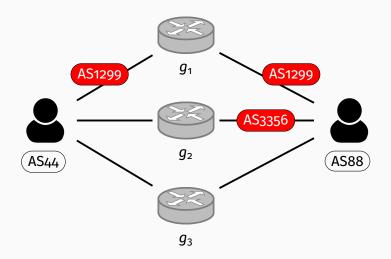


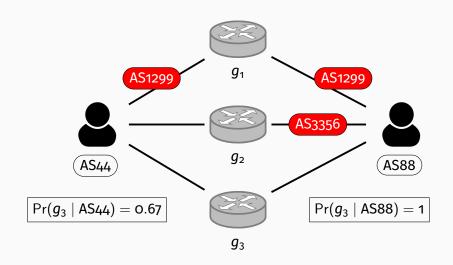


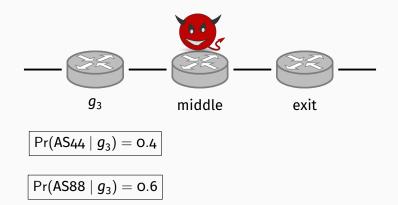


g-select leaks location information!









Our Tempest Attack

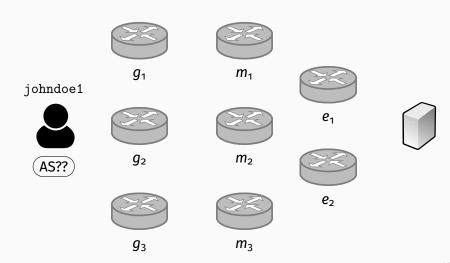
Leak worsens over time!

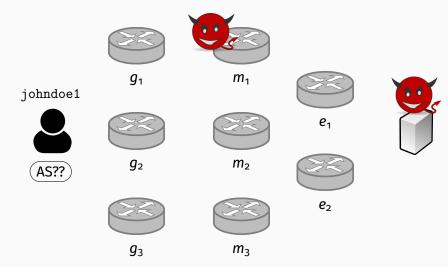
$$Pr(AS88 \mid G_1 \wedge G_2 \wedge \cdots \wedge G_N) \gg Pr(AS44 \mid G_1 \wedge G_2 \wedge \cdots \wedge G_N)$$

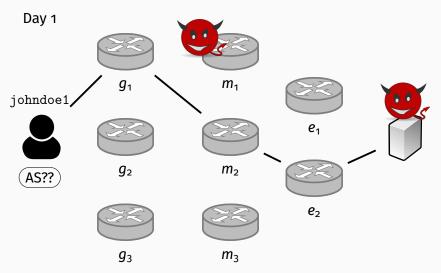
How can the adversary learn a client's guard history?

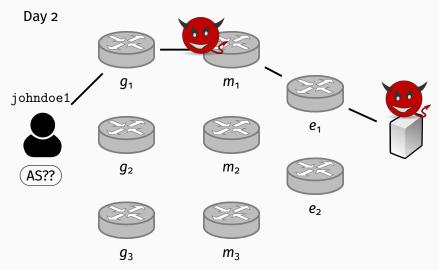
Adversary Model

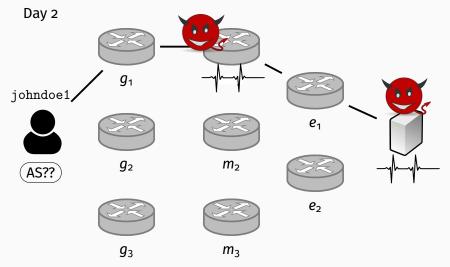
- Adversary runs a destination and some relays
- Passive
- · Goal: learn client AS

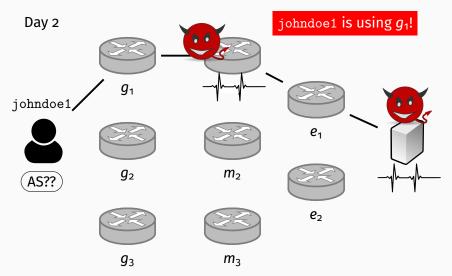




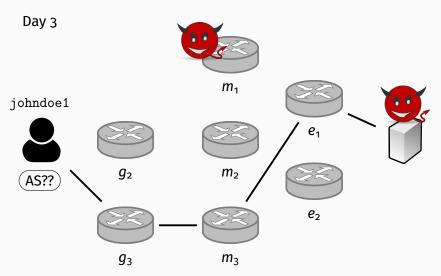


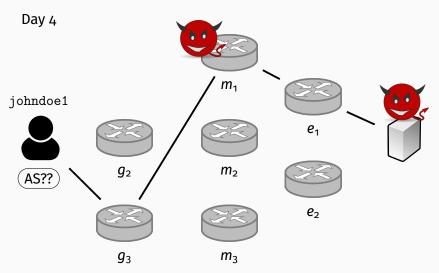


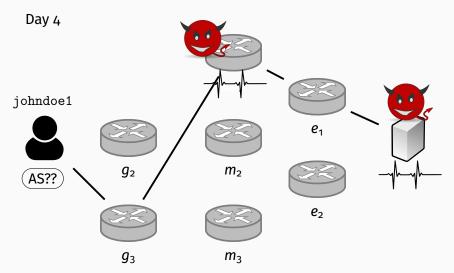


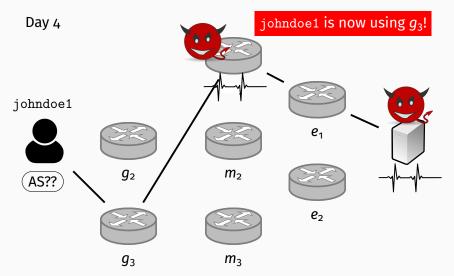


Day 3 m_1 johndoe1 e_1 g_2 m_2 AS?? e_2 g_3 m_3









Our Tempest Attack

Note:

- 1. Many other known guard discovery attacks.
- 2. Other ways to link client connections.

Our Tempest Attack

Adversary then computes posterior location distribution:

```
\Pr(\mathsf{AS1} \mid g_1 \land g_3)

\Pr(\mathsf{AS2} \mid g_1 \land g_3)

...

\Pr(\mathsf{AS60000} \mid g_1 \land g_3)
```

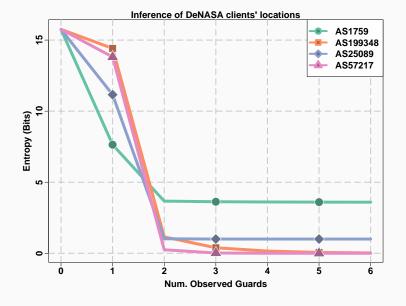
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- 2. Identified 10 "leaky" client locations
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- 4. Collected 100 samples for each location
- Computed average posterior entropy after adversary makes x guard observations



Points show average entropy after x guard observations

Takeaways

- 1. Small leaks can quickly become significant.
- 2. Important to consider the worst-case.

Conclusion

Temporal Dynamics & Anonymity Systems

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	Counter-RAPTOR	Novel X	Novel ✓	Known 🗸
	TAPS		Known ✓	Novel X
Net-Layer	HORNET	Novel X		Novel X
	LAP	Novel X		Novel X
	Dovetail		Novel X	
	PHI		Novel X	

Legend:

Vulnerability 🗡

Resistance ✓

Future Work

- 1. Explicitly accounting for temporal dynamics
- 2. Considering the long-lived adversary
- 3. Capturing time in evaluations and formalization

Thank you!