Profiling BGP Serial Hijackers: Capturing Persistent Misbehavior in the Global Routing Table

NANOG 80

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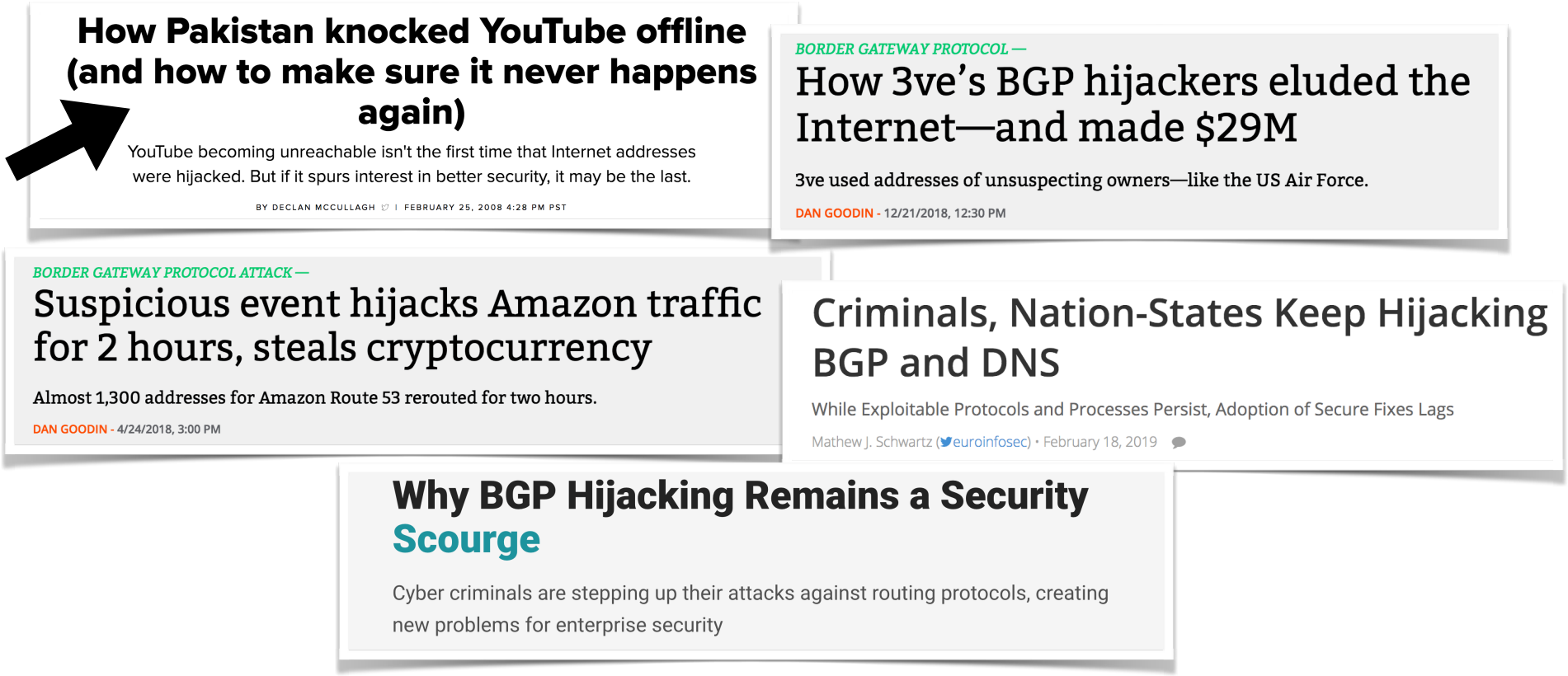
BGP hijacking is pervasive in the Internet



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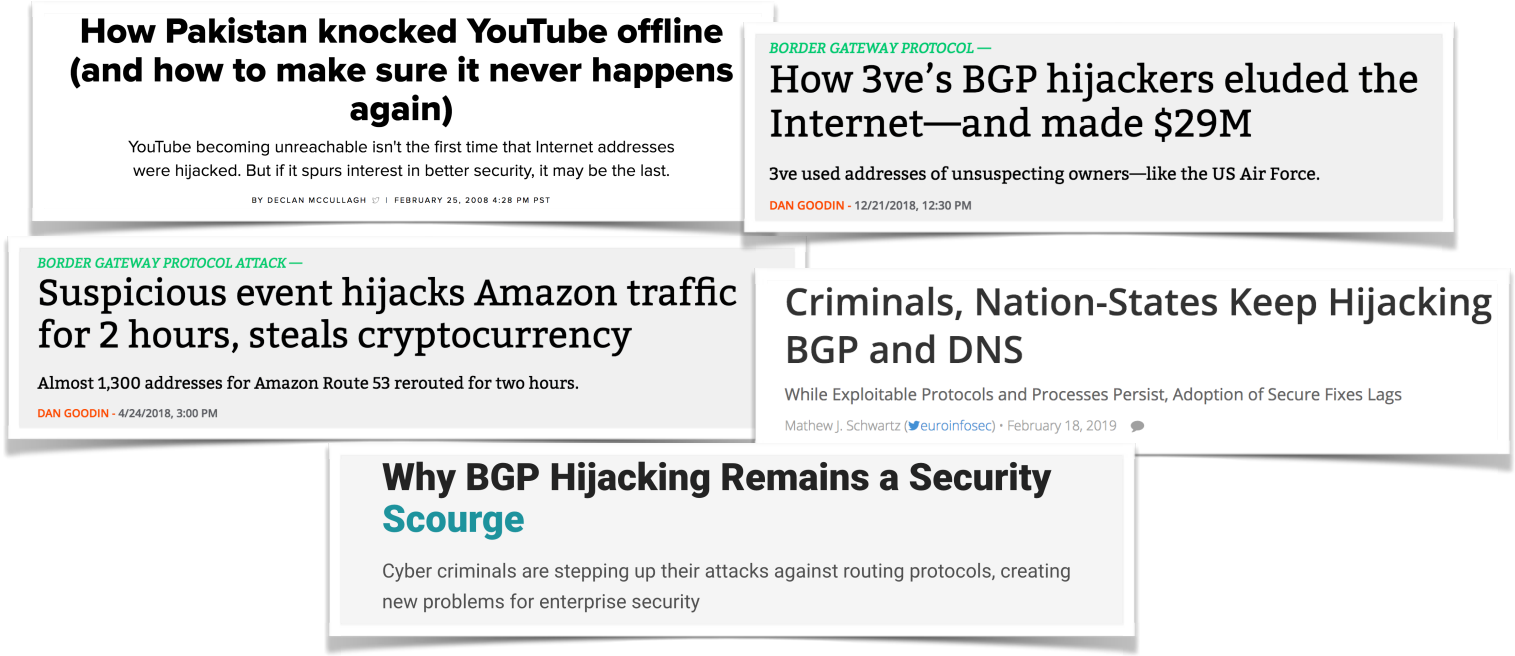
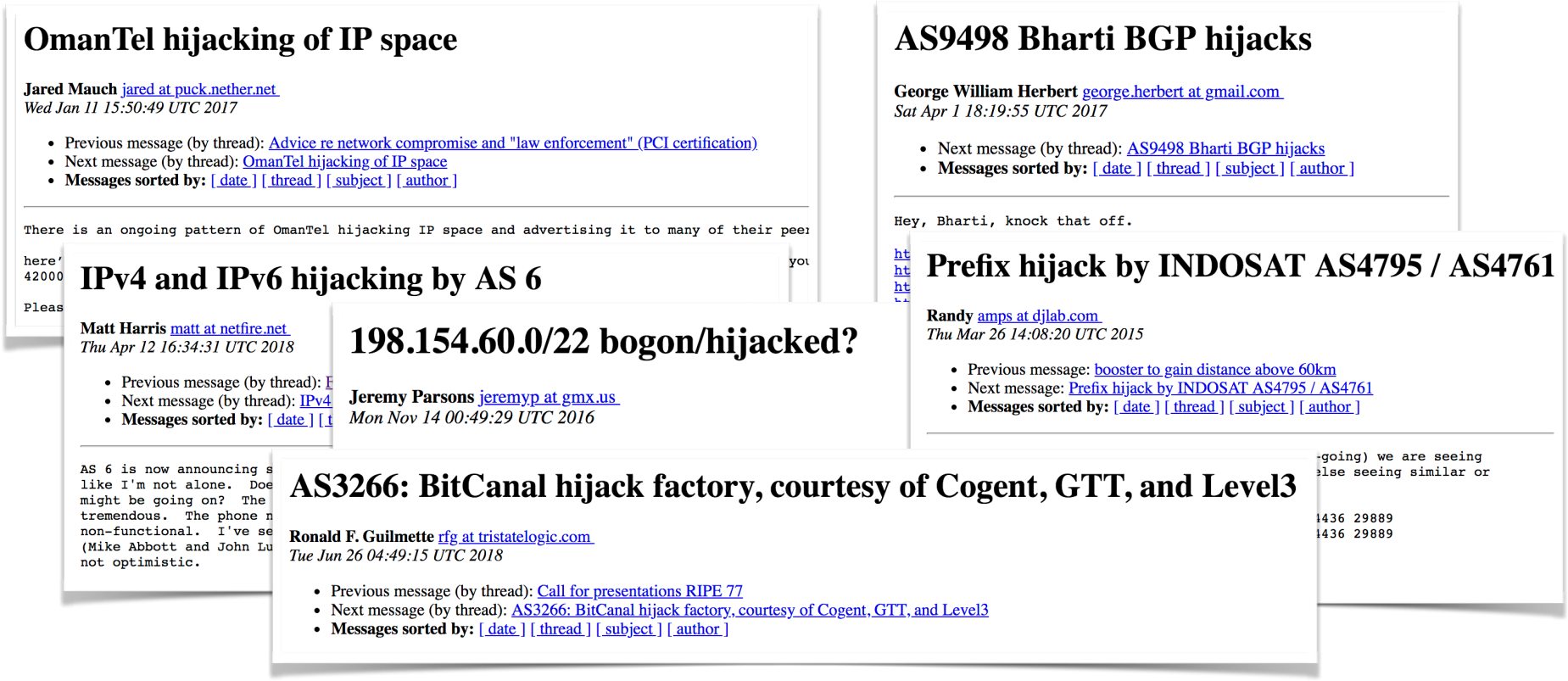


- The problem of BGP hijacking is **still** far from solved.

Hijack disclosure in mailing lists



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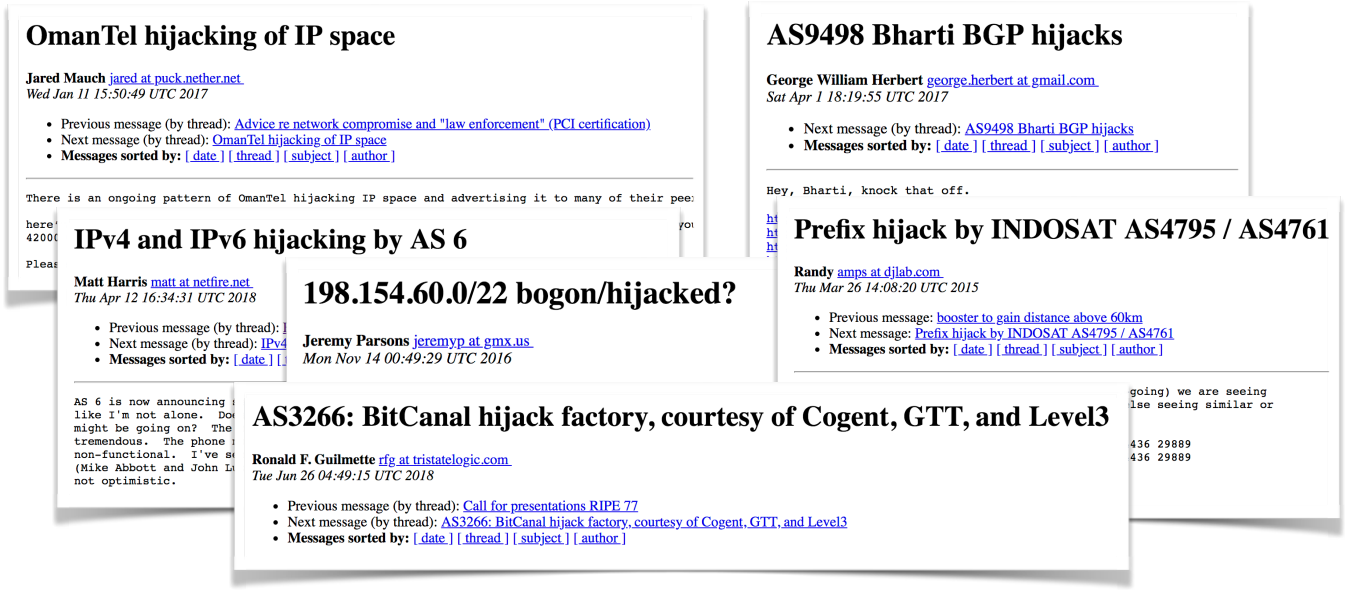
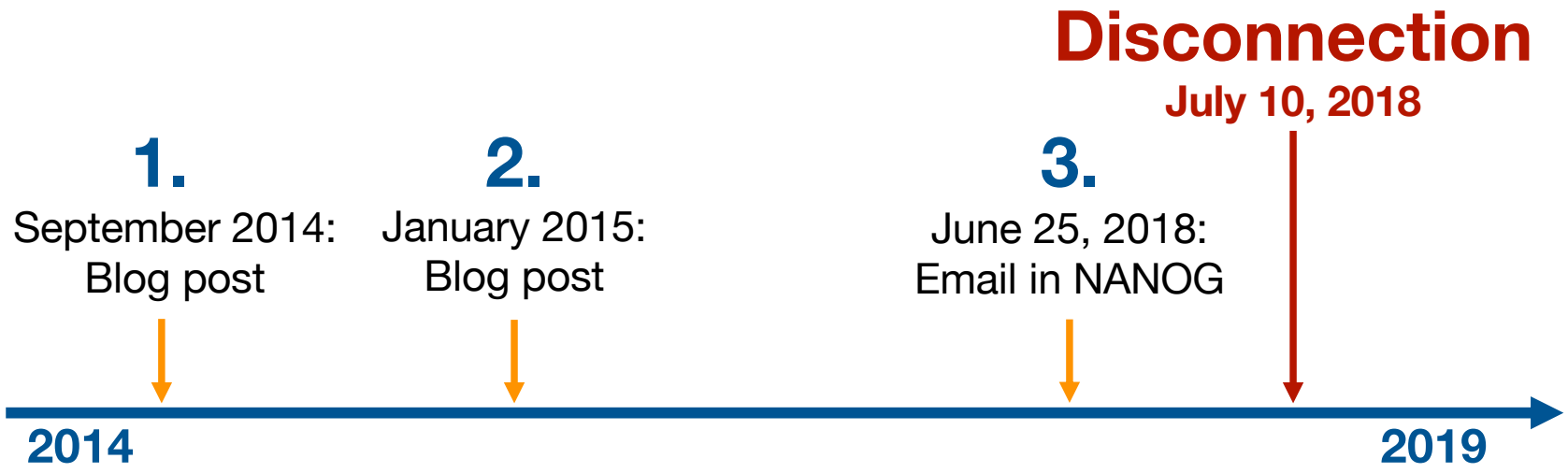


- ***Serial hijackers***: ASes that **repeatedly hijack** over **long periods** of time.

Bitcanal: an infamous serial hijacker



- It took **4 years** to disconnect this serial hijacker.

Research goals

Find serial hijackers in the Internet

(i) Identify hijackers distinctive routing characteristics

(ii) Build a machine learning system to flag suspicious ASes

(iii) Evaluate our results

What can we learn about serial hijackers?

Ground truth: serial hijackers

23 serial hijackers:

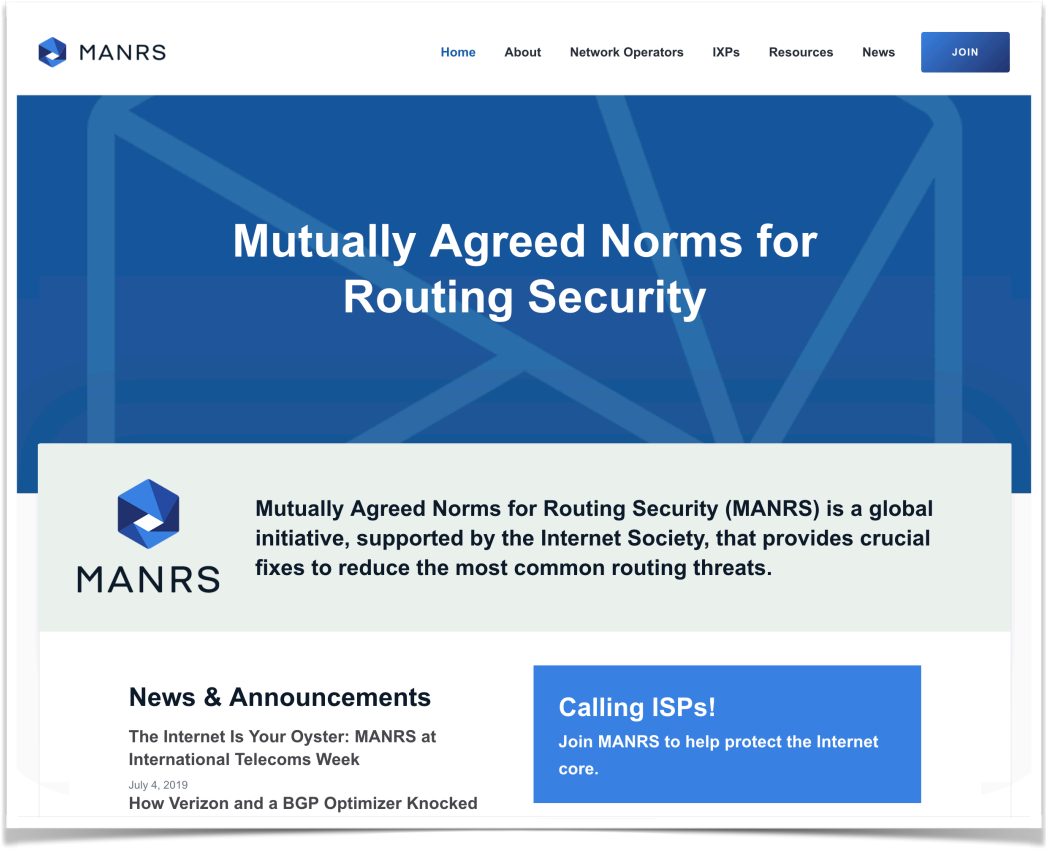
- 10+ hijacks
- Most have been active over a year
- Up to 30,000 originated prefixes

ASN country and RIR registration

Ground truth: legitimate ASes

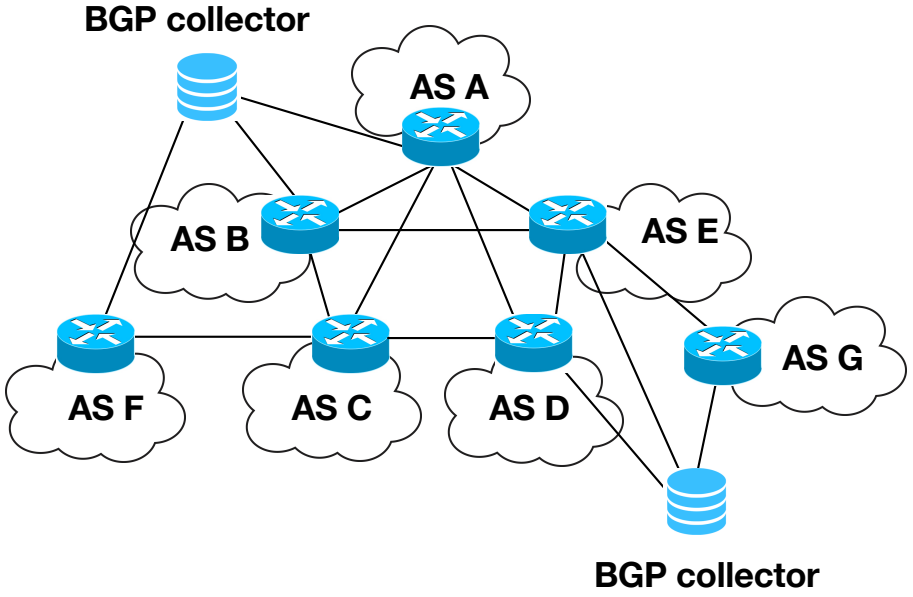
230 Legitimate ASes:

- 191 MANRS ASes
- 26 ASes manually selected



BGP dataset and processing

- RIPE RIS and RouteViews collectors (~40 col., ~1400+ col. peers)
- We process all **BGP updates** to reconstruct peer routing tables
- We extract **(prefix, origin AS)** pairs and the number of peers with each pair in their routing table **(visibility)**
- Data from Jan. 2014 to Dec. 2018



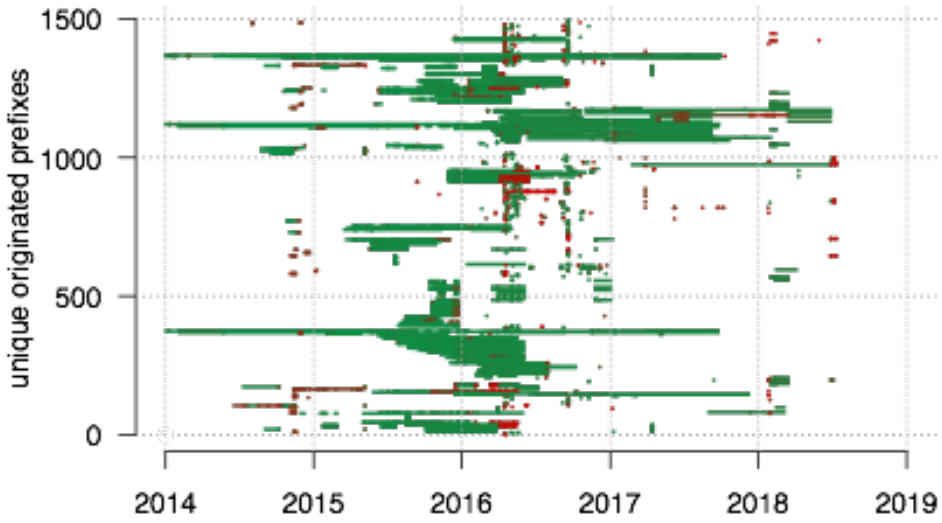
▶ **(prefix, origin AS, visibility, timestamp)** every 5 min.

BGP origination behavior: legitimate vs. serial hijacker

British Telecom (AS 5400)

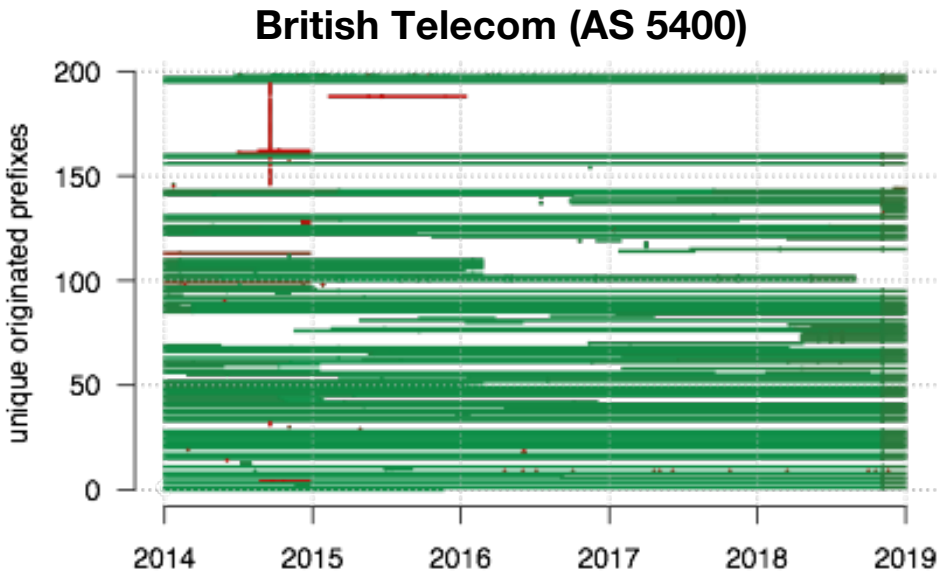
Legitimate ASes mostly show **stable** BGP behavior.

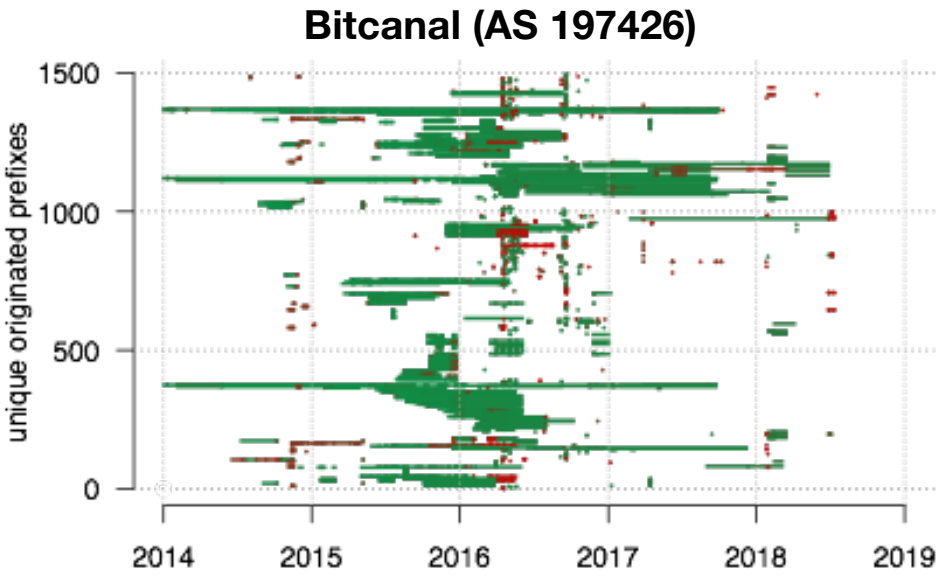
Bitcanal (AS 197426)



Serial hijackers BGP activity is **visually different.**

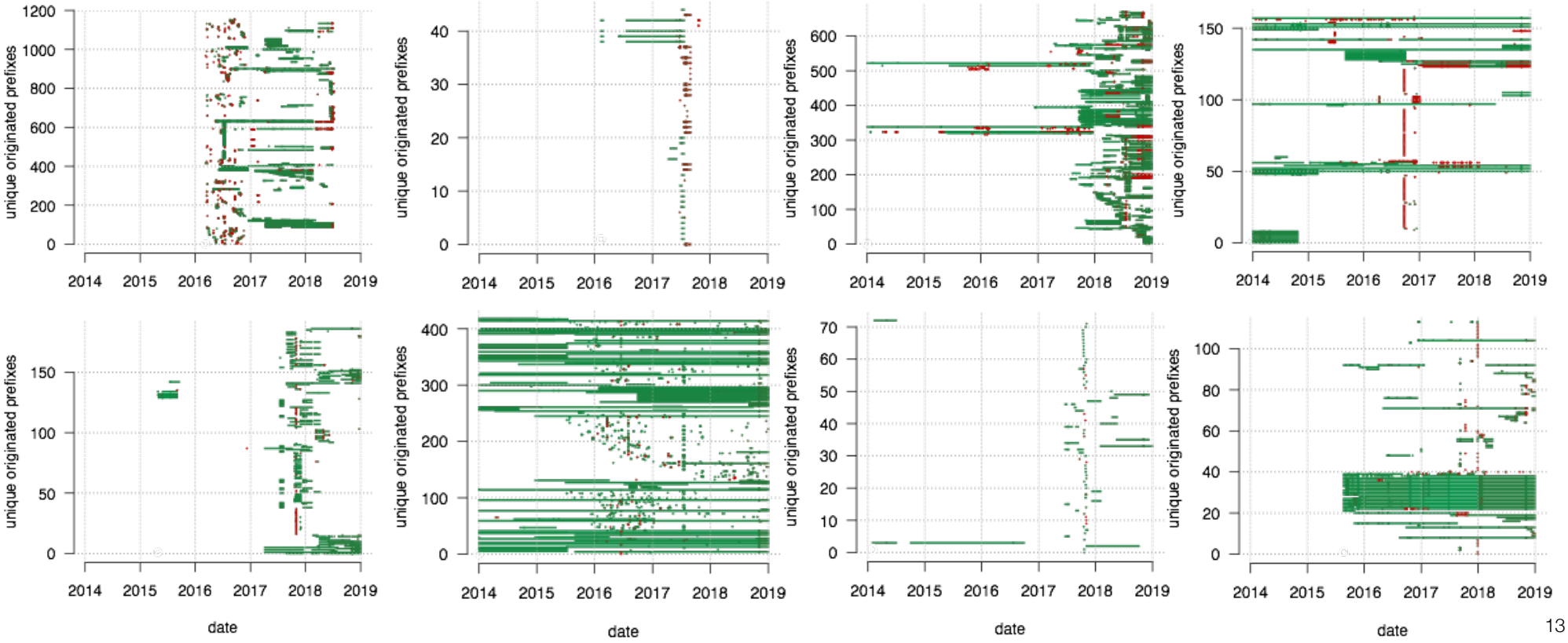
BGP origination behavior: legitimate vs. serial hijacker



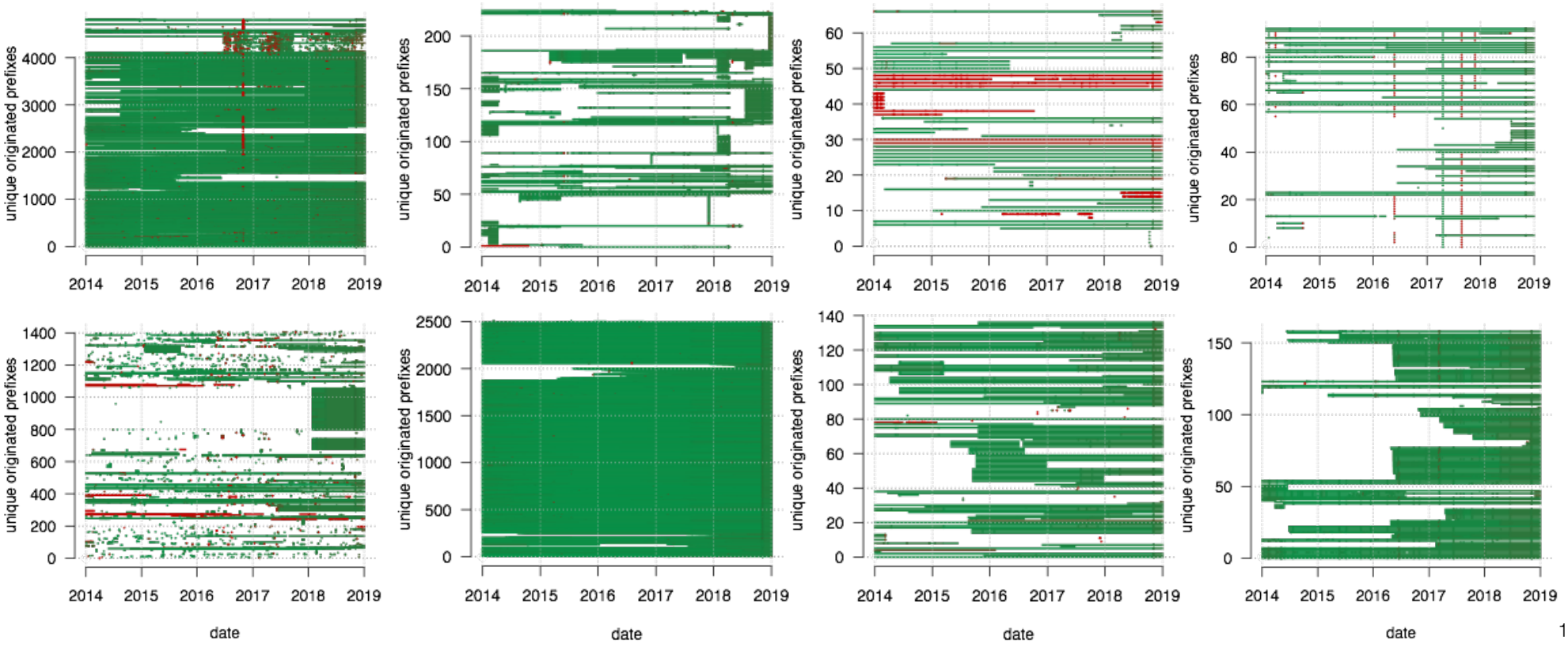


- We need features that **capture** this **behavioral difference.**

Variability of BGP behavior: serial hijackers



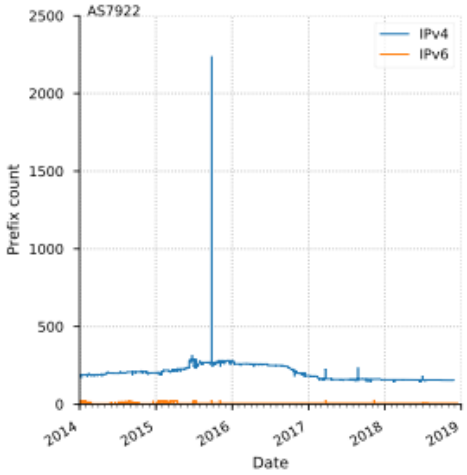
Variability of BGP behavior: legitimate ASes



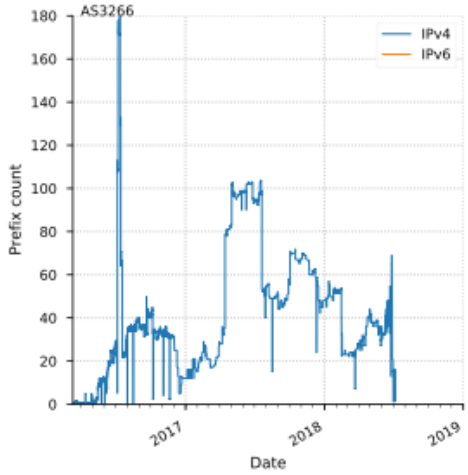
Expected serial hijacker behavior

- Repeated AS absence from the global routing table.
- Short prefix origination times.
- More multi-origin conflicts (MOAS).
- Volatile count of concurrently advertised prefixes.
- Broad geographical distribution of address space originated.

Comcast AS



Serial Hijacker AS



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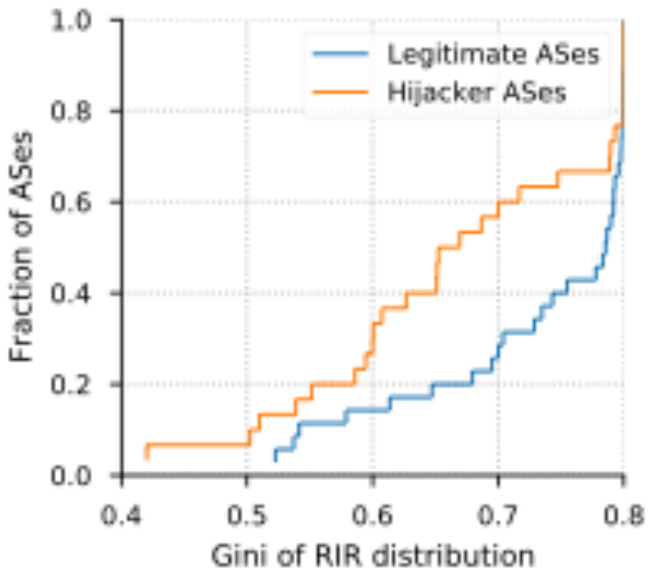

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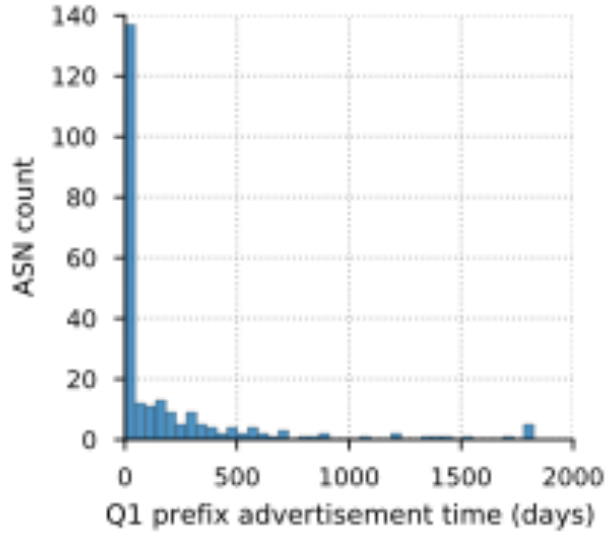
▶ We derived **52 features** to capture differences.

Challenges of applying ML to find more potential serial hijackers

- Heavy-tailed and skewed data:
  Monthly prefix changes [0,2600], Gini in [0,0.8]
- Very small ground truth:
  240 AS for 19,000 ASes
- Class Imbalance:
  23 serial hijacker vs. 217 legitimate networks



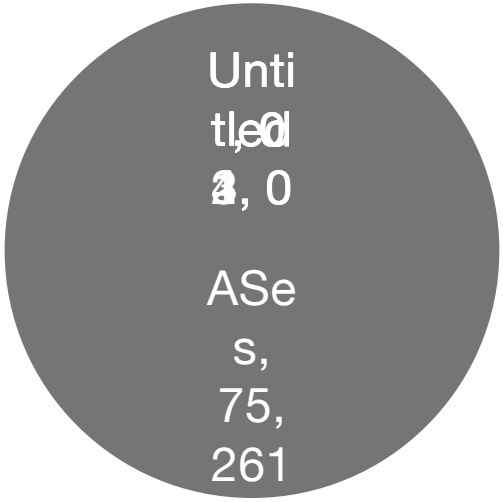
Our ML approach

- Tree based classifier.
- Voting ensemble of extremely randomized forests.
- 3 over-sampling techniques.
- All 52 features with positive median drop column importance.

▶ **79% precision** and **100% recall**
(in ground-truth using out-of-bag score)

Putting our classifier to work

- **Goal:** Find ASes exhibiting similar BGP behavior to serial hijackers in our ground truth.



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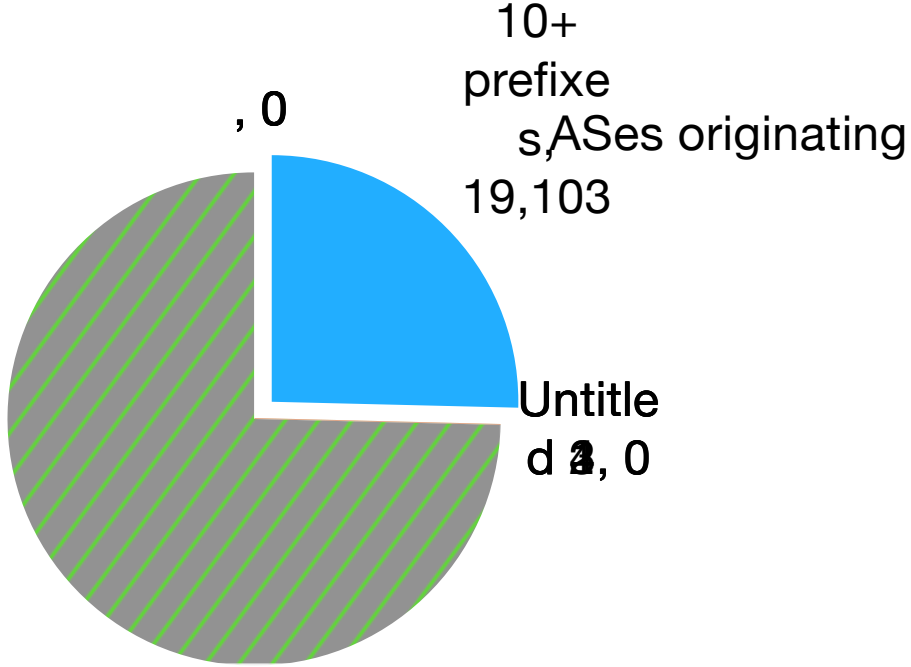

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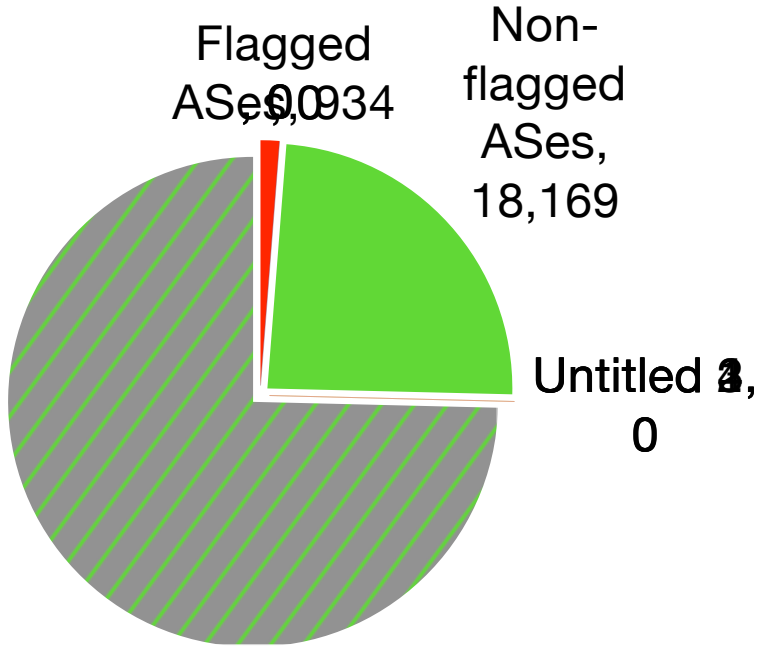

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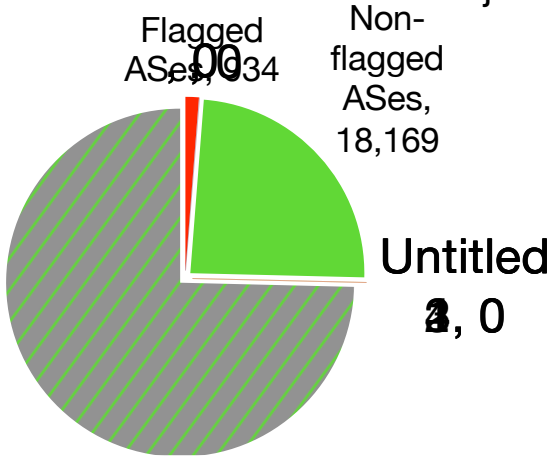
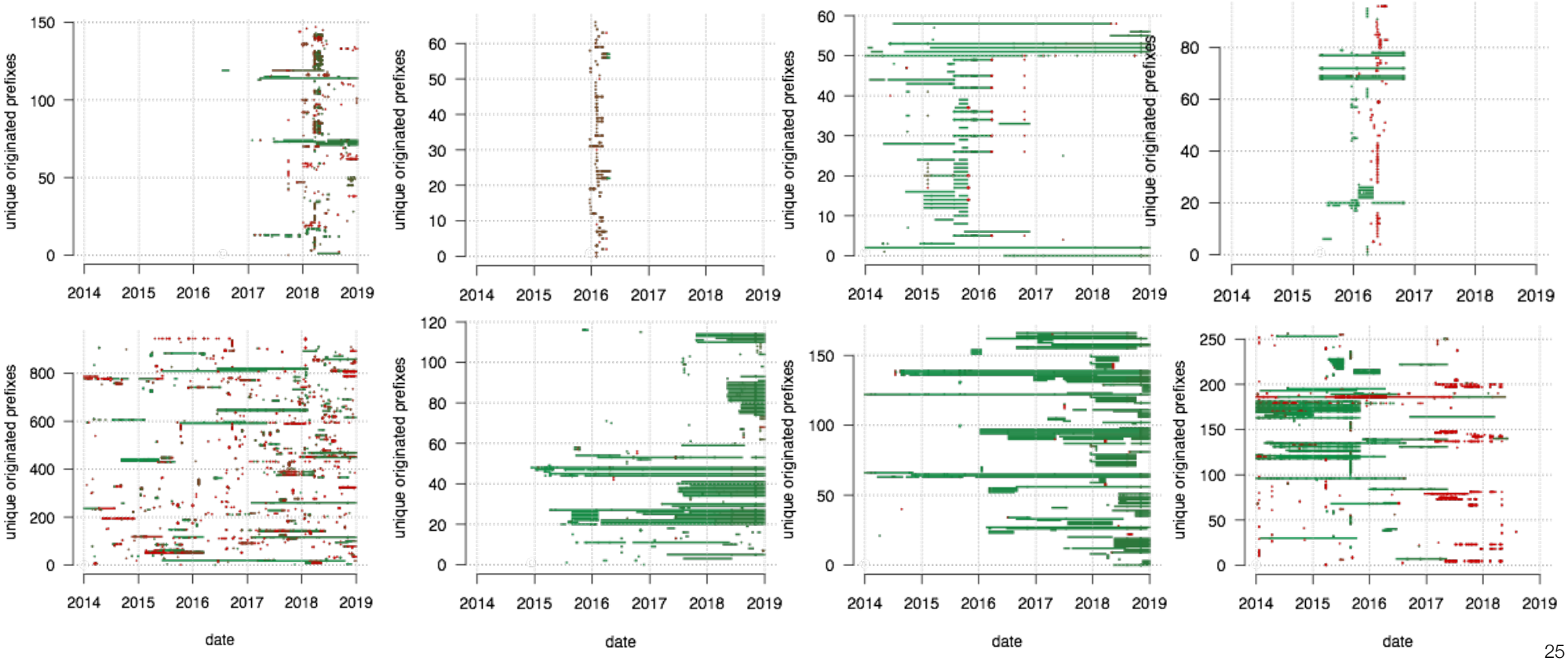


- Flagged ASes are:
  - **4.9%** of ASes originating 10+ prefixes
  - **1.2%** of all ASes.

BGP behavior of flagged ASes



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Flagged ASes are **10x** more likely to be block listed

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  - Spammer ASNs:

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- **Indication of malicious behavior**
  - Block listed ASNs: **84/290** ASes in *Spamhaus ASN DROP list*
  - Spammer ASNs: **33%** ASes have a prefix in UCE-PROTECT level 2 spam blacklist

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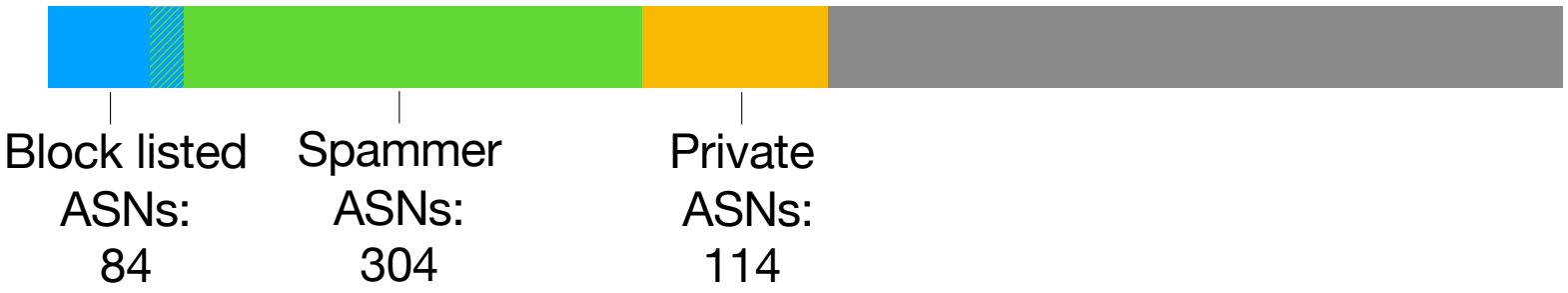
  - Private ASNs **12%**



What are ASes flagged by our classifier?

- Indication of malicious behavior
- **Indication of misconfigurations**
  - Private ASNs **12%**
  - Fat-finger error ASNs **1%**

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- **Known false positives**

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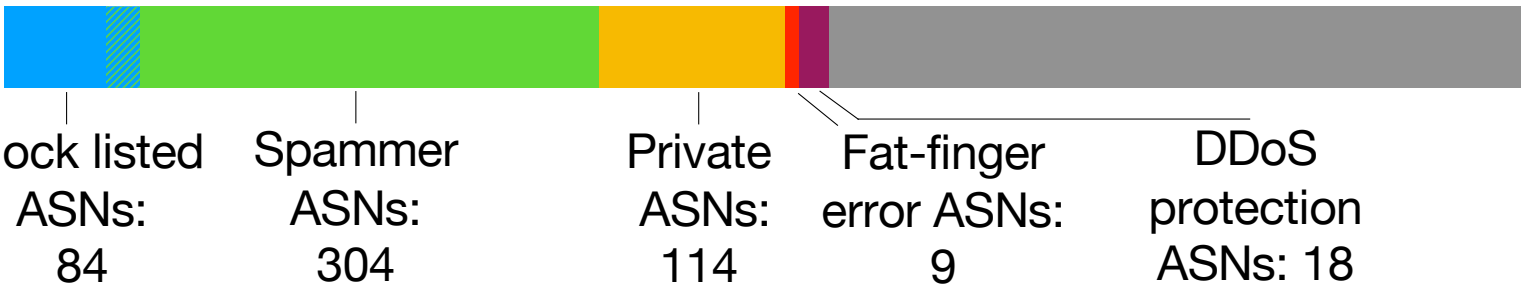
- **Known false positives**
  - DDos protection ASNs **2%**



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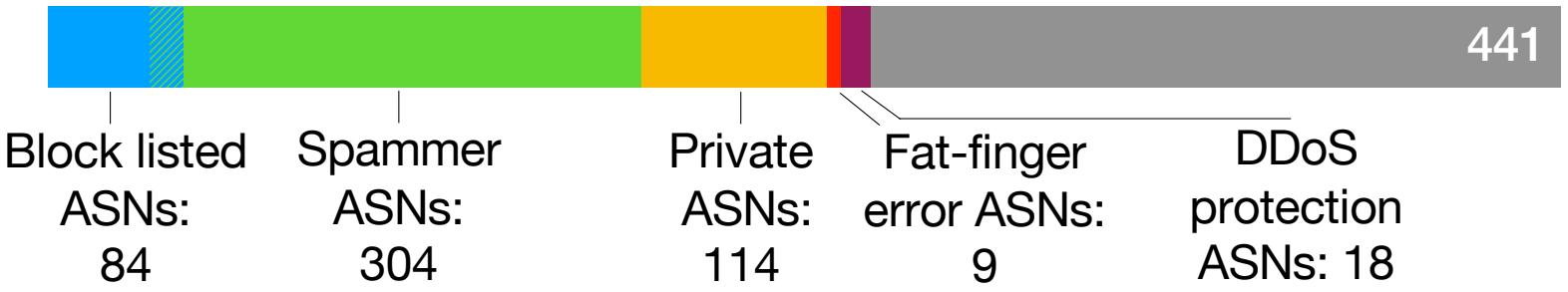

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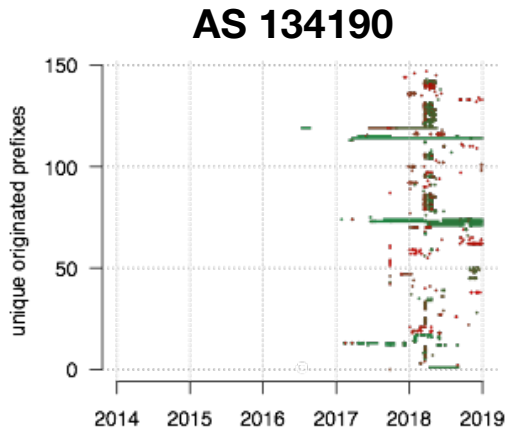

Block listed ASNs: 84

Spammer ASNs: 304

Private ASNs: 114

Fat-finger error ASNs: 9

DDoS protection ASNs: 18

What are ASes flagged by our classifier?

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- Indication of misconfigurations
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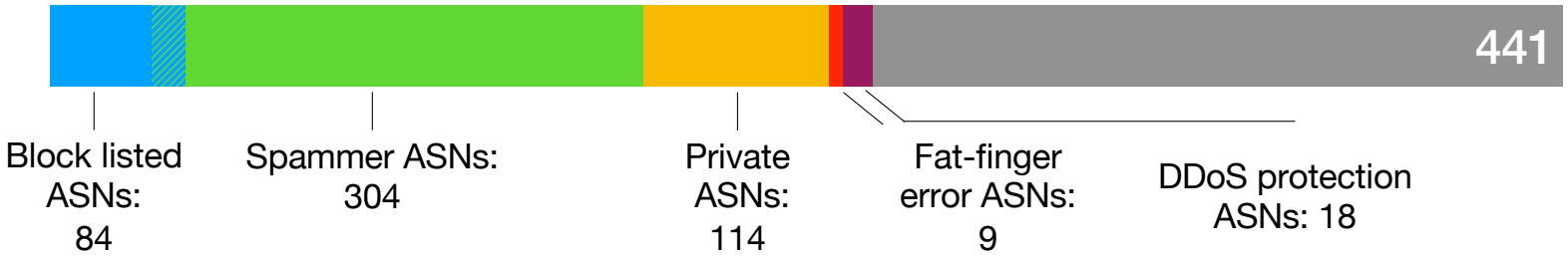

- 53% of flagged ASes are in known categories.
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What our classifier is not…

- A bulletproof identifier of malicious ASes.
- A system that exhaustively captures hijackers.

Key takeaways

- **First** longitudinal analysis of **serial hijacker** ASes.
- Features offer **state of affairs** of AS-wide **BGP behavior**.
- Classifier outcome provides **new data for network reputation** scoring systems.
- Effectively **narrows the focus on suspicious networks**, with much future work to be done.

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- **First** longitudinal analysis of serial hijacker ASes.
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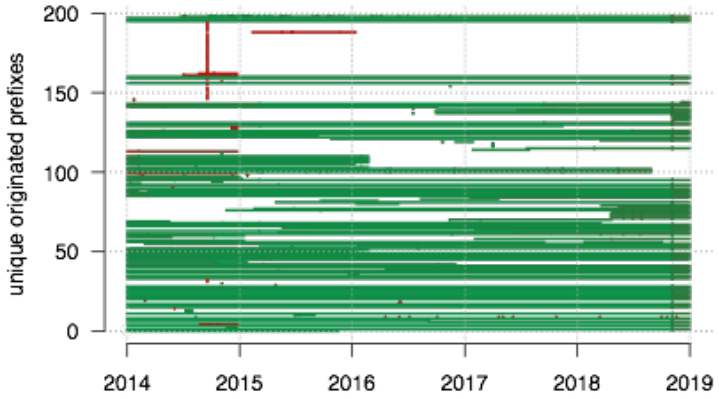
Legitimate AS



Serial hijacker AS