

PARGAV

Privacy Aware Research of Generic Anomalies and Visualisations

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Our Team

PARGAV

Privacy **Aware** Research of **Generic** **Anomalies** and **Visualisations**



Data Processing



Visualization
Driving

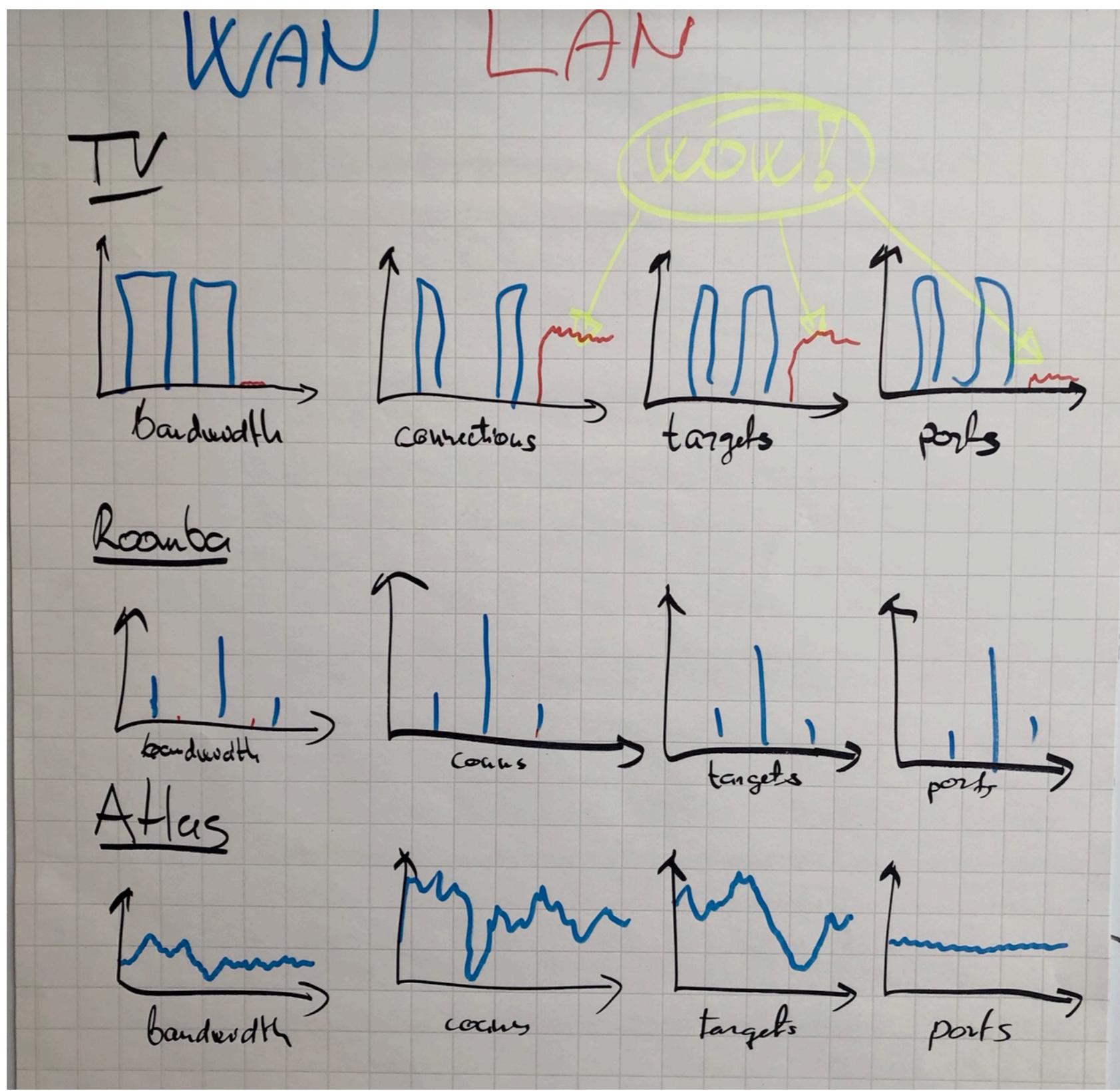


Anomalies
Detection

Goal

- Finding methods to identify network anomalies by analysing network traffic patterns.
- Assumption: IoT devices are connected to a (home) network doing their thing. We want to **visualise their behaviour** and **detect anomalies** in network activity.
- **Proof of concept**, based on real-life historical data, some hardly applicable frameworks (visualisation) and needs-a-lot-of-work integrations.

Concept Art



At a Glance

- Planned to work with real-life, already captured data
- Steps planned:
 1. Pre-processing
 2. Graphing
 3. Anomaly detection
 - A. Naïve
 - B. “Professional”
 - C. Perhaps both
 4. Future work: categorise, learn traffic patterns, adapt, block, ...

Data preparation

- In real life scenarios one does real-time pcaps, monitor (DHCP) clients, ... to produce streamed input
- Now we “replayed” existing data, converted it to fit graphing / anomaly detection

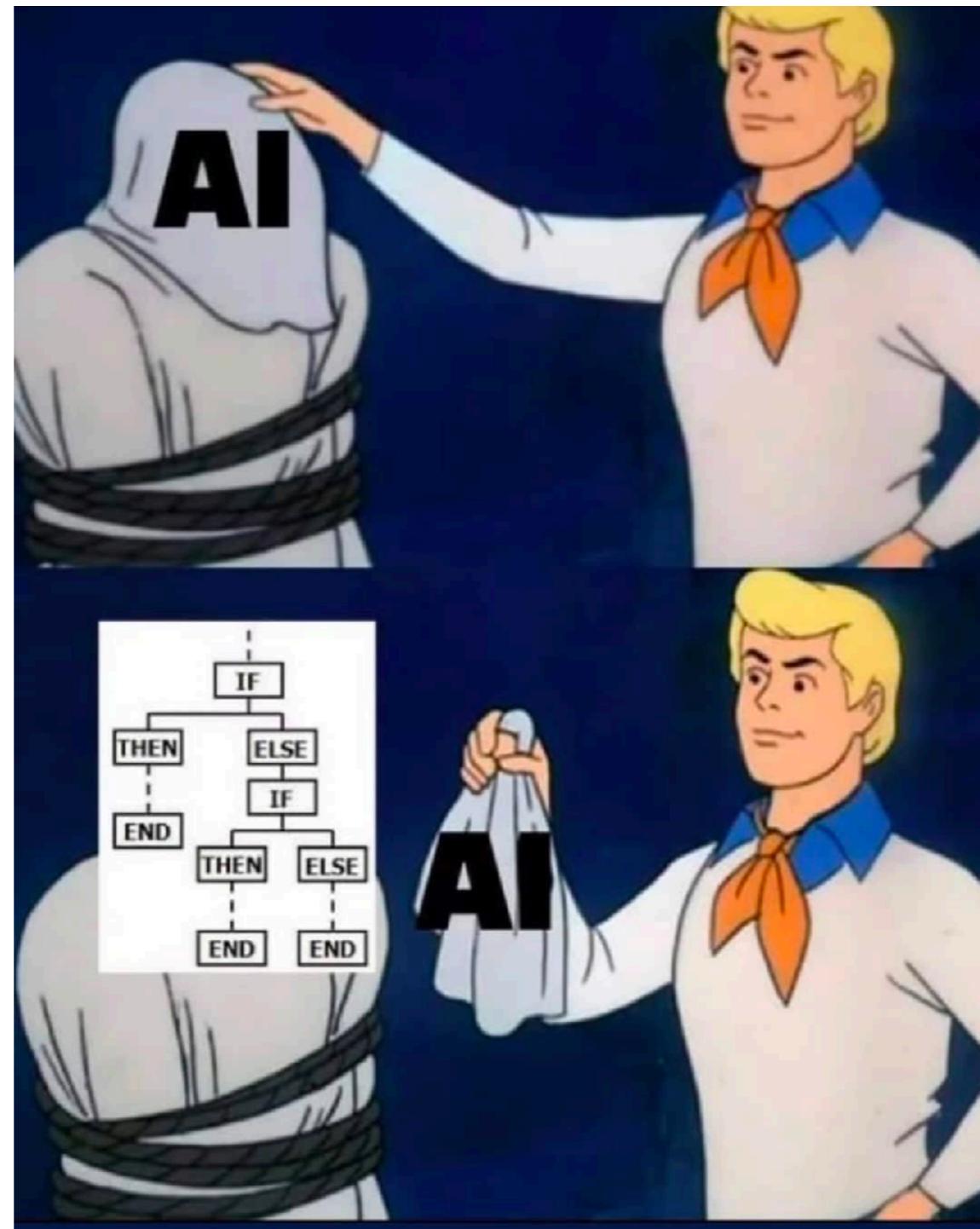
Graphing

- Could be done with RRD on constrained devices
- Maybe Grafana, InfluxDB, ... in less constrained environments
- We visualised historical data, went with something simpler (d3js/c3js)
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Anomaly Detection

- Naïve approach:
 - Compare current values to median / average of past few minutes / hours
- Professional tools:
 - `luminol` (<https://github.com/linkedin/luminol>) &
 - `prophet` (https://facebook.github.io/prophet/docs/quick_start.html)
- In real life combinations are possible (ie. naive detector has a signal, sends data off for confirmation to a different algorithm)

Anomaly Detection



Other ideas

- Categorise devices
 - A TV behaves differently (ie. has different traffic patterns) than a Roomba or a phone or a RIPE Atlas probe
 - “Learning/training mode” to establish normal behaviour
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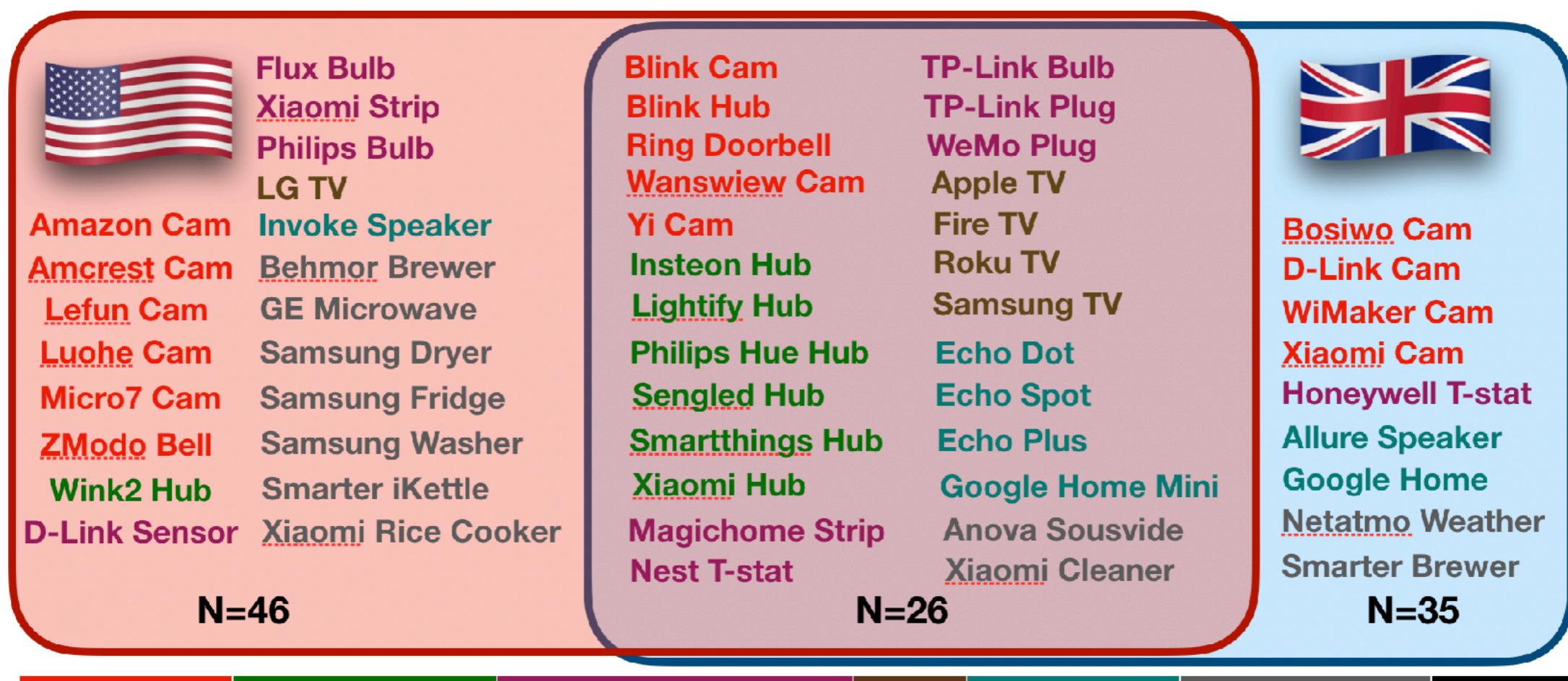
Data Used

- Mon(IoT)r Testbed
 - <https://moniotrlab.ccis.neu.edu/tools/>
- Also looked at BaloT attack data:
 - <https://archive.ics.uci.edu/ml/machine-learning-databases/00442/>

Data Used

Selecting IoT Devices

- **Criteria:** category; features; popularity; US & UK markets

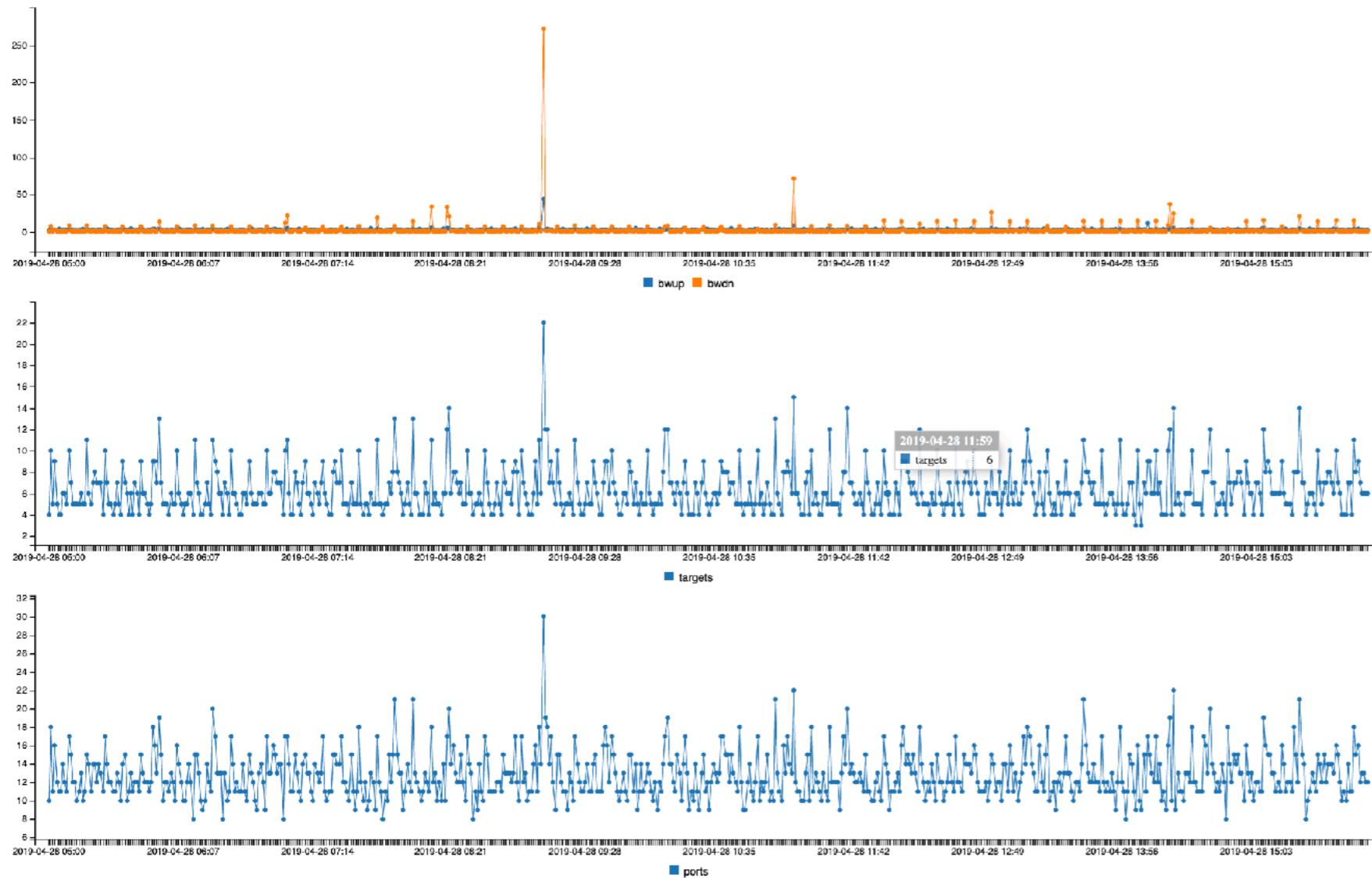


20 Cameras | 13 Smart Hubs | 15 Home Automation | 9 TVs | 11 Speakers | 13 Appliances | 81 Total



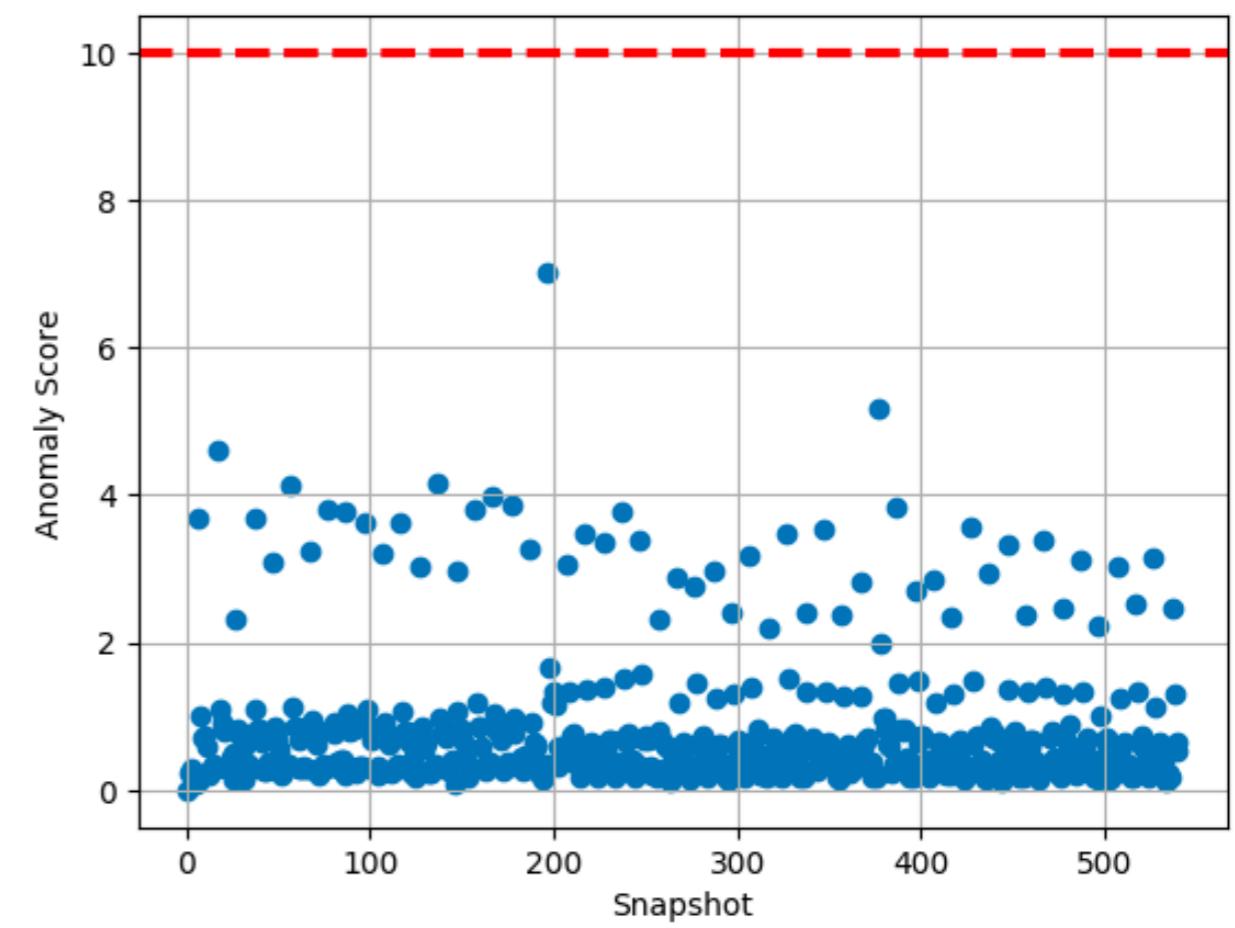
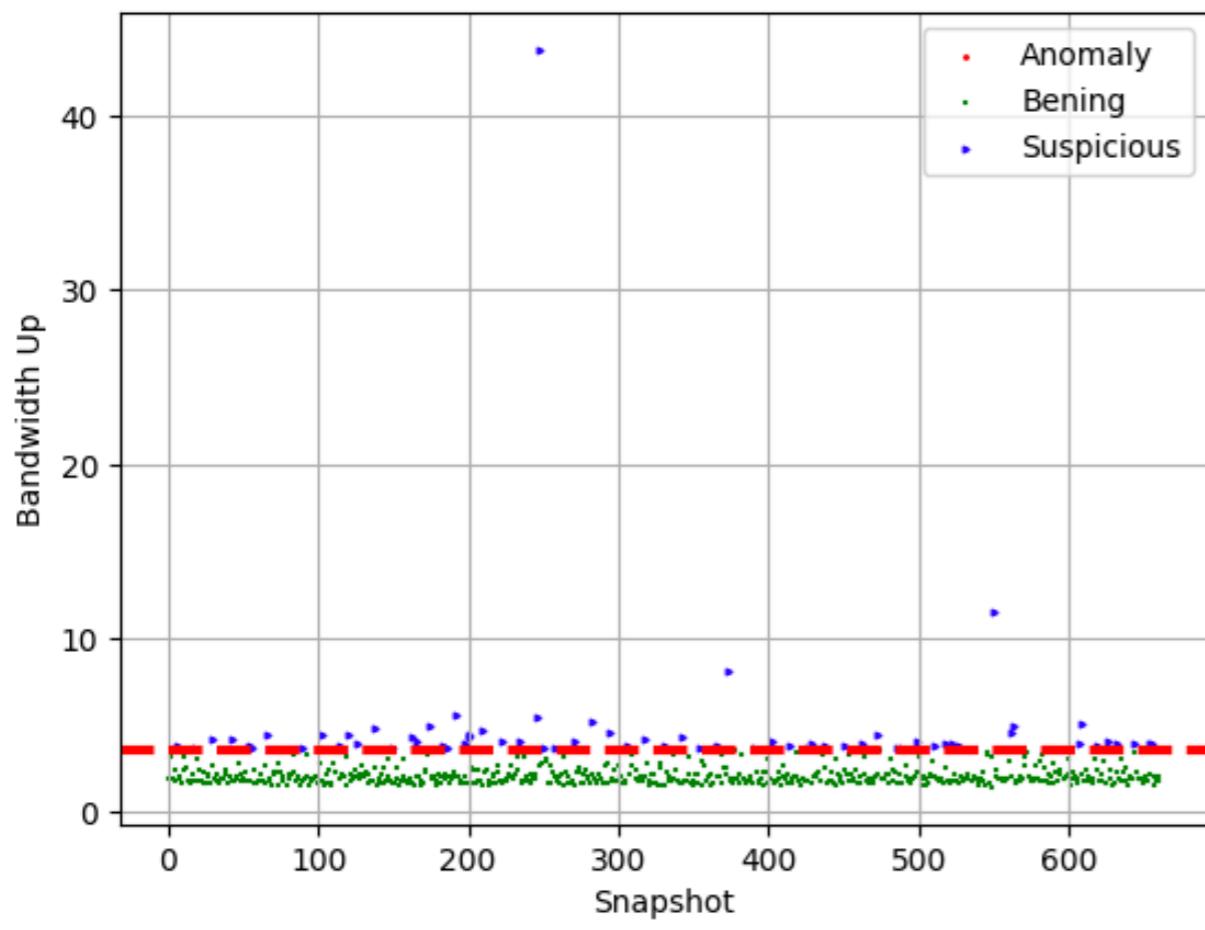
Idling

Samsung TV (US)



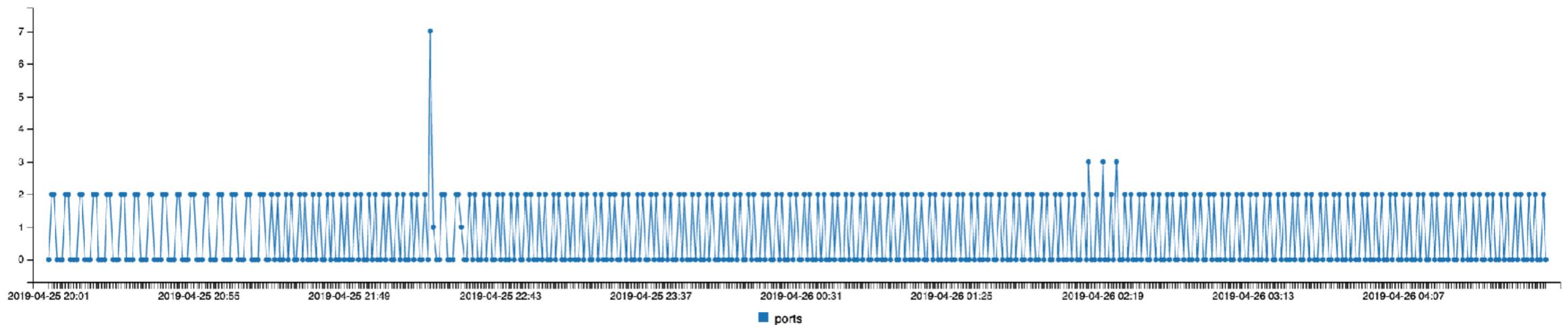
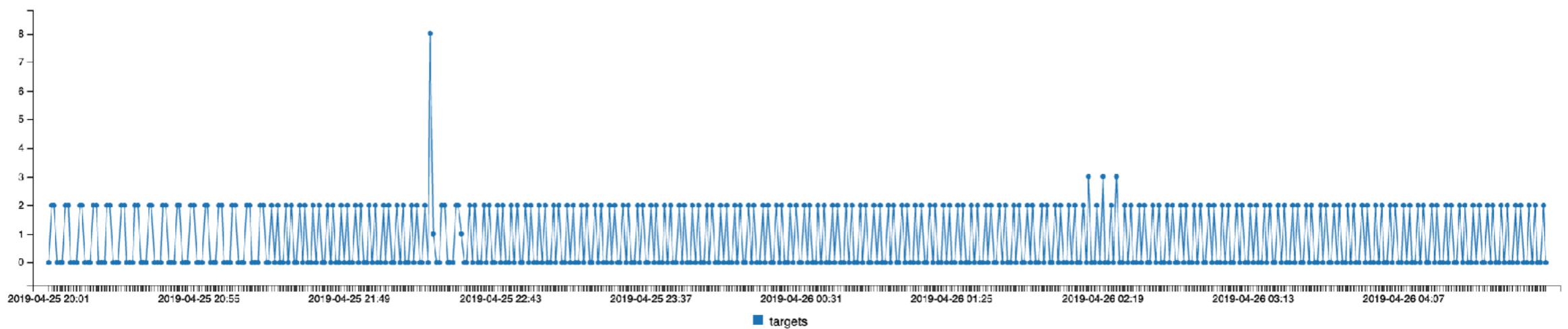
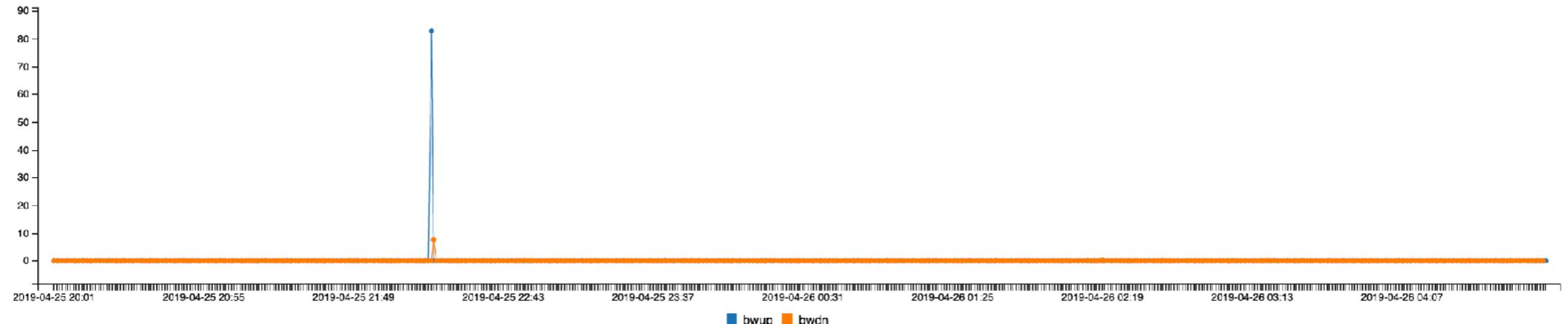
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Samsung TV



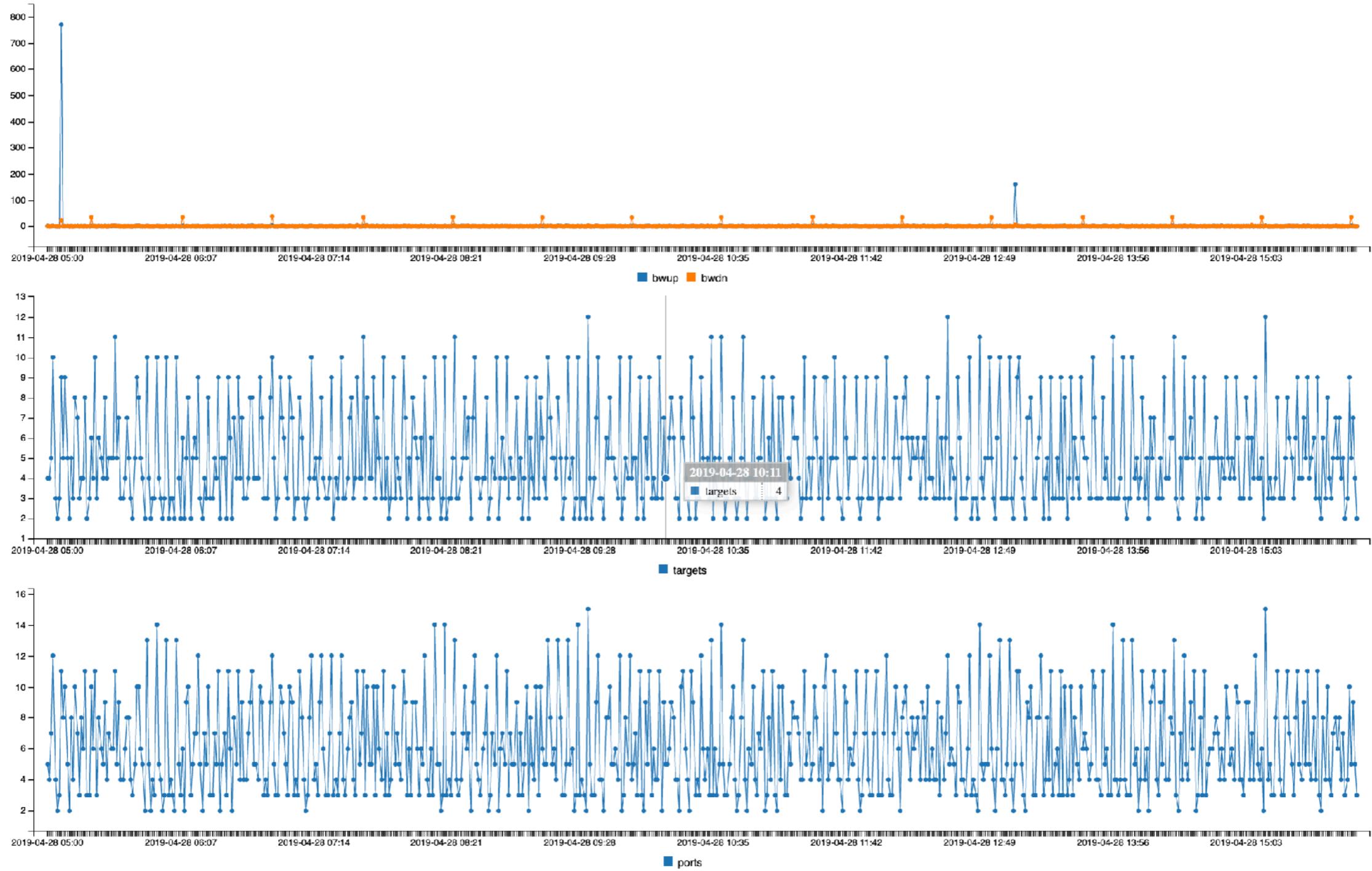
Idling

Ring



Idling

Echo Plus (US)



Attaaaaaaack!

BaIoT Mirai Thermostat attack

