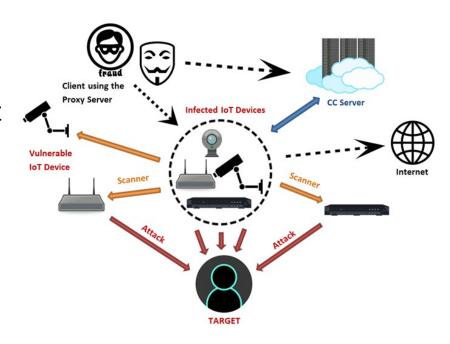


Machine Learning for IoT Network Monitoring

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Motivation

- > 75B connected devices by 2030
- Diversity of IoT devices: security camera, smart bulb, smart plug, smart thermostat, smart car, ...
- ► 600% increase in IoT attacks from 2016 to 2017 (ISTR Symantec 2018)
- Constantly evolving malware:
 - Mirai (2016)
 - Reaper (2017)
 - HideNSeek (2018)



IoT Botnet Source: www.fortinet.com



Proposal

- ➤ <u>SoTA.</u> Extensive academic research on the use of ML for intrusion detection in general purpose networks: but few actual deployments in production
- R. Sommer & V. Paxson: Outside the closed world, S&P'10
- ► <u>Key observation.</u> IoT devices perform very specific tasks making their networking behavior very stable and predictable
- Constraints. IoT devices may be resource-constrained so no monitoring on host
- Two types of monitoring methods are proposed:
 - ■IoT device identification: classification methods help enforce security, e.g. by applying device specific filtering rules.
 - Traffic anomaly detection: learning IoT signals to detect new and anomalous behaviors



Platform

► Experimental smart home network composed of Nest security camera, D-Link motion sensor, TP-Link smart bulb and smart plug.

► Traffic collected for 7 day	ys
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N = 10	, timeout =	600 seconds
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	train	test
Motion sensor	867	207
Security camera	839	216
Smart Bulb	821	219
Smart Plug	695	163
Total	3222	805



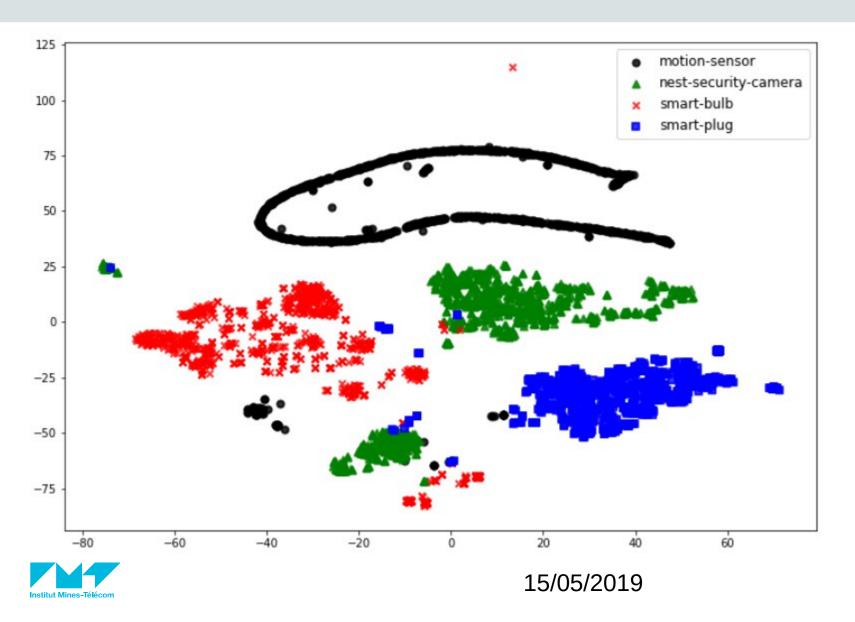


IoT Device Recognition

- Network Traffic Classification on TCP packet headers or flow metadata
- Assumption: 1 profile for each category of IoT device
- Training: 7 days of data, validation set of 25% to fine tune hyperparameters
- Focus on precision: need to accurately classify objects
- ► Tested algorithms: Random Forest, Decision Tree, SVM, k-Nearest Neighbors, Artificial Neural Network and Gaussian Naïve Bayes.

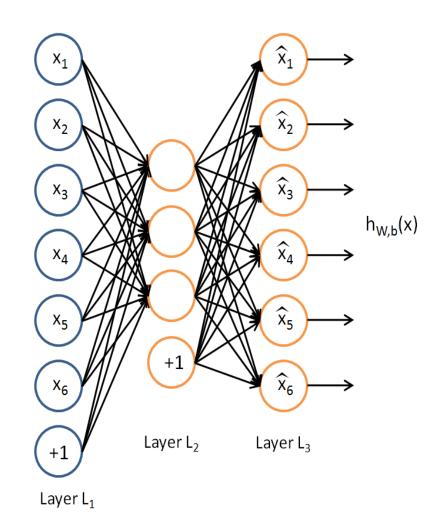


Features Visualization



Anomaly Detection

- Intuition: create a model that faithfully reproduce normal traffic
- Model attempts to minimize the reconstruction error of the signal
- One anomaly detector per device





Future Work

