Machine Learning-Based Fingerprinting of Network Traffic Using Programmable Forwarding Engines

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Goal

Explore the efficacy of classifying large amounts of network traffic using PFE-generated, rich flow records in two separate applications

- Ransomware identification and classification
- Censorship circumvention traffic fingerprinting

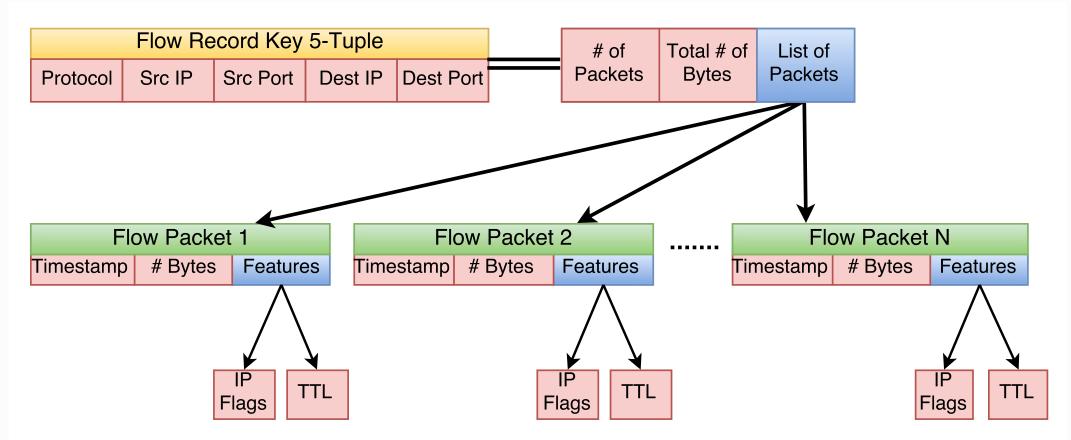
Programmable Forwarding Engines (PFEs)

- Allow commodity network equipment to support the scalable generation of rich flow records
- Stream processing systems utilize PFE switch hardware to process network data at high-rates of speed and extract vital, per-packet flow information
- Provide system designers with the data and speed necessary for network, flow-based traffic analysis and fingerprinting

Compact, per-packet flow records

PFE flow record overview

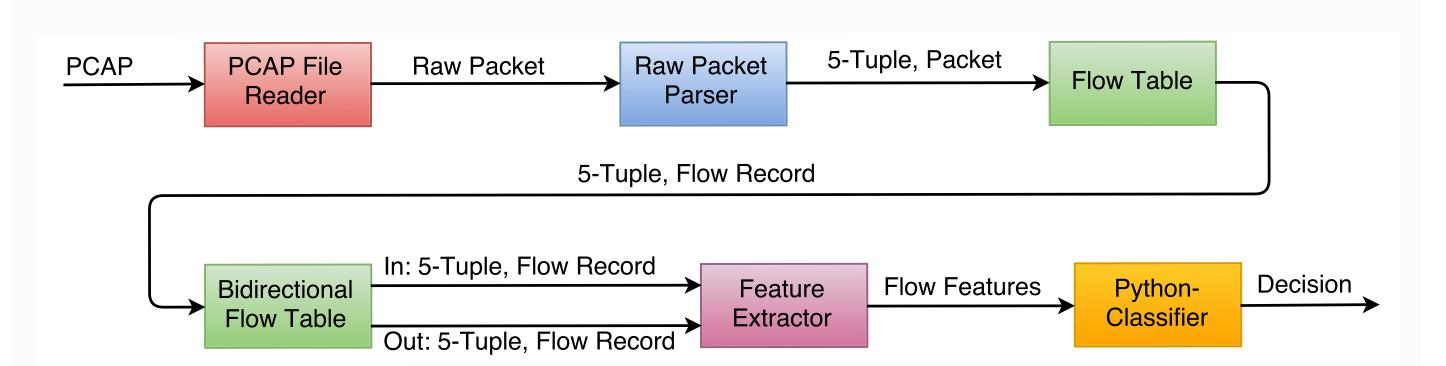
- The data extracted from a flow can be tailored to fit a user's specific application



Flow Records and Processing

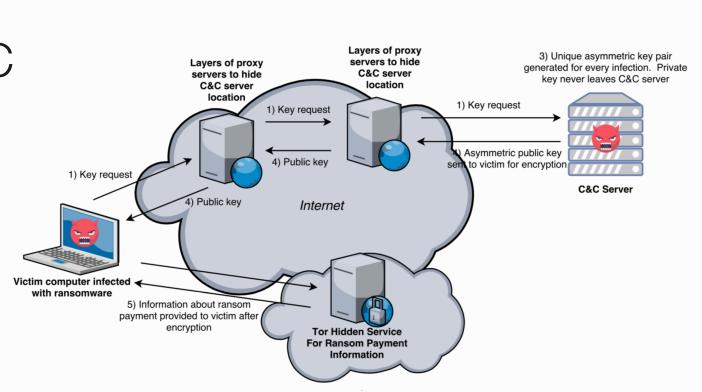
Stream processor

- 5 kernel stream processor
- Simulates PFE-generated compact, rich flow records
- Extracts vital features from flow records
- Feeds into Python classifier



Ransomware Overview

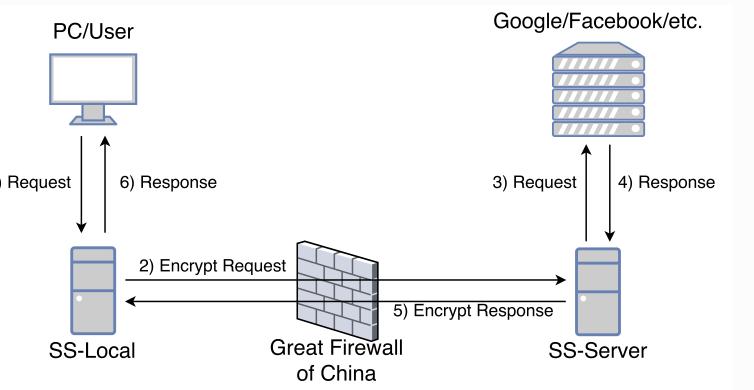
- Victim makes initial key request to C&C server
- C&C server returns encryption key
- Tor hidden service communicates method of payment



[Cabaj, Mazurczyk. CoRR '16]

Shadowsocks (SS) Overview

- Censorship circumvention tool
- User sets up own proxy server outside GFW domain



Ransomware Classification Results

Goal

- Minimize false negatives
- Balance FNR and FPR

Random Forest Classifier

- 40 decision trees with depth 15

Performance

- Precision: 0.89 F1 Score: 0.87
- Recall: 0.83 AUC of ROC: 0.93



Key Flow Features

- Packet interarrival times
- Inflow to outflow packet ratios
- Burst lengths
- Flow duration

Shadowsocks Classification Results

Goal

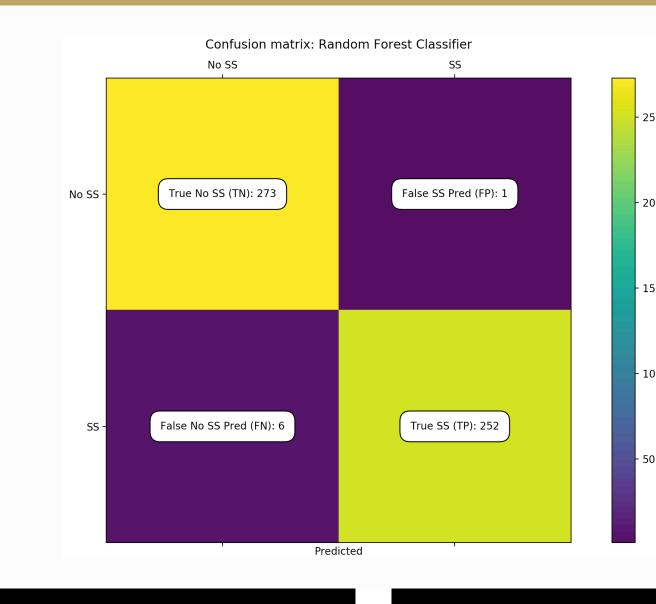
- Minimize false positives

Random Forest Classifier

- 10 decision trees with depth 10

Performance

- Precision: 0.996 F1 Score: 0.987
- Recall: 0.977 AUC of ROC: 0.99



Key Flow Features

- Packet interarrival times
- Traffic latency
- Payload entropy

Discussion

Takeaway

- Preliminary results show efficacy of utilizing high-rate PFEgenerated, rich flow records to fingerprint different types of web traffic

Future Work

- Write classifiers in C++ for line rate classification
- Continue to explore other classification techniques

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