



Visualizing Android Feature Maliciousness through time



Michael Tegegn

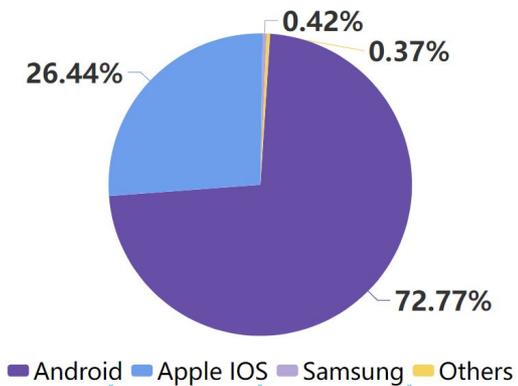


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Android

- Most widely used Mobile OS

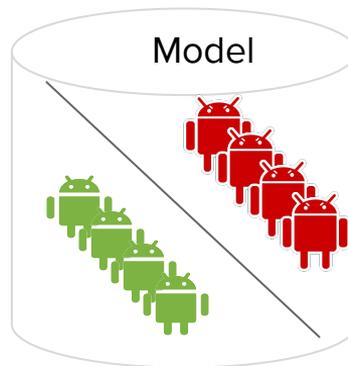
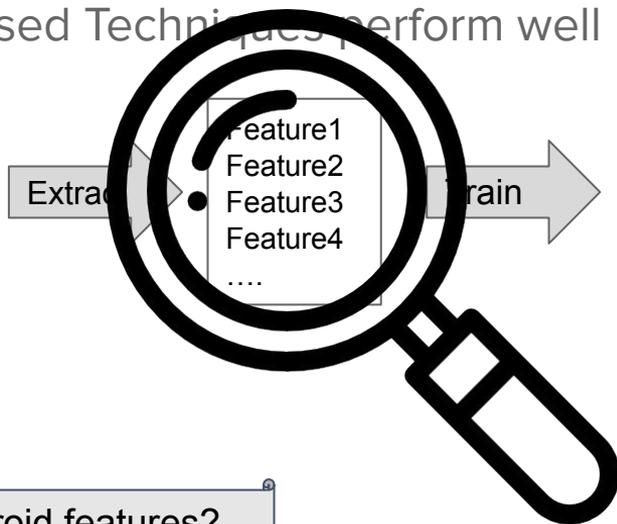
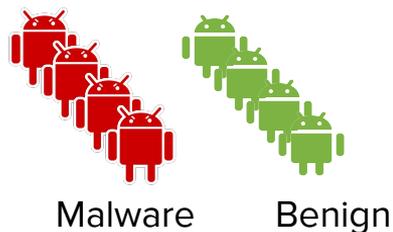


- 12,000 new Android malware instances every day. [unb](#)

Android Malware Detection

- Machine Learning Based Techniques perform well

Training Samples



Why visualize android features?

- Aid in Feature Selection
 - Eg. Selecting features more common in malware applications can help boost robustness [1]

Sample Selection: Which samples to train on?

- We want to identify future malware

Seen Samples



Future Samples



- We can just train on all existing benign and malware samples

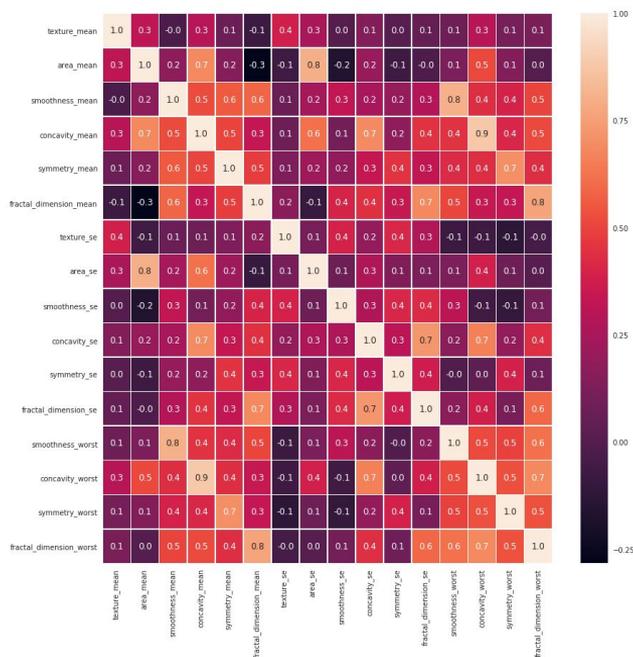


Why visualize android features through time?

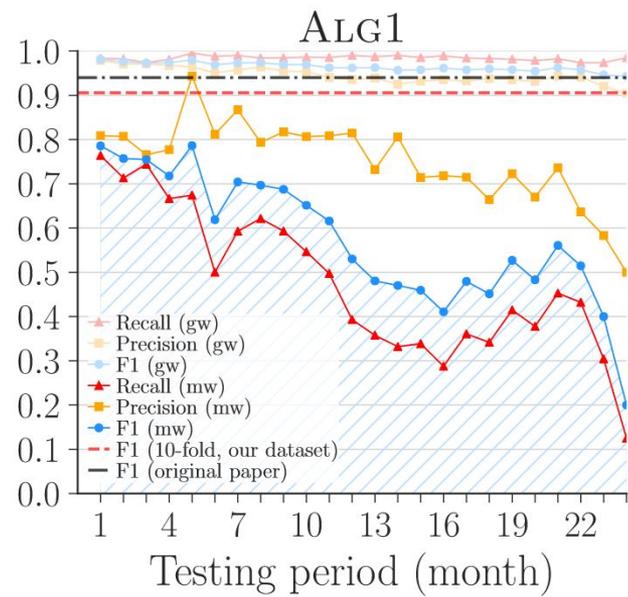
- Identify features that can help detect future malware [2]

Related Work: Visualizations

1. Correlation Between Features



2. Model Performance through time



Data:

- Android Applications represented as a set of (DREBIN) features
 - Benign and Malware Android applications
 - Eg. app1: { feature1, feature3, feature5 } label: **benign**
- Feature selection metrics and their results
 - Features ranked according the feature selection metrics
 - Eg. Mutual info metric: {feature3, feature1, feature2}

Derived Attributes:

$$\text{Maliciousness of Feature } x = \frac{\text{Malicious Apps with } x - \text{Benign Apps with } x}{\text{Total Number of Apps with } x}$$

- -1 => x only in malware apps
- 1 => x only in benign apps

$$\text{Normalized freq of Feature } x = \frac{\text{Apps with Feature } x}{\text{Total Number of Apps}}$$

- 1 => All apps contain x
- 0 => No apps contain x

Data Abstraction Summary

<u>Attribute</u>	<u>Kind</u>
Feature	Categorical
Feature Set / Feature Family	Categorical
App Development Year	Ordinal
Feature Maliciousness	Quantitative
Feature Normalized Frequency	Quantitative

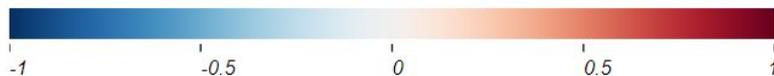
Goal: Identify feature trends in android applications

Target Group: Malware detection tool developers

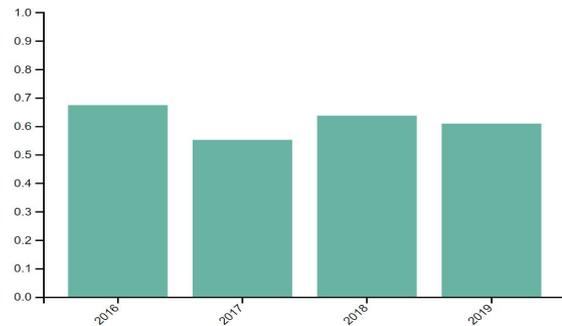


Visualization Design: Encodings

- Feature Maliciousness over time using heatmap
 - **y-axis:** Feature **x-axis:** Development year
 - Maliciousness encoded by **Blue-Red Diverging Scale**



- Normalized frequency over time using bar charts
 - **y-axis:** Feature **x-axis:** Development year
 - **Bar height:** Normalized frequency of a feature



Normalized Frequency per year of selected feature

Visualization Design: View Manipulation

- **Requirement 1:** Accommodate for large number of features
 - Design Solution
 - **Filter** features based on feature set
 - **Order** features based on popularity, maliciousness, ...
 - **Alter view** using scrolling

- **Requirement 2:** Show selected features

- Design solution
 - Show selection using **dashed border lines**



- **Requirement 3:** Display normalized frequency of a feature on demand
 - Design solution
 - Add side view bar chart upon selection

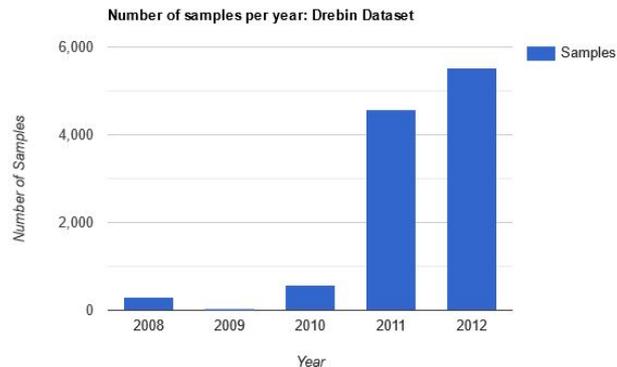
Demo

<http://127.0.0.1:5500/index.html>

Limitations and Future Work

Limitations:

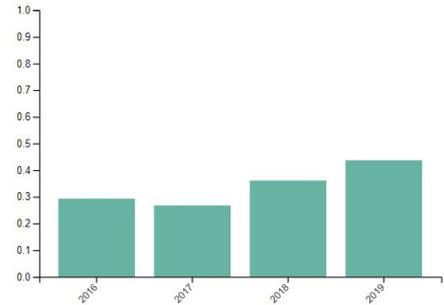
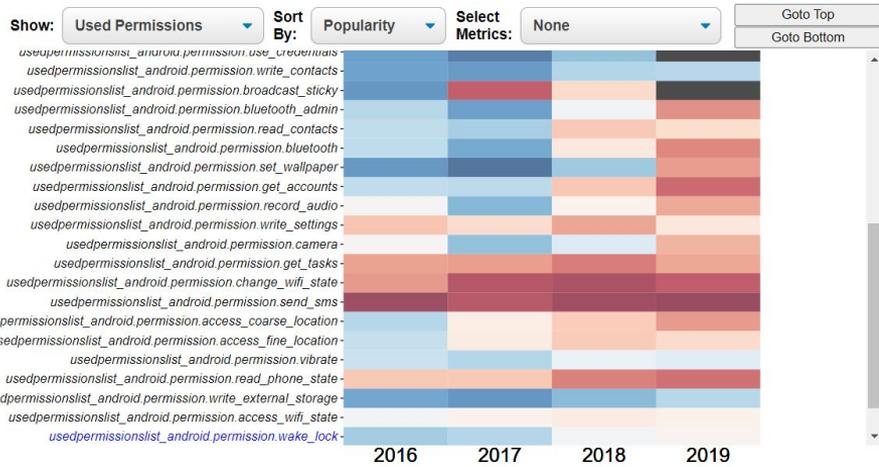
- Sampling bias in dataset
 - Eg. Drebin Dataset:



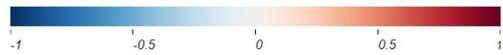
- Short time duration for the two datasets (VT: 4 years, DREBIN: 5 years)
- Still a large number of features (Scrolling required)

Future work:

- Hand select features and train models directly on the vis interface
- Extend for any domain that requires analysis of features for feature selection



Normalized Frequency per year of selected feature
usedpermissionslist_android.permission.wake_lock



Thank you!

Questions?