AnDarwin: Scalable Detection of Semantically Similar Android Applications

Jon Crussell, Clint Gibler, Hao Chen

UC Davis Computer Security Lab

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Talking Leonard Tig PhoneLiving LLC



Talking Christopher PhoneLiving LLC



Talking Diddy Dog PhoneLiving LLC



Talking Pepe Pengu PhoneLiving LLC

Smartphones Abound













App Plagiarism

Prior work¹ showed apps are often plagiarized to siphon ad revenue

· Leveraged mobile ad traffic from major US provider

On average, developers lost:

- 14% Ad revenue
- 10% User base



¹Gibler et. al, MobiSys 2013

Plagiarism Harms the App Ecosystem

Plagiarist Goals:

- Siphon ad revenue from legitimate apps
- Infect users with malware
- Phish users

Developers:

Lose revenue and incentive to develop apps

Markets:

Polluted search results

Users:

- Lose sensitive information
- Difficult to find useful, well-supported apps

Differences of App Plagiarism

- High-level bytecode
 - Significant modification of apps feasible²
 - Static analysis tractable
- Large scale
 - Previous work often focuses on localized problems:
 - Students' homeworks³
 - Synthetically created software "clones"
- Libraries embedded with app code
 - Difficult to differentiate without external knowledge

²Davis et. al, MobiSys 2013

³MOSS: http://theory.stanford.edu/~aiken/moss/

Previous Approaches: Not Robust

DroidMOSS (Zhou et. al, CODASPY 2012)

- Compares fuzzy hashes computed over the bytecode Juxapp (Hanna et. al, DIMVA 2012)
- Compares k-grams of opcodes in disassembled bytecode PiggyApp (Zhou et. al, CODASPY 2013)
 - Compares APKs using vectors of used API methods

Previous Approaches: Not Scalable

DNADroid4

- Pair-wise app comparison based on PDGs
- Uses subgraph isomorphism to compare PDGs (NP-C)
- Relies on meta data to pick candidate app pairs

⁴Crussell et. al, ESORICS 2012

App Cloning:

• Taking an existing app, modifying it and republishing it

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Our goal: find clones of Android apps based on code similarity

Android

Android

- Smartphone operating system
- Open-source, developed primarily by Google
- Official and third-party markets for apps
- Over 900 million devices and 975,000 apps⁵

Android apps

- Developed primarily in Java, run on Dalvik VM
- Some (< 10%) include native code



⁵http://www.android.com

AnDarwin

Goals:

- Scalably find clusters of apps with similar byte code
- Robust against plagiarist evasion techniques
- Low false-positive rate

Non-goals:

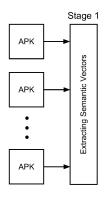
- Determine similarity of native code
- Determine which app is the original

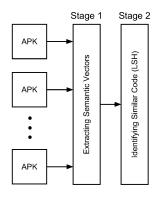


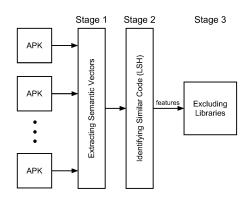


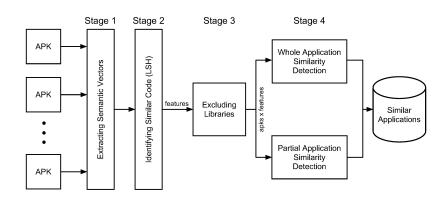


APK







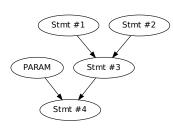


Program Dependency Graph (PDG)

PDG represents a method in a program

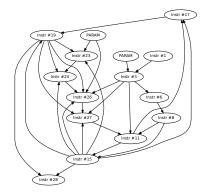
- Nodes: statement in the method, many different types:
 - Arithmetic operations
 - Branch, return, call
 - ...
- Edges: show data dependencies between statements

```
0: def foo(arg0)
1:    var x = 0
2:    var y = 1
3:    var z = x + y
4:    return z + arg0
```



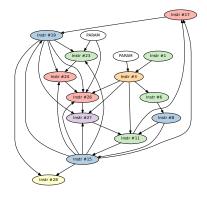
Semantic Vectors

PDG:



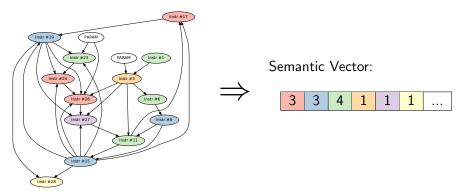
Semantic Vectors

PDG:



Semantic Vectors

PDG:



Clustering Semantic Vectors

Methodology:

- Extract semantic vectors from all apps
 - Use dex2jar to convert Dalvik byte code to Java byte code
 - Use WALA⁶ to construct PDGs
- Cluster vectors using Locality Sensitive Hashing

⁶T.J. Watson Libraries for Analysis

Clustering Semantic Vectors

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Clusters of semantic vectors used as features to detect similar apps

PDGs within clusters are likely similar

⁶T.J. Watson Libraries for Analysis

Locality Sensitive Hashing (LSH)⁷

A family of hash functions such that for any two points, p and q:

$$\begin{aligned} &\text{if } ||p-q|| \leq R \text{ then } Pr[h(p)=h(q)] \geq P_1 \\ &\text{if } ||p-q|| \geq cR \text{ then } Pr[h(p)=h(q)] \leq P_2 \\ &P_1 > P_2 \end{aligned}$$

Used as approximation algorithm to find nearest-neighbors (NN):

- Approximate ⇒ probability some NNs will be missed
- Preprocessing: compute hash values for all points
- Time complexity for single NN search: sublinear
- Form clusters by querying NNs of each point

⁷Andoni et. al, FOCS 2006

App Feature Matrix

			Apps						
		app ₁	app ₂	app ₃	app ₄	app ₅		app_N	
	f_1	✓			✓				
	f_2		✓					√	
	f_3	✓			√				
SS	f_4			✓		✓			
tur	f_5		\checkmark			\checkmark			
Features	f_6		\checkmark					\checkmark	
"	f ₇			√		✓			
	f_8							√	
	f_{M}	√			√				

App Feature Matrix

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		app ₁	app ₂	app ₃	app ₄	app ₅		app _N	
	f_1	✓			✓				
	f_2		√					√	
	f ₃	✓			✓				
SS	f ₄			✓		✓			
ture	f_5		✓			✓			
Features	f ₆		✓					✓	
-	f ₇			✓		✓			
	f ₈							✓	
	f_{M}	✓			✓				

Observation: Matrix is sparse

Full App Similarity Detection

Goal: Find apps with many shared features

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		Apps						
		app ₁	app ₂	app ₃	app ₄	app ₅		app∧
	f_1	✓			✓			
	f_2		✓					✓
	f ₃	✓			✓			
SS	f_4			✓		✓		
ture	f_5		√			✓		
Features	f_6		✓					✓
"	f ₇			✓		✓		
	f ₈							✓
	f_{M}	√			√			

Full App Similarity Detection

Goal: Find apps with many shared features

		Apps						
		app ₁	app ₂	app ₃	app ₄	app ₅		app∧
	f_1	\checkmark			\checkmark			
	f_2		\checkmark					✓
	f ₃	\checkmark			\checkmark			
SS	f_4			\checkmark		\checkmark		
Features	f_5		\checkmark			\checkmark		
ea.	f_6		\checkmark					√
-	f ₇			\checkmark		\checkmark		
	f ₈							✓
	f_{M}	√			√			

MinHash

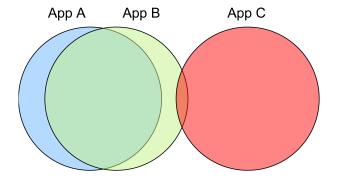
Broder et. al present MinHash to find similar sets⁸

- Member of LSH family
- Designed to estimate Jaccard coefficient of sets
- Requires a sparse feature matrix
- Time complexity: O(N log N)

⁸Compression and Complexity of Sequences 1997

Problem:

Jaccard coefficient
$$\equiv J(A, B) = \frac{|A \cap B|}{|A \cup B|}$$



Partial App Similarity Detection

Goal: Find apps that share common module of code

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		Apps						
		app ₁	app ₂	app ₃	app ₄	app ₅		app∧
	f_1	✓			✓			
	f_2		✓					✓
	f ₃	✓			✓			
SS	f_4			✓		✓		
ture	f_5		√			✓		
Features	f_6		✓					✓
"	f ₇			✓		✓		
	f ₈							✓
	f_{M}	√			√			

Partial App Similarity Detection

Goal: Find apps that share common module of code

			Apps						
		app ₁	app ₂	app ₃	app ₄	app ₅		app∧	
	f_1	✓			\checkmark				
	f_2		✓					\checkmark	
	f ₃	✓			✓				
SS	f_4			✓		√			
tur	f_5		√			√			
Features	f_6		\checkmark					\checkmark	
-	f ₇			✓		✓			
	f ₈							\checkmark	
	f_{M}	√			√				

Library Code

Apps may contain one or more libraries

Common library code should **not** be considered for clone detection

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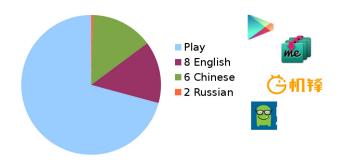
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Methodology:

- Calculate number of apps each feature appears in
- Exclude features (rows of matrix) using threshold
 - Must be larger than expected number of clones of single app
 - Must not be too large to avoid missing less common libraries

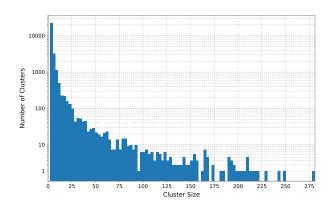
Dataset

We developed crawlers for Google Play and 16 Third-Party Markets We downloaded a total of 265,000 free apps:



Full App Similarity Detection Results

28,000 clusters consisting of 150,000 apps in ten hours



Finding: Rebranded Apps

Authors cloning their own apps to cater to different audiences:



Finding: Rebranded Apps

Authors cloning their own apps to cater to different audiences:



Total Number of Rebranded Apps: 36,106

Finding: Plagiarism Victims

Clusters of apps with multiple authors:

Market	Developer	Package	Key	Ad Library
Play	Rhythm Software	com.rhythm	fce850	admob
AndroidOnline	Rhythm Software	com.rhythm	fce850	admob
Freeware Lovers	Rhythm Software	com.rhythm	fce850	admob
Play	useful tools	com.uninstall	92cced	admob, airpush
Play	Android Tools	com.soft	92cced	admob

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Play	useful tools	com.uninstall	92cced	admob, airpush
Play	Android Tools	com.soft	92cced	admob

Total Number of Plagiarism Victims: 4,295

Verification

Leverage DNADroid

- Robust tool based on PDGs
- Scalability limited
- Experimental False Positive Rate: 0%
 - Manually compared 140 app pairs

Tested 25,000 pairs from 6,000 random clusters

Experimental False Positive Rate: 3.73%

Conclusion

Contributions:

- Developed approach for detecting similar apps
 - Analyzes apps at an unprecedented scale
 - Considers both full and partial similarity
 - Uses only code similarity without relying on metadata
 - Low false positive rate
- Use cases:
 - Single author: rebranded apps
 - Multiple authors: plagiarized apps

Coming soon: sherlockdroid.com