



Institut
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Identifying Unknown Android Malware with Feature Extractions and Classification Techniques

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Seminar at SAP



Many Android Applications (and Malware!)

Application repositories

- ▶ Google Play: 1.7 million+
- ▶ F-Droid, APPSAPK, APKTOP, ...



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Malware

- ▶ Aug. 2015. **2.7 millions+** malicious Android **samples**
- ▶ **2,000+** new malicious **Android samples every day**

Known Malware

I'M A MALWARE
e.g I pose as a Skype installer



We are **KNOWN MALWARE**
and it's usually easy to
detect us

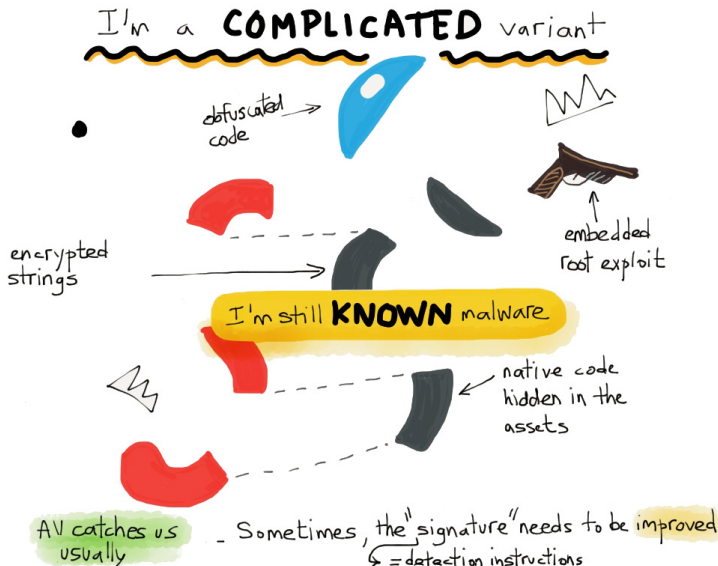


I'm a
minor
variant
"Trust me, I
install Opera
Mini"

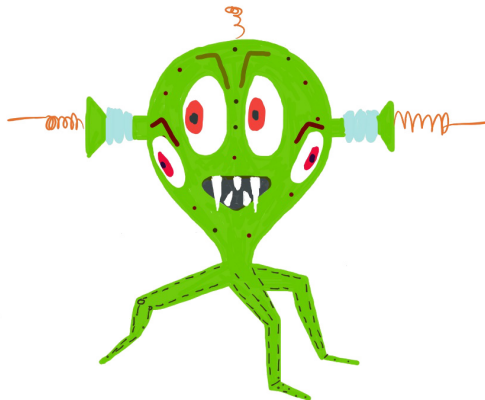


I'm
another
minor
variant
e.g I'm just repackaged
(like my new dress?)

Known Malware



Unknown Malware



Do they exist? YES

Malware: Android Carbon 14 Dating ;)



Shortest detection delay for some samples by **all AV vendors**

Name	Creation date	Detection date
Android/Wroba	June 16 2014	June 21 +5d
Android/Curesec	July 3 2014	July 11 +8d
Android/ScarePakage	July 13 2014	July 24 +11d

Malware: Android Carbon 14 Dating ;)



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Android/ScarePackage	July 13 2014	July 24 +11d
Android/Ganlet	Nov 1 2013	May 15 2014 +6 months!!!

What Are We Interested In?



KNOWN (and detected)
MALWARE



NOTHING TO DO
(BORING...)



MINOR VARIANT
(usually detected)



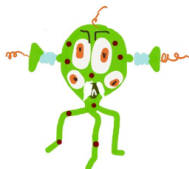
NOT VERY
INTERESTING
FOR RESEARCH



MAJOR VARIANT



AH HA ?!



UNKNOWN MALWARE



**TOP
INTEREST**

RESEARCH / PR
INNOVATION

Problems with Manual Search

Too many apps and marketplaces to crawl
Waste time on clean apps
Even a **team of 100 analysts** is insufficient

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Too many apps and marketplaces to crawl
Waste time on clean apps
Even a **team of 100 analysts** is insufficient

**We need an automated system that helps
identifying unknown malware with less effort**

→ SherlockDroid

SherlockDroid to the Rescue!

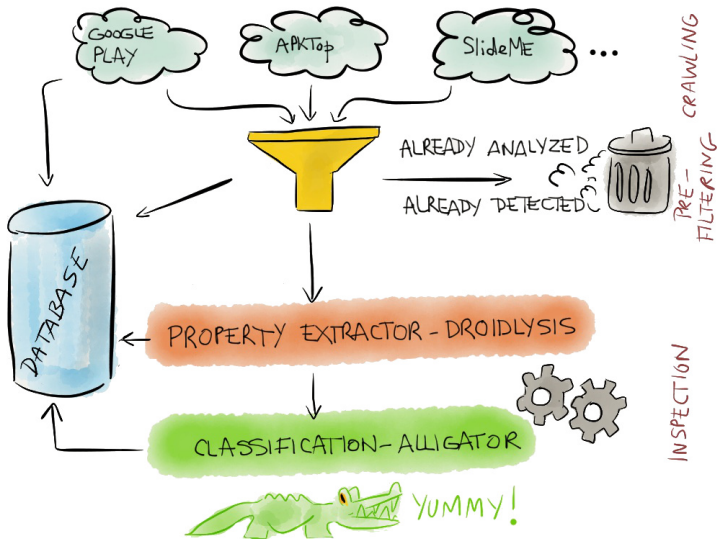


Crawl Android marketplaces

Spot suspicious apps

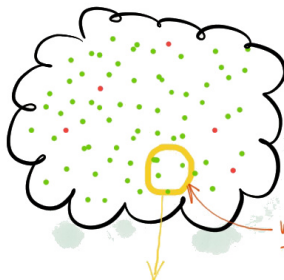
Focus on major variants and unknown malware

SherlockDroid Architecture



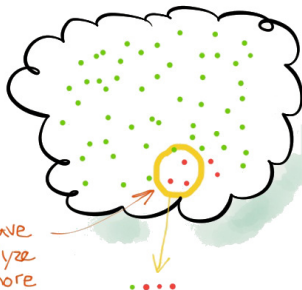
SherlockDroid (Unbiased) Benefits

WITHOUT
SHERLOCKDROID



WE WASTE OUR PRECIOUS
TIME ON CLEAN SAMPLES
(and usually don't have time
to find nasty samples)

WITH
SHERLOCKDROID



HIGHER CHANCES TO
SPOT INTERESTING
MALWARE
(it can't be perfect, though)



Remarks on SherlockDroid

It is not an AV scanner
because SherlockDroid does not handle known
malware / minor variants



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*But we would have missed them without
SherlockDroid too*

Crawlers - Evading Detection

Easy to implement
but constantly needs to be maintained :(

- Your IP: [REDACTED]
- URL: www.appsapk.com/android/all-apps
- Your Browser: libwww-perl/6.03
- Block ID: **BNP002**
- Block reason: Scanning tool access attempt.
- Time: Fri, 20 Jun 2014 05:30:21 -0400
- Server ID: **cp76**

- ▶ Search Limit
- ▶ Download activity per IP address
- ▶ User Agent verification
- ▶ Android ID verification <https://github.com/Akdeniz/google-play-crawler>

DroidLysis - Extracting Properties/Features

Static extraction of 289 properties



54 File-related properties

- ▶ Permissions, certificate, ...

22 Resource properties

- ▶ Native code, resource risky calls (*su*, *mount*, etc.), Javascript, URLs, ...

70 Dalvik code properties

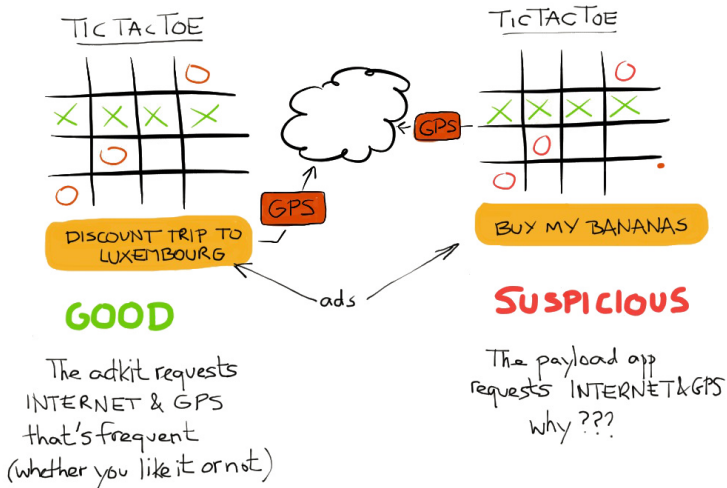
- ▶ API usage, actions, intents, constants, implementation techniques (e.g., *junk bytecode injection*)

143 Third party kits properties

- ▶ Advertisements, statistics reporting, error reporting

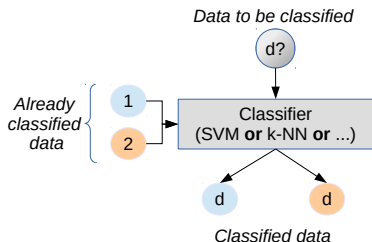
Ruling out Third Party Code

DIFFICULT TO SPOT THE DIFF

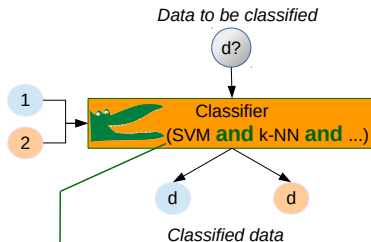


Alligator - Classification (1/2)

Usual classification approach



Alligator



Alligator automatically combines classification algorithms in order to obtain better classification results

Alligator - Classification (2/2)

Other capabilities

- ▶ Favor a cluster over another
- ▶ Forget/boost too abnormal elements

Other

- ▶ Shown to better classify than other classifiers (e.g., SVM) in various application domains (e.g., image classification)
- ▶ Free and open-source, easy to install and configure, scriptable



alligator.telecom-paristech.fr



SherlockDroid: Hall of “Fame”

- ▶ Android/MisoSMS.A!tr.spy
- ▶ Android/Odpa.A!tr.spy
- ▶ Adware/Geyser!Android
- ▶ Riskware/Flexion!Android
- ▶ Riskware/SmsControlSpy!Android
- ▶ Riskware/Zdchial!Android
- ▶ Riskware/SmsCred!Android
- ▶ Riskware/Blued!Android
- ▶ Riskware/SneakFont!Android

Descriptions: <http://www.fortiguard.com/encyclopedia/>



SherlockDroid: Unknown Malware Identified

*Do you know any other framework who identified
real unknown malware?*



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Answer: DroidRanger: 2



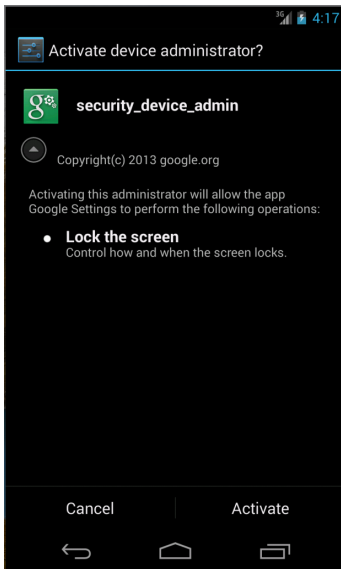
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Answer: DroidRanger: 2

AAS, Andromaly, CopperDroid, Crowdroid, Drebin, MADAM,
MAST, pBMDS, PUMA...
tested on *artificial or known malware*

Into Android/MisoSms Trojan Spyware



Android/MisoSms.A!tr.spy

- ▶ Poses as Google Settings app
- ▶ Sends 1 initial email with phone number of victim
- ▶ Listens to incoming SMS
- ▶ Forwards them by email to attackers

Into Geyser Adware

```
▼ Hypertext Transfer Protocol
  ▶ HEAD /?widgetid=[REDACTED]&guid=[REDACTED]&v=0.84.13498.7218&hid=null&tlat=0.0&tlon=0.0&test=1 HT
    User-Agent: Dalvik/1.2.0 (Linux; U; Android 2.2; sdk Build/FRF91)\r\n
    Host: ads.[REDACTED]ser.com\r\n
    Connection: Keep-Alive\r\n
    \r\n
```

Adware/Geyser!Android

Posts GPS location in clear text

[http://blog.fortinet.com/post/
alligator-detects-gps-leaking-adware](http://blog.fortinet.com/post/alligator-detects-gps-leaking-adware)

LOL - In falsepositives.txt:

"Reputable companies including banks, US Government/ Military sector are using our tools"

Learning and Classification Results

Typical results we expect

- ▶ FP/FN shall be as low as possible (Obviously)
- ▶ FP shall be much lower than FN (Missing a malware is not a big deal w.r.t. wasting time on false alerts)

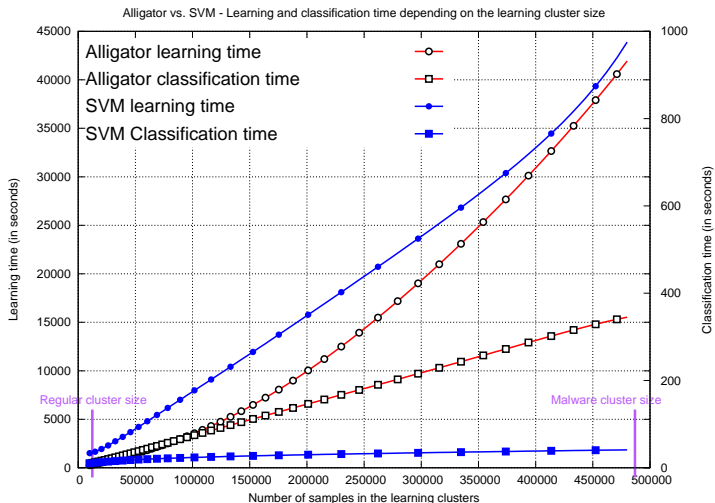
Samples

- ▶ **Learning clusters:** 500k samples used in the learning clusters
 - ▶ ~ 487k malware, ~ 12k clean
 - ▶ Gathered before June 2014
- ▶ **Testing clusters:** 1.5k clean and 3k malware gathered after Sept. 2014

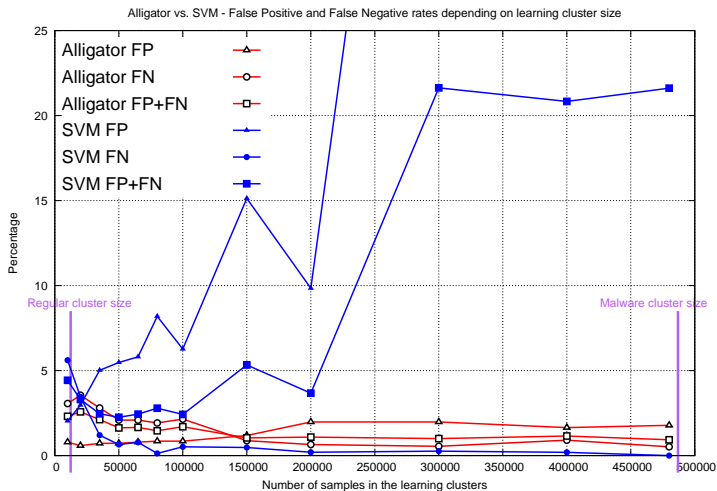
Learning and Classification Results (Cont.)

Learning cluster size	Learning time	Classification time	FP	FN	Average
480,000	~ 11 h	6 mn	1.78%	0.52%	0.93%
50,000	20 mn	~ 34 s	0.72%	2.1%	1.67%

Learning and Classification Results: Comparison



Learning and Classification Results: Comparison (Cont.)



What About the SAP Android App?

```
xterm
-,===,oo= 7/8 93% Computing list of interesting correlations of malware | 87M
-,===,oo< 7/8 93% Computing list of interesting correlations of malware | 87M
-,===,oo= 7/8 94% Computing list of interesting correlations of malware | 87M
-,===,oo< 7/8 94% Computing list of interesting correlations of malware | 87M
-,===,oo= 7/8 94% Computing list of interesting correlations of malware | 87M
-,===,oo< 7/8 94% Computing list of interesting correlations of malware | 87M
-,===,oo= 7/8 95% Computing list of interesting correlations of malware | 87M
-,===,oo< 7/8 95% Computing list of interesting correlations of malware | 87M
-,===,oo= 7/8 95% Computing list of interesting correlations of malware | 87M
-,===,oo< 7/8 95% Computing list of interesting correlations of malware | 87M
-,===,oo= 7/8 97% Computing list of interesting correlations of malware | 87M
-,===,oo< 7/8 97% Computing list of interesting correlations of malware | 87M
-,===,oo= 7/8 97% Computing list of interesting correlations of malware | 87M
-,===,oo< 7/8 97% Computing list of interesting correlations of malware | 87M
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-,===,oo= 7/8 98% Computing list of interesting correlations of malware | 87M
-,===,oo< 7/8 98% Computing list of interesting correlations of malware | 87M
-,===,oo= 7/8 1% Computing list of interesting correlations of malware | 87M
Regular min percentage: 94
Malware min percentage: 94
Regular weight : 700
Malware weight : 7
```

- Property extraction:
17 seconds
- Classification:
4 seconds

What About the SAP Android App? (Cont.)



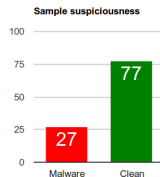
SHERLOCK<DROID ANALYSIS

Filename : b1.mobile.android.apk_38ccc092.apk

SHA256 : 2890685023c151dc35ee98d7cd4dd93772531e69a38bad0a505d28f98dda971f

Alligator

This sample looks clean
Analysis time: 318 seconds



→ Classified as regular ;-)

Conclusion and Future Work

- ▶ SherlockDroid is operational, tested on a huge number of applications from various application markets
- ▶ 9 unknown malware identified (Actually: 10!)
- ▶ For classification purpose, it relies on the Alligator meta-classifier

What's next?

- ▶ Feature extraction: mix contextual information (e.g., call stack) and the related features
 - ▶ Sending an email is not the same if it is for a bug report or for a connection to a Command&Control server
- ▶ Differentiate between malware and Potentially Unwanted Applications

Thank You

Contact info

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Alligator: ludovic dot apvrille at telecom minus paristech dot com

References

Alligator Release: alligator.telecom-paristech.fr

A. Apvrille, L. Apvrille, "SherlockDroid: a research assistant to spot unknown malware in Android marketplaces", Journal of Computer Virology and Hacking Techniques, vol. 11, No. 39, pages 1-11, pub. Springer, july 2015

Powerpoint slides? No way! This is L^AT_EX- Beamer !