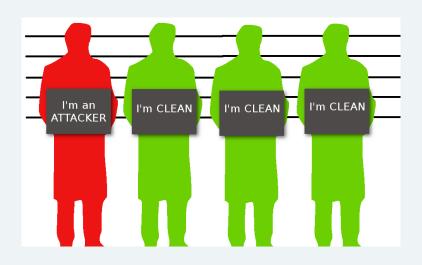
Crimina Profiling: Android Malware

Axelle Apyrille - FortiGuard Labs, Fortinet

Nuit du Hack, June 2015



Criminal Profiling





Plenty of **stats**



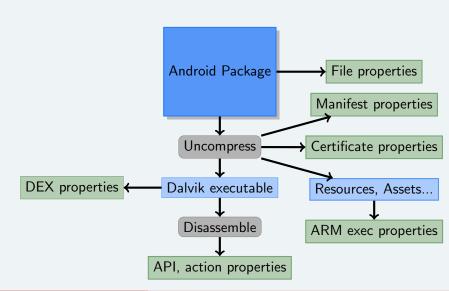






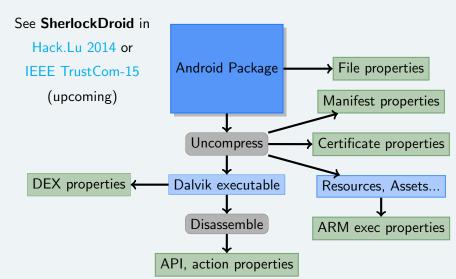
Please tweet stats correctly though:) Whenever possible, include how stats were computed: it matters (very much) Want to re-use? Sure - please credit (fair, isn't it?)

How are stats computed?

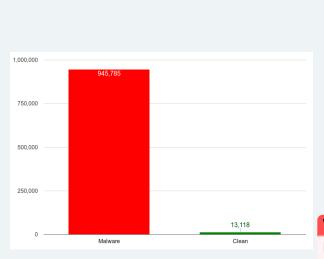


How are stats computed?

289 **static** properties



Datasets



- Malware: taken from Fortinet's DB

 unique & non damaged samples only
- ► Clean: apps we analyzed manually, open source apps, top apps with known developer in Play Store

Why so few clean?

Hey, it's very difficult (and long) to be sure it's clean!

Number of samples

Unless specified otherwise, we considered:

Property type	Nb of samples
Package properties	945,785
DEX format properties	945,785
API call properties etc	945,092
Manifest properties	617,942

Properties in 3rd party kits (AdMob, JUnit...) are ruled out

Why not all?

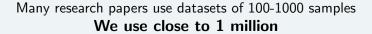
- ► Some samples are incomplete (e.g. just classes.dex)
- Some samples are damaged
- Some properties are 'optional' (e.g targetSDK)

Comparisons

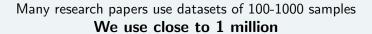
Many research papers use datasets of 100-1000 samples

We use close to 1 million





Android Malware Genome dates back to 2011 Our study is on samples collected before March 2015



Android Malware Genome dates back to 2011

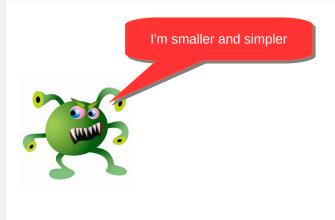
Our study is on samples collected before March 2015

Extensive work: Andrubis (BADGERS'14), PlayDrone (SIGMETRICS'14)

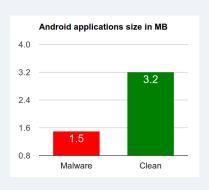
Our study focuses on malware with stats on code-level properties

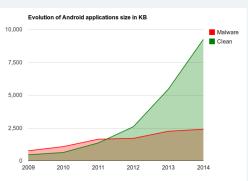
Criminal Profiling: What Do Malware Look Like?





Sample file size

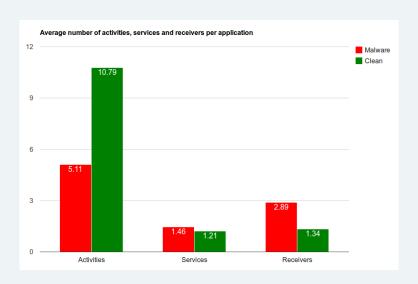




End of 2014

Clean: 9.2M average **4x bigger** than Malware: 2.4M average Malware don't need to implement all features

Activities, services, receivers





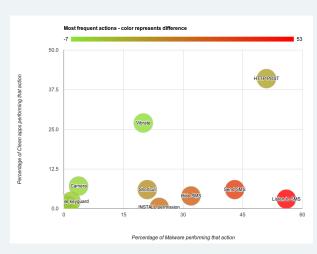
Criminal Profiling: What Do Malware Like?



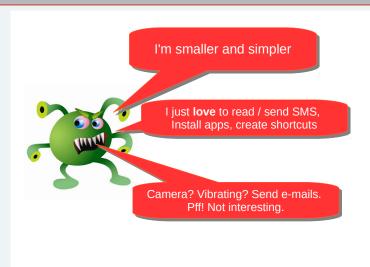


SMS: a strong indicator!

- ► 56% of malware implement a **SMS** receiver! (only 3% of clean)
- ► 43% of malware send SMS!
- 32% of malware use abortBroadcast() to conceal incoming SMS!



Criminal Profiling: Other Interests





What Malware Like / Don't Like

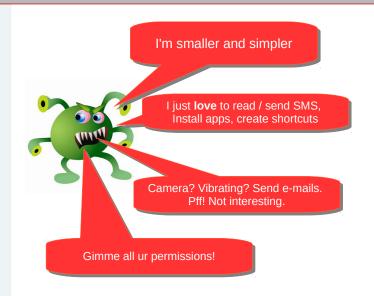




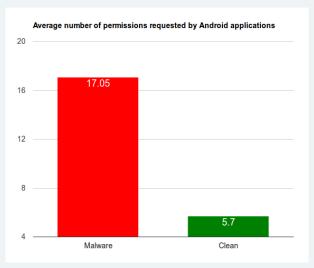
- INSTALL_PACKAGES: 24% malware ask for it. Only 0.4% clean apps do. NB. Works for system applications only.
- ► Install **shortcuts**: 21% malware, 6% clean apps.

- ► Emails. 14% malware < 29% clean (support/contact)
- ► **Vibrate**. 20% malware (ransomware?), 27% clean
- ► Is the era of **premium phone number** dialers over? 1%
- ► Camera. 3.7% malware, 7.1% clean. Only if you're a VIP ?;)
- Disable the keyguard. Malware can run background tasks as services...

Criminal Profiling: Your Permissions, or Your Life!

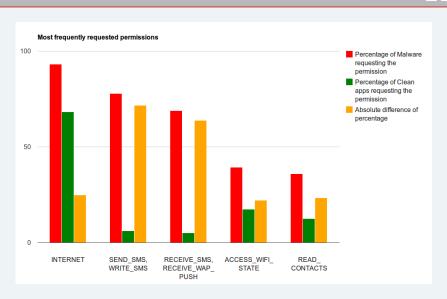


Permissions indicate evil will...



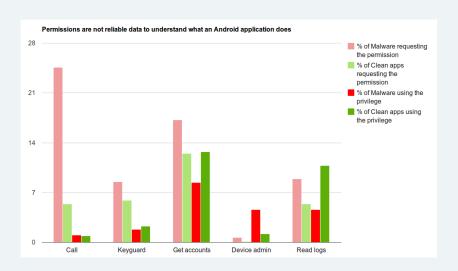
Clear over-use of permissions!!!

Top 5 permissions





Permissions are not so reliable





Why can't we rely on permission stats?

A permission may be requested but never used

Or the permission can be used within (legitimate?) third party code

Example: call permission vs ACTION_CALL/_DIAL

We don't have the manifest for all malware

Explains rare cases where use > request

Example: BIND_DEVICE_ADMIN permission vs

DeviceAdminReceiver

Bypassing permissions

- Call another app that has the permission
- ► Escalate privileges via updating
- Hijacking the Android installer
- ▶ Use an exploit...

Wide Target





Declared Target SDK

On average

- Malware target Gingerbread
- ► Clean apps target Jelly Bean

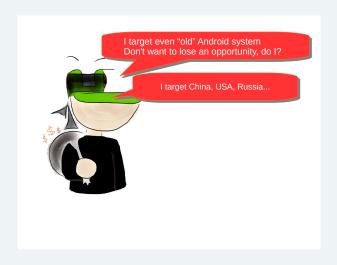
Stats

Considered 'only' 6,976 malware and 707 clean Why not 900K?

- ▶ All samples don't come with a manifest
- ► All manifests don't come with target SDK



Malware profiling: targets





Geographic attribution statistics

Amount of data

- Country of application's certificate (575,396)
- ▶ Rule out unknown countries, buggy and fake entries
 - e.g. GF is not a correct country code
 - e.g. VU is Vanuatu but this entry is probably fake: CN=VU
 OU=VU O=VU L=VU ST=VU C=VU
 - ▶ 63% ruled out!
- ▶ Rule out dev / debug certificates (12%)
- ► Remaining: **146,764** certificates. 14,919 in 2014, and only 6,308 in 2015 (incomplete).



Geographic attribution is complicated

Attribution script turned out to be tricky

Plenty of cases!

- ► Certificates using call codes (e.g. +86 for China) or zipcodes
- Match towns or 'states' to countries (e.g Gweru is in Zimbabwe)
- ▶ Deal with errors e.g C=CH for China, C=CA for California...
- Fixed several bugs, but probably others :((
 - ► C=gg-2 (fake country) was counting for ... Guernsay
 - ► C=asd3f21asdf was counting for American Samoa



Malware certificates: target or origin?

Examples

CN=Praveen Kumar Pendyala OU=Student O=IIT Bombay

L=Mumbai ST=Maharastra C=400076

CN=Dau Dinh Manh O=Song Vang L=Ha noi ST=Ha Noi C=84

 $CN=Zhong\ Zhang\ OU=Zhainanzhi\ Inc\ O=Zhainanzhi\ Inc$

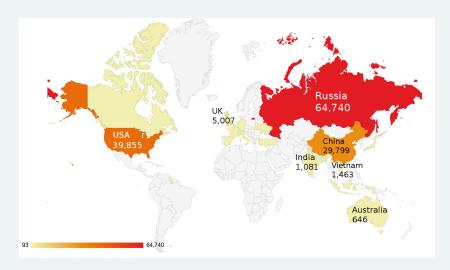
L=FuZhou ST=FuJian C=CN

- Many certificates with a seemingly valid identity
- Why mention a particular name?
 - ► For fame?
 - Because they don't believe their app is malicious?
 - Because they think we won't notice?
 - ► To complexify attribution?
 - Trojanized app where original certificate name was retained?

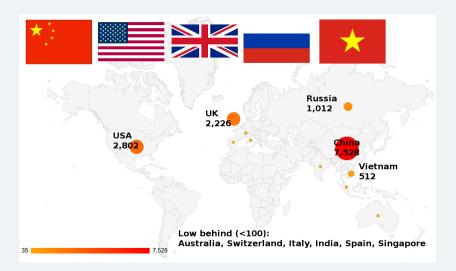


Presumed Targets of 146,764 malware



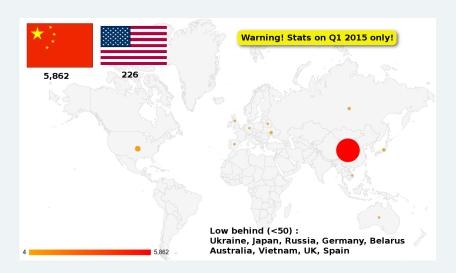


Top target countries in 2014





Top target countries in 2015



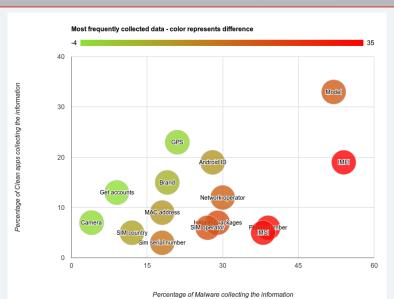


Information, I want information!





Most representative collected data





Collected Data











Not so obvious

We hadn't expected the diff with clean apps would be so strong:

- ▶ IMEI collected \approx 3 times more for malware
- ▶ Phone number, IMSI, S/N: **6 times more**
- ► List apps, SIM operator: 4 times more
- ► Android ID, MAC address: twice

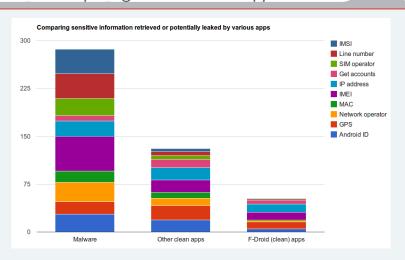
What reason for those???

- ▶ GPS (\approx 22% for both)
- ► Get accounts (9% malware, 13% clean)

Captain Obvious:

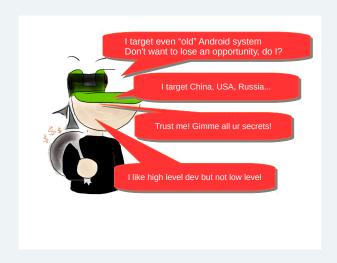
► IMEI, IMSI, Phone number...

Sidenote: comparing with F-Droid apps



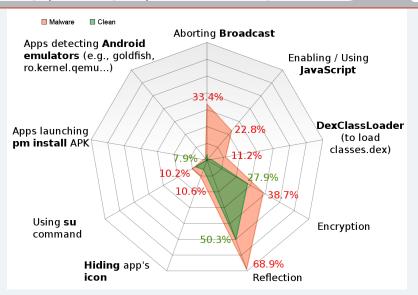
F-Droid (Free and Open Source Software Android apps) far cleaner than the average

Malware authors: how much skills?





Most frequent techniques



Reminder: code from third party kits are ruled out

Techniques: What Do We Make Out of It?

Malware authors are not Unix geeks:

- > su (8-10%), chmod (< 2%), mount (< 1%), busybox (\approx 1.5%)
- ► Command line installation pm install: only 2.2%
- ► Android emulator detection: only 1.4%

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Malware authors have development skills:

- Android SDK: abortBroadcast(), DexClassLoader, setComponentEnabledSetting()
- ▶ JavaScript (22.8% malware only 0.6% clean)



Techniques: surprises



Reflection: 68.9% malware, 50.3% clean

Encryption: 39.7% - 27.9%

Because they're old/well-known techniques?

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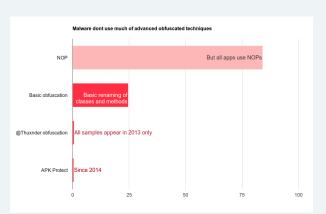
Because they're old/well-known techniques?

What are clean apps doing with openDexFile and loadDex?!

0.3% malware - 0.4% clean

Dalvik.system.DexFile - openDexFile() is private

Obfuscation: smaller than expected?



- NOPs are meaningless
- Basic obfuscation = ProGuard a, b, c renaming
- obfuscation (2012)
 = abusing linear
 sweep with
 fill-array-data
 = 0.5%. All 4,800
 samples in 2013.
- ► APKProtect: since 2014

Obfuscation (continued)

Reliable properties

nop opcode, APKProtect string, @thuxnder

if-eq v0, v0, +9
fill-array-data v0, +3
fill-array-data-payload

Unreliable property: basic obfuscation

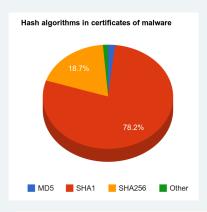
- ► AESObfuscator-1: used by Android LVL
- ► /a/a;->a: simplistic!!!

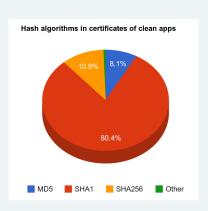
Issues

- ▶ NOPs mentioned by Mody (VB 2013)
- ► Lipovsky (CARO 2014) estimates all abusing linear sweep up to 30%
 - Seems too high
 - ► Unless I miss samples or case detections?



Hash algorithms of app certificates

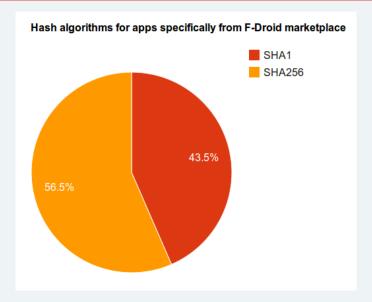




- ► malware: 617,942
- ▶ clean: 13,110
- ▶ Are malware authors more tech-savvy than regular developers?

Sidenote: F-Droid developers even more tech-savvy?





Use of exploits is not widespread

Detectors

Specific root exploits (Rage in the Cage, Levitator, Zerg Rush...) Generic (and very imperfect) exploit detector

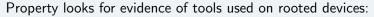
Result 1: my specific root exploit detectors don't work

Rage in the Cage	3
Exploid	4
Levitator	0
Mempodroid	0
Towel Root	0
Zerg Rush	0

Result 2: generic exploit detector works

Detected in 1.6% malware - I certainly miss cases though Yet, exploits are not widespread

Rooting is not specific to malware



- com.cyanogenmod
- com.noshufou.android.su
- Superuser.apk
- eu.chainfire.supersu

Both clean and malicious apps look for those $\approx 2\%$

ightharpoonup Stats computed on pprox 1 million malware. However, some properties (obfuscation, country...) are difficult to spot accurately.





- ► There's a general belief that malware are complicated (assembly, emulator detection, exploits etc). Statistically, this is wrong.
 - Rooting is not specific to malware
 - ▶ Unix commands, exploits, emulator detection < 2%</p>
 - Malware authors are skilled Android developers
 - ► They don't like low level dev + Unix





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 - Malware focus on their goals: money!
 - ▶ They are smaller (why code useless stuff?)





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 - They are smaller (why code useless stuff?)
- ➤ half malware read or send SMS, grab IMEI. They retrieve twice+ more sensitive data than clean apps
- ► Geographic attribution is difficult. Countries like China, Russia, USA, UK, Vietnam, Ukraine are top targets.



Thanks for your attention!



@cryptax or aapvrille (at) fortinet (dot) com

Thanks to



my husband Alligator, Lobster...

More

A. Apvrille, L. Apvrille, SherlockDroid: an Inspector for Android Marketplaces, Hack.Lu 2014

M. Lindorfer, M. Neugschwandtner et al ANDRUBIS - 1,000,000 Apps Later: A View on Current Android Malware Behaviors,

BADGERS 2014

N. Viennot, E. Garcia, J. Nieh, A Measurement Study of Google Play, SIGMETRICS 2014

