

# Using Static Analysis for Automatic Assessment and Mitigation of Unwanted and Malicious Activities Within Android Applications

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## Abstract

*In the last decade, smartphones have gained widespread usage. Since the advent of online application stores, hundreds of thousands of applications have become instantly available to millions of smartphone users. Within the Android ecosystem, application security is governed by digital signatures and a list of coarse-grained permissions. However, this mechanism is not fine-grained enough to provide the user with a sufficient means of control of the applications' activities. Abuse of highly sensible private information such as phone numbers without users' notice is the result. We show that there is a high frequency of privacy leaks even among widely popular applications. Together with the fact that the majority of the users are not proficient in computer security, this presents a challenge to the engineers developing security solutions for the platform. Our contribution is twofold: first, we propose a service which is able to assess Android Market applications via static analysis and provide detailed, but readable reports to the user. Second, we describe a means to mitigate security and privacy threats by automated reverse-engineering and refactoring binary application packages according to the users' security preferences.*

## 1. Introduction

The smartphone market has been growing steadily since the introduction of Apple iPhone in 2006, followed by a slew of Android devices since the end of

2008. In May 2011, the number of smartphone owners reached 76.8 million in the U.S., with Google's Android operating system running on 38.1% of those, making it the dominating mobile platform [2]. Akin to its competitors, applications for this platform are distributed through a central web service titled "Android Market", deployed and maintained by Google. Developers can submit executable binary packages to it, which are then published to a world-wide user base. No in-depth security analysis is performed by Google prior to market release, which makes the platform attractive for malicious and unwanted software writers.

Android is based on the Linux kernel, together with a strongly adapted userland. Applications run and maintain their data in isolated environments, and inter-process communication is strictly governed by a permission system. An application defines the list of capabilities it requires in a manifest file, which is presented to the user before installation. However, the user is not able to selectively grant or deny permissions to an application - instead, the only choice she has is to allow or deny installation. This presents a strong limitation of Android's security infrastructure, together with the fact that the permissions are very coarse-grained and do not provide sufficient insight of the actual, potentially malicious, activities performed after installation. Our analysis shows that at least 167 out of the top 1865 apps on the Android Market access private information such as the telephone number, accounting for around 9% of all applications. 114 applications of this set write private identifiers into a stream immediately after retrieval. However, the user is only informed of the fact that the application will

“read the phone’s state”. Since most end-users are not experts, this information is typically not considered suspicious, contributing to a high install base of such privacy-offending software.

In this work, we target the problem of detection and mitigation of such unwanted activities within Android applications. Our contribution is twofold - first, we describe a static analysis service which allows users to gain deep insight into applications’ internals while remaining readable and, in the first place, comprehensible for an end-user. Second, we show a novel mitigation strategy for this kind of activities. Since applications may still be useful to a user despite unwanted side-effects, we propose a method of disabling malicious features from an application while retaining its core functionality. This is a novel approach which works at application-binary-level, in contrast to popular system modification approaches prevalent in the current literature.

In section 2, we provide an overview of current research in the area of Android security and application analysis, as well as threat elimination. In section 3, our approach to detection and mitigation of malicious and unwanted activities within Android applications is presented. Section 4 provides an overview of a proof-of-concept prototype currently deployed by our research group. In section 5, we conclude and discuss future directions for follow-up research.

## 2. Related work

Estimating the impact of third-party software is a hard task which also fostered malware spread on smartphone platforms in the past. Several approaches to malware assessment exist and can be categorised in *dynamic analysis* and *static analysis* of binaries. While the static approach analyses the binary itself for malicious features, the dynamic rationale focuses on its behaviour while being executed.

Dynamic analysis can be applied to a production system - such as a user’s smartphone - as means of detecting malicious and/or anomalous activity at runtime. The analysis itself may be executed both off-device to preserve battery power [8], or on device to provide instant feedback, as demonstrated by TaintDroid introduced by Enck et al. [3]. This tool provides real-time analysis for identifying potential misuse of

users’ private information by analyzing flow of sensitive data in the system. However, these solutions allow potentially anomalous binaries to be executed on users’ devices. This limitation can be addressed by sandboxing and virtualization techniques [1], which allow thorough analysis of the applications’ effects in an isolated and controlled environment. However, when executed automatically, e.g. with random input, this technique may not be able to emulate the behavior of a user well enough to provide realistic results.

Static analysis of executable binaries is a well established technique [7]. Static analysis techniques have been applied to Android executables as well. Schmidt et al. propose a system which detects malicious native binaries by applying machine learning techniques to function call histograms. Enck et al. [4] focus on the more typical Java-based applications and provide an off-line analysis of 1,100 most popular applications from the Android Market together with a list of 27 security-related findings obtained by de-compilation and static code analysis. We tie in with this research and provide a comprehensive methodology for streamlined, automated binary assessment.

Malware mitigation strategies for Android proposed in current literature commonly focus on modifications to the operating system itself. Schmidt et al. [9] propose usage of native userland security methods known from traditional Linux systems. Shabtai et al. [10] focus on the kernel and describe the benefits of using SE Linux in Android to provide more robustness to the platform. Ongtang et al. [6] aim at the upper levels of Android application stack and describe modifications to its high-level security system to allow dynamic permission management. A more business-oriented approach is presented by Kuntze et al. [5], including usage of a trusted computing platform as a basis for an enterprise-grade secure smartphone based on Android. While creating a fork of Android for a specific purpose remains feasible, the history of large and frequent updates to Google’s official source mainline<sup>1</sup> has shown that maintaining an own development branch is a tedious and costly task.

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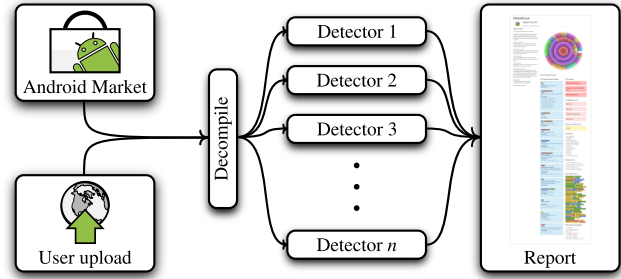
<sup>1</sup><http://developer.android.com/sdk/>

### 3. Static Analysis for Automated Binary Assessment and Malicious Activity Mitigation

In this work, we focus on two shortcomings of the Android platform, or, more specifically, its third-party software distribution. First, we provide a means of in-depth static analysis of Android applications, bundled with a human-readable reporting to support users in assessment of application. This way, we try to make it easier for the user to make the significant security decision - to install or not to install an application. Second, we provide a method to overcome the security and privacy threats introduced by individual applications. While the efforts mentioned in Section 2 focus on mitigation of the shortcomings of the platform, malicious and unwanted functionality may remain in third-party software. We address this problem by an unorthodox approach - modifying the application binaries instead of the system. Therefore, our approach does not require changes in the operating system and the underlying security methodology, enabling to preserve the vast amount of already existing third-party software without compromising security. In our approach, the software is analyzed for malicious and unwanted patterns in its decompiled form, and then modified at source-code level to match the user's preference or policy.

#### 3.1. Automated Binary Assessment

Our incentive is to provide extended static analysis of Android applications as a service to allow users to inform themselves about the potential security threats and privacy violations. Currently, all information which a user receives about the application prior to installation is comprised of two sources: *a)* free-text application description and screenshots provided by the developer and not checked by Google or any other authority, and therefore *not trustworthy*, and *b)* list of permissions required by the application to run, which is obtained from the package's metadata by the operating system itself and can therefore be considered *trustworthy*. However, the latter is very coarse-grained and only provides the user with generic information of what information or hardware functionality the application is allowed to access on the device. Neither is any information provided on what exactly the application intends to perform with the obtained data, nor



**Figure 1. Proposed workflow of the static analysis service.**

is the user informed of which - potentially private - data is concerned. We address this issue by providing deep static analysis of application as a service to the user. Our proposed system is depicted in Figure 1 and is comprised of the following steps:

1. The user requests a report on an application, which is obtained from the Android Market or uploaded by the user.
2. The application is unpacked and decompiled.
3. A set of pluggable detectors perform data mining and analysis operations
4. A report is generated from the detection results and provided to the user in a human-readable, comprehensible form.

The aforementioned detectors can be designed for both simple analysis tasks such as detection and identification of third-party libraries included in the distribution, as well as more sophisticated source-code analysis tasks which involve rich pattern matching. This should allow the user to obtain information about the internal workings of the application on a level previously unreachable by currently available solutions. For example, such a report can contain a list of included third-party advertising and analytical packages, as well as privacy leaks detected in the source code such as “*reads IMEI and writes it to a stream*” or security warnings such as “*uses non-standard networking API*” and “*package includes a native executable*”.

### 3.2. Automatic Mitigation

While in the analysis phase, we try to identify security related information in the decompiled binary, the obvious next step is mitigation of the detected potentially malicious activities within the application. The majority of privacy breaches we have detected among the most popular free Android applications can be mitigated by applying a patch to the decompiled binary without any impact on its core functionality.

In our vision, the user should be able to create a personal profile which contains her preferences regarding which of the detected suspicious activities have to be removed from the application. Then, given a set of detectors  $D := \{d_1, \dots, d_n\}$  defined as  $d_i : A \rightarrow \{T, F\}$  for the set of all applications  $A$  and  $1 \leq i \leq n$ , we define full set of mitigation strategies as a function:

$$m\_strategy : D \rightarrow M \cup \{deny\},$$

where  $M$  is a set of code modifications  $\{m_1, \dots, m_l\}$ . Now, assuming a set of personal preferences is denoted as a subset of  $D$ :

$$P \in \mathcal{P}(D),$$

we define a mitigation function:

$$mitigate : A \times \mathcal{P}(D) \rightarrow \mathcal{P}(M) \cup \{deny\}$$

for  $a \in A$  and  $P \subseteq D$  as follows:

$$mitigate(a, P) = \begin{cases} f(a, P) & \text{if } deny \notin f(a, P) \\ deny & \text{otherwise,} \end{cases}$$

where  $f : A \times \mathcal{P}(D)$  is a helper function defined as:

$$f(a, P) = \bigcup_{d \in P, d(a)=T} m\_strategy(d)$$

This means that if the users policy requires certain detected activities to be removed, but no mitigation strategy is available for at least one of these (i.e. *deny*), then the mitigation strategy for an application results in denial of installation. Otherwise, the patches are applied as intended, and a new, “sanitized” version of the application is created and provided to the user. As a small example, consider a set of detectors  $D = \{\text{uses\_reflection}, \text{reads\_IMEI}\}$  and a mitigation strategy defined as  $strategy(\text{uses\_reflection}) =$

*deny* and  $strategy(\text{reads\_IMEI}) = \{\text{patch}_1\}$ . Consider user Alice with a preference set is  $P_{\text{Alice}} = \{\text{reads\_IMEI}\}$ , user Bob with a preference set  $P_{\text{Bob}} = \{\text{uses\_reflection}\}$  and an application  $X$  with  $\text{uses\_reflection}(X) = T$  and  $\text{reads\_IMEI}(X) = T$ . Then, the mitigation function would return  $\{\text{patch}_1\}$  for Alice, and *deny* for Bob.

Modifying and redistributing application binaries is not permitted by most non-open-source licences. Since the Android Market does not provide a licencing identification information for the application it hosts, and its Terms of Service forbid redistribution of its contents, we are left with the following two possible deployment versions: *a)* the user obtains an application by means other than the Android Market, such as direct download from the developer’s site, and uploads it to a server, which provides analysis and performs mitigation actions, or *b)* the user deploys a daemon on her device which detects installation requests, interrupts those and performs the actions required for decompilation, static analysis and threat mitigation on-device. The latter method requires modifications to the operating system - namely, a daemon must be installed and a hook has to be injected into the normal installation process. However, these changes are very subtle if compared to what is proposed in [6, 3] and Whisper Systems’ “Whisper Core” effort<sup>2</sup>.

## 4. Experimental Prototype Solution

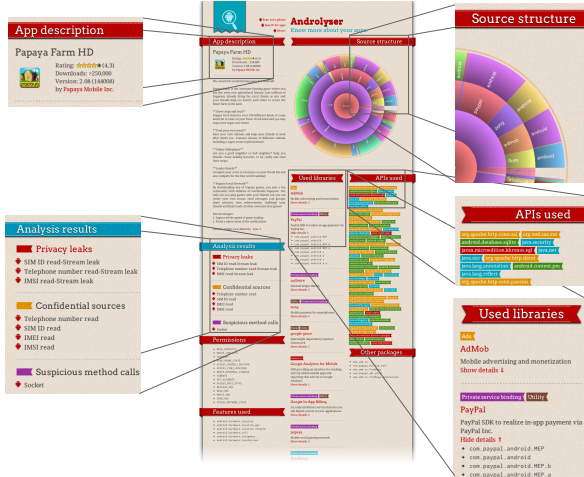
To prove the feasibility of the proposed solution, we have implemented a working prototype for the analysis service, as well as a proof-of-concept prototype for automated mitigation.

### 4.1. Binary Assessment via Static Analysis

Android applications are distributed in self-contained packages called *APKs* which contain both the executable bytecode as well as other binary and XML-based resources required by the application to run. Prior to the actual analysis, we perform a preprocessing step using `apktool`<sup>3</sup>, an open-source decompilation tool. It decodes binary resources to their initial

<sup>2</sup><http://www.whispersys.com/>

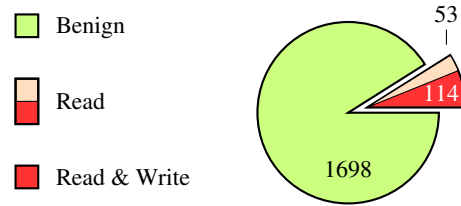
<sup>3</sup><http://code.google.com/p/android-apktool/>



**Figure 2. A sample report for an application with obvious privacy violations.**

XML form and decompiles Java bytecode to Smali<sup>4</sup>, an assembler language targeted at Android’s Dalvik VM. An example of the code produced by the decompiler is shown in Listing 1.

After decompilation, a set of detectors written in Python perform static analysis mainly using regular expressions and statistic analysis of the obtained source tree. The detectors run in a distributed manner and report their results to a central CouchDB-based<sup>5</sup> non-relational database. CouchDB’s ability to serve dynamic HTML rendered from detectors’ results is then used to present comprehensible reports to the user. A sample report for an seemingly harmless game is presented in Figure 2. On the left, the application description from the Market is shown along its title and ID. Below, a list of identified privacy leaks is presented, followed by potential leaks, list of permissions and list of device features required by the application. On the right, the structure of the source code is presented as an interactive tree map, with segments corresponding to Java packages: wider segments show packages with a greater count of opcodes. Below, you can see a list of identified third-party packages, tagged by their functionality, e.g. “Ads”, “Analytics”, a list of Android APIs being in use is provided, as well as



**Figure 3. Number of privacy violating applications among the top 1865 free market items.**

the list of unrecognized third-party packages. It can be seen that the application in this example contains a slew of advertising and analytics libraries and reads the IMEI, IMSI, the phone number and the SIM ID, while writing the former three directly to a stream. Please note that this application is a popular freeware game with an installation base of well above 250.000 users.

We have used a custom Android Market retriever, which was able to obtain 1865 top free applications evenly distributed among the 22 categories available, e.g. “Social”, “Weather” or “Productivity”. After performing automated static analysis on this set of applications, we have found out that at least 167 of these access private identifiers such as IMEI, IMSI, phone number, etc. (see Figure 3). 114 of these applications write the private information to a stream immediately after reading it, which presents a major privacy concern. Please note that the applications under investigation are highly popular free applications with a huge user base of up to millions of users each. The percentage of privacy-offending spyware among all available applications may be even higher than this.

#### 4.2. Malicious activity mitigation

As already mentioned, some malicious activities of the applications are not easy to detect and even harder to overcome. However, the vast majority of the unwanted functionality which we have found in our bulk analysis experiment is related to access, storage and transmission of private identifiers such as phone number or IMSI. This sort of activity is easy to define as regular expressions: since internal system resources like IMEI, IMSI etc. are accessed through an Android system service the resulting bytecode for

<sup>4</sup><http://code.google.com/p/smali/>

<sup>5</sup><http://couchdb.apache.org/>

```

1 .method private benignMethod()V
2   .locals 2
3   .prologue
4   .line 26
5   const-string v0, "Hello there!"
6   #v0=(Reference);
7   const/4 v1, 0x0
8   #v1=(Null);
9   invoke-static {p0, v0, v1}, Landroid/
    widget/Toast;->makeText(Landroid/
    content/Context;Ljava/lang/
    CharSequence;I)Landroid/widget/
    Toast;
10  move-result-object v0
11  invoke-virtual {v0}, Landroid/widget/
    Toast;->show()V
12  .line 27
13  return-void
14 .end method

```

**Listing 1. Smali code for displaying a Toast message**

such accesses always looks very similar. Because of this fact, the decompiled Smali code for such calls will also look the same for every application except for register or parameter identifiers. The following Java code to display a short on-screen notification `Toast.makeText(this, "Hello there!", Toast.LENGTH_SHORT).show();` would result in the Smali code shown in Listing 1. Depending on the surrounding code, the local registers  $v_x$  and parameter registers  $p_x$  may differ, however, the rest of the code stays the same for each instance of this API call. Due to this fact, it is possible to extract a pattern for most API method calls, and implement a detector and the corresponding mitigation strategy with help of regular expressions. The aforementioned way to retrieve system internals via Android's system services like the `TelephonyManager` makes it easy to exactly identify and substitute those unwanted accesses. In order to obtain the device's IMEI, the application has to call the method `getDeviceId()` of a `TelephonyManager` singleton. The corresponding Java code may look as follows:

```

1 TelephonyManager tm = (TelephonyManager)
    getSystemService(TELEPHONY_SERVICE);
2 mId = tm.getDeviceId();

```

First, a reference to the `TelephonyManager` singleton is obtained to be able to communicate with the

```

1 const-string v3, "phone"
2
3 #v3=(Reference);
4 invoke-virtual {p0, v3}, Levil/app/EvilApp
    ;->getService(Ljava/lang/String;)
    Ljava/lang/Object;
5
6 move-result-object v1
7
8 #v1=(Reference);
9 check-cast v1, Landroid/telephony/
    TelephonyManager;
10
11 .line 32
12 .local v1, tm:Landroid/telephony/
    TelephonyManager;
13 invoke-virtual {v1}, Landroid/telephony/
    TelephonyManager;->getDeviceId()Ljava/
    lang/String;
14
15 move-result-object v3
16
17 iput-object v3, p0, Levil/app/EvilApp;->mId:
    Ljava/lang/String;

```

**Listing 2. Malicious code obtaining the IMEI**

service. Then the actual call `getDeviceId()` takes place, revealing the IMEI. In this case, an obvious mitigation strategy would replace the IMEI with a UUID. Prior to the implementation of this strategy, the opcodes corresponding to the malicious pattern as well as to its replacement have to be identified. In this example, the crucial opcode is located on the line 13 of Listing 2. Listing 3 shows an example replacement, where `TelephonyManager` has been replaced by an instance of the UUID generator. We have tested this particular patch with a small number of applications, and none of the applications has shown any indications of missing or broken functionality after the patch has been applied. For more complex substitutions it would be crucial to take care of the used registers. But as most accesses to internal resources are realized through single `Service` methods, an automated substitution is easy to implement.

After the mitigation strategy has been executed, the patched Smali code is compiled back to an APK using the aforementioned `apktool`, and then signed with `jarsigner`. Obviously, the certificate used for the APK signature will differ from the original binary from the Market. However, Android only checks signatures during an upgrade, but not if the application is not yet installed on the device. This makes deploy-

```

1 invoke-static {}, Ljava/util/UUID;->
    randomUUID()Ljava/util/UUID;
2
3 move-result-object v1
4
5 invoke-virtual {v1}, Ljava/util/UUID;->
    toString()Ljava/lang/String;
6
7 move-result-object v3
8
9 iput-object v3, p0, Levil/app/EvilApp;->mId:
    Ljava/lang/String;

```

**Listing 3. Patched code using UUID instead of IMEI**

ment of self-signed, patched applications an easy task for the user.

## 5. Conclusion and Future Work

In this work, we have proposed a system which combines deep static analysis of Android applications with comprehensible reporting functionality and a means of security and privacy threat mitigation on the application level. An end user can benefit from our solution in a number of ways. First, the user gains an insight in the security and privacy related internals of any application with a level of detail unseen before. Second, the user can decide if certain potentially malicious functionality is unwanted, and it can be removed from the application while retaining the core functionality intact. We have implemented a proof-of-concept prototype and provided its implementation details to the reader.

In future work, more effort can be put into pattern mining and other malicious code detection techniques to help us identify a greater number of security and privacy threats as well as other malicious activities within Android applications. Another open problem is the identification of obfuscated malicious activities which have shown to be not easy to detect with static analysis. However, since Android applications run as bytecode in the Dalvik VM, we have to investigate the possibilities available to an attacker in terms of self-modification and polymorphic code, as well as the corresponding mitigation strategies.

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