

# CS 598 - Computer Security in the Physical World: Project Submission #2

Professor Adam Bates Fall 2016

Security & Privacy Research at Illinois (SPRAI)

# Last Deliverable (BG+RW) I

- I will have feedback back to you by this evening.
- Generally positive impressions. Did people have trouble writing these?
- Poor scores were mostly due to people failing to follow instructions.
- If you fix the issues that I identified prior to the final submission, the first deliverable will not have a major impact on your final course project.

# Oct 18th Deliverable

- Last submission we "locked in" the background and related work sections of our paper. This submission will be a living document as you begin to do the real work.
- Purpose: Tell me specifically what are you doing to DO in your project
- Format: Add a new section called "Experimental Proposal" to the LaTeX Two Column ACM document.
- Submission: Email me the PDF (include [cs598] in subject)

# Oct 18th Deliverable

Be sure to include:

- Hypothesis: Based on what you've learned so far,
  "commit" to a prediction that is the basis of your paper.
  - What does a hypothesis in a defensive paper look like?
- Methodology and/or Design:What techniques are you going to use? How are you going to use them? Will you leverage existing tools? Convince me that you will succeed in executing your methodology.
- Evaluation and/or Analysis: How will you determine the extent to which you have succeeded in your goal?

## Using Provenance Patterns to Vet Sensitive Behaviors in Android Apps --C. Yang, et al. SecureComm '16

Presented By : Wajih

#### UNIVERSITY OF ILLINOIS

AT URBANA-CHAMPAIGN



## Problem

How can you tell whether software you - Develop - Buy -Install is **safe** to run?



## What is Provenance?

- Provenance, a.k.a. lineage of data
  - Data's life cycle
    - Origins
    - Accesses
    - Deletion
  - Provides confidence in authenticity
- Was the latest data used in the computation?
- Was the data deleted after its use?
- Was the sensitive data sent on network?

#### **Open Provenance Model (OPM)**

**Process Vertices**: Represent dynamic entities e.g. operating system processes

PID 1002 UID 3002

Artifact Vertices: Represent static elements that are consumed or produced by processes.

e.g. files, sockets etc.

File1 Sock5

Edges: Represent dependency between pair of vertices; Directed

- WasTriggeredBy: from a process to another process
- WasGeneratedBy: from an artifact to a process
- Used: from a process to an artifact
- WasDerivedFrom: from an artifact to another artifact.



#### **Provenance Graph**

- Log all the system events and then make OPM relationships to generate provenance graphs
- **Example:** Download file from network and executes it:



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### **Program Behavior Analysis**



#### **Traditional Analysis methods**

**Static Analysis:** Investigates the properties that can be investigated by inspecting the downloaded app and its source code only. Eg: Signature based inspection used by anti-virus technologies.

**Dynamic Analysis:** In this method, app is run in a secure environment such as sand-box and logs every relevant operation of the app.



#### **Static vs Dynamic Analysis**

Static -

- Consider all possible inputs
- Can prove absence of bugs/vulnerability
- Cannot handle code obfuscation (Java reflections, source encryption etc)
- Cannot find vulnerabilities in RT
- HUGE overhead, reason about every input

Dynamic -

- Consider some inputs
- Can prove presence of bugs/vulnerability
- Cannot guarantee the full coverage of the source code
- May require virtual Machine instrumentation



#### **Example: Taint Checking Solution for Behavior Analysis**





#### **Example: Taint Checking Solution for Behavior Analysis**



### Paper's Approach

- Collect provenance of the applications behaviors <u>without</u> modifications and generate provenance graphs
- Pattern match generated provenance graphs with sensitive behavior patterns



### Dagger

- A lightweight system to dynamically vet sensitive behaviors in Android apps by making provenance graphs.
  - No VMI instrumentation
  - Less Overhead
  - Just tracks apps interactions with underlying platform
- Dagger uses the open source **SPADE** provenance middleware to collect three types of low-level execution information:
  - Linux System Calls -- via Strace
  - Android Binder transactions -- via sysfs
  - App process details -- via /prof fs



### System Design

- App Executor:
  - Executes the app in a sandbox
- Syscall Collector:
  - Collects the system call invocation
- ProvEst Daemon:
  - Collects information from binder transactions; Builds relationships between artifacts and process
- Graph Reporter:
  - Outputs Data provenance graph from the relationships made by ProvEst daemon
- Behavior Identifier:
  - Detects sensitive behaviors from the provenance graph



Dagger runs each app, collect provenance records and perform pattern matching





#### **Collection of Sensitive Behaviours**

- They ran Android apps with selected input that is known *a priori* to trigger sensitive behavior
- Dependent on internal working of Android

Table 2. Fined-grained	d sensitive behaviors a	associated with	malicious behaviors
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Index	Malicious Behaviors	Sensitive Behaviors	Index	Malicious Behaviors	Sensitive Behaviors
1	Phone Call	Phone Call	5	Steal Contact	Read Contact and Net
2	Send SMS	Send SMS	6	Track Location	Read Location and Net
3	Block SMS	Receive SMS, not Write SMSDB	7	Execute Shell	Execute Shell
4	Steal SMS	Read SMSDB and Net	8	Net	Net



#### **Example Sensitive Behaviour**

#### Pattern : Read Geolocation:

- An app attempts to read the geographic location
- Interacts with the Location Manager Service
- Requests the location from the GpsLocationProvider.





#### **Provenance Generated from the Behaviour**





## **Evaluations**

Evaluated **EFFECTIVENESS** from three perspectives:

- 1. Vetting real-world malware case studies
  - 1.1. Gamex: Code Encryption
  - 1.2. Gone60: Privacy Leakage
  - 1.3. Zsone: SMS Service Usage
- 2. Vetting Android Genome Project malware [S&P 2012]
  - 2.1. 1,260 real-world malware samples collected from the Genome Project
- 3. Vetting official market (Google Play) apps
  - 3.1. 1000 apps sampled from 18,527 official market (Google Play)



## **Evaluations**

- Evaluated EFFICIENCY from three perspectives:
  (i) CPU overhead
  (ii) Memory overhead
  (iii) Runtime overhead
- Used AnTuTu (android benchmark app) to test performance
- Both CPU and Memory overhead was less than 10 %
- The runtime overhead is less than 63%
  - Strace overhead



### Discussion

What are the key contributions of this paper?

What are the limitations of this paper?

### Can we use Provenance Patterns for something else?



## **BACKUP SLIDES**



#### **Taint Analysis**

Explicit – Passed in assignments

Implicit - Passed in control flow structures (ICC, Broadcasts, Media)

#### **Data-Flow Analysis**

Most use some form of Data Flow Analysis with Taint Analysis Sources: Location Data, Unique IDs, Call State, Authentication Data and Contact/Calendar Data Sinks: SMS Communications, File Output, Network Communication, Intents, Content Resolver Examples: TaintDroid (2010), AndroidLeaks (2012)



