# **Understanding Real-World Malwares and Anti-Virus Engines**

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#### **ABSTRACT**

Your abstract should go here. You will also need to upload a plaintext abstract into the web submission form.

#### CCS CONCEPTS

• Security and privacy → Use https://dl.acm.org/ccs.cfm to generate actual concepts section for your paper;

#### **KEYWORDS**

template; formatting; pickling

#### 1 INTRODUCTION

Combating malware is important.

Combating malware needs efforts from the whole community. VirusTotal is a website, combining vendors? new detection techniques. VirusTotal is widely used in industry.

Beyond industry, academia also widely uses VirusTotal for different purposes.

Our measurement shows that research community is using Virus-Total in a wrong way.

VirusTotal is used in a wrong way.

The question we want to ask is whether academia uses VirusTotal in a correct way. If not, how should we use VirusTotal.

Contribution:

- a. We survey more than 100 academic paper and summarize how researchers use Virus Total. We find two usage patterns.
- b. We collect big data from VirusTotal and use these data to show that the current usage of VirusTotal is wrong.
  - c. We build a prediction model to help better use VirusTotal

# 2 EMPIRICAL STUDY ON HOW VIRUSTOTAL IS USED IN ACADEMIC PAPERS

# 2.1 How we collect paper and characteristics of collected paper

- a. Year distribution
  - b. Conference distribution
  - c. Topic distribution

#### 2.2 Findings

- a. Do not wait until results become stableb. Treat vendors equally
- 2.3 Discussion

What if the current usage is not correct?

# 3 METHODOLOGY TO COLLECT VIRUSTOTAL DATA

#### 3.1 The large data set

How the data set is built?

What information we can get? Basically, we need to explain the data format.

Basic properties of the data set

- a. How many submissions every data?
- b. Submission type distribution
- c. The number of submissions for the same file
- d. Engines used to scan a submission

Advantage:

Across categorization, such as file types

Covering a longer time.

#### 3.2 The small data set

How the data set is built?

What information we can get? I mean the data format.

Basic properties of the data set

- a. Detection results from the first scan
- b. Vendor distribution
- c. How Virus Total update engines? Scanning time vs. update vs. version  $\,$

#### 3.3 Caveats

Discuss errors during our data collection.

#### 4 FLIPPING AND STABLE STUDY

## 4.1 Hazard discussion

Todo: add hazard discussion in this part before discussing flipping. Hazard caused by VT API? NO

Hazard caused by vendor misfunction? NO

conclusion: randomly appear, but quite frequently, affect lots of files

Conclusion: remove hazard before studying flipping? Run experiments both with and without hazard

#### 4.2 Flipping

a. For each file, we count of flipping vendors and of flipping times
 b. For each vendor, we count of flipping, of flipping files, average flipping per file

#### 4.3 Stable

a. We need some metrics to support our definition of ?stable? is reasonable.

b. Some metrics to show how long we need to wait until results become stable  $\,$ 

#### 4.4 Conclusions

#### 5 INFLUENCE MODEL

### 5.1 Influence graph

How we model vendors? influence

1

# 5.2 Influence measurement

The four metrics

#### 5.3 Results

a. On the small data set
 b. On the large data set\*\*
 Justify file category
 Justify length of the data

#### 5.4 Discussion

## **6 PREDICTION MODEL**

#### 6.1 Overview

What do we want to predict?

a. Whether a file is stable now
b. decision tree

# 6.2 Experiments and implications

## 7 RELATED WORKS

The AISec Paper

# 8 CONCLUSIONS

**REFERENCES**