

## A Framework for the Analysis of File Infection Malware





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Promo 2024 – Digital Security

#### Malware

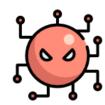
#### **Categories:**



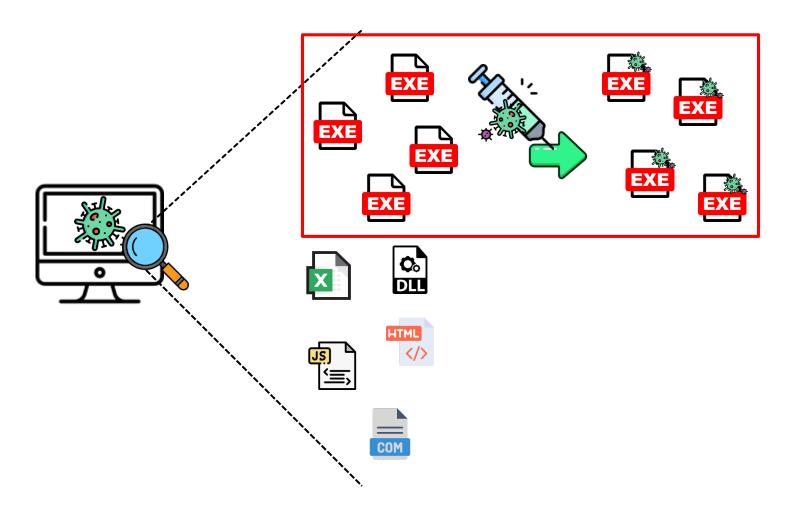
- Worm
- Grayware
- Backdoor
- Ransomware
- Clickwa

#### **Families:**

- Expiro
- Melissa
- Zeus
- Emotet
- Conficker
- Nimda
- Ramnit
- Slammer
- Wannacry
- Maze
- Ryuk



### **File Infection**



### **Detection & Classification**





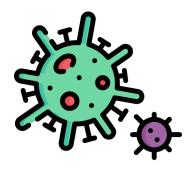


















#### Families:

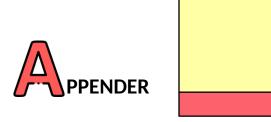
- Rovnix
- Zlob
- Gator
- Look2Me
- Rustock

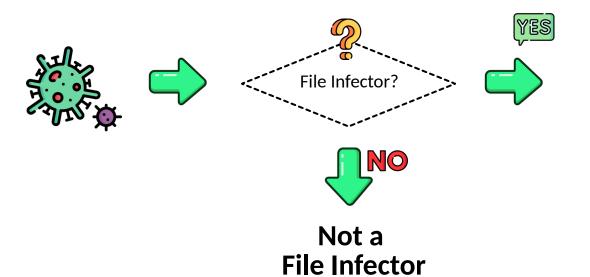
# Findectors are Hard Petect/Classify

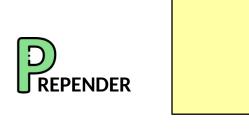
Class	7	y class. Dyn.	Recall Comb.	Far S'		core Com.
Adware		-	0.981		11د	0.925
Backdoor	b.	,	0.0		J.730	0.838
Clicker	0.9,		•		0.692	0.821
Dialer	0.994			<b>68</b>	0.888	0.984
Downloader	0.974	0.		J.864	0.695	0.874
Grayware	0.932			832	0.675	0.852
Miner	0.989			77	0.807	0.962
Ransomware	0.9		<u> </u>		0.580	0.853
Rogueware			0.95		401	0.663
Spyware		وم	0.998		4	0.879
Tool		J.929	1.000	L.		0.830
Virus		0.939	0.971	0.81		809
Wor	/8	0.899	0.996	0.922		Y
	0.967	0.920	0.9907	0.848	0.	

Dambra et al. «Decoding the Secrets of Machine Learning in Malware Classification: A Deep Dive into Datasets, Feature Extra don, and Model Performance». CCS '23.

#### **Our Problem**

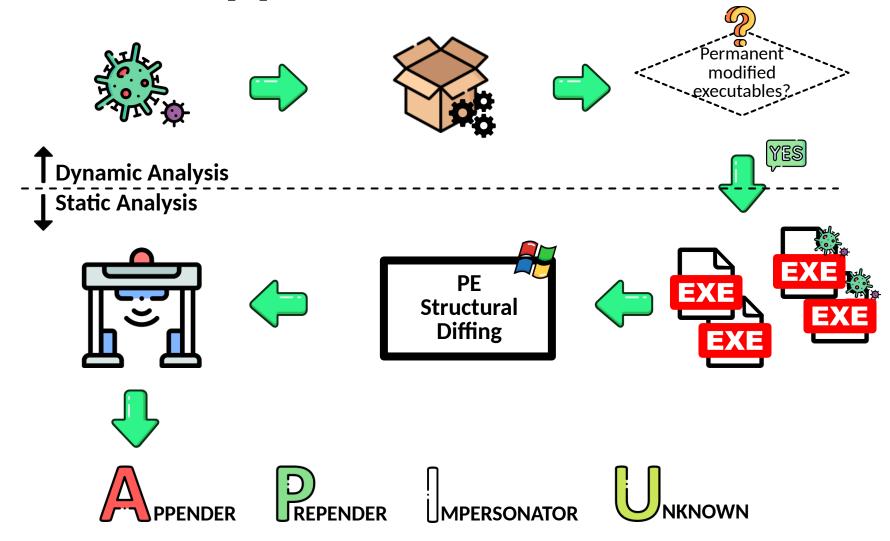








### **Approach Overview**



### PE Structural Diffing



DOS Header

DOS Stub

Rich Header (OPT)

COFF File Header
Standard Fields
Windows Specific Fields
Data Directories

Section Table

Section 1

Section 2

Attribute Certificate Table (OPT)

Overlay (OPT)



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Overlay (OPT)

DOS Header
DOS Stub
Rich Header (OPT)

COFF File Header
Standard Fields
Windows Specific Fields
Data Directories

Section Table
Section 1
Section 2
Section 3
Section added

#### **Related Work**

#### File infectors

- Vast literature on manual analysis of file infector families (i.e., Memery, Neshta, Triusor [VirusBulletin])
- Concept of Computer Virus [Cohen1987, Szor2005, Filiol2006]
- Overview of malwares [Skoudis2004]

#### Executable Diffing

 Unlike existing tool such as BinDiff, we developed a component based PE Structural Diffing

#### Fuzzy Hashes

We implied TLSH [Azab2014] and SSDeep [Kornblum] for PE Structural Diffing

#### Our work

 To our knowledge, there are no existing studies on automated file infector classification based on infection behavior

#### **Dataset**

#### Dambra et al. Dataset

- 67,000 samples
- 670 families
- 100 samples per families

Collected between 2021-2022

**AVClass2** 



#### **Subset Dataset**

- 7,000 samples
- 70 families
- 100 samples per families

#### **Analyzed Dataset**

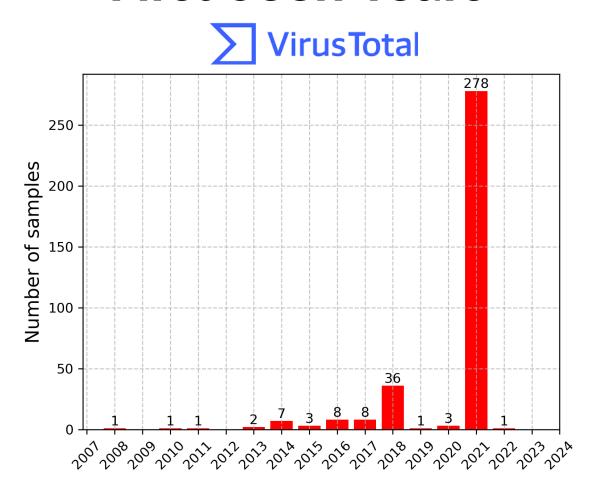
• 350 samples

**Analysis** 

- 70 families
- 5 samples per families



#### **First Seen Years**



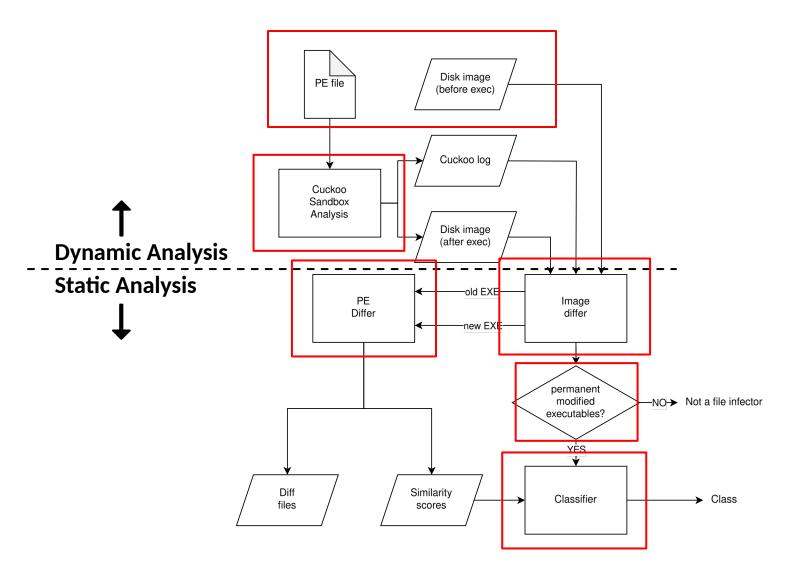
#### **Contributions**

- A novel framework for the analysis of file infection malware
- A PE Executable Differ
- Evaluation on 350 malware samples

### **Outline**

- Introduction
- → Approach
- Results

### **Approach**



#### **Cuckoo Sandbox**

- Input : Sample
- Output:

Cuckoo report

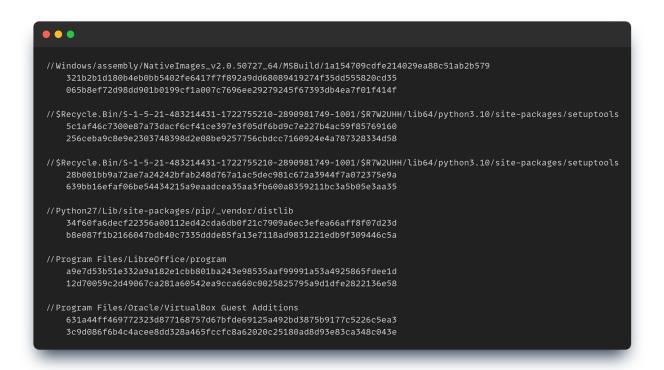
```
JS report.js
      "added": 1707408441.368206,
      "started": 1707408454.89419,
      "ended": 1707408880.598961,
      "owner": null,
      "score": 3.8,
      "id": 1,
      "category": "file",
      "git": {
           "head": "13cbe0d9e457be3673304533043e992ead1ea9b2",
          "fetch_head": "13cbe0d9e457be3673304533043e992ead1ea9b2"
      "monitor": "2deb9ccd75d5a7a3fe05b2625b03a8639d6ee36b",
      "package": "",
      "route": "internet",
      "custom": null,
      "machine": {
          "status": "stopped",
          "name": "Win7",
          "label": "Win7",
          "manager": "VirtualBox",
          "started_on": "2024-02-08 16:07:35",
          "shutdown_on": "2024-02-08 16:14:40"
```

Disk Image (after exe)

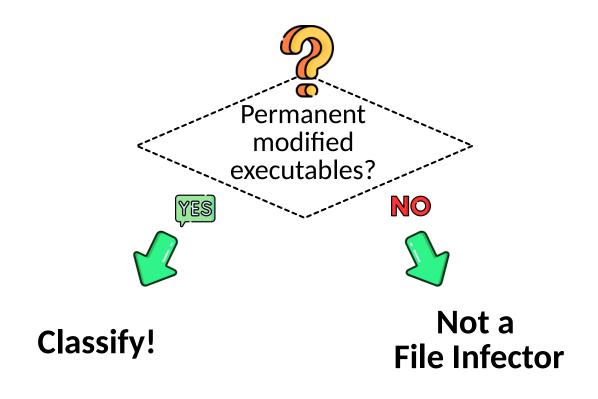


### **Image Differ**

- Input : Disk Image (orig.) & Disk Image (infect.) & Cuckoo report
- Output : Permanent modified executable dictionary



#### Is it a File Infector?



### **Fuzzy Hashes**

- Fuzzy hashes allow to compute similarities between executables
- Similarity score range from 0 to 100
- Fuzzy hashes used are TLSH and SSDeep

T134537C21B981C073C446107A592DC6B19F 7BBC312675C983BB961BBB9F313D1E72E24A





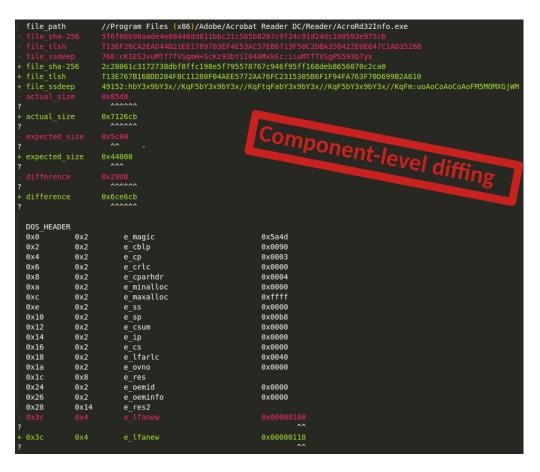


Similarity: 67

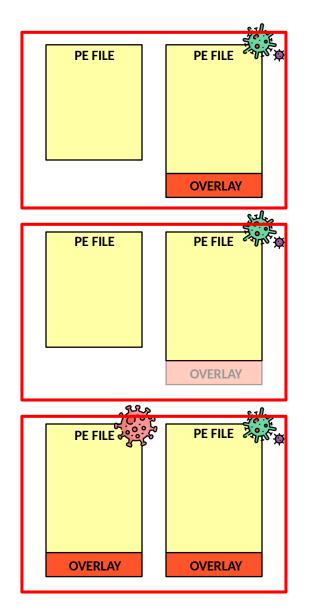
T1BA666B02B69DBCF8C4765030477793F25B 29FC211560EA5F73D4BB252E34683BA29B26

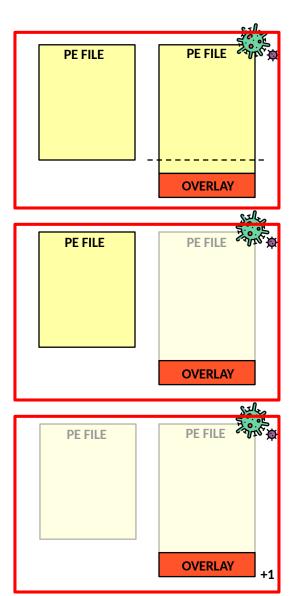
#### **PE Differ**

- Input : Original EXE & Infected EXE
- Output : Diff file



### **Classifier Features**





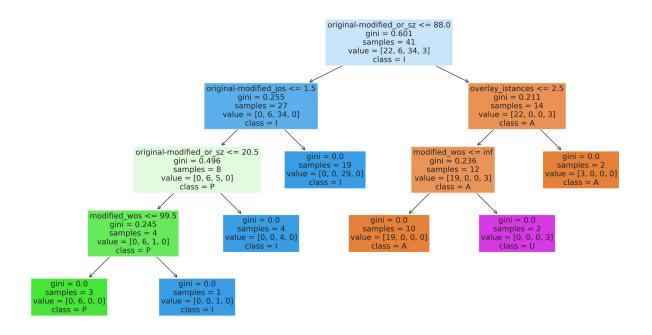
#### **Ground Truth Labels**

family	type	$\mathbf{st}$	$\mathbf{st}$	$\mathbf{st}$	orig	$\mathbf{e}\mathbf{p}$	overlay	
		added	extend	remov	es	$\operatorname{mod}$	ratio	
					$\operatorname{mod}$			
expiro	Α	0	1	0	✓	Х	Х	
triusor	A	4	0	1	Х	<b>✓</b>	Х	
wapomi	Α	1	0	0	X	✓	Х	
wlksm	A	0	1	0	✓	Х	Х	
lamer	Р	3	0	all	-	-	637.2	
induc	Р	3	0	all	-	-	1.0	
neshta	Р	8	0	all	-	-	1.0	
shodi	Р	4	0	all	-	-	6.4	
sinau	Р	8	0	all	-	-	1.0	
sivis	Р	3	0	all	-	-	4.8	
soulclose	Р	3	0	all	-	-	1.0	
xiaobaminer	Р	7	0	all	-	-	53.3	
memery	Р	4	0	all	-	-	55.9	
pidgeon	Р	26	0	all	-	-	0.8	
detroie	Р	8	0	all	-	-	20.3	
gogo	Р	3	0	all	-	-	33.1	
lmir	Р	8	0	all	-	-	0.1	
stihat	Р	18	0	all	-	-	13.4	
xolxo	Р	70	0	all	-	-	82.6	
xorer	Р	3	0	all	-	-	1.8	
virlock	Ι	2	0	all	-	-	Х	
grenam	I	10	0	all	-	-	Х	

22 file infector families 4 Appenders, 16 Prependers, 2 Impersonators

#### Classifier

- Random Forest Classifier on 4 different labels (A, P, I, U)
- Training Set and Testing Set divided as 70 % and 30 %



### **Outline**

- Introduction
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#### **Results: Execution**

- 350 samples analyzed, 5 for each family
- 2 families with 0 Windows API calls
- About 97% of the samples detonated (>50 Win API)
- 94 samples of 22 families permanently modified executables, thus labelled as file infectors

### **Results: Classifier**

	Accuracy	Macro	Weighted	$\mathbf{A}$	P	I	U
		Avg	Avg				
Precision	1.0	1.0	1.0	1.0	1.0	1.0	1.0
Recall	1.0	1.0	1.0	1.0	1.0	1.0	1.0
F1-score	1.0	1.0	1.0	1.0	1.0	1.0	1.0



#### Limitations

- A fraction of samples may not have detonated
- The 70 families may not capture all file infector types, e.g., we did not encounter Cavity infectors
- Only 350 samples analyzed
- Overfitting in the random forest classifier

#### Conclusions

- A novel framework for the analysis of file infection malware
- A PE Executable Differ
- Evaluation on 350 malware samples

#### **Future work**

- Analysis of a more extensive sample pool
- Additional features beyond similarity scores
- Improving the capabilities of PE Differ

**Questions?** 

Thanks!