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When Malware is Packin' Heat;

Limits of Machine Learning Classifiers Based on Static Analysis Features

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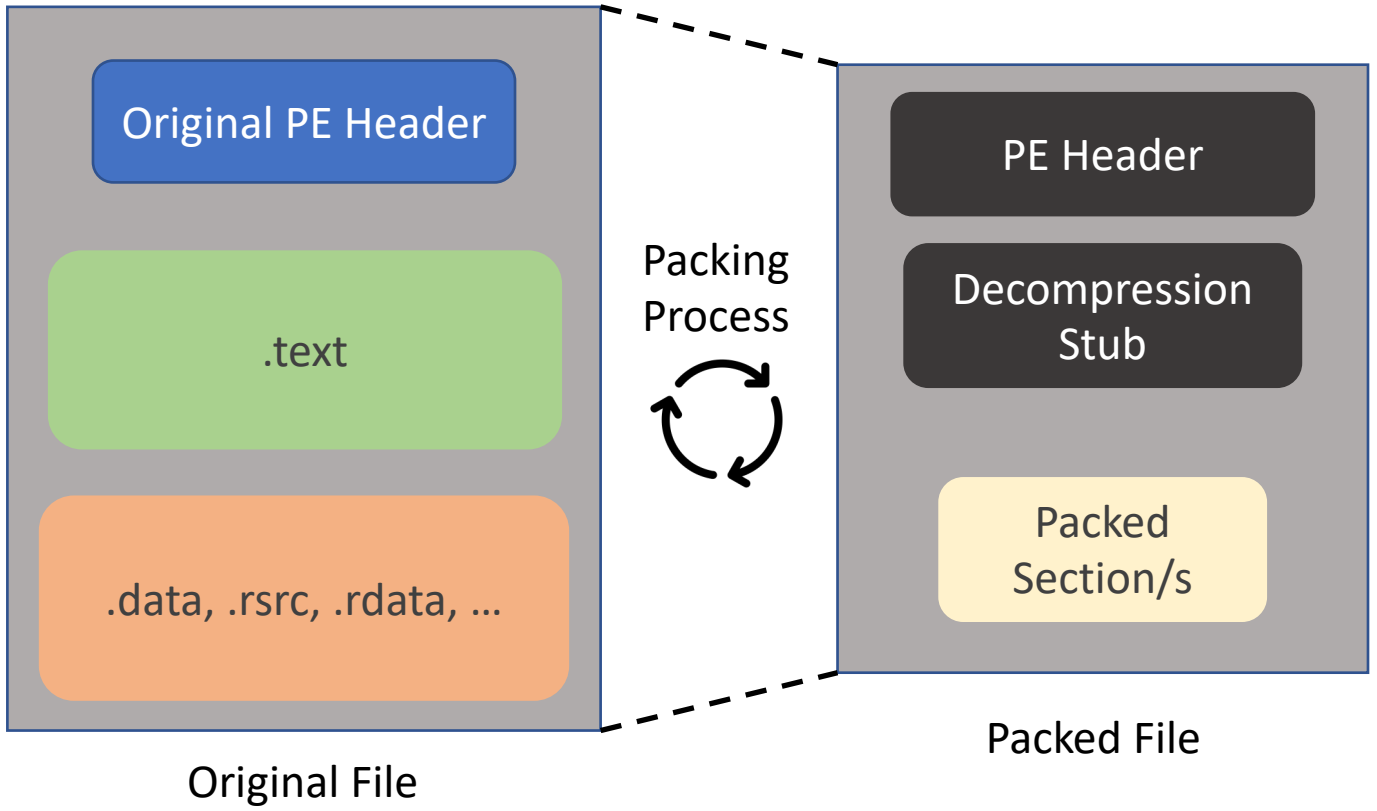
Packing



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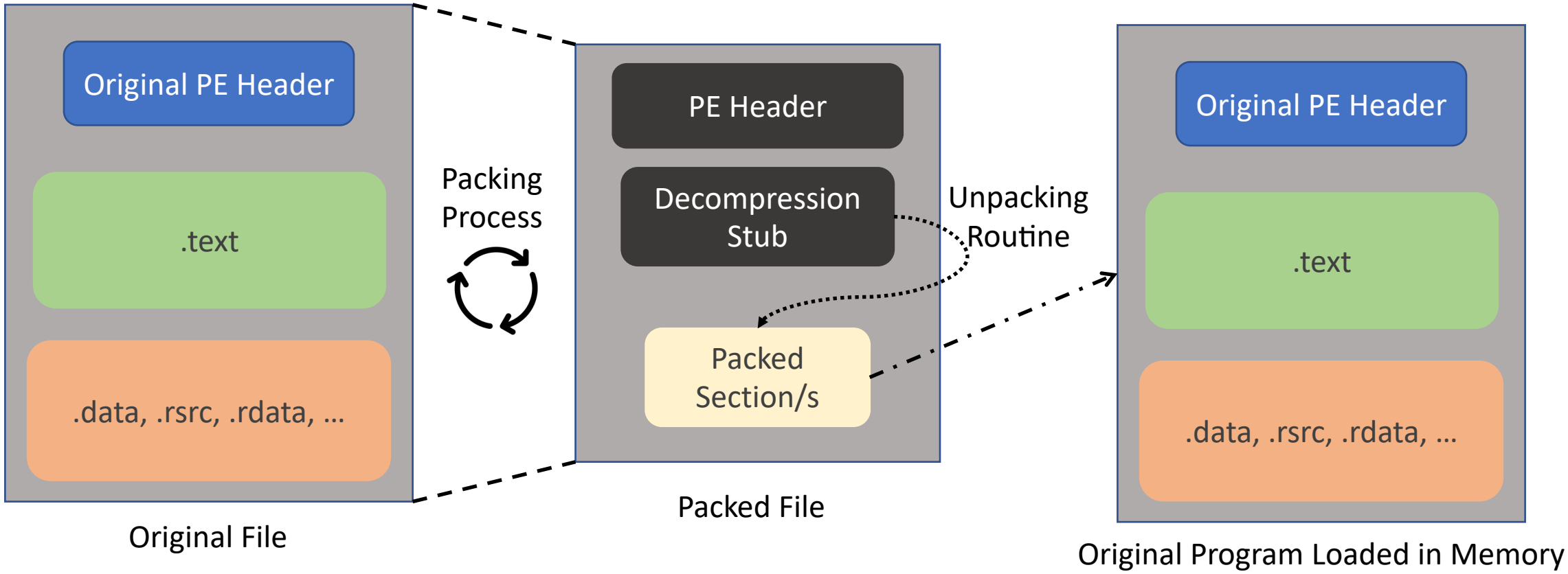
Original PE Header
.text
.data, .rsrc, .rdata, …
Original File
Packing Process
PE Header
Decompression Stub
Packed Section/s
Packed File
Unpacking Routine
Original PE Header
.text
.data, .rsrc, .rdata, …
Original Program Loaded in Memory

Packing Employed By Malware Authors

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Packing Evolution

- Most packers are not this simple anymore...

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  - Different methods of obfuscation or encryption are being used

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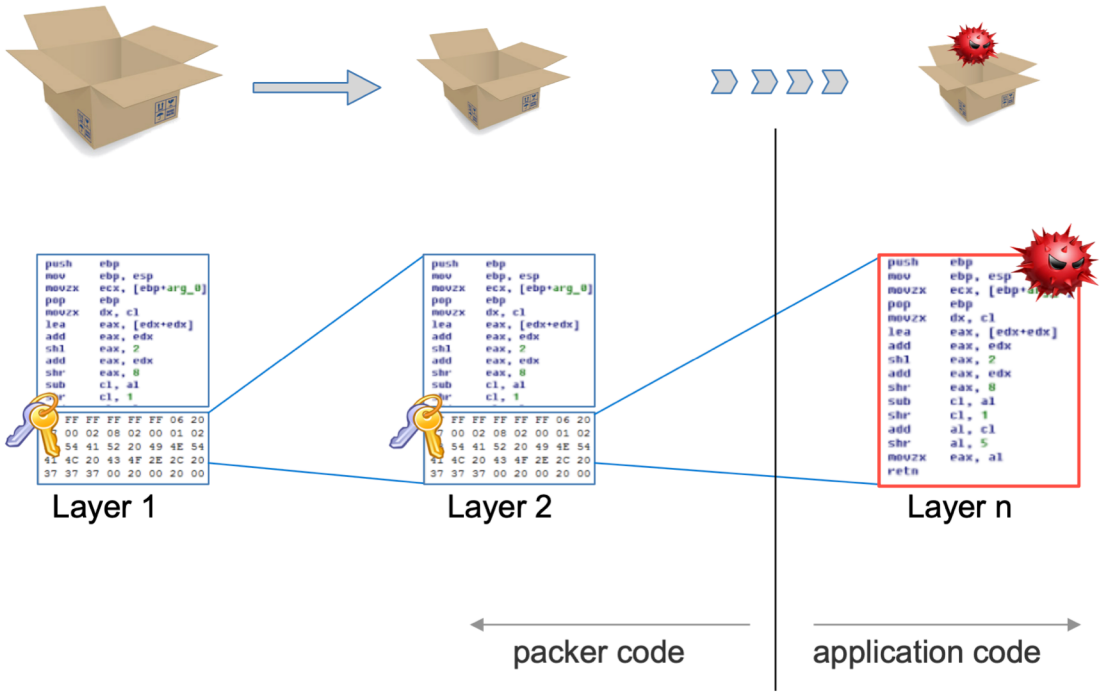
  - Packing happens at multiple layers



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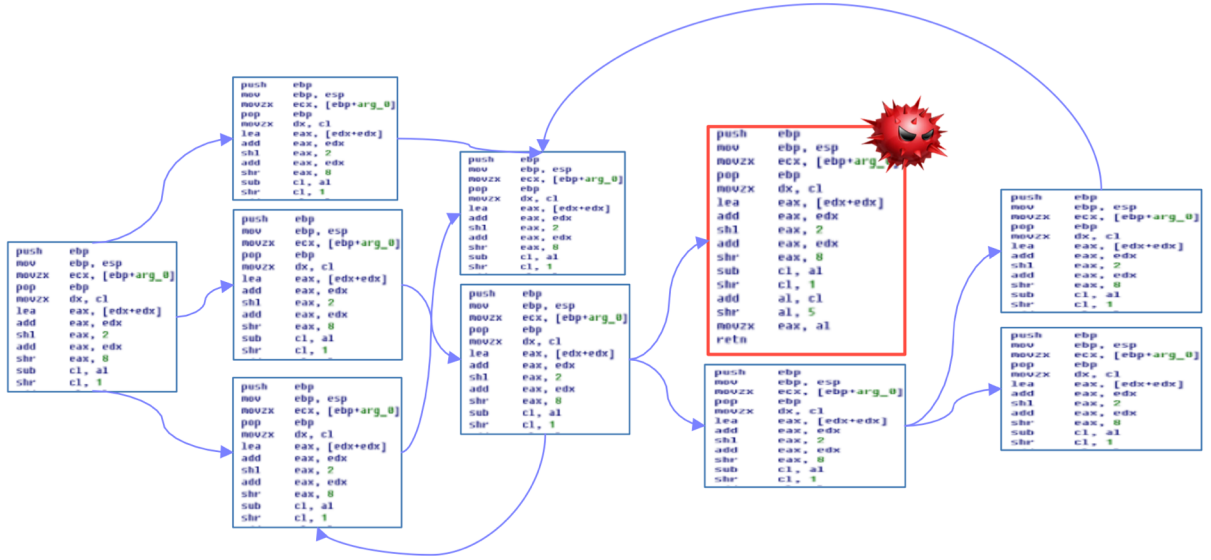
  - Packing happens at multiple layers
  - Unpacking routines are not necessarily executed in a straight line



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Packing Evolution

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Packing Evolution

- Most packers are not this simple anymore...
  - Different methods of obfuscation or encryption are being used
  - Packing happens at multiple layers
  - Unpacking routines are not necessarily executed in a straight line
  - Only a single fragment of the original code at any given time
  - Usually anti-debugging or anti-reverse-engineering techniques are employed

Why Does Packing Matter?

- It hampers the analysis of the code

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- Makes malware classification more challenging!

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- Makes malware classification more challenging!
  - Especially, when using only static analysis

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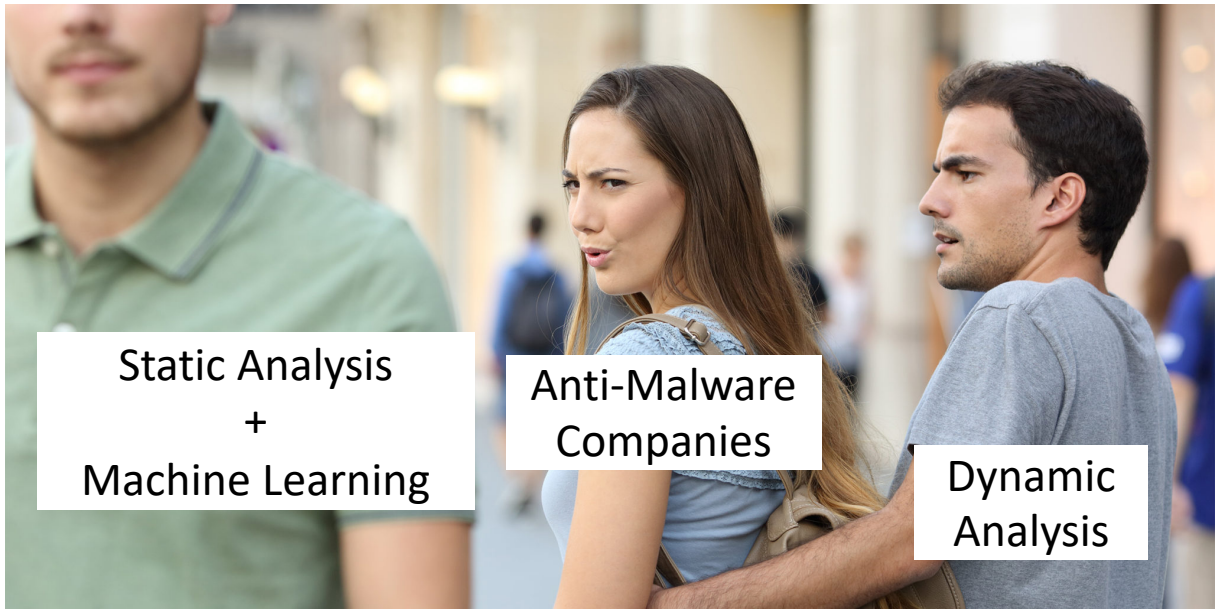
Malware Classification Using Static Analysis

ENDGAME.

CYLANCE

cybereason®



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Acronis

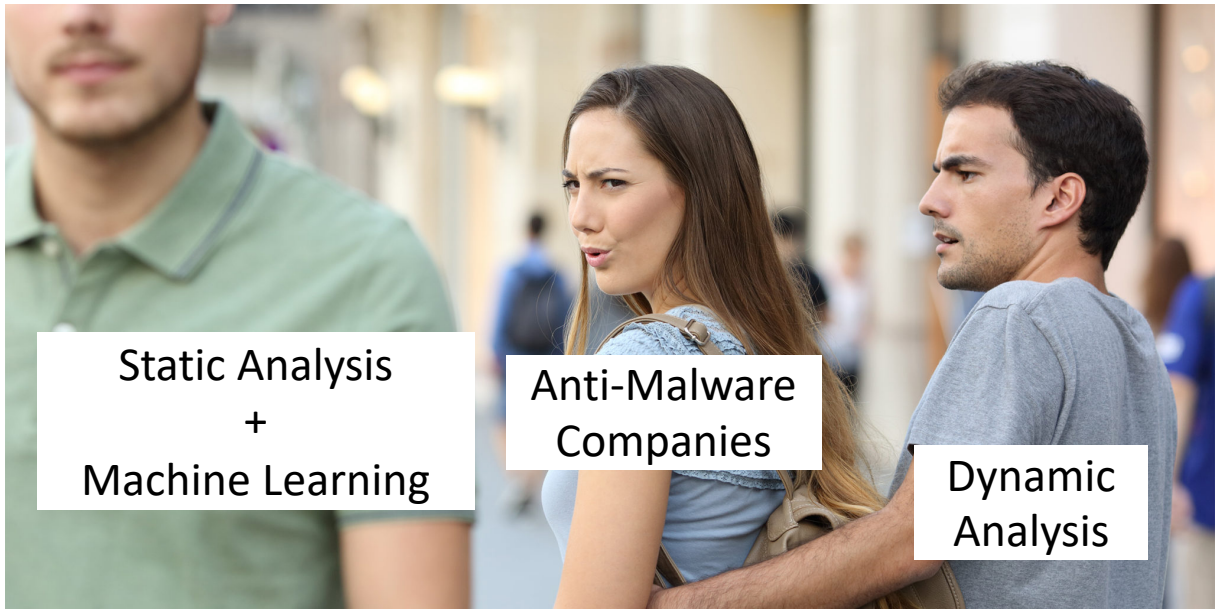
TRAPMINE

Malware Classification Using Static Analysis

ENDGAME.

CYLANCE

cybereason®



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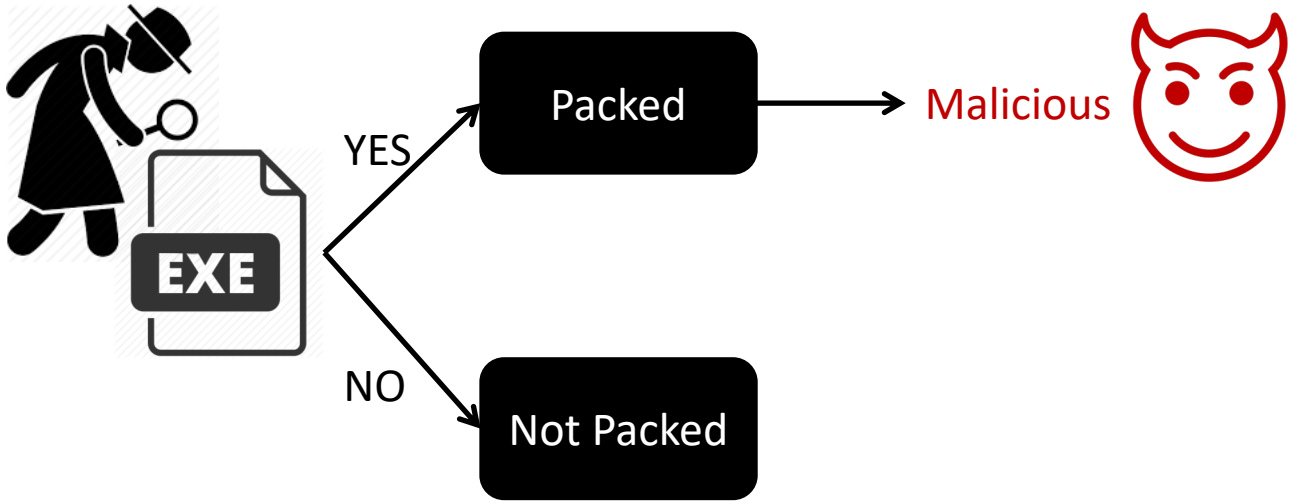
Acronis

TRAPMINE

- What happens if the program is packed, i.e., the features are obfuscated?

Do Benign Software Programs Use Packing?



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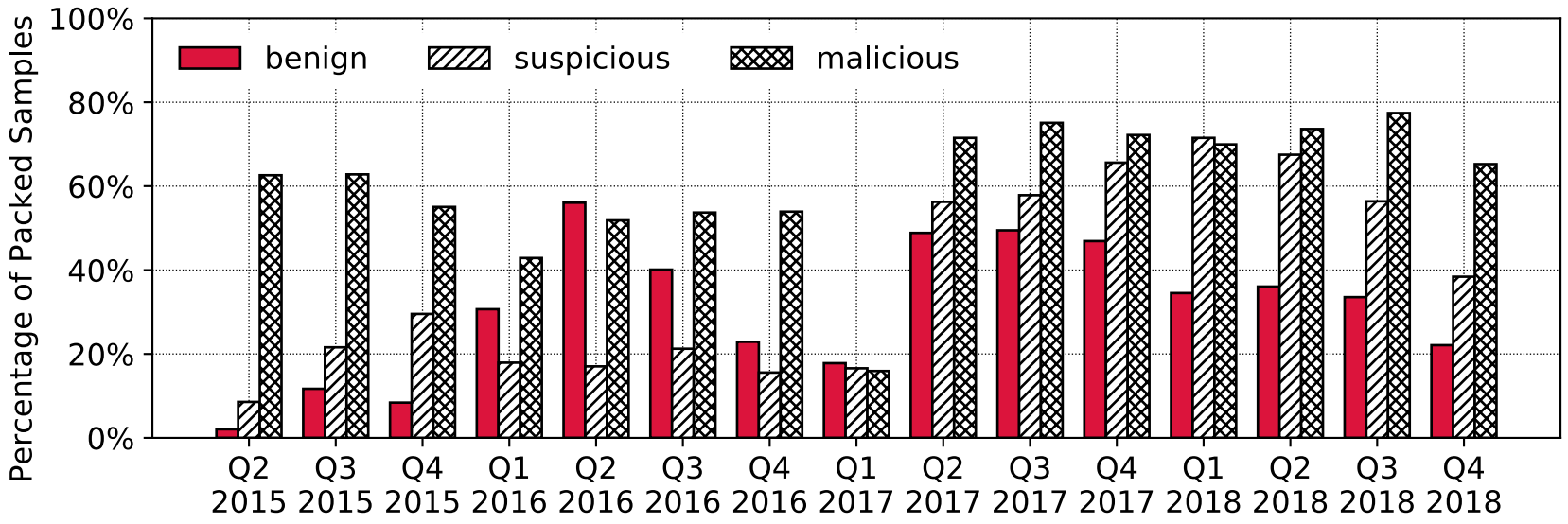
Packing Is Common in Benign Programs



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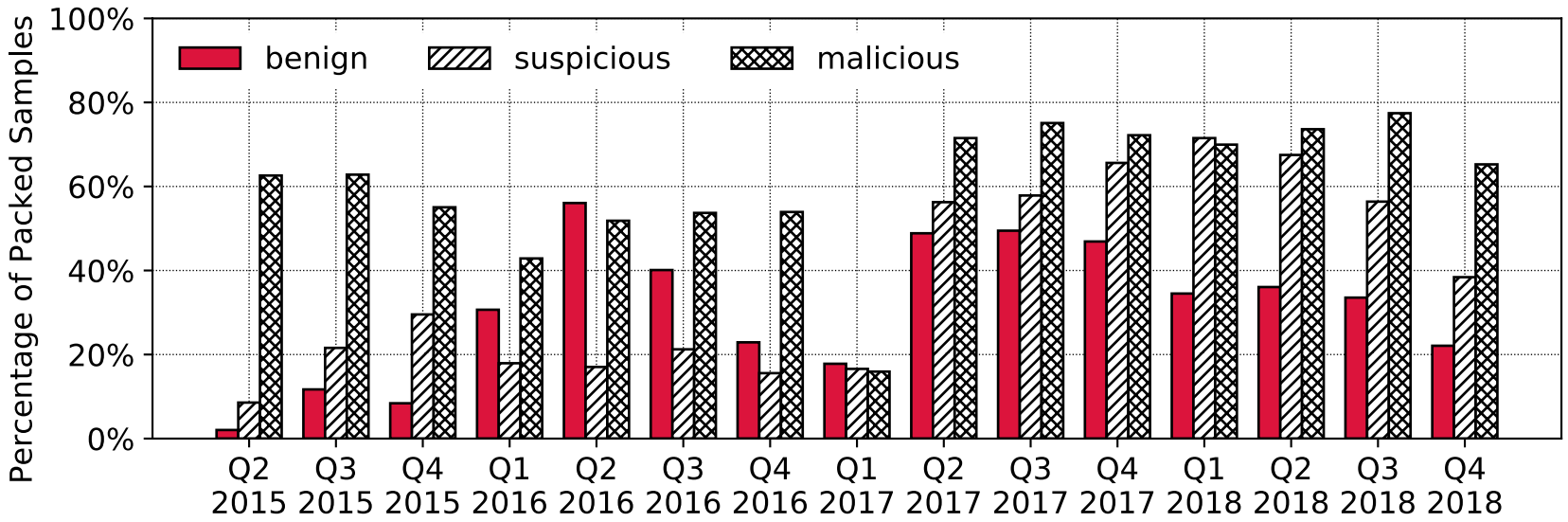
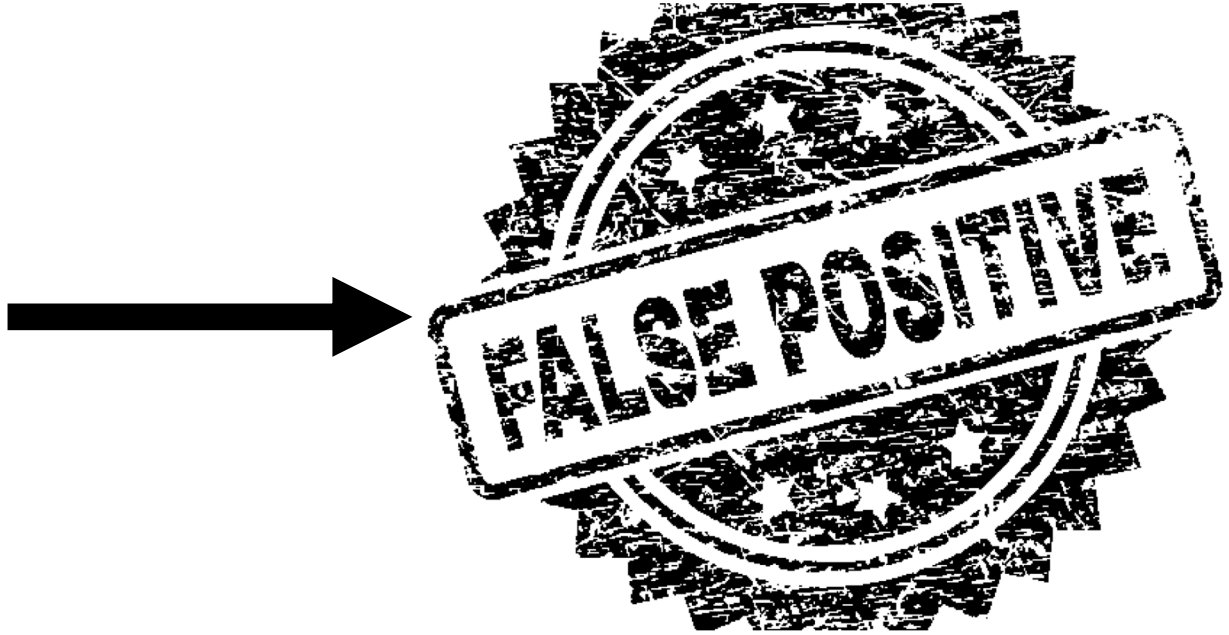


- Rahbarinia et al. [84], who studied 3 million web-based software downloads over 7 months in 2014, found that *both* malicious and benign files use known packers (58% and 54%, respectively)

B. Rahbarinia, M. Balduzzi, and R. Perdisci, "Exploring the Long Tail of (Malicious) Software Downloads," in *Proc. of the International Conference on Dependable Systems and Networks (DSN)*, 2017.

“Packing == Malicious” 

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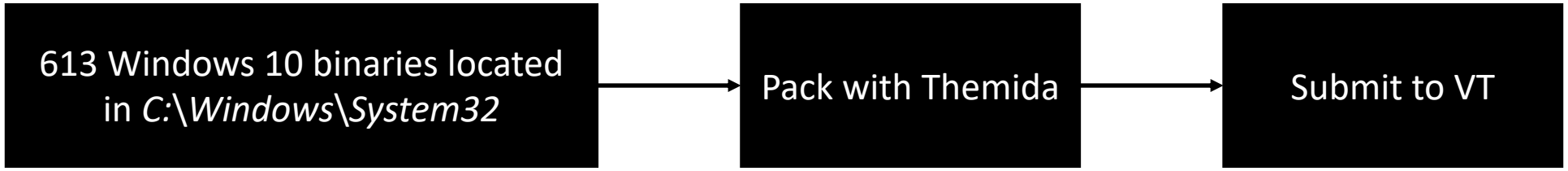
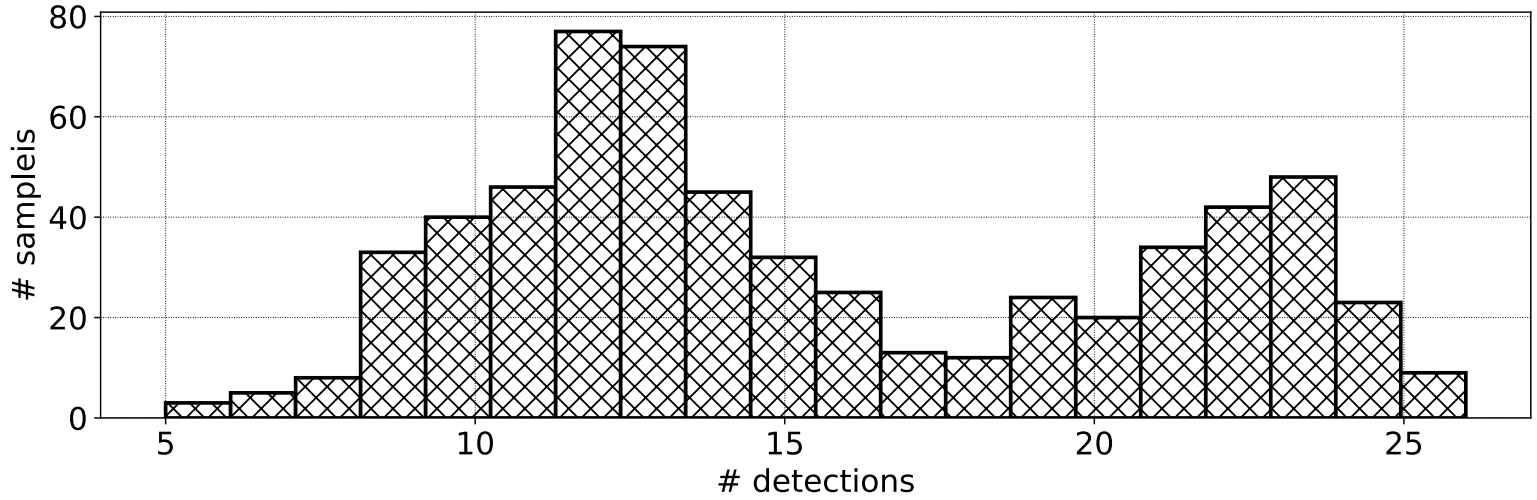
“Packing == Malicious” on VirusTotal?





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Dataset Pollution

Does static analysis on packed binaries provide *rich enough* features to a malware classifier?

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Datasets

1. Wild Dataset (50,724 executables):
   - 4,396 unpacked benign
   - 12,647 packed benign
   - 33,681 packed malicious

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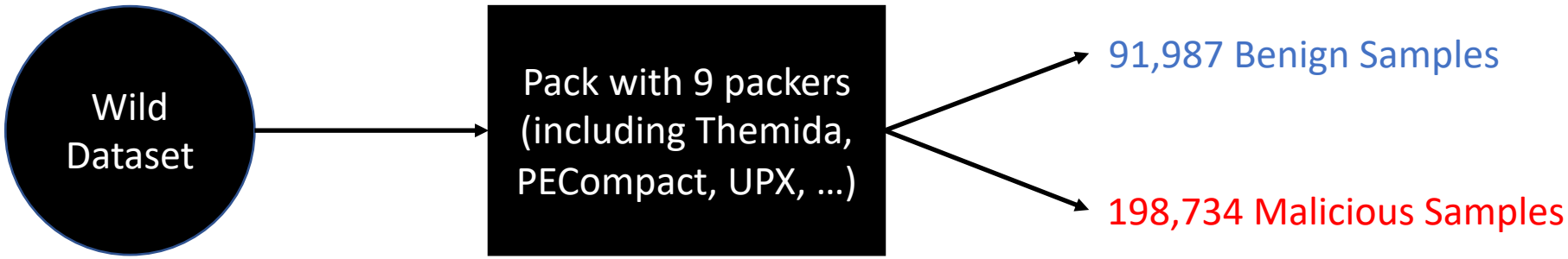
2. Lab Dataset:



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Nine Feature Categories

Category	# Features
PE headers	28
PE sections	570
DLL imports	4,305
API imports	19,168
Rich Header	66
Byte n-grams	13,000
Opcode n-grams	2,500
Strings	16,900
File generic	2

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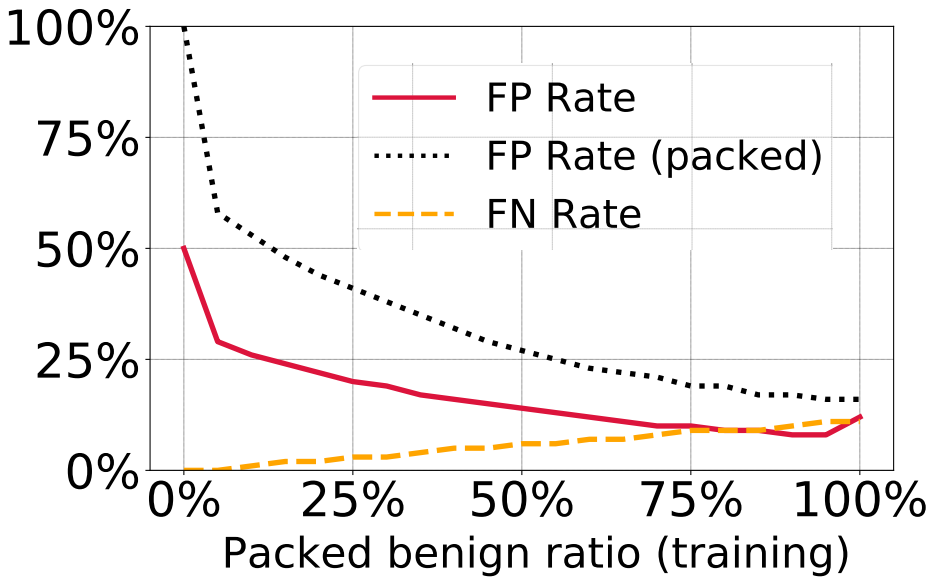
Our Research Questions

1. Do packers preserve static analysis features that are useful for malware classification?

Experiment “Different Packed Ratios (lab)”

1. We exclude packed benign samples from the training set
2. Then, we keep adding more packed benign samples to the training set



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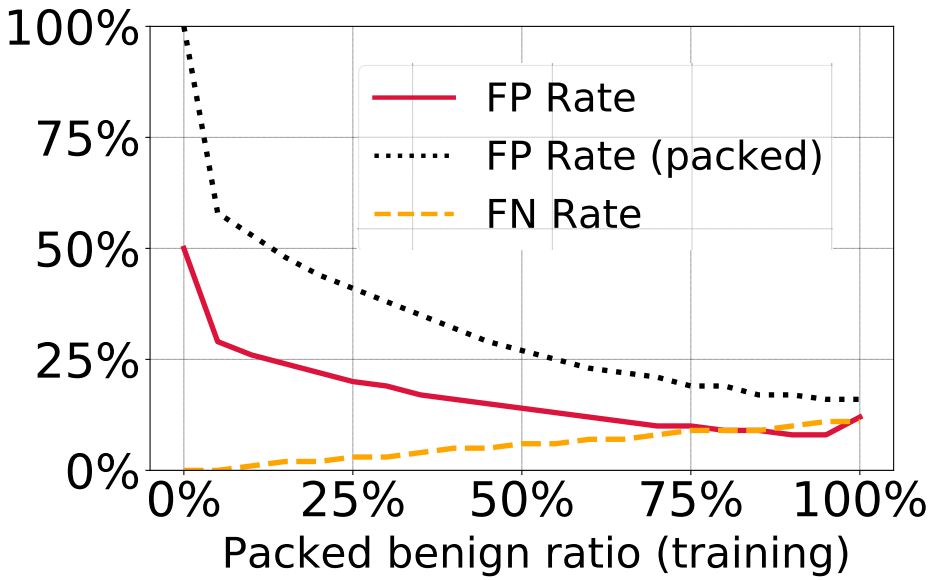

- Surprisingly, the classifier is doing ok!

But, How??

- We focused on one packer at a time to identify ***useful features*** for each packer!

1. Some packers (e.g., Themida) often keep the Rich Header.
2. Packers often keep .CAB file headers in the resource sections of the executables.
3. UPX keeps one API for each DLL.

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Our Research Questions

1. Do packers preserve static analysis features that are useful for malware classification?

Packers preserve some information when packing programs that may be “useful” for malware classification, however, such information does not necessarily represent the real nature of samples

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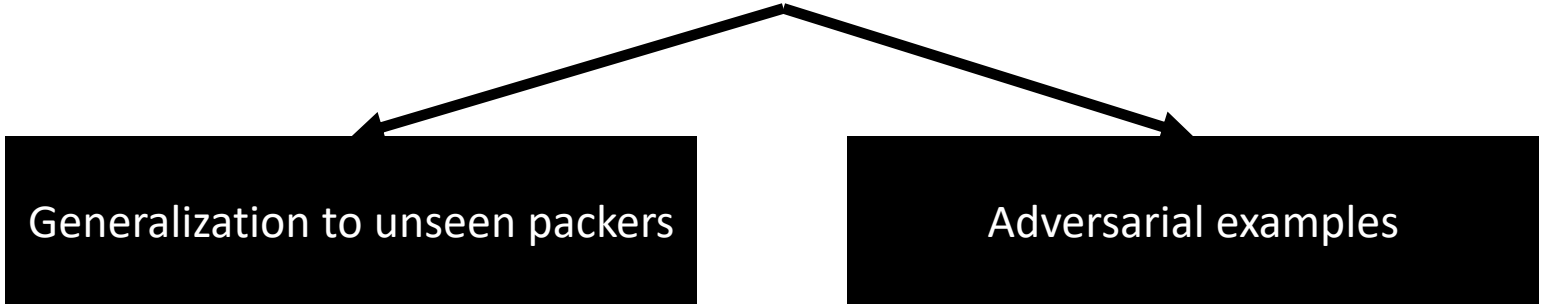

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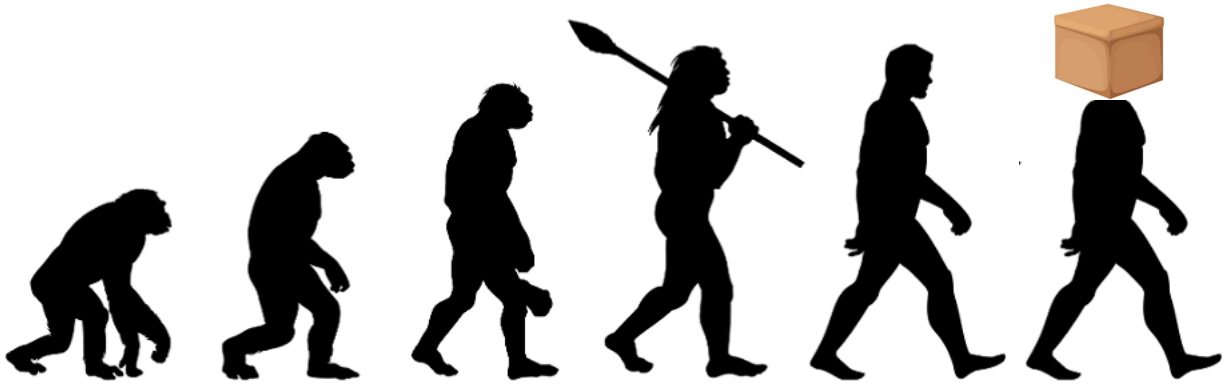
Generalization To Unseen Packers

- Runtime packers are evolving, and malware authors often tend to use their own custom packers



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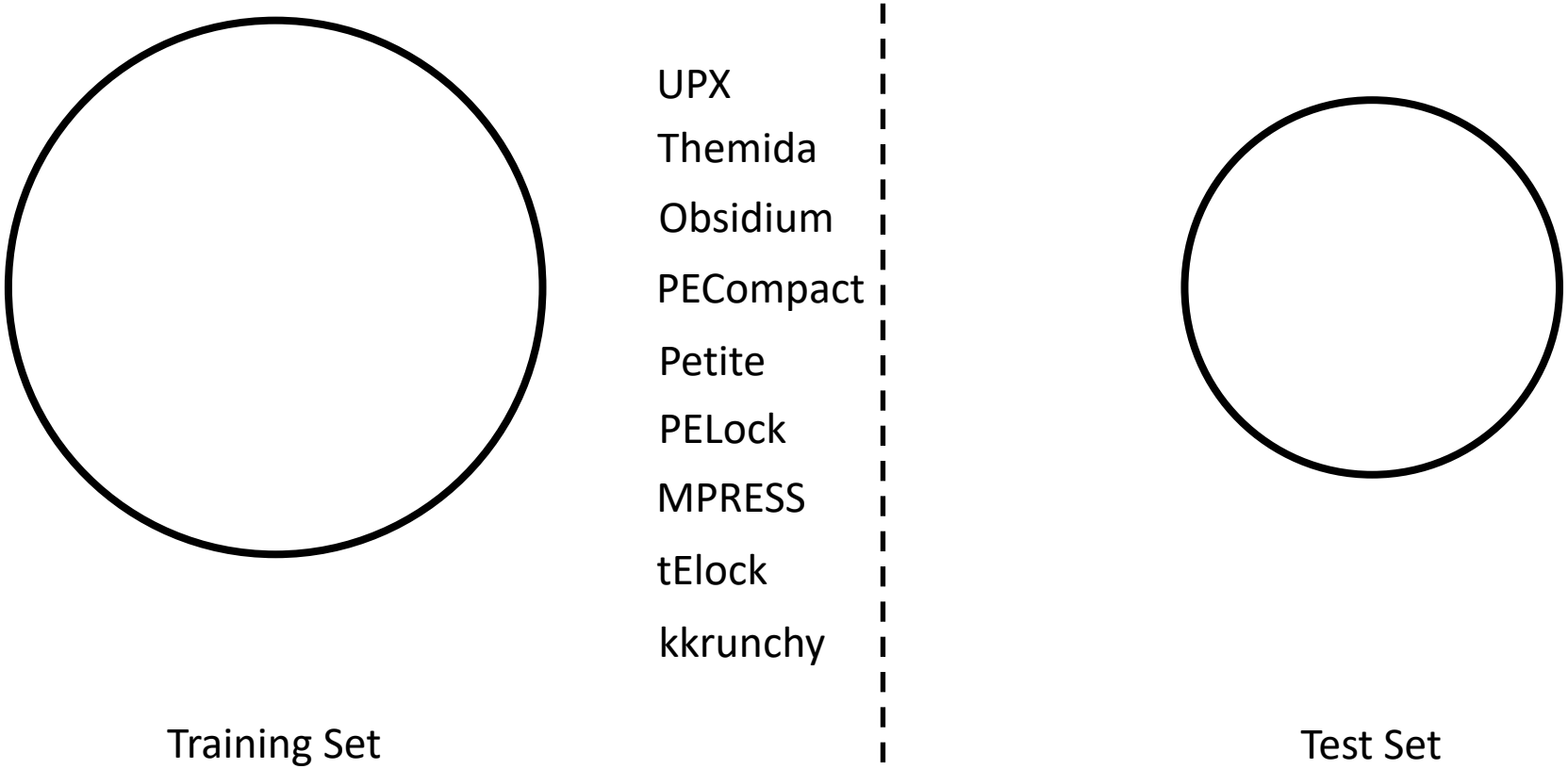
Generalization To Unseen Packers

1. Experiment “withheld packer”



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Generalization To Unseen Packers

1. Experiment “withheld packer”

Withheld Packer	FPR (%)	FNR (%)
PELock	7.30	3.74
PECompact	47.49	2.14
Obsidium	17.42	3.32
Petite	5.16	4.47
tElock	43.65	2.02
Themida	6.21	3.29
MPRESS	5.43	4.53
kkrunchy	83.06	2.50
UPX	11.21	4.34

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Generalization To Unseen Packers

2. Experiment “lab against wild”

- We train the classifier on Lab Dataset
- And evaluate it on packed executables in Wild Dataset

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Generalization To Unseen Packers

2. Experiment “lab against wild”

- We train the classifier on Lab Dataset
- And evaluate it on packed executables in Wild Dataset

- We observed the false negative rate of 41.84%, and false positive rate of 7.27%

Poor Generalization To Unseen Packers

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Adversarial Examples

- Machine-learning-based malware detectors shown to be vulnerable to adversarial samples

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- Packing produces features not directly deriving from the actual (unpacked) program
- Generating such adversarial samples would be easier for an adversary

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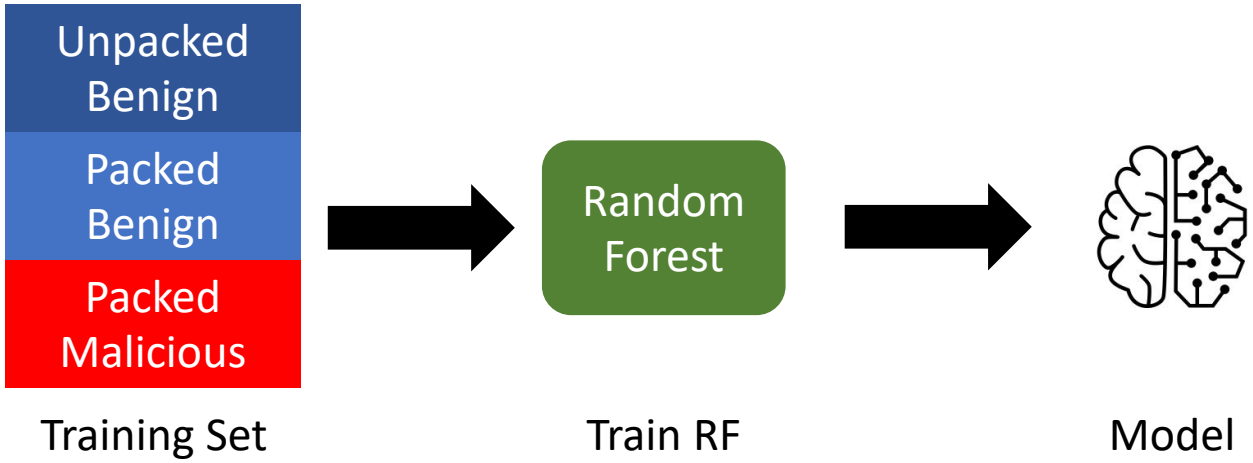
Adversarial Examples

Training:



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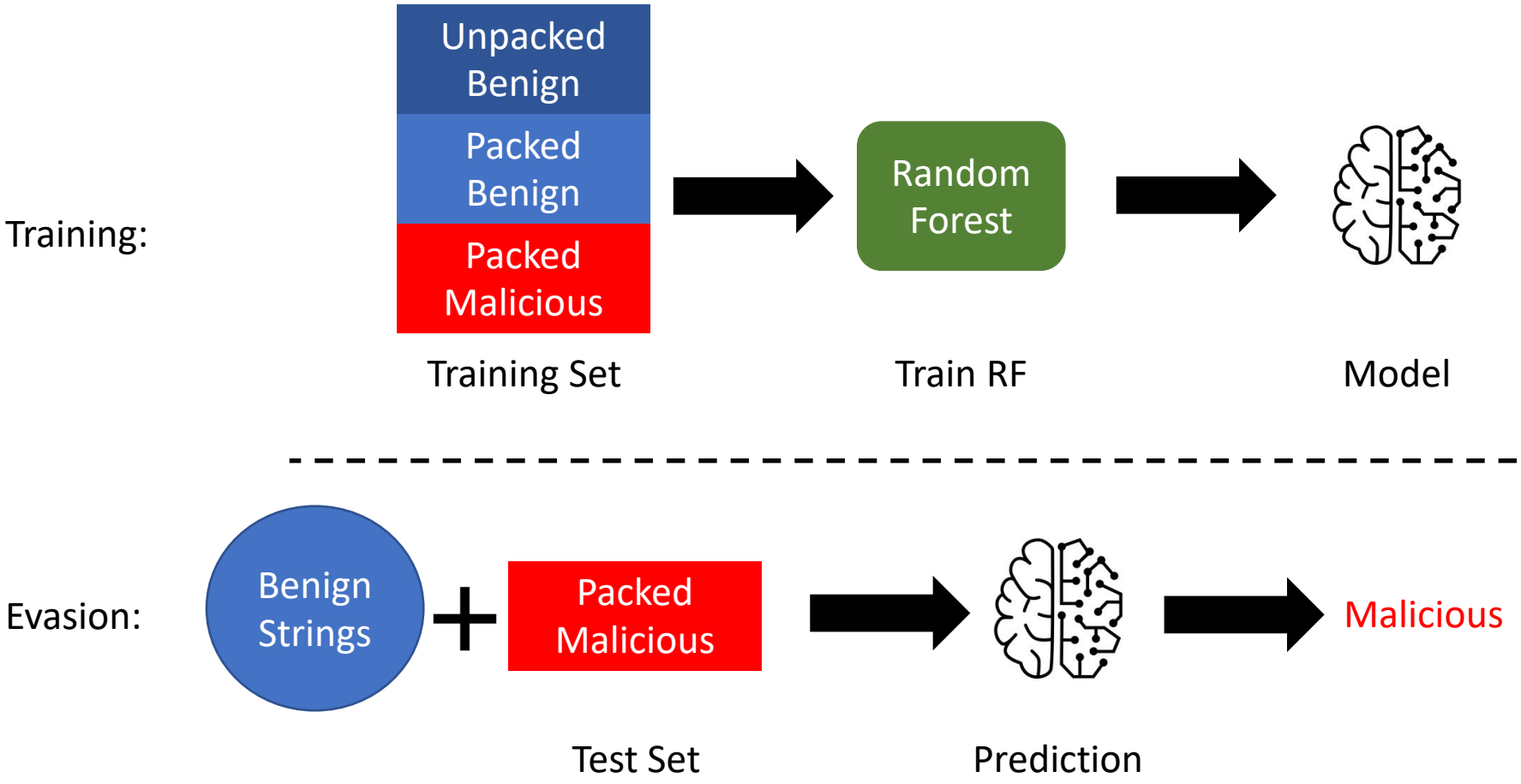

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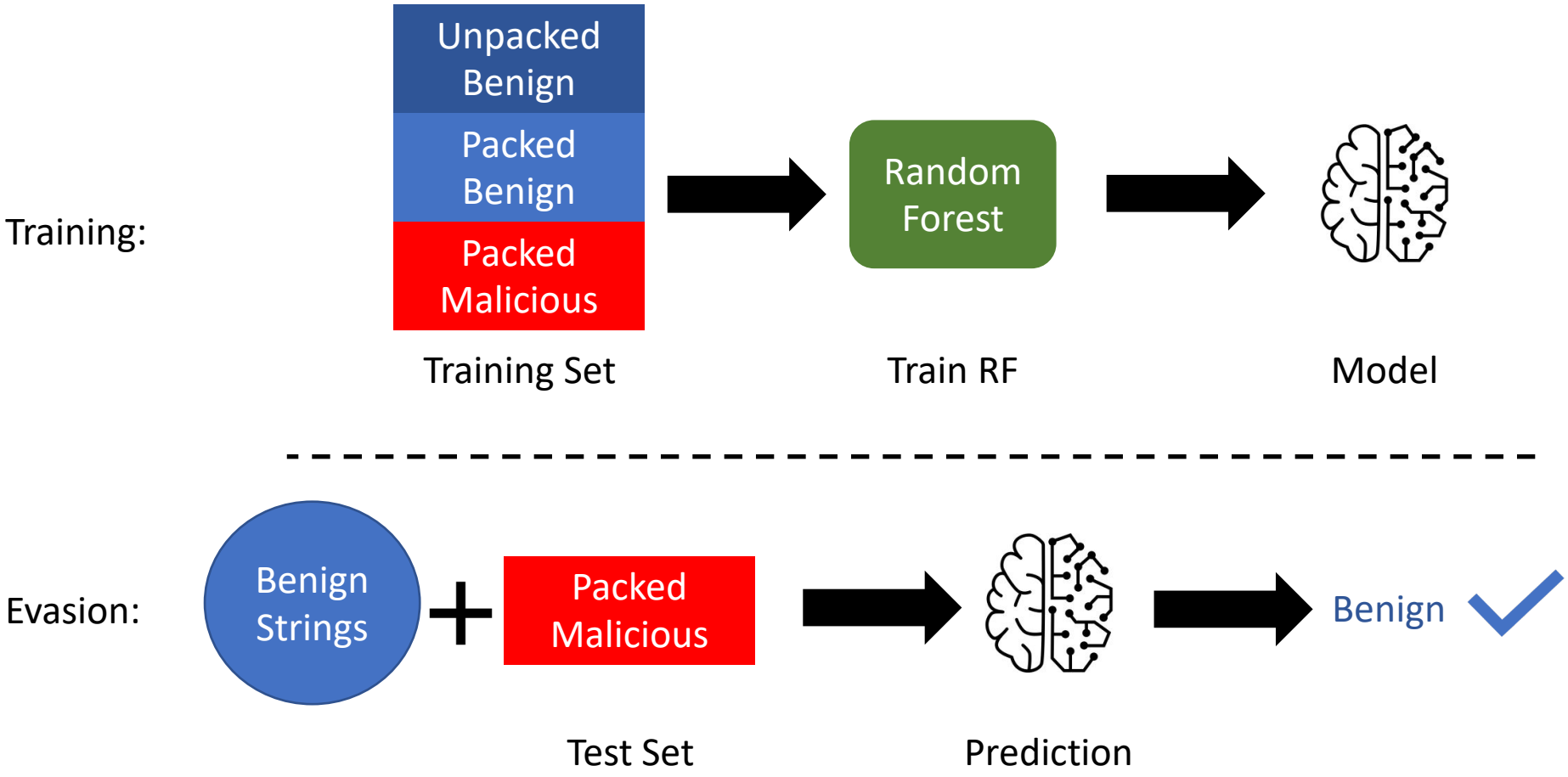

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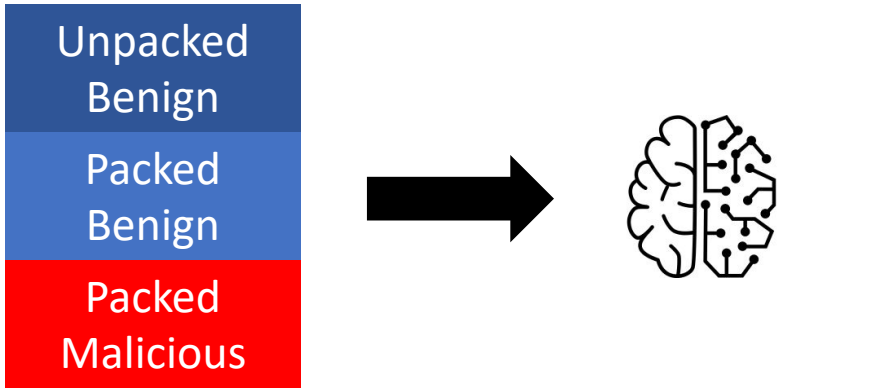
Machine Learning Static Evasion Competition

Machine Learning Static Evasion Competition

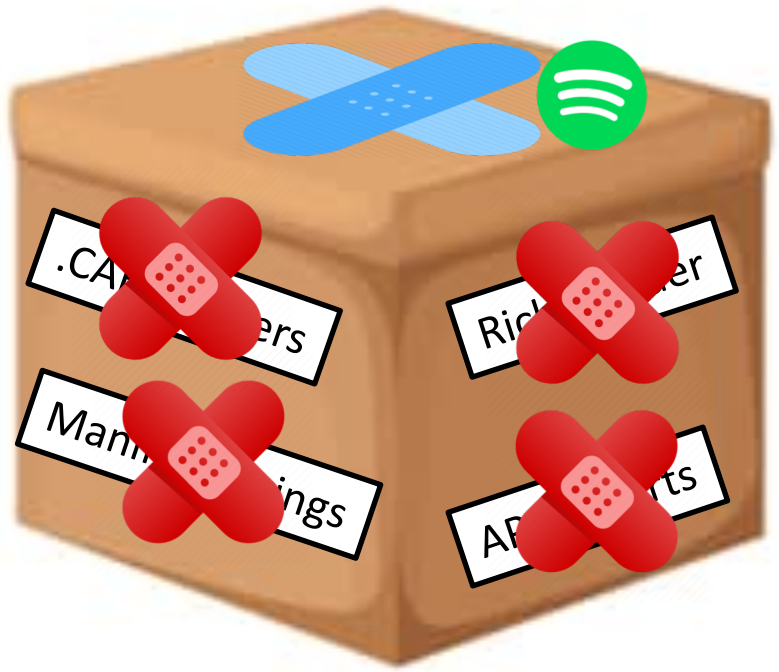
- Recently, a group of researchers found a very similar way to subvert an AI-based anti-malware engine
- By simply taking strings from an online gaming program and appending them to known malware, like WannaCry

Vulnerable To Trivial Adversarial Examples

Conclusion


Unpacked
Benign
Packed
Benign
Packed
Malicious
Training Set
Not Biased Model

Conclusion



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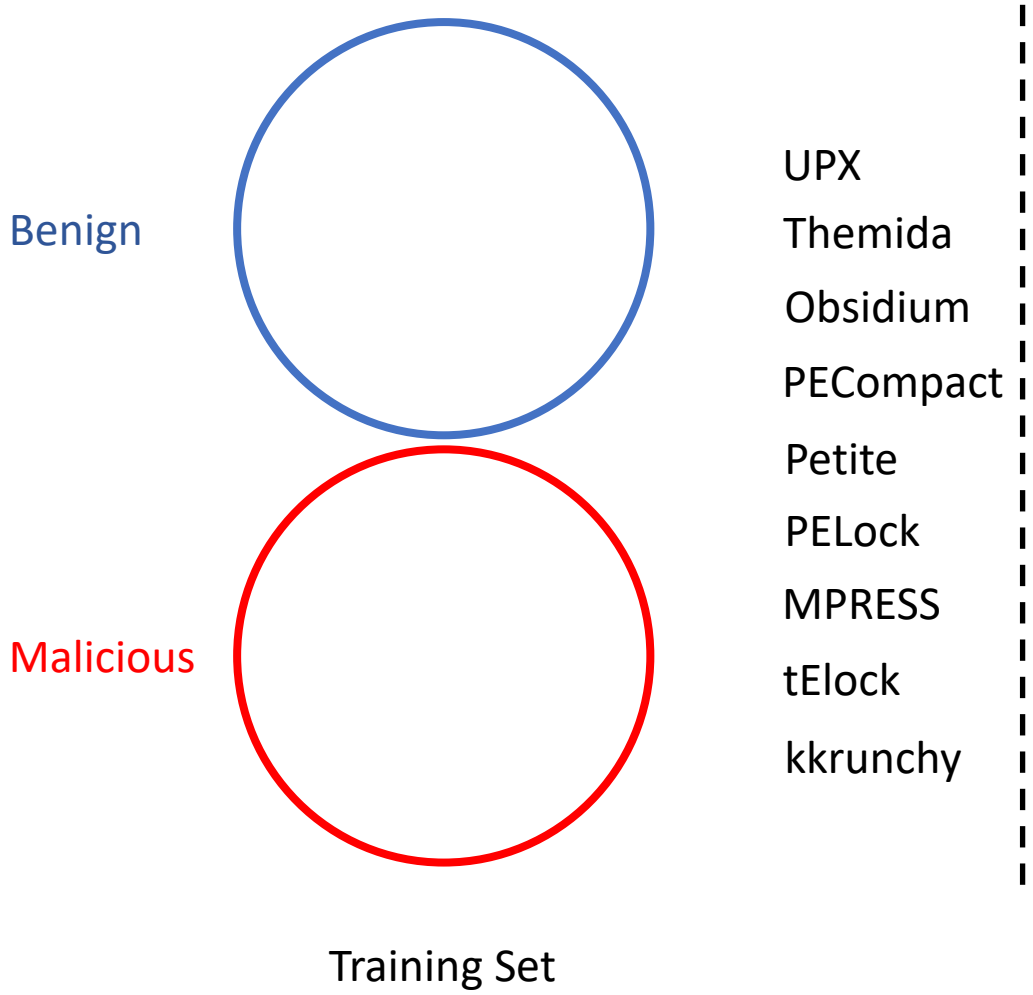
Reproducibility

- The source code and our datasets of 392,168 executables are available at https://github.com/ucsb-seclab/packware
- All experiments can be simply executed in the provided Docker image

Any Questions?

Experiment “Good-Bad Packers”



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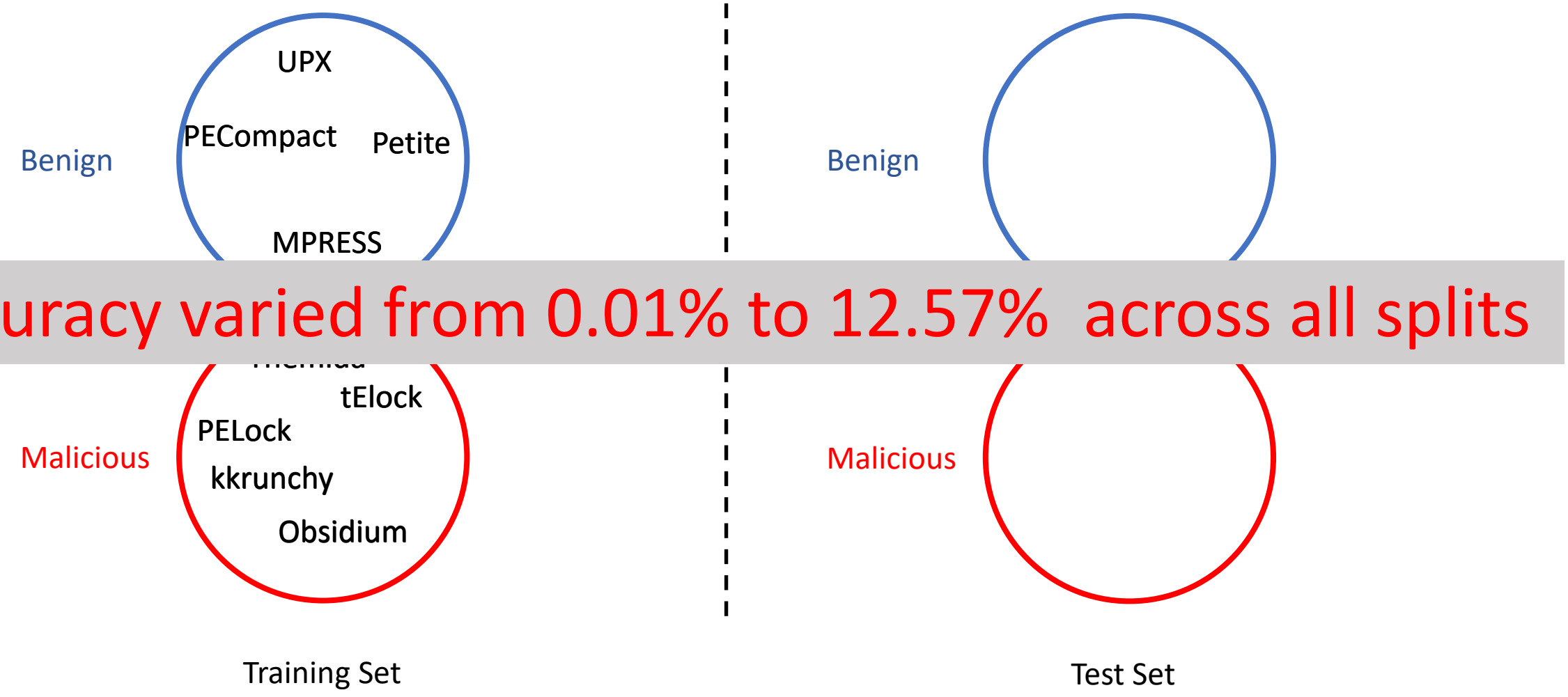

Accuracy varied from 0.01% to 12.57% across all splits