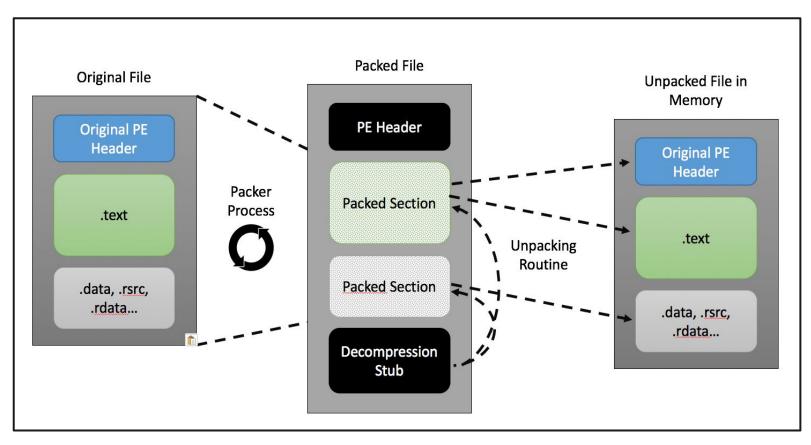
Prevalence and Impact of Low-Entropy Packing Schemes in the Malware Ecosystem

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Packing



Scope / Packing Definition

(Our definition of) packing implies

- Original code present, but NOT in an executable form
- Real code recovered at run-time

(Our definition of) packing does NOT include

- JIT compilers
- Droppers
- Emulators (Themida)
- Shellcode

Packed or not packed: that is the question



- Fundamental in malware analysis
- Wrong classification \Rightarrow
 - costly and time-consuming dynamic analysis trying to unpack the sample
 - pollute the datasets used in many malware analysis studies
 - even worse, EVASION
- Our (false) friend: the entropy
 - compressed/encrypted data has high entropy levels

Our Agenda

- 1. The propagation of low-entropy packed samples
- 2. The adopted schemes
- 3. Current tools/approaches vs. low-entropy packed malware

Dataset



Do malware authors use low-entropy schemes to evade entropy checks?

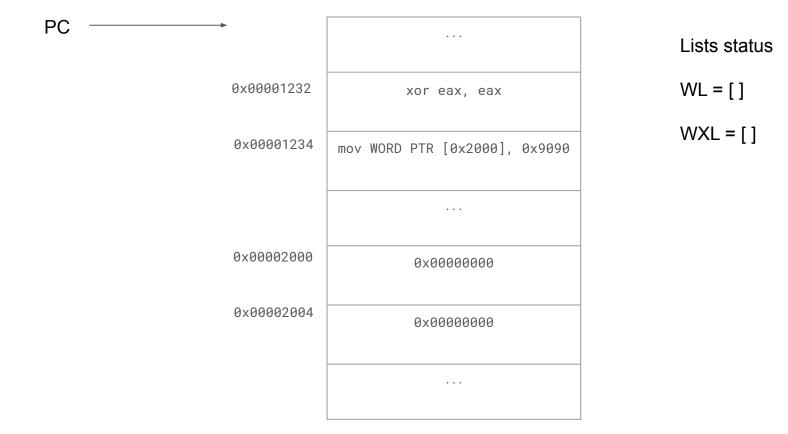
- 50.000 Portable Executable files (excluding libraries and .Net applications)
- 2013 2019
- Classified as malicious by more than 20 antivirus engines
- Entropy H < 7.0
 - entire file [1]
 - each section [2]
 - overlay data

Ugarte-Pedrero, Balzarotti, Santos, Bringas. Deep packer inspection: A longitudinal study of the complexity of run-time (2015)
pefile Python module Manalyze static analyzer for PE executables

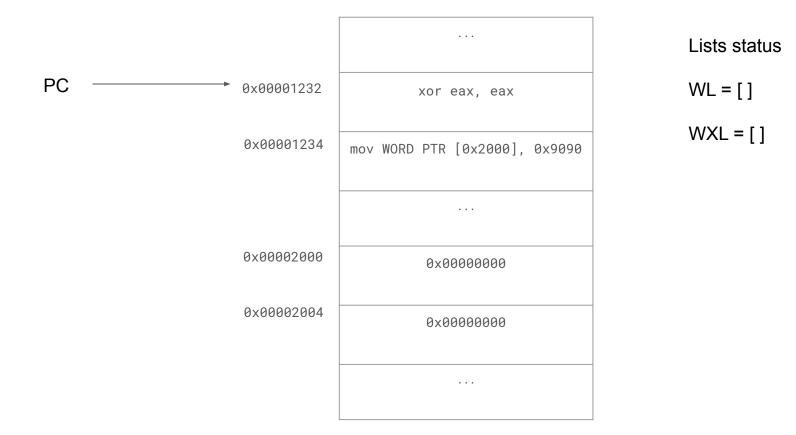
[1] Lyda and Hamrock. Using entropy analysis to find encrypted and packed malware (2007).

[2] Han and Lee. Packed PE file detection for malware forensics (2009).

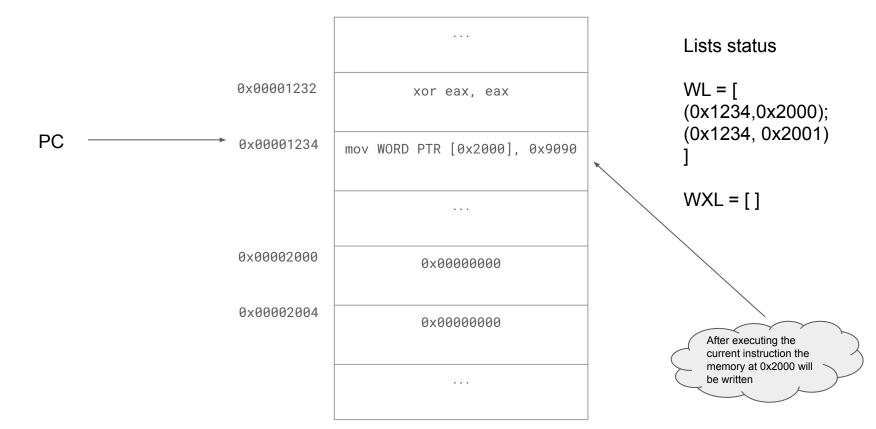
Packer Detector (1/5)



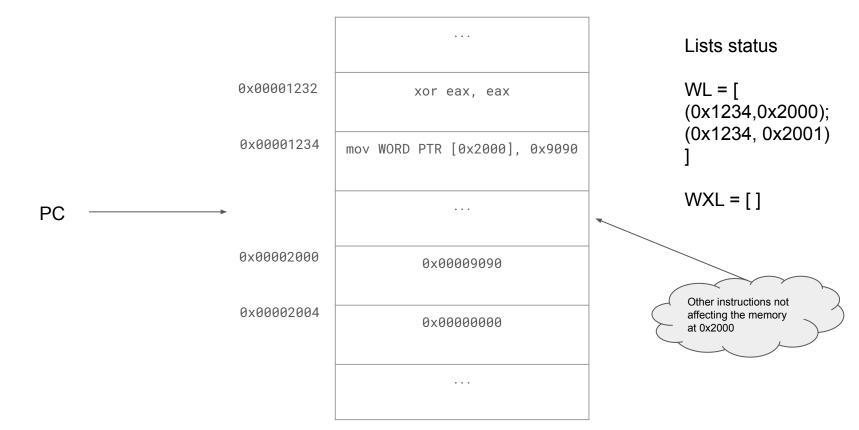
Packer Detector (%)



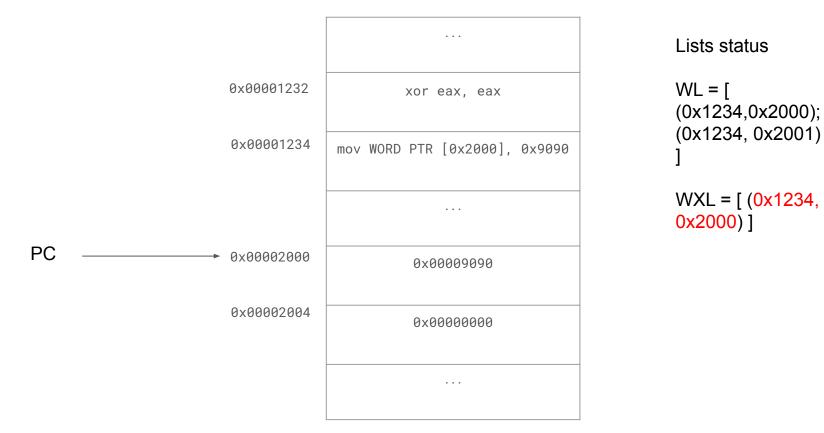
Packer Detector (3/5)



Packer Detector (%)



Packer Detector (5/5)



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Packer Detector - False Negatives

- False Negatives -- packed samples detected as not packed
 - unexpected crash
 - virtual environment detection
 - missing dependencies
 - incorrect command line arguments
- We discarded the samples that did not exhibit a sufficient runtime behavior
 - did not invoke at least 10 disk or network-related syscalls
 - samples whose executed instructions did not span at least five memory pages
- 50.000 **3.705** = 46.295

Hidden high-entropy data

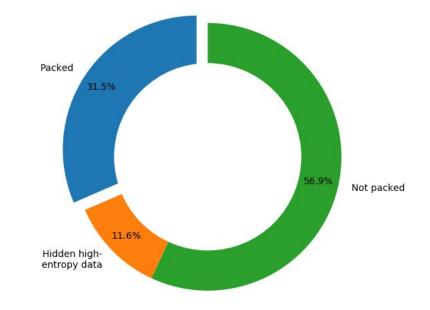
While packed with a high-entropy scheme, these samples evaded our set of filters

- Encrypted data, but the data was
 - not stored in any of the section
 - nor in the overlay area
- **11.6%** (5.386/46.295)
 - o dominated by two families: *hematite* and *hworld*
- E.g., *hematite*
 - \circ file infector
 - area created between the PE header and the first section

PE header
Encrypted data
.text
Encrypted data
.data

Packer Detector - Results

31.5% (14.583/46.295) \Rightarrow entropy alone is a very poor metric to select packed samples



Schemes Taxonomy w.r.t. Entropy

1. Decreasing

- Byte Padding
- \circ Encoding

2. Unchanged

- Transposition
- Monoalphabetic Substitution
- 3. Slightly Increasing
 - Polyalphabetic Substitution

Scheme Classifier

Relies on the output of Packer Detector ⇒ Written and eXecuted List [WXL]

- Every packing scheme needs to follow the same steps while unpacking
 - locate and access the source buffer that contains the packed data
 - perform operations on such data
 - write the unpacked data in the destination buffer
- We use PANDA to perform deterministic record and replay of a sample
 - $\circ \quad \langle \mathbf{PCx} , AWy \rangle \in [\mathbf{WXL}]$
 - backward data-flow analysis to locate the source buffer
- Decision making based on the byte distribution of source and destination buffers

Scheme Classifier - Results

Scheme	Туре	%
Padding	-	8.0
Encoding	standard	3.9
Lifeounig	custom	0.5
Mono-alphabetic Substitution	XOR	29.8
	ADD	5.2
	ROL/ROR	0.5
Transposition	-	0.3
Poly-alphabetic Substitution	XOR	46.9
i ory-alphabetic Substitution	ADD	2.8
Unknown		2.1

Case Study: Custom Encoding (Emotet)

Two layers of packing

- The second layer uses a custom high-entropy encryption with an 8-bytes long key
- The first layer reduces the entropy from 7.63 to 6.57
- Custom encoding + byte padding
- Packed data and keys stored in the sections: ".rsrc" and ".rdata"

Signature and Rule-Based Packing Detection

- Detect It Easy (DIE)
 - signatures based on a scripting language
- PEiD
 - signatures only contain low-level byte patterns
- Manalyze
 - signatures
 - PE structure heuristics
 - unusual section names
 - sections WX
 - low number of imported functions
 - resources bigger than the file itself
 - sections with H > 7.0

Signature and Rule-Based Packing Detection - Results

Dataset	Manalyze (signatures)	Manalyze (heuristics)	PEiD	Manalyze Sig ^ PEiD
Packed	242 (1.7%)	8358 (57.3%)	386 (2.6%)	214 (1.5%)
Not Packed	2518 (9.6%)	6023 (22.9%)	3438 (13.1%)	2487 (9.4%)
Hidden H-E data	0 (0%)	14 (0.3%)	2 (0.1%)	0 (0%)

- DIE detects no well-known packer in our entire dataset
- PEiD and Manalyze generated a large number of false positives
 - detected the presence of packing more often in unpacked samples than in the packed group
- Manalyze alerts are based on sections names used by some off-the-shelf packers
 - why the malware authors used those names?
 - they could be fake clues used on purpose to deceive automated tools

ML Packing Detection

- 15 approaches deal with this problem (SOTA)
- Several features categories
 - PE structure, heuristics, opcodes, n-grams, statistics, entropy
- Features vector (W): union of all features from previous studies
 - A separate features vector excluding the entropy (\tilde{W}) $\boldsymbol{\mathfrak{S}}$
- The most popular classifiers: SVM, RF, MLP
- Dataset: low entropy packed + not packed (~40K)

ML Packing Detection - Results

Err _{not} Packe	$d = \frac{1}{ TeS }$	FP notPacked	$Err_{packed} = \frac{ FN }{ TeS_{packed} }$			
	Classifier	Training-Testing	$Err_{notPacked}(W)$	$Err_{packed}(W)$	$Err_{notPacked}(\tilde{W})$	$Err_{packed}(\tilde{W})$
	SVM	75%-25% 50%-50% 25%-75%	4.43% 4.31% 4.44%	25.01% 28.41% 32.01%	4.12% 3.97% 4.11%	24.57% 26.20% 29.85%
	MLP	75%-25% 50%-50% 25%-75%	6.34% 6.87% 6.89%	12.70% 16.14% 11.91%	5.86% 6.24% 6.33%	12.15% 14.73% 12.93%
	RF	75%-25% 50%-50% 25%-75%	0.20% 0.18% 0.21%	32.77% 29.46% 28.84%	0.23% 0.20% 0.20%	31.54% 28.46% 26.83%
	ja da a		Consideri	ng H	Not Consi	dering H

NO classifier was able to identify accurately low-entropy packed malware!

Conclusions



- Low-entropy packing schemes are a real and widespread problem
- Existing static analysis techniques are unsuccessful against them
 - Entropy 🗙
 - Signature and Rule-Based X
 - \circ Machine Learning \mathbf{X}
- There is need for new solutions
- Low-entropy packing schemes must be considered in future experiments

-- Thank you for your attention --