

Padding Matters – Exploring Function Detection in PE Files

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Motivation

Motivation

- 300k+ new hash-unique malware samples daily
- automatic binary code analysis is essential
- 30 out of 61 binary similarity methods operate at the function level according to Haq and Caballero [\[1\]](#)

Detecting functions in binaries is a critical component of cyber defence.

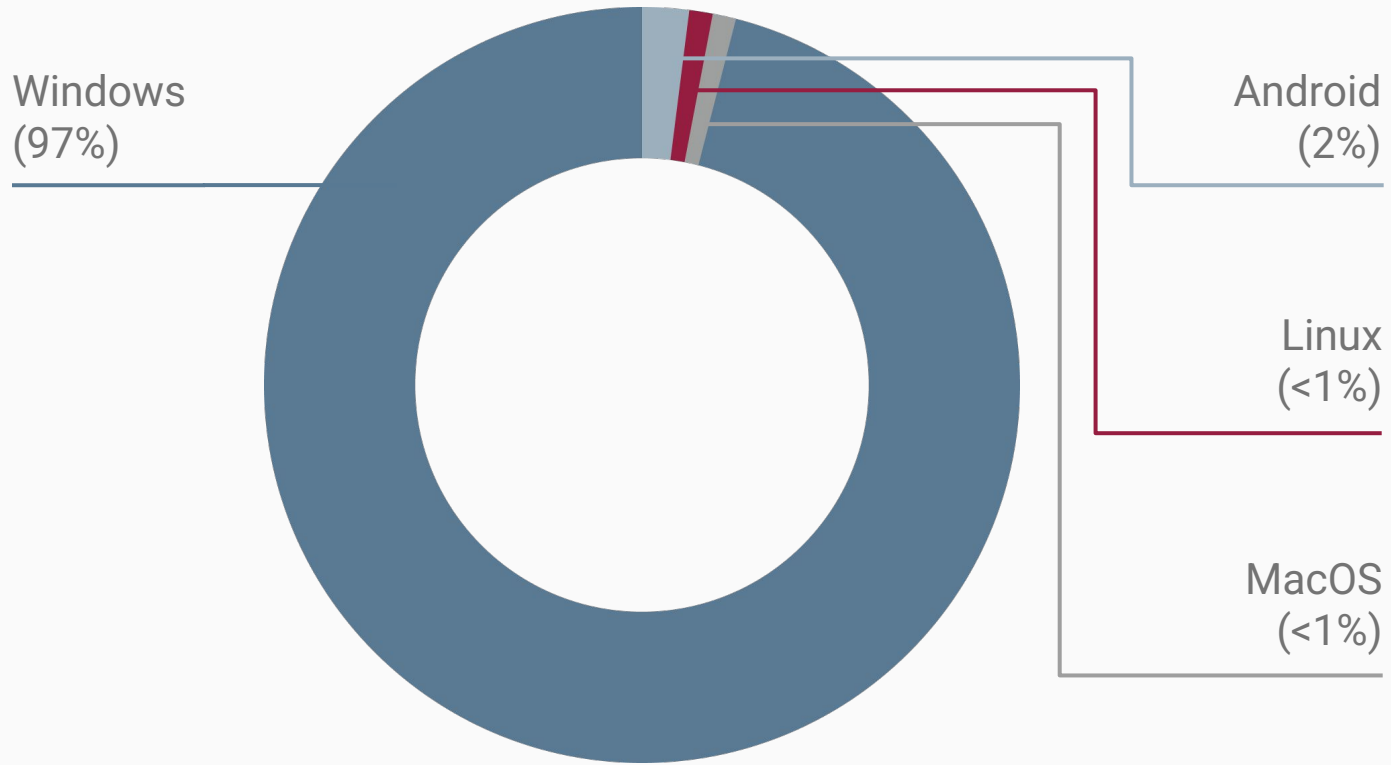
```
f function start()  
  
    var today =  
    var h = today.  
    var m = today.  
    var s = today.  
    m = correctTime  
    s = correctTime  
    document.getElementById  
    //calling the  
    var t = setTime  
  
    //adding the zero  
function correctTime
```

Tool/Paper	Samples		Functions	
	ELF	PE	ELF	PE
FunProbe [2]	19.09k	0	3.06m	0
DeepDi [3]	1.4k	688	n/a	n/a
Koo et al. [4]	152	0	769k	0
FETCH [5]	43	0	1.11m	0
XDA [6]	2.6k	528	n/a	n/a
Jima [7]	3.8k	0	4.91m	0
LEMNA [8]	2.1k	0	n/a	n/a
Nucleus [9]	324	152	n/a	379k
rev.ng [10]	1.9k	0	n/a	0
Andriesse et al. [11]	829	152	1.53m	379k
Shin et al. [12]	2.1k	136	589k	188k
BAP/ByteWeight [13]	2.1k	136	589k	188k
Rosenblum et al. [14]	728	443	284k	100k

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Malware distribution according to AV-TEST



Function Detection

Identifying sequences of bytes in a compiled binary that correspond to individual functions in the original source code.

Byte Sequence B

55 48 89 e5 48 83 ec 10 89 c1 31 c0 89 4d f8 e8 b4 ff ff ff 48 83 c4 10 5d c3 48 83 ec 10 55 48
/ |
 $B[0]$ $B[1]$... $B[30]$ $B[31]$

Byte Sequence B

55 48 89 e5 48 83 ec 10 89 c1 31 c0 89 4d f8 e8 b4 ff ff ff 48 83 c4 10 5d c3 48 83 ec 10 55 48

$F_1 = [B[0], B[1], B[2], B[3], B[4], B[5], B[6], B[7], B[11], B[12], B[13], B[14], B[15], B[16], B[17]]$

Byte Sequence B

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$F_1 = [B[0], B[1], B[2], B[3], B[4], B[5], B[6], B[7], B[11], B[12], B[13], B[14], B[15], B[16], B[17]]$

$F_2 = [B[8], B[9], B[10], B[20], B[21], B[22], B[23], B[24], B[25], B[26]]$

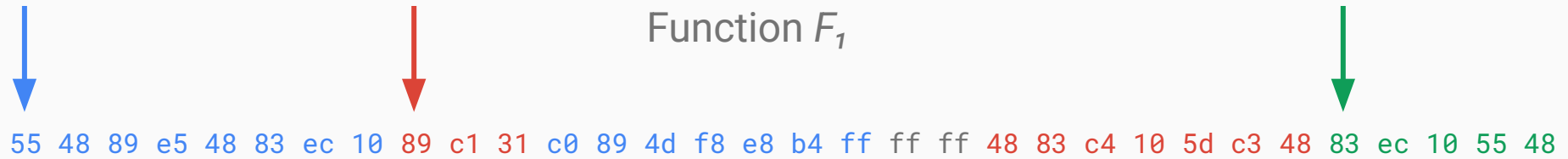
Byte Sequence B

55 48 89 e5 48 83 ec 10 89 c1 31 c0 89 4d f8 e8 b4 ff ff ff 48 83 c4 10 5d c3 48 83 ec 10 55 48

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$F_2 = [B[8], B[9], B[10], B[20], B[21], B[22], B[23], B[24], B[25], B[26]]$

$F_3 = [B[27], B[28], B[29], B[30], B[31]]$



$F_1 = [B[0], B[1], B[2], B[3], B[4], B[5], B[6], B[7], B[11], B[12], B[13], B[14], B[15], B[16], B[17]]$

Function start: $B[0]$
Function end: $B[17]$
Function boundary: $(B[0], B[17])$

Pattern Recognition

- characteristic byte sequences form signatures
- prologue patterns for function starts
- epilogue patterns for function ends

Prologue

55	PUSH RBP
4889e5	MOV RBP,RSP
4883ec10	SUB RSP,0x10
554889e54883ec10	

Pattern Recognition

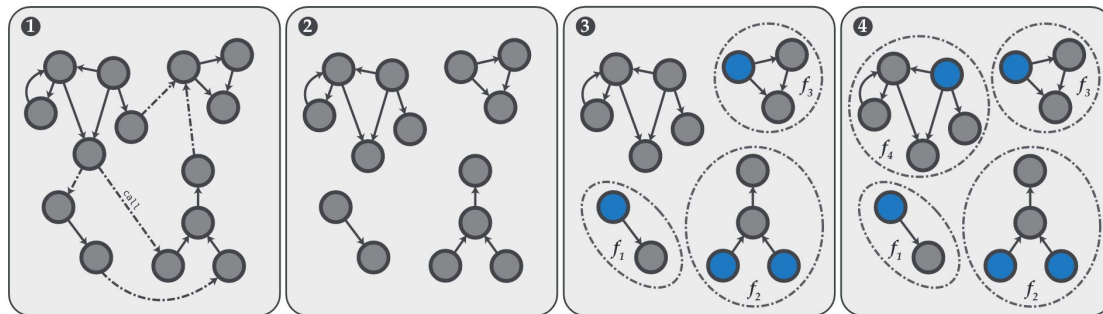
- recognizing prologue and epilogue
- prologue pattern for function starts
- epilogue pattern for function ends
- normalization of immediate values

Prologue

```
55          PUSH RBP
4889e5      MOV RBP,RSP
4883ec10    SUB RSP,0x10
554889e54883ec??
```

Static Analysis

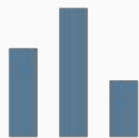
- disassemble
- construct control-flow graph (CFG)
- find call targets
- recognize and analyze jump tables
- some tools incorporate machine learning



Function Detection of Nucleus [9]

FuncPEval

FuncPEval



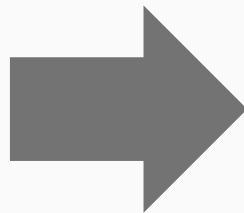
Four samples

Chromium

- x86
- x64

Conti
Ransomware

- x86
- x64



1,092,820
functions

FuncPEval – Chromium

- open source browser Chromium version 109
- compiled and linked with LLVM Clang
 - Ox, Os, Oy, O1, Ot, and O2 optimization levels
- public snapshot published by Google
 - `chrome.dll`
 - `chrome.dll.pdb`

http://commondatastorage.googleapis.com/chromium-browser-snapshots/index.html?prefix=Win_x64/1069922/

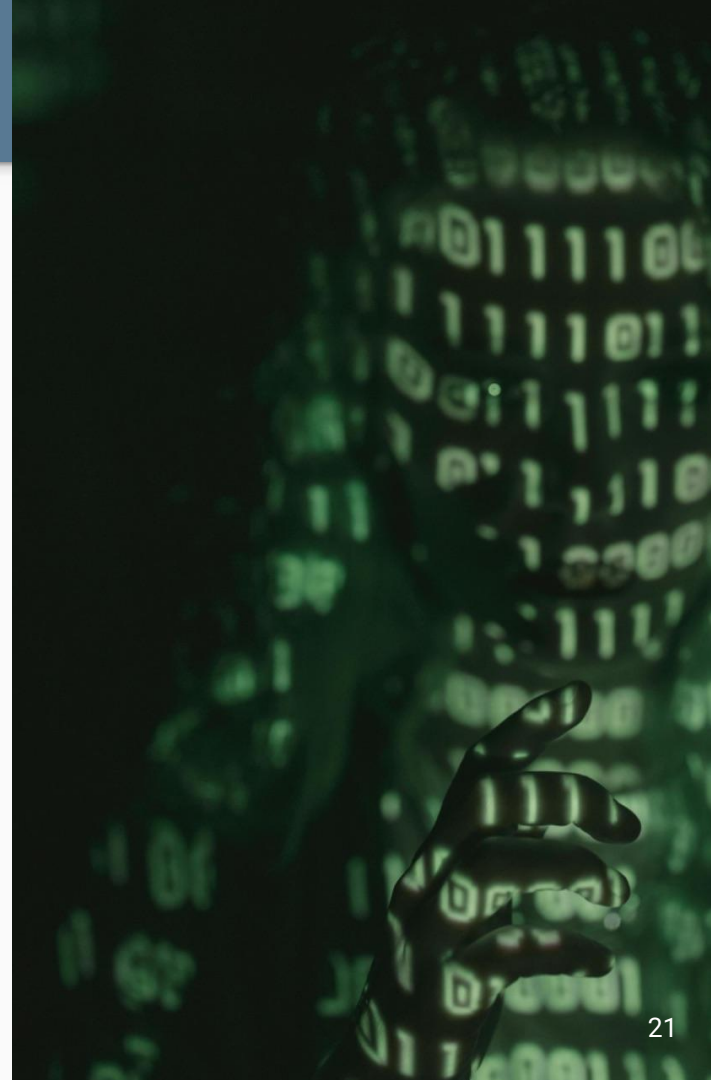
<http://commondatastorage.googleapis.com/chromium-browser-snapshots/index.html?prefix=Win/1069956/>



chromium

FuncPEval – Conti

- Conti Ransomware version 3
- source code leaked in early 2022
- `cryptor.exe`
 - component that encrypts machine
- compiled and linked using Visual Studio 2022



Dataset	Functions		
	total	unique	normalized
Chromium x64	542,902	536,182 (99%)	315,745 (58%)
Conti x64	662	659 (99%)	450 (68%)
BAP PE [13]	94,548	65,733 (70%)	18,169 (19%)

Evaluation

Evaluation – Tools



Industry Standards

- IDA Pro 7.7
- Ghidra 10.0.4



Scientific Publications

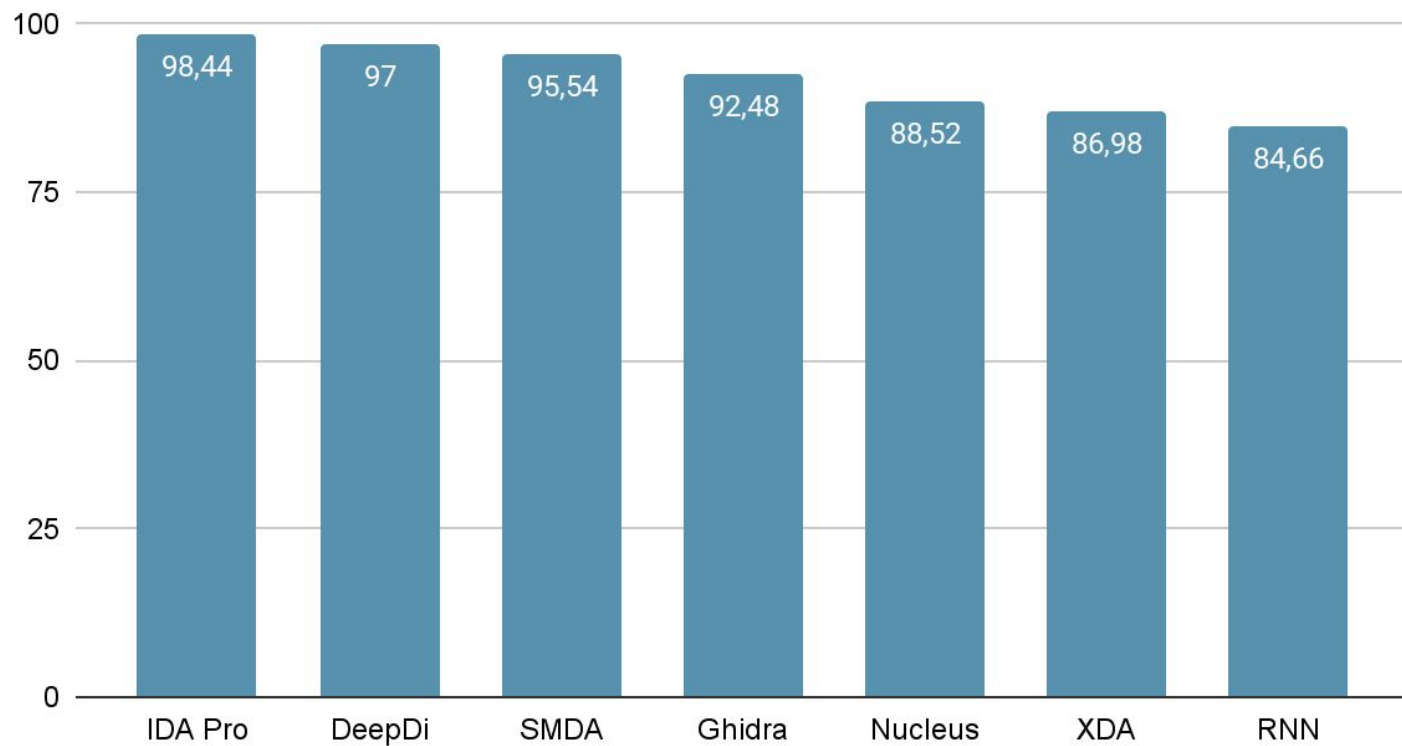
Non-ML-based

- Nucleus [\[9\]](#)
- SMDA [\[15\]](#)

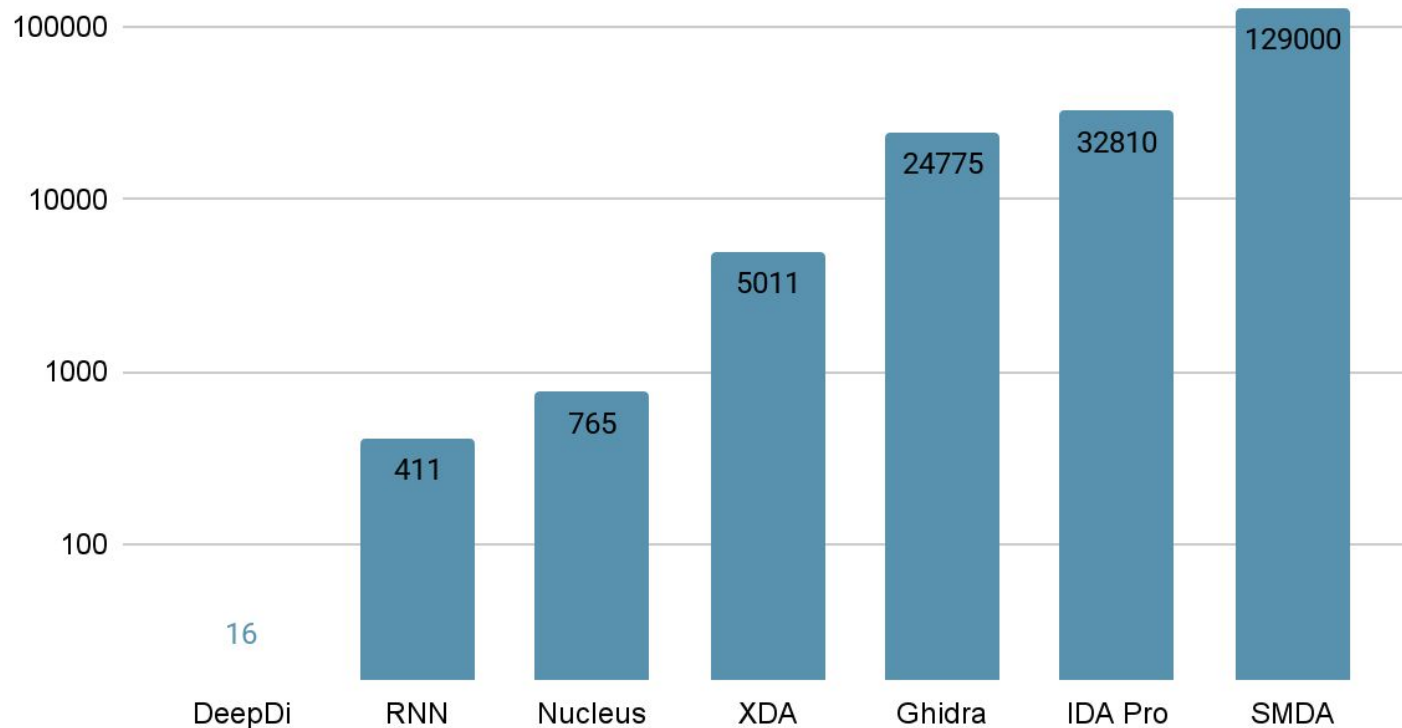
ML-based

- RNN [\[12\]](#)
- XDA [\[6\]](#)
- DeepDi [\[3\]](#)

Chromium x64 – F1-score



Chromium x64 – Execution time in seconds



Padding

Padding – Motivation

What features
do ML-based tools
learn?

How do they
generalize over different
compiler versions?



Padding

- compiler align functions at multiples of 16
- padding fills bytes in between
 - 0xCC int3
 - 0x90 NOP
- padding not part of control flow
 - can be changed arbitrarily

```
140018fd7 49 8b c1      MOV     RAX,R9
140018fda c3           RET
140018fdb cc         ??     CCh
140018fdc cc         ??     CCh
140018fdd cc         ??     CCh
140018fde cc         ??     CCh
140018fdf cc         ??     CCh
```

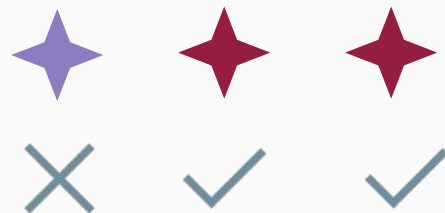
```
*****
*                                                     FUN
*****
byte * __fastcall FUN_140018fe0
    assume GS_OFFSET = 0xff0000
    byte * RAX:8 <RETURN>
    char * RCX:8 param_1
    FUN_140018fe0
140018fe0 0f b6 01      MOVZX   EAX,byte ptr
140018fe3 4c 8d 41 01    LEA     R8,[param_1 +
140018fe7 84 c0         TEST    AL,AL
140018fe9 75 64         JNZ     LAB_14001904f
```

Padding – Randomization

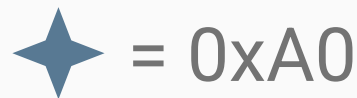
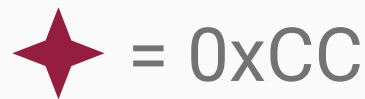
- 1 Collect up to 20 bytes before the beginning of the function.



- 2 Consider only those bytes that do not belong to a preceding function.



- 3 Replace each padding byte of value 0xCC with a random arbitrary byte value.



140018fd7	49 8b c1	MOV	RAX,R9
140018fda	c3	RET	
140018fdb	cc	??	CCh
140018fdc	cc	??	CCh
140018fdd	cc	??	CCh
140018fde	cc	??	CCh
140018fdf	cc	??	CCh

140018fd7	49 8b c1	MOV	RAX,R9
140018fda	c3	RET	
140018fdb	c0	??	C0h
140018fdc	15	??	15h
140018fdd	35	??	35h
140018fde	29	??	29h
140018fdf	63	??	63h

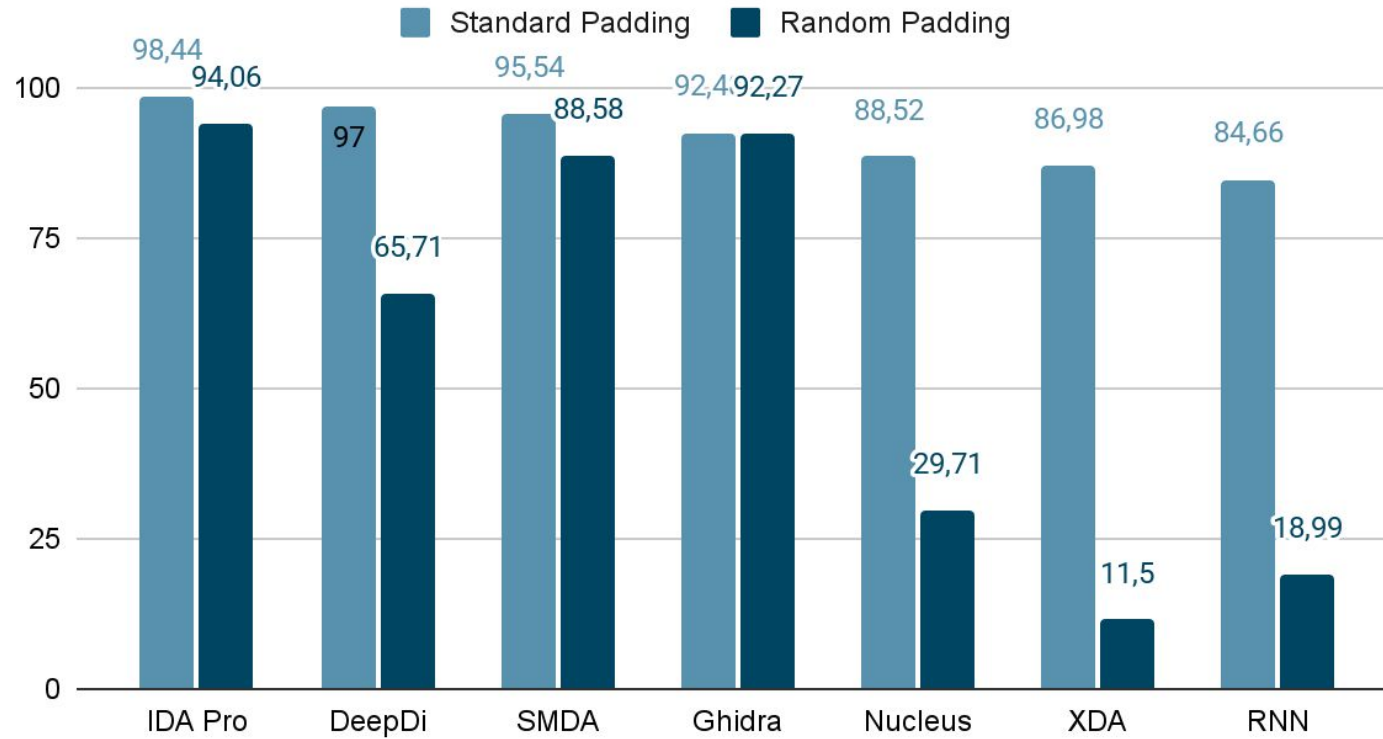
standard padding

```
*****
*
*****
byte * __fastcall FUN_140018f
    assume GS_OFFSET = 0xff000
    byte *      RAX:8      <RETURN>
    char *      RCX:8      param_1
    FUN_140018fe0
140018fe0 0f b6 01      MOVZX    EAX,byte ptr
140018fe3 4c 8d 41 01    LEA      R8,[param_1 +
140018fe7 84 c0          TEST     AL,AL
140018fe9 75 64          JNZ     LAB_14001904f
```

random padding

```
*****
*
*****
byte * __fastcall FUN_140018f
    assume GS_OFFSET = 0xff000
    byte *      RAX:8      <RETURN>
    char *      RCX:8      param_1
    FUN_140018fe0
140018fe0 0f b6 01      MOVZX    EAX,byte ptr
140018fe3 4c 8d 41 01    LEA      R8,[param_1 +
140018fe7 84 c0          TEST     AL,AL
140018fe9 75 64          JNZ     LAB_14001904f
```

Chromium x64 – F1-score



Conclusion

Conclusion

- new PE dataset spanning 1,092,820 functions
 - evaluated seven tools
 - IDA Pro highest F1-score
 - DeepDi fastest execution time
 - ML-based tools sensitive to padding modification
-
- Artifacts
 - FuncPEval
 - tooling for ground truth extraction
 - all trained models and training data
 - <https://github.com/internet-sicherheit/Padding-Matters---Exploring-Function-Detection-in-PE-Files>

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Thank you.

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<https://threatlab.if-is.net>

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