Compiler-Agnostic Function Detection in Binaries

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Disassembly in Systems Security

Disassembly is the backbone of all binary-level systems security work (and more)

- Control-Flow Integrity
- Automatic Vulnerability/Bug Search
- Lifting binaries to LLVM/IR (e.g., for reoptimization)
- Malware Analysis
- Binary Hardening
- Binary Instrumentation
- . . .

Results from Previous Work

Function detection currently the main disassembly challenge

- Even function start detection yields many FPs/FNs (20%+)
- Complex cases: non-standard prologues, tailcalls, inlining, ...
- Binary analysis commonly requires function information



Function Detection: False Negative

Listing: False negative indirectly called function for IDA Pro 6.7 (gcc compiled with gcc at 03 for x64 ELF)

6caf10 <ix86_fp_compare_mode>:

6caf10: mov 0x3f0dde(%rip),%eax 6caf16: and \$0x10,%eax 6caf19: cmp \$0x1,%eax 6caf1c: sbb %eax,%eax 6caf1e: add \$0x3a,%eax 6caf21: retq Z

Function Detection: False Positive

Listing: False positive function (shaded) for Dyninst (perlbench compiled with gcc at 03 for x64 ELF)

6b990 <perl_pp_enterloop>:</perl_pp_enterloop>			
		[]	
	46ba02:	ja	46bb50 <perl_pp_enterloop+0x1c0></perl_pp_enterloop+0x1c0>
	46ba08:	mov	%rsi,%rdi
	46ba0b:	shl	%cl,%rdi
	46ba0e:	mov	%rdi,%rcx
	46ba11:	and	\$0x46,%ecx
	46ba14:	je	46bb50 <perl_pp_enterloop+0x1c0></perl_pp_enterloop+0x1c0>
		[]	
	46bb47:	рор	%r12
	46bb49:	retq	
	46bb4a:	nopw	0x0(%rax,%rax,1)
	46bb50:	sub	\$0x90,%rax

Signature-Based Function Detection

- Most current approaches scan for prologue/epilogue signatures
 - IDA Pro, Dyninst, ByteWeight (Bao et al. 2014), (Shin et al. 2015)
- Error-prone: sigs may be missing/optimized away
- Non-scalable: new sigs needed for every compiler version/platform
- Even machine learning approaches need continuous retraining

Compiler-Agnostic Function Detection

- We propose a signature-less approach based on structural analysis of the Control-Flow Graph (CFG)
- Basic premise: Weakly Connected Components Analysis
- Compiler-agnostic: no training/maintenance needed
- Able to detect all basic blocks of a function
- Inherent support for corner cases such as non-contiguous functions



(1) Disassemble binary and generate interprocedural CFG (linear disassembly + switch/inline data detection)



(2) Hide edges $e \in E_{call}$



(3) Locate directly called entry points and expand functions by following control flow (ignoring direction)



④ Find remaining functions using Connected Components Analysis, analyze control-flow to find entry points

Evaluation



Function Start Detection

- Overall average F-score of 0.96 for SPEC CPU 2006 (similar for servers)
- Stable performance across compiler/platform/optimization level
- Main improvement over others: higher recall (fewer FNs)

Evaluation



Function Boundary Detection

- Overall average F-score of 0.90 for SPEC CPU 2006
- Even better for C-only server tests (average F-score 0.97)
- Again, more stable than other approaches
- Best alternative: IDA Pro, average F-score of 0.84

More Results

- In-depth analysis of results (including FPs/FNs) in paper
- Most complex cases handled correctly (non-contiguous functions, multi-entry functions, ...)
- Main problematic case: tail calls

Evaluation



Runtime

• On par with fastest alternatives

Resistance to Obfuscation

- Although this talk is in the Malware session, we do not explicitly target malware
- That said, our approach is agnostic of some basic obfuscation approaches
 - Instruction-level polymorphism
 - Mangling of function prologues/epilogues
 - Some control flow obfuscations (e.g., converting direct calls to indirect, branching functions, ...)
- But we make no promises for arbitrary obfuscations!

Performance Discrepancies

- During our evaluation, noticed far lower performance for ByteWeight than previously reported (Bao et al. 2014)
- Mean F-score 0.32 points lower than expected
- Observation persists for gcc (v4.7-v5.1), clang, and Visual Studio
- Upon closer inspection, discovered issues with test suite used to evaluate *all* major machine learning-based function detection work (Bao et al. 2014 and Shin et al. 2015)

Test Suite Issues

- Both Bao et al. and Shin et al. use ten-fold cross-validation to evaluate their work
- Partition test suite into training set (B_T , 90% of binaries) and evaluation set (B_E)
- Repeat ten times such that each binary is in B_E exactly once
- Crucial to ensure sufficient variation in test suite to prevent overfitting!

Test Suite Issues

- Linux test suite used by Bao et al. and Shin et al. consists of coreutils (106 binaries), binutils (16 binaries), and findutils (7 binaries)
- Average coreutils binary shares 54% of its functions with *all* other coreutils binaries
- Average coreutils binary shares 94% of its functions with *at least one other* coreutils binary
- For the average coreutils binary in B_E , at least 86% of its functions are expected to occur in B_T
- Large degree of overfitting in evaluation of machine learning approaches, re-evaluation needed

- We introduced a novel compiler-agnostic function detector
- No maintenance/learning phase required
- More accurate results than existing approaches
- Inherent support for complex cases
- Available open source: https://www.vusec.net/projects/function-detection/
- Features export to IDA $\mathsf{Pro} \to \mathsf{easy}$ to use in real-world setting