CE 815 – Secure Software Systems

ML-Based Vulnerability Detection Methods (GraphSPD)

Mohammad Haddadian/Mehdi Kharrazi Department of Computer Engineering Sharif University of Technology



Acknowledgments: Some of the slides are fully or partially obtained from other sources. A reference is noted on the bottom of each slide, when the content is fully obtained from another source. Otherwise a full list of references is provided on the last slide. Thanks to Mohammad Haddadian for the help on the slides.

Introduction



- As announced by the 2021 report, 98% of codebases contain open source components
- Meanwhile, 84% of code-bases have at least one open-source vulnerability
- 60% of them contain high-risk vulnerabilities
- By exploiting the OSS vulnerabilities reported in the vulnerability databases (e.g., NVD), attackers can perform "N-day" attacks against unpatched software systems

• Average users need to timely detect and apply security patches before being exploited by armored attackers. Fall 1403 CE 815 - Secure Software Systems [GraphSPD] 3

• A large volume of OSS security patches (e.g., GitHub commits fixing vulnerabilities) are silently released.

From 7f9822a48213dd2feca845dbbb6bcb8beb9550de Subject: [PATCH] Add blinding to a DSA signature

This is based on side channel attacks demonstrated by (NCC Group) for ECDSA which are likely to be able to be applied to DSA.

From 41bdc78544b8a93a9c6814b8bbbfef966272abbe
Subject: [PATCH] x86/tls: Validate TLS entries to protect espfix

Installing a 16-bit RW data segment into the GDT defeats espfix. AFAICT this will not affect glibc, Wine, or dosemu at all.

Not provide explicit description

Not report to NVD



Problem

Previous Solutions and Limitations



Mining security keywords
 Requiring well-maintained doc.

Regarding code as sequential data
 Losing important semantics.

Our solution: representing code as graph
 Retaining rich patch structural info.

A Graph-Based Security Patch Detection System





- PatchCPG: a new graph representation of inherent code change structures.
 - Syntax and semantics: AST + control & data dependency graph.
 - Changes and relations with context: pre-patch + post-patch graph.
- PatchGNN: a tailored GNN model to capture diverse patch structural information



• Challenge: how to construct PatchCPG?



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[GraphSPD]



• A joint graph encodes rich patch structural information.



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Reducing Noisy Information by Slicing





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- Challenge 1: how to embed the PatchCPGs?
- Challenge 2: how to learn multiple attributes (CDG/DDG/AST/pre/post)?

PatchCPG Embeddings

- Node Embedding
 - 20-dimensional vulnerability features.
 - code snippet metadata
 - identifier and literal features
 - control flow features
 - operator features
 - API features



• Edge Embedding

 \circ 5-dimensional binary vector.



e.g., [1,1,0,1,0] means the edge is a context edge of data dependency.

PatchCPG Embeddings



TABLE I: The involved tokens or sub-tokens of the control flow features, the operator features, and the API features.

Features	Matched Tokens or Sub-tokens			
condition	if, switch			
loop	for, while			
jump	return, break, continue, goto, throw, assert			
arithmetic [†]	++,, +, -, *, /, %, =, +=, -=, *=, /=, %=			
relational	==, !=, >=, <=, >, <			
logical	&&, $ $, $!$, and, or, not			
bitwise*	&, $, <<, >>, \sim, \land$, bitand, bitor, oxr			
memory API	alloc, free, mem, copy, new, open, close, delete, create, release, sizeof, remove, clear, dequene, enquene, detach, attach			
string API	str, string			
lock API	lock, mutex, spin			
system API	init, register, disable, enable, put, get, up, down, inc, dec, add, sub, set, map, stop, start, prepare, suspend, resume, connect,			

[†] Operator * is determined as dereference operator or arithmetic operator.

* Operator & is determined as address-of operator or bitwise operator.

[GraphSPD]







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[GraphSPD]





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Implementation

• **5K new LoC** in Scala and Python on top of *Joern* parser and *PyTorch* library.

Datasets:

- PatchDB: 12K security patches from 300+ GitHub repos.
- SPI-DB: 10K security patches from FFmpeg and QEMU.

Evaluation:

- Compared with sequential-based patch detector.
- Compared with vulnerability detection methods.
- Case study on four popular OSS repos.



(a) Patch Code.



(c) Data Dependency Subgraph.

(b) Control Dependency Subgraph.



(d) Abstract Syntax Tree Subgraph.



. [GraphSPD]

Implementation (Demo)



 https://github.com/liquidware/liquidware_beagleboard_linux/commit/ 176df2457ef6207156ca1a40991c54ca01fef567

~	✓ arch/x86_64/kernel/ptrace.c □ ÷						
. <u>†</u>	@@ −232,10 +232,6 @@ static int putreg(struct task_struct *child,						
232	{	232	{				
233	unsigned long tmp;	233		unsigned long tmp;			
234		234					
235	- /* Some code in the 64bit emulation may not be 64bit						
	clean.						
236	Don't take any chances. */						
237	<pre>- if (test_tsk_thread_flag(child, TIF_IA32))</pre>						
238	<pre>- value &= 0xfffffff;</pre>						
239	<pre>switch (regno) {</pre>	235		<pre>switch (regno) {</pre>			
240	<pre>case offsetof(struct user_regs_struct,fs):</pre>	236		<pre>case offsetof(struct user_regs_struct,fs):</pre>			
241	if (value && (value & 3) != 3)	237		if (value && (value & 3) != 3)			
+							

Compared with Sequential-based Solution



- Accuracy 10.8%↑
- F-1 score: 0.096↑

- Precision: 28.82%↑
- False Positive Rate: 14.62%↓

Method	Dataset	General	Metrics	Special Metrics	
		Accuracy	F1-score	Precision	FP Rate
TwinRNN	PatchDB	69.60%	0.461	48.45%	19.67%
[1][2]	SPI-DB	56.37%	0.512	49.07%	41.57%
GraphSPD	PatchDB	80.39%	0.557	77.27%	5.05%
	SPI-DB	63.04%	0.503	63.96%	19.16%

[1] PatchRNN: A Deep Learning-Based System for Security Patch Identification.

[2] SPI: Automated Identification of Security Patches via Commits.



• **2.5 - 50x** detection rate of vulnerability detectors.

Method	# Vul _{pre-patch}	# Vul _{post-patch}	# Patch _{security}	TP Rate
Cppcheck[3]	3	1	2	0.54%
flawfinder[4]	109	108	1	0.27%
ReDeBug[5]	29	29	0	0.00%
VUDDY[6]	22	16	21	5.71%
VulDeePecker[7]	3	0	3	0.82%
GraphSPD	-	-	53	14.40%

[3] Cppcheck. https://cppcheck.sourceforge.io.

- [4] flawfinder. https://dwheeler.com/flawfinder/.
- [5] Redebug: finding unpatched code clones in entire os distributions.
- [6] VUDDY: A scalable approach for vulnerable code clone discovery.
- [7] VulDeePecker: A deep learning- based system for vulnerability detection.

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```
commit 247d30a7dba6684ccce4508424f35fd58465e535
  if (!s1->current frame.data[0]
3
      ||s->width != s1->width
      ||s->height!= s1->height) {
4
5
      if (s != s1)
6
         copy_fields(s, s1, golden_frame, current_frame);
7
         copy fields(s, s1, golden frame, keyframe);
 +
8
     return -1;
9
```

Listing 4: Security patch for a double free (CVE-2011-3934).

```
1 commit 360e95d45ac4123255a4c796db96337f332160ad
2 if (priv->cac_id_len) {
3    serial->len=MIN(priv->cac_id_len, SC_MAX_SERIALNR);
4 - memcpy(serial->val,priv->cac_id,priv->cac_id_len);
5 + memcpy(serial->val,priv->cac_id,serial->len);
6    SC_RETURN(card->ctx,SC_DEBUG_NORMAL,SC_SUCCESS);
7 }
```

Listing 5: A patch with similar patterns (CVE-2018-16393). CE 815 - Secure Software Systems [GraphSPD]

Case Study on OSS Repos



• **NGINX**: detect 21 security patches (Precision: 78%).

Changes w/	CVE	Total Commits	Valid Commits	Detected S.P.	Confirmed S.P.	Precision
1.19.x 1.17.x 1.15.x 1.13.x	3 3 1 1	180 134 203 270	127 82 120 157	7 4 7 9	6 3 4 8	86% 75% 57% 89%
Sum.	8	787	486	27	21	78%

- Xen: detect 29 security patches (Precision: 55%).
- **OpenSSL**: detect 45 security patches (Precision: 66%).
- ImageMagick: detect 6 security patches (Precision: 46.2%).

Slicing Depth



Slicing	General	Metrics	Special Metrics		
Iteration No. (N)	Accuracy	F1-score	Precision	FP Rate	
0	80.19%	0.555	76.11%	5.42%	
1	80.39%	0.557	77.27%	5.05%	
2	79.24%	0.531	73.61%	5.87%	
∞	76.90%	0.501	64.42%	8.95%	

Conclusion



- Silent security patches can be leveraged by attackers to launch N-day attacks.
- GraphSPD presents patches as graphs and identifies security patches with graph learning, achieving higher accuracy and fewer false alarms.
- GraphSPD can be extended to other programming languages.





- PatchDB only collects security patches from NVD
- SPI-DB contains security patches only from two repositories
- GraphSPD mainly learns from the existing patterns and may not apply to the unseen ones
- Cannot distinguish specific security patch types:
 - Some patches are too rare to be included in training set
 - Data is imbalanced in security patch datasets. According to the analysis on NVD, 24.6% of vulnerabilities are related to code execution, whereas only 0.1% of vulnerabilities are HTTP response splitting
- Single function support!

Acknowledgments



• [GraphSPD] GraphSPD: Graph-Based Security Patch Detection with Enriched Code Semantics, S. Wang, X. Wang, K. Sun, S. Jajodia, H. Wang, Q. Li, IEEE S&P 2023.