# CE 815 – Secure Software Systems

ML-Based Vulnerability Detection Methods (Vulchecker)

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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.

#### Prior Works Limitations



- Detects vulnerability at function level
- Can't find vulnerability type

#### VulChecker



- Precisely locate vulnerabilities in source code (down to the exact instruction)
- Classify vulnerabilities type
- Low-cost dataset augmentation

### Insights



- Broad Program Slicing
  - Location of the vulnerability instead of a region or function
- Incomplete Code Representations
  - enhanced-PDG
- Manifestation distance
  - Manifestation vs root cause
- The Lack of Labeled Data
  - datasets that only label code regions or functions
- Level of program representation
  - Source code or machine instructions

```
1 short concat(char *a, char *b, char **out) {
    short al = strlen(a);
    short bl = strlen(b);
    *out = (char *) malloc(al+bl);
    if (al)
      memcpy(*out, a, al);
    if (bl)
      memcpy(*out+al,b,bl);
    return al + bl;
  AST →
  NCS -
  CFG -
  DFG -
               char *a
   short
                           char *b
 ReturnType
              Parameter
                          Parameter
   char *
                               char *
ParameterType
                Identifier
                                           Identifier
                           ParameterType
```



### Prior Works



			(1) Code Representation						(2) Sample Selection					(3) Feature Extraction							(4) Model indu		(5) Application								
								Code Slicing																		be,					
			Le	evel Structure					Plane			PoI		C	Cut		Node		Edge	Input		Model	Utilizes Edge Type	De	etectio	n lev	el	Type?			
Year	Cite	Name	Source Code	IR	Linear	CFG	PDG	CPG	ncsCPG	ePDG	Function	Control-flow	Data-flow	Generic	Manifestation	Region	Scoped	One-hot Enc.	Word2Vec	Doc2Vec	Explicit features	Dtype feature	Sequence	Graph			Function	Code Region	Line	Instruction	Classifies Vuln.
2018	[28]	Russle'18				<u> </u>	_	_			•		$\stackrel{\frown}{\rightarrow}$	_	_		-			_					CNN,RF				=		
2018	[23]	Vuldeepecker	•		•						•					•							•		BiLSTM		•				•
2018	[40]	μVulDeePecker	•		•											•							•		BiLSTM						•
2019	[39]	•			•				_				•	•		•		•					•		GCN,DNN			•			•
		Devign	•			_			•		•					•			•					•	· ·	•	•	_			
2019 2019	[14] [31]	VGDetector NW-LCS	•			•					•					•				•				•	GCN,DNN LCS Scores			•			
2019	[19]	Li'20	•			•					•					•		•						•	CNN		•				•
2020	[38]		•		•							•		•		•			•				•		DNN			•			•
2020	[32]	Zagane'20 Funded	•		•				_			•		•		•		0	_				•	_	GNN,GRU			•			
2020	[30]	AI4VA						_	•							•			•					•	GNN,GRU						•
2020	[22]	SySeVR						•			•					•		١.	•						BiRNN	•	•				
2021	[20]	Li'21	_	•												•									CNN+RNN,DNN						
2021	[21]	Vuldeelocator												•	•										BiRNN			•			
2021	[13]	DeepWukong	•															•						•	GCN,DNN						•
2021	[35]	Wu'21										•		•						•					GNN,DNN			•			
2021	[ <del>9</del> ]	BGNN4VD					•									•									GNN,GRU						
2021	[11]	Reveal						•								•			•						GCN,DNN						
	[**]	VulChecker	_	•						•		•	$\dashv$		•		•	0			•	•		•	GN (S2V)	•			•	•	•

#### VulChecker



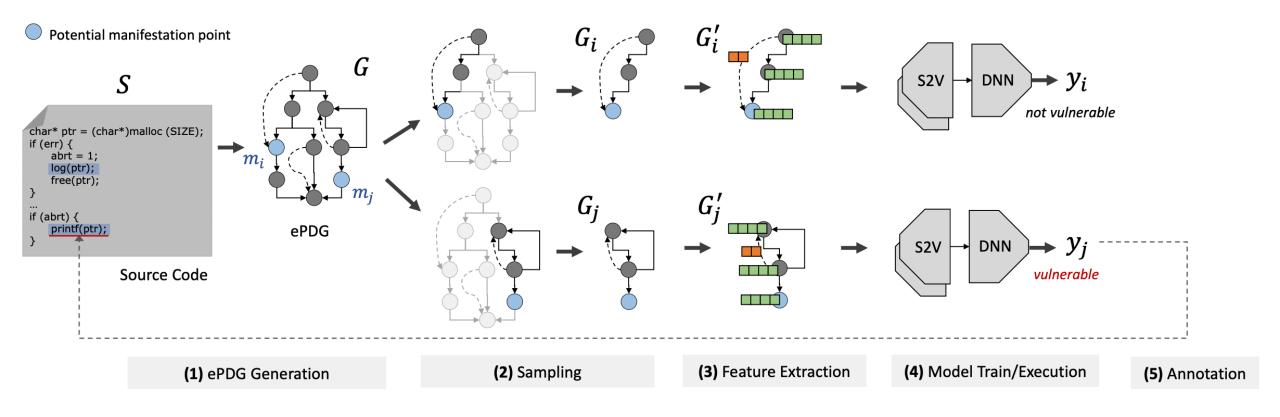


Figure 2: A diagram showing the steps of VulChecker's pipeline for one CWE. Note that the real graphs are significantly larger than what is visualized (e.g., projects like libgit2-v0.26.1 have over 18 million nodes in G). Solid edges represent control-flow and dashed edges are data dependencies.

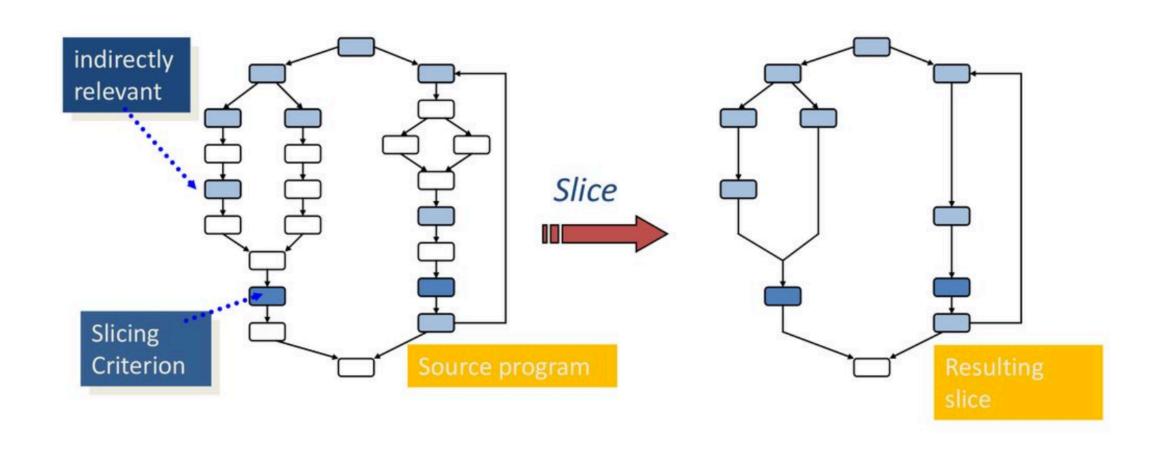
#### ePDG



 ePDGs are graph structures in which nodes represent atomic machine-level instructions and edges represent control- and data-flow dependencies between instructions

# Program Slicing





# Program Slicing (cont.)



```
public class SimpleExample {
    static int add(int a, int b){
        return(a+b);
    }
    public static void main(final String[] arg){
        int i = 1;
        int sum = 0;
        while (i < 11) {
            sum = add(sum, i);
            i = add(i, 1);
        }
        System.out.println("sum = " + sum);
        System.out.println("i = " + i);
    }
}</pre>
```

```
public class SimpleExample {
    static int add(int a, int b){
        return(a+b);
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    public static void main(final String[] arg){
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#### VulChecker



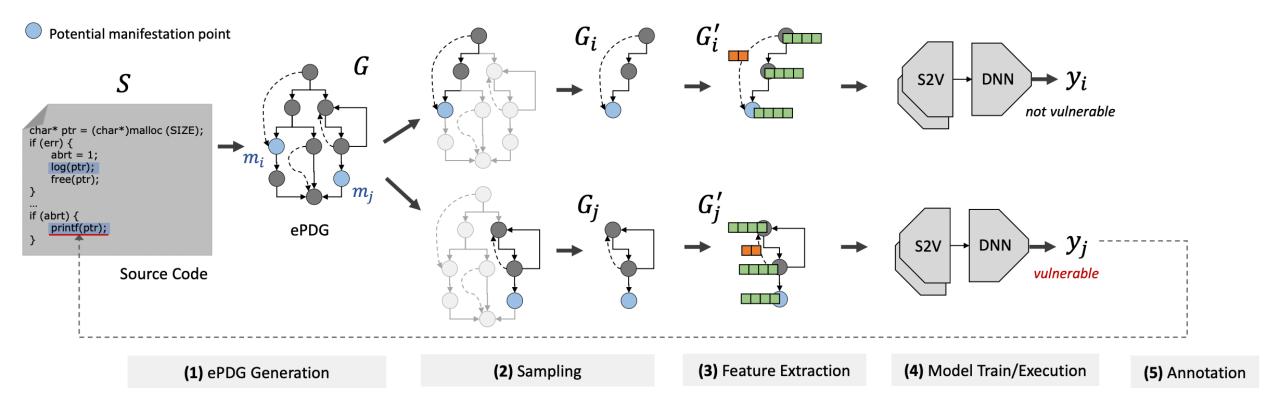


Figure 2: A diagram showing the steps of VulChecker's pipeline for one CWE. Note that the real graphs are significantly larger than what is visualized (e.g., projects like libgit2-v0.26.1 have over 18 million nodes in G). Solid edges represent control-flow and dashed edges are data dependencies.

#### ePDG Generation



- Lowering the source code S to LLVM IR
- Extracting G based on the structure and flows it contains

### Lowering Code to LLVM IR



- Simplifies the program representation:
  - Control-flow: complicated branching constructs in source code are reduced to conditional jumps that test a single condition
  - Data-flow: definition-use chains are shorter and less complex as they are based on virtual register values rather than source code variables
- During lowering, VulChecker instructs Clang to embed debug information in the IR, which enables traceability of IR instructions back to source code instructions

### Lowering Code to LLVM IR (cont.)



- Using semantic-preserving compiler optimizations provided by LLVM to simplify and better express the code in G:
  - Function inlining to replace function call sites in the IR with a concrete copy of the called function body
  - Indirect branch expansion to eliminate indirect branching constructs
  - Dead code elimination to reduce the size of the output graph

# Sampling



- Pol Criteria
- Program Slicing
  - Crawls G backwards from  $m_i$  using breadth first search (BFS)
- Labeling

#### Feature Extraction



- Operational Node Features
- Structural Node Features
  - Distance from the nearest potential root cause
  - Betweenness centrality measure (BEC)
- Semantic Node Features
- Edge Features

Table 2: Summary of Features used in  $G'_i$ 

	Name	Type	Count
		Bool Num. Categ.	
	Has static value?	•	1
	Static value	•	1
	Operation {+, *, %,}	•	54
	Basic function {malloc, read,}	•	1228
	Part of IF clause	•	1
<b>.</b> ⊿	Number of data dependents	•	1
Vertex	Number of control dependents	•	1
Ver	Betweeness centrality measure	•	1
,	Distance to $m_i$	•	1
	Distance to nearest r	•	1
	Operation of nearest <i>r</i>	•	54
	Output dtype {int, float,}	•	6
	Node tag $\{r, m, none\}$	•	2
		Total	1352
<u>و</u>	Output dtype {float, pointer}		6
Edge	Edge type {CFG, DFG}		2
		Total	8

### Embedding



- Some embeddings include one hot encodings and pre-processed embeddings (e.g., Word2Vec)
- In some cases entire portions of code are summarized using Doc2Vec
- The issue with these representations:
  - nodes in  $G_i$  would likely capture multiple operations in a single line of source code resulting in a loss in semantic precision
  - the use of pre-processed embeddings prevents the model from learning the best representation to optimize the learning objective
  - Hence use S2Vec (structure to vector)

### Data Augmentation



- Data augmentation is a technique for creating new training examples from existing ones. VulChecker augments its training dataset by adding synthetic vulnerabilities to "clean" projects.
- Validity: Since augmentation process splices multiple ePDGs, it may produce samples where a vulnerability ePDG subgraph lies on an infeasible path in the augmented ePDG
  - Ignore such samples

### Data Augmentation (cont.)



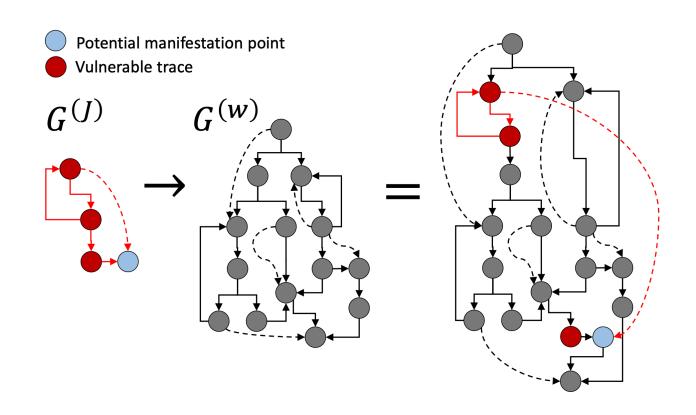
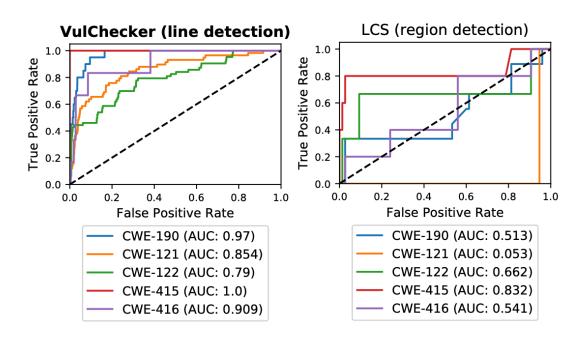
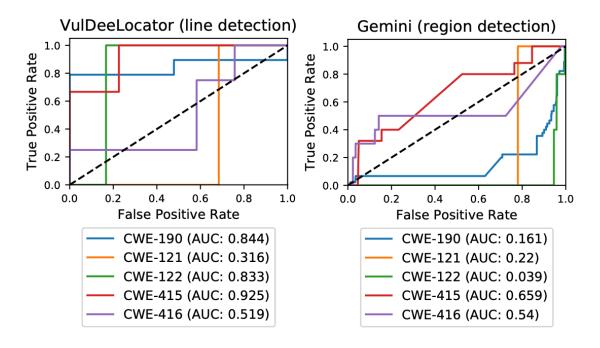


Figure 3: An illustration of an ePDG from the wild  $G^{(w)}$  being augmented with a synthetic vulnerability trace from Juliet  $G_i^{(J)}$ .

#### Evaluation







### Evaluation (cont.)



Table 3: Baseline comparison against a commercial SAST tool in detecting CVEs in the wild.

	VulChecker @ FPR 0.05				Chec	ker @ FPR 0.1	Helix QAC				
	Lines		CVEs	Lin		CVEs	Li	nes	CVEs		
<b>CWE</b>	TP	FP	TP	TP	FP	TP	TP	FP	TP		
190	9	55	3	12	112	6	1	2	1		
121	7	33	7	9	112	9	4	230	1		
122	1	6	1	1	6	1	4	241	1		
415	3	0	2	3	0	2	0	5	0		
416	4	6	4	6	228	6	0	0	1		
Total	24	100	17	31	458	24	9	478	4		

### Evaluation (cont.)



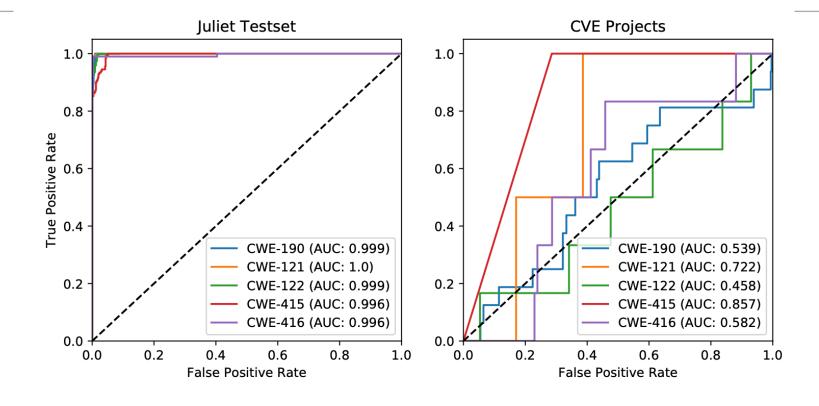


Figure 6: Performance of VulChecker when trained on synthetic data, then either tested on synthetic (left) or tested on real data (right).

#### Conclusion



- VulChecker precisely locates vulnerabilities in source code down to the exact instruction.
- Classifies vulnerabilities according to the Common Vulnerabilities and Exposures (CVE) taxonomy.
- Employs a novel data augmentation technique to enrich the training dataset and enhance generalization ability.
- Achieves near-zero false positives in vulnerability detection, outperforming commercial tools.
- VulChecker successfully detects a previously unknown zero-day vulnerability, highlighting its ability to identify novel vulnerabilities.

# Acknowledgments



- [VulChecker] VulChecker: Graph-based Vulnerability Localization in Source Code, Y. Mirsky, G. Macon, M. Brown, C. Yagemann, M. Pruett, E. Downing, S. Mertoguno, and W. Lee, Usenix Security 2023.
- [Alves] Program Slicing. SwE 455, Alves, E., Federal University of Pernambuco, 2015.