

Vestige: Identifying Binary Code Provenance for Vulnerability Detection

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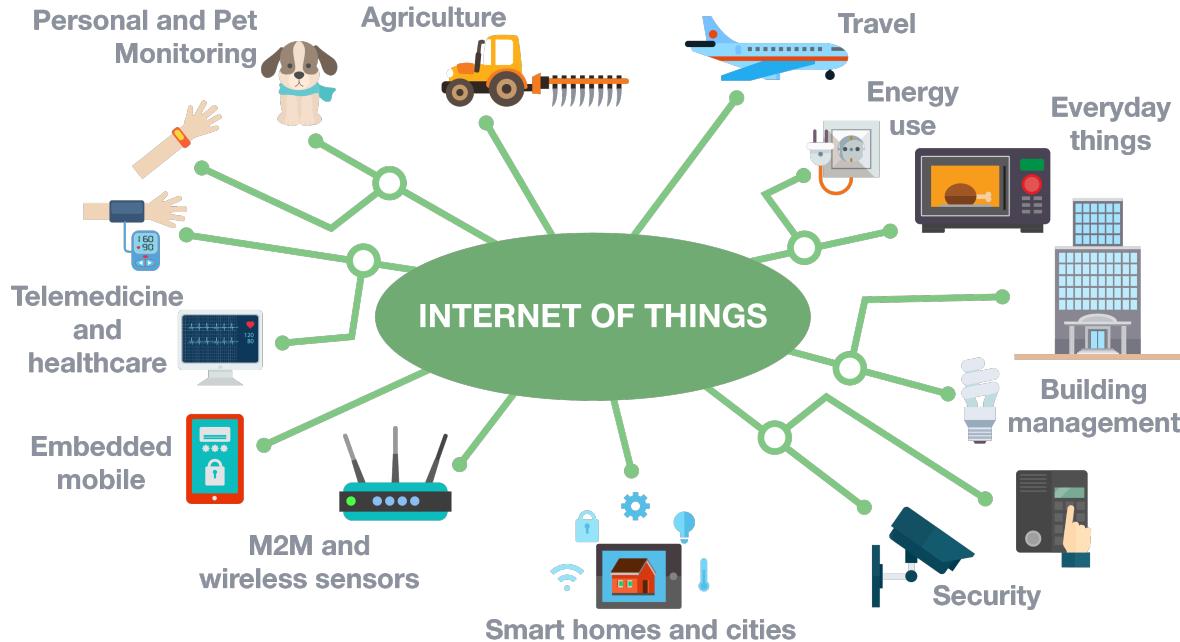
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June 24th, 2021

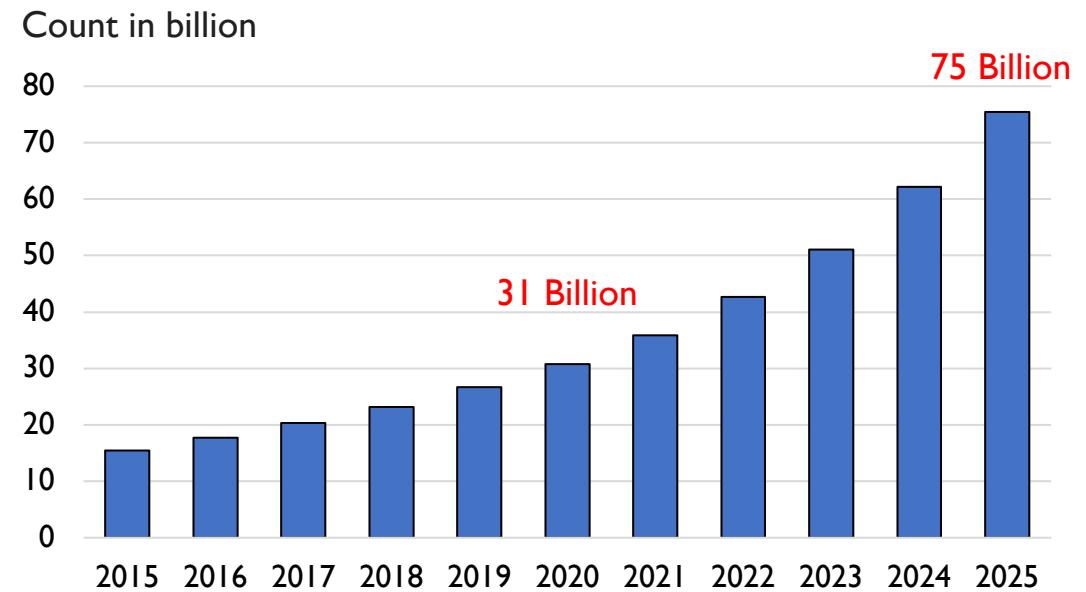


Binary Code is Prevalent

- Software vendors usually do not share the source code.
- A significant number of binaries are running in the wild, e.g., firmware of IoT devices.

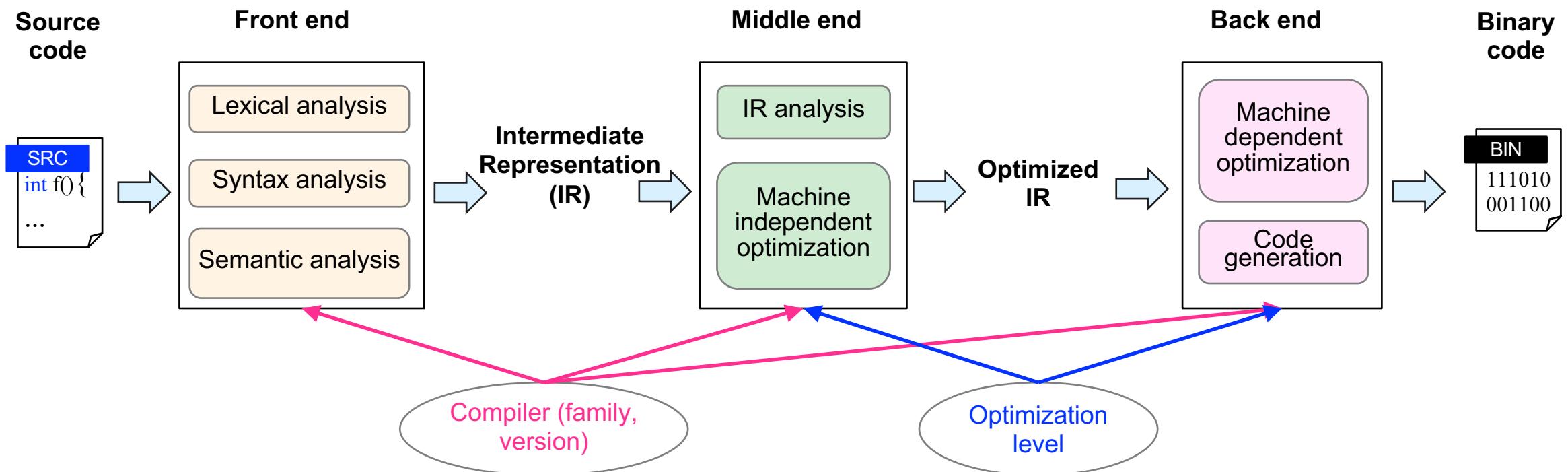


Num. of Connected IoT Devices from 2015 to 2025



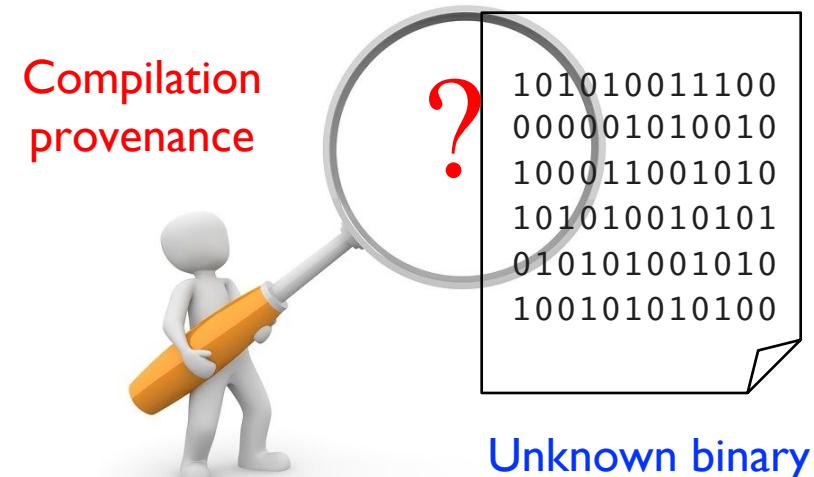
Binary Code Compilation Provenance

- Compilation *provenance*: the *compiler* and *compilation configurations*.
- Used in malware analysis, code vulnerability detection, and code authorship identification.



Research Problem and Existing Solutions

- Research Problem
 - Identify the compilation provenance of an unknown binary.
 - The provenance is regarded as a 3-tuple, <*compiler family*, *version*, *optimization level*>.



- Existing Solutions [Rosenblum et al. ISSTA'11], [Massarelli et al. BAR'19], [Rahimian et al. DFRWS'15]
 - Convert the provenance identification problem to a classification problem.
 - Extract features (patterns) from binary code.
 - Instructions features, control flow graphs.
 - Build a machine learning-based prediction model.

Motivation

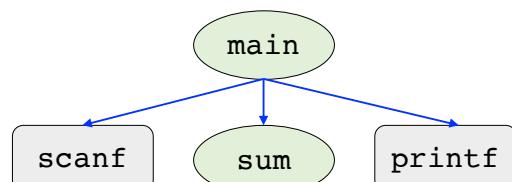
```
1 #include <stdio.h>
2
3 int sum(int a, int b) {
4     return a + b;
5 }
6
7 int main() {
8     int a, b;
9     scanf("%d%d", &a, &b);
10    int c = sum(a, b);
11    printf("c = %d\n", c);
12    return 0;
13 }
```

(a) An example source code
with inter overflow

- **GCC-4.8.4-O0 v.s. GCC-4.8.4-O2**
 - *Control flow graphs* are same
 - *Instructions* have minor differences
 - *Function call graphs* are obviously different

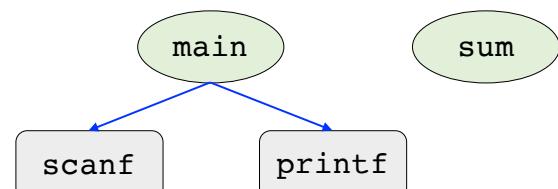
```
1 <sum:>
2 push    ebp
3 mov     ebp, esp
4 mov     eax, dword ptr [ebp+0Ch]
5 mov     edx, dword ptr [ebp+8]
6 add     eax, edx
7 pop     ebp
8 ret
9
10 <main:>
11 push    ebp
12 mov     ebp, esp
13 and    esp, 0FFFFFFF0h
14 sub    esp, 20h
15 lea     eax, [esp+18h]
16 mov     [esp+8], eax
17 lea     eax, [esp+14h]
18 mov     [esp+4], eax
19 mov     dword ptr [esp], offset aDD
20 call    __isoc99_scanf
21 mov     edx, [esp+18h]
22 mov     eax, [esp+14h]
23 mov     [esp+4], edx
24 mov     [esp], eax
25 call    sum
26 mov     [esp+1Ch], eax
27 mov     eax, [esp+1Ch]
28 mov     [esp+4], eax
29 mov     dword ptr [esp], offset format
30 call    _printf
31 mov     eax, 0
32 leave
33 ret
```

(b) Assembly code of GCC-4.8.4-O0



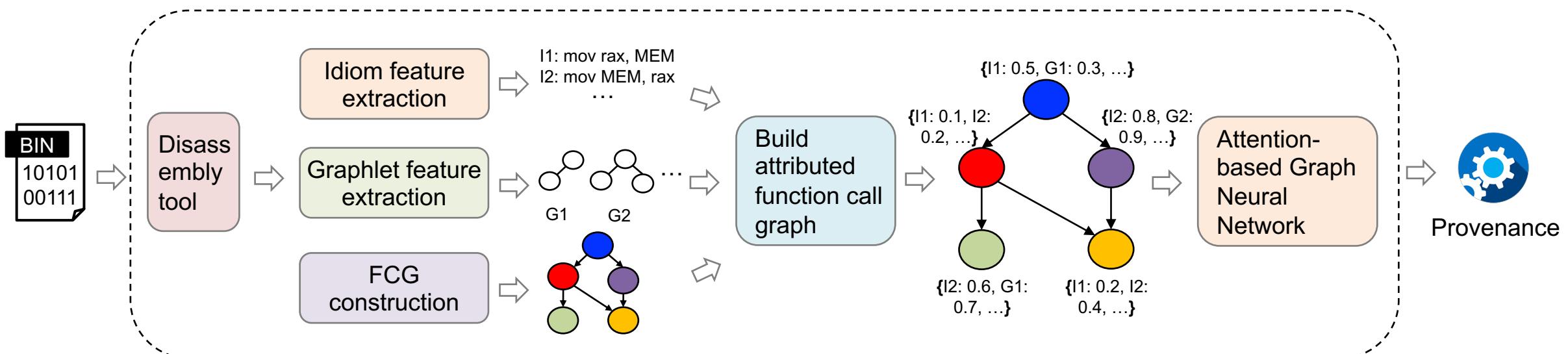
```
1 <sum:>
2 mov     eax, dword ptr [esp+4]
3 add     eax, dword ptr [esp+8]
4 ret
5
6 <main:>
7 push    ebp
8 mov     ebp, esp
9 and    esp, 0FFFFFFF0h
10 sub   esp, 20h
11 lea    eax, [esp+1Ch]
12 mov   [esp+8], eax
13 lea    eax, [esp+18h]
14 mov   [esp+4], eax
15 mov   dword ptr [esp], offset aDD
16 call  __isoc99_scanf
17 mov   eax, [esp+1Ch]
18 add   eax, [esp+18h]
19 mov   dword ptr [esp+4], offset aCD
20 mov   dword ptr [esp], 1
21 mov   [esp+8], eax
22 call  __printf_chk
23 xor   eax, eax
24 leave
25 ret
```

(c) Assembly code of GCC-4.8.4-O2



Overview of Vestige

- Vestige: a graph neural network-based binary code provenance identification method.
 - Represent a binary code as *attributed function call graph (ACFG)*.
 - Use *graph attention network* to learn a provenance prediction model.



Attributed Function Call Graph

```
1 ...  
2 do {  
3     tbio = BIO_pop(f);  
4     BIO_free(f);  
5     f = tbio;  
6 } while (f != upto)  
7 ...
```

Source code fragment of CVE-2015-1792

```
loc_816C6F2:  
mov    [esp+1Ch+var_1C],  
      ebx  
call   BIO_pop  
mov    [esp+1Ch+var_1C],  
      ebx
```

loc_816C6F0:
...

Provenance 1: *GCC-4.8.4-02*

```
loc_824A343:  
mov    eax, [esp+1Ch+arg_0]  
mov    [esp+1Ch+var_1C],  
      eax  
call   BIO_pop  
mov    [esp+1Ch+var_1C],  
      eax
```

...

Provenance 2: *LLVM-5.0-00*

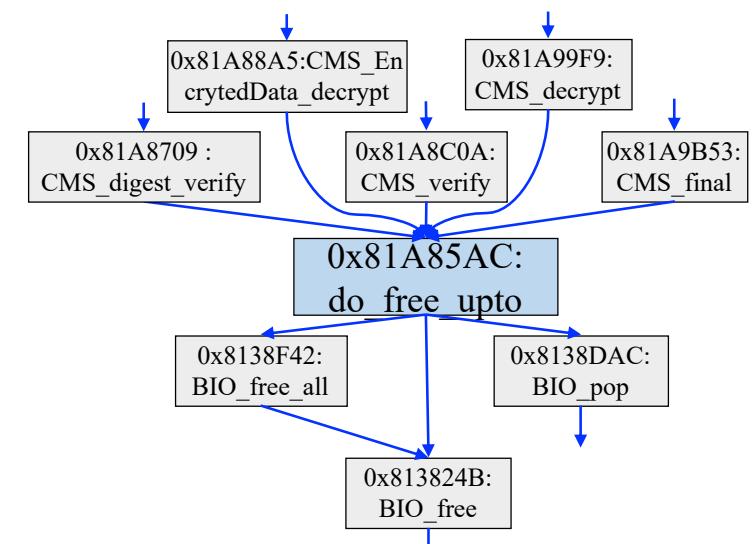
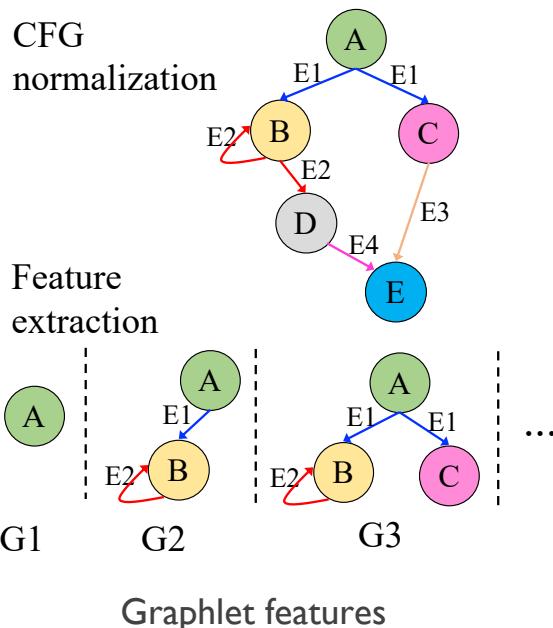
Instruction normalization

```
mov    rax, MEM  
mov    MEM, rax  
call   rip  
mov    MEM, rax  
...  
...
```

Feature extraction

```
I1: mov rax, MEM  
I2: mov rax, MEM | mov MEM, rax  
I3: mov rax, MEM | * | mov MEM, rax  
...  
...
```

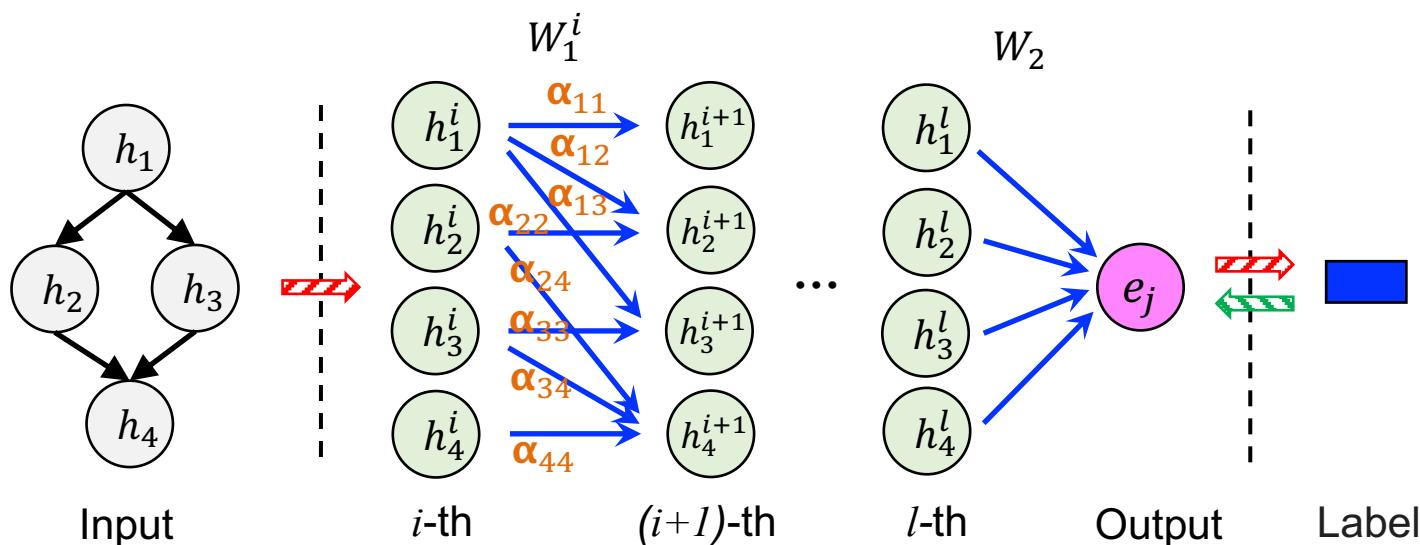
Idiom features



Attributed Graph Embedding

- Graph Embedding
 - Graph attention network (GAT) [Velicković et al. ICLR'18]

- Objective function:
$$h_v^{i+1} = \sigma \left(\alpha_{vv} h_v^i + \sum_{u \in N(v)} \alpha_{vu} W_1^i h_u^i \right)$$



Experiment

- Dataset

Dataset	Software	Compiler	Optimization	# Binaries
I (Baseline)	Bash-4.3, Diffutils-3.3, Grep-2.16, Tar-1.27.1, Wget-1.15			336
II (Code similarity dataset)	SNNS-4.2, PostgreSQL- 7.2, Binutils-{2.25, 2.30}, Coreutils-{8.21, 8.29}	GCC-{4.6.4, 4.8.4, 5.4.1}, LLVM-{3.3, 3.5, 5.0}	O0 – O3	6,168
III (Vulnerability dataset)	OpenSSL-{0.9.7f, 1.0.1f, 1.0.1n}			24

- Evaluation Metrics
 - Accuracy
 - Top-k hit rate

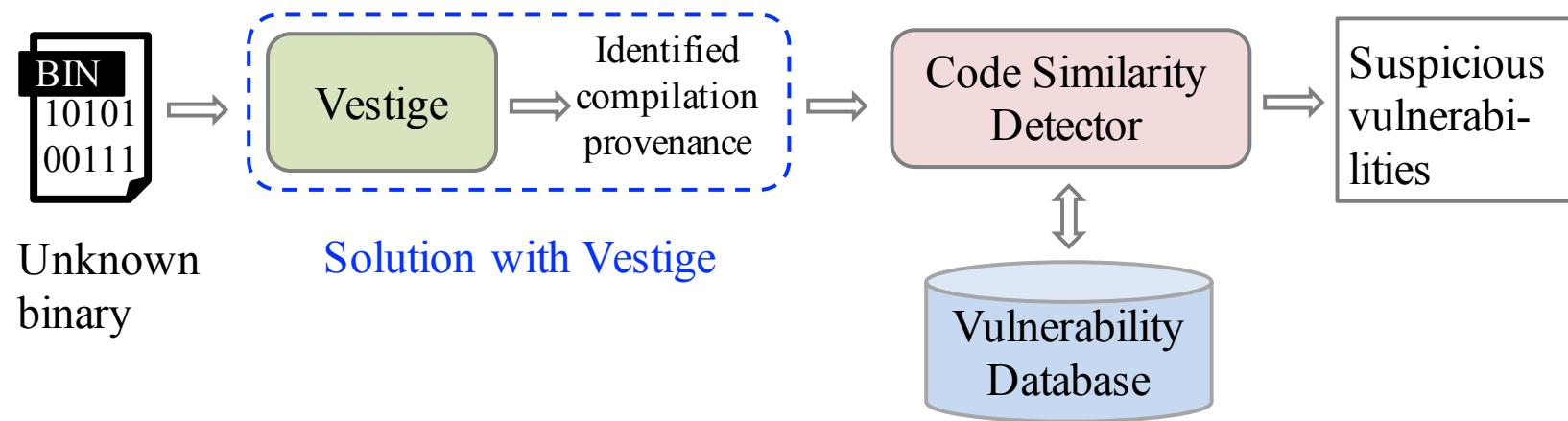
Accuracy of Provenance Identification

- Setting
 - 6,504 binaries from dataset I and II.
 - 10-fold cross validation.
- Compared Works
 - Origin [Rosenblum et al. ISSTA'11]
 - Vestige-S2V: a different graph neural network, structure2vec, on AFCG.

	Origin	Vestige-S2V	Vestige-GAT
Optimization level	92.2%	98.7%	99%
Compiler version	96.8%	95.5%	97.9%
Compiler family	99.5%	99.5%	99.5%
Overall	90.2%	93.3%	96.1%

Case Study #1: Code Similarity Detection

- Two-step Solution with Vestige
 - First step: identify compilation provenance of the unknown binary code.
 - Second step: compare the unknown binary code with vulnerable code sharing the same compilation provenance.
- Code Similarity Detector
 - BGM [*Bi*partite *G*raph *M*atching]
 - Genius [*Qian et al. CCS'16*]
 - Gemini [*Xu et al. CCS'17*]



Case Study #1: Code Similarity Detection

- Dataset II
 - 6K+ binaries
 - 24 compilation provenances
- Training and testing use different software
- Search 1,000 functions

	Software	Compilation Provenance	# Binaries
Train	SNNS-4.2, PostgreSQL-7.2	Compiler: GCC-{4.6.4, 4.8.4, 5.4.1}, LLVM-{3.3, 3.5, 5.0} Optimization: O0 – O3	600 5,568
Test	Binutils-{2.25, 2.30}, Coreutils-{8.21, 8.29}		
Total	-	24	6,168

	Top-1		Top-5	
	Original	Original + Vestige	Original	Original + Vestige
BGM	29%	56% (+27%)	45%	89% (+44%)
Genius	51%	64% (+13%)	69%	91% (+22%)
Gemini	66%	87% (+21%)	77%	94% (+17%)

Case Study #2: Vulnerability Detection

- OpenSSL dataset (III)
 - 20 vulnerable functions
 - Query against 24 variants
- Top-1 hit rate
- Result
 - BGM: 33% → 49% (16%)
 - Genius: 39% → 58% (19%)
 - Gemini: 50% → 76% (26%)

CVE	Query	BGM:+Vestige	Genius:+Vestige	Gemini:+Vestige
2015-0209	1.0.lf	46:67	50:71	50:88
2014-0195	1.0.lf	33:42	42:58	54:92
2016-2106	1.0.lf	58:63	58:63	63:83
2012-0027	1.0.lf	42:58	58:67	63:88
2014-3513	1.0.lf	46:71	50:83	67:92
2015-1791	1.0.lf	50:67	50:83	71:96
2015-3196	1.0.lf	42:67	38:75	58:92
2014-3567	1.0.lf	22:56	33:67	50:79
2016-0797	1.0.ln	21:41	25:41	38:83
2016-2180	1.0.ln	25:42	29:83	46:95
2016-2105	0.9.7f	58:67	47:58	58:83
2016-2176	1.0.ln	38:42	38:46	50:67
2016-2109	0.9.7f	10:29	30:38	40:54
2015-3195	0.9.7f	25:42	50:63	58:83
2016-2182	0.9.7f	25:42	33:50	46:58
2016-2178	0.9.7f	13:25	19:42	25:63
2015-0292	0.9.7f	37:42	46:54	50:67
2016-2105	0.9.7f	58:63	58:63	67:71
2016-2842	1.0.ln	5:25	10:33	19:50
2016-0705	1.0.ln	13:21	17:21	19:42
Average	-	33:49	39:58	50:76

Summary

- Key Takeaway:
 - *Identifying compilation provenance* can effectively improve code similarity detection.
 - The *attributed function call graph* (AFCG) is an efficient representation for binary code analysis.
- Impact:
 - Improve *top-1* hit rate of recent code vulnerability detection methods by up to **26%**.

Thank You

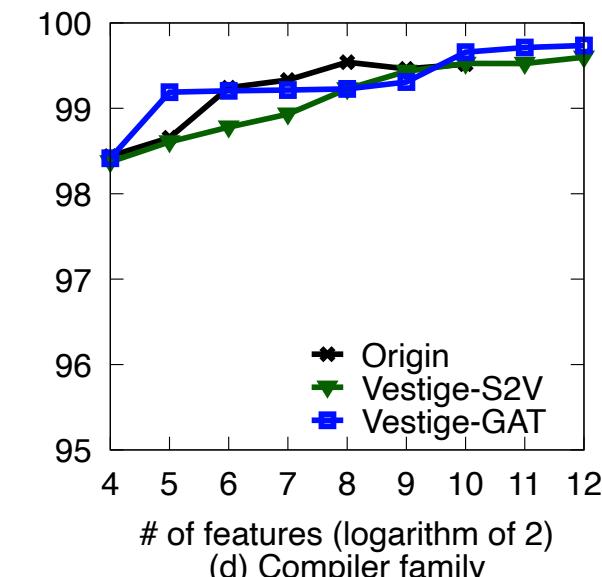
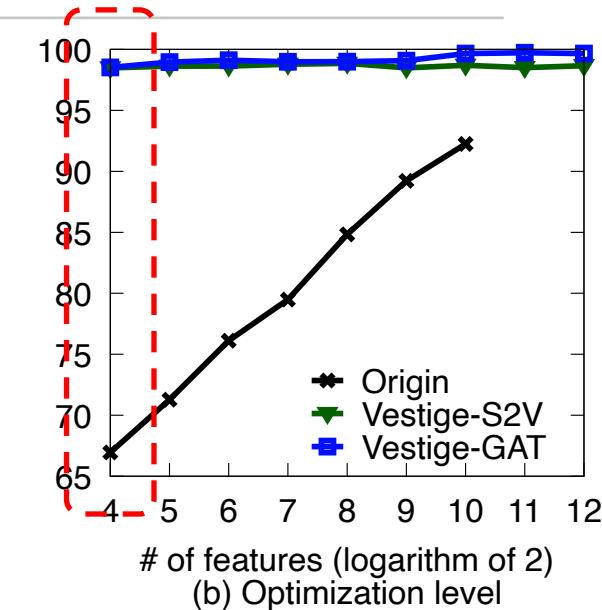
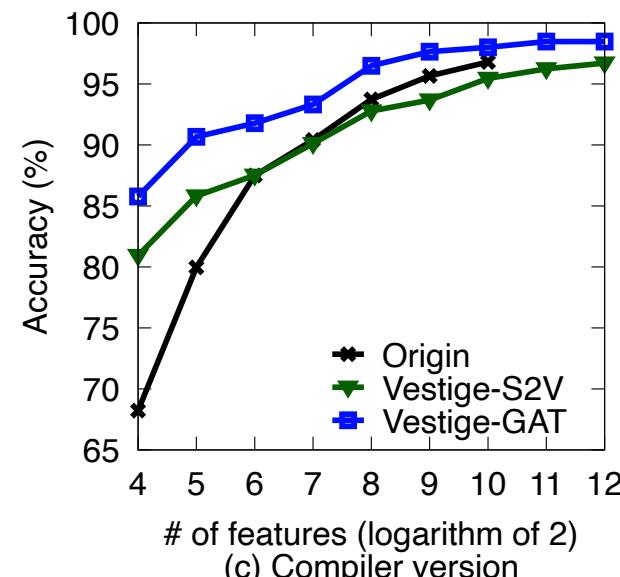
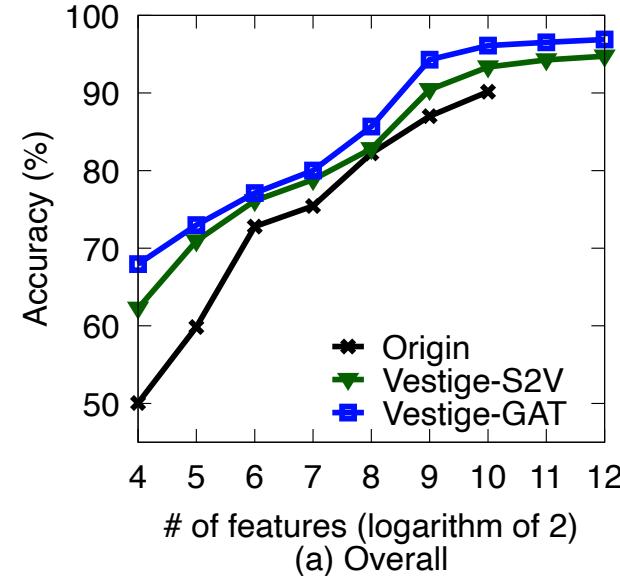
Contact us: yuedeji@gwu.edu, leicui@gwu.edu, howie@gwu.edu



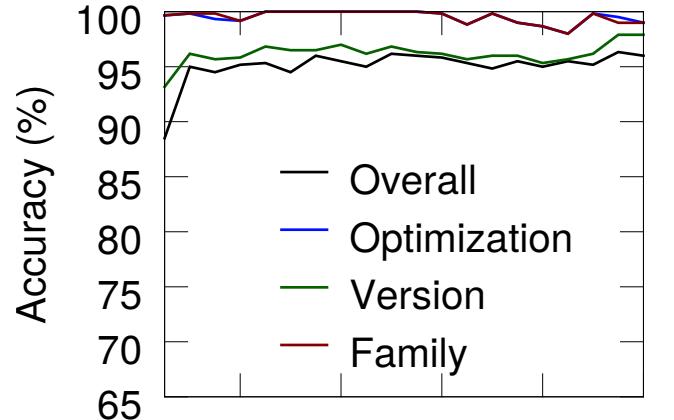
Backup Slides

Accuracy of Provenance Identification

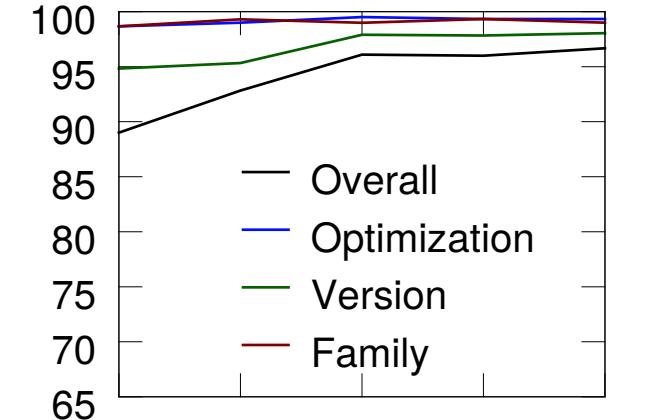
- Accuracies against number of features
 - Function call graph is efficient for optimization level



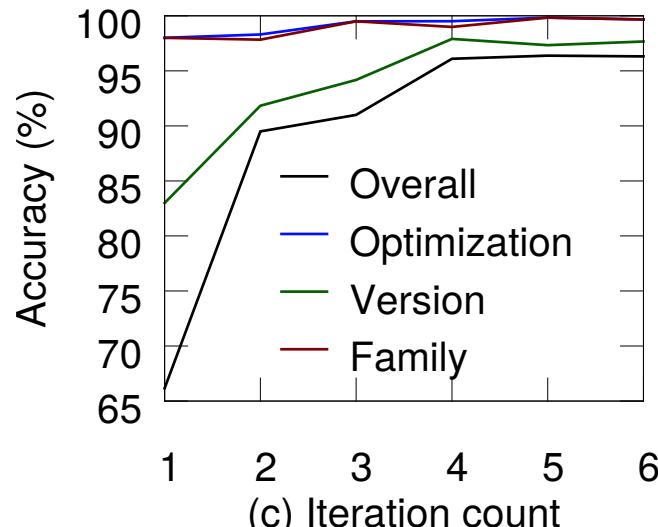
Sensitivity Study



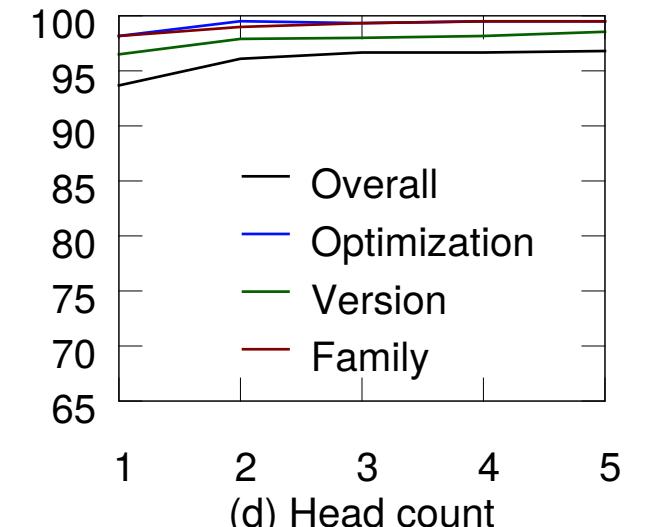
(a) Epoch number



(b) Embedding size



(c) Iteration count



(d) Head count