SymLM: Predicting Function Names in Stripped Binaries via Context-Sensitive Execution-Aware Code Embeddings

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Predicting binary function names is extremely useful

► Function names: summary of <u>function behavior and semantics</u>.

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```
void FUN_001092f3(byte *param_1) {
   byte *local_48;
   ulong local_40;
   ...
   if (*local_48 == 10) {
        local_48 = local_48 + 1;
   }
   else if (*local_48 == 0x3a) {
        DAT_00119470 = '\0';
        bVar3 = *local_48;
   }
   ...
}
```

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   }
}
```

```
Function Name
Prediction
```

```
void DNS_flood(byte *param_1) {
    byte *local_48;
    ulong local_40;
    ...
    if (*local_48 == 10) {
        local_48 = local_48 + 1;
    }
    else if (*local_48 == 0x3a) {
        DAT_00119470 = '\0';
        bVar3 = *local_48;
    }
}
```

Challenges

Missing Semantics. Very limited semantic information.

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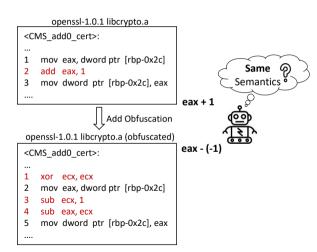
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Missing Semantics

Names of **identifiers** and **function parameters** use the same words as function names [LWN21].

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- Binary Variation. Semantically similar code appearing differently.
- Noisy Function Names. Different developers naming functions differently.

Reasons

- Synonyms and abbreviations are ubiquitous in function names.
- Even single letters can be meaningful when probably used [BGOF17].
- Probability (two developers select same names for the same function) = 6.9% [FMN+20].

- Missing Semantics. Very limited semantic information.
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- **4 OOV Issues**. Out-of-vocabulary words widely used.

Challenges

- Missing Semantics. Very limited semantic information.
- Binary Variation. Semantically similar code appearing differently.
- Noisy Function Names. Different developers naming functions differently.
- OOV Issues. Out-of-vocabulary words widely used.

OOV Words

	Category	Ratio	Examples
1	Abbreviation concatenation	29.9%	statinfo, streq
2	Clean word concatenation	22.3%	sharefile, startpoints
3	Misspelling	14.6%	anewer, tac, sb
4	Clean word	12.1%	dependent, specifer
5	Abbreviation	7.0%	utils, pred
6	Inflection	9.6%	addresses, using
7	Digits in word	4.5%	add32, merge2

- Missing Semantics. Very limited semantic information.
- Binary Variation. Semantically similar code appearing differently.
- Noisy Function Names. Different developers naming functions differently.
- OOV Issues. Out-of-vocabulary words widely used.
- Comprehensive semantic modeling. Semantics preserved in calling context.

- Missing Semantics. Very limited semantic information.
- Binary Variation. Semantically similar code appearing differently.
- **3** Noisy Function Names. Different developers naming functions differently.
- OOV Issues. Out-of-vocabulary words widely used.
- **6** Comprehensive semantic modeling. Semantics preserved in calling context.

Prior Works and Limitations

- ► Limited features, e.g.,
 - handcrafted features: Debin [HIT+18] and Punstrip [PECK20].
 - **2** partial function semantics: NERO [DAY20] and NFRE [GCXZ21].

Prior Works and Limitations

- ► Limited features, e.g.,
 - handcrafted features: Debin [HIT+18] and Punstrip [PECK20].
 - partial function semantics: NERO [DAY20] and NFRE [GCXZ21].

- 1 xor ecx, ecx
- 2 mov eax, dword ptr [rbp-0x2c]
- 3 sub eax, ecx
- 4 mov dword ptr [rbp-0x2c], eax



1 INST_1534 2 INST_5741 3 INST_7745 4 INST_2573

Preprocessing Step of NFRE

Predicting function names

▶ Program semantics is manifested in execution behavior.

openssl-1.0.1 libcrypto.a

```
<CMS_add0_cert>:
...

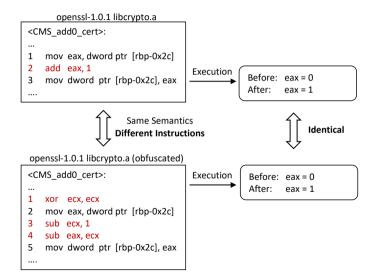
1 mov eax, dword ptr [rbp-0x2c]
2 add eax, 1
3 mov dword ptr [rbp-0x2c], eax
....
```



openssl-1.0.1 libcrypto.a (obfuscated)

```
<CMS_add0_cert>:
...

1    xor    ecx, ecx
2    mov   eax, dword ptr [rbp-0x2c]
3    sub   ecx, 1
4    sub   eax, ecx
5    mov   dword ptr [rbp-0x2c], eax
....
```



Predicting function names

- ▶ Program semantics is manifested in execution behavior.
- ► Learning semantics requires understanding both function instructions and calling context.

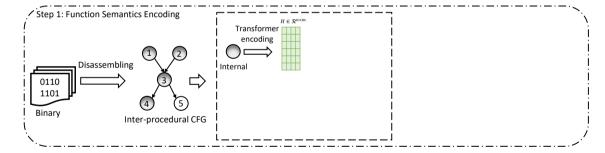
Predicting function names

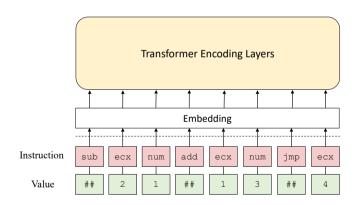
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- ► Learning semantics requires understanding both function instructions and calling context.
- Function name semantic similarity needs to be measured.

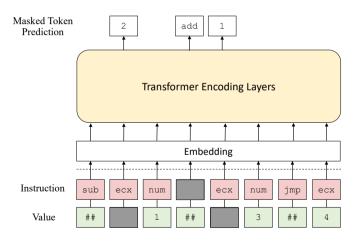
Predicting function names

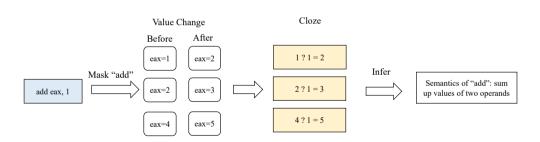
- ▶ Program semantics is manifested in execution behavior. ⇒ C1 and C2
- ► Learning semantics requires understanding both function instructions and calling context. ⇒ C5
- ► Function name semantic similarity needs to be measured. ⇒ C3 and C4

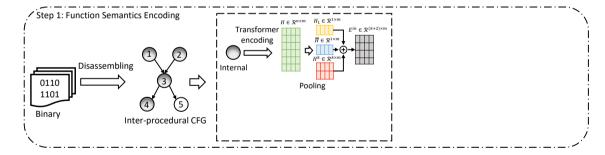


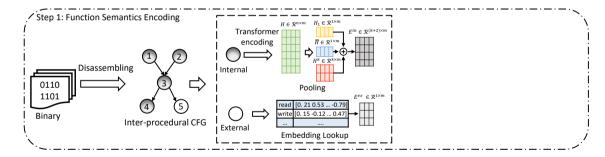


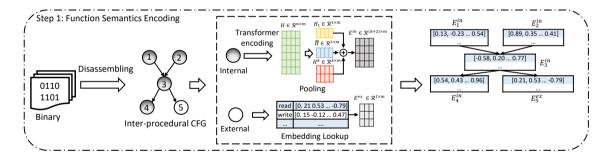


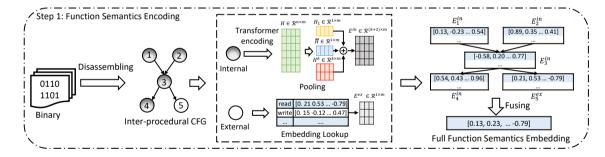


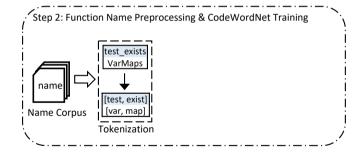


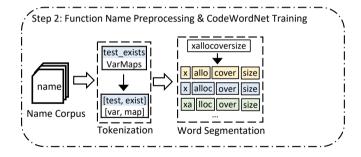


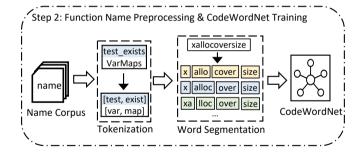


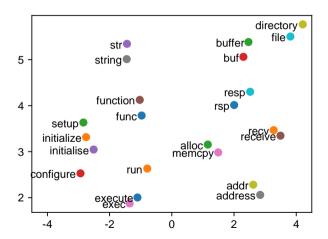


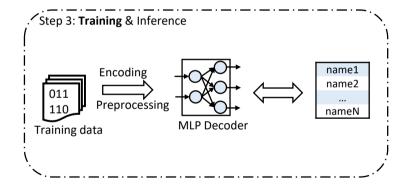




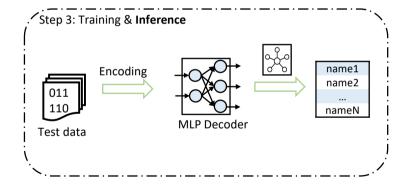








System Workflow: Step 3



Dataset: 1,431,169 functions

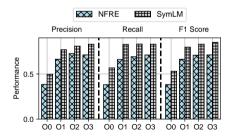
- **Dataset**: 1,431,169 functions
 - ► 27 open-source projects.
 - ► 4 architectures: x86, x64, ARM, and MIPS.
 - ▶ 4 optimization levels: O0, O1, O2, and O3 (GCC-7.5).
 - ▶ 4 obfuscation options: bcf, cff, sub, and split (LLVM obfuscator).

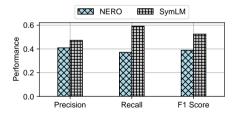
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- Metrics: precision, recall, and F1-score.

Baseline Comparison

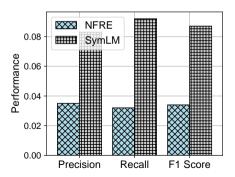
➤ SYMLM outperforms the state-of-the-art works (up to 35% improvement on F1 score).

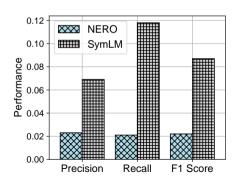




Generalizability

➤ SYMLM is more generalizable to unseen binary functions (295.5% better F1 score).





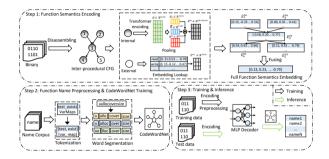
Use Case

► SYMLM successfully infers function semantics of IoT firmware image [Gat].

Ground Truth

Prediction

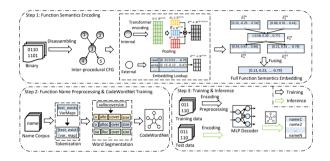
Takeaway



SYMLM

- ▶ A novel neural architecture that generates execution-aware context-sensitive code embeddings.
- ▶ Effective modules, function name preprocessing and CodeWordNet, to calculate function name similarity.
- ► Advancing the state-of-the-art and practical use cases.

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- ► Advancing the state-of-the-art and practical use cases.

The source code is available at https://github.com/OSUSecLab/SymLM.

References I



Gal Beniamini, Sarah Gingichashvili, Alon Klein Orbach, and Dror G Feitelson, Meaningful identifier names: the case of single-letter variables, 2017 IEEE/ACM 25th International Conference on Program Comprehension (ICPC), IEEE, 2017, pp. 45–54.



Kevin Burk, Fabio Pagani, Christopher Kruegel, and Giovanni Vigna, *Decomperson: How humans decompile and what we can learn from it*, 31st USENIX Security Symposium (USENIX Security 22), 2022, pp. 2765–2782.



Sanchuan Chen, Zhiqiang Lin, and Yinqian Zhang, Selectivetaint: Efficient data flow tracking with static binary rewriting, 30th USENIX Security Symposium (USENIX Security Symposium (USENIX Security 21), USENIX Association, August 2021.



Yaniv David, Uri Alon, and Eran Yahav, Neural reverse engineering of stripped binaries using augmented control flow graphs, Proceedings of the ACM on Programming Languages 4 (2020), no. OOPSLA, 1–28.



Lesly-Ann Daniel, Sébastien Bardin, and Tamara Rezk, Binsec/rel: Efficient relational symbolic execution for constant-time at binary-level, 2020 IEEE Symposium on Security and Privacy (SP), IEEE, 2020, pp. 1021–1038.



Dror Feitelson, Ayelet Mizrahi, Nofar Noy, Aviad Ben Shabat, Or Eliyahu, and Roy Sheffer, *How developers choose names*, IEEE Transactions on Software Engineering (2020).



Gateway, https://github.com/RiS3-Lab/p2im-real_firmware/blob/master/binary/Gateway, Accessed: 2022-04-26.



Han Gao, Shaoyin Cheng, Yinxing Xue, and Weiming Zhang, A lightweight framework for function name reassignment based on large-scale stripped binaries, Proceedings of the 30th ACM SIGSOFT International Symposium on Software Testing and Analysis, 2021, pp. 607–619.

References II



Masoud Ghaffarinia and Kevin W Hamlen, *Binary control-flow trimming*, Proceedings of the 2019 ACM SIGSAC Conference on Computer and Communications Security, 2019, pp. 1009–1022.



Jingxuan He, Pesho Ivanov, Petar Tsankov, Veselin Raychev, and Martin Vechev, *Debin: Predicting debug information in stripped binaries*, Proceedings of the 2018 ACM SIGSAC Conference on Computer and Communications Security, 2018, pp. 1667–1680.



Yi Li, Shaohua Wang, and Tien N Nguyen, *A context-based automated approach for method name consistency checking and suggestion*, 2021 IEEE/ACM 43rd International Conference on Software Engineering (ICSE), IEEE, 2021, pp. 574–586.



Tomas Mikolov, Ilya Sutskever, Kai Chen, Greg S Corrado, and Jeff Dean, *Distributed representations of words and phrases and their compositionality*. Advances in neural information processing systems **26** (2013).



James Patrick-Evans, Lorenzo Cavallaro, and Johannes Kinder, *Probabilistic naming of functions in stripped binaries*, Annual Computer Security Applications Conference, 2020, pp. 373–385.