DeLink: Source File Information Recovery in Binaries

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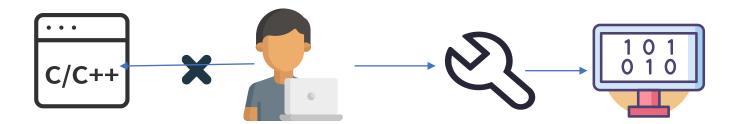
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Introduction



Program Comprehension



Program Comprehension in Binary Analysis include:

• Instruction-level analysis







• Function-level analysis (Focus of previous work, Debin(ccs 18), NFRE(issta 21), SymLM(ccs 22), Exact(ACSAC 20))



• **High-level analysis** (previous work: BCD(ccs 18))



C1: Lack of Explicit File Boundaries

C2: Difficulty in Feature Selection for File Name Prediction

Background

Terminologies

File Structure: Indicates which binary functions belong to a source file.



• **Homologous Functions**: Binary functions originating from the same source file.



Continuous Distribution: Homologous functions are placed continuously in a binary by the compiler.



Static Function: Defined with the static keyword, visible only within the same file.



External Function: Implemented in other binaries, such as APIs and system calls.



Background

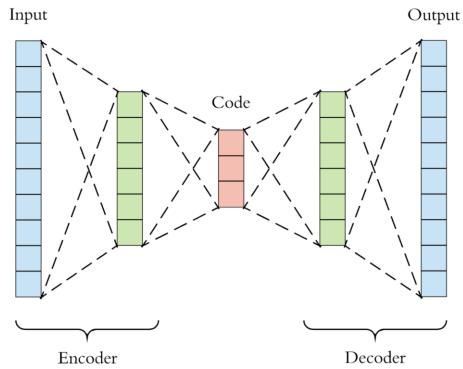
Key Techniques

• **Dynamic Programming (DP)**: Used to locate file boundaries by optimizing the file structure recovery process.

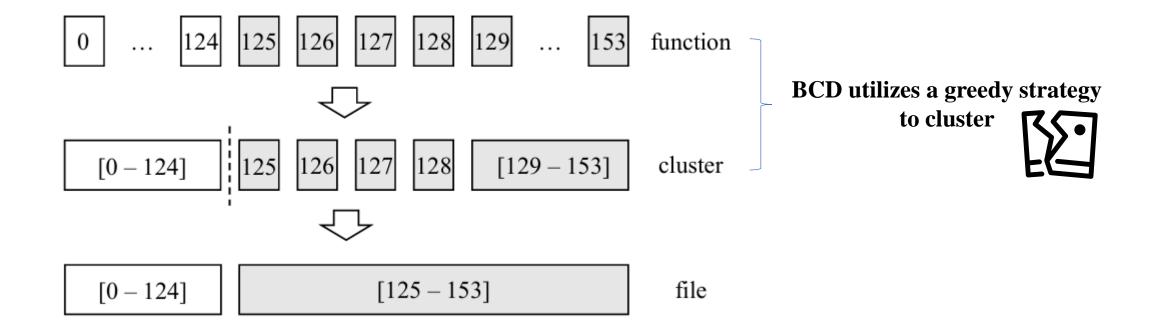


• Encoder-Decoder Model: Utilizes a graph neural network (GNN) to encode graph structure features for file

name prediction.

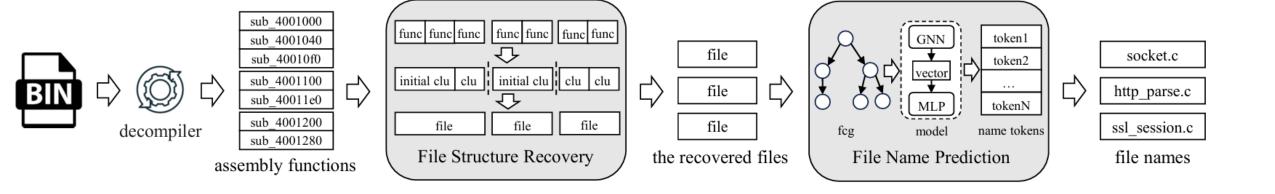


Motivating Example



An asynchronous event library *libev.so* is compiled from two source files, with one file containing 125 out of 154 functions, making it difficult to accurately identify all binary functions belonging to this file within the library.

Overview

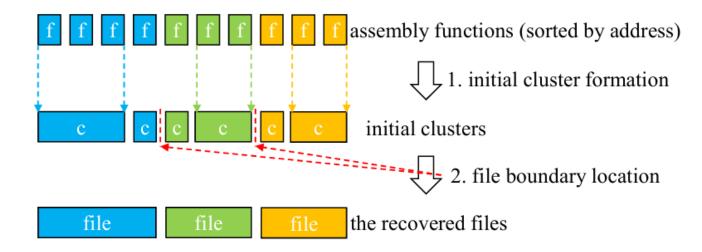


Overall Workflow of DeLink

- File Structure Recovery
- Core Idea
 - Static Function Visibility
 - Continuous Function Placement

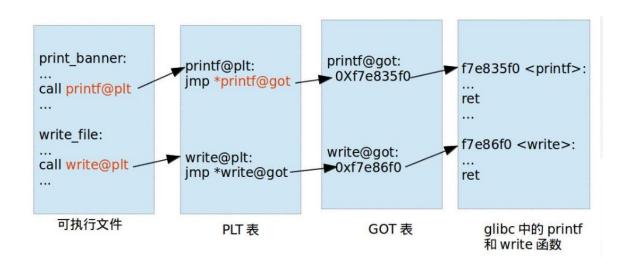
All binary functions between a static callee function and its caller function are from the same source file.



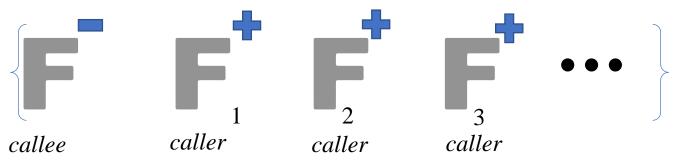


Overview of File Structure Recovery

- Identify static functions within dynamic libraries
 - Non-static functions are typically called through the Procedure Linkage Table (PLT) and Global Offset Table (GOT)
 - A static global variable has the same visibility as a static function, meaning that functions which access this variable are homologous



- Identify static functions within executable program
 - A three-step heuristic approach
 - 1. Function Set Construction
 - 2. Filtering Based on Index Distance
 - 3.Outlier Removal Using API References



 $D = \{d_1, d_2, ..., d_n\}$, where d_i is the function index distance from the caller function f_i^+ to its callee function f^- (i.e., $d_i = i_i^+ - i^-$). Besides, average function index distance $aver_dis$ is calculated by $\frac{1}{n} \sum_{i=1}^{n} |d_i|$.

Key Insights

- •Static functions tend to be located close to their caller functions.
- •Non-static functions might have larger, sometimes negative distances due to cross-file calls.
- •Header file analysis helps differentiate valid function relationships from outliers.

File Boundary Location

Algorithm 1: File Boundary Location

```
Input: the target binary B, the cluster set C.
  Output: the file boundary results R.
1 Function file_boundary_location(B, C):
       func\_feats \leftarrow get\_func\_feature(B)
       feat_matrix \leftarrow gen_feat_matrix(func_feats)
3
       while e clu not at the end of C do
            s\_clu, e\_clu \leftarrow get\_clusters(C, file\_bdy\_pos)
            bdy\_array \leftarrow dynamic\_programming(s\_clu, e\_clu,
              feat_matrix)
            file\_bdy\_pos \leftarrow backtrack\_pos(bdy\_array)
7
            if number(file\_bdy\_pos) > 0 then
                 R \leftarrow add\_boundary(file\ bdy\ pos)
            end
10
       end
11
       return R
12
```

Steps

- **1.Cluster Formation**:
- **2.Function Relationship Matrix**
- **3.Subset Processing**
- 4. Dynamic Programming (DP) Optimization
- **5.**Backtracking for Final Boundaries:

$$Q = \frac{1}{2W} \sum_{i,j} \left[w_{ij} - \frac{k_i^{out} k_j^{in}}{2W} \right] \delta(C_i, C_j)$$

File Name Prediction

• Code Feature Selection(22.458 source files from 663 open-source software projects)

Total Files	Files Named with Func Tokens	Proportion
22, 458	20, 387	90.78%

Statistics on Files Named with Function Name Tokens.

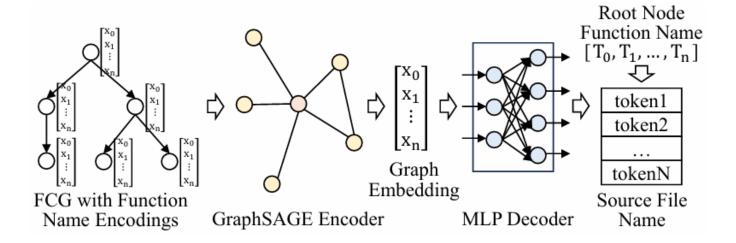
Function Level	File Token Number	Proportion		
First Level	32, 738	58.57%		
Second Level	28, 609	51.18%		
Third Level	27, 240	48.73%		
Fourth Level	23, 098	41.32%		

FCG && The root node function

eg : in ssl_session.c, first level :ssl_session_init()

Statistics on File Name Token Occurrences across FunctionLevels.

Prediction Model



- Training use source code directly
- Prediction when faced with binary with no symbols, use SymLM to predicate function name

File Structure Recovery Quality Evaluation

metrics:

1.Precision(P),Recall(R),and F1score(F1) to evaluate the **file structure recovery results**

$$P = \sum_{i=1}^{N_f} \frac{n_i}{n_f} P_i, \quad R = \sum_{i=1}^{N_f} \frac{n_i}{n_f} R_i, \quad F_1 = \sum_{i=1}^{N_f} \frac{n_i}{n_f} F_1^i$$

2.Jaccard(J),Fowllkes-Mallows(FM),Rand(RI) to evaluate ~ from clustering perspective

$$JC = \frac{a}{a+b+c}$$

$$FMI = \sqrt{\frac{a}{a+b} \cdot \frac{a}{a+c}}$$

$$RI = \frac{a+d}{a+b+c+d}$$

• File Structure Recovery Quality Evaluation

Dataset 1: consists of 106C/C++ executable programs from ModX and 14 C++ executable programs from BCD

C

Dataset 2: composed of 45 dynamic libraries, with 6 of them compiled using the "-fvisibility=hidden" optimization.

Results on Dataset I (%)			Results on Dataset II (%)									
Approach	Precision	Recall	F_1 score	JC	FMI	RI	Precision	Recall	F_1 score	JC	FMI	RI
BCD	58.28	69.51	63.17	41.81	60.34	95.51	67.02	75.53	70.79	37.37	55.79	92.36
ModX	69.71	62.88	66.01	8.81	17.47	92.94	75.99	67.39	71.23	11.56	22.73	89.61
DeLink	77.77	83.63	80.51	49.03	66.12	96.02	82.70	86.59	84.54	72.27	82.53	97.25

• File Name Prediction Accuracy

Approach	Precision(%)	Recall(%)	$F_1 \text{ score}(\%)$
TF-IDF	57.22	47.44	51.87
Graph2Seq	63.37	51.32	56.71
DeLink	70.09	63.91	66.86

File Name Prediction Accuracy Comparison.

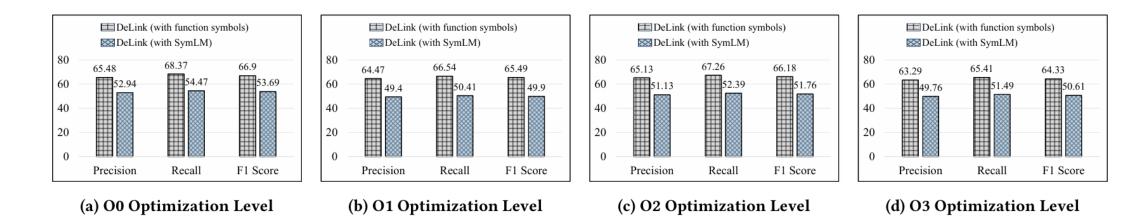
• Limitation Analysis

TF-IDF: Fixed Output Length && Vocabulary Coverage Issue

Graph2Seq: Error Propagation && Local Perspective Limitation

• File Name Prediction Without Acess to function symbols

Dataset IV: consists of 1,118 files from 60 stripped binaries.



The Accuracy of DeLink with Original Function Symbols and with SymLM

• File Structure Recovery Time Comparison

Recovery Time (s)	Dataset I	Dataset II	Average
BCD	44.99	7.26	34.43
ModX	170.25	33.01	131.79
DeLink	20.78	6.41	16.74

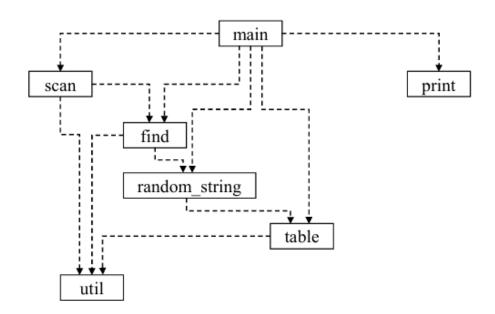
• File Name Prediction Time Comparison

Prediction Time (ms)	Average
TF-IDF	1.28
Graph2Seq	0.33
DeLink	0.32

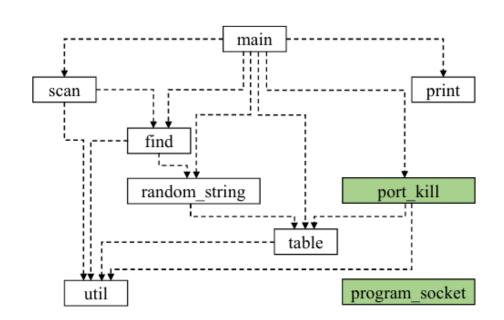
Application

Malware homology analysis

Mirai: a well-known botnet program family that can be utilized for large-scale and destructive distributed denial of service (DDoS) attacks



(a) Mirai Sample A



(b) Mirai Sample B

End.