Assembly Code Search Engine for Reverse Engineer

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Technical Details

Efficient inexact assembly code search: The assembly code vector space is highly skewed. Small blocks tend to be similar to each other and large blocks tend to be sparsely distributed in the space. Original hyperplane hashing with banding technique equally partitions the space and does not handle the unevenly distributed data well. We propose a new adaptive locality sensitive hashing (ALSH) scheme to approximate the cosine similarity. ALSH organizes

buckets into a tree structure. To our best knowledge, ALSH is the First incremental locality sensitive hashing scheme that

solves this issue specifically for cosine space with theoretical guarantee.

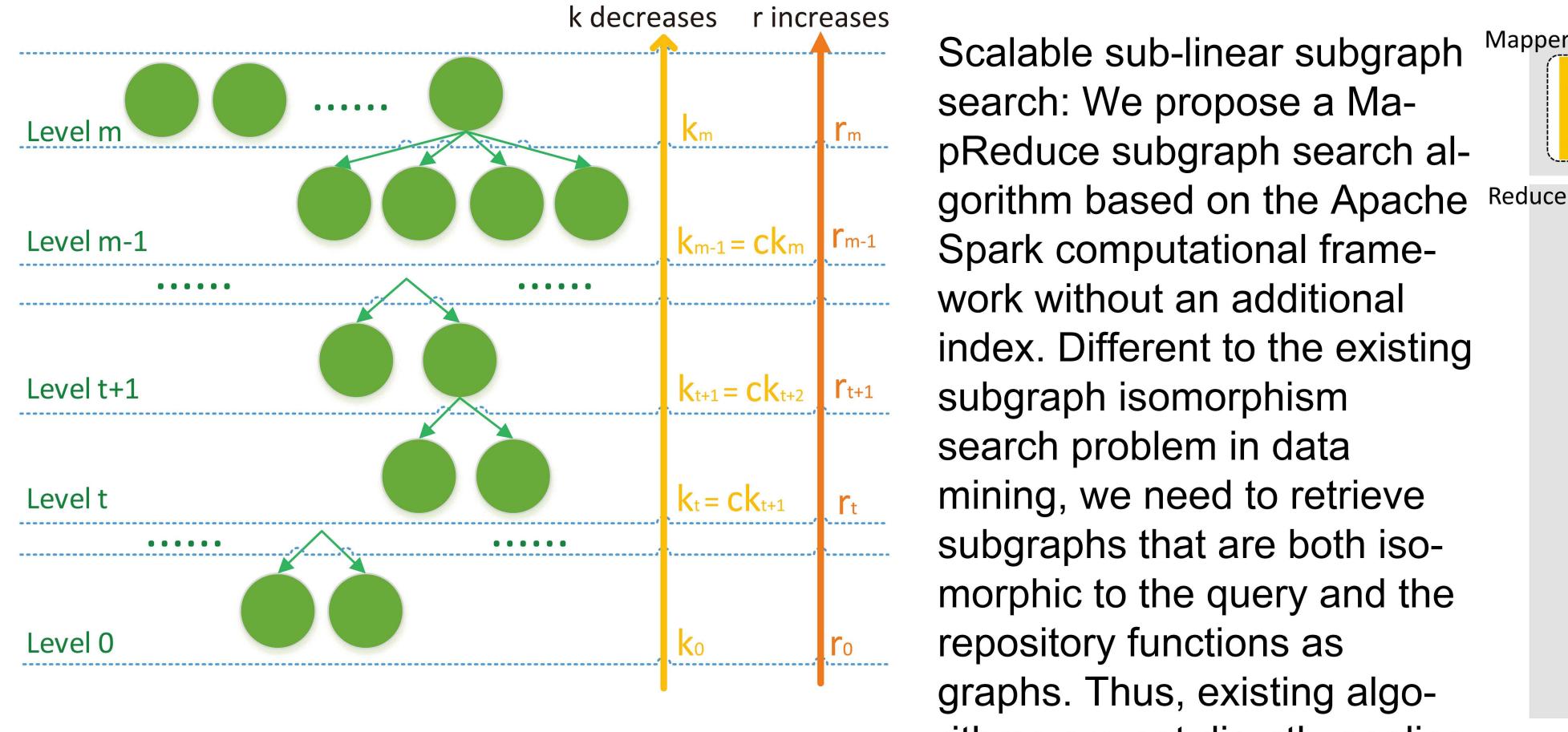


Figure 5. The index structure for the Adaptive Locality Sensitive Hashing (ALSH).

graphs. Thus, existing algorithms are not directly applica-Figure 6. The index structure for the Adaptive Loble. Algorithmically, our apcality Sensitive Hashing (ALSH). proach is bounded by polynomial complexity. However, our

Experimental Evaluation

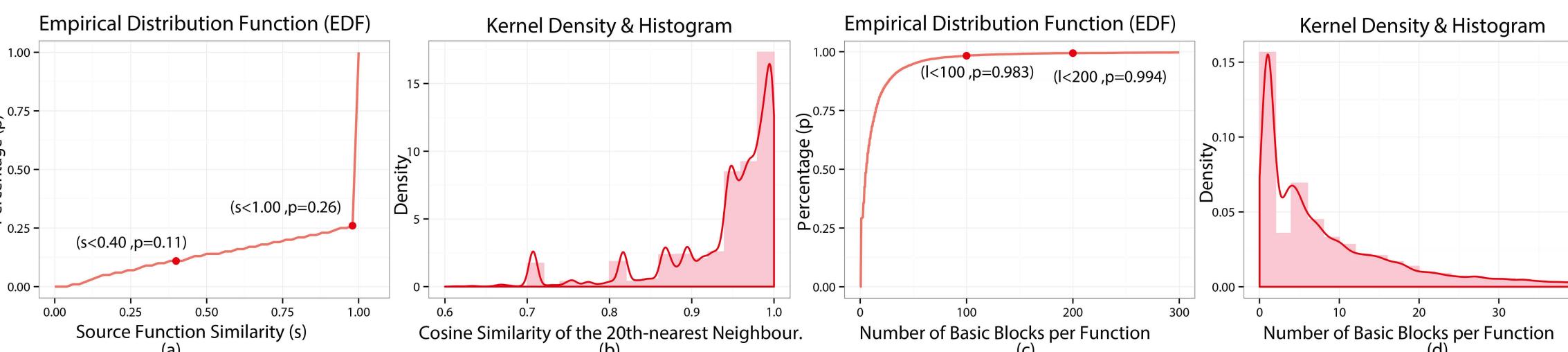
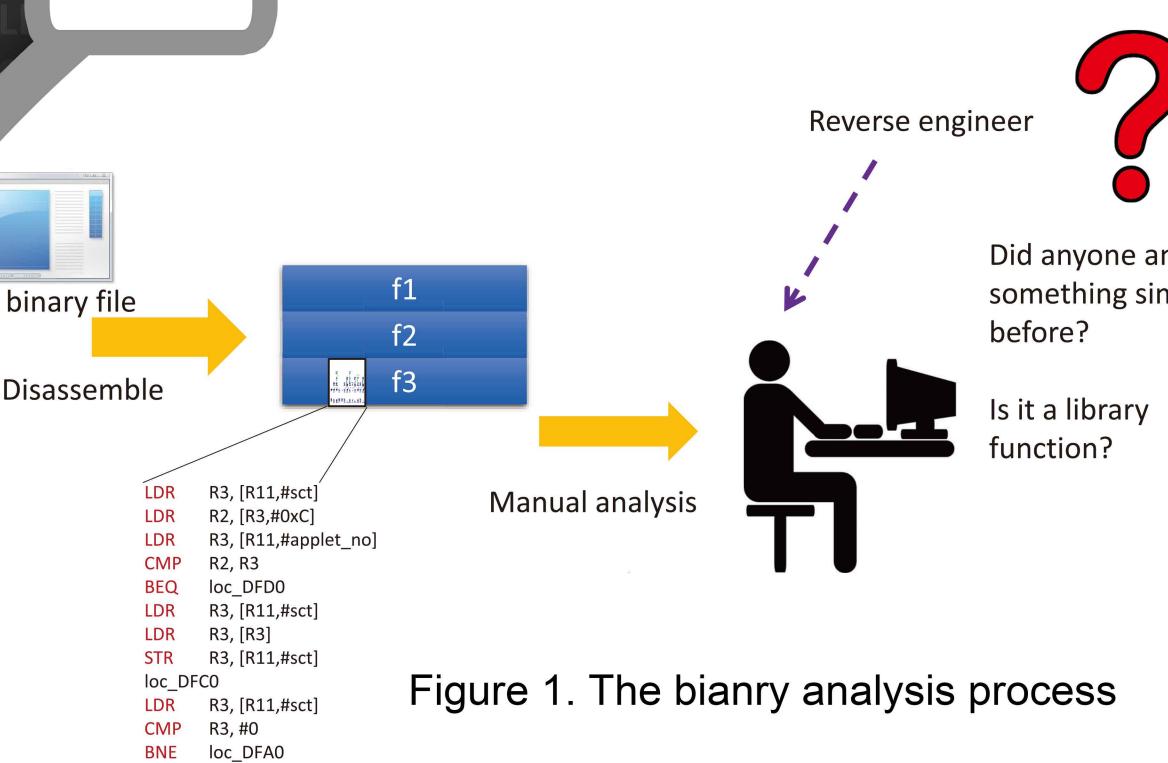


Figure 7: (a) the EDF function on repository function clone pair similarity, (b) the kernel density & histogram of the cosine similarity of each basic block's 20th-nearest neighbor, (c) the EDF on per assembly function block count, and (d) the kernel density & histogram on assembly function block count < 40.

Research Problem



A binary file can be disassembled

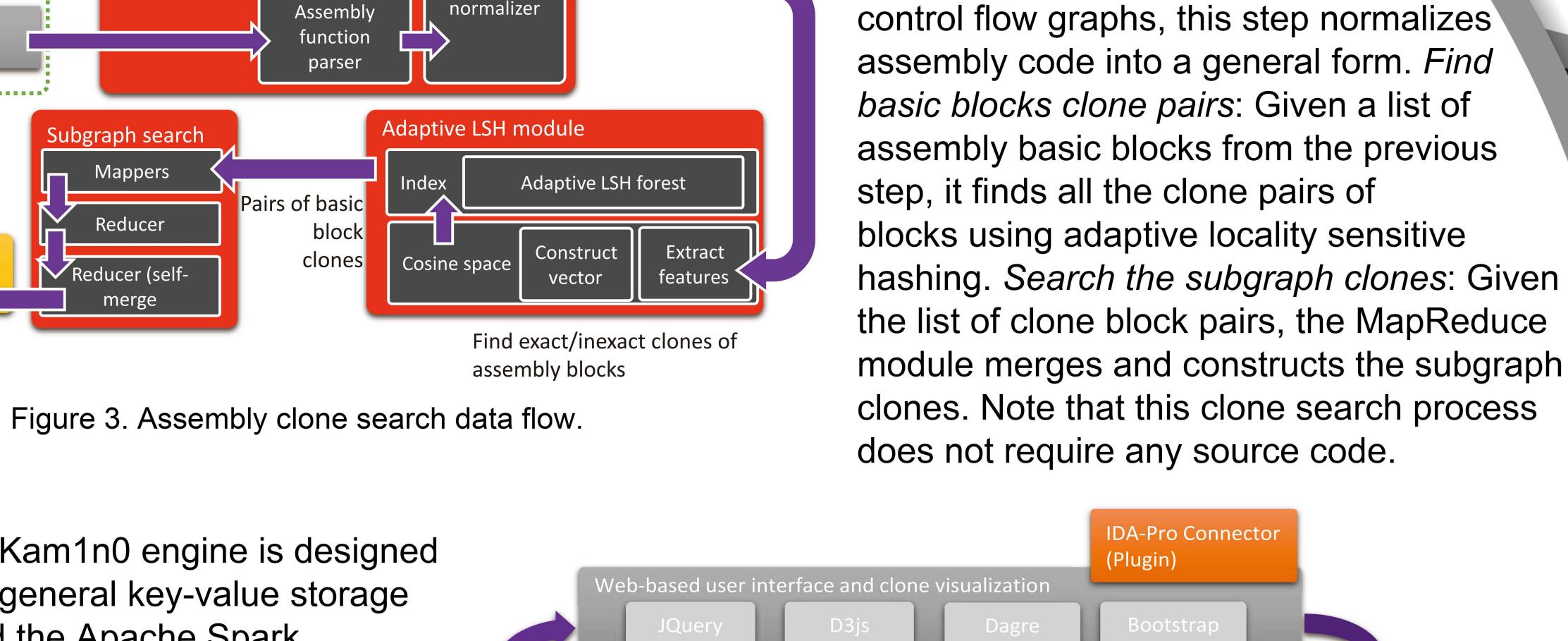
Assembly code analysis is one of the critical processes for detecting and proving software plagiarism and software patent infringements when the source code is unavailable. It is also a common practice to discover exploits and vulnerabilities in existing software. However, it is a manually intensive and time-consuming process.

subgraphs of

The Kam1n0 engine is designed for general key-value storage and the Apache Spark computational framework. Its solution stack, as shown in Figure 2, consists of three layers. The data storage layer is concerned with how the data is stored and indexed. The distributed/local execution layer manages and executes The jobs submitted by the engine. The Kam1n0 engine splits a search query into multiple jobs and

coordinates their execution flow.

Overall Architecture



The search process consists of

the three steps. *Preprocessing*:

After parsing the input (either a

binary file or assembly functions) into

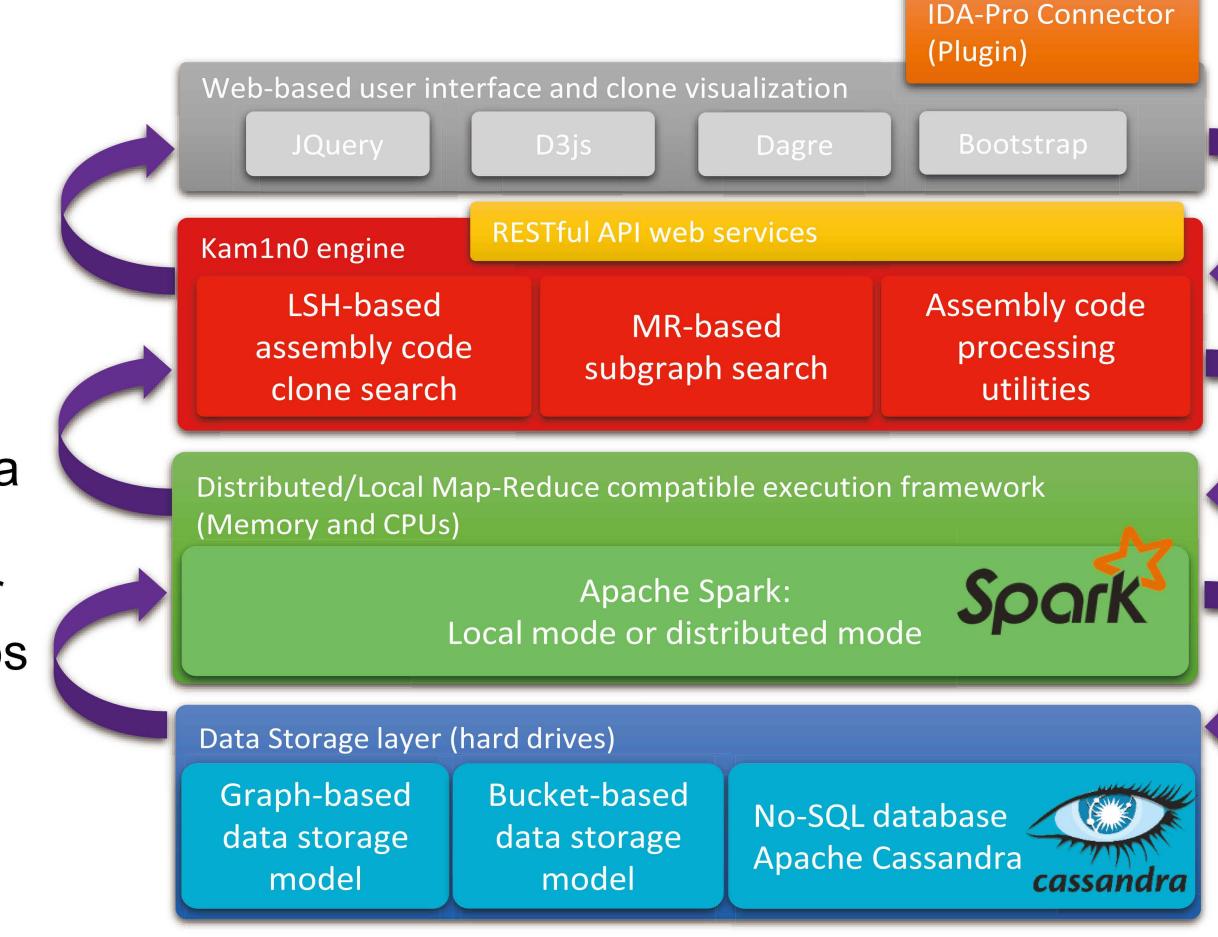


Figure 4. The solution stack of implementation.

Kam1n0 is open-source and available on GitHub.

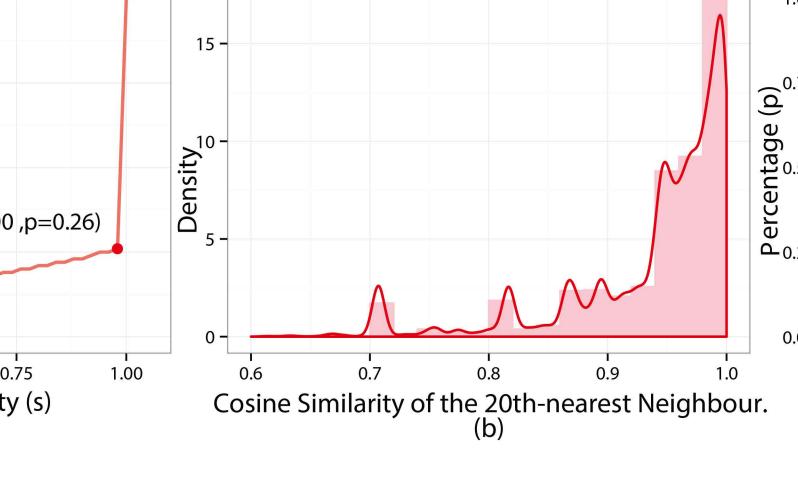
Scan the code to check out Kam1n0 and the paper.

in to as list of assembly functions. A function can be represented as a control flow graph (see Figure 2). Per our discussion with a practical search engine should be able to decompose the given query assembly function to different known subgraph clones which can help reverse engineers better understand the function's composition. A subgraph consists short loc_401100 short loc_401 of several interconnected *basic* blocks; and there are three types ------> An explanation label A block to block clone pa of clones between basic blocks. _____ A jump link between two

Figure 2. An example cloned subgraph

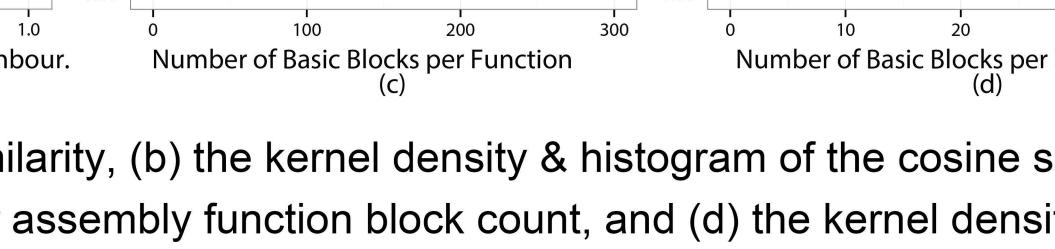
Average Indexing Time per-Function Average Query Response Time per-Function (b) Number of Functions in Repository

Figure 8. Scalability study. (a): Average Indexing Time vs. Number of Functions in the Repository. (b): Average Query Response Time vs. Number of Functions in the Repository.



experiment suggests that it is

sub-linear in practice.



A block-to-block

A cloned subgraph

Cloned subgraphs

We construct a new labeled one-to-many assembly code clone dataset that is available to the research community by linking the source code and assembly function

We benchmark the performance of twelve existing state-of-the-art solutions on the dataset using three typical information retrieval metrics (Table 1). Kam1n0 boosts the clone search quality and yields stable results across different datasets and metrics.

level clones.

We also setup a mini-cluster (4 nodes) to evaluate the scalability of Kam1n0 (Figure 8).

metrics: the Area Under the Receiver Operating Characteristic Curve (AUROC), the Area Under the Precision-Recall Curve (AUPR), and the Mean Average Precision at Position 10 (MAP@10). Ø denotes that the method s not scalable and we cannot obtain a result for this dataset within 24 hours