

Greybox Program Synthesis: A New Approach to Attack Dataflow Obfuscation

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About me



- Software Security Engineer @ Quarkslab
- Primarily interested in attacking **obfuscation** and **automating bug discovery**

I. Introduction

II. Synthesis Primer

- Usages
- Application to software deobfuscation

III. Greybox Synthesis

- Algorithm overview
- black-box I/O oracle
- whitebox AST search

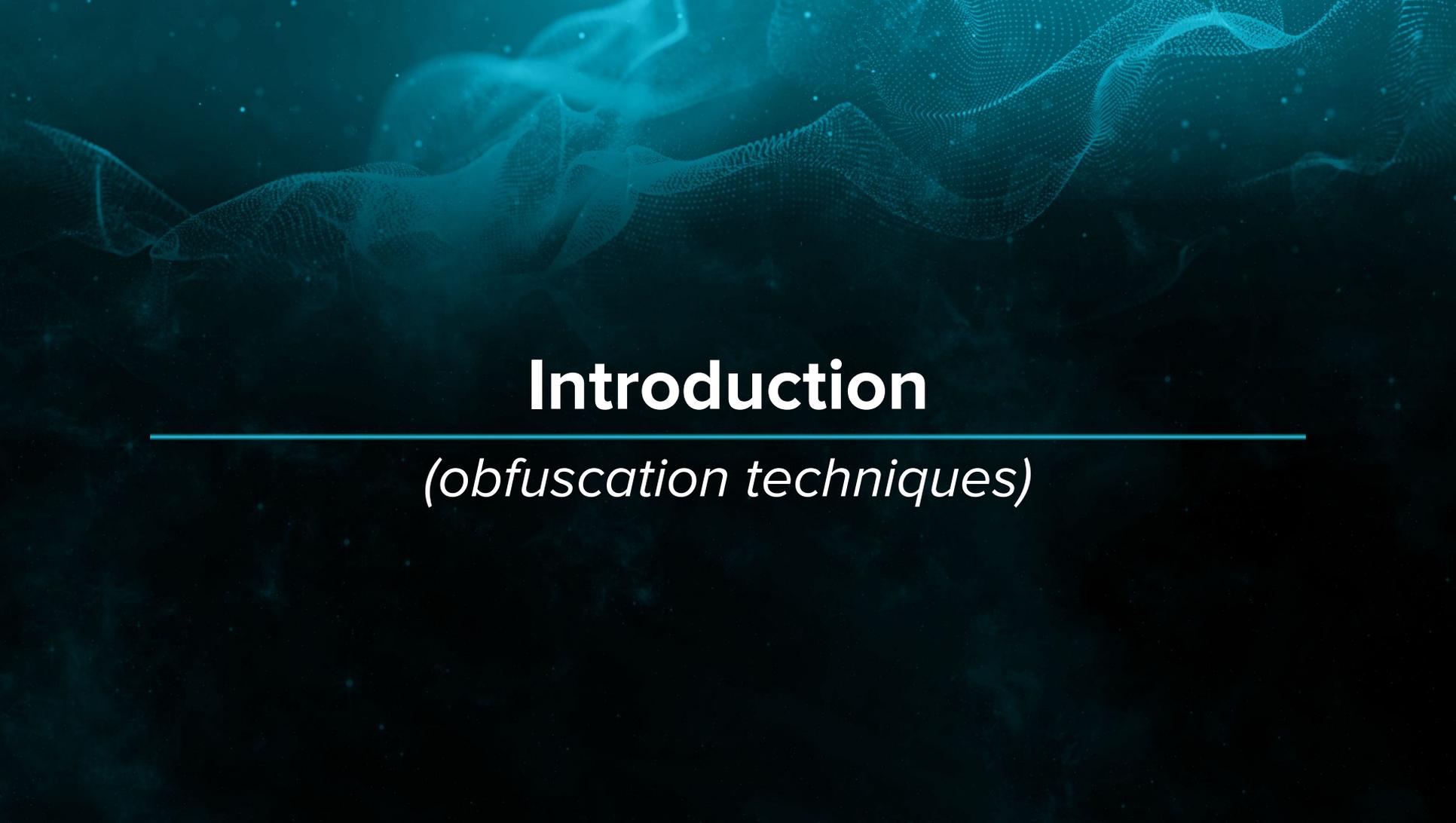
IV. Table generation

V. Implementation in **QSynthesis** *(deobfuscation up-to reassembled instructions)*

- implementation & reassembly
- IDA integration

VI. Use-cases

VII. Conclusion



Introduction

(obfuscation techniques)

What ?

Transformation of a program P in a **semantically equivalent** P'
harder to understand

Why ?

To protect **intellectual property** from
reverse-engineering

How ?

By hiding **valuable assets** of the program
(which are usually)

program logic
algorithms

(referred as control-flow)

program data

keys, strings, constants...
(referred as data-flow)

Control-Flow Obfuscation

Hiding the **logic** and algorithm of the program

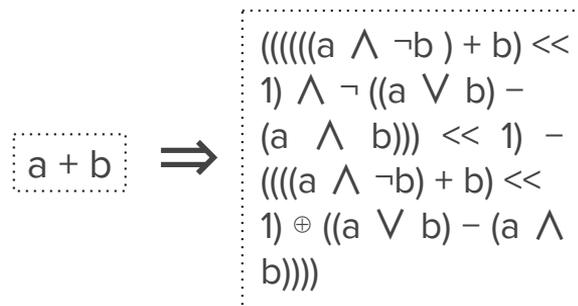
virtualization, opaque predicates, CFG-flattening, split, merge, packing, implicit flow, MBA, loop-unrolling...



Data-Flow Obfuscation

Hiding **data**: constants, strings, APIs, keys etc.

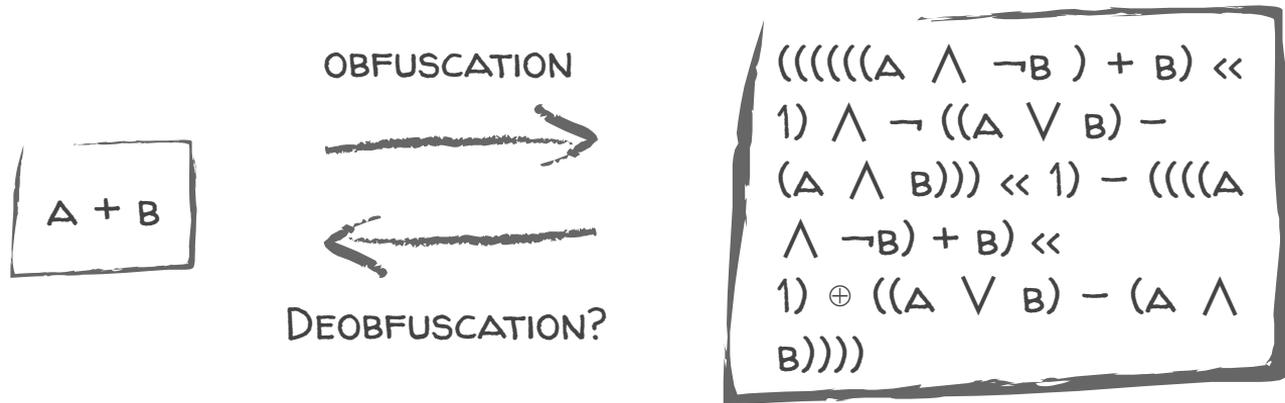
data encoding, MBA, arithmetic encoding, whitebox, array split/fold/merge, variable splitting...



Data Obfuscation (*data-flow*)



⇒ This work focuses on data-flow and more especially **MBA** (Mixed Boolean Arithmetic)
(but many other transformation exists like: data encoding, whitebox, variable splitting/merging ..)



! Problem

Reversing the transformation is hard (*unlike many control-flow obfuscation, solution is not boolean*)



Deobfuscating data-flow expressions on real-world obfuscated programs yield **two distinct** research problems.

PB #1

Locating the data to deobfuscate and knowing **what to deobfuscate** (*depends on what you're looking for in the binary*).

(This is specific to each binary and is mostly manual)

PB #2

Deobfuscating the data obtained after it gets located (*in our context a data-flow expression*).

*(Synthesis **only** addresses this issue !)*

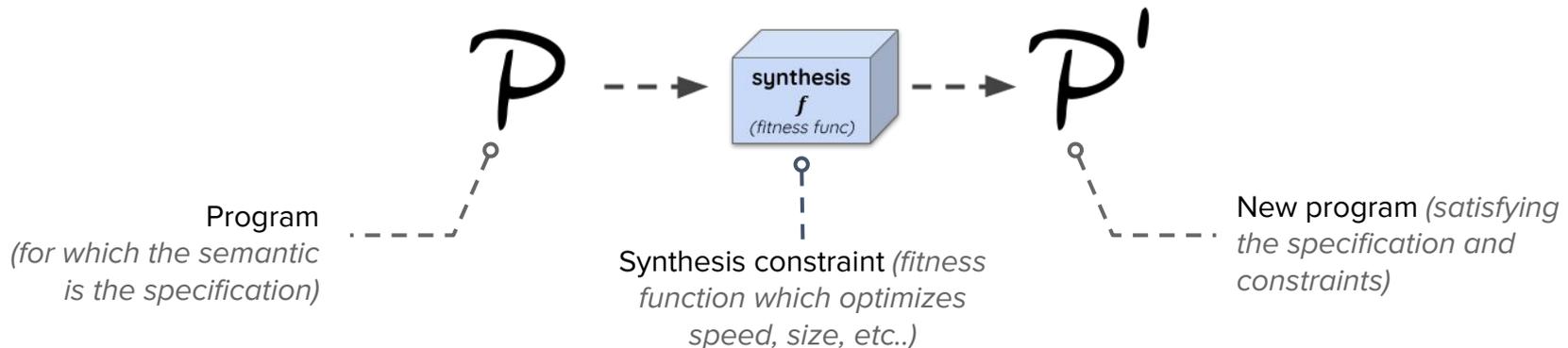
Synthesis primer

Program Synthesis



⇒ Program synthesis consists in automatically deriving a program from

- A high-level **specification** (typically its semantic through its I/O behaviour)
- Additional constraints:
 - Compilation: a **faster** program
 - Deobfuscation: a **smaller** or more readable program



Synthesis for Superoptimization



Synthesis is used in a **variety** of **domains**.
Applied on program analysis it is mostly
used for **optimization** (known as *super-optimization*)
or **deobfuscation**.



A Synthesizing Superoptimizer

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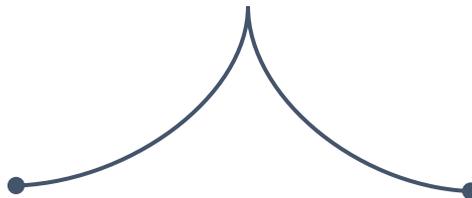
Abstract

If we can automatically derive compiler optimizations, we might be able to sidestep some of the substantial engineering challenges involved in creating and maintaining a high-quality compiler. We developed Souper, a synthesizing superoptimizer, to see how far these ideas might be pushed in the context of LLVM. Along the way, we discovered that Souper's intermediate representation was sufficiently similar to the one in Microsoft Visual C++ that we applied Souper to that compiler as well. Shipping, or about-to-ship, versions of both compilers contain optimizations suggested by Souper but implemented by hand. Alternately, when Souper is used as a fully automated optimization pass it compiles a Clang compiler binary that is about 3 MB (4.4%) smaller than the one compiled by LLVM.

signed for LLVM [12] but we have also used it to find new optimizations for the Microsoft Visual C++ compiler. Several trends convinced us that it was time to write a new superoptimizer. There has been increased pressure on compiler developers due to the adoption of higher-level programming languages and a proliferation of interesting hardware platforms. SAT and SMT solvers continue to improve; they are already more than capable of discovering equivalence proofs necessary to verify compiler optimizations involving tens to hundreds of instructions. Solvers are also a key enabler for program synthesis, which supports the discovery of new optimizations that are out of reach for naïve search. Finally, verified compilers appear to be much more difficult to extend than are traditional compilers. Though we have not yet done so, a natural extension of superoptimization research would be to use a proof-producing solver to

Xiv:1711.04422v2 [cs.PL] 6 Apr 2018

Superoptimizers



STOKE

A stochastic superoptimizer and program synthesizer
STOKE is a stochastic optimizer and program synthesizer for the x86-64 instruction set. STOKE uses random search to explore the extremely high-dimensional space of all possible program transformations. Although any one random transformation is unlikely to produce a code sequence that is desirable, the repeated application of millions of transformations is sufficient to produce novel and non-obvious code sequences. STOKE can be used in many different scenarios, such as optimizing code for performance or size, synthesizing an implementation from scratch or to trade accuracy of floating point computations for performance. As a superoptimizer, STOKE has been shown to outperform the code produced by general-purpose and domain-specific compilers, and in some cases expert hand-written code.

Publications

STOKE has appeared in a number of publications.

- **Stochastic Superoptimization** – ASPLOS 2013
- **Data-Driven Equivalence Checking** – OOPSLA 2013
- **Stochastic Optimization of Floating-Point Programs with Tunable Precision** – PLDI 2014
- **Conditionally Correct Superoptimization** – OOPSLA 2015
- **Stochastic Program Optimization** – CACM 2016
- **Stratified Synthesis: Automatically Learning the x86-64 Instruction Set** – PLDI 2016
- **Sound Loop Superoptimization for Google Native Client** – ASPLOS 2017

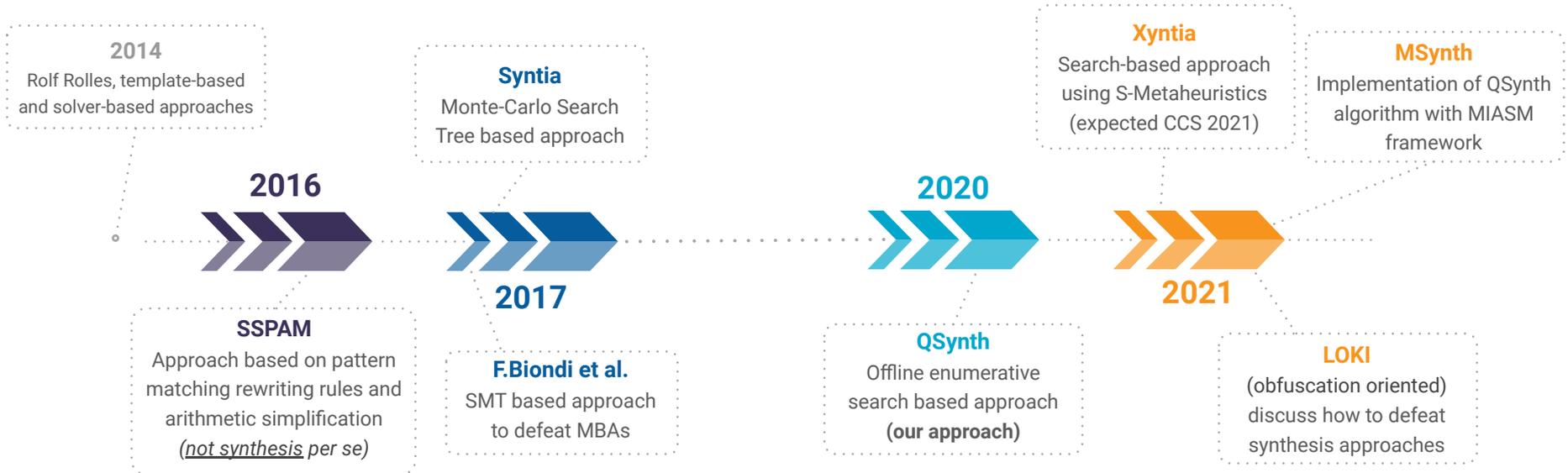
Souper: superoptimizer for LLVM IR
(backed by SMT solving)

STOKE: stochastic superoptimizer at
assembly level (x86_64)

Synthesis for Deobfuscation



Multiple approaches exist, **templates**, **stochastics** (e.g *MCTS*), **solver-based**, **enumerative** approaches, **search-based** (*S-Metaheuristics*) etc...



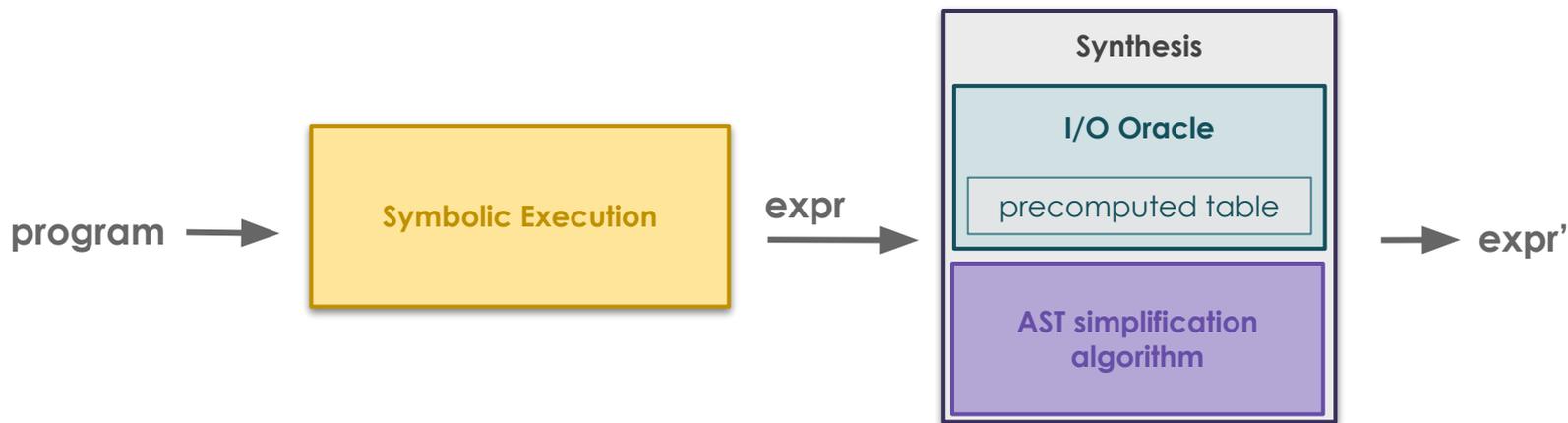
Greybox Synthesis

(design & principles of our algorithm)

Synthesis algorithm



Our algorithm is based on an **enumerative approach** backed by **symbolic execution** and a **synthesis** (*itself based on two sub-components*)



Symbolic Execution



⇒ We use symbolic execution as a means of extracting **data-flow expressions** of registers or memory at arbitrary locations in the program. The symbolic execution can either be **static** or **dynamic**.

Can backtrack expressions up to program entry

Avoid having to execute the program

Assembly

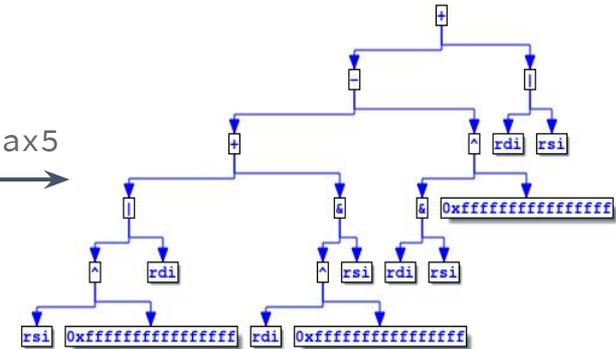
```
mov    rax, rsi
xor    rax, 0xFFFFFFFF
or     rax, rdi
mov    rcx, rdi
xor    rcx, 0xFFFFFFFF
and    rcx, rsi
mov    rdx, rdi
and    rdx, rsi
xor    rdx, 0xFFFFFFFF
or     rdi, rsi
or     rdi, rsi
add    rax, rcx
sub    rax, rdx
add    rax, rdi
retn
```

Intermediate Representation

```
rax0 := rsi
rax1 := rax ⊗ 0xFFFFFFFF
rax2 := rax1 | rdi
rcx0 := rdi
rcx1 := rcx0 ⊗ 0xFFFFFFFF
rcx2 := rcx1 & rsi
rdx0 := rdi
rdx1 := rdx0 & rsi
rdx2 := rdx1 ⊗ 0xFFFFFFFF
rdi0 := rdi | rsi
rax3 := rax2 + rcx2
rax4 := rax3 - rdx2
rax5 := rax4 + rdi0
```

SE →

rax5 →



AST

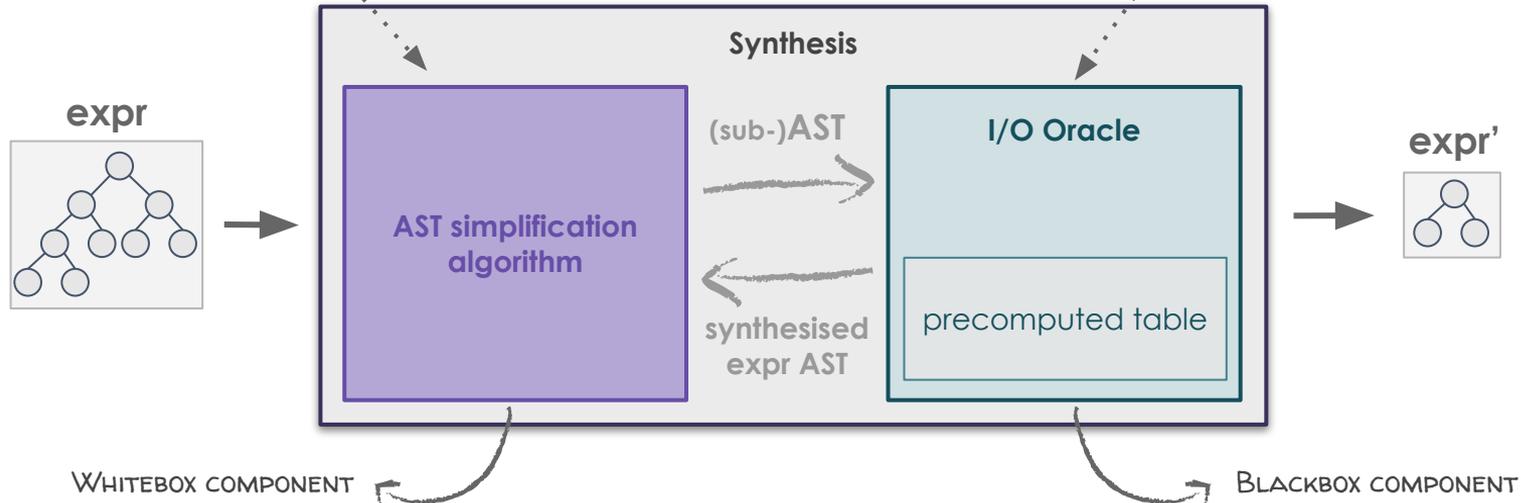
Our synthesis algorithm



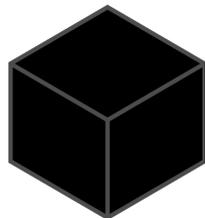
Our algorithm is a **greybox synthesizer** based on two components

An **AST simplification** algorithm that can use **various strategies**

An **I/O oracle** based on an **offline enumerative search** backed by a **pre-computed table**



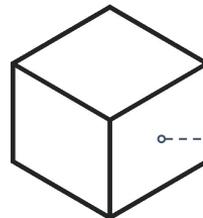
Blackbox vs Whitebox in Synthesis (for deobfuscation)



Blackbox

relates to approaches considering expressions to synthesize as blackboxes and only interacting with them through their **input/output behavior**

- + only influenced by semantic complexity
- large search space
- boolean result (*fully synthesized or not at all*)



Whitebox

relates to approaches manipulating the semantic of the expression through its syntactic representation (*usually the AST of the semantic*)

- + the exact semantic is considered
- influenced by syntactic complexity
- + enable sub-expressions synthesis

$$\begin{aligned} & (((((a \wedge \neg b) + b) \ll 1) \wedge \neg ((a \vee b) - (a \wedge b))) \ll 1) - (((a \wedge \neg b) + b) \ll 1) \oplus ((a \vee b) - (a \wedge b))) \end{aligned}$$

Blackbox I/O Synthesis Oracle



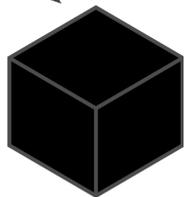
Blackbox I/O Oracle

set of pseudo-random inputs

$V_{in} =$

	A	B
i1	0	1
i2	-1	3
i3	4	1

A + B



$V_{out} =$

o1	o2	o3
1	2	5

o1	o2	o3
1	2	5

Equivalent !

Pre-computed tables

Given a grammar with some **operators** (+, -, |, &, ⊕, ..), and **variables** (a, b, c..), derives all possible expressions (up to a given bound) and evaluate them on V_{in} to obtain a function:

$$V_{out} \mapsto \text{expr}$$

V_{out}	expr
<1, 2, 5>	A + B
<-1, -4, 3>	A - B
<1, -1, 5>	A B
...	...

- o **generated once**, and ensures **O(log(n))** synthesis
- o **Unsound** but equivalence can be checked by SMT

⇒ What happens if it cannot synthesize the root node ?

Whitebox AST search

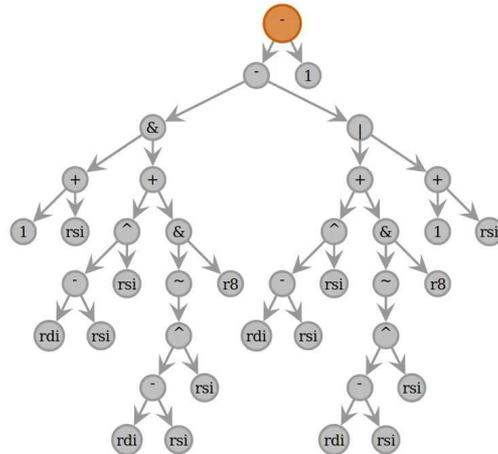


- If it cannot synthesize root node it aims at simplifying sub-expressions to obtain **at least a partial synthesis** (while with an I/O oracle the result is boolean).
- Thus an **AST search algorithm** will iterate through the graph looking for sub-nodes to synthesize.

Algorithm

1. Search a node to synthesize
2. if find one, replaces it by a temporary placeholder
3. if not, replaces it also
4. repeat the search until having substituted all nodes
5. recursively replace placeholders by the corresponding AST (synthesized or original)

Original strategy



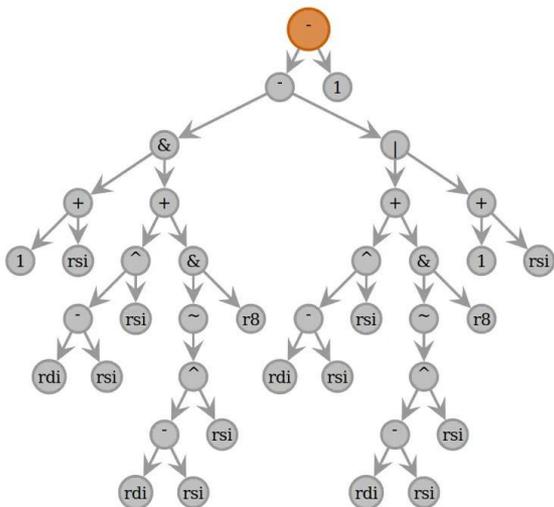
This simplification strategy have some **complexity issues** (yet it provides optimal results)

New AST search strategies



Top-Down *(Divide & Conquer)*

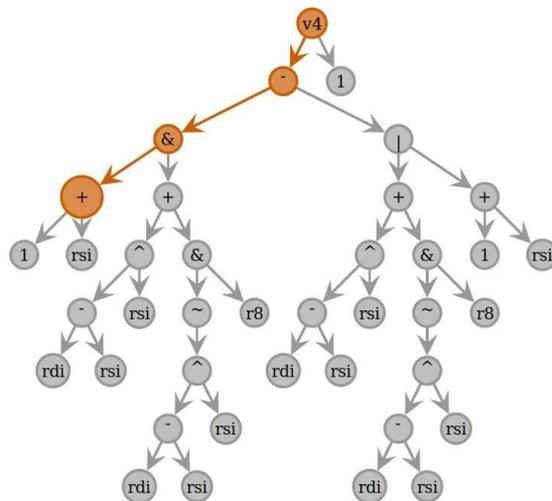
Single **DFS** traversal of the AST. Ensures linearity of the simplification of the algorithm *(while original one was quadratic in the worst case)*.



<https://youtu.be/V0Rg3LHC6Lw>

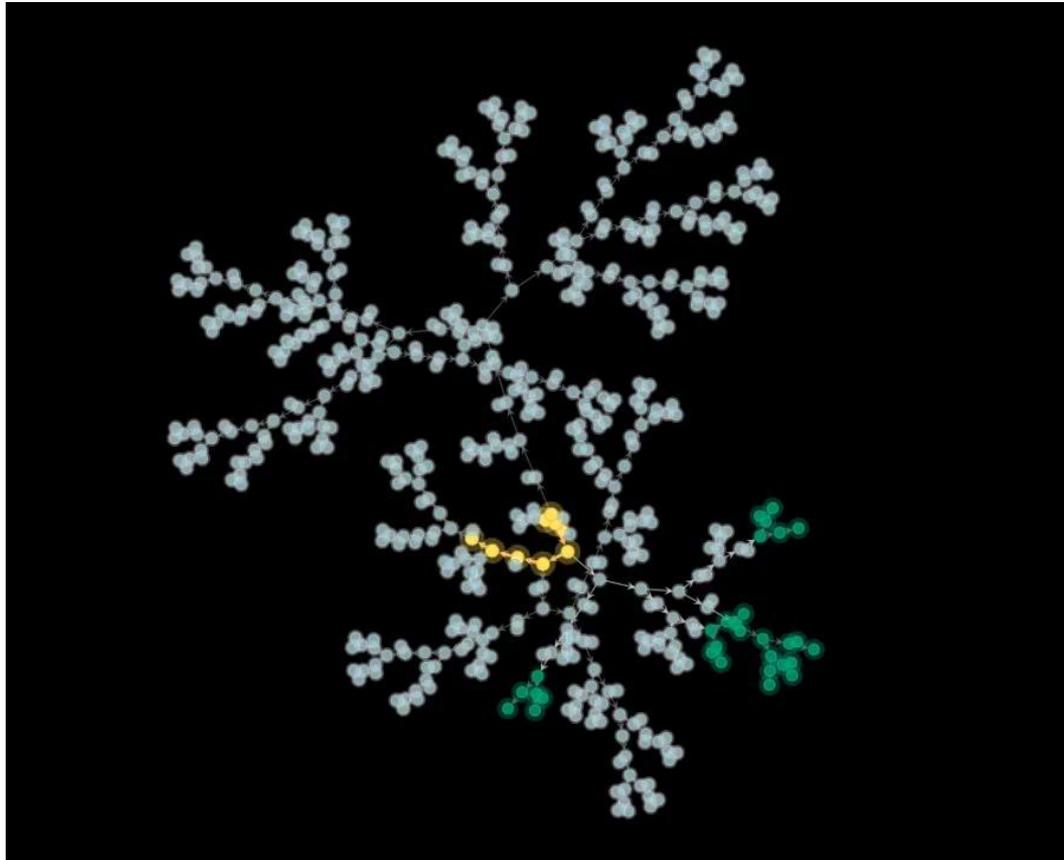
Top-Down & Bottom-Up

Like Top-Down but if a node gets synthesized attempts to re-synthesize its parents by means of reducing the variable cardinal.



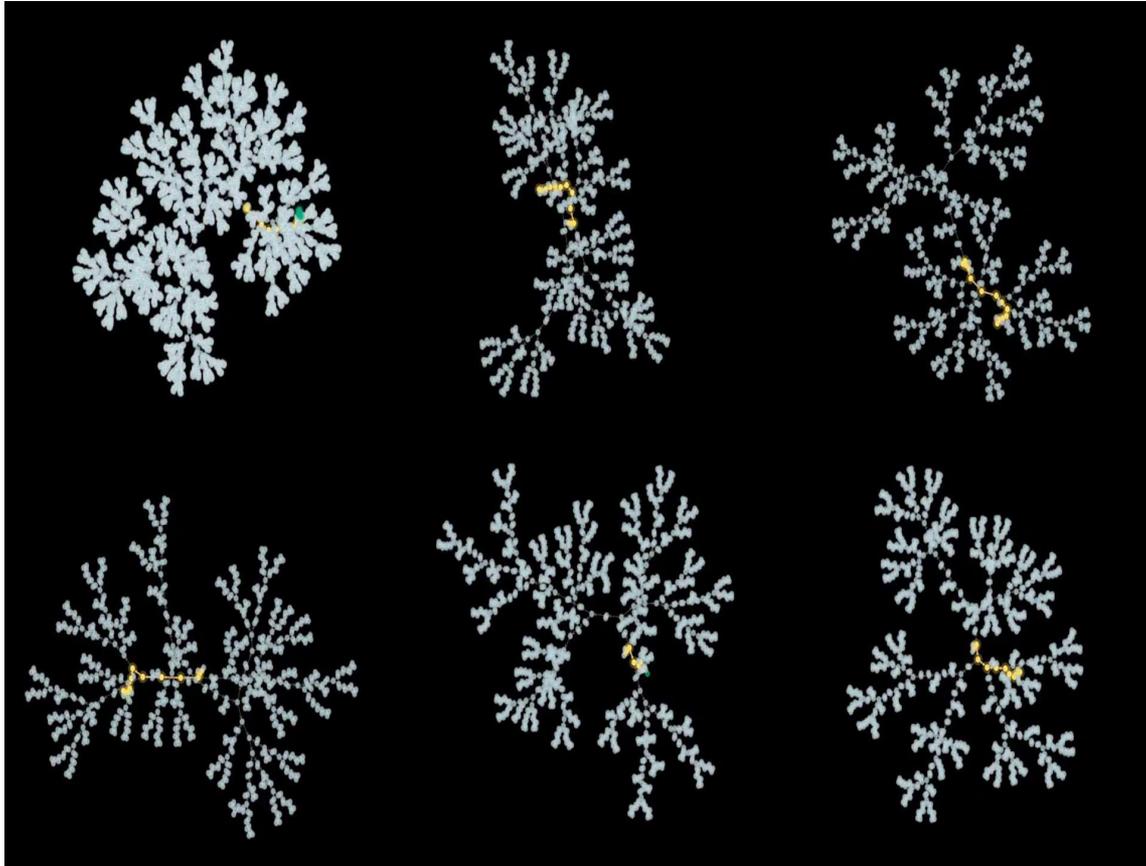
<https://youtu.be/G1lB0qmWLaI>

Algorithm Visualization



<https://youtu.be/Nz8KC1HtgiI>

Algorithm Visualization



<https://youtu.be/9MHeGtc3Uhc>

Algorithm Visualization

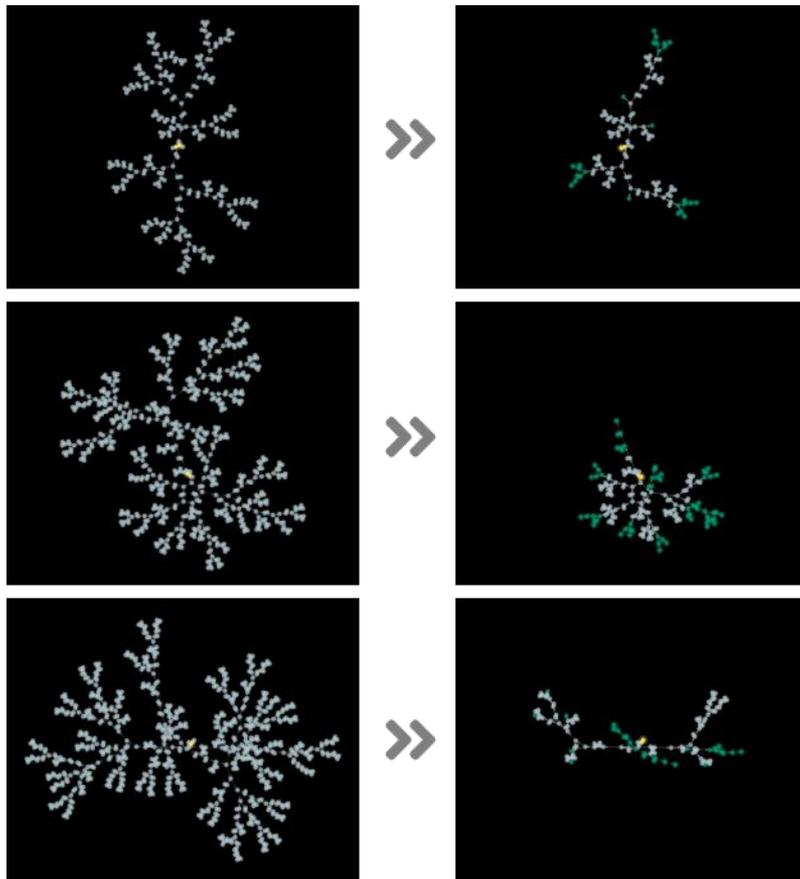


Table generation

(aka generating a potent I/O oracle)

Table Generation



⇒ Table generation requires **evaluating millions** of expressions and keeping millions of V_{out} vectors to ignore identical ones (*by construction we generate from smaller to larger expressions*).

Improvements:

- **Memoization** of all evaluated expressions (*thus $A+B$ is evaluated only once for all, when combined with another expression like $A+B-C$ the memoized result is reused for evaluation*)
- **JITting** of expressions evaluation. Evaluation made on native integers (*not using Python*). For that uses **dragonFFI** (*could also have used numpy*).

reach
25K exprs/sec



We now have a table with **375 million entries**
(last year we had ~3 millions)
(Generated with a 235 GB RAM machine :p)

Table Storage



pickle

Python object
serialization module

- Requires loading the whole table
- Parsing is slow on large object

⇒ Ok for small tables but limited for larger ones

(format used by MSynth)



Python ORM for
databases like sqlite

- If V_{out} primary key, insertion is linear in number of entries.
- If not, lookup is linear in the number of entries

⇒ Not suitable for such large tables



levelDB

Key Value database
(by Google)

- Store keys as “tries” to ensure **$O(\log(n))$** access
- Automatic caching mechanism

⇒ Best suited for our need

↪ 122 μ s

⇒ We also made a REST API (using FastAPI) to serve Level-DB database content

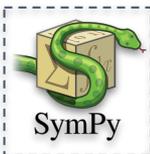
Expression Normalization



⇒ Tables are limited by the enumerative approach, combining some variables ($a, b, c..$) with some operators ($+, -, & \dots$). Thus no constants in sight. To improve expression diversity we performed two experiments.

Expression Linearization

Goal: Representing expressions as **normalized equations**. For that, uses SymPy a library for symbolic maths.



Original	Linearized
$a - (c - a)$	$2*a - c$
$(a-b) - (a + a)$	$-a - b$
$a + (b * b)$	$b^2 + a$
...	...



Pros/Cons:

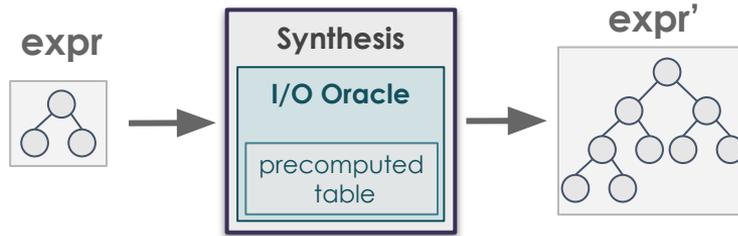
- introduces constants !
- **annihilates generation performances**
- introduces power operators
- only works on pure arithmetic expressions



we thus do not use it in practice

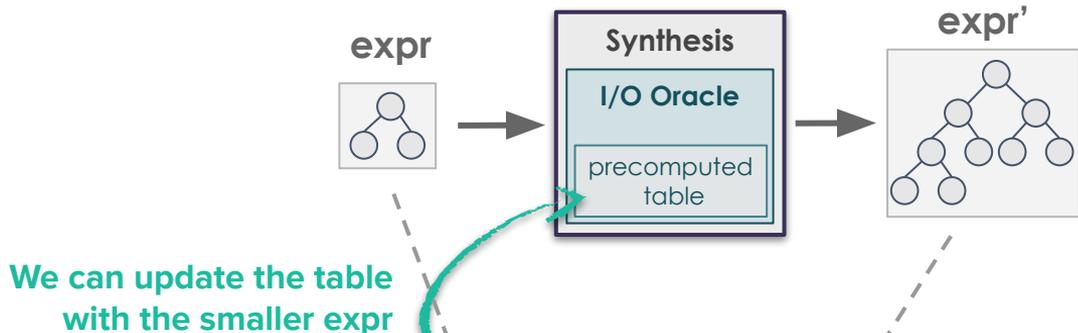
Problem

What if the synthesized expression is **larger** than the one in input ?



Problem

What if the synthesized expression is **larger** than the one in input ?



We can update the table with the smaller expr



Input Expr

$$(a * a) - 1$$

$$-1 + a$$

$$(b \wedge a) - 1$$

⇒

⇒

⇒

Output Expr' (in table)

$$\cancel{((a * a) + a) + (\sim a)}$$

$$\cancel{(\sim a + a) + (\sim(-a))}$$

$$\cancel{(\sim a + a) + (b \wedge a)}$$

⇒ We also now introduce simple constants in our table generation process

Benchmarks

Paper benchmarks



Comparison with Syntia

Simplification

	Mean expr. size			Simplification			Mean scale factor	
	Orig	Obf _B	Synt	∅	Partial	Full	Obf _S /Orig	Synt/Orig
Syntia	/	/	/	52	0	448	/	/
QSynth	3.97	203.19	3.71	0	500	500	x35.03	x0.94

Orig, Obf_S, Obf_B, Synt are rsp. original, obfuscated (source, binary level) and synthesized exprs

Accuracy & Speed

	Semantic	Time			
		Sym.Ex	Synthesis	Total	per fun.
Syntia	/	/	/	34 min	4.08s
QSynth	500	1m20s	15s	1m35s	0.19s

Against Tigress

Simplification

	Mean expr. size			Simplification			Mean Scale factor	
	Orig	Obf _B	Synt	∅	Partial	Full	Obf _S /Orig	Synt/Orig
Dataset 2 EA	13.5	245.81	21.92	0	500	354 (70.80%)	x18.34	x1.64
Dataset 3 VR-EA	13.5	443.64	25.42	0	500	375 (75.00%)	-	x1.90
Dataset 4 EA-ED	13.5	9223.46	3812.84	5	234	133 (55.65%)	x405.25	x234.44

Orig, Obf_S, Obf_B, Synt are respectively original, obfuscated (source, binary level) and synthesized expressions

Accuracy & Speed

	Semantic	Time			
		Sym.Ex	Synthesis	Total	per fun.
Dataset 2 EA	OK: 413 KO: 4	1m7s	1m42s	2m49s	0.34s
Dataset 3 VR-EA	OK: 401 KO: 43	17m10s	2m46s	19m56s	2.39s
Dataset 4 EA-ED	-	13m18s	2h7m	2h21m	35.47s

⇒ Results were promising !

Benchmarks improvements



	Algorithm Evolution	Mean size Synt Expr.	Simplification			Mean Scale factor			Time			
			∅	Partial	Full	Obf _S /Orig	Synt/Obf _B	Synt/Orig	Sym.Ex	Synthesis	Total	per fun.
Dataset 1 Syntia	Paper	3.71	0	500	500	x35.03	x0.02	x0.94	1m20s	15s	1m35s	0.19s
	New	3.71	0	500	500	x35.03	x0.01	x0.94	57s	6s	64.05s	0.13s
	Mul	3.71	0	500	500	x35.03	x0.02	x0.94	54s	4s	59.50s	0.12s
	Concat	3.71	0	500	500	x35.03	x0.02	x0.94	60s	4s	64.90s	0.13s
	LDB	3.71	0	500	500	x35.03	x0.02	x0.94	60s	4s	64.91s	0.13s
	370M	3.85	0	500	471	x35.03	x0.02	x0.97	61s	4s	65.73s	0.13s
Dataset 2 EA	Paper	21.92	0	500	354	x18.34	x0.17	x1.64	67s	1m42s	2m49s	0.34s
	New	19.93	0	500	324	x18.34	x0.12	x1.49	37s	26s	63.89s	0.13s
	Mul	19.48	1	499	324	x18.34	x0.15	x1.45	37s	23s	60.59s	0.12s
	Concat	19.48	1	499	324	x18.34	x0.15	x1.45	39s	23s	62.71s	0.13s
	LDB	19.48	1	499	324	x18.34	x0.15	x1.45	40s	17s	58.39s	0.12s
	370M	17.37	2	498	343	x18.34	x0.13	x1.30	39s	16s	55.94s	0.11s
Dataset 3 VR-EA	Paper	25.42	0	500	375	-	x0.06	x1.90	17m10s	2m46s	19m56s	2.39s
	New	75.14	14	486	296	-	x0.16	x5.61	11m55s	36s	12m31s	1.50s
	Mul	73.98	18	482	296	-	x0.19	x5.52	11m46s	35s	12m21s	1.48s
	Concat	21.50	0	500	324	-	x0.06	x1.60	12m2s	16s	12m18s	1.48s
	LDB	21.52	0	500	324	-	x0.06	x1.61	10m2s	8s	10m11s	1.61s
	370M	19.07	0	500	346	-	x0.05	x1.42	9m57s	9s	10m6s	1.21s
Dataset 4 EA-ED	Paper	3812.84	5	234	133	x405.25	x0.41	x234.44_	13m18s	2h7m	2h21m	35.47s
	New	483.26	0	239	133	x458.47	x0.03	x35.87_	9m22s	2h19m	2h28m	37.29s
	Mul	375.36	0	239	133	x458.47	x0.04	x27.86_	9m20s	1h34m	1h43m	26.01s
	Concat	375.36	0	239	133	x458.47	x0.04	x27.86_	9m15s	1h21m	1h30m	22.88s
	LDB	375.45	0	239	133	x458.47	x0.04	x27.87_	9m34s	1h16m	1h26m	21.64s
	370M	315.01	0	239	149	x458.47	x0.04	x23.38_	9m30s	1h21m	1h30m	22.79s

Benchmarks improvements



- Paper: Original results

Dataset	Algorithm Evolution	Mean size Synt Expr.	Simplification			Mean Scale factor			Time			
			∅	Partial	Full	Obf _S /Orig	Synt/Obf _B	Synt/Orig	Sym.Ex	Synthesis	Total	per fun.
Dataset 1 Syntia	Paper	3.71	0	500	500	x35.03	x0.02	x0.94	1m20s	15s	1m35s	0.19s
	New	3.71	0	500	500	x35.03	x0.01	x0.94	57s	6s	64.05s	0.13s
	Mul	3.71	0	500	500	x35.03	x0.02	x0.94	54s	4s	59.50s	0.12s
	Concat	3.71	0	500	500	x35.03	x0.02	x0.94	60s	4s	64.90s	0.13s
	LDB	3.71	0	500	500	x35.03	x0.02	x0.94	60s	4s	64.91s	0.13s
	370M	3.85	0	500	471	x35.03	x0.02	x0.97	61s	4s	65.73s	0.13s
Dataset 2 EA	Paper	21.92	0	500	354	x18.34	x0.17	x1.64	67s	1m42s	2m49s	0.34s
	New	19.93	0	500	324	x18.34	x0.12	x1.49	37s	26s	63.89s	0.13s
	Mul	19.48	1	499	324	x18.34	x0.15	x1.45	37s	23s	60.59s	0.12s
	Concat	19.48	1	499	324	x18.34	x0.15	x1.45	39s	23s	62.71s	0.13s
	LDB	19.48	1	499	324	x18.34	x0.15	x1.45	40s	17s	58.39s	0.12s
	370M	17.37	2	498	343	x18.34	x0.13	x1.30	39s	16s	55.94s	0.11s
Dataset 3 VR-EA	Paper	25.42	0	500	375	-	x0.06	x1.90	17m10s	2m46s	19m56s	2.39s
	New	75.14	14	486	296	-	x0.16	x5.61	11m55s	36s	12m31s	1.50s
	Mul	73.98	18	482	296	-	x0.19	x5.52	11m46s	35s	12m21s	1.48s
	Concat	21.50	0	500	324	-	x0.06	x1.60	12m2s	16s	12m18s	1.48s
	LDB	21.52	0	500	324	-	x0.06	x1.61	10m2s	8s	10m11s	1.61s
	370M	19.07	0	500	346	-	x0.05	x1.42	9m57s	9s	10m6s	1.21s
Dataset 4 EA-ED	Paper	3812.84	5	234	133	x405.25	x0.41	x234.44_	13m18s	2h7m	2h21m	35.47s
	New	483.26	0	239	133	x458.47	x0.03	x35.87_	9m22s	2h19m	2h28m	37.29s
	Mul	375.36	0	239	133	x458.47	x0.04	x27.86_	9m20s	1h34m	1h43m	26.01s
	Concat	375.36	0	239	133	x458.47	x0.04	x27.86_	9m15s	1h21m	1h30m	22.88s
	LDB	375.45	0	239	133	x458.47	x0.04	x27.87_	9m34s	1h16m	1h26m	21.64s
	370M	315.01	0	239	149	x458.47	x0.04	x23.38_	9m30s	1h21m	1h30m	22.79s

- Syntia: ED + EA (very simple)
- EA: EncodeArithmetic ⇒ MBA
- VR-EA: Virtualization + EA
- EA-ED: EA + EncodeData

Benchmarks improvements



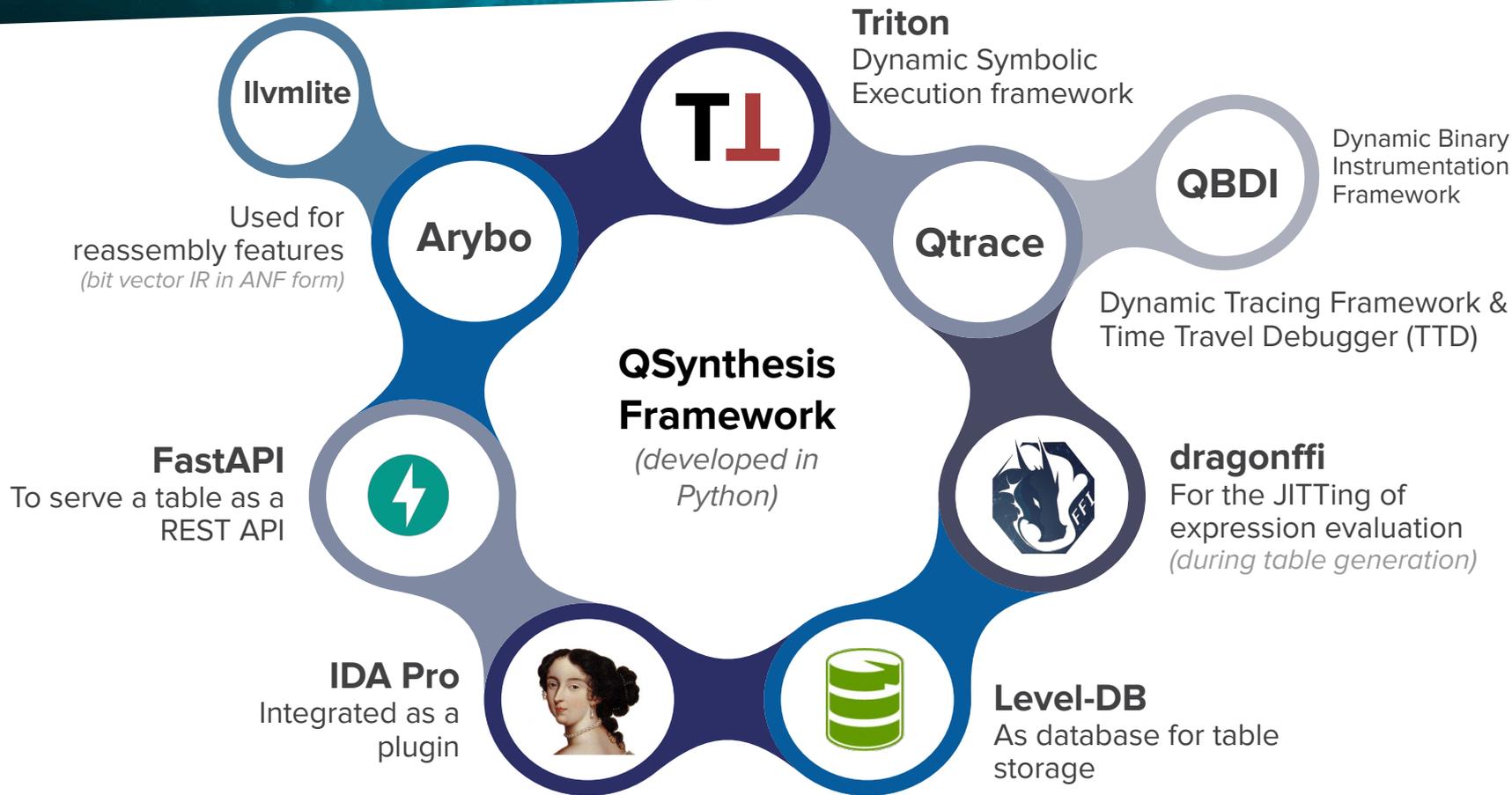
	Algorithm Evolution	Mean size Synt Expr.	Simplification			Mean Scale factor			Time			
			∅	Partial	Full	Obf _S /Orig	Synt/Obf _B	Synt/Orig	Sym.Ex	Synthesis	Total	per fun.
Dataset 1 Syntia	Paper	3.71	0	500	500	x35.03	x0.02	x0.94	1m20s	15s	1m35s	0.19s
	New	3.71	0	500	500	x35.03	x0.01	x0.94	57s	6s	64.05s	0.13s
	Mul	3.71	0	500	500	x35.03	x0.02	x0.94	54s	4s	59.50s	0.12s
	Concat	3.71	0	500	500	x35.03	x0.02	x0.94	60s	4s	64.90s	0.13s
	LDB	3.71	0	500	500	x35.03	x0.02	x0.94	60s	4s	64.91s	0.13s
	370M	3.85	0	500	471	x35.03	x0.02	x0.97	61s	4s	65.73s	0.13s
Dataset 2 EA	Paper	21.92	0	500	354	x18.34	x0.17	x1.64	67s	1m47s	2m49s	0.34s
	New	19.93	0	500	324	x18.34	x0.12	x1.49	37s	26s	63.89s	0.13s
	Concat	19.48	1	499	324	x18.34	x0	x0	-	23s	60.59s	0.12s
	LDB	19.48	1	499	324	x18.34	x0	x0	-	23s	62.71s	0.13s
	Mul	19.48	1	499	324	x18.34	x0	x0	-	17s	58.39s	0.12s
	370M	17.37	2	498	343	x18.34	x0	x0	-	16s	55.94s	0.11s
	Paper	25.42	0	500	375	-	x0	x0	-	1m46s	19m56s	2.39s
	New	75.14	14	486	296	-	x0	x0	-	36s	12m31s	1.50s
	Mul	73.98	18	482	296	-	x0.19	x5.52	11m46s	35s	12m21s	1.48s
	Concat	21.50	0	500	324	-	x0.06	x1.60	12m2s	16s	12m18s	1.48s
Dataset 3 EA	LDB	21.52	0	500	324	-	x0.06	x1.61	10m2s	8s	10m11s	1.51s
	370M	19.07	0	500	346	-	x0.05	x1.42	9m57s	9s	10m6s	1.21s
Dataset 4 EA-ED	Paper	3812.84	5	234	133	x405.25	x0.41	x234.44_	13m18s	2h7m	2h21m	35.47s
	New	483.26	0	239	133	x458.47	x0.03	x35.87_	9m22s	2h19m	2h28m	37.29s
	Mul	375.36	0	239	133	x458.47	x0.04	x27.86_	9m20s	1h34m	1h43m	26.01s
	Concat	375.36	0	239	133	x458.47	x0.04	x27.86_	9m15s	1h21m	1h30m	22.88s
	LDB	375.45	0	239	133	x458.47	x0.04	x27.87_	9m34s	1h16m	1h26m	21.64s
	370M	315.01	0	239	149	x458.47	x0.04	x23.38_	9m30s	1h21m	1h30m	22.79s

Better average simplification than original implementation (90% for EA-ED)

Speed improvement ranging from 31% to 67%

Implementation

(in the QSynthesis utility)



IDA Integration



The screenshot displays the IDA Pro interface with the QSynthesis plugin active. The main window shows assembly code for a function named 'target_77'. A 'Reassembly options' dialog box is open, with the following settings:

- patch function bytes
- shrink function
- move some instruction instead of filling with NOPS.
- Can break disassembly for relative instructions. (Works only for linear blocks)
- Snapshot database before patching

The 'QSynthesis' configuration window is also visible, showing the following settings:

- From: 0x4011a2 To: 0x40128a Target: REG RAX
- Table: LEVELDB /tmp/lts/final_table_leveldb
- Algorithm: Top-Down Type: FULL_SYMBOLIC FULL_SYMBOLIC

The 'Run Synthesis' window displays the following statistics:

Node count	Depth	Scale
124	12	-92.74%

The 'Simplified: Yes Synthesized Expression' is shown as: $((((rcx + rdi) \& rdi) \wedge ((rdx + rdi)))$. The 'Inputs' section lists: rdi 64, rcx 64, rdx 64.

The 'Output window' at the bottom shows the following text:

```
Python
Python 3.9.1+ (default, Feb 5 2021, 13:46:56)
[GCC 10.2.1 20210110]
IDAPython 64-bit v7.4.0 final (serial 0) (c) The IDAPython Team <idapython@googlegroups.com>
Propagating type information...
Function argument information has been propagated
lumina: applied metadata to 3 functions.
The initial autoanalysis has been finished.
Running @Synthesis
```

Use-Cases

(getting our hands dirty!)

Attacking YANSOllvm



Transforms:

- **VM:** transforms basic operators (+, ⊕..) with function calls
- **Merge:** merges all internal linkage functions in a single one
- **Flattening:** CFG flattening
- **Connect:** splits basic blocks and uses switch to add false branches
- **ObfCon:** obfuscates constants with MBAs
- **BB2func:** splits & extracts basic blocks in new functions
- **ObfCall:** changes internal linkage function calling convention

master 1 branch 0 tags Go to file Code

emc2314 Update README.md 1001331 on Jun 20, 2020 65 commits

README.md

YANSOllvm

Yet Another Not So Obfuscated LLVM

LLVM Version

Based on the release version 9.0.1. Other version might work as well, but one has to merge/rebase the X86 related code.

Build

```
wget https://github.com/llvm/llvm-project/releases/download/llvmorg-9.0.1/llvm-9.0.1.src.tar.xz
tar xf llvm-9.0.1.src.tar.xz && cd llvm-9.0.1.src
git init
git remote add origin https://github.com/emc2314/YANSOllvm.git
```

About
Yet Another Not So Obfuscated LLVM
obfuscation llvm
Readme
GPL-3.0 License

Releases
No releases published

Packages
No packages published

Languages
C++ 94.1% Python 3.2%
CMake 2.4% C 0.3%

<https://github.com/emc2314/YANSOllvm>

⇒ There are plenty of other Obfuscator-LLVM derivatives used in the wild

YANSOLLvm: VM obfuscation



```
var_8= dword ptr -8
var_2= word ptr -2

; __unwind {
push rbp
mov rbp, rsp
sub rsp, 10h
mov [rbp+var_8], edi
cmp [rbp+var_8], 0
jnz short loc_40128E

call d
movsx eax, word ptr [rax+2]
mov edi, eax
mov eax, 0Ch
mov esi, eax
call __YANSOLLVM_VM_Xor
mov [rbp+var_2], ax
jmp short loc_4012B7

loc_40128E:
mov edi, [rbp+var_8]
mov eax, 1
mov esi, eax
call __YANSOLLVM_VM_Sub
mov edi, eax
call c
cwde
mov edi, eax
mov eax, 1
mov esi, eax
call __YANSOLLVM_VM_Add
mov [rbp+var_2], ax
```

```
__YANSOLLVM_VM_Add proc near
; __unwind {
mov rax, rsi
xor rax, 0FFFFFFFFFFFFFFFh
or rax, rdi
mov rcx, rdi
xor rcx, 0FFFFFFFFFFFFFFFh
and rcx, rsi
mov rdx, rdi
and rdx, rsi
xor rdx, 0FFFFFFFFFFFFFFFh
or rdi, rsi
add rax, rcx
sub rax, rdx
add rax, rdi
retn
; } // starts at 4012E0
__YANSOLLVM_VM_Add endp
```

Synthesized and reassembled to

```
lea rax, [rsi+rdi]
ret
```

⇒ We then could go further by removing calls and replacing them by the operation directly



OpaqueConstant

- $((\sim x \mid 0x7AFafa69) \& 0xA061440) + ((x \& 0x1050504) \mid 0x1010104) == 185013572$
- $p1*(x \mid any)**2 \neq p2*(y \mid any)**2$
- $x + y = x \hat{y} + 2*(x \& y)$
- $x \hat{y} = (x \mid \sim y) - 3*(\sim(x \mid y)) + 2*(\sim x) - y$

MBAs

$x + y$	$(x \mid \sim y) + (\sim x \& y) - (\sim(x \& y)) + (x \mid y)$
$x - y$	$x + \sim y + 1$
$x \ll y$	/
$x > a \ y$	/
$x > \bar{l} \ y$	/
$x \& y$	$-(\sim(x \& y)) + (\sim x \mid y) + (x \& \sim y)$
$x \mid y$	$(x \hat{y}) + y - (\sim x \& y)$
$x \wedge y$	$x + y - ((x \& y) \ll 1)$

About MBA & constants:

expression using constants: $a \& 0xdeadbeef \Rightarrow \text{✗ tables do not contains constants}$

constants: $0xd00dfeed \Rightarrow \text{✓ can synthesize it !}$

Example: Opaque Constant



```
push    rbp
mov     rbp, rsp
mov     edx, edi
not     edx
mov     eax, edx
or      eax, 0A021040h
and     eax, 0A061440h
mov     ecx, edi
and     ecx, 40400h
lea     eax, [rcx+rax+1010104h]
mov     r9d, eax
xor     r9d, 0B071544h
mov     esi, r9d
or      esi, edx
mov     edx, r9d
or      edx, edi
not     edx
lea     r8d, [rdx+rdx*2]
mov     eax, edx
xor     edx, 74F8EABh
add     edx, edx
sub     edi, ecx
xor     ecx, edi
sub     ecx, esi
add     ecx, r8d
xor     ecx, edx
mov     edx, ecx
add     edx, 9054CB9h
xor     eax, edx
xor     eax, 20259FCh
lea     rax, [rax+rax+0Fh]
and     rax, 0FFFFFFFFFFFF0h
mov     r8, rsp
sub     r8, rax
mov     rsp, r8
or      ecx, 0EEh
movzx  eax, cl
```

blackbox I/O optimization

If the evaluation of all inputs produces the same output, thus the expression encodes a constant.

```
imul   eax, eax
imul   esi, eax, 37F1h
xor     edx, edx
mov     eax, esi
sub     eax, 203D2640h
setz   dl
neg     edx
xor     edx, 0F88BA899h
mov     r10d, esi
or      r10d, 0DFC2D9BFh
mov     eax, esi
not     eax
mov     ecx, eax
and     ecx, 0DFC2D9BFh
lea     ecx, [rcx+rcx*2]
lea     eax, [rax+rax-203D2640h]
xor     esi, 203D2640h
sub     esi, r10d
add     esi, ecx
xor     esi, eax
mov     eax, esi
xor     eax, 0BDC2BA9h
imul   edx, eax
lea     rcx, ds:0Fh[rdx*8]
and     rcx, 0FFFFFFFFFFFF0h
mov     rax, rsp
sub     rax, rcx
mov     rsp, rax
mov     [rax], rdi
mov     r10, [rax]
mov     ecx, esi
add     ecx, r9d
mov     edx, esi
xor     edx, r9d
sub     ecx, edx
and     esi, r9d
shl    esi, 1
xor     ecx, esi
mov     ecx, ecx
cmp     r10, rcx
setz   cl
test   cl, 1
jnz    short loc_401D60
```

Value synthesized

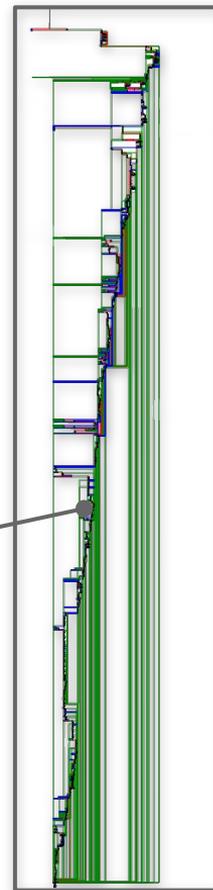
⇒ 0x0

Windows Warbird



⇒ Part of the Windows kernel is known to be obfuscated with a framework called **Warbird**. More specifically **PatchGuard** features are obfuscated. We gave a very quick look at the PatchGuardInit function.

```
loc_140A3AF54:  
rdtsc  
shl     rdx, 20h  
mov     rdi, 7010008004002001h  
or      rax, rdx  
mov     r12d, 5  
mov     rcx, rax  
ror     rax, 3  
xor     rcx, rax  
mov     rax, rdi  
mul     rcx  
mov     rcx, rdx  
mov     [rsp+24B8h+var_1858], rdx  
xor     rcx, rax  
mov     rax, 2E8BA2E8BA2E8BA3h  
mul     rcx  
shr     rdx, 1  
imul   rax, rdx, 0Bh  
sub     rcx, rax  
cmp     ecx, r12d  
ja      loc_140A3B062
```



*thanks Damien for pinpointing me that function

Windows Warbird



The screenshot displays four windows from IDA Pro:

- IDA View-A:** Shows assembly code for a function named `rdtsc`. The code includes instructions like `shl rdx, 20h`, `mov r8, 7010008004002001h`, `or rax, rdx`, `mov ebx, 5`, `mov rcx, rax`, `ror rax, 3`, `xor rcx, rax`, `mov rax, r8`, `mul rcx`, `mov rcx, rdx`, `mov [rsp+24B8h+var_1AD0], rdx`, `xor rcx, rax`, `mov rax, 2E8BA2E8BA2E8BA3h`, `mul rcx`, `shr rdx, 1`, `imul rax, rdx, 0Bh`, `sub rcx, rax`, `cmp ecx, ebx`, and `ja loc_140A2B288`.
- Reassembly:** Shows the deobfuscated assembly code: `shl rdx, 0x20`, `or rdx, rax`, `mov rax, rdx`, `rol rax, 0x3d`, `xor rax, rdx`, `movabs rcx, 0x7010008004002001`, and `imul rcx, rax`.
- Triton AST:** A complex tree diagram representing the original assembly code's structure.
- Synthesized AST:** A simplified tree diagram representing the deobfuscated assembly code's structure.

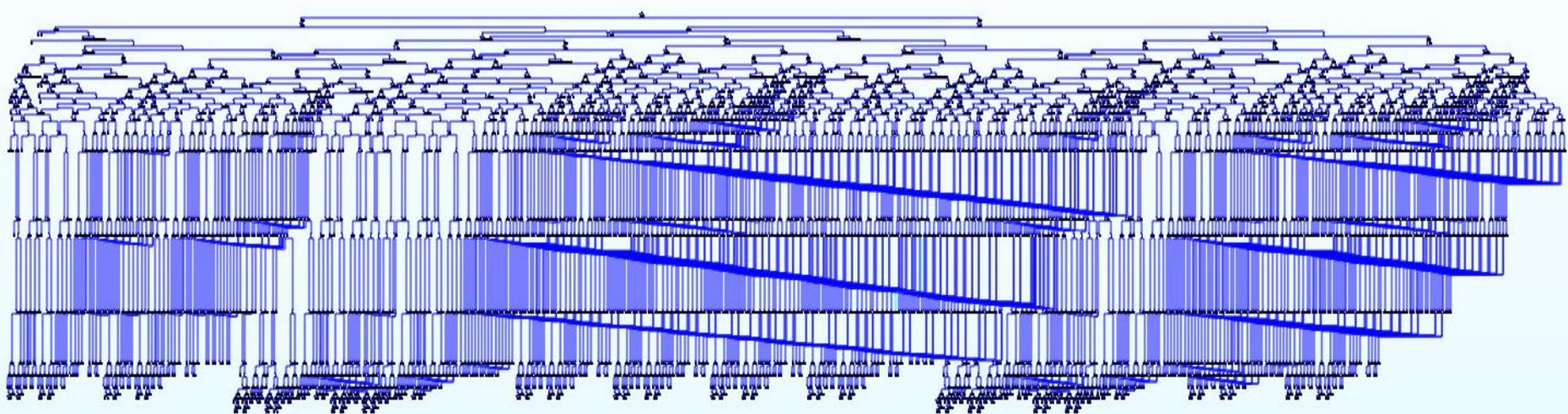
⇒ Deobfuscating it would require a deeper understanding of the function **and more time!**

(more detailed analyses of Warbird [here](#) & [here](#))

Messaging Application



Contains beautiful MBAs



Messaging Application



The screenshot shows the QSynthesis tool interface. On the left, there is a list of assembly instructions for a function named `loc_9B7AEC`. The instructions include `MOV`, `SMULH`, `MOV`, `ASR`, `MOVK`, `ADD`, `MSUB`, `MOV`, `AND`, `MOVK`, `ADD`, `MOVK`, `SUB`, `MOVK`, `MOV`, `MOV`, `SMULH`, `MOVK`, `MOV`, `SMULH`, `ASR`, `MOVK`, `MOV`, `MOVK`, `ASR`, `ADD`, `MOVK`, `MOVK`, `ADD`, `MUL`, `SMULH`, `SMULH`, `ASR`, `ADD`, `ADD`, `MOV`, `SUB`, `MOVK`, `ADD`, `STR`, `MUL`, `SUB`, `AND`, `SUB`, `SMULH`, `MOV`, `ADD`, `MOVK`, `ASR`, `MOVK`, `ADD`, `MOV`, `MOVK`, `MOV`, `ADD`, `MOV`, `ADD`, `MOV`, `B`.

On the right, the QSynthesis configuration panel is shown. The synthesis configuration includes:

- From: `0x9b7aec` To: `0x9b7bfc` Target: `REG` X20
- Table: `LEVELDB` /home/robin/Quarkslab/synthesis/its/final_table_leveldb
- Algorithm: `Top-Down` Type: `FULL_SYMBOLIC` `FULL_SYMBOLIC`

Buttons for `Run Triton` and `Run Synthesis` are visible. The synthesis results are:

Node count	Depth
5093	37

Inputs: `x24` `64`

Simplified: `Yes`
Synthesized Expression: `0x7B2`

Node count	Depth	Scale
1	1	-99.98%

⇒ We managed to synthesize many MBAs (but as usual it is mixed with other transformations and we do not really know what we are synthesizing)

Conclusion



Greybox algorithm

The greybox algorithm strongly **reduces** the need for huge tables and enable opportunistically **synthesizing sub-expressions**

*(thus tables shall be more **representative** than exhaustive introducing constants etc)*

Next plans

- Breaking MBA using constants *(we have ideas on mechanisms that can be integrated within the synthesis algorithm but with some ad-hoc checks)*
- Restoring original simplification algorithm potency *(by fixing some complexity induced by Triton)*



- Breaking the obfuscation is crucial as it is the first step before further reversing
- Synthesis only help on a sub-part of the deobfuscation process:
 - it addresses PB#2: deobfuscating a data-flow expression
 - but **do not** addresses PB#1: **locating the data** to deobfuscate
- We do use these techniques to **assess** and continuously **improve** the strength of our own obfuscator (*Quarks AppShield*)
- (*As usual*) what makes obfuscation potent is **carefully mixing** obfuscation passes

Acknowledgement



- **Luigi Coniglio** how kickstarted that approach in our dynamic tracing framework
Qtrace [↗](#)
- **Jonathan Salwan** that tweaked Triton to make it more efficient on this kind of use-cases
- My Quarkslab's colleagues, and people of the synthesis community with whom I had stimulating discussions

Thank you !

Q & A



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