

Blockchain Superoptimizer

Julian Nagele **Maria A. Schett**

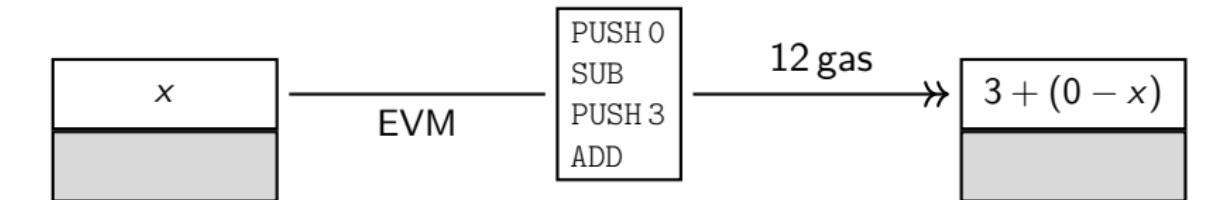
Overview

```
PUSH 0  
SUB  
PUSH 3  
ADD
```

- ▶ Ethereum smart contracts are executed as bytecode on the Ethereum Virtual Machine (EVM)



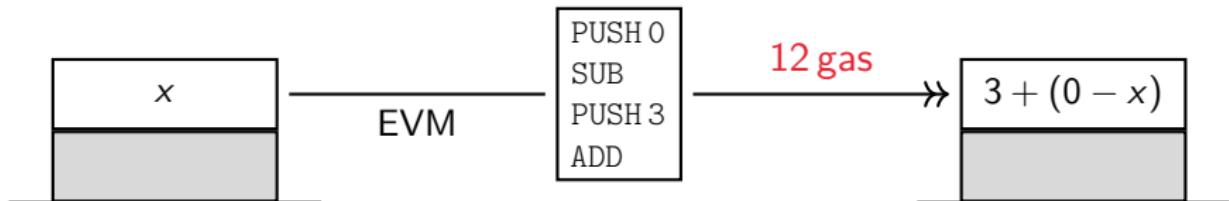
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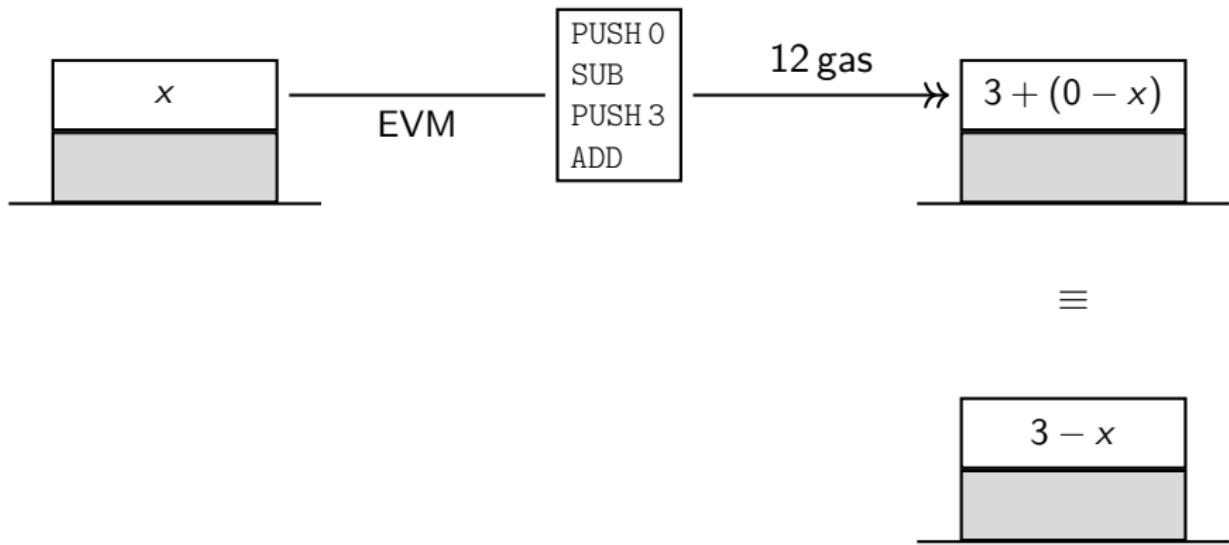
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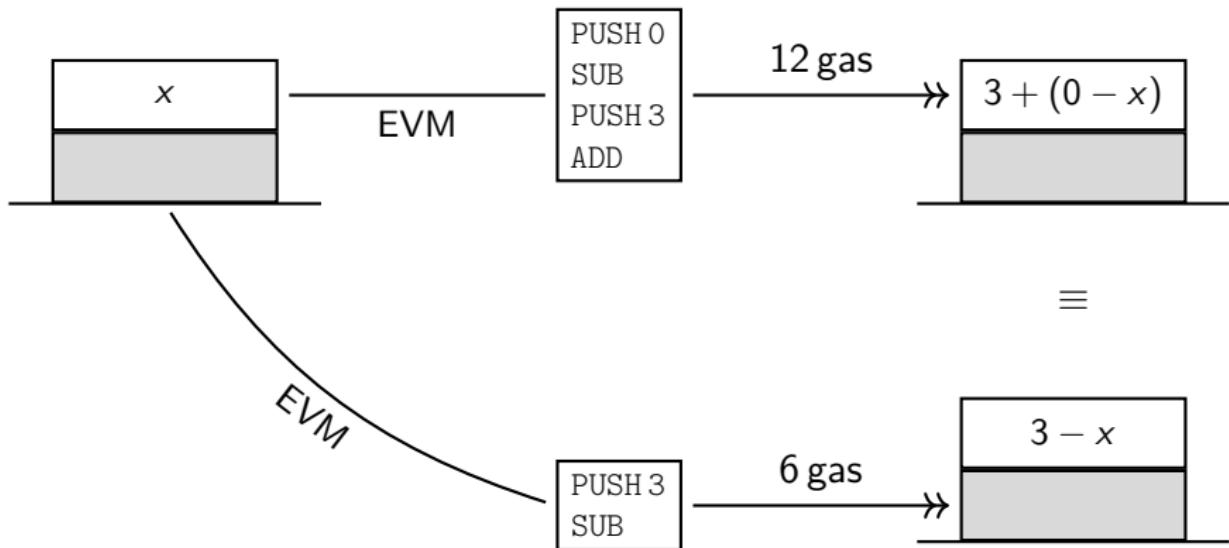
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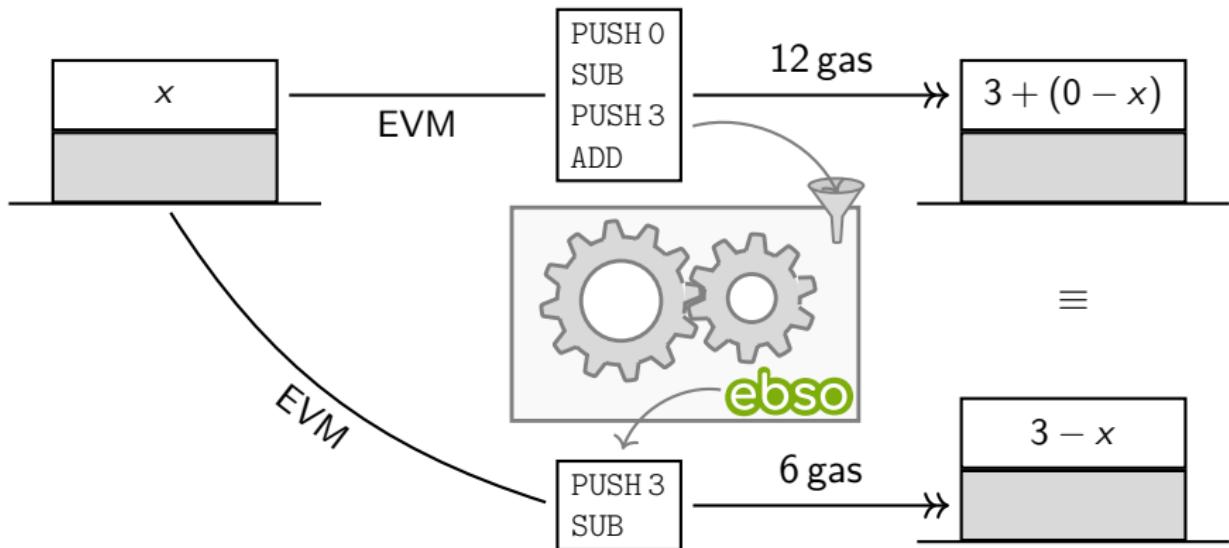
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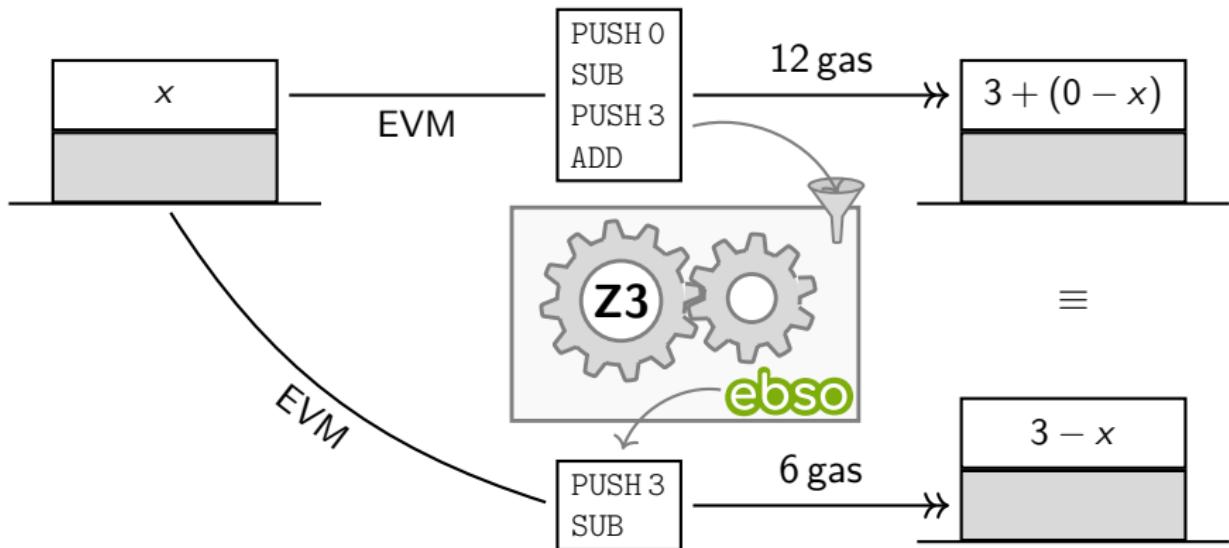


Overview



- ▶ our tool **ebso** superoptimizes EVM bytecode

Overview



Interface

```
$ ./ebs0 -direct "600003600301"
Optimized PUSH 0 SUB PUSH 3 ADD to
PUSH 3 SUB
Saved 6 gas,
this instruction sequence is optimal.
```

Superoptimization

Superoptimization [Massalin 1987]

- ▶ **given:** source program p_s & cost function C
- ▶ **find:** target program p_t that
 1. correctly implements p_s
 2. has minimal cost
- ▶ **flavors:** basic and unbounded superoptimization

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Ethereum

- ▶ formal semantics [Yellow Paper, 2018]
- ▶ EVM gas provides **clear cost model**
- ▶ \exists data sets for evaluation

Basic Superoptimization (1)

```
1: function BASICSO( $p_s, C$ )
2:    $n \leftarrow 0$ 
3:   while true do
4:     for all  $p_t \in \{p \mid C(p) = n\}$  do
5:        $\chi \leftarrow \text{ENCBSO}(p_s, p_t)$ 
6:       if not SATISFIABLE( $\chi$ ) then
7:         return  $p_t$ 
8:        $n \leftarrow n + 1$ 
```

ENCBSO: $\exists \vec{x}$ to distinguish p_s & p_t ?

Templates [Gulwani et al. 2011]

- ▶ PUSH w where w is a 256 bit word $\implies 2^{256}$ candidates

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Example

PUSH 0 SUB PUSH 3 ADD

$$a(0) = 0 \quad a(2) = 3 \quad a(j) = \underline{}$$

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Example

PUSH 0 SUB PUSH 3 ADD

$$a(0) = 0 \quad a(2) = 3 \quad a(j) = \underline{\hspace{2cm}}$$

ENCBSO: $\exists(a : \text{position} \rightarrow \text{word}) \forall \vec{x} \text{ s.t. } p_t \text{ implements } p_s?$

Basic Superoptimization (2)

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7:          $m \leftarrow \text{GETMODEL}(\chi)$ 
8:          $p_t \leftarrow \text{DECBSO}(m)$ 
9:       return  $p_t$ 
10:       $n \leftarrow n + 1$ 
```

Unbounded Superoptimization [Jangda&Yorsh 2017]

ENCUso: $\exists(\text{instr} : \text{position} \rightarrow \text{instruction}) \forall \vec{x} \text{ s.t. } p_t \text{ implements } p_s?$

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```
1: function UNBOUNDEDSo( $p_s, C$ )
2:    $p_t \leftarrow p_s$ 
3:    $\chi \leftarrow \text{ENCUSO}(p_t) \wedge \text{BOUND}(p_t, C)$ 
4:   while SATISFIABLE( $\chi$ ) do
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- ▶ can stop early with non-optimal solution

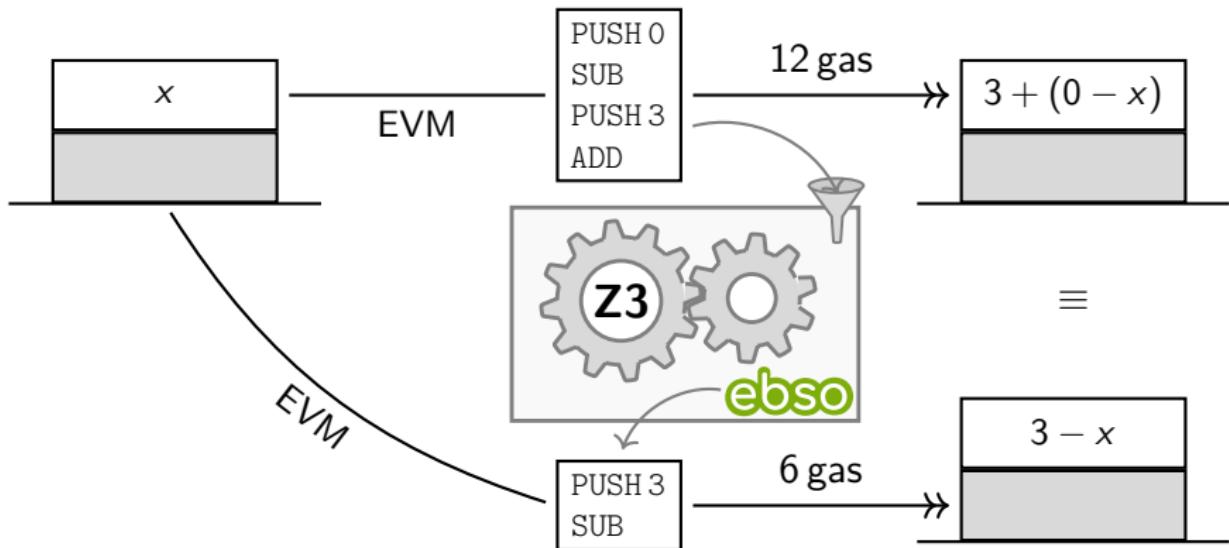
SMT Encoding

Encoding

Satisfiability Modulo Theories

- ▶ first-order logic with background theories
 - ★ bit vectors, integers, uninterpreted functions, arrays, ...
- ▶ \exists powerful off-the-shelf solvers
- ▶ ■

Overview



Ingredients

- ▶ state, i.e., stack, used gas, ...
- ▶ \rightarrow_E^P , i.e., operational semantics of EVM
- ▶ \equiv , i.e, equality on states

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- ▶ \equiv , i.e, equality on states
- ▶ for UNBOUNDEDSo: encoding of search space

State

state $\sigma = \langle \text{st}, \text{c}, \text{g} \rangle$ consists of

- ▶ function $\text{st}(\vec{x}, j, n)$: n -th word on stack after j instructions on input \vec{x}
- ▶ function $\text{c}(j)$: number of words on stack after j instructions
- ▶ function $\text{g}(j)$: amount of gas consumed by first j instructions.

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$\text{st}(\vec{x}, j, \text{c}(j) - 1)$ returns top of stack $\sim \top$

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Example

PUSH 41 PUSH 1 ADD

$$j = 2 : \quad \text{st}(j, 0) = 41 \quad \text{st}(j, 1) = 1 \quad \text{c}(j) = 2 \quad \text{g}(j) = 6$$

Instructions

semantics of instruction ι

$$\tau_{\text{st}}(\text{ADD}, \sigma, j) \equiv \text{st}_\sigma(j + 1, \top) = \text{st}_\sigma(j, \top) +_{bv} \text{st}_\sigma(j, \top - 1)$$

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$$\tau_{\text{st}}(\text{SWAP2}, \sigma, j) \equiv \text{st}_\sigma(j + 1, \top) = \text{st}_\sigma(j, \top - 2)$$

$$\wedge \text{st}_\sigma(j + 1, \top - 1) = \text{st}_\sigma(j, \top - 1)$$

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$$\tau_c(\iota, \sigma, j) \equiv c_\sigma(j + 1) = c_\sigma(j) + \alpha(\iota) - \delta(\iota)$$

Instructions

number of elements added (α) to and deleted (δ) from stack

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for program $p = \iota_0, \dots, \iota_n$ we define

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Equality

- equality of states after j_1 and $j_2 \rightarrow_{\text{EVM}}$ steps

$$\begin{aligned}\epsilon(\vec{x}, \sigma_1, \sigma_2, j_1, j_2) \equiv & c_{\sigma_1}(j_1) = c_{\sigma_2}(j_2) \\ \wedge \forall n < c_{\sigma_1}(j_1). \text{st}_{\sigma_1}(\vec{x}, j_1, n) = & \text{st}_{\sigma_2}(\vec{x}, j_2, n)\end{aligned}$$

Superoptimization

Basic

$$\begin{aligned}\text{ENCBSO}(p_s, p_t) \equiv & \forall \vec{x}. \tau(p_s, \vec{x}, \sigma) \wedge \tau(p_t, \vec{x}, \sigma') \\ & \wedge \epsilon(\vec{x}, \sigma, \sigma', 0, 0) \wedge \epsilon(\vec{x}, \sigma, \sigma', |p_s|, |p_t|)\end{aligned}$$

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Unbounded

Superoptimization

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Unbounded

$$\text{ENCUSO}(p) = \forall \vec{x}. \tau(p, \vec{x}, \sigma) \wedge \epsilon(\vec{x}, \sigma, \sigma', 0, 0) \wedge \epsilon(\vec{x}, \sigma, \sigma', |p|, n)$$

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Uninterpreted Instructions

- ▶ BLOCKHASH, ADDRESS, BALANCE, ... arbitrary, but fixed, result

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- ▶ **idea:** add \forall -quantified variables u_ℓ
- ▶ e.g. $\tau_{\text{st}}(\text{ADDRESS}, \sigma, j) \equiv \text{st}_\sigma(j + 1, \top) = u_{\text{ADDRESS}}$

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Non-constant

- ▶ **idea:** add f_ι and \forall -quantified variables u_{ι_j} for j occurrences of ι
- ▶ initialize f_ι as read-only memory with arguments from stack

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+ Storage

Initialize

- ▶ BLOCKHASH at positions j_1, \dots, j_ℓ
- ▶ variables u_1, \dots, u_ℓ for result
- ▶ arguments $a_{j_i} = \text{st}_\sigma(j_i, \top)$

$$\begin{aligned}\forall w. f_{\text{BLOCKHASH}}(w) = & \text{ite}(w = a_{j_1}, u_1, \\ & \text{ite}(w = a_{j_2}, u_2, \\ & \dots \\ & \text{ite}(w = a_{j_\ell}, u_\ell, \underline{})))\end{aligned}$$

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Implementation

Implementation

- ▶ available at

github.com/juliannagele/ebso

- ▶ implemented in OCaml
- ▶ ~1.6 kloc, 635 tests
- ▶ using Z3 as SMT solver



Translation Validation

- ▶ **given:** large word size of EVM 256 bit \implies scalability problems
- ▶ **solution:** find model for small word size & validate for 256 bit

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$$\begin{aligned}\text{TRANSVAL}(p_s, p_t) = \exists \vec{x}. \tau(p_s, \vec{x}, \sigma) \wedge \tau(p_t, \vec{x}, \sigma') \\ \wedge \epsilon(\vec{x}, \sigma, \sigma', 0, 0) \wedge \neg\epsilon(\vec{x}, \sigma, \sigma', |p_s|, |p_t|)\end{aligned}$$

Evaluation

EvaluationS

1. “Optimize the Optimized”

- ★ Gas Golfing Contest: 199 Solidity contracts
- ★ ⇒ 2743 ebso blocks

2. “BasicSo vs. UnboundedSo”

- ★ bytecode of 2500 most called contracts from Ethereum Blockchain
- ★ ⇒ 61217 ebso blocks

- ▶ 60 min/15 min time-out on 1 core at 2.40 GHz with 1 GiB RAM
- ▶ validation with pseudo-random input on go-ethereum EVM

Optimize The Optimized (1)

	#	%
(O) optimized (optimal)		
(P) proved optimal		
(T) time-out (transl. val. fail)		

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	#	%
(O) optimized (optimal)	19 (10)	0.69 %
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Optimize The Optimized (1)

	#	%
(O) optimized (optimal)	19 (10)	0.69 %
(P) proved optimal	481	17.54 %
(T) time-out (transl. val. fail)		

Optimize The Optimized (1)

	#	%
(O) optimized (optimal)	19 (10)	0.69 %
(P) proved optimal	481	17.54 %
(T) time-out (transl. val. fail)	2243 (196)	81.77 %

Optimize The Optimized (2)

- (12) CALLVALUE DUP1 ISZERO PUSH 81 and
 - (13) CALLVALUE DUP1 ISZERO PUSH 298
- to CALLVALUE CALLVALUE ISZERO PUSH _

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(17) POP PUSH 1 SWAP1 POP PUSH 0

to SLT DUP1 EQ PUSH 0

BASICSo vs. UNBOUNDEDSo (1)

UNBOUNDEDSo	#	%	BASICSo	#	%
(O)					
(P)					
(T)					

- ▶ (O) optimized (optimal)
- ▶ (P) proved optimal
- ▶ (T) time-out (translation validation failed)

BASICSo vs. UNBOUNDEDSo (1)

	UNBOUNDEDSo		BASICSo	
	#	%	#	%
(O)	943 (393)	1.54 % (0.64 %)	184	0.3 %
(P)				
(T)				

- ▶ (O) optimized (optimal)
- ▶ (P) proved optimal
- ▶ (T) time-out (translation validation failed)

BASICSo vs. UNBOUNDEDSo (1)

	UNBOUNDEDSo		BASICSo	
	#	%	#	%
(O)	943 (393)	1.54 % (0.64 %)	184	0.3 %
(P)	3882	6.34 %	348	0.57 %
(T)				

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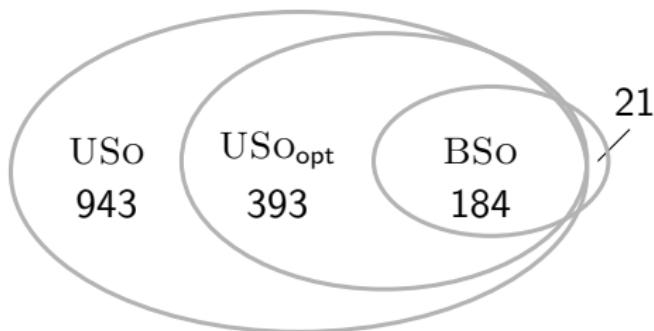
BASICSo vs. UNBOUNDEDSo (1)

	UNBOUNDEDSo		BASICSo	
	#	%	#	%
(O)	943 (393)	1.54 % (0.64 %)	184	0.3 %
(P)	3882	6.34 %	348	0.57 %
(T)	56 392 (1467)	92.12 % (2.4 %)	60 685	99.13 %

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- ▶ (P) proved optimal
- ▶ (T) time-out (translation validation failed)

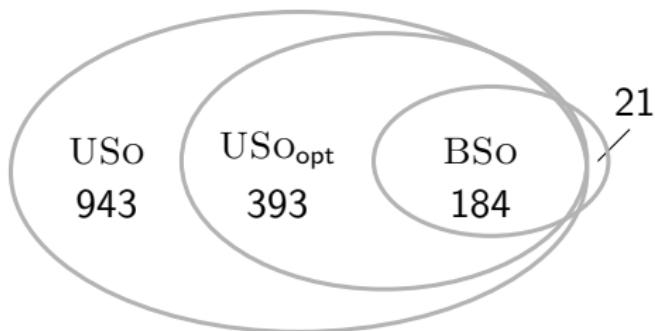
BASICSo vs. UNBOUNDEDSo (2)

- ▶ optimizations for basic (BSo) and unbounded (USo) superoptimization



BASICSo vs. UNBOUNDEDSo (2)

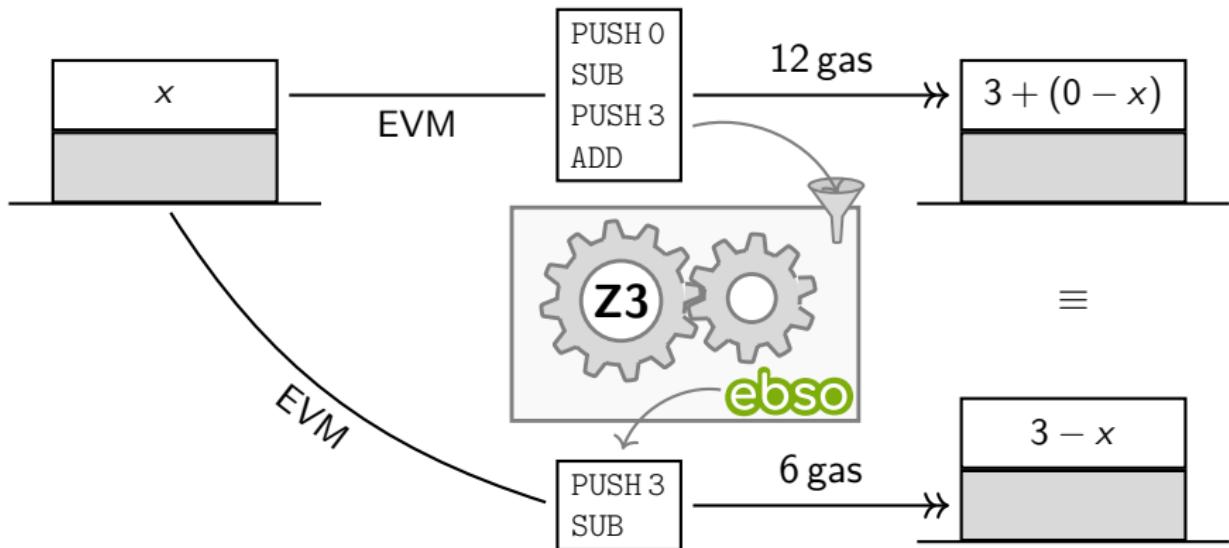
- ▶ optimizations for basic (BSo) and unbounded (USo) superoptimization



- ▶ PUSH 0 PUSH 12 SLOAD LT ISZERO ISZERO ISZERO
PUSH 12250 to PUSH 1 PUSH 12250

Conclusion

Overview



Future

Application

- ▶ generalize patterns to rewrite
- ▶ beyond straight-line code with control flow analysis
- ▶ eWasm

Future

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- ▶ generalize patterns to rewrite
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SMT

- ▶ extract benchmarks
- ▶ strategies to guide solver(s)

Thank you & questions?



available at github.com/juliannagele/ebso

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Bibliography



H. Massalin

Superoptimizer: A Look at the Smallest Program

ASPLOS II, 1987



S. Gulwani, S. Jha, A. Tiwari, and R. Venkatesan

Synthesis of Loop-free Programs

Proc. PLDI 2011



A. Jangda and G. Yorsh

Unbounded Superoptimization

Proc. Onward! 2017



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