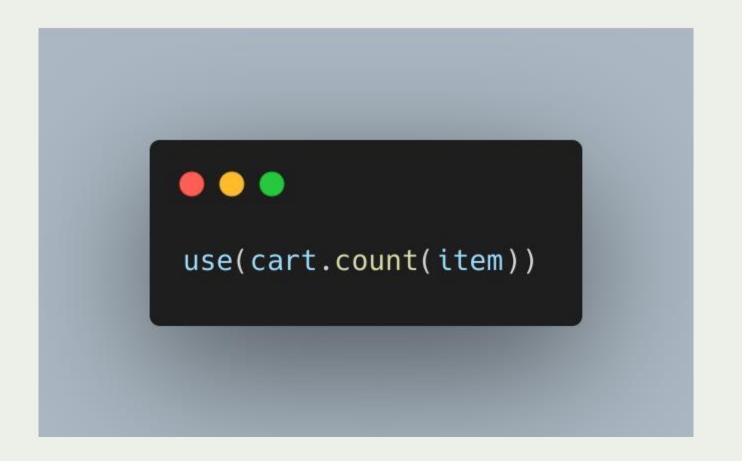
SEMANTIC CODE SEARCH VIA EQUATIONAL REASONING

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```
• • •
count = 0
for a in cart:
    if a == item
        count += 1
use (count)
```

```
• • •
count = 0
for a in cart:
    if a == item
        count += 1
use (count)
```



```
count = 0
for i in cart:
    if debug:
       print(cart[i])
    if cart[i] == item:
       count += 1
use(count)
```

```
count = 0
for i in range(len(arr)):
    if item != arr[i]:
     continue
    count += 1
use(count)
```

```
• • •
count = 0
i = 0
while i < len(cart):</pre>
  if cart[i] == k:
    count += 1
   i += 1
use(count)
```



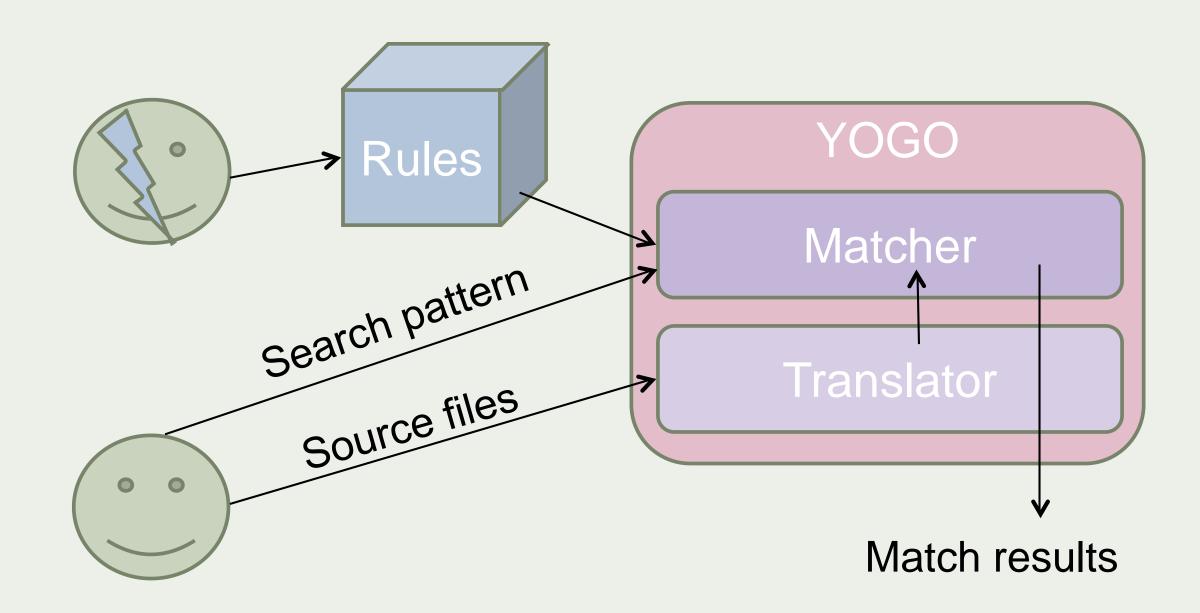
Semantic code search

Scope

Query type

Performance needs

You Only Grep Once



Operational evaluation

Search patterns gleaned from common StackOverflow questions

YOGO times out on 1% of the methods searched

False negatives due to tool incompleteness and reasoning

limitations

Graal

13 static analyzers on top of Graal's bytecode analysis

YOGO analyzed 1.2 million LOC in 2.5 hours on 30 AWS instances

Generalizing one of the analyzers' queries revealed a bug it

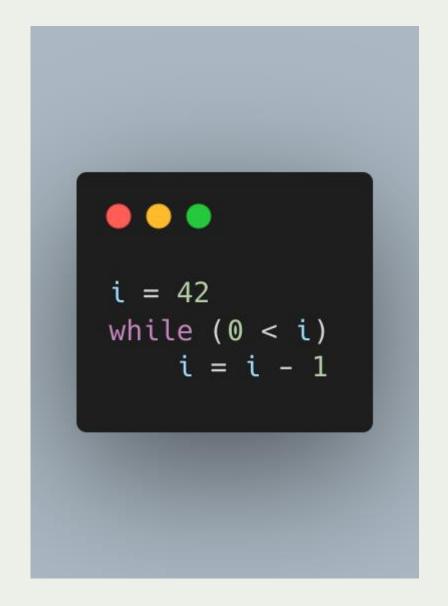
missed!

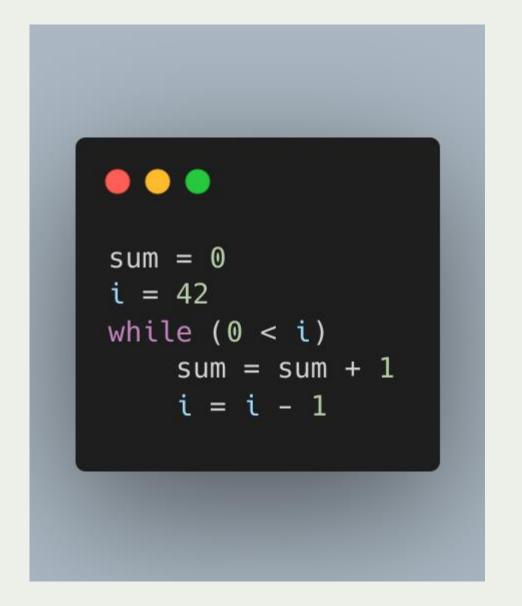
YOGO query was 60 lines; the analyzer was 330 lines

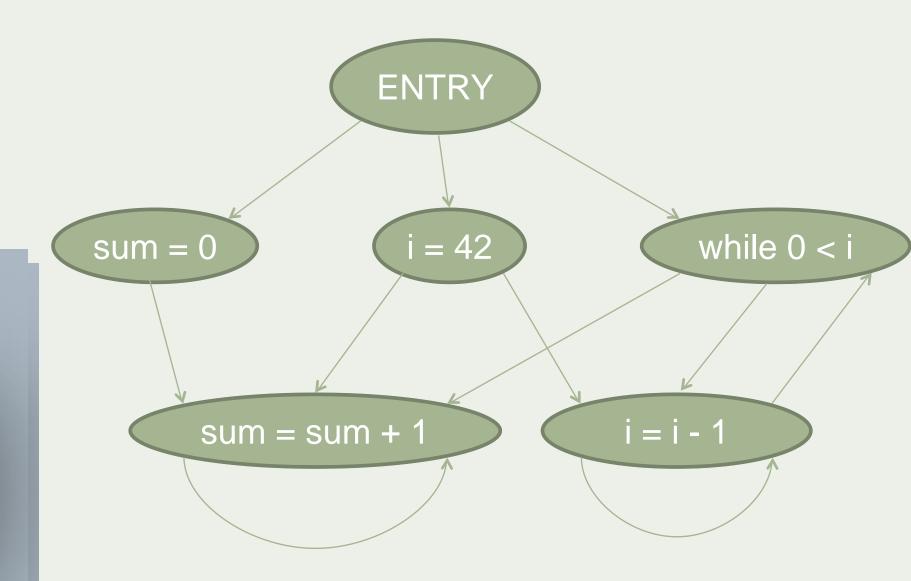
Usability evaluation

YOGO was straightforward enough to use that the authors

outsourced Graal query coding; it only took 3 days to learn!

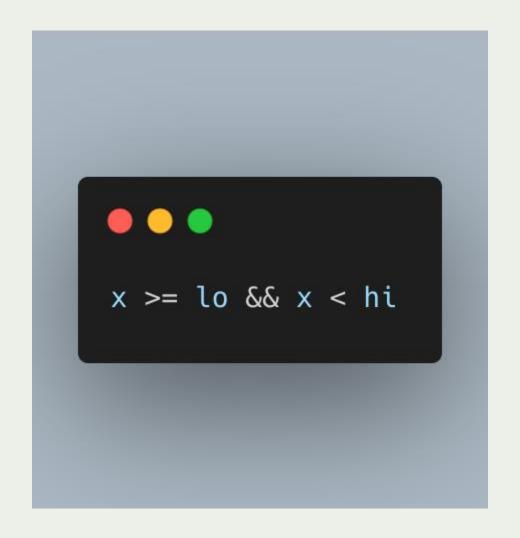




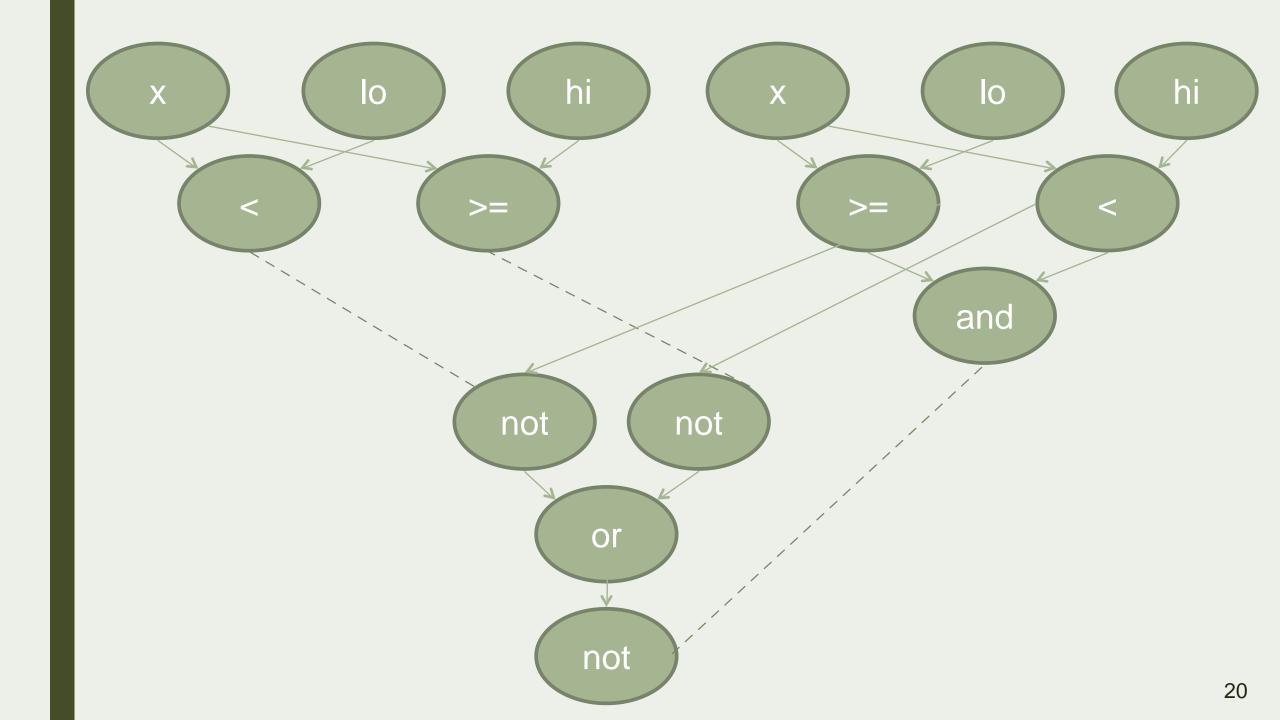




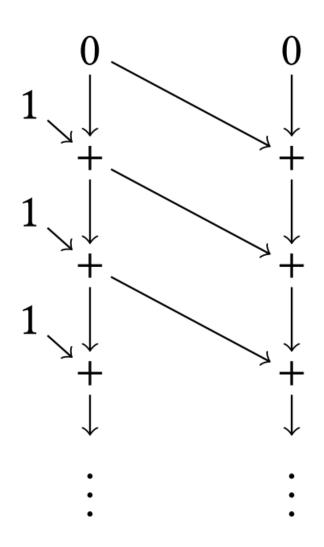
Program Expression Graphs give us an efficient procedure for code equivalence.



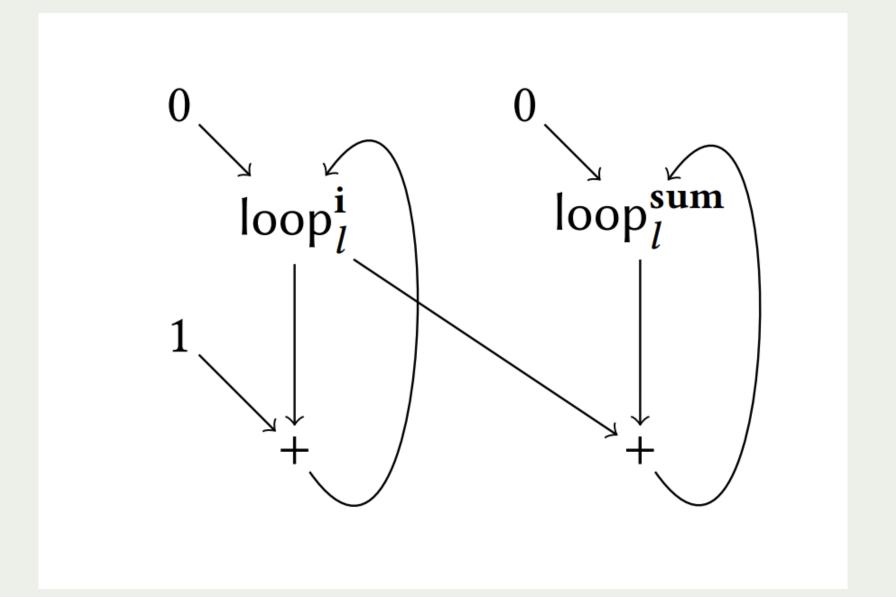
```
(defsearch bound-checking
    (root <- (generic/binop :or (generic/binop :< x lo)</pre>
                                  (generic/binop :>= x hi))))
```



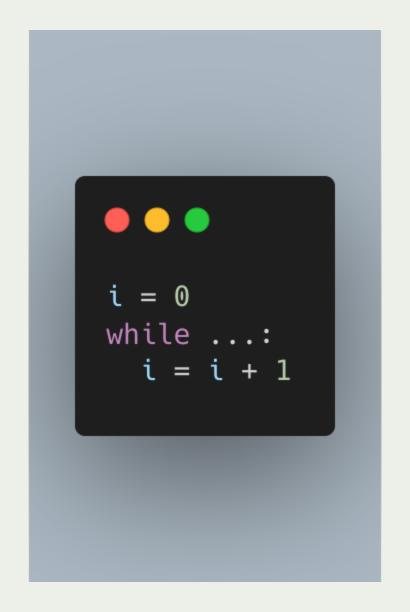
```
• • •
sum = 0;
for(i = 0; ...; i++) {
  sum += i;
```



loopsum

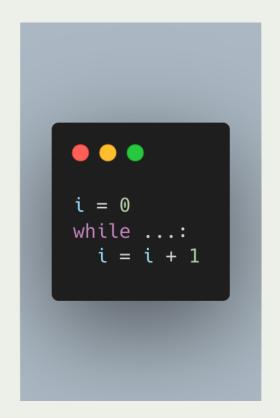


Node	Denotation
$loop(e_0, e)$	$\lambda i. \begin{cases} e_0 & i = 0 \\ e(i-1) & \text{otherwise} \end{cases}$
final(e, l)	$\lambda i.l(\min_{j\in\mathbb{N}} e(j) = \mathbf{false})$



High-level concepts can be identified as dataflow patterns.

Node	Denotation
seq(i, k)	$(i, i + K, i + 2k, \ldots)$



Node	Denotation
$iterV((t_0,t_1,\dots))$	$\lambda i.$ $\begin{cases} t_i & t_i \neq \bot \\ \bot & \text{otherwise} \end{cases}$
$iterP((t_0, t_1, \dots))$	$\lambda i.$ $\begin{cases} \mathbf{true} & \mathbf{t}_i \neq \bot \\ \mathbf{false} & \text{otherwise} \end{cases}$

INVARIANT(l, e)

PURE(f)

INDEPENDENT (λ_1, λ_2)

```
• • •
count = 0
for a in cart:
    if a == item
        count += 1
use (count)
```

 $\mathsf{counter} \leftarrow \mathsf{loop}_l(0,\mathsf{next})$

counter $\leftarrow loop_1(0, next)$ $next \leftarrow cond(iterV_l(coll) = k,$ counter + 1, counter) answer \leftarrow final_l(iterP_l(coll), counter) INVARIANT $_l(k)$

Conceptual rewrite soundness

WTS: Code refines behavior of concept specification

Rewrites must be sound wrt their concept's state specification

Rule writers prove soundness; users enjoy the benefits!

Key contributions

Coarser notion of equality supports hierarchical abstraction

Higher-level denotations in the DSL = more abstract

representations

Not a fixed set of equational rules! User-defined rewrite rules

Limitations

Rewriting is hard: syntactic differences, interleaving, overlapping

DSL cannot handle arbitrarily-nested expressions i.e. f(g(x))

Impure function calls assume no relation between memory states

Interprocedural patterns not supported; cause graph explosion

Comparison to other code search methods

Static analysis abstractions are too coarse

SMT methods struggle with loops and function calls

Other searches do not scale well

Reflections

Code searcher + bug finder

Generality makes abilities hard to characterize

Presents a technique with a lot of reusability