#17: BlinkFill and Constraintbased Search

Sankha Narayan Guria

EECS 700: Introduction to Program Synthesis



- Strengths? Weaknesses?
 - differences between FlashFill and BlinkFill language? which one is more expressive?

- What does BlinkFill use as behavioral constraints? Structural constraints? Search strategy?
 - input-output examples (+ input examples); custom string DSL; VSA
- What is the main technical insight of BlinkFill wrt FlashFill?
 - BlinkFill uses the available inputs (with no outputs) to infer structure (segmentation) common to all inputs
 - it uses this structure to shrink the DAG and to rank substring expressions

Example

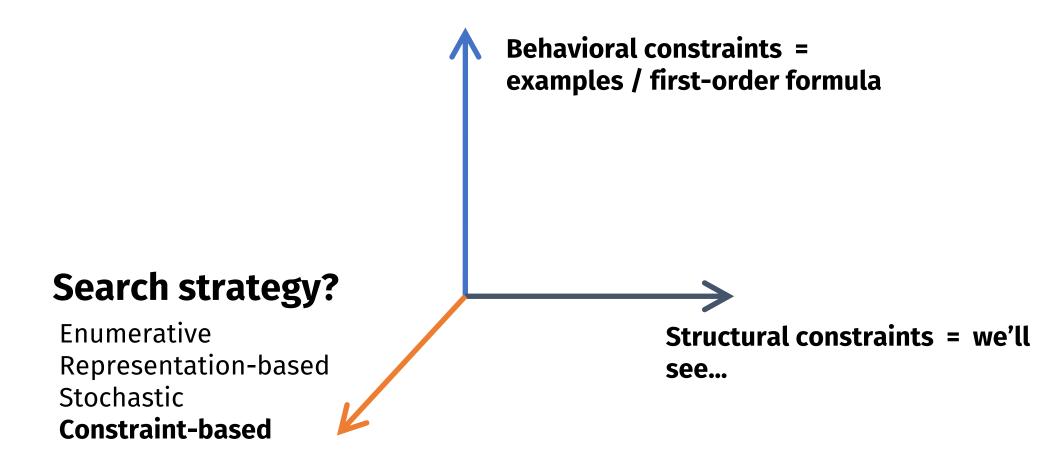
"Los Angeles, United States" • learn_F <"Mumbai, India" → "India"> learn_P sub⋈ {cstr("India")} <"Mumbai, India" **→** 9> {cpos(9)} ranked higher (", ", 1, End) ("Mumbai, ", 1, End) (ws, 1, End)

- Write a BlinkFill program that satisfies:
 - "Programming Language Design and Implementation (PLDI), 2019, Phoenix AZ" ->
 "PLDI 2019"
 - "Principles of Programming Languages (POPL), 2020, New Orleans LA" -> "POPL 2020"
 - Between first parentheses and between first and last comma:
 Concat(SubStr(v1, ("(", 1, End), (")", 1, Start)),
 SubStr(v1, (",", 1, End), (",", -1, Start)))

- Could we extend the algorithm to support sequences of tokens?
 - More expressive:
 - "Programming Language Design and Implementation: PLDI 2019" -> "PLDI 2019"
 - "POPL 2020 started on January 22" -> "POPL 2020"
 - SubStr(v1, (C ws d, 1, Start), (C ws d, 1, End))
 - Each edge of the single-string IDG would have more labels
 - Extra edges from 0 and to the last node
 - More edges left after intersection (might be a problem, but unclear)
 - Need fewer primitive tokens (no need for ProperCase)

On to today's topic

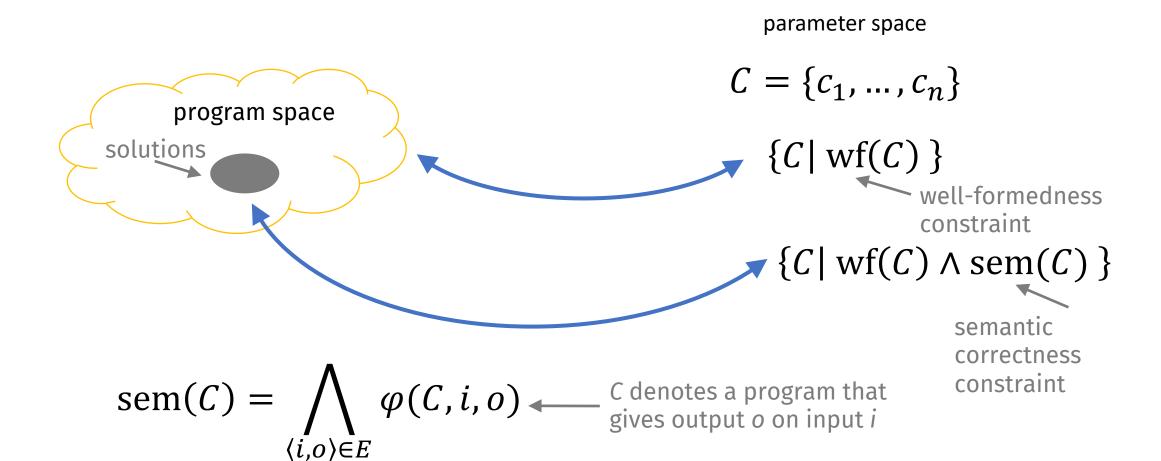
The problem statement



Constraint-based search

• Idea: encode the synthesis problem as a SAT/SMT problem and let a solver deal with it

What is an encoding?



How to define an encoding

- Define the parameter space $C = \{c_1, ..., c_n\}$
 - decode: C → Prog (might not be defined for all C)
- Define a formula $wf(c_1, ..., c_n)$
 - that holds iff decode[C] is defined
- Define a formula $\varphi(c_1, ..., c_n, i, o)$
 - that holds iff (decode[C])(i) = o

Constraint-based search

```
constraint-based (wf, \varphi, E = [i \rightarrow o]) { Find a satisfying match SAT(wf(\mathcal{C}) \land \land_{(i,o)\in E} \varphi(\mathcal{C},i,o)) with \longleftarrow assignment for c_1, ..., c_n (i and o are fixed) Model C* -> return decode[C*]
```

SAT encoding: example

```
x is a two-bit word
                                                                             parameter space
      program space
                                       (x = x_h x_l)
                                                                          C = \{c : Bool\}
                                       E = \begin{bmatrix} 11 \rightarrow 01 \end{bmatrix}
                                                                          decode[0] \rightarrow x
                                                                          decode[1] \rightarrow x \& 1
wf(c) \equiv T
\varphi(c, i_h, i_l, o_h, o_l) \equiv (\neg c \Rightarrow o_h = i_h \land o_l = i_l)
\wedge (c \Rightarrow o_h = 0 \wedge o_l = i_l)
SAT(\varphi(c, 1, 1, 0, 1))
                                                                                         SAT solver
SAT((\neg c \Rightarrow 0 = 1 \land 1 = 1) \land (c \Rightarrow 0 = 0 \land 1 = 1))
                                                                                                           Model \{c \rightarrow 1\}
                                       return decode[1] i.e. x & 1
```

SMT encoding: example

```
N is an in integer literal
                                                                                      parameter
       program space
                                                                               C = \{c_{op}^{\text{space}}: \text{Bool}, c_{N}: \text{Int}\}
                                          x is an integer input
                                          E = \begin{bmatrix} 2 \rightarrow 9 \end{bmatrix}
                                                                                decode[0,N] \rightarrow x + N
                                                                                decode[1,N] \rightarrow x * N
\operatorname{wf}(c_{op}, c_N) \equiv \mathsf{T}
\varphi(c_{op}, c_N, i, o) \equiv (\neg c_{op} \Rightarrow o = i + c_N) \land (c_{op} \Rightarrow o = i * c_N)
SAT(\varphi(c_{op}, c_N, 2, 9))
                                                                                                SMT solver
SAT((\neg c_{op} \Rightarrow 9 = 2 + c_N) \land (c_{op} \Rightarrow 9 = 2 * c_N))
                                                                                                                 Model \{c_{op} \rightarrow 0,
                                                                                                                               c_N \rightarrow 7
                                          return decode [0,7] i.e. x + 7
```

What is a good encoding?

- Sound
 - if wf(C) ∧ sem(C) then decode[C] is a solution
- Complete
 - if decode[C] is a solution then $wf(C) \wedge sem(C)$
- Small parameter space
 - avoid symmetries
- Solver-friendly
 - decidable logic, compact constraint

DSL limitations

- Program space can be parameterized with a finite set of parameters
 - Counterexample:

```
| L + L | [N] | x
N ::= find(L,N) | 0
```

Workaround

 $L ::= sort(L) \mid L[N..N]$

- Program semantics $\varphi(C,i,o)$ is expressible as a (decidable) SAT/SMT formula
 - Counterexample

Comparison of search strategies

