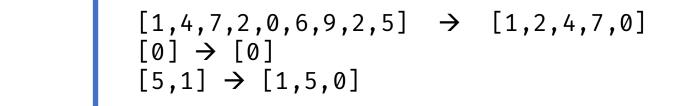
#9: Version Space Algebra

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EECS 700: Introduction to Program Synthesis



The problem statement



Behavioral constraints = examples

Search strategy?

Enumerative
Representation-based
Stochastic
Constraint-based

Structural constraints = grammar

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

Representation-based search

Idea:

- 1. build a data structure that compactly represents good parts of the program space
- 2. extract solution from that data structure

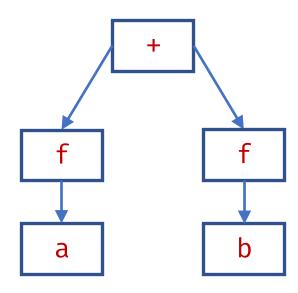
Compact term representation

Consider the space of 9 programs:

$$f(a) + f(a)$$
 $f(a) + f(b)$ $f(a) + f(c)$
 $f(b) + f(a)$ $f(b) + f(b)$ $f(b) + f(c)$
 $f(c) + f(a)$ $f(c) + f(b)$ $f(c) + f(c)$

Can we represent this compactly?

• observation 1: same top-level structure, independent subterms



Compact term representation

Consider the space of 9 programs:

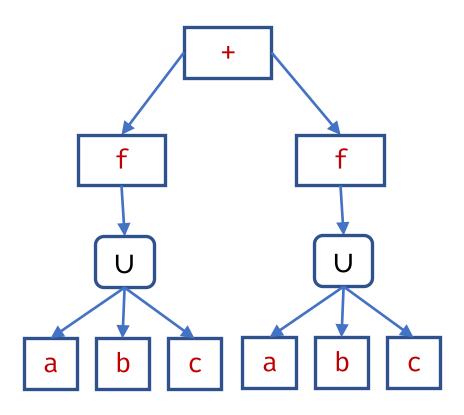
```
f(a) + f(a) f(a) + f(b) f(a) + f(c)

f(b) + f(a) f(b) + f(b) f(b) + f(c)

f(c) + f(a) f(c) + f(b) f(c) + f(c)
```

Can we represent this compactly?

- observation 1: same top level structure, independent subterms
- observation 2: shared sub-spaces



Compact term representation

Consider the space of 9 programs:

```
f(a) + f(a) f(a) + f(b) f(a) + f(c)

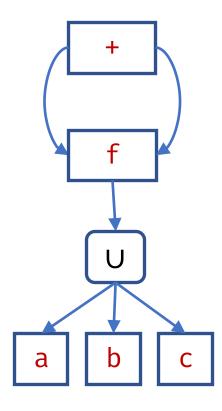
f(b) + f(a) f(b) + f(b) f(b) + f(c)

f(c) + f(a) f(c) + f(b) f(c) + f(c)
```

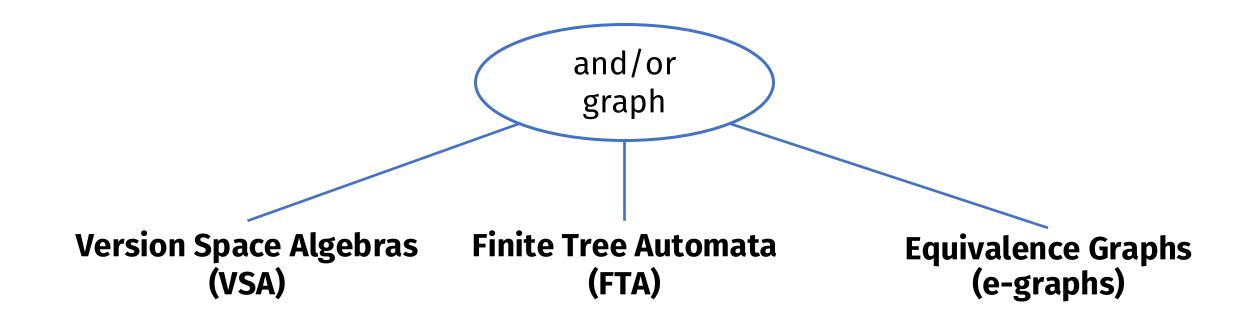
Can we represent this compactly?

- observation 1: same top level structure, independent subterms
- observation 2: shared sub-spaces

Key idea: use and-or graph!



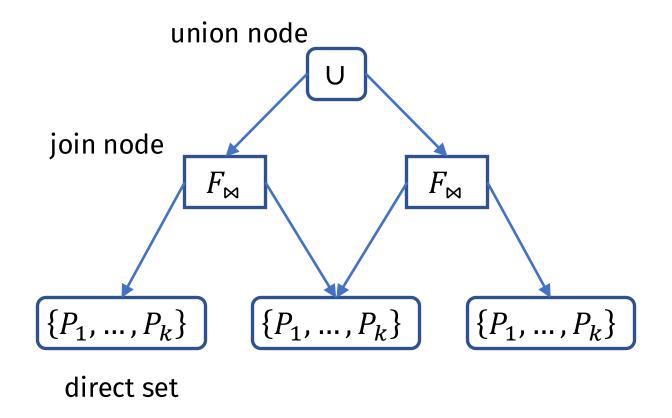
Representation-based search



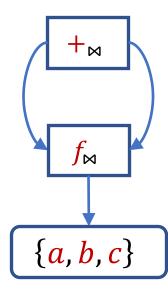
Version Space Algebra

- Idea: build a graph that succinctly represents the space of all programs consistent with examples
 - called a version space
- Operations on version spaces:
 - learn $\langle i, o \rangle \rightarrow VS$
 - $VS_1 \cap VS_2 \rightarrow VS$
 - extract VS → program
- Algorithm:
 - 1. learn a VS for each example
 - 2. intersect them all
 - 3. extract any (or best) program

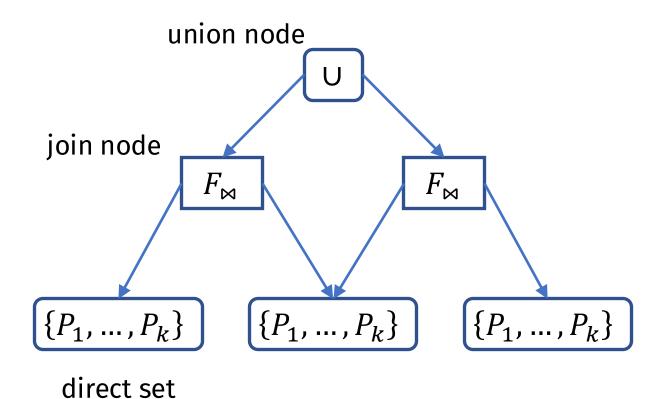
Version Space Algebra



example:



Version Space Algebra



Volume of a VSA V(VSA) (the number of nodes)

Size of a VSA (the number of programs)

$$V(VSA) = O(\log|VSA|)$$

VSA-based search

- Mitchell: Generalization as search. AI 1982
- Lau, Domingos, Weld. Version space algebra and its application to programming by example. ICML 2000
- Gulwani: Automating string processing in spreadsheets using input-output examples. POPL 2011.
 - Follow-up work: BlinkFill, FlashExtract, FlashRelate, ...
 - generalized in the PROSE framework

FlashFill

Simplified grammar:

```
E ::= F | concat(F, E) "Trace" expression

F ::= cstr(str) | sub(P, P) Atomic expression

P ::= cpos(num) | pos(R, R) Position expression

R ::= tokens(T_1, ..., T_n) Regular expression

T ::= c | c | Token expression

C ::= c | c | Alpha | c | Alpha
```

FlashFill: example

```
"Hello POPL 2023" → "POPL'2023"

"Goodbye PLDI 2021" → "PLDI'2021"

concat(
    sub(pos(ws, Alpha), pos(Alpha, ws)),
    concat(
        cstr("'"),
        sub(pos(ws, digit), pos(digit, $))))
```

```
E ::= F | concat(F, E)
F ::= cstr(str) | sub(P, P)
P ::= cpos(num) | pos(R, R)
R ::= tokens(T<sub>1</sub>, ..., T<sub>n</sub>)
T ::= C | C+
```

VSAs for Flashfill

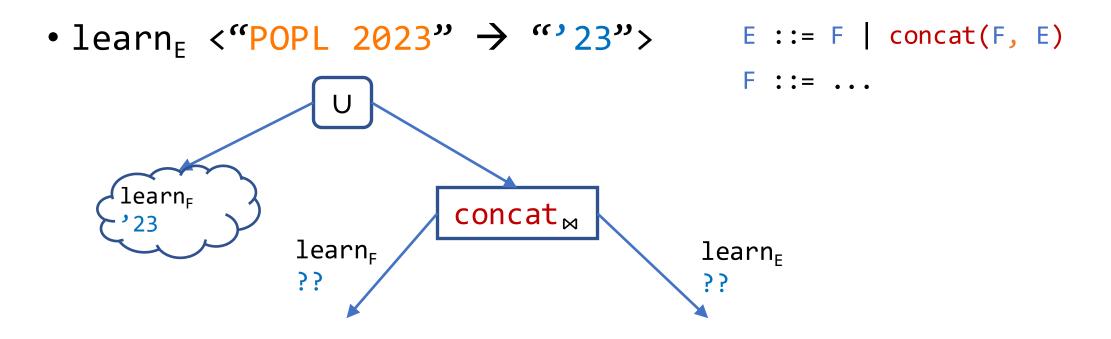
- Recall operations on version spaces:
 - learn $\langle i, o \rangle \rightarrow VS$
 - $VS_1 \cap VS_2 \rightarrow VS$
 - extract VS → program
- How do we implement learn?
 - define learn_N <i, o> for every non-terminal N
 - build VS top-down,
 propagating <i, o> the example

```
E ::= F | concat(F, E)
F ::= cstr(str) | sub(P, P)
P ::= cpos(num) | pos(R, R)
R ::= tokens(T<sub>1</sub>, ..., T<sub>n</sub>)
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```

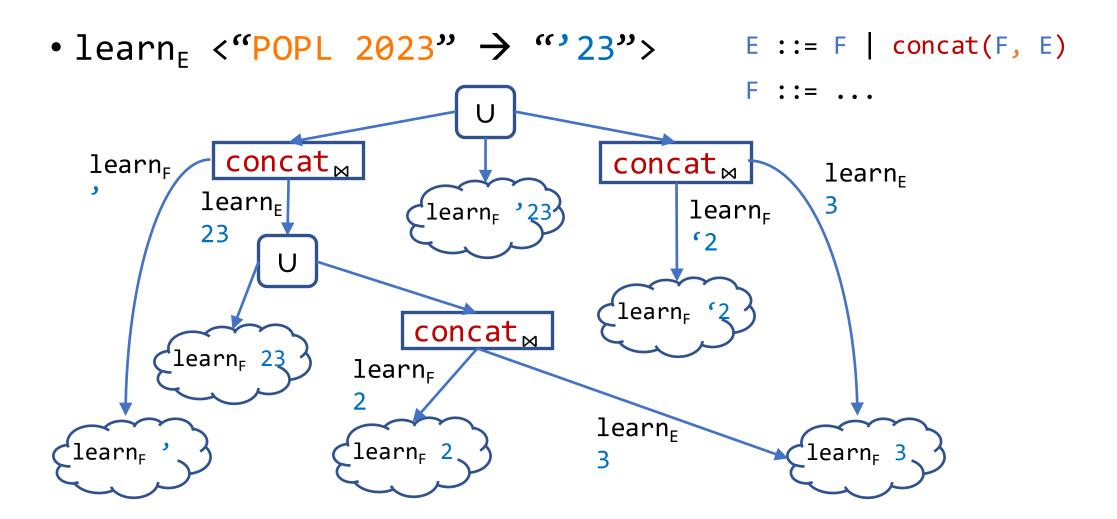
Learning atomic expressions

```
• learn<sub>F</sub> <"POPL 2023" → "2023">
                                                                  F ::= cstr(str) \mid sub(P_1, P_2)
                                                                  P ::= cpos(num) \mid pos(R_1, R_2)
                                                                  R ::= tokens(T_1, \ldots, T_n)
                          learn<sub>p</sub>
                                                                  T ::= C | C+
                          <"POPL 2023"
{cstr("2023")}
                                           sub⋈
                          \rightarrow 5>
                                                                       learn<sub>p</sub>
                                                                       <"POPL 2023"
                                                                       \rightarrow 9>
{cpos(5)}
                learn<sub>R</sub>
                                                        learn<sub>R</sub>
                                       pos⋈
                match "POPL
                                                        match "2023"
                                                       {digit+}
                {ws, alpha+ ws}
```

Learning trace expressions



Learning trace expressions

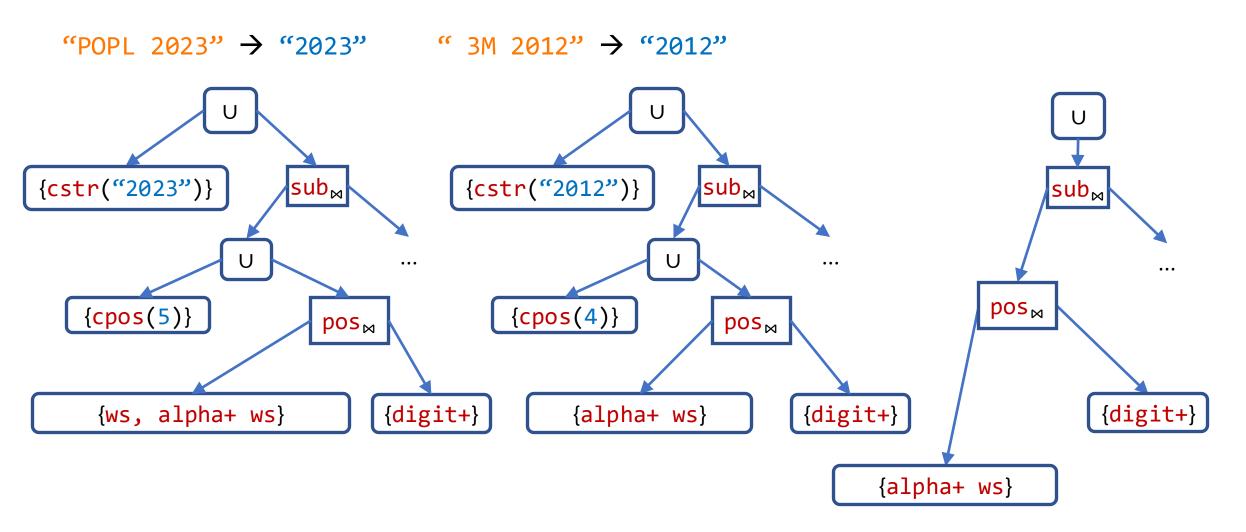


VSAs for Flashfill

- Recall operations on version spaces:
 - learn $\langle i, o \rangle \rightarrow VS$
 - $VS_1 \cap VS_2 \rightarrow VS$
 - extract VS → program
- How do we implement intersection?
 - top-down
 - union: intersect all pairs of children
 - join: intersect children pairwise

```
E ::= F | concat(F, E)
F ::= cstr(str) | sub(P, P)
P ::= cpos(num) | pos(R, R)
R ::= tokens(T<sub>1</sub>, ..., T<sub>n</sub>)
T ::= C | C+
```

Intersection



VSAs for Flashfill

- Recall operations on version spaces:
 - learn $\langle i, o \rangle \rightarrow VS$
 - $VS_1 \cap VS_2 \rightarrow VS$
 - extract VS → program
- How do we implement extract?
 - any program: just pick one child from every union
 - best program: shortest path in a DAG

```
E ::= F | concat(F, E)
F ::= cstr(str) | sub(P, P)
P ::= cpos(num) | pos(R, R)
R ::= tokens(T<sub>1</sub>, ..., T<sub>n</sub>)
T ::= C | C+
```

Discussion

- What do VSAs remind you of in the enumerative world?
 - VSA learning ~ top-down search with top-down propagation
- How are they different?
 - Caching of sub-problems (DAG!)
 - Can construct one per example and intersect
 - This allows it to guess arbitrary constants!

Discussion

- Why could we build a finite representation of all solutions?
 - Could we do it for this language?

```
E ::= F + F

k \in \mathbb{Z} + is integer addition
```

What about this language?

```
E ::= E + 1 | x
```

DSL restrictions: efficiently invertible

- Every operator has a small, easily computable inverse
 - Example when an inverse is small but hard to compute?
- The space of sub-specs is finite
 - either non-recursive grammar
 - or finite space of values for the recursive non-terminal (e.g. bit-vectors)
 - or every recursive production generates a strictly smaller spec

```
E ::= F | concat(F, E)

learn<sub>E</sub> '18

learn<sub>E</sub> 18
```

PROSE

Grammar

A ::=
$$f(B, C) \mid ...$$

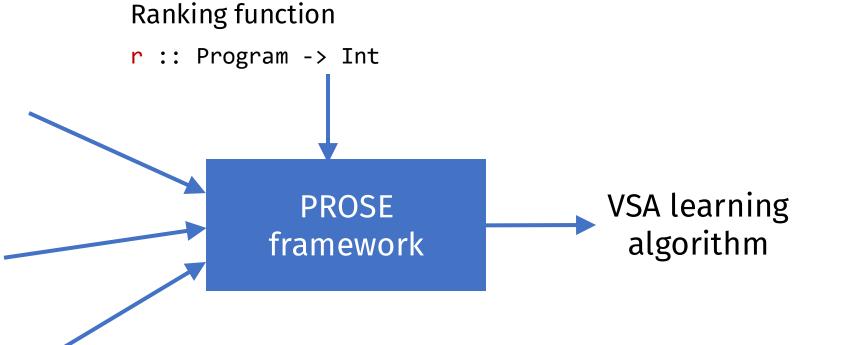
B ::= $g(D, E) \mid ...$

Semantics

Inverse semantics

$$f^{-1} :: A \rightarrow \{(B, C)\}$$

 $f^{-1} a = ...$



https://microsoft.github.io/prose/