

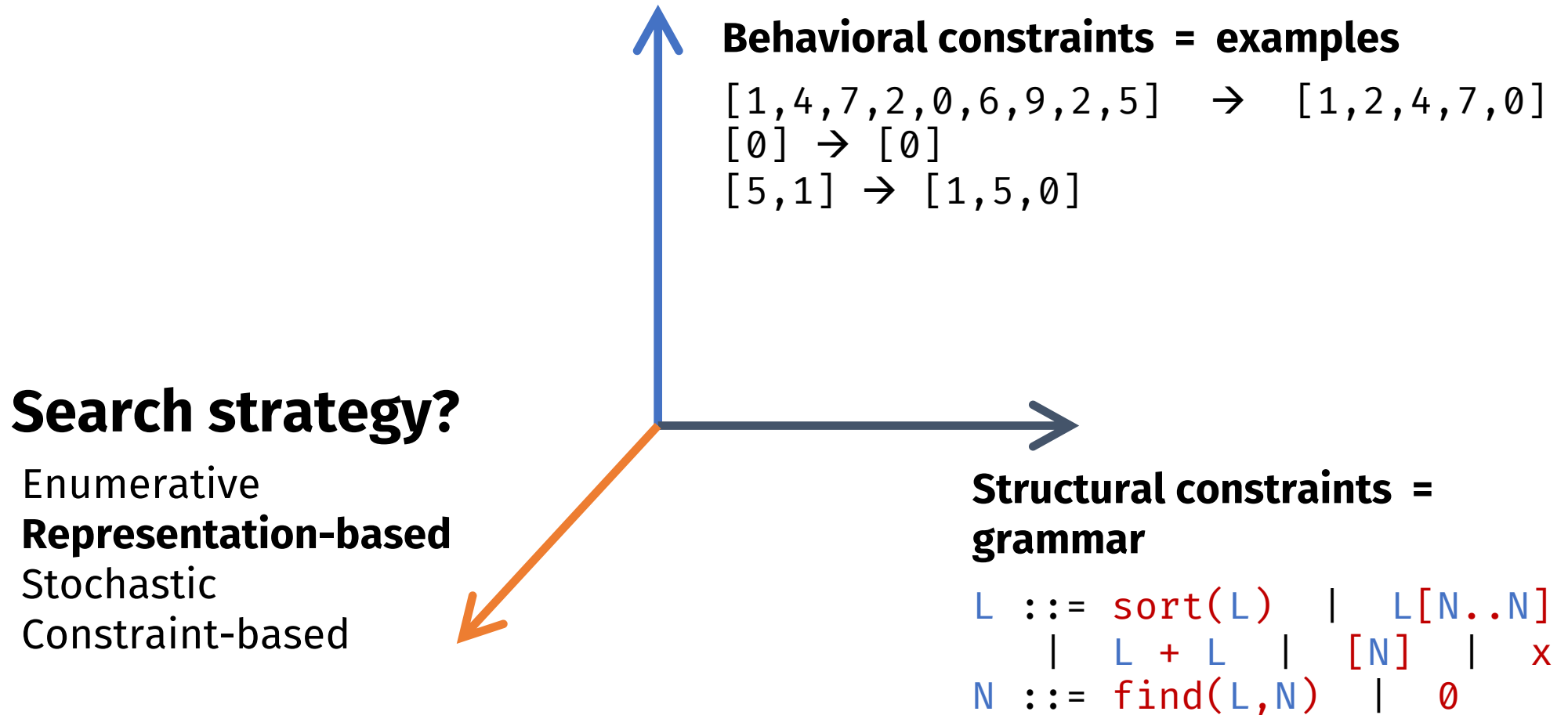
#9: Version Space Algebra

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EECS 700: Introduction to Program Synthesis



The problem statement



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

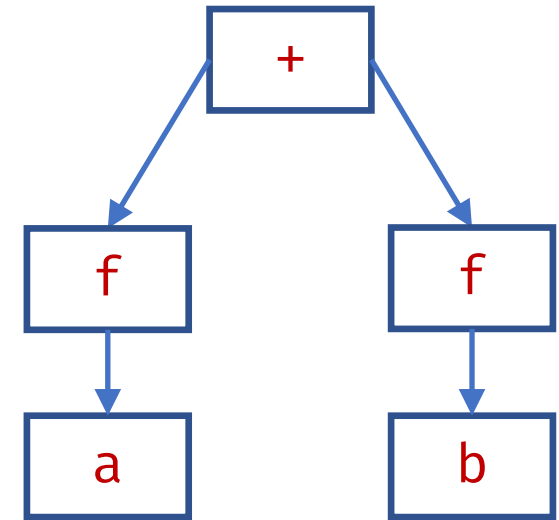
Compact term representation

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



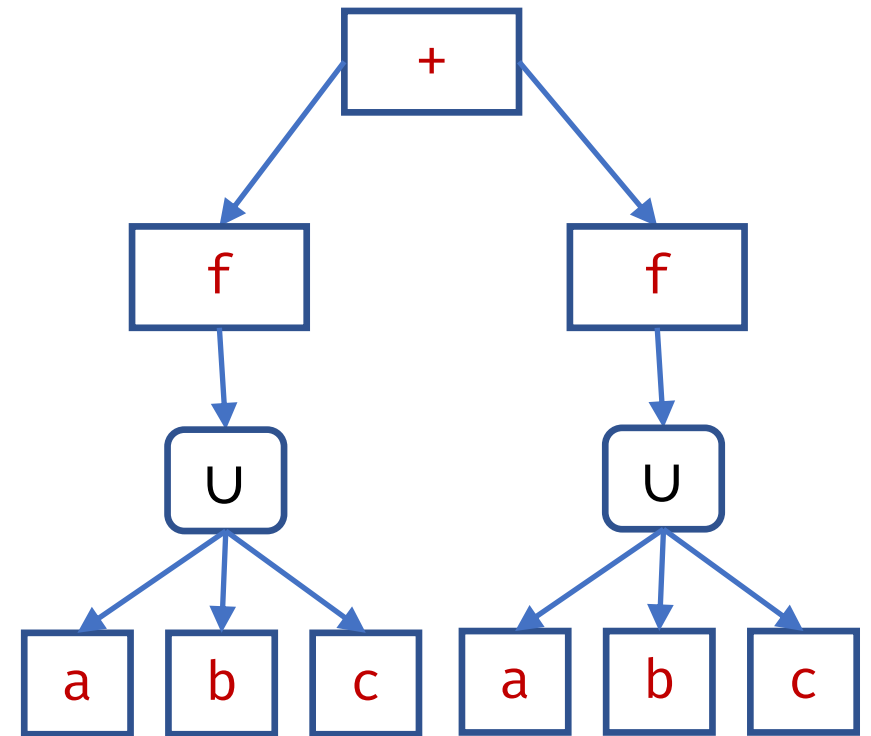
Compact term representation

Consider the space of 9 programs:

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Can we represent this compactly?

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



Compact term representation

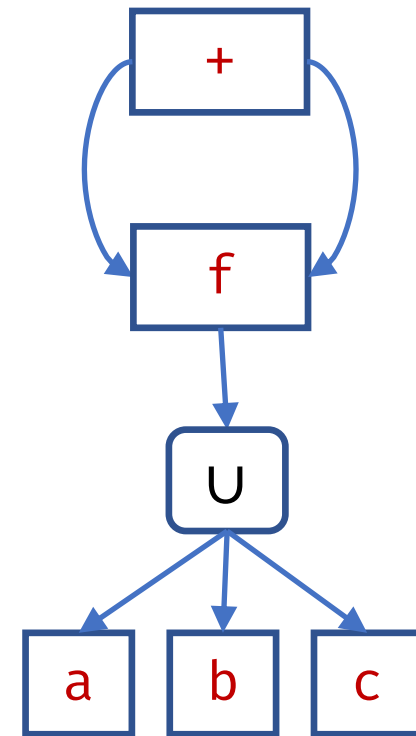
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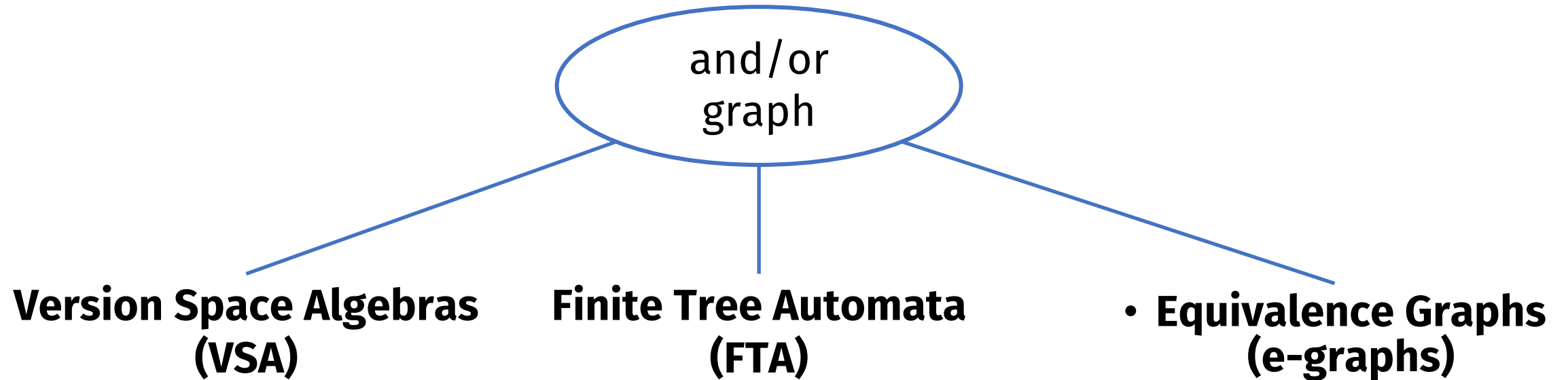
Can we represent this compactly?

- observation 1: same top level structure, independent subterms
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Key idea: use and-or graph!



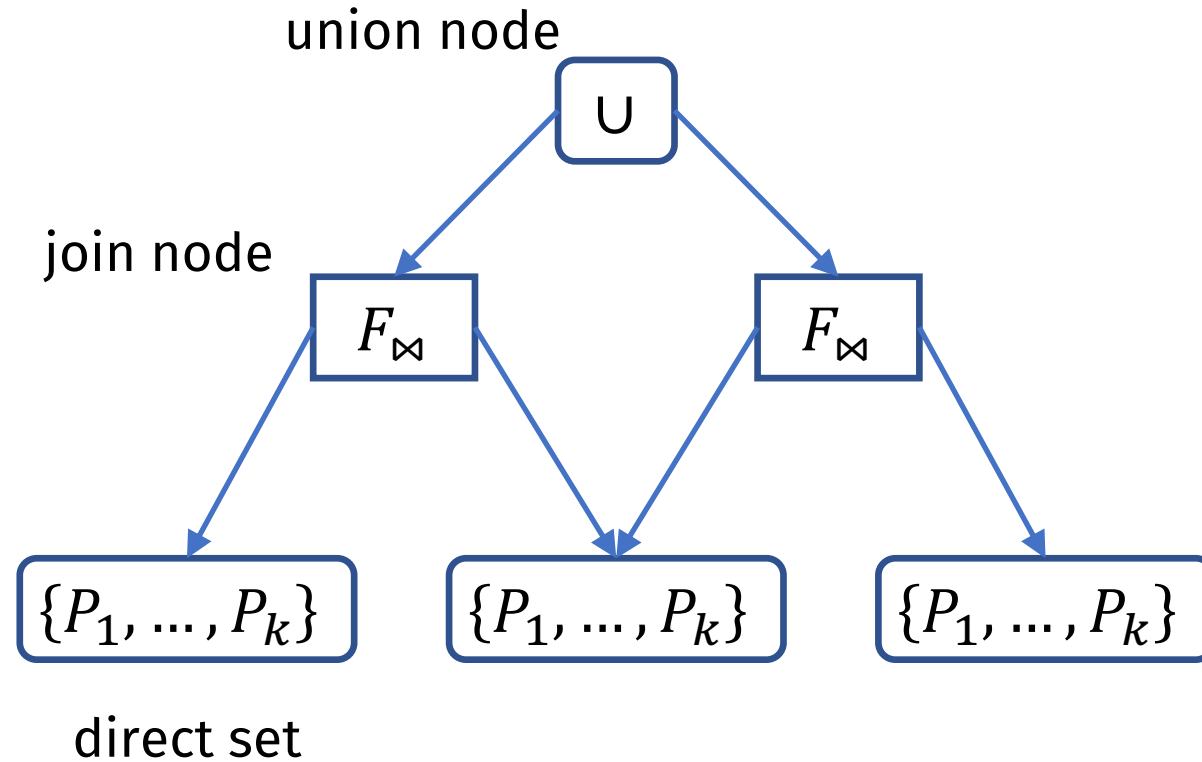
Representation-based search



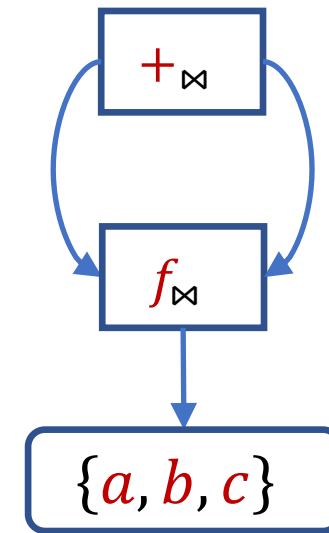
Version Space Algebra

- **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 $VS \rightarrow \text{program}$
- Algorithm:
 1. learn a VS for each example
 2. intersect them all
 3. extract any (or best) program

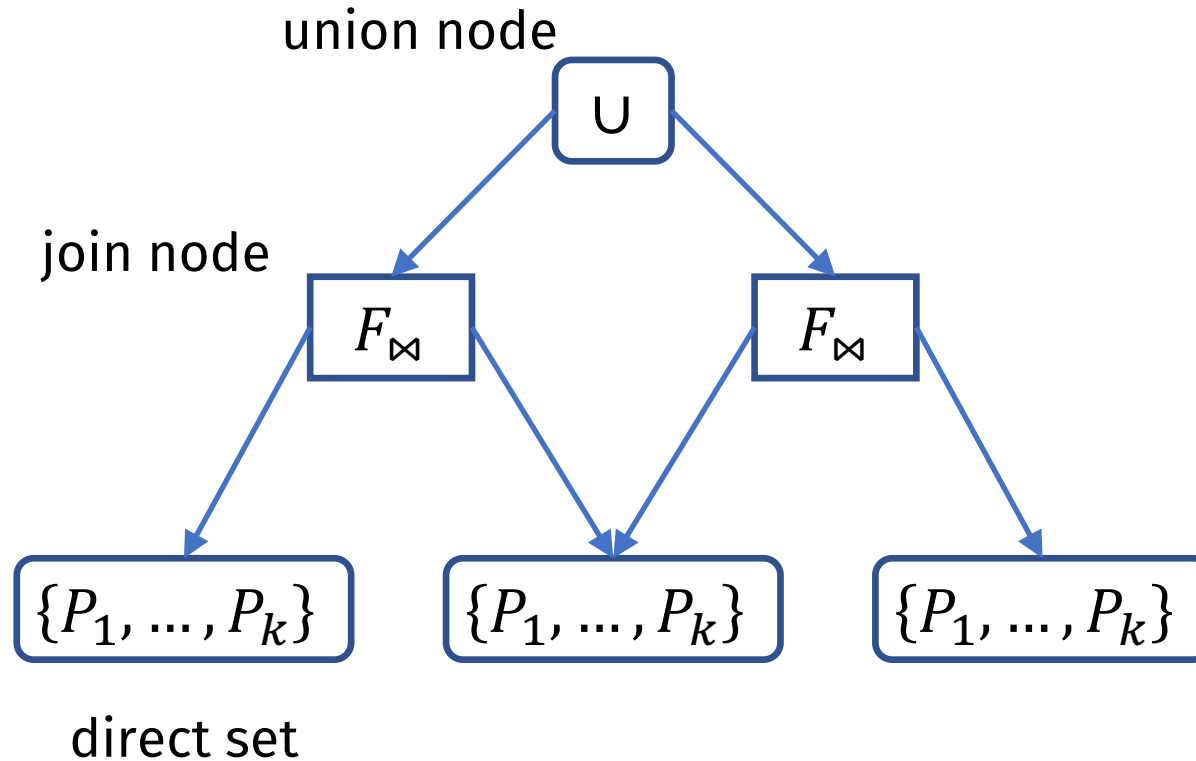
Version Space Algebra



example:



Version Space Algebra



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

Size of a VSA
(the number of programs) $|VSA|$

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

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

[Gulwani '11]

Simplified
grammar:

$E ::= F \mid \text{concat}(F, E)$	“Trace” expression
$F ::= \text{cstr}(str) \mid \text{sub}(P, P)$	Atomic expression
$P ::= \text{cpos}(num) \mid \text{pos}(R, R)$	Position expression
$R ::= \text{tokens}(T_1, \dots, T_n)$	Regular expression
$T ::= C \mid C^+$	Token expression
$C ::= ws \mid digit \mid alpha \mid Alpha \mid \$ \mid ^$ $\mid \dots$	

FlashFill: example

0 1 2 3 4 5 6 7 8 9 ...

“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, $))))
```

$E ::= F \mid \text{concat}(F, E)$

$F ::= \text{cstr}(str) \mid \text{sub}(P, P)$

$P ::= \text{cpos}(num) \mid \text{pos}(R, R)$

$R ::= \text{tokens}(T_1, \dots, T_n)$

$T ::= C \mid C+$

VSAs for Flashfill

- Recall operations on version spaces:

- $\text{learn } \langle i, o \rangle \rightarrow \text{VS}$
- $\text{VS}_1 \cap \text{VS}_2 \rightarrow \text{VS}$
- $\text{extract VS} \rightarrow \text{program}$

- How do we implement learn?

- define $\text{learn}_N \langle i, o \rangle$
for every non-terminal N
- build VS top-down,
propagating $\langle i, o \rangle$ the example

$E ::= F \mid \text{concat}(F, E)$

$F ::= \text{cstr}(str) \mid \text{sub}(P, P)$

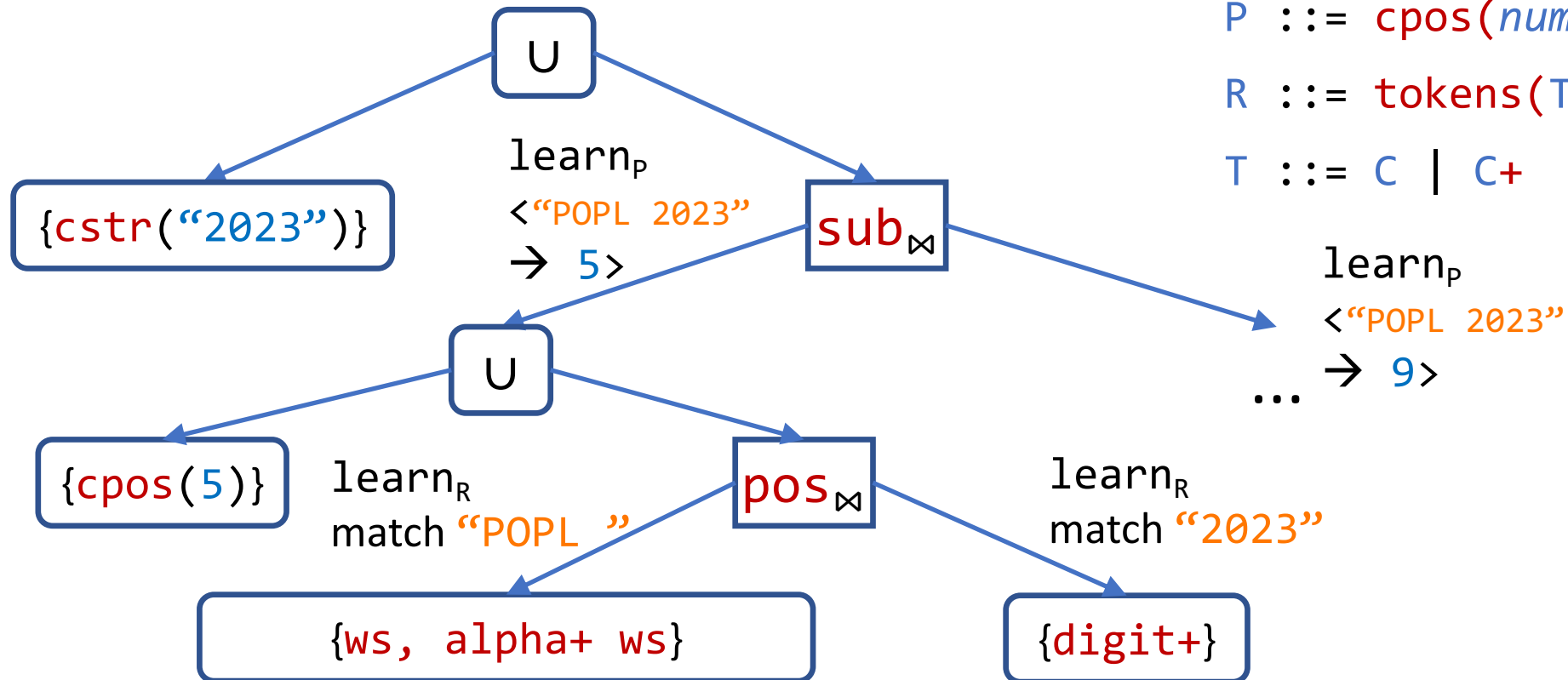
$P ::= \text{cpos}(num) \mid \text{pos}(R, R)$

$R ::= \text{tokens}(T_1, \dots, T_n)$

$T ::= C \mid C+$

Learning atomic expressions

- $\text{learn}_F \langle \text{"POPL 2023"} \rangle \rightarrow \text{"2023"} \rangle$



$F ::= cstr(str) \mid sub(P_1, P_2)$

$P ::= cpos(num) \mid pos(R_1, R_2)$

$R ::= tokens(T_1, \dots, T_n)$

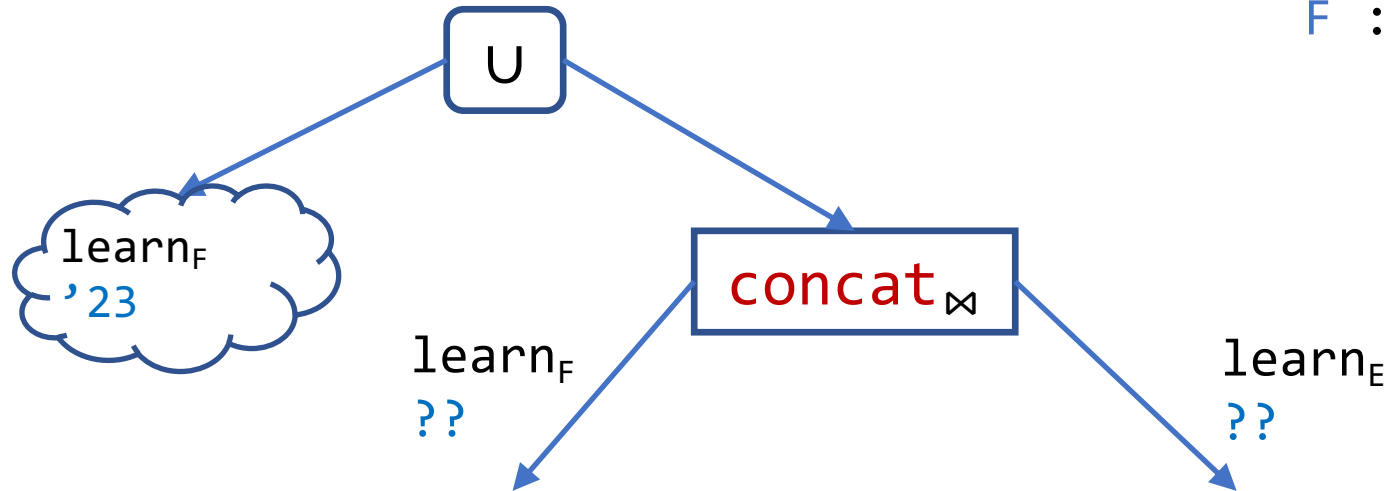
$T ::= C \mid C+$

Learning trace expressions

- $\text{learn}_E \langle \text{"POPL 2023"} \rightarrow \text{"'23"} \rangle$

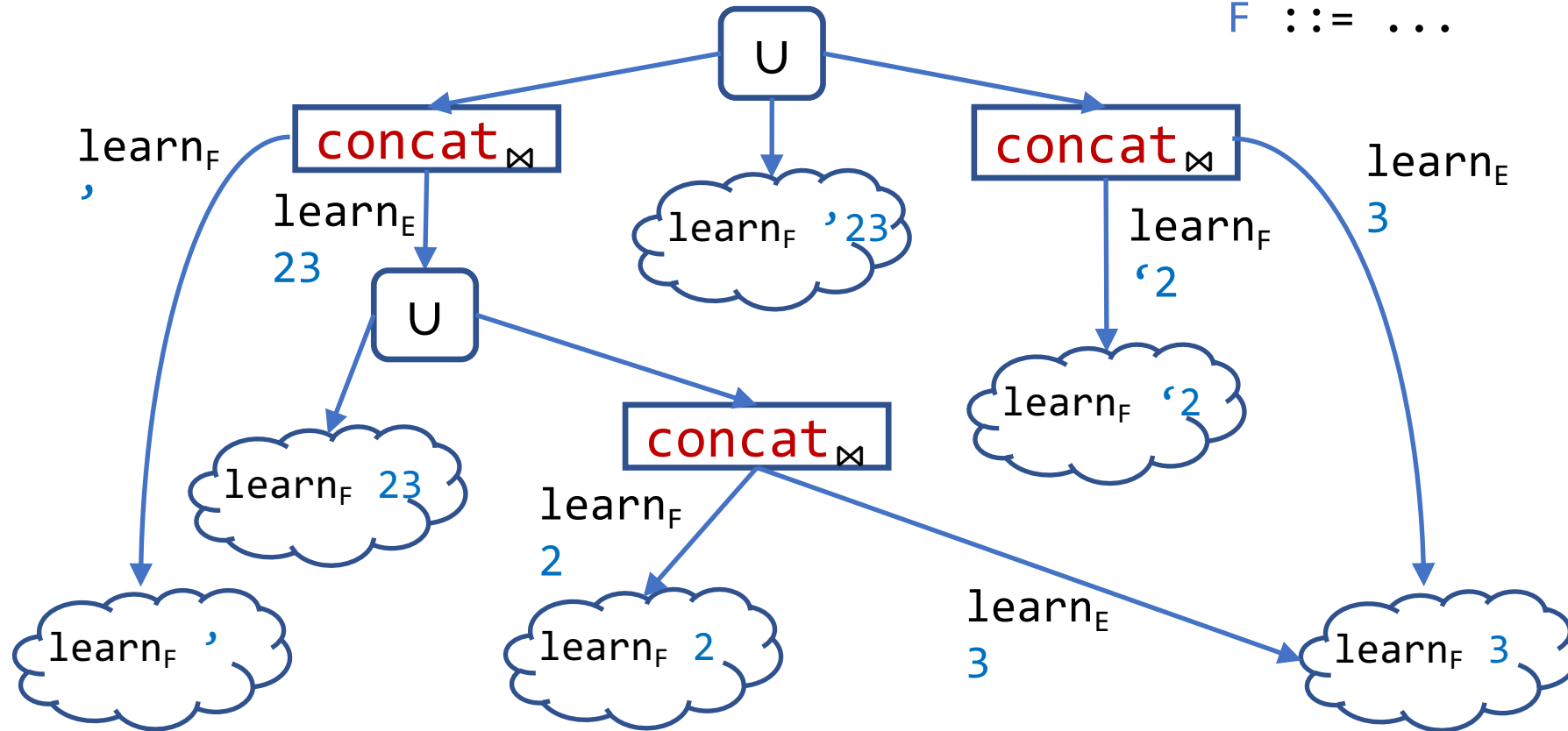


$E ::= F \mid \text{concat}(F, E)$
 $F ::= \dots$



Learning trace expressions

- $\text{learn}_E \langle \text{"POPL 2023"} \rightarrow \text{"'23"} \rangle$ $E ::= F \mid \text{concat}(F, E)$
 $F ::= \dots$



VSAs for Flashfill

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

$E ::= F \mid \text{concat}(F, E)$

$F ::= \text{cstr}(str) \mid \text{sub}(P, P)$

$P ::= \text{cpos}(num) \mid \text{pos}(R, R)$

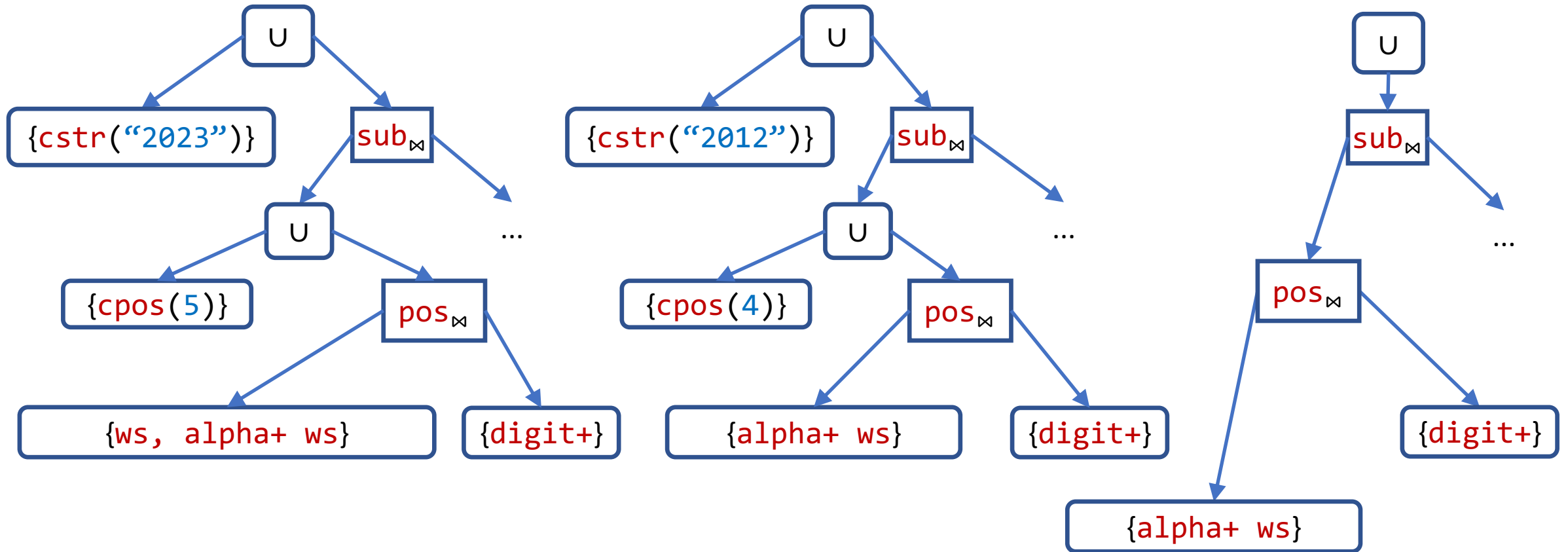
$R ::= \text{tokens}(T_1, \dots, T_n)$

$T ::= C \mid C+$

Intersection

“POPL 2023” → “2023”

“ 3M 2012” → “2012”



VSAs for Flashfill

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

$E ::= F \mid \text{concat}(F, E)$

$F ::= \text{cstr}(str) \mid \text{sub}(P, P)$

$P ::= \text{cpos}(num) \mid \text{pos}(R, R)$

$R ::= \text{tokens}(T_1, \dots, T_n)$

$T ::= C \mid C+$

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?

$E ::= F + F$

$F ::= k \mid x$

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

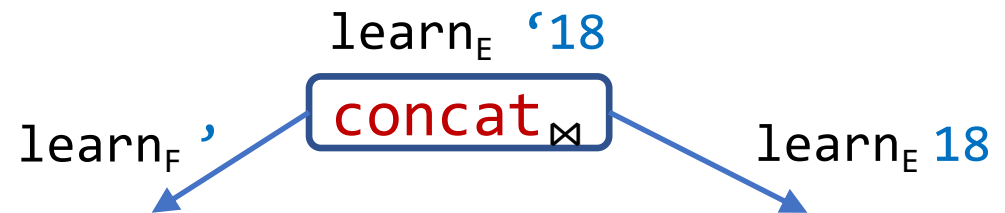
- What about this language?

$E ::= E + 1 \mid x$

DSL restrictions: efficiently invertible

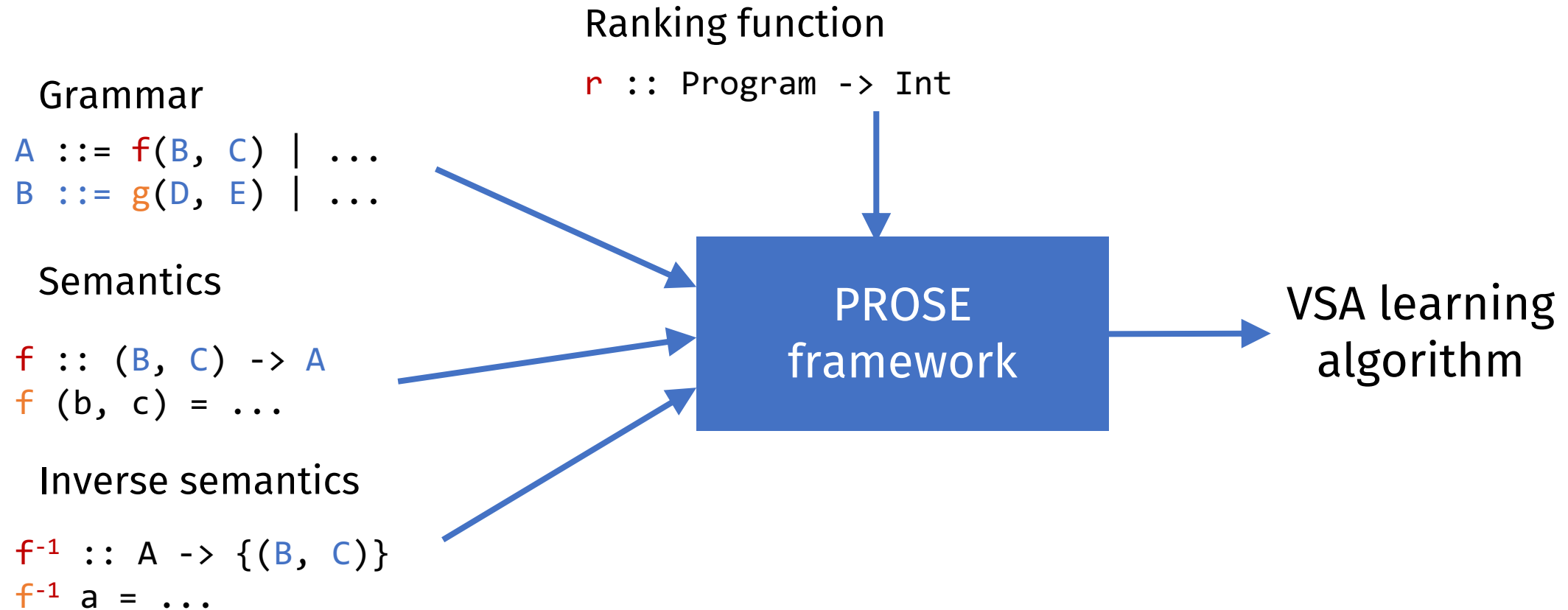
- Every operator has a small, easily computable inverse
 - Example when an inverse is small but hard to compute?
- The space of sub-specs is finite
 - either non-recursive grammar
 - or finite space of values for the recursive non-terminal (e.g. bit-vectors)
 - or every recursive production generates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

$E ::= F \mid \text{concat}(F, E)$



PROSE

[Polozov, Gulwani '15]



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