

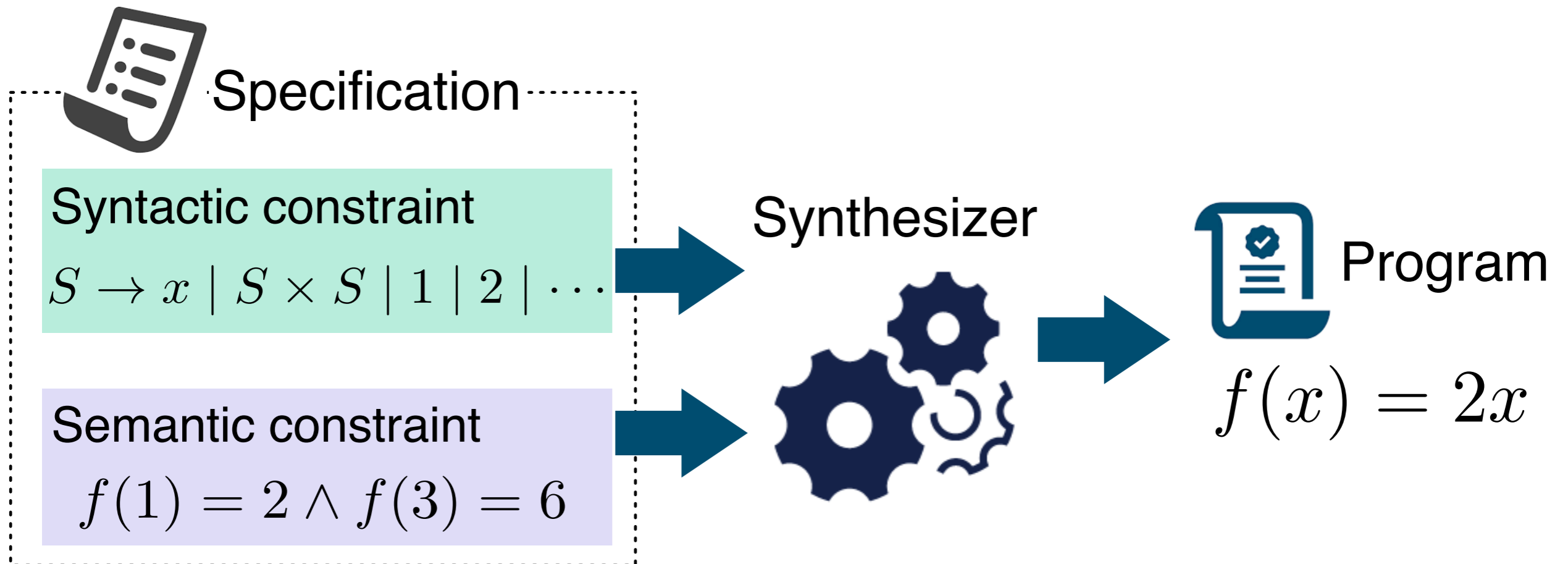
Accelerating Search-Based Synthesis Using Learned Probabilistic Models

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Syntax-Guided Program Synthesis (SyGuS)



Existing General-Purpose Strategies

- **Enumerative:** search with pruning
 - EUSolver: Udupa et al. (PLDI'13)
- **Symbolic:** constraint solving
 - CVC4: Reynolds et al. (CAV'15)
- **Stochastic:** probabilistic walk
 - STOKE: Schkufza et al. (ASPLOS'13)

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Key limitation:
search not guided towards *likely* programs

Statistical Regularities in Programs

- Programs contain repetitive and predictable patterns.

```
for (i = 0; i < 100; ??)
```

- Statistical program models define a probability distribution over programs.

$$Pr(?? \rightarrow i++ \mid \text{for } (i = 0; i < 100; ??)) = 0.80$$

$$Pr(?? \rightarrow i-- \mid \text{for } (i = 0; i < 100; ??)) = 0.01$$

– e.g., n-gram, probabilistic context-free grammar (PCFG), ...

- Applications: code completion, deobfuscation, program repair...

Exploiting Statistical Regularities

Can we leverage statistical program models to accelerate program synthesis?

Key Challenges:

1. How to guide the search given a statistical model?
2. How to learn a good statistical model?

Our Contributions

- A general approach to accelerate *CEGIS*-based program synthesis
 - by using a probabilistic model to guide the search
 - supports a wide range of models (e.g., *n*-gram, PCFG, PHOG, ...)
- Transfer learning-based method to mitigate overfitting
- Tool (Euphony) and evaluation on widely applicable domains

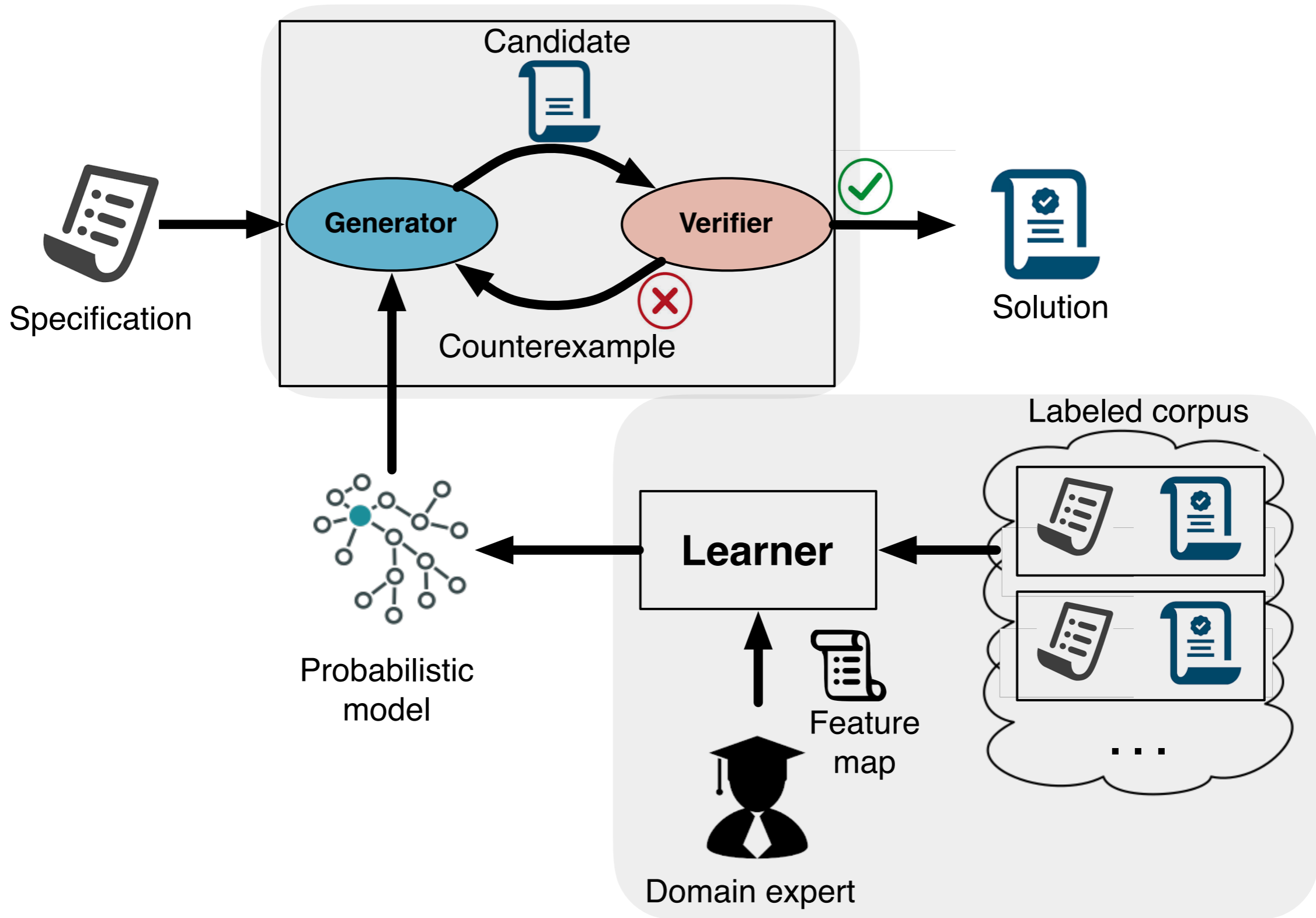
<https://github.com/wslee/euphony>



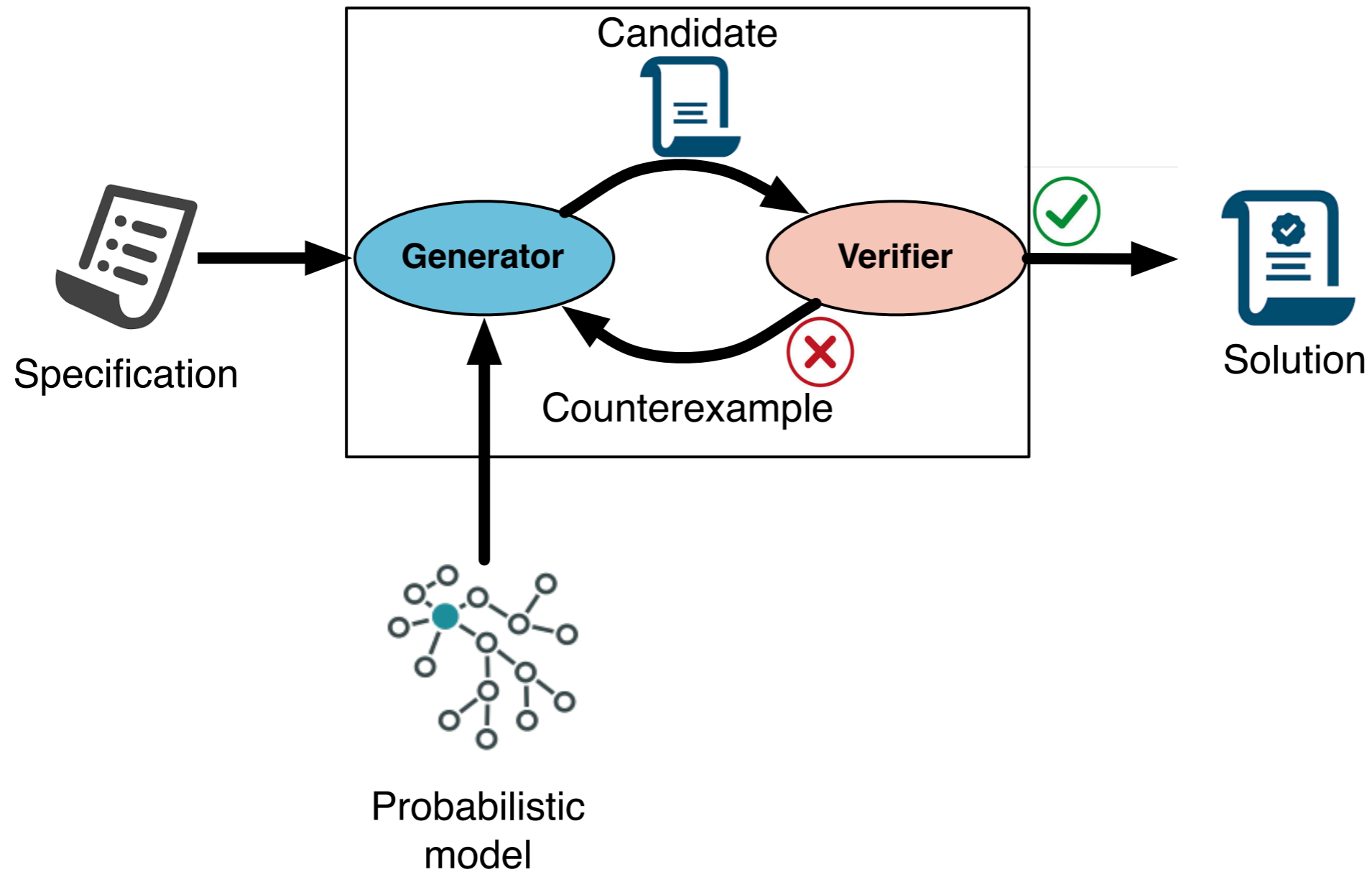
Talk Outline

- Overall Architecture
- Illustrative Example
- Empirical Evaluation

Overall Architecture

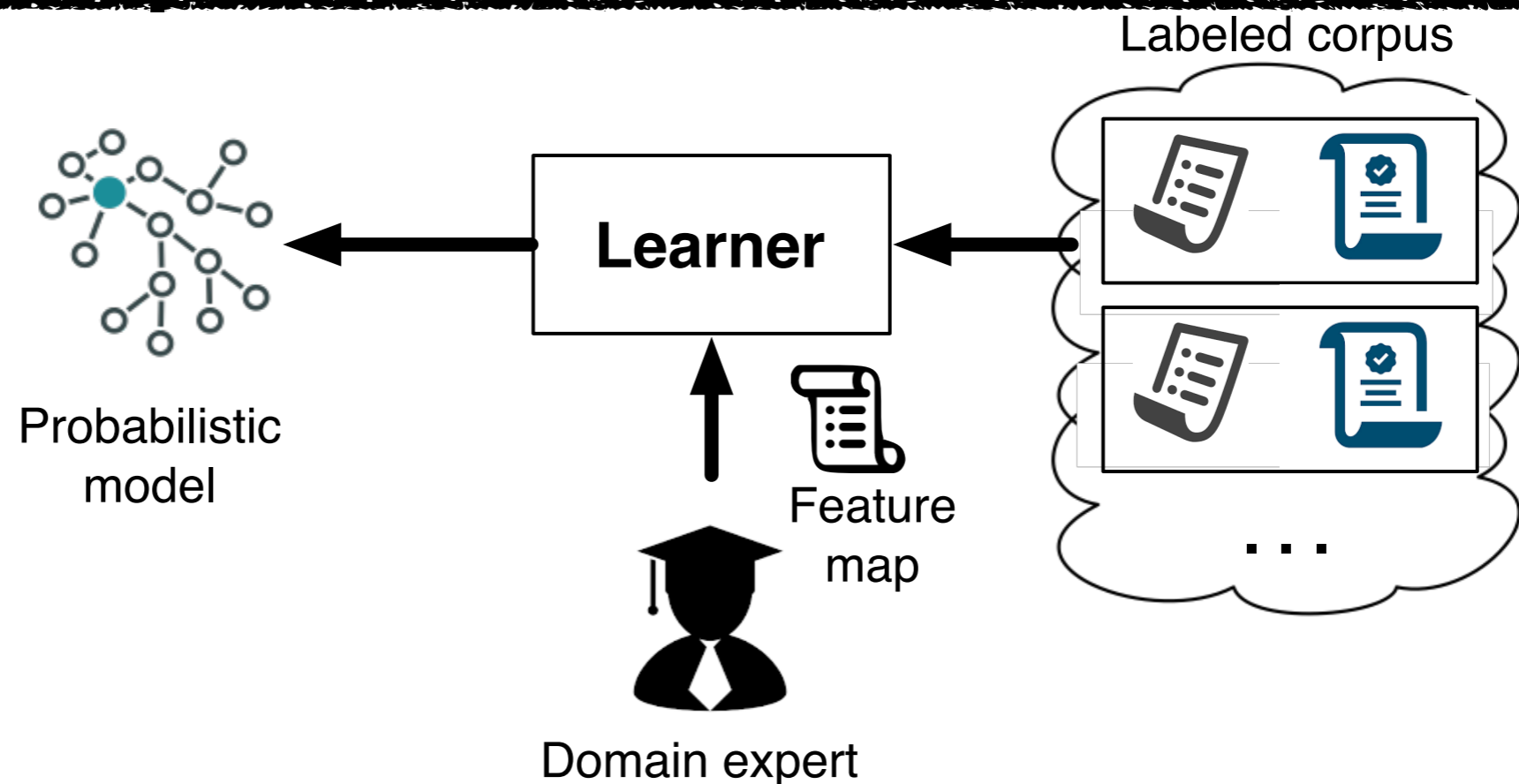


CEGIS with Guided Search



Transfer Learning

- Problem: **overfitting**
- Our solution: generalize to unseen programs better using a **feature map** designed by domain expert



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Example

- **Goal:** Replacing a hyphen (-) by a dot (.) in a given string x

-  **Specification**

Syntactic specification:

$S \rightarrow x \mid \text{"-"} \mid \text{"."} \mid S + S \mid \text{Rep}(S, S, S)$

String concatenation

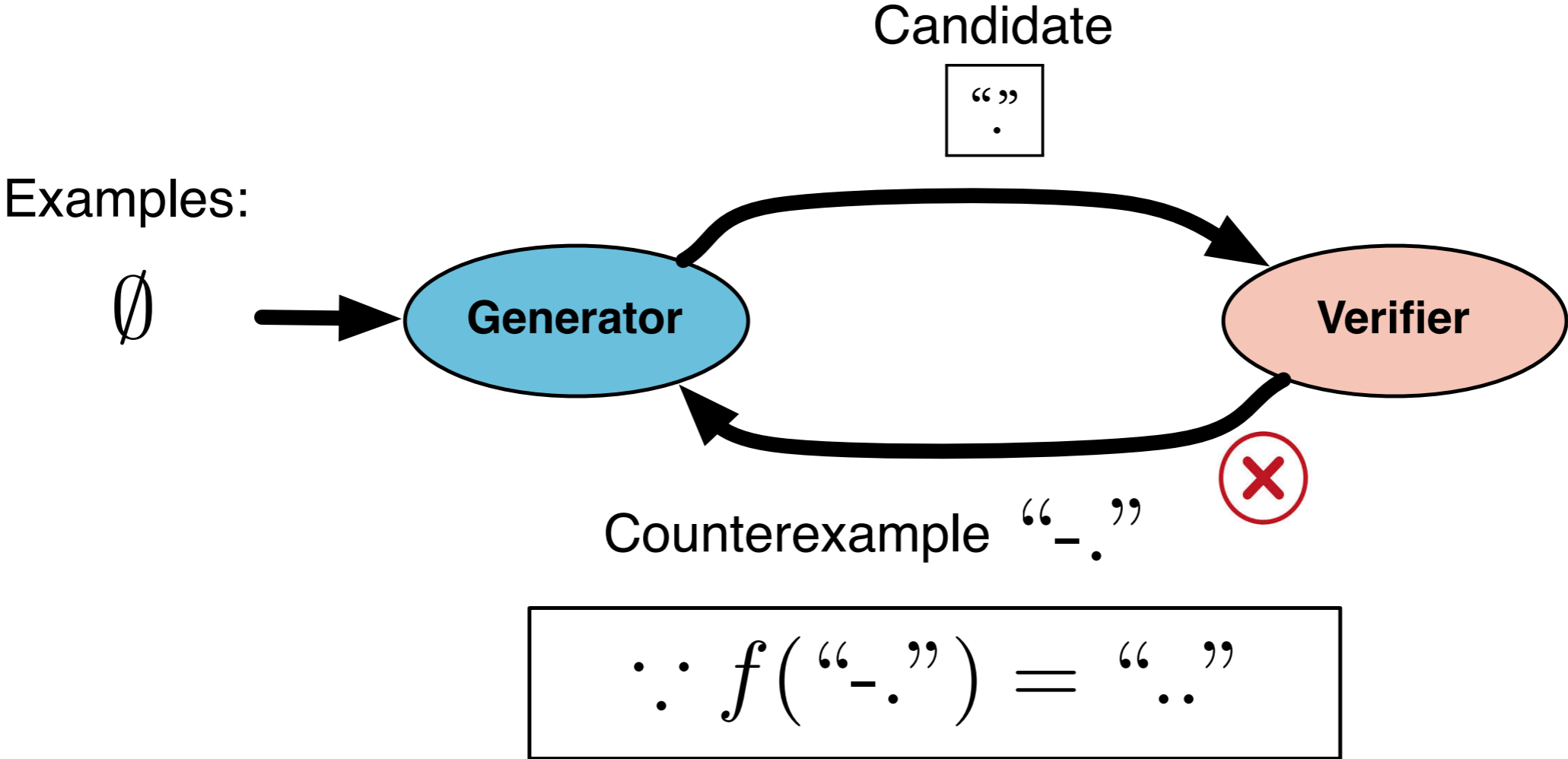
Semantic specification:

Rep(s, t1, t2): t1 in s is replaced by t2

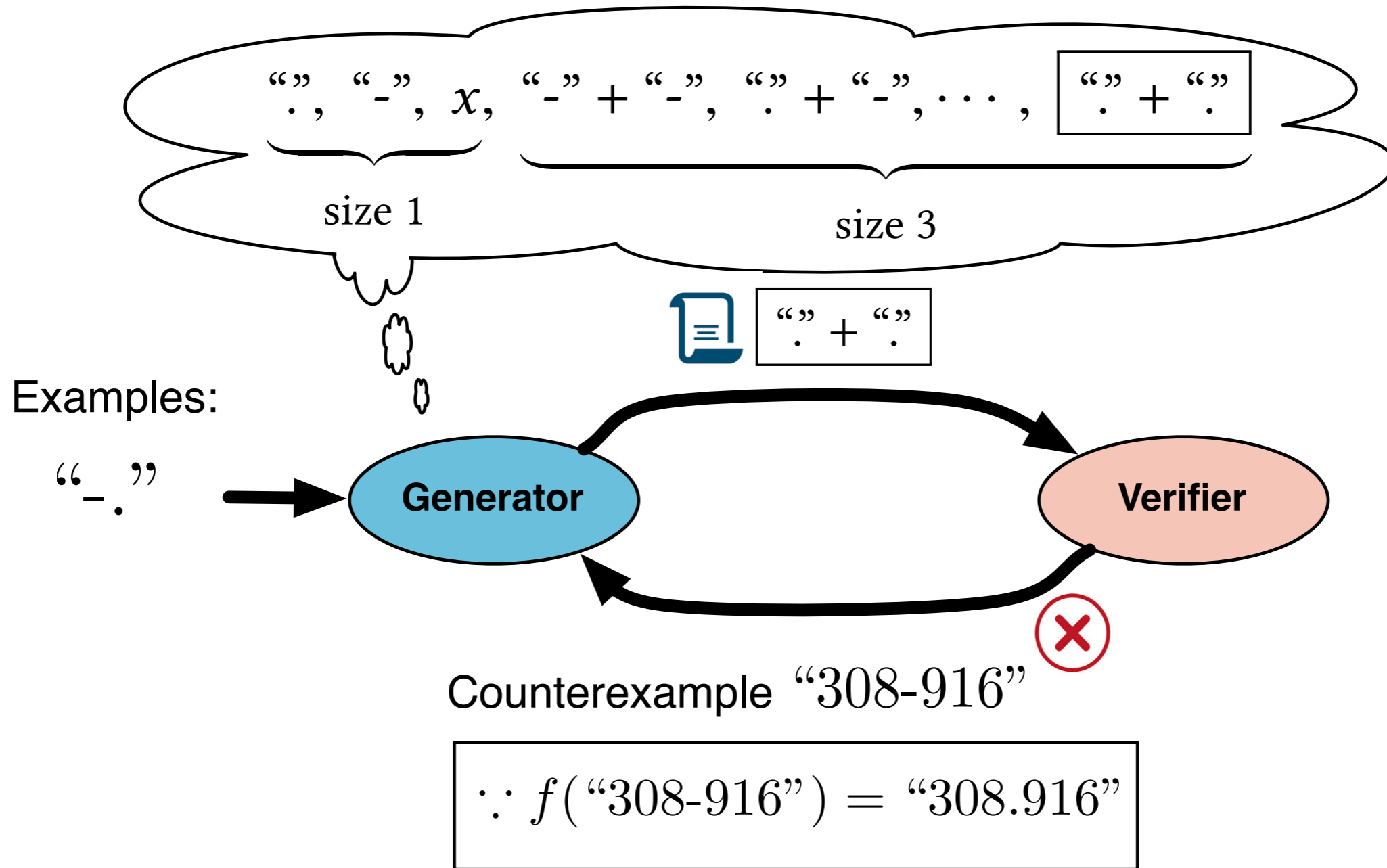
$f(\text{"-."}) = \text{".."} \wedge f(\text{"308-916"}) = \text{"308.916"} \wedge f(\text{"1"}) = \text{"1"}$

-  **Solution:** $\text{Rep}(x, \text{"-"}, \text{"."})$.

Enumerative Search: Unguided

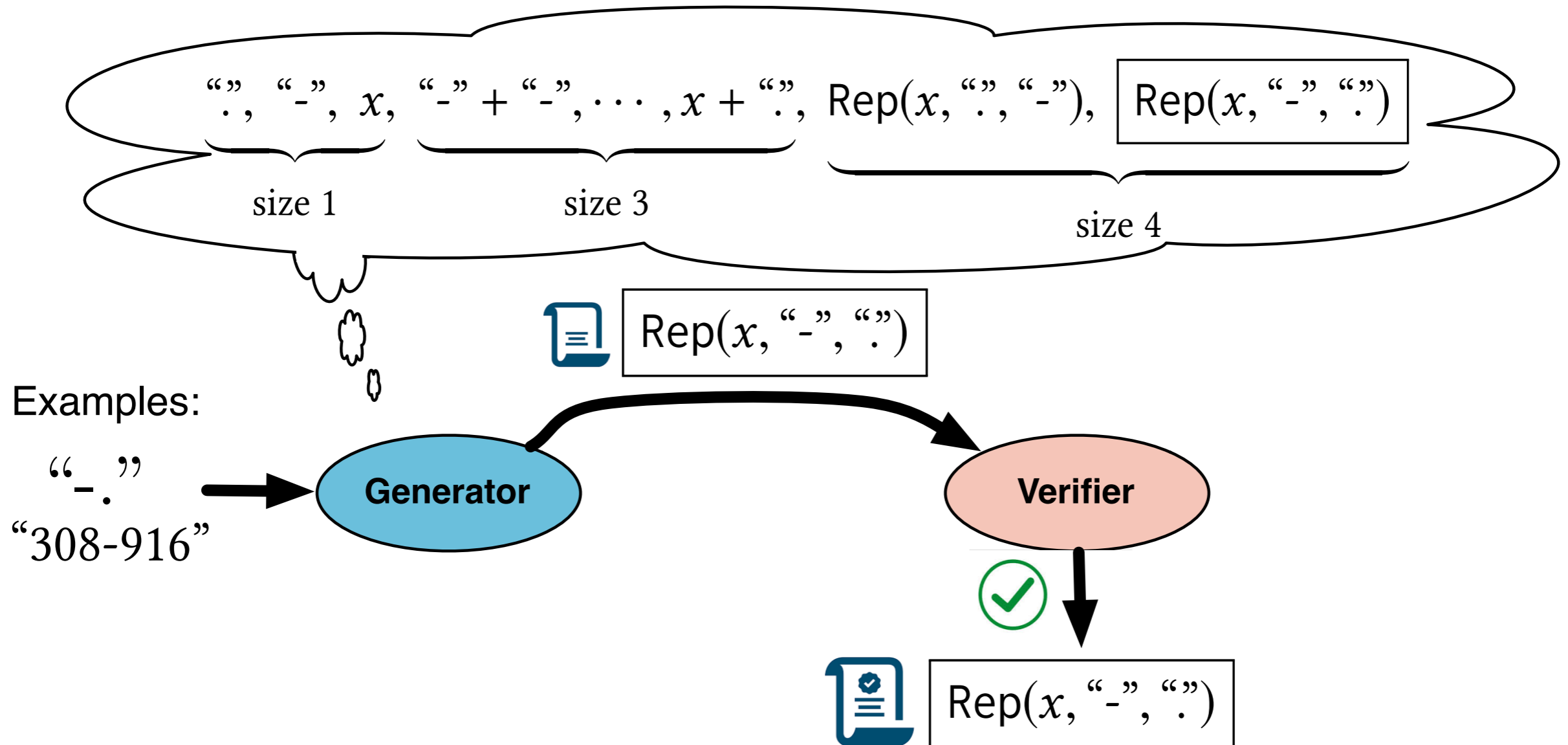


Enumerative Search: Unguided



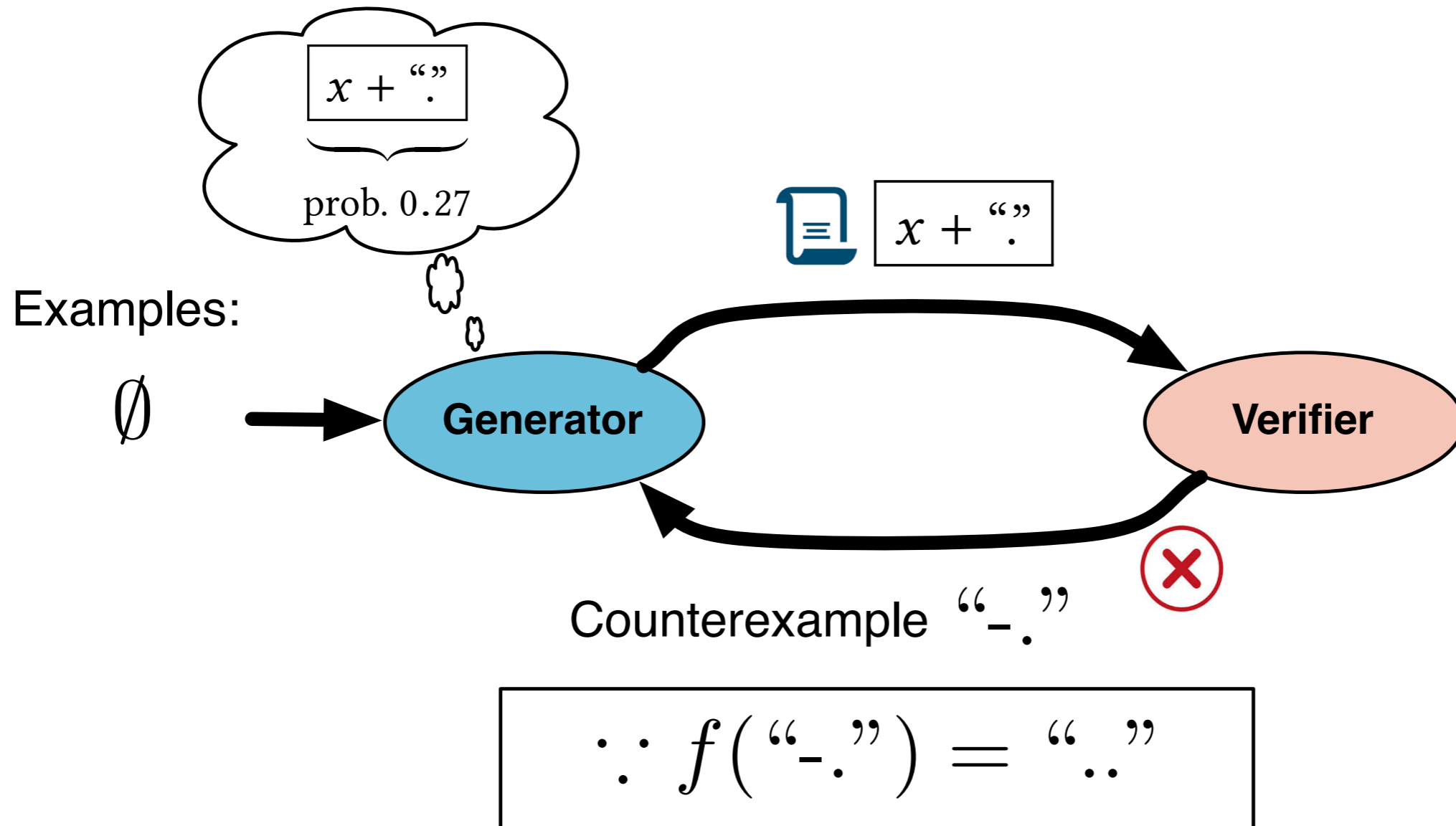
Pruning optimization: “-” + “.” is not explored.

Enumerative Search: Unguided



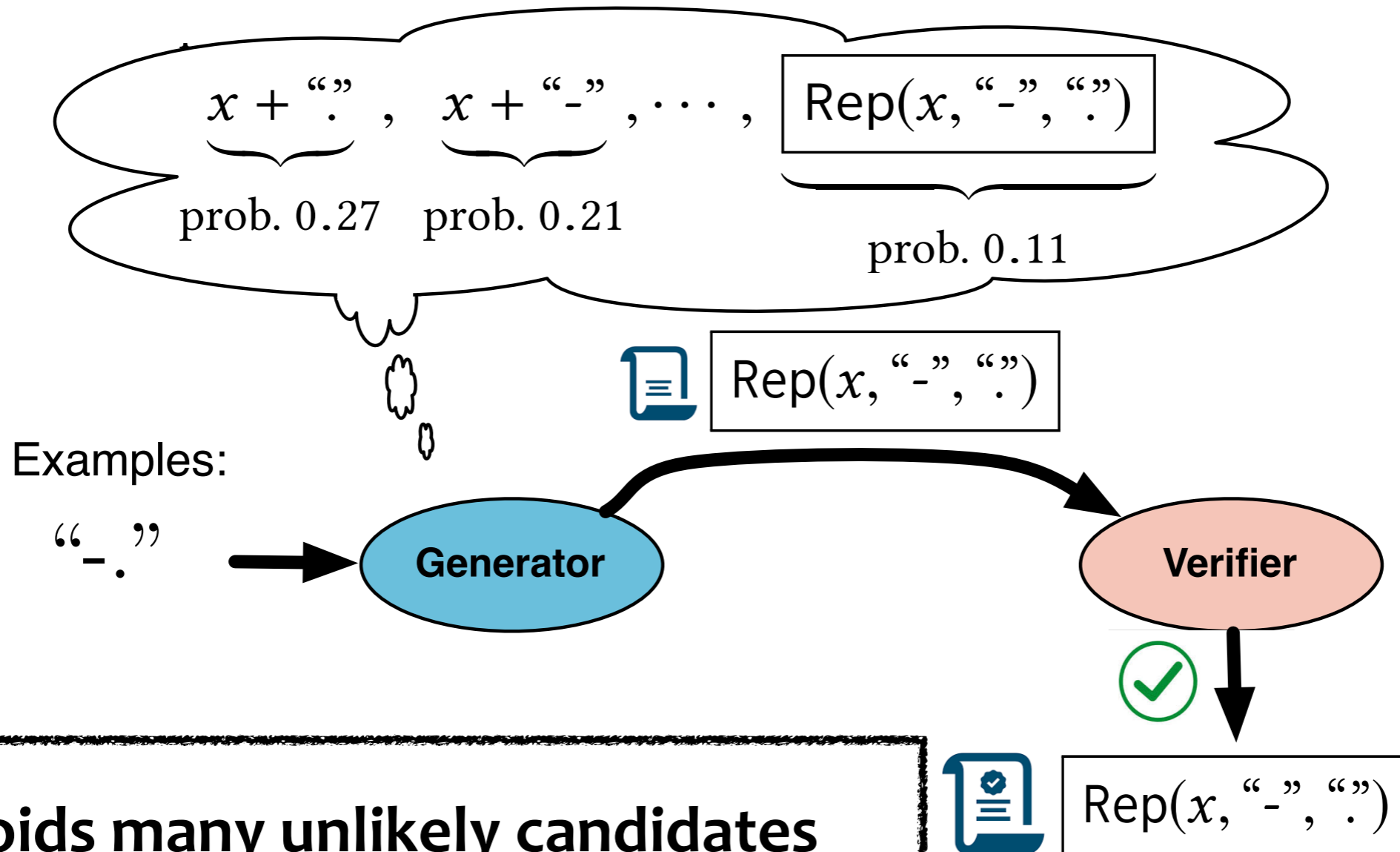
Many unlikely programs (e.g., `“.” + “.”`) are explored.

Enumerative Search: Guided



Enumerates in order of **likelihood** instead of **size**

Enumerative Search: Guided



- Avoids many unlikely candidates
- Preserves the pruning optimization

A Uniform Interface to Statistical Program Models

- Given a sequence of terminal/nonterminal symbols (i.e., sentential form), provide a probability for each production rule

$$Pr(S \rightarrow "." \mid \text{Rep}(x, "-", S)) = 0.72$$

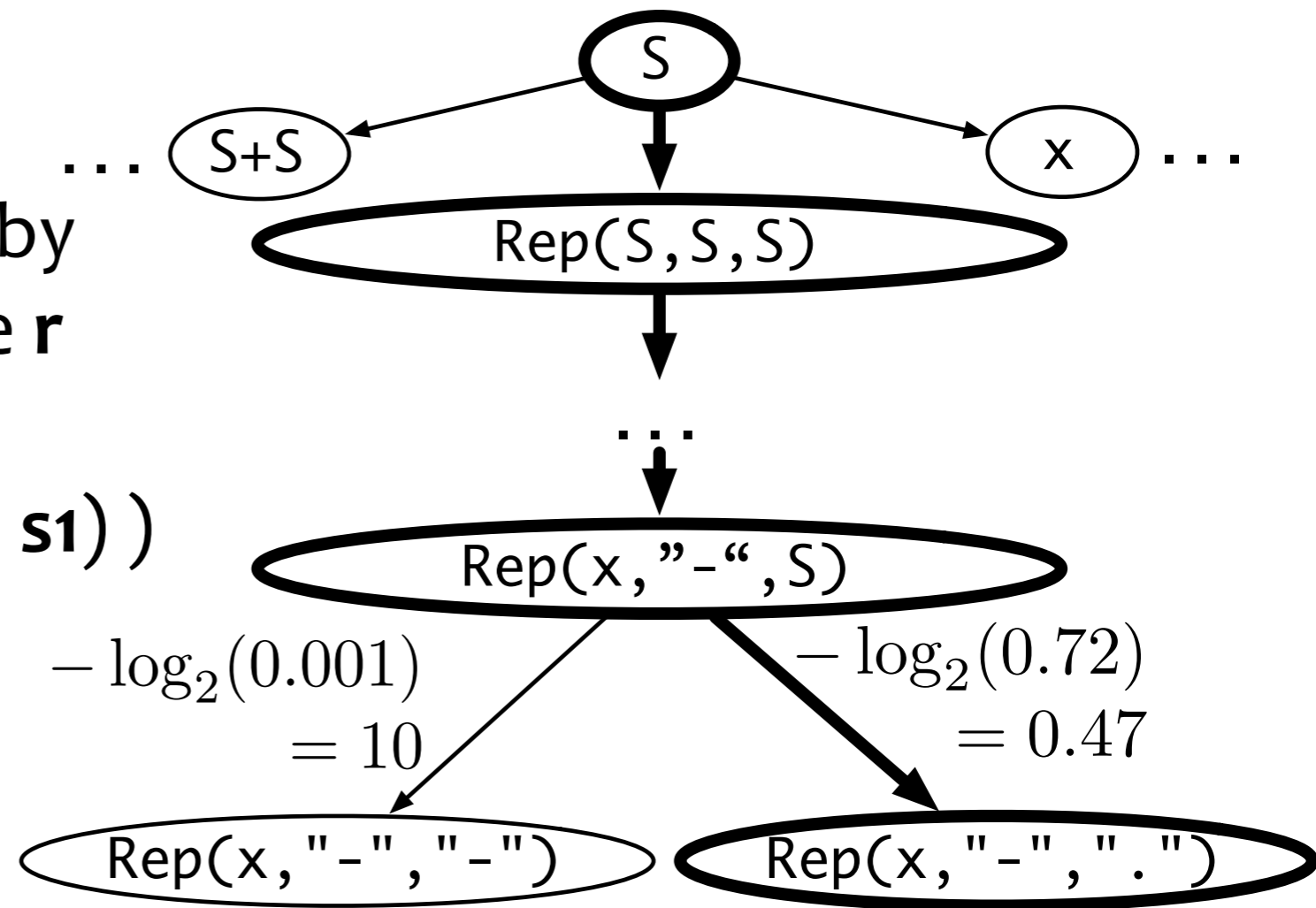
$$Pr(S \rightarrow "-" \mid \text{Rep}(x, "-", S)) = 0.001$$

...

Guided Enumeration via Path Finding

Given a model, we construct a directed graph.

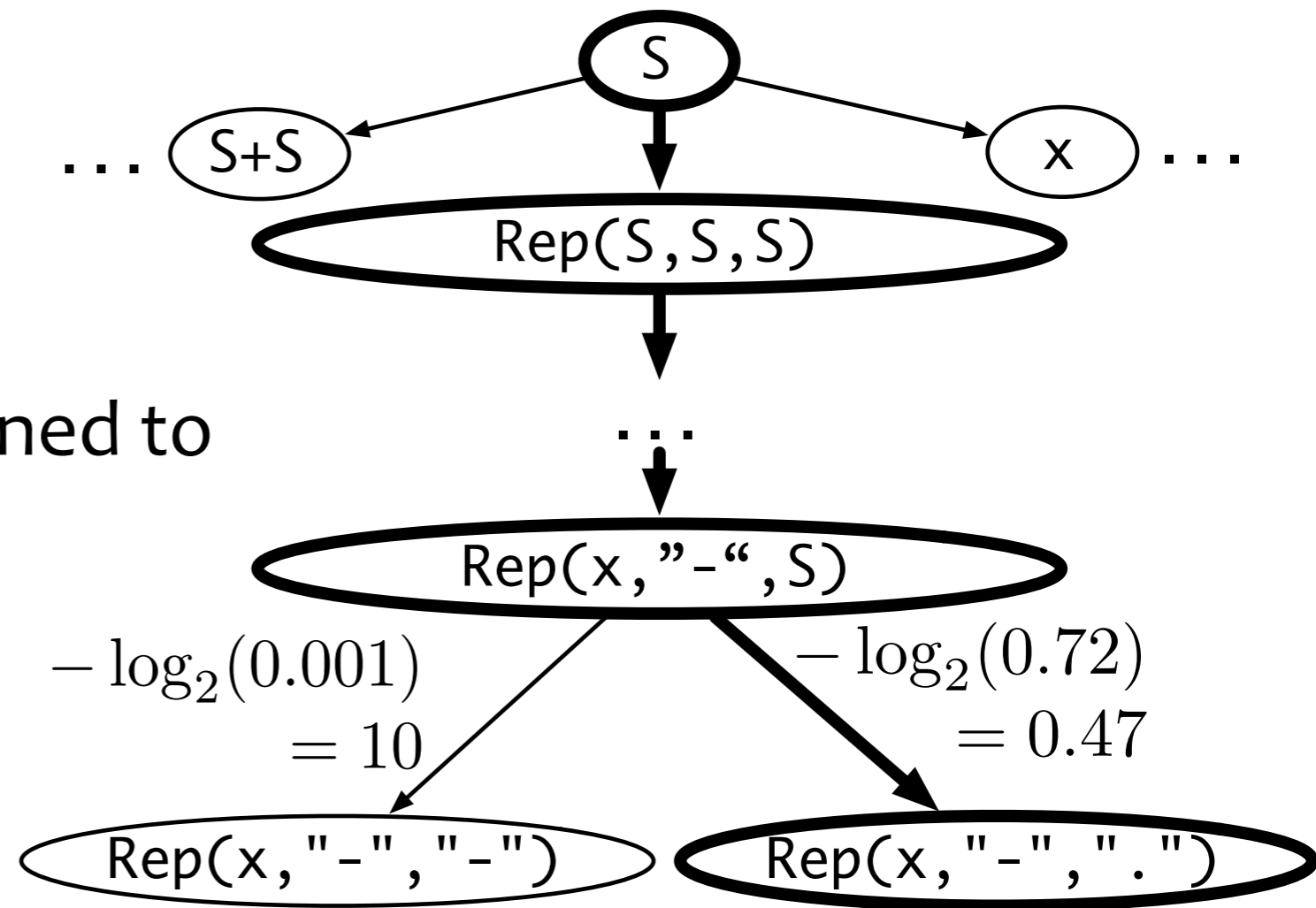
- Nodes: sentential forms
- $s_1 \xrightarrow{r} s_2$: s_1 expands to s_2 by applying a production rule r
- $w(s_1 \xrightarrow{r} s_2) = -\log (Pr(r | s_1))$



Guided Enumeration via Path Finding

Idea: solving a shortest pathfinding problem via A* search

- Start node: **S**
- Goal nodes: all programs
- A **heuristic function** designed to work with any model



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Evaluation Setup

- Benchmarks:
 - **1,167** problems from 2017 SyGuS competition and online forums
- Comparison to two baselines:
 - **EUSolver** (general-purpose): winner of 2017 SyGuS competition
 - **FlashFill** (domain-specific): string processing in spreadsheets

Benchmarks

	A	B	C	D
1	Number	Phone		
2	02082012225	020-8201-2225		
3	02072221236	020-7222-1236		
4	0208123654	020-8123-654		
5	0207236523	020-7236-523		
6	02082012222	020-8201-2222		
7				
8				

STRING: End-user Programming
205 problems

complement

```
~ 010100011101011100000000000001111
   1010111000101000111111111110000
```

bitwise and

```
010100011101011100000000000001111
& 00110001011011100011000101101110
   00010001010001100000000000001110
```

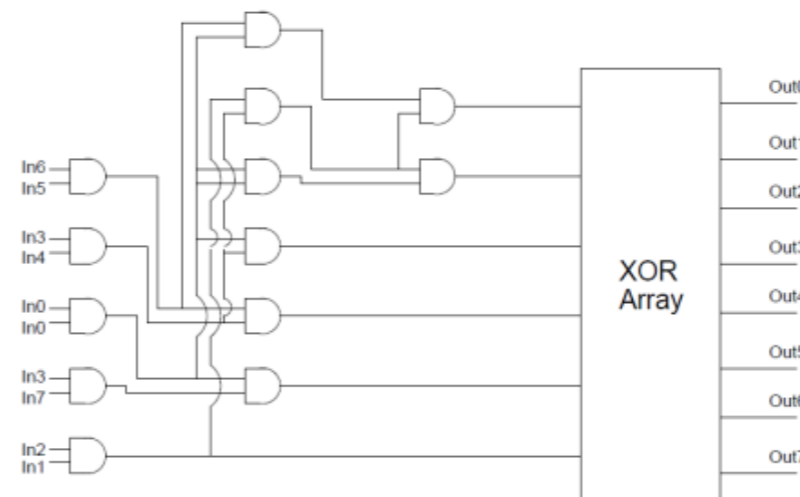
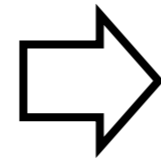
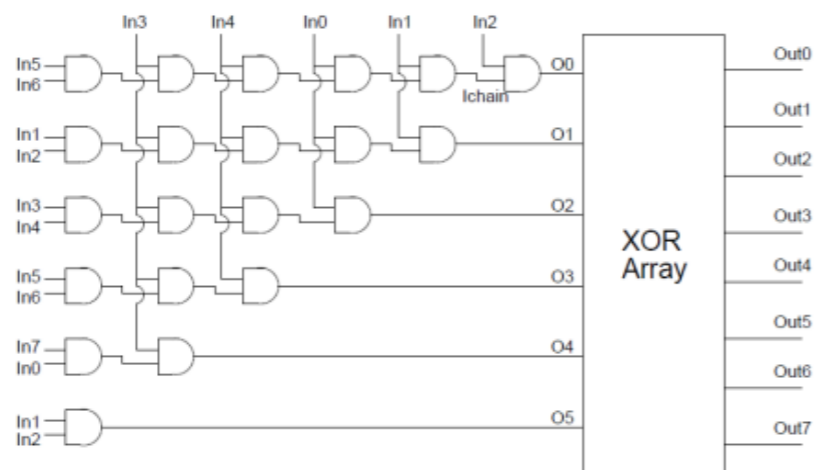
bitwise or

```
010100011101011100000000000001111
| 00110001011011100011000101101110
   011100011111111110011000101101111
```

bitwise xor

```
010100011101011100000000000001111
^ 00110001011011100011000101101110
   01100000101110010011000101100001
```

BITVEC: Efficient low-level algorithm
750 problems

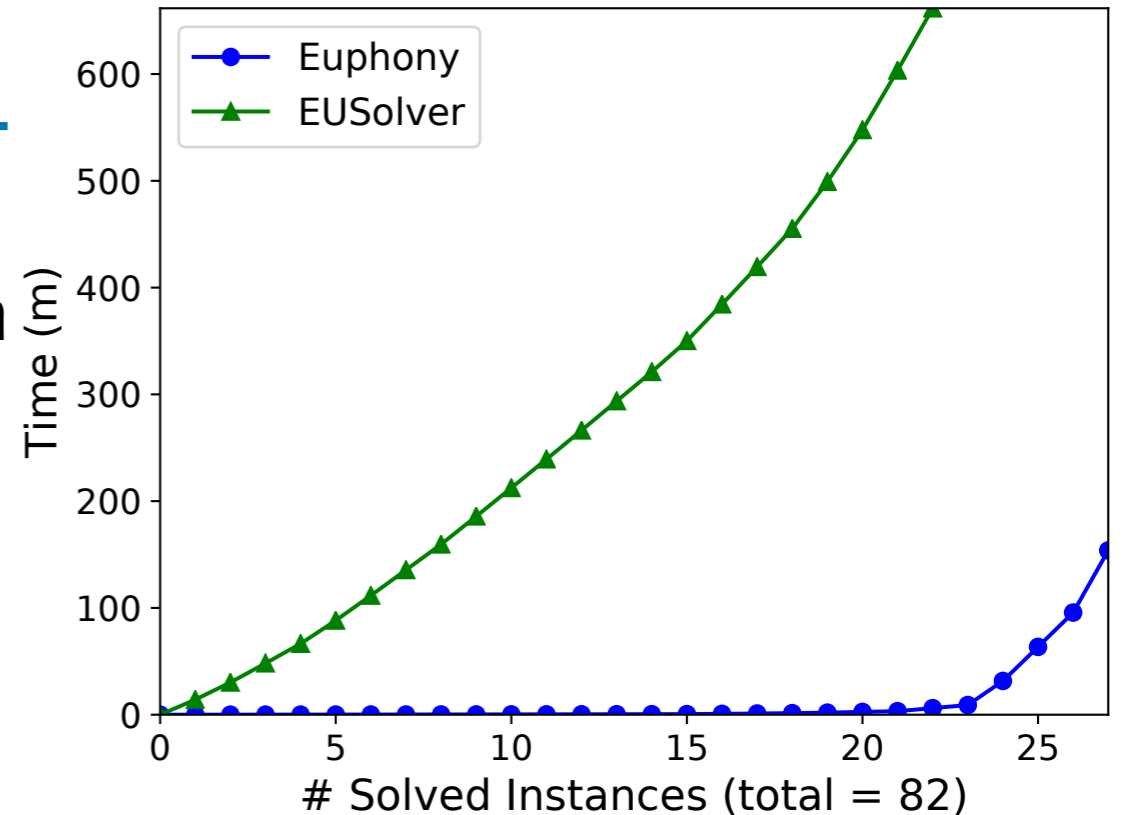


CIRCUIT: Attack-resistant crypto circuits
212 problems

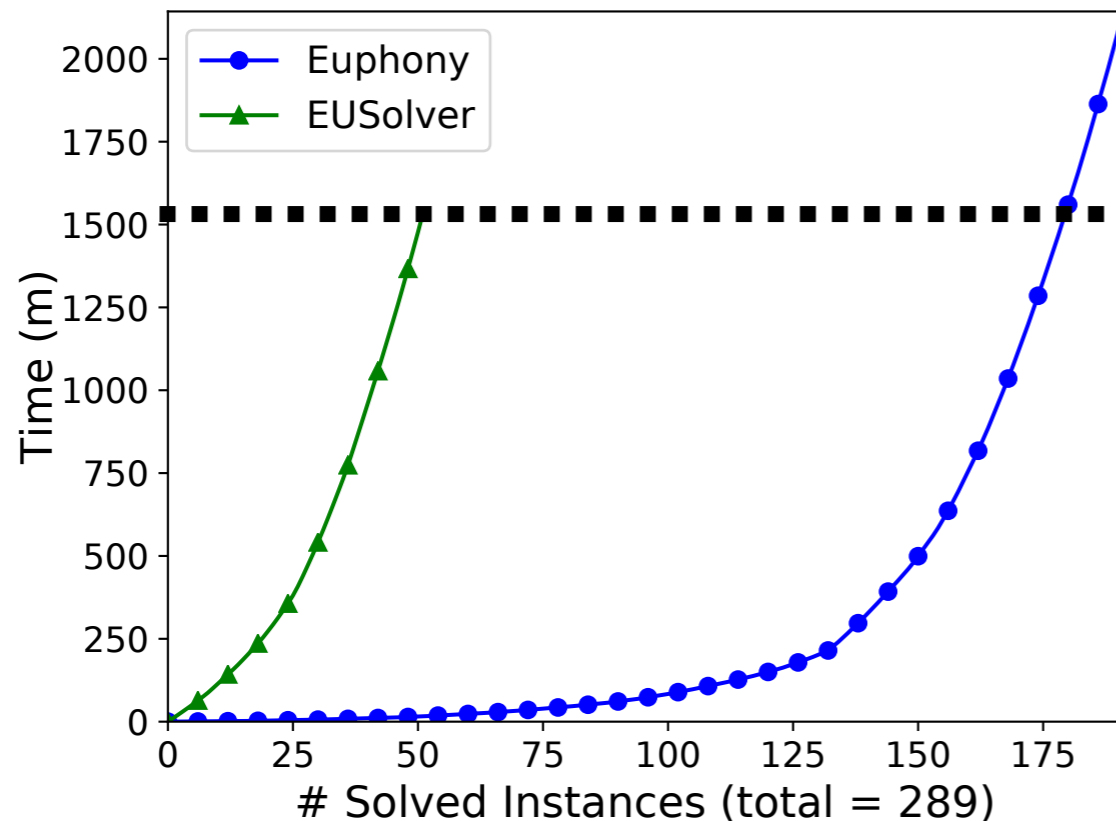
Comparison with EUSolver

- Training: **762** solved by EUSolver in 10 m
- Testing: **405** (timeout: 1 hour)
- # solved: Euphony **236**, EUSolver **87**

STRING



BITVEC

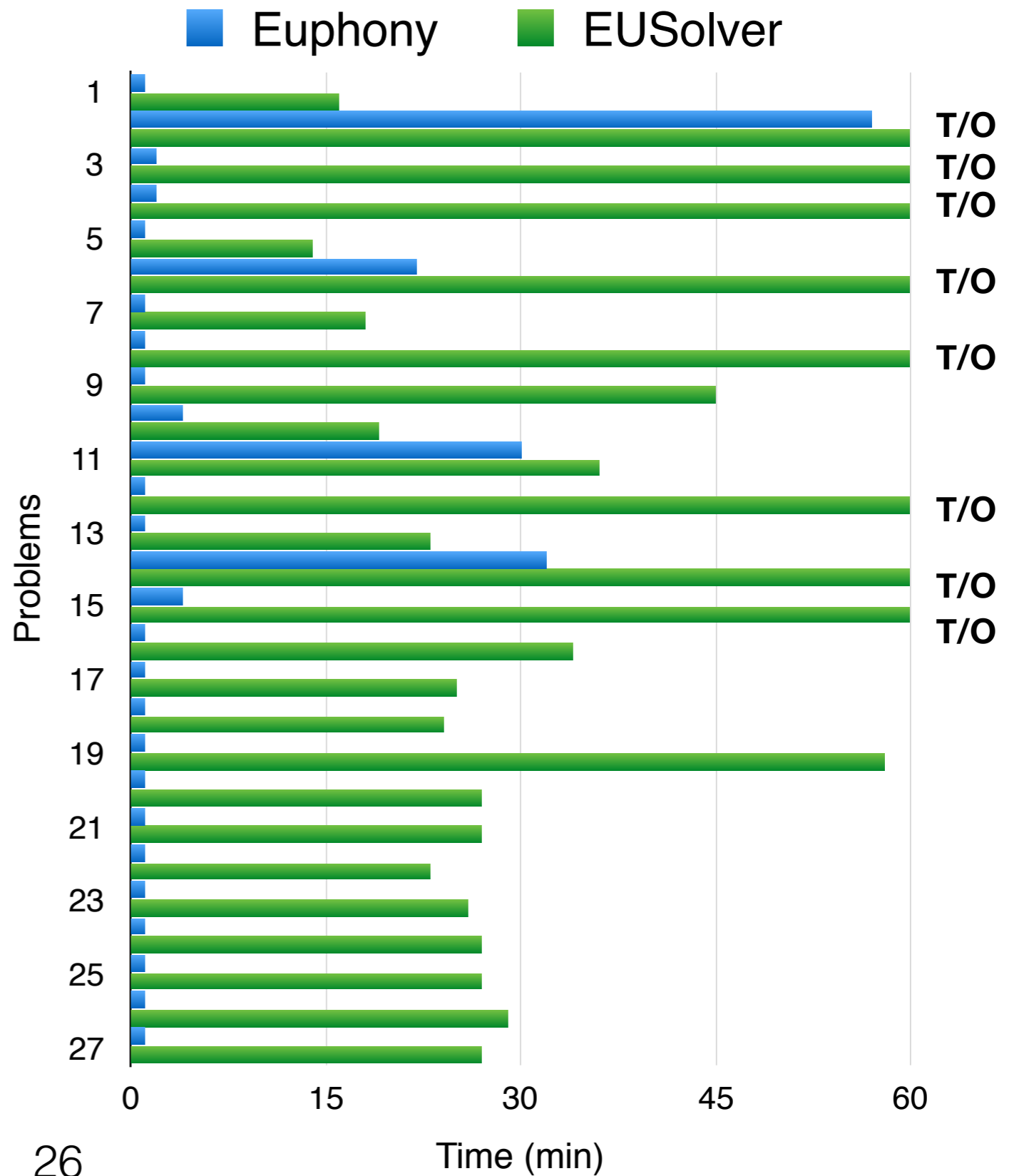


Result for STRING benchmarks

	A	B	C	D
1	Number	Phone		
2	02082012225	020-8201-2225		
3	02072221236	020-7222-1236		
4	0208123654	020-8123-654		
5	0207236523	020-7236-523		
6	02082012222	020-8201-2222		
7				
8				
9				

205 problems (training 123 / testing 82)

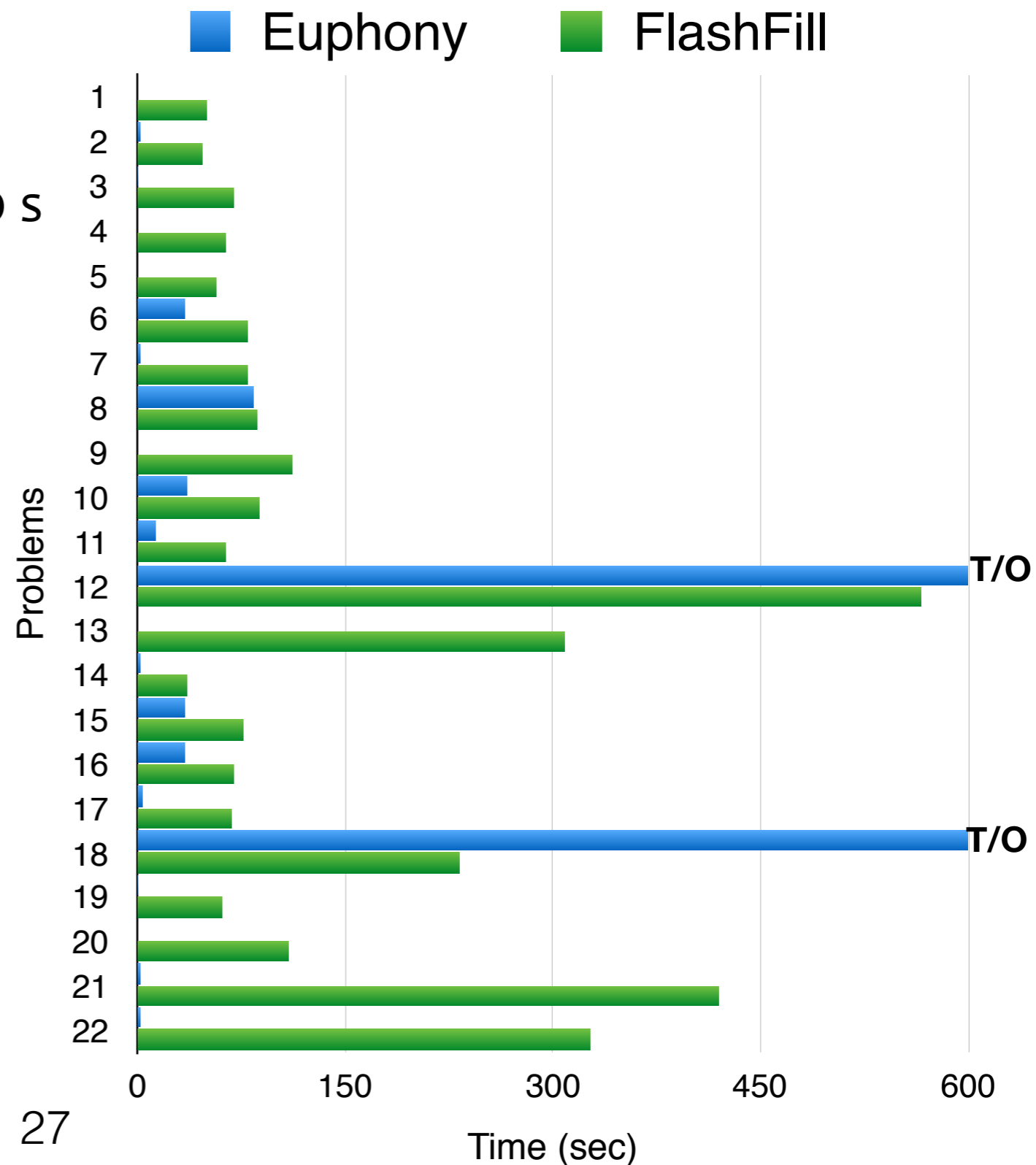
- Euphony solved **78%** within 1 min
- solved **8** on which EUSolver timed out
- outperformed EUSolver on all



Comparison with FlashFill (STRING)

- 113 problems handled by FlashFill
- Training: 91 solved by FlashFill in 10 s
- Testing: 22 (timeout: 10 m)
- Euphony outperforms in 20 / 22

	Average	Median
Euphony	13 s	3 s
Flashfill	140 s	78 s



In the paper ...

- General heuristic function for A* search
- How to preserve orthogonal search optimizations
- Feature maps for the three application domains
- Effectiveness of different models

Thank you!