V-Star: Learning Visibly Pushdown Grammars from Program Inputs

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Accepted by PLDI 2024
NJPLS 2024

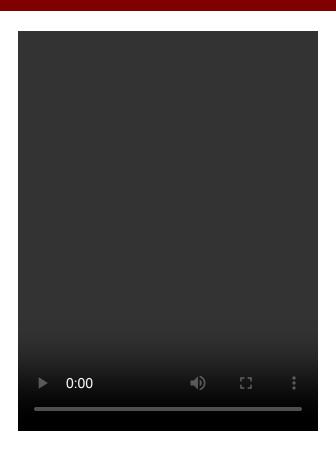
Overview

- Background
- Key Contributions of V-Star
- Evaluation
- Future Work

Program Input Learning

Infer formal grammars of inputs from black-box programs and sample valid inputs.

- Oracle: The black-box program
 - A calculator
 - JSON/XML parser binaries
 - Web server that accepts HTTP requests
- Seed Strings: Given valid inputs

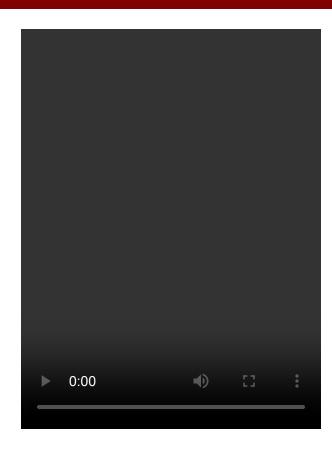


Importance

- Widely applicable
- Improves security and robustness

Application

- Grammar-based fuzzing
- Program validation
- Program comprehension
- Reverse engineering



Related Work

Existing Tools

- Glade (PLDI '17)
- Arvada (ASE '22)
- Learn context-free grammars

• Limitations:

- Limited accuracy for many practical grammars
- Do not fully utilize the *nesting* structures in program inputs

V-Star (PLDI '24)

- **Observation**: Using nesting structures can significantly increase model accuracy.
- Approach: Exploits nesting structures in program inputs to improve accuracy.

Nesting Structures

```
Program input:  Hello 
Tokenization: OPEN_TAG TEXT CLOSE_TAG TEXT
```

Recursion is delimited by special paired symbols, namely call symbols and return symbols.

```
Recursion

XML -> OPEN_TAG XML CLOSE_TAG XML | TEXT | ε

↑ ↑

Call Return
```

Nesting Structures

```
Program input :  Hello 
Tokenization : OPEN_TAG TEXT CLOSE_TAG TEXT
```

Recursion is delimited by special paired symbols, namely call symbols and return symbols.

```
Recursion Tail recursion does not count

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Nesting Structures

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↑ ↑

Call Return
```

Visibly Pushdown Grammars (VPGs) =

Regular Grammars + Nesting Structures

$$egin{array}{ll} L
ightarrow \epsilon \ L
ightarrow cA & ext{Regular Grammar} \ L
ightarrow aAbB & ext{Models Nesting Structures} \end{array}$$

Denote Sentences of VPGs

 Sentences in VPGs are normal strings with explicitly denoted call and return symbols, known as tagging.

```
 Hello  World!
```

• **V-Star** learns which substrings need to be colorized, also known as the *tagging function* for program inputs.

From Program Input Learning to Active Learning

• Oracle: Answers membership queries

$$\mathcal{O}: ext{string } s \mapsto \{ ext{true}, ext{false}\}$$

- Challenges: With finite queries, exact learning is not guaranteed.
- Minimally Adequate Teacher (MAT): Answers equivalence queries

```
\mathcal{E}: \operatorname{grammar} G \mapsto \{\operatorname{exact} 
oting s\}
```

• Active Learning: learn grammar from a MAT.

Exact Learning: Achievable?

| | Regular | VPG | CFG |
|---------------------|--------------------------------|------------|------------------|
| Positive Examples | Impossible | | |
| Positive + Negative | NP-complete | | |
| Membership Queries | Very Likely NP | | |
| MQ + Seed Strings | Very Likely NP | (V-Star) | (Glade & Arvada) |
| MAT (MQ + EQ) | Polynomial (Angluin's L^st) | Polynomial | Very Likely NP |

V-Star's Contribution

- Learns how to model program inputs as VPGs.
- Learns VPGs using active learning methods.

Key Contributions of V-Star

V-Star Workflow

1. Identify Call and Return Symbols:

- Use oracle and seed strings to infer nesting structures.
- Develop a tagging function to recognize call and return symbols.

2. Learn VPA and Convert to VPG:

- Use an L*-like algorithm to learn a Visibly Pushdown Automaton (VPA).
- Convert the VPA into a Visibly Pushdown Grammar (VPG).

V-Star Example: Arithmetic Formula

Seed String:

$$(1+(2\times3)/4)$$

What are the Call and Return Symbols?

Hypothesize recursion as:

```
expr -> "(" expr ")" expr | number | ...
```

• Therefore, (and) are the call and return symbols.

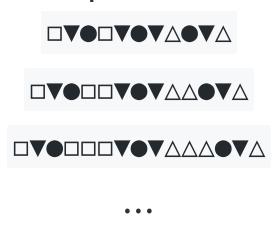
V-Star Example: Arithmetic Formula

Seed String (Encrypted):



What are the Call and Return Symbols?

• Nesting Patterns: Two substrings (x, y) that can be repeated at the same time, and must be repeated at the same time.



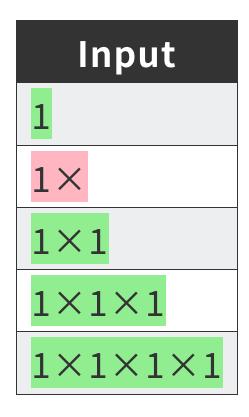
V-Star Example: Arithmetic Formula

- From valid strings $\neg \bullet$ (\neg) k $\bullet \bullet \bullet$, the nesting pattern is $(u,x,z,y,v)=(\neg \bullet \bullet$, \neg , $\bullet \bullet \bullet$), or simply $(x,y)=(\Box,\triangle)$
- Lemma (V-Star): each nesting pattern (x, y) must contain a call symbol in x, and a return symbol in y.
- Therefore, □ and △ are the call and return symbols.



The tagging function: For any program input, tags □ as call, and
 △ as return symbols.





- Conjecture: regular expression $(1 \times)^*$ 1 specifies valid inputs.
- You have learned a formal grammar

- Learn regular expression
 - Regular expression is equivalent to Finite State Automata (FSA).
- ullet \rightarrow Learn FSA
 - Each state in the automata is an equivalence class.
- → Learn Equivalence Classes
 - $\circ \to \mathsf{Angluin's}\ L^*$ (1987): There is a table-based method!

Fill the Prefix-Suffix Table

| Prefix | Suffix | | | | | | |
|-----------------------|-----------------------|------------------------|--------------------------------|--|--|--|--|
| PIEIIX | ε | 1 | × 1 | | | | |
| € | E | 1 | × 1 | | | | |
| 1 | 1 | 11 | 1×1 | | | | |
| 1 × | 1 × | 1×1 | $1 \times \times 1$ | | | | |
| 1 × 1 | 1×1 | 1 × 11 | $1 \times 1 \times 1$ | | | | |
| $1 \times 1 \times 1$ | $1 \times 1 \times 1$ | $1 \times 1 \times 11$ | $1 \times 1 \times 1 \times 1$ | | | | |

Fill the Prefix-Suffix Table

| Prefix | | Suffix | |
|-----------------------|---|--------|-----|
| I I CIIX | € | 1 | × 1 |
| E | | | |
| 1 | | | |
| 1 × | | | |
| 1 × 1 | | | |
| $1 \times 1 \times 1$ | | | |

• Each color sequence represents an equivalence class.

Fill the Prefix-Suffix Table

| Prefix | Suffix | | | | | | | |
|-----------------------|--------|---|-----|--|--|--|--|--|
| PICIIX | ε | 1 | × 1 | | | | | |
| E | | | | | | | | |
| 1 | | | | | | | | |
| 1 × | | | | | | | | |
| 1 × 1 | | | | | | | | |
| $1 \times 1 \times 1$ | | | | | | | | |

• ϵ is equivalent to 1 \times .

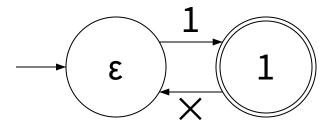
Fill the Prefix-Suffix Table

| Prefix | | Suffix | |
|-----------------------|---|--------|-----|
| PICIIX | ε | 1 | × 1 |
| E | | | |
| 1 | | | |
| 1 × | | | |
| 1 × 1 | | | |
| $1 \times 1 \times 1$ | | | |

• 1 is equivalent to 1 × 1 and 1 × 1 × 1.

Build an FSA

• The equivalence classes can be convert to an FSA directly.



• The FSA above is equivalent to regular expression $(1 \times)^* 1$.

Doesn't work for VPG

| Prefix | Suffix | | | | | | | |
|--------|--------|------------------|----------------------------|-------|-------------------------------------|-------|--|--|
| Prenx | € | ×1) | $\times 1) \times 1)$ | ••• | ×1)×1) | ••• | | |
| E | E | ×1) | ×1) ×1) | • • • | \times 1) \times 1) \times 1) | • • • | | |
| (1 | (1 | (1×1) | $(1 \times 1) \times 1)$ | • • • | $(1 \times 1) \times 1) \times 1)$ | • • • | | |
| ((1 | ((1 | $((1 \times 1)$ | $((1 \times 1) \times 1)$ | • • • | $((1 \times 1) \times 1) \times 1)$ | • • • | | |
| • • • | • • • | • • • | • • • | • • • | • • • | ••• | | |
| ((1 | (((1 | $(((1 \times 1)$ | $(((1 \times 1) \times 1)$ | • • • | ((1 ×1) ×1) | ••• | | |
| • • • | • • • | • • • | • • • | • • • | • • • | • • • | | |

• Infinite number of equivalence classes.

| Infix | Left | Left Context, Right Context | | | | | |
|----------------|----------------|-----------------------------|---------------------------|--|--|--|--|
| IIIIIX | €, € | ε, × 1 | (, × 1) | | | | |
| E | E | × 1 | (× 1) | | | | |
| 1 | 1 | 1 × 1 | (1×1) | | | | |
| 1 × | 1 × | $1 \times \times 1$ | $(1 \times \times 1)$ | | | | |
| (1×1) | (1×1) | $(1 \times 1) \times 1$ | $((1 \times 1) \times 1)$ | | | | |

- Each infix must be well-matched.
- The number of equivalence classes becomes finite.

More Info in the Paper

• Hierarchy:

- Regular Grammar < VPGs < CFGs
- Formal Definition of VPGs and VPA
- Conversions and Learning:
 - From Finite State Automata (FSA) to Visibly Pushdown Automata (VPA)
 - How to identify nesting structures using nesting patterns
 - How to learn Visibly Pushdown Automata (VPA)

• Partial Tokenizer:

- What is it?
- Our How does it help?

Evaluation Overview

Evaluation Methodology

- Replicated the artifact of Arvada (ASE '22).
- o Includes oracle grammars, evaluation datasets, and seed strings.

Selected Grammars

- Five Grammars: JSON, LISP, XML, While, MathExpr
- Chosen because they are VPGs

Further exploration of additional program inputs is reserved for future research.

Evaluation: Accuracy

- Recall $\frac{|L\cap L_{\mathcal{O}}|}{|L|}$: Probability that a string of the oracle is accepted by the learned grammar G.
- ullet Precision $rac{|L\cap L_{\mathcal{O}}|}{|L_{\mathcal{O}}|}$: Probability that a string in G is accepted by the oracle.
- F-1 Score $\frac{2}{1/\text{recall}+1/\text{prec}}$: Harmonic mean of precision and recall, indicating overall accuracy.

| Crammar | GLADE | | Arvada | | | V-Star | | | |
|----------|--------|------|--------|-----------------|-----------------|-----------------|--------|------|------|
| Grammar | Recall | Prec | F-1 | Recall | Prec | F-1 | Recall | Prec | F-1 |
| JSON | 0.42 | 0.98 | 0.59 | 0.97 ± 0.09 | 0.92 ± 0.08 | 0.94 ± 0.05 | 1.00 | 1.00 | 1.00 |
| Lisp | 0.23 | 1.00 | 0.38 | 0.38 ± 0.26 | 0.95 ± 0.08 | 0.50 ± 0.18 | 1.00 | 1.00 | 1.00 |
| XML | 0.26 | 1.00 | 0.42 | 0.99 ± 0.02 | 1.00 ± 0.00 | 1.00 ± 0.01 | 1.00 | 1.00 | 1.00 |
| While | 0.01 | 1.00 | 0.02 | 0.91 ± 0.20 | 1.00 ± 0.00 | 0.94 ± 0.14 | 1.00 | 1.00 | 1.00 |
| MathExpr | 0.18 | 0.98 | 0.31 | 0.72 ± 0.24 | 0.96 ± 0.03 | 0.80 ± 0.16 | 1.00 | 1.00 | 1.00 |

Evaluation: Efficiency

• #Queries: The number of unique membership queries made during the learning process.

• **Time**: The overall running time for each tool.

• #Seeds: The number of seed strings (i.e., valid strings) shared by all tools.

| Crammar | | GLADE | | Arva | da | V-Star | | |
|----------|--------|----------|------|-------------------------|----------------|----------|--------|--|
| Grammar | #Seeds | #Queries | Time | #Queries | Time | #Queries | Time | |
| JSON | 71 | 11 K | 21 s | 6.8 K ± 394 | 25 s ± 2 s | 541 K | 33 min | |
| Lisp | 26 | 3.8 K | 7 s | $2.2~{\rm K}~{\pm}~307$ | 8 s ± 2 s | 16 K | 77 s | |
| XML | 62 | 15 K | 21 s | 12 K \pm 1 K | $61 s \pm 5 s$ | 208 K | 16 min | |
| While | 10 | 9.2 K | 13 s | 5.4 K ± 563 | 15 s ± 1 s | 1440 K | 1.5 h | |
| MathExpr | 40 | 19 K | 42 s | 6.6 K ± 421 | 24 s ± 2 s | 4738 K | 6 h | |

Evaluation: More Statistics of V-Star

- %Queries(Token): Percentage of membership queries made for token inference.
- %Queries(VPA): Percentage of membership queries made for VPA learning.
- #TestString: The number of test strings sampled from the seed strings by V-Star.

| Grammar | #Queries | %Q(Token) | %Q(VPA) | #TestString | Time |
|----------|----------|-----------|---------|-------------|--------|
| JSON | 541 K | 2.71% | 97.29% | 8043 | 33 min |
| Lisp | 16 K | 1.37% | 98.63% | 693 | 77 s |
| XML | 208 K | 94.93% | 5.07% | 682 | 16 min |
| While | 1440 K | 9.40% | 90.60% | 119 | 1.5 h |
| MathExpr | 4738 K | 0.11% | 99.89% | 2602 | 6 h |

Most time is spent on VPA learning, not token inference, as short seed strings lead to shorter nesting patterns and a smaller search space.

Conclusion and Future Work

• **Conclusion**: V-Star is more accurate compared to other tools but requires more time due to VPA learning.

• Future Work:

- **Performance**: Improve V-Star's efficiency to reduce VPA learning time.
- Evaluation: Test V-Star on more practical grammars.
- Alternative Algorithms: Explore other VPA learning methods.
- Readability: Enhance the readability of the learned grammar.
- CFG Learning: Use V-Star as a starting point for learning Context-Free Grammars.