# V-Star: Learning Visibly Pushdown Grammars from Program Inputs

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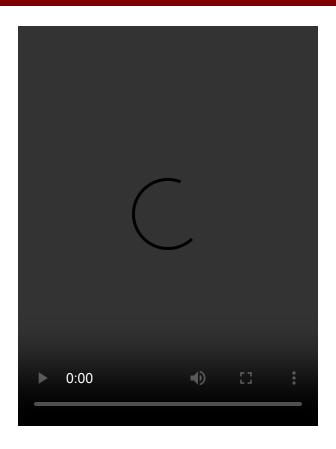
#### Overview

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

# **Program Input Learning**

Infer grammars of input 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

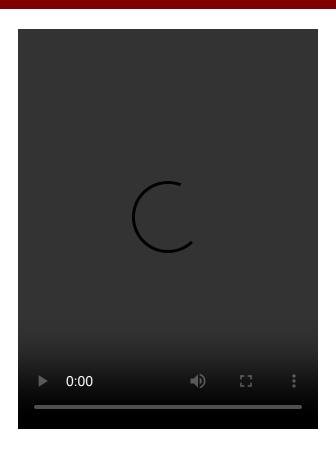


#### **Importance**

- 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 Tail recursion does not count

XML -> OPEN_TAG XML CLOSE_TAG XML | TEXT | ε

Λ Λ

Call Return
```

# **Nesting Structures**

```
Recursion Tail recursion does not count

XML -> OPEN_TAG XML CLOSE_TAG XML | TEXT | ε

↑ ↑

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} \ggg, \operatorname{counterexample string} 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.

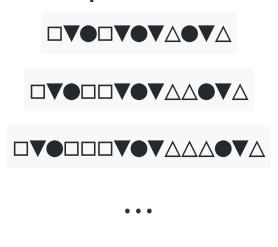
#### V-Star Example: Arithmetic Formula

## **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.



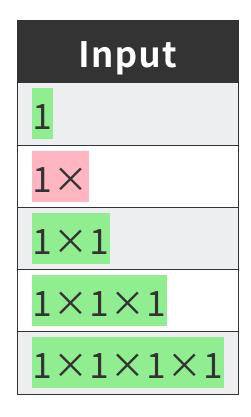
## V-Star Example: Arithmetic Formula

- 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 the first 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
  - $\circ \to \mathsf{Angluin's}\ L^*$  (1979): There is a table-based method!

#### Fill the Prefix-Suffix Table

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

• Each color sequence represents an equivalence class.

#### Fill the Prefix-Suffix Table

Prefix		Suffix	
FICHA	ε	1	× 1
E			
1			
1 ×			
1 × 1			
$1 \times 1 \times 1$			

•  $\epsilon$  is equivalent to 1  $\times$  .

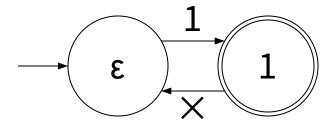
#### Fill the Prefix-Suffix Table

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	$(1 \times 1)$	$(1 \times 1) \times 1)$	• • •	$(1 \times 1) \times 1) \times 1)$	• • •		
((1	((1	$((1 \times 1)$	$((1 \times 1) \times 1)$	• • •	$((1 \times 1) \times 1) \times 1)$	• • •		
• • •	• • •	• • •	• • •	• • •	• • •	•••		
((1	(((1	$(((1 \times 1)$	$(((1 \times 1) \times 1)$	• • •	((1 ×1) ×1)	•••		
• • •	• • •	• • •	• • •	• • •	• • •	• • •		

• Infinite number of equivalence classes.

Infix	ht Context		
IIIIIX	€, €	ε, × 1	(, × 1)
E	E	× 1	( × 1)
1	1	1 × 1	$(1 \times 1)$
1 ×	1 ×	$1 \times \times 1$	$(1 \times \times 1)$
$(1 \times 1)$	$(1 \times 1)$	$(1 \times 1) \times 1$	$((1 \times 1) \times 1)$

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

## **More Info in the Paper**

#### • Hierarchy:

- Regular Grammar < VPGs < CFGs</li>
- 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	$0.97 \pm 0.09$	$0.92 \pm 0.08$	0.94 ± 0.05	1.00	1.00	1.00
Lisp	0.23	1.00	0.38	$0.38 \pm 0.26$	$0.95 \pm 0.08$	$0.50 \pm 0.18$	1.00	1.00	1.00
XML	0.26	1.00	0.42	$0.99 \pm 0.02$	$1.00 \pm 0.00$	$1.00 \pm 0.01$	1.00	1.00	1.00
While	0.01	1.00	0.02	$0.91 \pm 0.20$	$1.00 \pm 0.00$	$0.94 \pm 0.14$	1.00	1.00	1.00
MathExpr	0.18	0.98	0.31	$0.72 \pm 0.24$	$0.96 \pm 0.03$	$0.80 \pm 0.16$	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.

Ckammak		GLADE		Arva	da	V-Star		
Grammar	#Seeds	#Queries	Time	#Queries	Time	#Queries	Time	
JSON	71	11 K	21 s	$6.8~{ m K}\pm394$	25 s ± 2 s	541 K	33 min	
Lisp	26	3.8 K	7 s	2.2 K ± 307	8 s ± 2 s	16 K	77 s	
XML	62	15 K	21 s	12 K $\pm$ 1 K	61 s ± 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.