

Interval Counterexamples for Loop Invariant Learning

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① Background

② Motivation

③ Approach

④ Evaluation

⑤ Conclusion

1 Background

2 Motivation

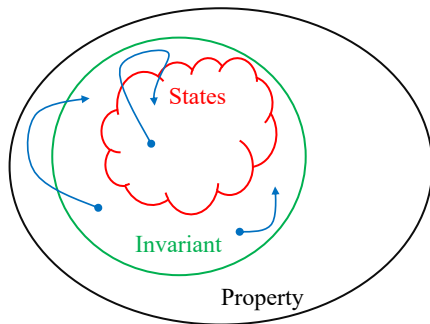
3 Approach

4 Evaluation

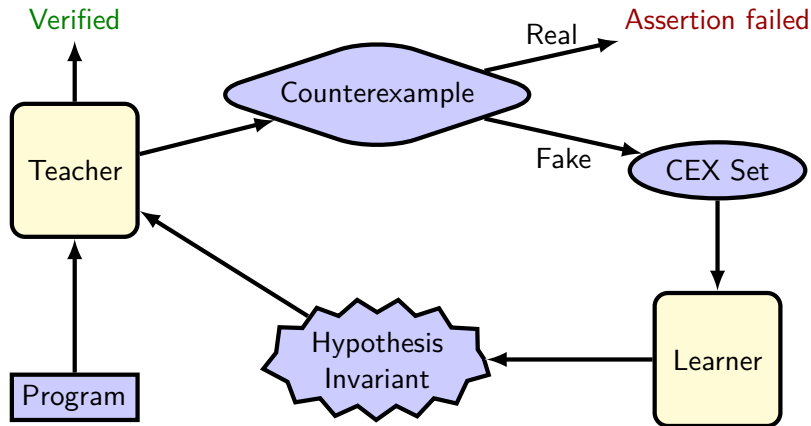
5 Conclusion

Loop invariants

- The loop invariant is a property that is always maintained when the program enters, iterates, and exits the loop.

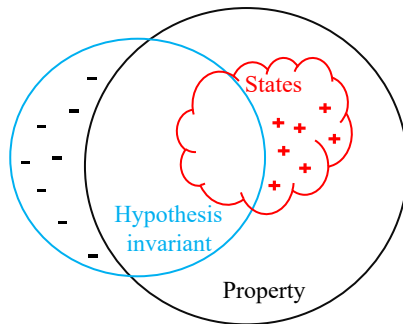


Black-box loop invariant inference



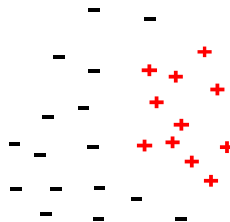
Positive and negative counterexamples

- Positive counterexample weakens the hypothesis invariant.
- Negative counterexample strengthens the hypothesis invariant.



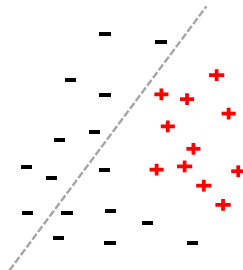
Decision tree learner

- We use decision tree to split all positive/negative examples and use the predicate of the positive as hypothesis invariant.



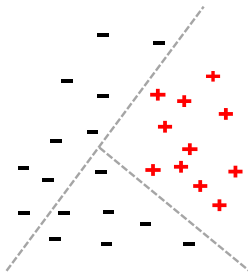
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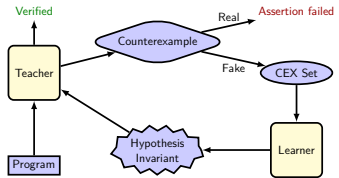
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Motivation

```
1  int x, y;  
2  assume(x <= 2);  
3  assume(y >= 0 && y <= 1);  
4  while(*){  
5      if (x > 0) {  
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9  }  
10 assert(y >= 0 && y <= 3);
```

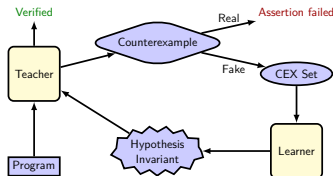
- Invariant hypothesis: *True*.



Motivation

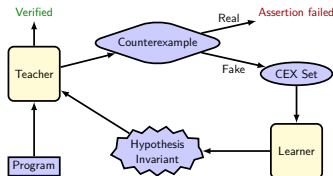
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$$\langle x = 0, y = 9 \rangle$$


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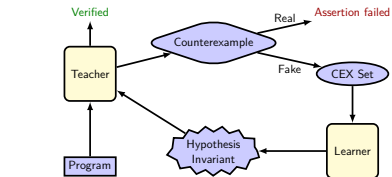
x	0
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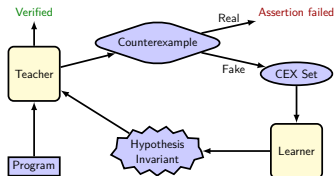
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- Modified hypothesis: *False*.

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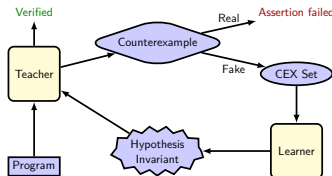


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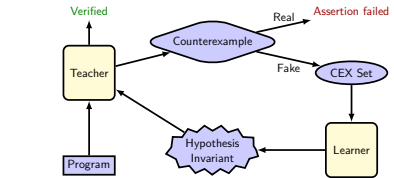
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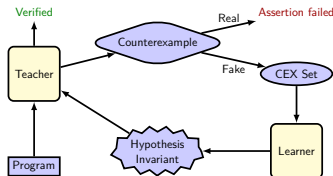


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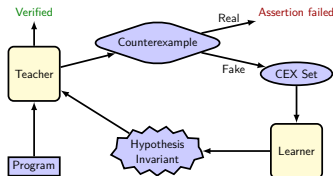
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- Modified hypothesis: $y \leq 8$.

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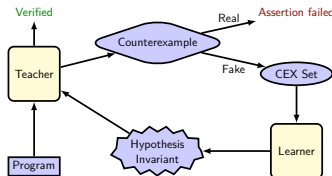


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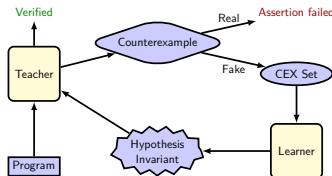
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x	0	2	0
y	9	1	8
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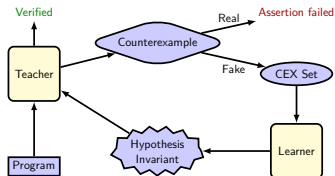


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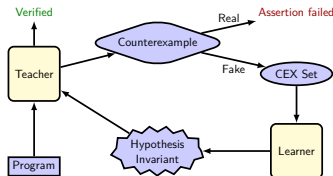
- Modified hypothesis: $y \leq 7$.

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- Invariant hypothesis: $y \leq 7$
- Negative cex provided by teacher to strengthen the invariant:

$$\langle x = 0, y = 7 \rangle$$

- Current cex set:

x	0	2	0	0
y	9	1	8	7
Label	Neg	Pos	Neg	Neg

- Modified hypothesis: $y \leq 6$

Potential improvement

x	0	2	0	0	...	0
y	9	1	8	7	...	4
Label	Neg	Pos	Neg	Neg	...	Neg

- A original counterexample is an *information-limited* single program state.

Potential improvement

x	0	2	0	0	...	0
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- A original counterexample is an *information-limited* single program state.
- The generalized counterexamples we call them **interval counterexamples** look like:

$$\langle x \in (-\infty, \infty), y \in [4, \infty) \rangle$$

is a **set** of program states but not a single.

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Core issues

- How to get a generalized interval counterexample?
- How to learn an invariant using interval counterexamples?

Generalization

- Suppose current hypothesis invariant by Learner: *True*.

```
1  int x, y;  
2  assume(x <= 2);  
3  assume(y >= 0 && y <= 1);  
4  while(*){ // Inv: True  
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Generalization

- Suppose current hypothesis invariant by Learner: *True*.
- The verification condition to prove the assertion via the loop invariant:
 $True \rightarrow y \geq 0 \wedge y \leq 3$. (Simplify to $y \geq 0 \wedge y \leq 3$)

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Generalization by variable elimination

- Suppose the negative counterexample provided by Teacher:

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- Consider the constraint:

$$\underset{\text{(VC)}}{y \geq 0 \wedge y \leq 3} \wedge x = 0 \wedge y = 9$$

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- Suppose the negative counterexample provided by Teacher:

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- The UNSAT core is $\{y = 9\}$.

Generalization by variable elimination

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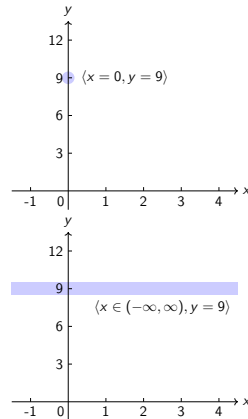
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- The UNSAT core is $\{y = 9\}$.
- We get a generalized counterexample:

$$\langle x \in (-\infty, \infty), y = 9 \rangle$$

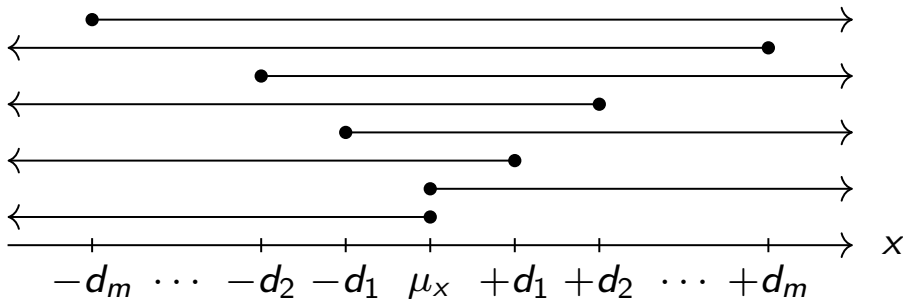


Generalization by boundary constraints

- Variable elimination: Rough and not widely general.

Generalization by boundary constraints

- Variable elimination: Rough and not widely general.
- Intuition: Find the generalization boundaries among some pre-defined distances.



Generalization by boundary constraints

- Consider distances (0,5) and new constraints to replace $x = 0$ and $y = 9$:

$$y \geq 0 \wedge y \leq 3 \underset{\text{(VC)}}{\wedge} \bigwedge \left\{ \begin{array}{l} x \geq -5 \\ x \geq 0 \\ x \leq 0 \\ x \leq 5 \end{array} \right\} \wedge \bigwedge \left\{ \begin{array}{l} y \geq 4 \\ y \geq 9 \\ y \leq 9 \\ y \leq 14 \end{array} \right\}$$

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- The UNSAT core may be: $y \geq 4$.
- More general counterexample:

$$\langle x \in (-\infty, \infty), y \in [4, \infty) \rangle$$

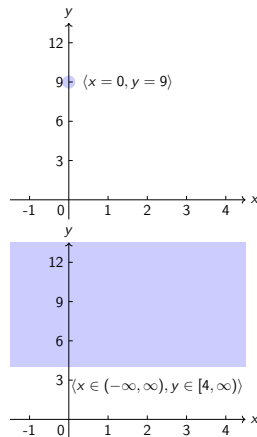
Generalization by boundary constraints

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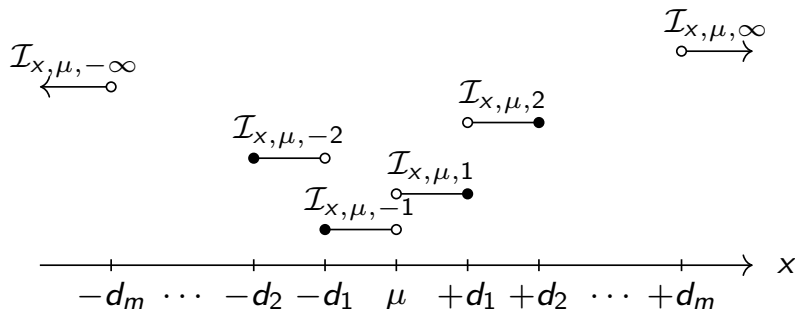


Generalization by interval digging

- UNSAT core may not be the best range. ($y \geq 4$ or $y \geq 9$?)

Generalization by interval digging

- UNSAT core may not be the best range. ($y \geq 4$ or $y \geq 9$?)
- Intuition: Dig out all intervals which contains program states can **not** be generalized.



Generalization by interval digging

- Rearrange the constraints:

$$\begin{array}{c} y \geq 0 \wedge y \leq 3 \\ \text{(VC)} \end{array} \wedge \bigwedge \left\{ \begin{array}{c} \neg(x > 5) \\ \neg(x > 0 \wedge x \leq 5) \\ \neg(x < 0 \wedge x \geq -5) \\ \neg(x < -5) \end{array} \right\} \wedge \bigwedge \left\{ \begin{array}{c} \neg(y > 14) \\ \neg(y > 9 \wedge y \leq 14) \\ \neg(y < 9 \wedge y \geq 4) \\ \neg(y < 4) \end{array} \right\}$$

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- Now the UNSAT core is **unique**: $\neg(y < 4)$.

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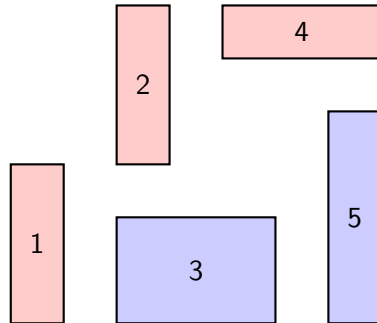
$$y \geq 0 \wedge y \leq 3 \underset{\text{(VC)}}{\wedge} \bigwedge \left\{ \begin{array}{c} \neg(x > 5) \\ \neg(x > 0 \wedge x \leq 5) \\ \neg(x < 0 \wedge x \geq -5) \\ \neg(x < -5) \end{array} \right\} \wedge \bigwedge \left\{ \begin{array}{c} \neg(y > 14) \\ \neg(y > 9 \wedge y \leq 14) \\ \neg(y < 9 \wedge y \geq 4) \\ \neg(y < 4) \end{array} \right\}$$

- Now the UNSAT core is **unique**: $\neg(y < 4)$.
- The counterexample is $\langle x \in (-\infty, \infty), y \in [4, \infty) \rangle$.

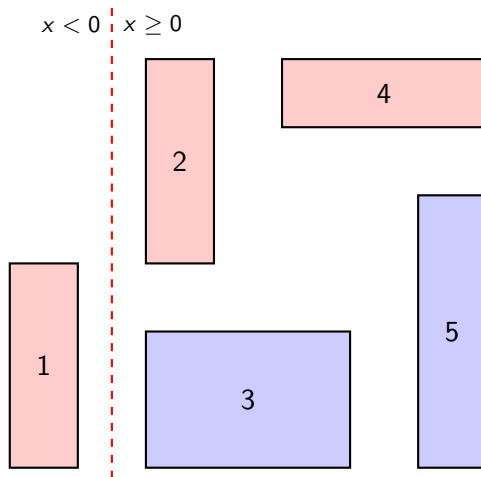
How to split interval examples?

- Avoid crossing examples first
- But also allow examples crossing

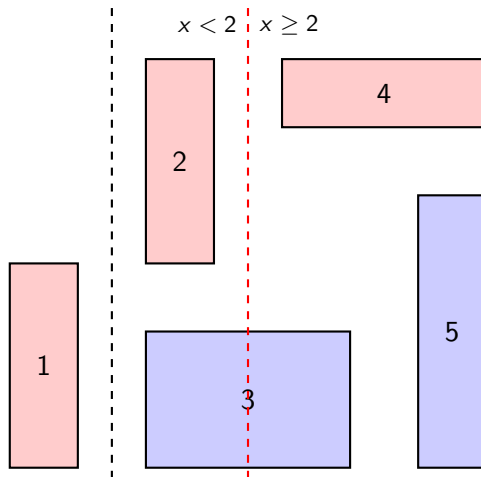
	x	y	Label
1	$[-2, -1]$	$[-2, 1]$	+
2	$[0, 1]$	$[1, 4]$	+
3	$[0, 3]$	$[-2, 0]$	-
4	$[2, 5]$	$[3, 4]$	+
5	$[4, 5]$	$[-2, 2]$	-



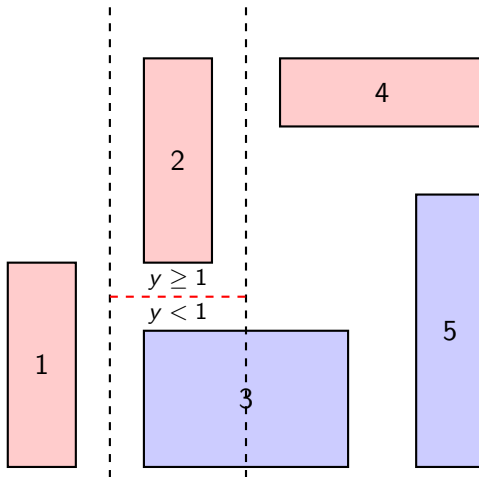
Interval decision tree



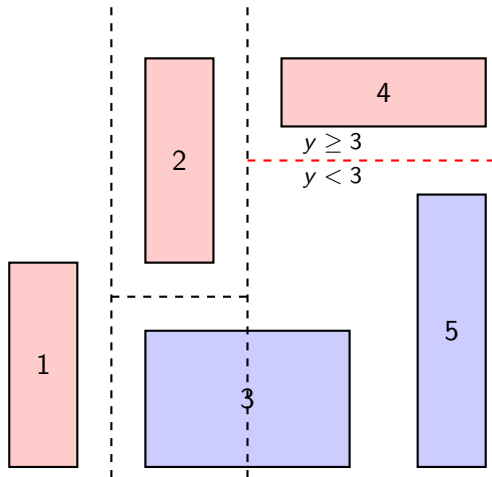
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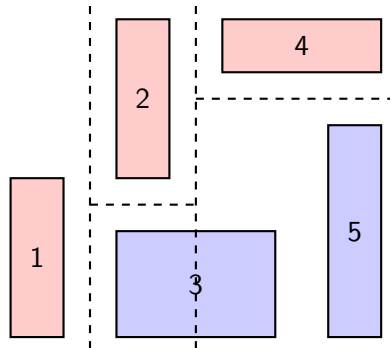


Interval decision tree



Interval decision tree

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2	$[0, 1]$	$[1, 4]$	+
3	$[0, 3]$	$[-2, 0]$	-
4	$[2, 5]$	$[3, 4]$	+
5	$[4, 5]$	$[-2, 2]$	-



- Predicted invariant:

$$x < 0 \vee (x \geq 0 \wedge x < 2 \wedge y \geq 1) \vee (x \geq 0 \wedge x \geq 2 \wedge y \geq 3)$$

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Experiment on generalization methods

- We implement prototype¹ of three generalization methods and also an interval counterexample supported decision tree.

¹ Available at: <https://doi.org/10.5281/zenodo.3898483>.

² Pranav Garg et al. Learning Invariants Using Decision Trees and Implication Counterexamples. POPL 2016

³ Available at: <https://github.com/sosy-lab/sv-benchmarks/>

Experiment on generalization methods

- We implement prototype¹ of three generalization methods and also an interval counterexample supported decision tree.
- Total 94 benchmarks are from ICE-DT² and SV-COMP 2019³.

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	Solved	R	$T(s)$	$T_C(s)$	$T_L(s)$
ICE-DT	90	3007	135.79	20.95	97.51
Elim	91	1916(36.3%)	87.47(35.6%)	17.48(16.6%)	52.46(46.2%)
Boundary	91	1717(42.9%)	72.27(46.8%)	15.15(27.7%)	42.41(56.5%)
Digging	91	1712(43.1%)	73.41(45.9%)	15.87(24.2%)	42.55(56.4%)

Table 1: Summary results on interval generalization methods

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³ Available at: <https://github.com/sosy-lab/sv-benchmarks/>

Experiment on learning methods

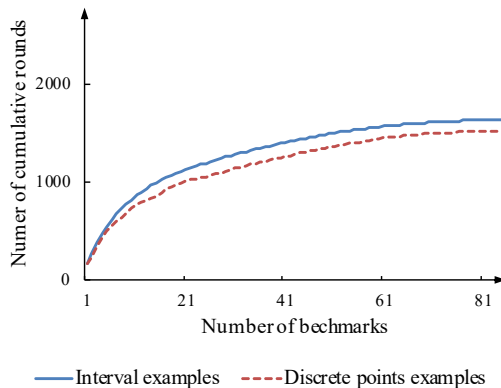
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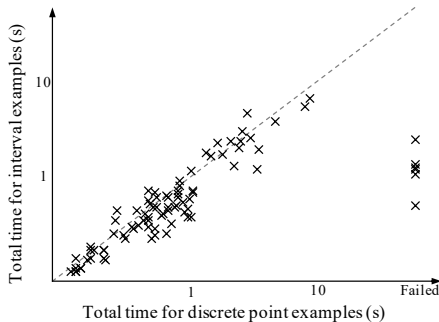
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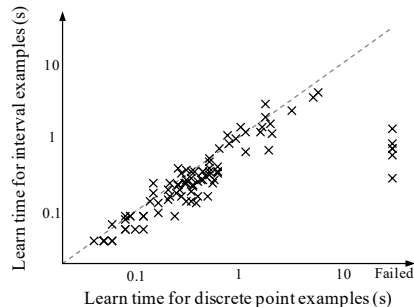


Experiment on learning methods

- Iteration rounds are similar, but time?



(a) Total time comparison



(b) Learn time comparison

- ① Background
- ② Motivation
- ③ Approach
- ④ Evaluation
- ⑤ Conclusion

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- Three different generalization techniques to compute interval counterexamples.
- An improved decision tree algorithm to adapt interval counterexamples.
- A prototype and over 40% improvement on learning rounds and verification time compared with existing methods.

Thanks!