

CE 815 – Secure Software Systems

ML-Based Vulnerability Detection Methods (Hoppity)

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Acknowledgments: Some of the slides are fully or partially obtained from other sources. A reference is noted on the bottom of each slide, when the content is fully obtained from another source. Otherwise a full list of references is provided on the last slide. Thanks to Mohammad Haddadian for the help on the slides.



Introduction

- Vulnerability detection as first step
 - Then, Vulnerability repair
-
- Compiler VS Interpreter
 - Vulnerability VS Bug
 - Security VS non-Security

HOPPITY: Learning Graph Transformations to Detect and Fix Bugs in Programs, ICLR 2020.

Problem



Source-code analysis is:

- Undecidable
- Noisy
- Rules are hand written
- Tailored to specific code bases / bug patterns



Javascript Challenges

- Incorrect operators
- Incorrect identifiers
- Accessing undefined properties
- Mishandling variable scopes
- Type incompatibilities

Example



```
function clearEmployeeListOnLinkClick() {
  document.querySelector("a").addEventListener("click",
    function(event) {
      document.querySelector("ul").InnerHTML = "";
    }
  );
}
```

(a) InnerHTML should have been innerHTML.

```
if (matches) {
  return {
    episode: Number(matches.groups.episode),
    hosts: matches.groups.hosts.split(/([, &]+|\sand\s)/)
      .map(el => S(el).trim().s)
  };
}
```

(b) Highlighted parentheses should have been removed.

```
module.exports = function (grunt) {
  grunt.initConfig({
    execute: {...}, copy: {...}, checktextdomain: {...}
    wp_readme_to_markdown: {...}, makepot: {...}})
  ...
  grunt.registerTask('default', ['wp_readme_to_markdown',
    'makepot', 'execute', 'checktextdomain'])
};
```

(c) copy function should have also been included in the highlighted list.

```
export default {
  computed: {
    level () {
      return dictMap.skillLevel[
        parseInt((this.value === 0 ? 1 : this.value)/20)];
    }, ...
  }
}
```

(d) parseInt should have been removed because === implies this.value is an integer.

Solution



Leverage large amounts of Javascript fixes on Github to locate and repair bugs

Steps



- Represent source code
- Represent fixes
- Learning

Model



- Problem of detecting and repairing bugs in programs is a structured prediction problem on a graph-based representation of programs.

$P(g_1, g_2, \dots, g_T \mid G_0, \theta)$, where θ is the model parameter

$$p(g_{fix} | g_{bug}; \theta) = p(g_1 | g_{bug}; \theta) p(g_2 | g_1; \theta) \dots p(g_{fix} | g_{T-1}; \theta)$$

Goal

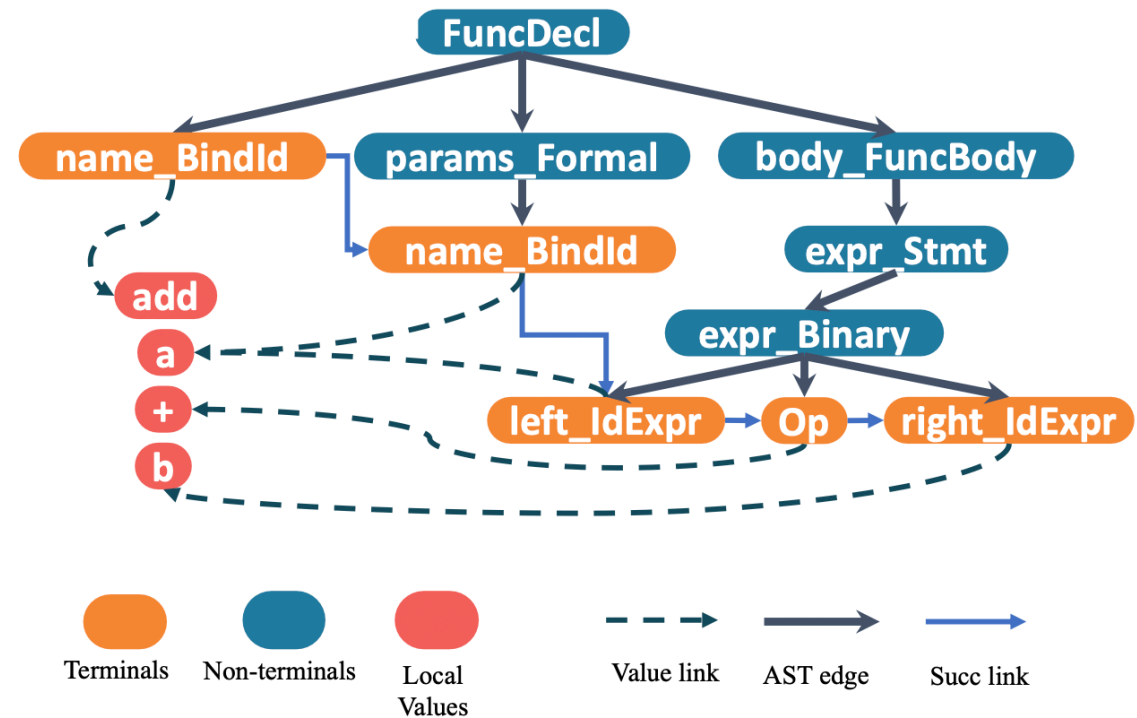


-
- 1- function `add (a) {a + b;}` Buggy
 - 2- function `add (a, b) {a + b;}` Step 1
 - 3- function `add (a, b) { return a + b;}` Step 2

Source code representation



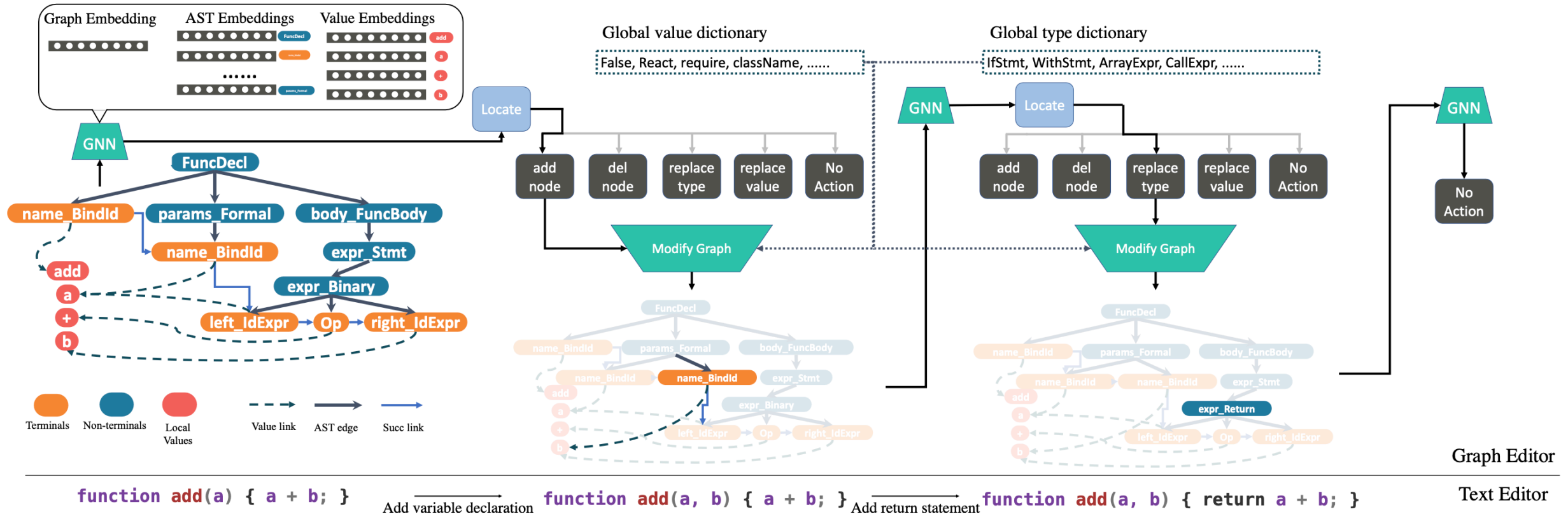
- AST
- ValueLink



```
function add(a) { a + b; }
```

Fix representation

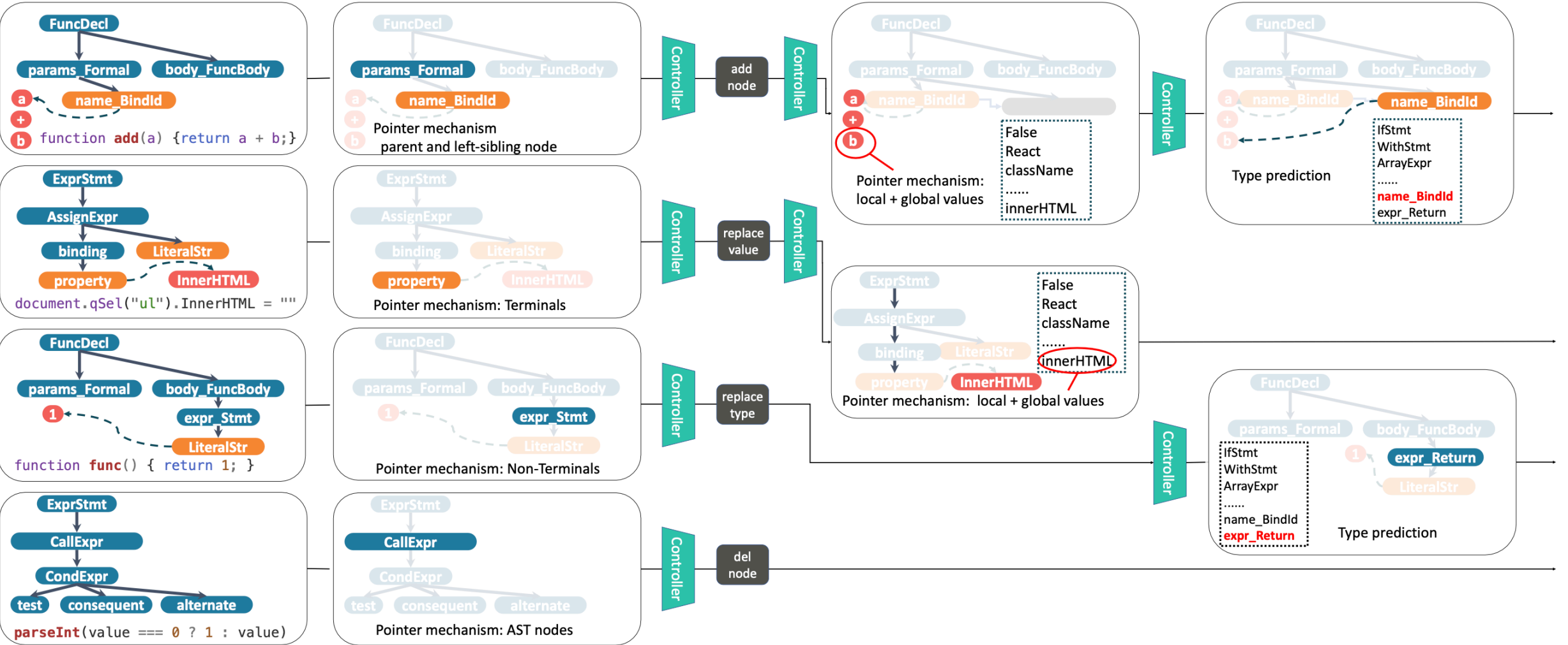
• Graph Edits



Location

Value

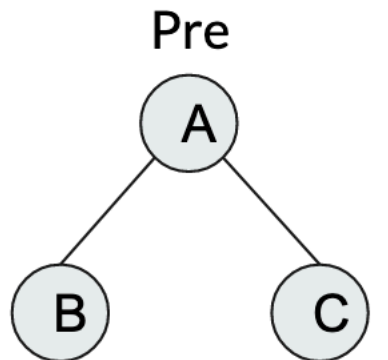
Type



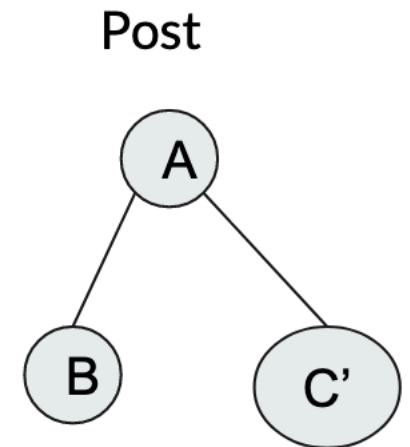
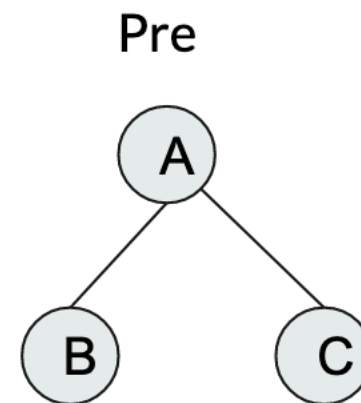
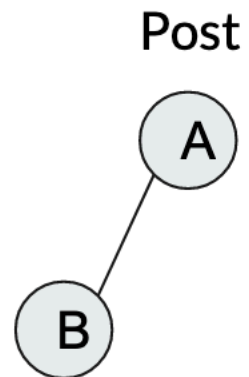
Low level primitives



- Location
- Value
- Type



Delete



Replace



Low level primitives: Value

Choose from either the values appearing in the current file (local value table), or a collection of global values that are common for the specific language

Let D_{val} be the global dictionary of commonly used leaf-node values in the language, where each item $i \in D$ is associated with a vector representation:

$$\vec{i}_v \in \mathbb{R}^d$$



Low level primitives: Type

- As the total possible number of types is finite and fixed for a given language, the type prediction is simply a multi-class classification problem.
- But utilize the AST grammar checker with contextual information to prune the output space.

Graph edit operators



add
node

del
node

replace
type

replace
value

No
Action



Anatomy of a graph edit

“replace_val”

1. Predict **Location**
2. Predict **Value**

“replace_type”

1. Predict **Location**
2. Predict **Type**

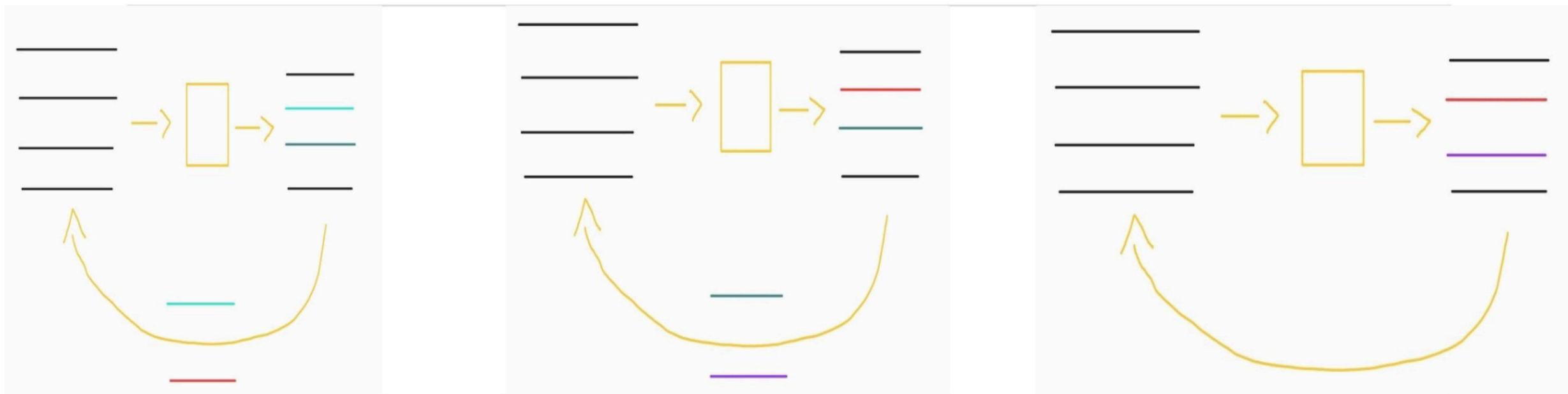
“remove”

1. Predict **Location**

“add”

1. Predict **Location**
2. Predict **Value**
3. Predict **Type**
4. Predict **Child Number**

Graph transformation

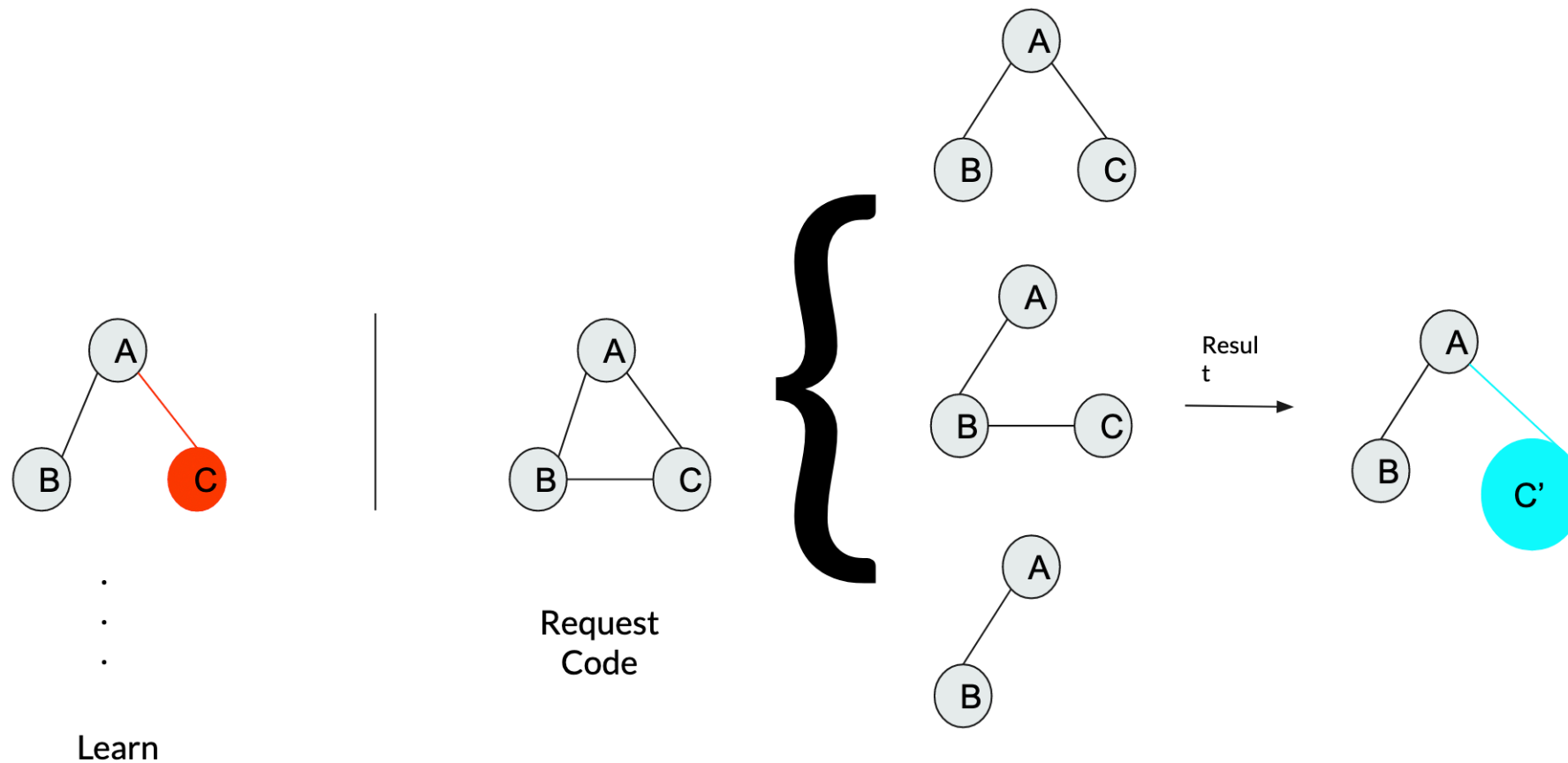


1

2

3

Inference



Dataset



- OneDiff (just one change)
- ZeroOneDiff (zero or one edit)
- ZeroOneTwoDiff (zero, one or two edits)

	ADD	REP_TYPE	REP_VAL	DEL	total
train	6,473	1,864	251,097	31,281	290,715
validate	790	245	31,357	3,957	36,349
test	796	233	31,387	3,945	36,361

Table 1: Statistic of `OneDiff` dataset. See appendix for more information of other dataset.

Evaluation



	Total		<i>Location</i>		<i>Operator</i>	<i>Value</i>		<i>Type</i>	
	Top-3	Top-1	Top-3	Top-1	Top-1	Top-3	Top-1	Top-3	Top-1
TOTAL	26.1	14.2	35.5	20.4	34.4	52.3	29.1	76.1	66.7
ADD	52.9	39.2	69.6	51.4	70.6	65.7	55.1	76.8	68.5
REP_VAL	23.4	11.9	33.3	18.5	31.7	53.0	28.8	-	-
REP_TYPE	71.7	52.4	73.0	52.8	79.4	-	-	74.7	61.0
DEL	39.6	24.8	44.0	27.5	45.8	-	-	-	-
Random	.08	.07	2.28	1.4	27.7	.01	.01	.27	0

Table 2: Evaluation of model on the OneDiff dataset: accuracy (%).

Evaluation (cont.)



Type	GGNN-Rep	GGNN-Cls	HOPPITY
Top-1	53.2%	99.6%	90.0%
Top-3	85.8%	99.6%	94.8%

Table 3: REP_TYPE accuracies with location+op.

Value	GGNN-Rep	GGNN-RNN	HOPPITY
Top-1	63.8%	60.3%	69.1%
Top-3	67.6%	63.6%	73.4%

Table 4: REP_VAL accuracies with location+op.

	Top-1	Top-3
HOPPITY	67.7%	73.3%
SequenceR	64.2%	68.6%

Table 5: Overall OneDiff accuracy with location.

Bug Type	Amount	TAJS	HOPPITY
Undefined Property	7	0	1
Functional Bug	11	0	3
Refactoring	12	0	1
Total	30	0	5

Table 6: Comparison with TAJS.

Acknowledgments



- [HOPPITY] HOPPITY: Learning Graph Transformations to Detect and Fix Bugs in Programs, ICLR 2020.