

Toward General Diagnosis of Static Errors

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Static Program Analysis

- Many flavors
 - Type system
 - Dataflow analysis
 - Information-flow analysis
- Useful properties
 - Type safety
 - Memory safety
 - Information-flow security
- But, (sometimes) confusing error messages make static analyses hard to use

Example 1: ML Type Inference

- OCaml

```
1 let foo(lst: int list): (float*float) list =  
2   ...  
3   let rec loop lst x y dir acc =  
4     if lst = [] then  
5       acc  
6     else  
7       print_string "foo"  
8   in  
9   List.rev (loop lst 0.0 0.0 0.0 [(0.0,0.0)])
```

Mistake

Locating the error cause is

- Time-consuming
- Difficult

OCaml: This expression has type 'a list but is here used with type unit

Example 2: Information-Flow Analysis

- Jif: Java + Information-Flow control

Mistake

```
1 public final byte[{}] {this} encText;  
2 ...  
3 public void m(FileOutputStream[{}]{this} encFos)  
4   throws (IOException) {  
5   try {  
6     for (int i=0; i<encText.length; i++)  
7       encFos.write(encText[i]);  
8   } catch (IoException e) {}  
9 }
```

Jif: This label is too restrictive

Better error report is needed

Toward Better Error Reports

- Limitations of previous work
 - Methods reporting full explanation – **Verbose reports**
 - Analysis-specific methods – **Tailored heuristics**
 - Methods diagnosing false alarms – **No diagnosis of true errors**
- Our approach
 - Applies to a large class of program analyses
 - Diagnoses the cause of both true errors and false alarms
 - Reports error causes more accurately than existing tools

Approach Overview

Language-Specific

Programs	
Jif	OCaml
Others	<pre> let foo(lst: int list):(float*float) list = let rec loop lst x y dir acc = if lst = [] then acc else print_string "foo" in List.rev(loop lst 0.0 0.0 0.0 [(0.0,0.0)]) </pre>

Constraints
<pre> unit = acc5 acc5 = acc3 acc3 = (float*float) list unit = loopret loopret = α list α list = (float*float) list loopret = acc5 </pre>

Language-Agnostic

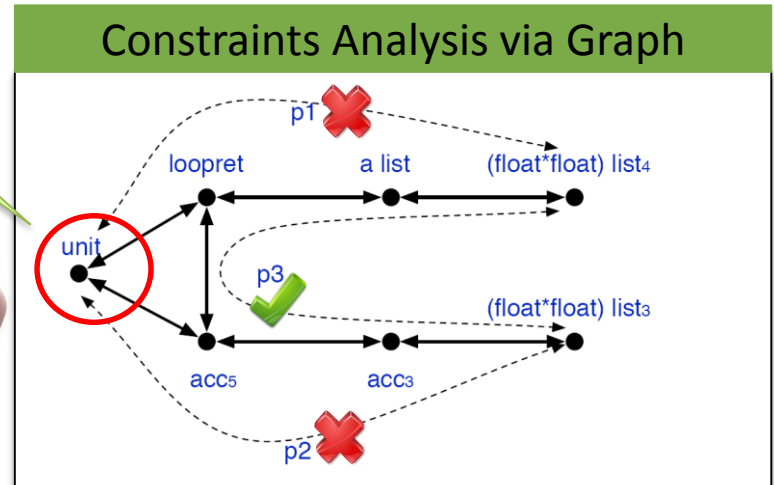
Based on Bayesian interpretation

Cause

General Diagnosis Heuristics

The error cause is likely to be

- Simple
- Able to explain all errors
- Not used often on correct paths
- (false alarm) weak and simple

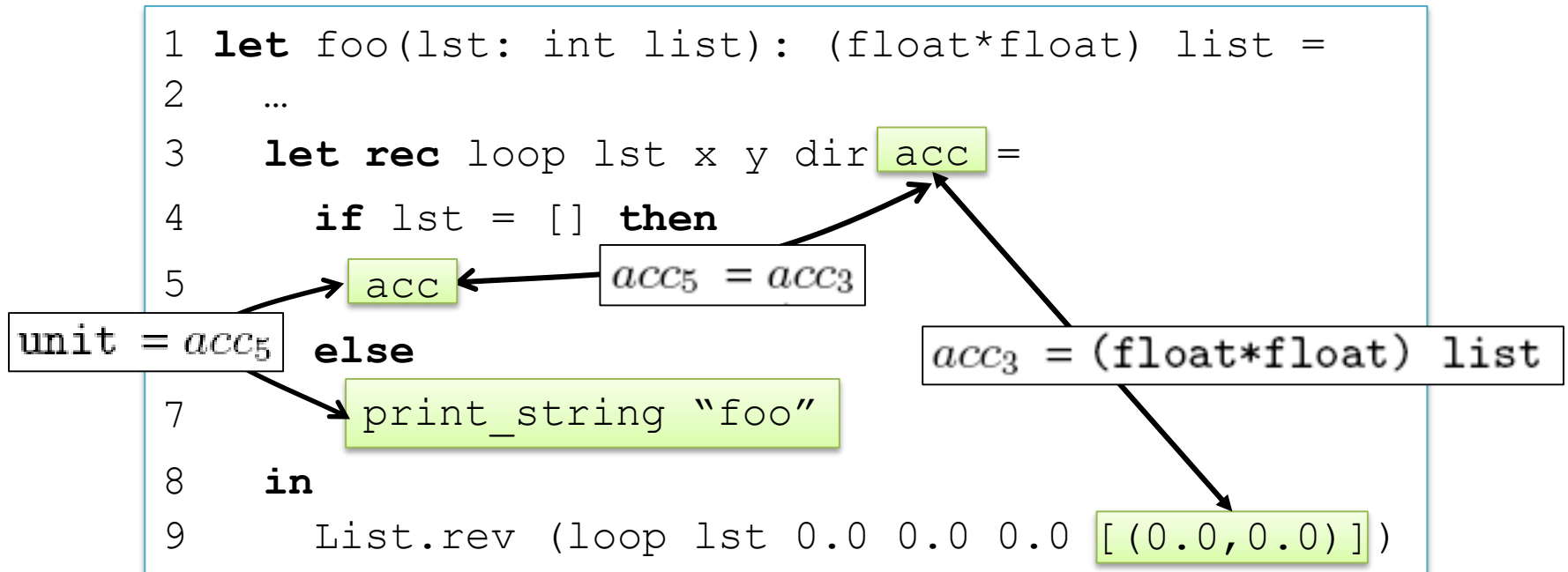


From Programs to Constraints

- ML type inference

- Constraint elements: types
- Constraints: type equalities

Constructors: unit, float, list, *
Variables: acc_3 , acc_5



A General Constraint Language

Syntax of Constraints

$$E ::= \alpha | c(E_1, \dots, E_n) | \bar{c}^i(E) | E_1 \sqcup E_2 | E_1 \sqcap E_2 | \perp | \top$$
$$I ::= E_1 \leq E_2 \qquad C ::= \bigwedge_i I_{1i} \vdash \bigwedge_j I_{2j}$$

- Element (E): form a lattice, with an ordering \leq
- Inequality (I): a partial order on elements
 - E.g., “subtype of”, “subset of”, “less confidential than”
- Constraint (Hypothesis \vdash Conclusion)
 - Hypothesis captures programmer assumptions
 - Variable-free constraint is valid when all \leq in conclusion can be derived from hypothesis

Properties of the Constraint Language

- Expressive
 - ML type inference with polymorphism
 - Information-flow analysis with complex security model
 - Dataflow analysis(See formal translations in paper)
- Practical to calculate satisfiable/unsatisfiable subsets of constraints

Approach Overview

Programs

Jif

OCaml

```
let foo(lst: int list):(float*float) list =  
let rec loop lst x y dir acc =  
  if lst = [] then  
    acc  
  else  
    print_string "foo"  
in  
List.rev(loop lst 0.0 0.0 0.0 [(0.0,0.0)])
```

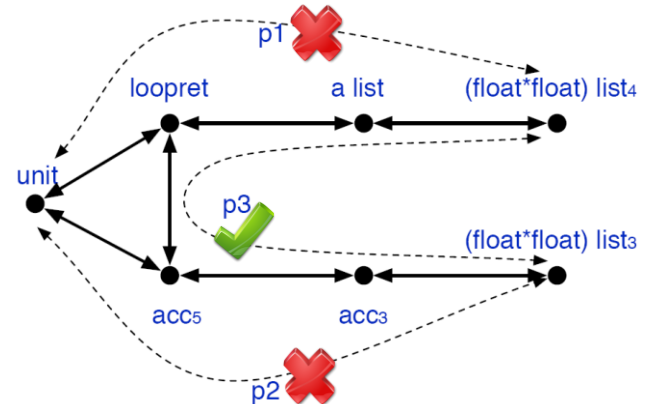
Others

Constraints

```
unit = acc5  
acc5 = acc3  
acc3 = (float*float) list  
unit = loopret  
loopret = α list  
α list = (float*float) list  
loopret = acc5
```

Language-Agnostic

Constraints Analysis via Graph



Constraint Graph in a Nutshell

- Graph construction (simple case)
 - Node: constraint element
 - Directed edge: partial ordering

1. $unit = acc_5$

2. $acc_5 = acc_3$

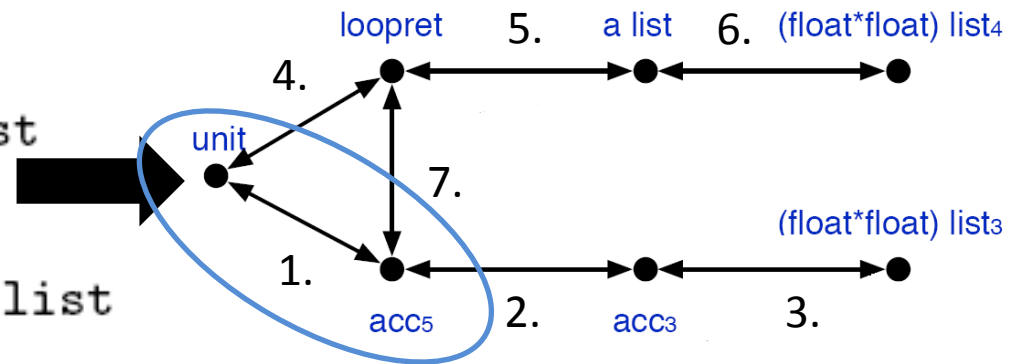
3. $acc_3 = (float*float) list$

4. $unit = loopret$

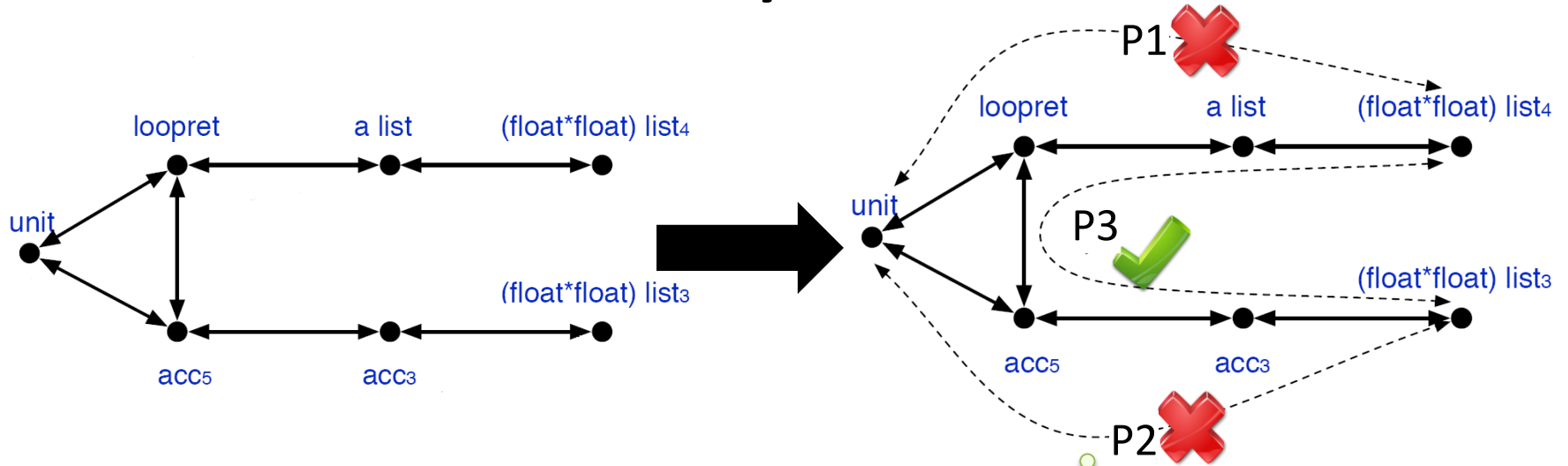
5. $loopret = \alpha list$

6. $\alpha list = (float*float) list$

7. $loopret = acc_5$



Constraint Analysis in a Nutshell



```

3  let rec loop lst x y dir acc =
4    if lst = [] then
5      acc ← acc5 = acc3
6    else
7      print_string "foo"
8  in
9  List.rev (loop lst 0.0 0.0 0.0 [(0.0, 0.0)])

```

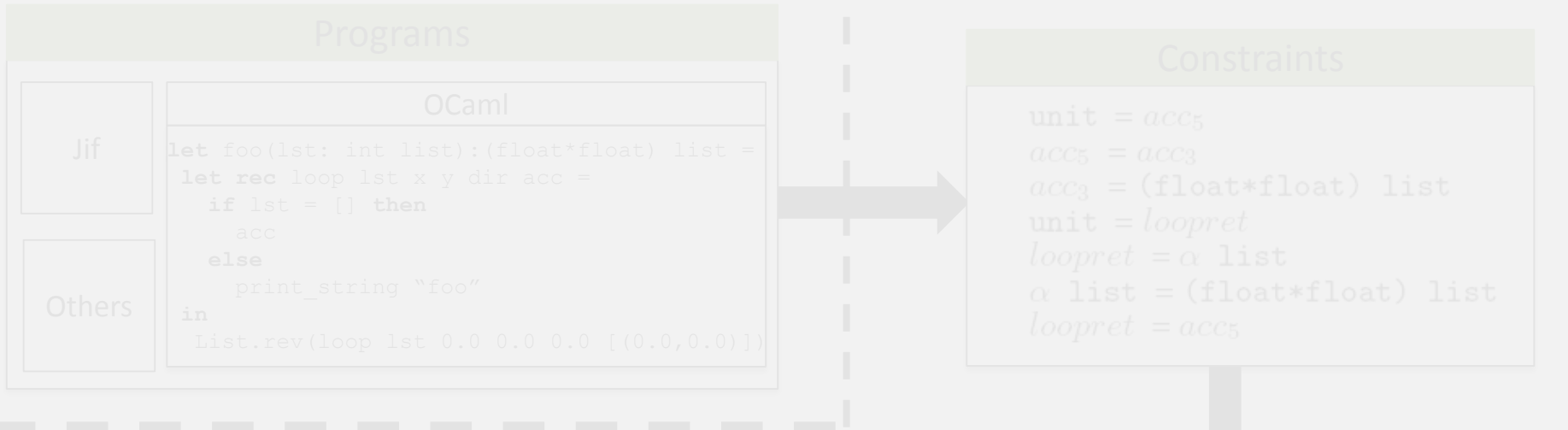
Annotations in the code block:

- A box around `acc` on line 3 points to a red 'X'.
- A box around `acc5 = acc3` on line 5 points to a green box containing `acc`.
- A box around `acc3 = (float*float) list` on line 6 points to a red 'X'.
- A box around `[(0.0, 0.0)]` on line 9 points to a green box containing `[(0.0, 0.0)]`.
- A green callout bubble with a red 'X' contains the text "Type mismatch".

Constraint Analysis for the Full Constraint Language

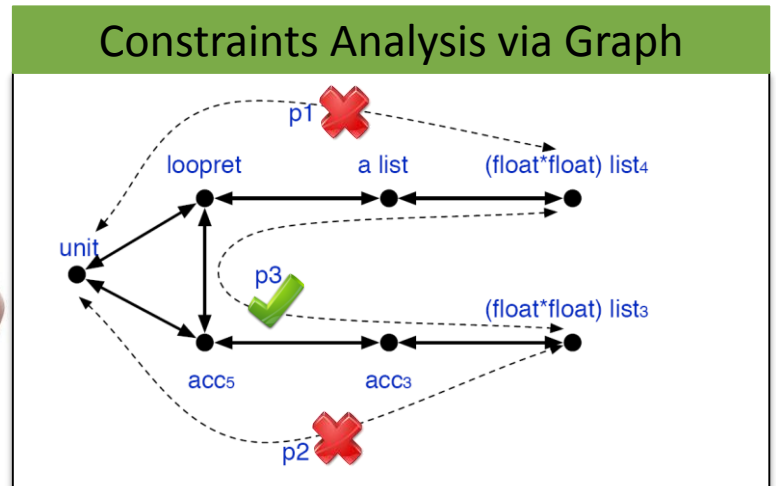
- Handling constructors, hypotheses
 - CFG Reachability [Barrett et al. 2000, Melski&Reps 2000]
 - Also handles join/meet operations(See details in paper)
- Performance
 - Scalable: quadratic w.r.t. # graph nodes in practice

Error Diagnosis



Language-Agnostic

Bayesian reasoning



Possible Explanations

- When an analysis reports an error, either
 - The program being analyzed is wrong (true alarm)
 - E.g., an expression is wrong in OCaml program
 - The program analysis reports an false alarm (false alarm)
 - E.g., an assumption is missing in Jif program
- Explanations to find
 - Wrong expressions
 - Missing hypotheses

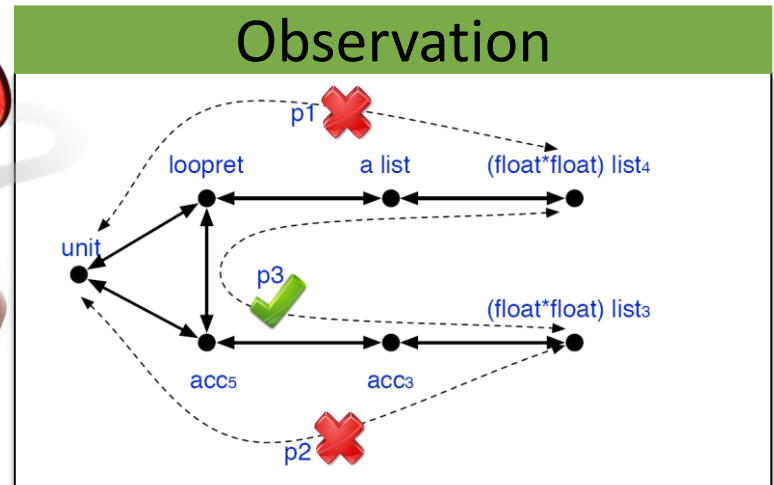
**Key insight:
Bayesian reasoning**

Inferring Most-Likely Error Cause

- The most likely explanation


$$\operatorname{argmax}_{(E,H) \in \mathcal{G}} P(E, H | o)$$

- \mathcal{G} : explanation (pair of constraint elements and hypotheses)
- o : observation (structure of a constraint graph)



Likelihood Estimation

$$\operatorname{argmax}_{(E,H) \in \mathcal{G}} P_{\Omega}(E)P(o|E,H)P_{\Psi}(H)$$



MAP
estimation

Likelihood Estimation

sat paths use elements in E

$$\operatorname{argmax}_{(E,H) \in \mathcal{G}} P_1^{|E|} \left(\frac{P_2}{1 - P_2} \right)^{k_E} P_\Psi(H)$$

- Simplifying assumptions:
 - All expressions are equally likely to be wrong (with P_1)
 - Errors are unlikely (with $P_2 < 0.5$) to appear on satisfiable paths
- Intuitively,

General Diagnosis Heuristics

The error cause is likely to be

- Simple
- Able to explain all errors
- Not used often on correct paths
- (missing hypotheses) weak and simple



Explain later

Inferring Likely Wrong Expressions

$$\operatorname{argmax}_E P_1^{|E|} \left(\frac{P_2}{1 - P_2} \right)^{k_E}$$

- Search space
 - all **subsets** of expressions (nodes in constraint graph)
- A* search
 - Optimal: all most likely wrong expressions are returned
 - Efficient: 10 seconds when the search space is over 2^{1000}

Evaluation suggests the accuracy is not sensitive to the value of P_1 and P_2

Inferring Likely Missing Hypotheses

$$\operatorname{argmax}_H P_\Psi(H)$$

- Simplicity is not the only metric
 - $T \leq \perp$ “explains” all errors
- Likely missing hypotheses are both ***weak*** and ***simple***
 - Minimal weakest hypothesis

Bob \leq Carol \vdash Alice \leq Bob
Bob \leq Carol \vdash Alice \leq Carol
Bob \leq Carol \vdash Alice \leq Carol $\sqcup \perp$

Minimal weakest hypothesis
Alice \leq Bob

Formal definition & search algorithm in paper

Evaluation

- Implementation

- Translation from analyses to constraints

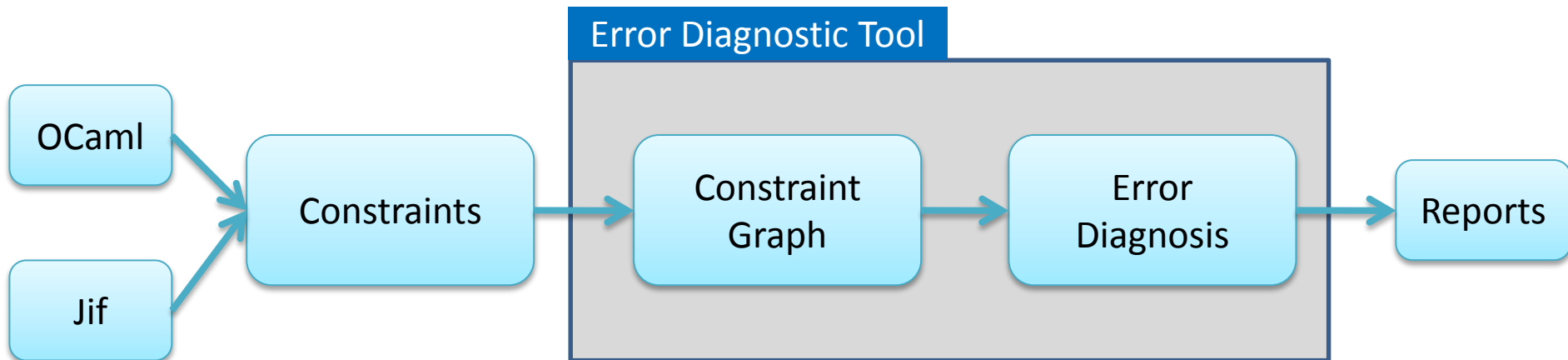
- OCaml: modified EasyOCaml (500 on top of 9,000LoC)

- Jif: modified Jif (300 on top of 45,000LoC)

- General error diagnostic tool

- ~5,500 LoC in Java

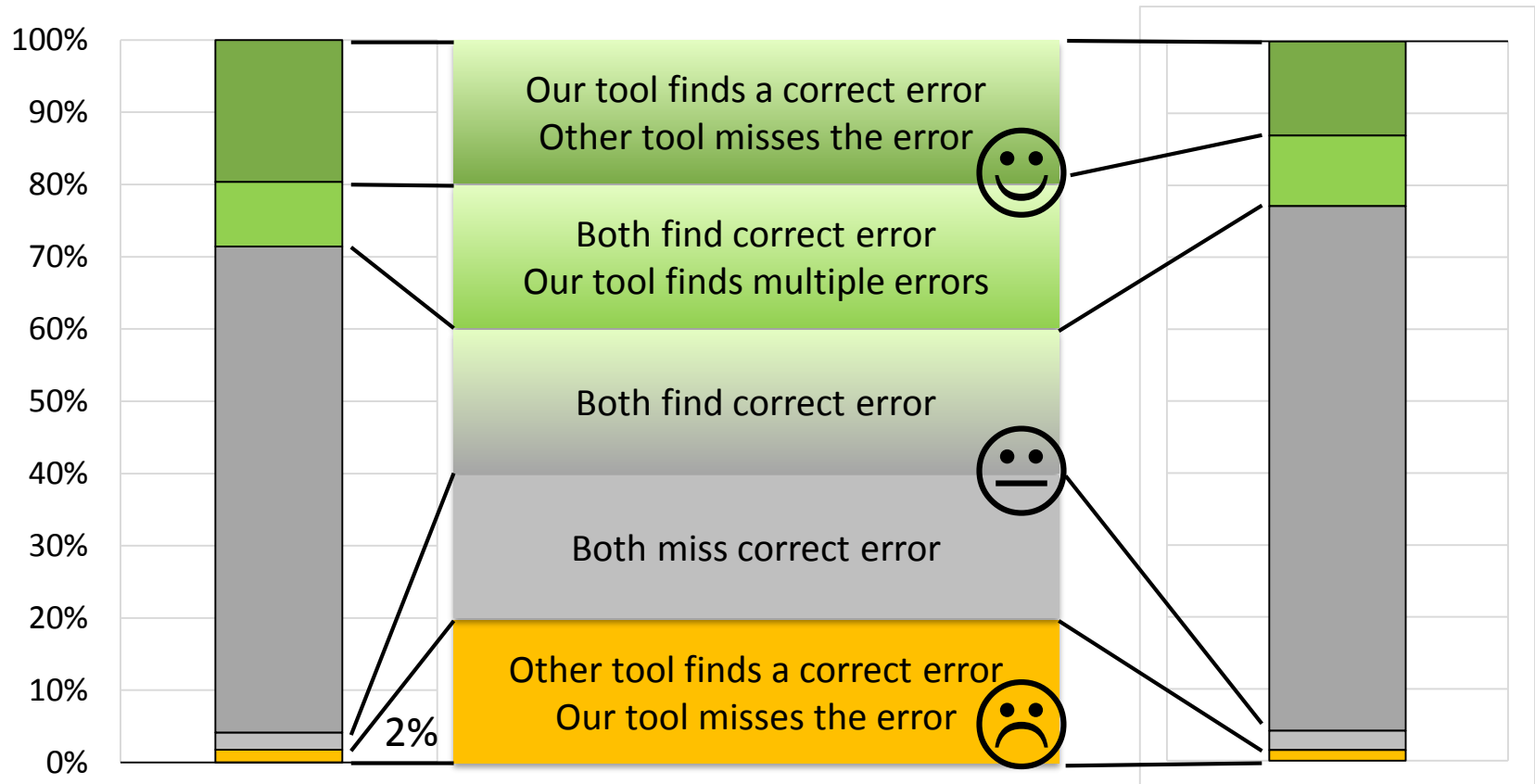
Modest effort



Accuracy of Error Reports: OCaml

- Data
 - A corpus of previously collected programs [Lerner et al.'07]
 - Analyzed 336 programs with type mismatch errors
- Metric of report quality
 - Location of programmer mistake: user's fix with larger timestamp
 - Correctness: only when the programmer mistake is returned

Comparison with OCaml and Seminal



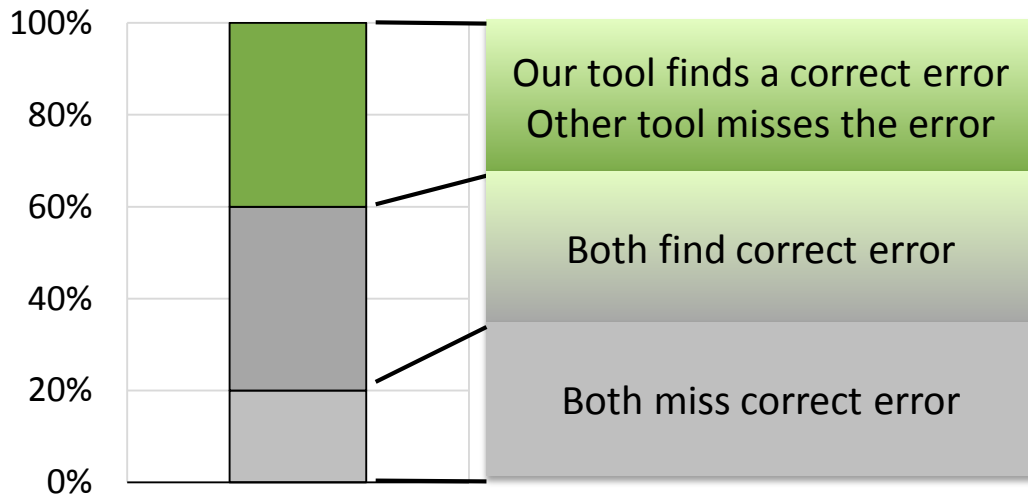
Comparison with the OCaml compiler

Comparison with the Seminal tool

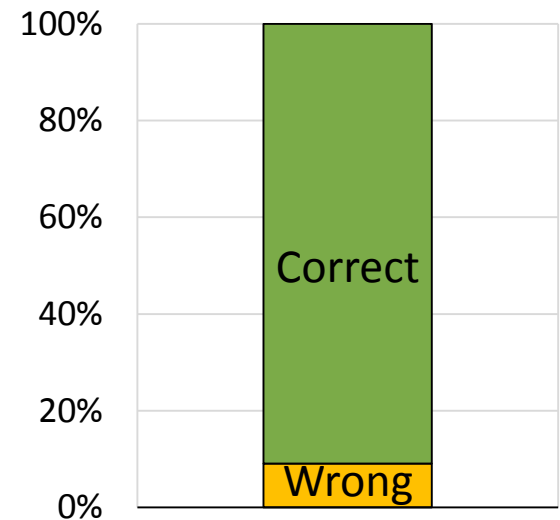
[Lerner et al.'07]

Comparison with Jif

- 16 previously collected buggy programs
 - An application with real-world security concern [Arden et al.'12]
 - Errors clearly marked by the application developer
 - Contains both error types



Comparison with the Jif compiler
(Wrong expression)



Accuracy on missing hypothesis

Related Work

- Program analyses as constraint solving [e.g., Aiken'99, Foster et al.'06]
 - No support for hypothesis; error report is verbose
- Diagnosing ML/Jif errors [e.g., McAdam'98, Heeren'05, Lerner'07, King'08, Chen&Erwig'14]
 - Tailored to specific program analysis
- Probabilistic inference [e.g., Ball et al.'03, Kremenek et al.'06, Livshits et al.'09]
 - Different contexts; errors are considered in isolation
- Diagnosing false alarms [e.g., Dillig et al.'12, Blackshear and Lahiri'13]
 - Does not diagnose true errors in program

Future Work

- More expressive language
 - Add arithmetic to the language
- Refine the simplifying assumptions
 - Remove assumptions on error independence
 - Incorporate domain specific knowledge

Conclusion



Program Analyses

ML Type Inference

Information-flow analysis

Dataflow analysis

General diagnosis of static errors

- Applies to a large class of program analyses
- Diagnoses the cause of both true errors and false alarms
- Bayesian reasoning => more accurate reports than with existing tools

A demo is available at: <http://apl.cs.cornell.edu/~zhangdf/diagnostic>