

# Automated Error Diagnosis Using Abductive Inference

Isil Dillig<sup>1</sup>   Thomas Dillig<sup>1</sup>   Alex Aiken<sup>2</sup>

<sup>1</sup>Department of Computer Science  
College of William & Mary, Virginia, USA

<sup>2</sup>Department of Computer Science  
Stanford University, CA, USA

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Severin Heiniger

# An Ordinary Day in a Developer's Life

```
1 void foo(int flag, unsigned int n) {
2     int k = 0, i = 0, j = 0, z = 0;
3     if (flag) k = n;
4     else      k = 1;
5
6     while (i <= n) {
7         i = i + 1;
8         j = j + i;
9     }
10    int z = k + i + j;
11    assert(z > 2 * n);
12 }
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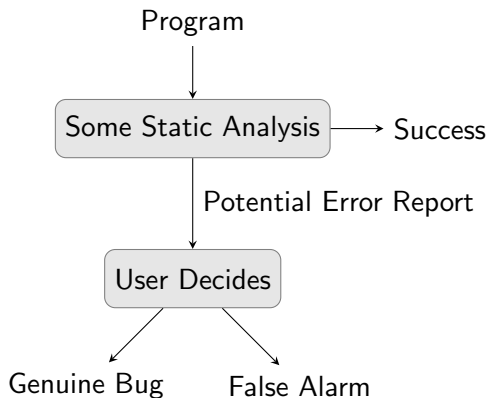
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```

## Static analysis tool error report

Assertion  $z > 2 * n$  may not always hold.

# Manual Report Classification

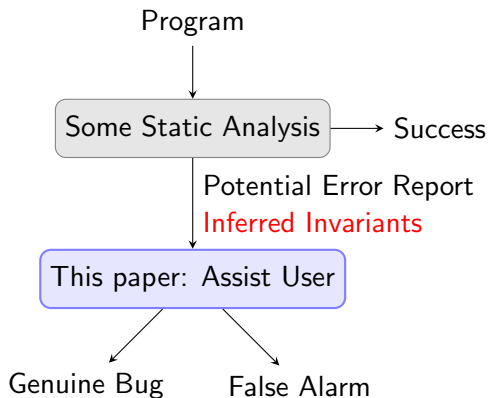


- Time-consuming
- User repeats all successful reasoning by tool
- Error-prone

## Effect

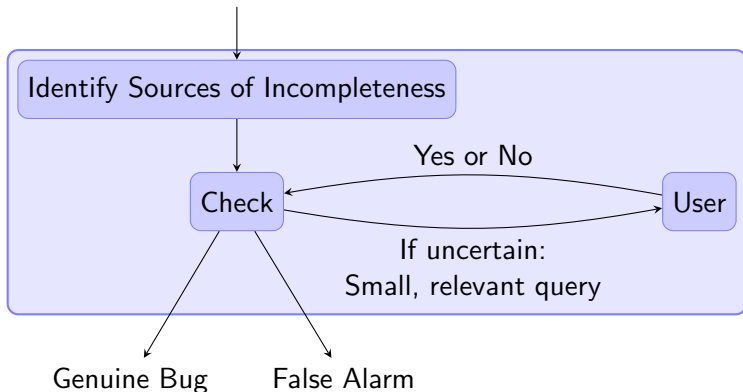
Major impediment to adoption of static analysis tools

# Semi-Automated Report Classification



# Semi-Automated Report Classification

Program with Inferred Invariants  
and Potential Error Report



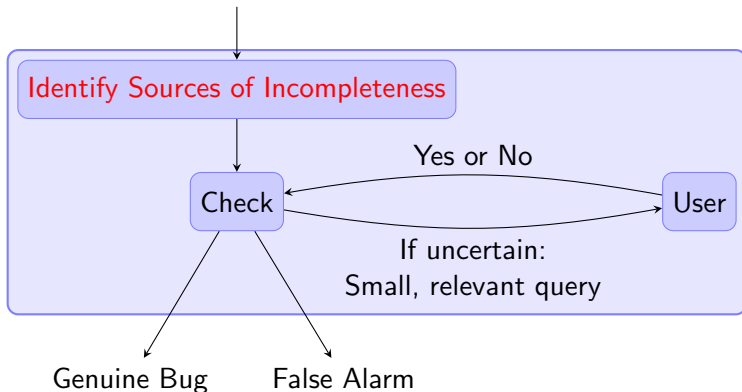
- Proof Obligation Query: *Is property  $P$  an invariant?*
  - If yes, the program is certainly error-free (false alarm)
- Failure Witness Query: *Can property  $P$  arise in some execution?*
  - If yes, the program is certainly buggy

## Strategy

Pose queries in order of increasing cost (easiest first) to minimize the amount of trusted information the user must supply



Program with Inferred Invariants  
and Potential Error Report



- Program with parameters, local variables, conditionals and while loops
- Only linear arithmetic, no function calls
- While loops annotated with inferred post-condition  $p'$ :  
`while(p) { s } [p']`
- Program ends with an `assert(p)`

# Identify Sources of Incompleteness

*Symbolically evaluate* the program. At each point in the program, environment  $\mathbb{S}$  maps program variables to *symbolic value sets*.

$\mathbb{S}(i) = \{\dots, (\pi, \phi), \dots\}$  Under constraint  $\phi$ , the value of variable  $i$  is the symbolic expression  $\pi$

Constraints  $\phi$  keep values from different paths separate.  $\pi$  can contain

**Input Variables  $\nu$**  For unknown program inputs

**Abstraction Variables  $\alpha$**  For unknown values due to imprecisions, e.g., after loops

# Example

```
1 void foo(int flag, unsigned int n) {
2     int k = 0, i = 0, j = 0, z = 0;
3                                      $\mathbb{S}(k) = \{(0, true)\}$     $\mathbb{S}(i) = \{(0, true)\}$    ...
4     if (flag) k = n;
5     else      k = 1;
6
7                                      $\mathbb{S}(k) = \{(1, \neg \nu_{flag}), (\nu_n, \nu_{flag})\}$ 
8     while (i <= n) {
9         i = i + 1;
10        j = j + i;
11    }
12    int z = k + i + j;
13    assert(z > 2 * n);
14 }
```

$\mathbb{S}(i) = \{(\alpha_i, true)\}$     $\mathbb{S}(j) = \{(\alpha_j, true)\}$   
 $\mathbb{S}(z) = \{(1 + \alpha_i + \alpha_j, \neg \nu_{flag}), (\nu_n + \alpha_i + \alpha_j, \nu_{flag})\}$

# Example

```
1 void foo(int flag, unsigned int n) {
2     int k = 0, i = 0, j = 0, z = 0;
3                                      $\mathbb{S}(k) = \{(0, true)\}$     $\mathbb{S}(i) = \{(0, true)\}$    ...
4     if (flag) k = n;
5     else      k = 1;
6
7                                      $\mathbb{S}(k) = \{(1, \neg \nu_{flag}), (\nu_n, \nu_{flag})\}$ 
8     while (i <= n) {
9         i = i + 1;
10        j = j + i;
11    } [ $i \geq 0 \wedge i > n$ ]
12    int z = k + i + j;
13    assert(z > 2 * n);
14 }
```

Propagate inferred invariants as constraints on abstract variables

$$\mathcal{I} = (\alpha_i \geq 0 \wedge \alpha_i > \nu_n \wedge \nu_n \geq 0)$$

# Example

```
1 void foo(int flag, unsigned int n) {
2   int k = 0, i = 0, j = 0, z = 0;
3                                     S(k) = {(0, true)}  S(i) = {(0, true)}  ...
4   if (flag) k = n;
5   else      k = 1;
6
7                                     S(k) = {(1, ¬νflag), (νn, νflag)}
8   while (i <= n) {
9     i = i + 1;
10    j = j + i;
11 } [i ≥ 0 ∧ i > n]
12 int z = k + i + j;
13 assert(z > 2 * n);
14 }
```

Symbolically evaluate the assertion predicate

$$\phi = (1 + \alpha_i + \alpha_j > 2 * \nu_n \wedge \neg \nu_{flag}) \vee (\nu_n + \alpha_i + \alpha_j > 2 * \nu_n \wedge \nu_{flag})$$

The result is a pair of symbolic constraints

- $\mathcal{I}$  All known invariants on abstract variables
- $\phi$  Condition under which the assertion evaluates to *true*

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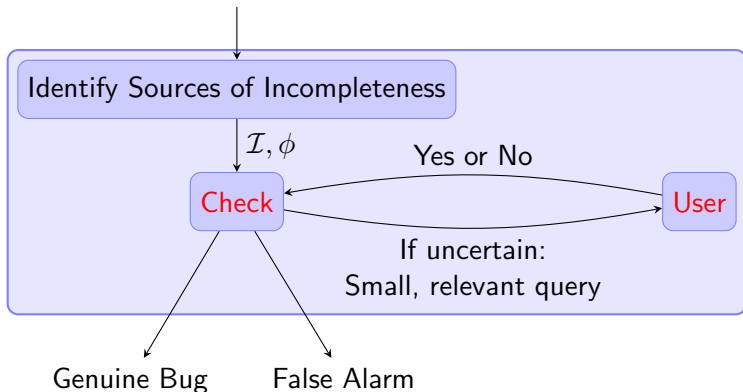
## Lemma

*If  $\mathcal{I} \models \phi$ , then the program is error-free (assertion always succeeds)*

*If  $\mathcal{I} \models \neg\phi$ , then the program must be buggy (assertion always fails)*



Program with Inferred Invariants  
and Potential Error Report



# Proof Obligation

Given known facts  $\mathcal{I}$  and success condition  $\phi$ , a *proof obligation* is a formula  $\Gamma$  that – together with  $\mathcal{I}$  – proves  $\phi$ :

$$\Gamma \wedge \mathcal{I} \models \phi \quad \text{and} \quad \text{SAT}(\Gamma \wedge \mathcal{I})$$

Given known facts  $\mathcal{I}$  and success condition  $\phi$ , a *proof obligation* is a formula  $\Gamma$  that – together with  $\mathcal{I}$  – proves  $\phi$ :

$$\Gamma \wedge \mathcal{I} \models \phi \quad \text{and} \quad \text{SAT}(\Gamma \wedge \mathcal{I})$$

## *Cost*( $\Gamma$ )

$$1 \cdot \# \text{abstraction variables } \alpha \in \text{Vars}(\Gamma) \\ + |\text{Vars}(\phi) \cup \text{Vars}(\mathcal{I})| \cdot \# \text{input variables } \nu \in \text{Vars}(\Gamma)$$

- The fewer variables, the better
- No input variables if possible

Given known facts  $\mathcal{I}$  and success condition  $\phi$ , a *failure witness* is a formula  $\Upsilon$  that – together with  $\mathcal{I}$  – proves  $\neg\phi$ :

$$\Upsilon \wedge \mathcal{I} \models \neg\phi \quad \text{and} \quad \text{SAT}(\Upsilon \wedge \mathcal{I})$$

## $Cost(\Upsilon)$

$$|Vars(\phi) \cup Vars(\mathcal{I})| \cdot \# \text{ abstraction variables } \alpha \in Vars(\Upsilon) \\ + 1 \cdot \# \text{ input variables } \nu \in Vars(\Upsilon)$$

- The fewer variables, the better
- Prefer input variables

## Weakest Minimum Proof Obligation $\Gamma$

- costs less than or equal to any other proof obligation, and
- is no stronger than any other proof obligations with same cost

## Weakest Minimum Failure Witness $\Upsilon$ Dito

Ask the user the one with lower cost

- *Does  $\Gamma$  hold in all program executions?*

Yes Program is error-free (because  $\Gamma \wedge \mathcal{I} \models \phi$ )

No Add  $\neg\Gamma$  to known witnesses and maybe ask another query

- *May  $\Upsilon$  arise in some execution?*

Yes Programm is buggy (because  $\Upsilon \wedge \mathcal{I} \models \neg\phi$ )

No Add  $\neg\Upsilon$  to known facts  $\mathcal{I}$  and maybe ask another query

# Example

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```

$$\mathcal{I} = (\alpha_i \geq 0 \wedge \alpha_i > \nu_n \wedge \nu_n \geq 0)$$

$$\phi = (\mathbf{1} + \alpha_i + \alpha_j > 2 * \nu_n \wedge \neg \nu_{flag}) \vee$$
$$(\nu_n + \alpha_i + \alpha_j > 2 * \nu_n \wedge \nu_{flag})$$

Weakest Minimum Proof Obligation  $\Gamma = (\alpha_j \geq \nu_n)$

Weakest Minimum Failure Witness  $\Upsilon = (\neg \nu_{flag} \wedge \alpha_i + \alpha_j < 0)$

# Example

```
1 void foo(int flag, unsigned int n) {
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$$\mathcal{I} = (\alpha_i \geq 0 \wedge \alpha_i > \nu_n \wedge \nu_n \geq 0)$$

$$\phi = (\mathbf{1} + \alpha_i + \alpha_j > 2 * \nu_n \wedge \neg \nu_{flag}) \vee$$
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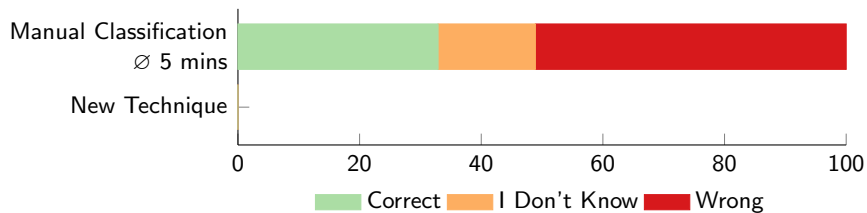
Weakest Minimum Proof Obligation  $\Gamma = (\alpha_j \geq \nu_n)$  ✓ (false alarm!)

Weakest Minimum Failure Witness  $\Upsilon = (\neg \nu_{flag} \wedge \alpha_i + \alpha_j < 0)$

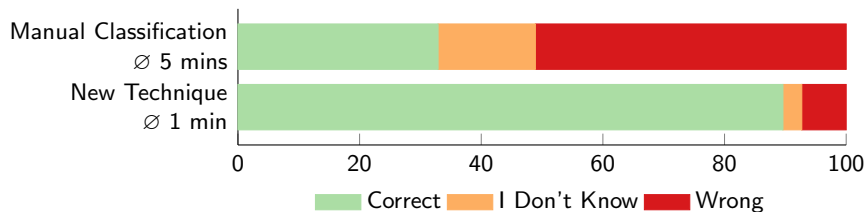


- 56 professional C programmers
- Classify 11 uncertain error reports for real-world code as
  - Genuine bugs (5), or
  - False alarms (6), or
  - *I don't know*
- Randomly assigned to classify manually or using the new technique

# User Study: Results



# User Study: Results



## Explaining Error Traces in Model Checking

Requires counter-example, does not address false alarms

## Counterexample-Guided Abstraction Refinement (CEGAR)

Learn new predicates from concrete counter-example trace

Fully automatic, but not guaranteed to terminate

- Implementation not (yet) publicly available
- Practical technique to help programmers classify error reports
- Tool-agnostic

# Questions

Program  $P$          $:=$     $\lambda \vec{a}. (\text{let } \vec{v} \text{ in } (s; \text{check}(p)))$   
Statement  $s$         $:=$     $v = e \mid \text{skip} \mid s_1; s_2$   
                          $\mid \text{if}(p) \text{ then } s_1 \text{ else } s_2$   
                          $\mid \text{while}^\rho(p)\{s\}[@p']?$   
Expression  $e$       $:=$     $v \mid c \mid c * e \mid e_1 \oplus e_2 \ (\oplus \in \{+, -\})$   
Predicate  $p$         $:=$     $e_1 \oslash e_2 \ (\oslash \in \{<, >, =\})$   
                          $\mid p_1 \wedge p_2 \mid p_1 \vee p_2 \mid \neg p$

# Operational Semantics of the Language

$$\begin{array}{c}
 \frac{}{S \vdash v : S(v)} \quad \frac{}{S \vdash c : c} \quad \frac{\oplus \in \{+, -, *\}}{S \vdash e_1 : c_1 \quad S \vdash e_2 : c_2} \\
 \frac{}{S \vdash e_1 : c_1 \quad S \vdash e_2 : c_2} \\
 b = \begin{cases} \text{true} & \text{if } c_1 \odot c_2 \\ \text{false} & \text{otherwise} \end{cases} \quad \frac{\text{lop} \in \{\wedge, \vee\}}{S \vdash p_1 : b_1 \quad S \vdash p_2 : b_2} \\
 \frac{}{S \vdash e_1 \odot e_2 : b} \quad \frac{}{S \vdash p_1 \text{ lop } p_2 : b_1 \text{ lop } b_2} \\
 \frac{S \vdash p : b}{S \vdash \neg p : \neg b} \quad \frac{S \vdash e : c}{S \vdash v = e : S[c/v]} \quad \frac{}{S \vdash \text{skip} : S} \\
 \frac{S \vdash p : \text{true} \quad S \vdash s_1 : S_1}{S \vdash \text{if}(p) \text{ then } s_1 \text{ else } s_2 : S_1} \quad \frac{S \vdash p : \text{false} \quad S \vdash s_2 : S_2}{S \vdash \text{if}(p) \text{ then } s_1 \text{ else } s_2 : S_2} \\
 \frac{S \vdash s_1 : S_1 \quad S_1 \vdash s_2 : S_2}{S \vdash s_1 ; s_2 : S_2} \quad \frac{S \vdash p : \text{true} \quad S \vdash s : S'}{S' \vdash \text{loop}^\rho(p)\{s\} : S''} \\
 \frac{}{S \vdash \text{loop}^\rho(p)\{s\} : S''} \\
 \frac{S \vdash \text{loop}^\rho(p)\{s\} : S' \quad S' \vdash p' : \text{true}}{S \vdash \text{while}^\rho(p)\{s\}[\text{@}p'] : S'} \quad \frac{S \vdash p : \text{false}}{S \vdash \text{loop}^\rho(p)\{s\} : S} \\
 \frac{S = [c_1/a_1, \dots, c_k/a_k][0/v_1, \dots, 0/v_n]}{S \vdash s : S' \quad S' \vdash p : b} \\
 \frac{}{S \vdash \lambda \vec{a}. (\text{let } \vec{v} \text{ in } (s; \text{check}(p))) (c_1, \dots, c_k) : b}
 \end{array}$$



$$\begin{aligned}\theta_1 &= \{(\pi_1, \phi_1), \dots, (\pi_k, \phi_k)\} \\ \theta_2 &= \{(\pi'_1, \phi'_1), \dots, (\pi'_n, \phi'_n)\} \\ \theta &= \bigcup_{ij} ((\pi_i \oplus \pi'_j), (\phi_i \wedge \phi'_j))\end{aligned}$$

---

$$\vdash \theta_1 \oplus \theta_2 : \theta$$

$$\begin{aligned}\theta_1 &= \{(\pi_1, \phi_1), \dots, (\pi_k, \phi_k)\} \\ \theta_2 &= \{(\pi'_1, \phi'_1), \dots, (\pi'_n, \phi'_n)\} \\ \phi &= \bigvee_{ij} ((\pi_i \otimes \pi'_j) \wedge \phi_i \wedge \phi'_j)\end{aligned}$$

---

$$\vdash \theta_1 \otimes \theta_2 : \phi$$

$$\theta' = \bigcup_{(\pi_i, \phi_i) \in \theta} (\pi_i, (\phi_i \wedge \phi))$$

---

$$\vdash \theta \wedge \phi : \theta'$$

# Symbolic Evaluation Rules for Expressions and Predicates

$$\begin{array}{c} \overline{\mathbb{S} \vdash v : \mathbb{S}(v)} \quad \overline{\mathbb{S} \vdash c : (c, \text{true})} \quad \frac{\oplus \in \{+, -, *\} \quad \mathbb{S} \vdash e_1 : \theta_1 \quad \mathbb{S} \vdash e_2 : \theta_2}{\mathbb{S} \vdash e_1 \oplus e_2 : \theta_1 \oplus \theta_2} \\ \\ \frac{\mathbb{S} \vdash e_1 : \theta_1 \quad \mathbb{S} \vdash e_2 : \theta_2}{\mathbb{S} \vdash e_1 \odot e_2 : \theta_1 \odot \theta_2} \quad \frac{\text{lop} \in \{\wedge, \vee\} \quad \mathbb{S} \vdash p_1 : \phi_1 \quad \mathbb{S} \vdash p_2 : \phi_2}{\mathbb{S} \vdash p_1 \text{ lop } p_2 : \phi_1 \text{ lop } \phi_2} \quad \frac{\mathbb{S} \vdash p : \phi}{\mathbb{S} \vdash \neg p : \neg \phi} \end{array}$$

# Transformers for the Symbolic Evaluation

$$\frac{\mathbb{S} \vdash e : \theta \quad \mathbb{S}' = \mathbb{S}[\theta/v]}{\mathbb{S}, \mathcal{I} \vdash v = e : \mathbb{S}', \mathcal{I}} \quad \frac{}{\mathbb{S}, \mathcal{I} \vdash \text{skip} : \mathbb{S}, \mathcal{I}} \quad \frac{\mathbb{S}, \mathcal{I} \vdash s_1 : \mathbb{S}_1, \mathcal{I}_1 \quad \mathbb{S}_1, \mathcal{I}_1 \vdash s_2 : \mathbb{S}_2, \mathcal{I}_2}{\mathbb{S}, \mathcal{I} \vdash s_1; s_2 : \mathbb{S}_2, \mathcal{I}_2}$$

$$\frac{\mathbb{S} \vdash p : \phi \quad \mathbb{S}, \mathcal{I} \vdash s_1 : \mathbb{S}_1, \mathcal{I}_1 \quad \mathbb{S}, \mathcal{I} \vdash s_2 : \mathbb{S}_2, \mathcal{I}_2 \quad \mathbb{S}' = (\mathbb{S}_1 \wedge \phi) \sqcup (\mathbb{S}_2 \wedge \neg\phi) \quad \mathcal{I}' = ((\phi \Rightarrow \mathcal{I}_1) \wedge (\neg\phi \Rightarrow \mathcal{I}_2))}{\mathbb{S}, \mathcal{I} \vdash \text{if}(p) \text{ then } s_1 \text{ else } s_2 : \mathbb{S}', \mathcal{I}'}$$

$$\frac{\mathbb{S}' = \mathbb{S}[(\alpha_1^\rho, \text{true})/v_1, \dots, (\alpha_k^\rho, \text{true})/v_k](\vec{v} \text{ modified in } s)}{\mathbb{S}, \mathcal{I} \vdash \text{loop}^\rho(p)\{s\} : \mathbb{S}', \mathcal{I}}$$

$$\frac{\mathbb{S}, \mathcal{I} \vdash \text{loop}^\rho(p)\{s\} : \mathbb{S}', \mathcal{I} \quad \mathbb{S}' \vdash p' : \phi}{\mathbb{S}, \mathcal{I} \vdash \text{while}^\rho(p)\{s\}[@p'] : \mathbb{S}', \mathcal{I} \wedge \phi}$$

$$\frac{\mathbb{S} = [(\nu_1, \text{true})/a_1, \dots, (\nu_k, \text{true})/a_k] \quad \mathbb{S}' = \mathbb{S}[(0, \text{true})/v_1, \dots, (0, \text{true})/v_n] \quad \mathbb{S}', \text{true} \vdash s : \mathbb{S}'', \mathcal{I} \quad \mathbb{S}'' \vdash p : \phi}{\vdash \lambda \vec{a}. (\text{let } \vec{v} \text{ in } (s; \text{check}(p))) : \mathcal{I}, \phi}$$

## Proof Obligation

Given known facts  $\mathcal{I}$  and success condition  $\phi$ , a *proof obligation* is a formula  $\Gamma$  such that

$$\Gamma \wedge \mathcal{I} \models \phi \quad \text{and} \quad \text{SAT}(\Gamma \wedge \mathcal{I})$$

## Cost of Proof Obligation

Let  $\Gamma$  be a proof obligation query for  $\mathcal{I}, \phi$ , and let  $\Pi_p$  be a mapping from variables to costs such that  $\Pi_p(\alpha) = 1$  for abstraction variable  $\alpha$  and  $\Pi_p(\nu) = |\text{Vars}(\phi) \cup \text{Vars}(\mathcal{I})|$  for input variable  $\nu$ . Then,

$$\text{Cost}(\Gamma) = \sum_{v \in \text{Vars}(\Gamma)} \Pi_p(v)$$

## Weakest Minimum Proof Obligation

Given known facts  $\mathcal{I}$  and success condition  $\phi$ , a *weakest minimum proof obligation* is a formula  $\Gamma$  such that

- 1  $\Gamma \wedge \mathcal{I} \models \phi$  and  $SAT(\Gamma \wedge \mathcal{I})$
- 2 For any other  $\Gamma'$  that satisfies 1, either  $Cost(\Gamma) < Cost(\Gamma')$  or  $Cost(\Gamma) = Cost(\Gamma') \wedge (\Gamma \not\Rightarrow \Gamma' \vee \Gamma \Leftrightarrow \Gamma')$

# Computing Weakest Minimum Proof Obligations

First, rewrite  $\Gamma \wedge \mathcal{I} \models \phi$  as  $\Gamma \models \mathcal{I} \Rightarrow \phi$ .

## Cost of Partial Assignment

Let  $\sigma$  be a partial assignment for a formula  $\phi$  and let  $\Pi$  be a mapping from variables in  $\phi$  to non-negative integers. The cost of partial assignment  $\sigma$  is

$$\text{Cost}(\sigma) = \sum_{v \in \text{Vars}(\sigma)} \Pi(v)$$

## Minimum Satisfying Assignment

Given mapping  $\Pi$  from variables to costs, a minimum satisfying assignment of formula  $\varphi$  is a partial assignment  $\sigma$  to a subset of the variables in  $\varphi$  such that

- $\sigma(\varphi) \equiv \text{true}$
- $\forall \sigma'$  such that  $\sigma'(\varphi) \equiv \text{true}$ ,  $\text{Cost}(\sigma) \leq \text{Cost}(\sigma')$

Minimum satisfying assignments help determine the minimum set of variables that any proof obligation  $\Gamma$  must contain.

## Consistent Minimum Satisfying Assignment

A minimum satisfying assignment  $\sigma$  of  $\varphi$  is consistent with  $\varphi'$  if  $\sigma(\varphi')$  is satisfiable.

Assignments that falsify  $\mathcal{I}$  are not interesting. We want a minimum satisfying assignment to  $\mathcal{I} \Rightarrow \phi$  that is consistent with  $\mathcal{I}$ .

Interpret  $\sigma$  as a logical formula  $F_\sigma$ .  $F_\sigma$  is a *strongest* proof obligation. It assigns each variable to a concrete value.

We want the *weakest sufficient condition* of  $\mathcal{I} \Rightarrow \phi$  containing only variables in  $\sigma$ .

## Lemma

*Let  $V$  be the set of variables in a minimum satisfying assignment of  $\mathcal{I} \Rightarrow \phi$  consistent with  $\mathcal{I}$ , and let  $\bar{V}$  be the set of variables in  $\mathcal{I} \Rightarrow \phi$  but not in  $V$ . We can obtain a weakest minimum proof obligation by eliminating the quantifiers from the formula*

$$\forall \bar{V}. (\mathcal{I} \Rightarrow \phi)$$



## Valid Answer to Proof Obligation Query

We say that the answer to a proof obligation query  $\Gamma$  is valid iff:

- The answer is either yes or no
- If the answer is yes, then  $\Gamma$  holds on *all* program executions (i.e.,  $\Gamma$  is a program invariant)
- If the answer is no, then there is at least one execution in which  $\Gamma$  is violated

## Lemma

*Let  $\Gamma$  be a proof obligation query and suppose yes is a valid answer to this query. Then, the program is error-free.*

- Translate analysis variables into program expressions (easy)
- Decompose complex queries to a series of simpler queries
  - If  $\phi_1 \wedge \phi_2$  is an invariant, so are  $\phi_1$  and  $\phi_2$
  - If  $\phi_1 \vee \phi_2$  is a witness, so are  $\phi_1$  and  $\phi_2$
  - Convert invariant queries to CNF and witness queries to DNF
  - Treat each clause as separate, independent query
- We learn additional facts for every subquery

# Algorithm (Given $\mathcal{I}$ and $\phi$ )

```
1  W :=  $\emptyset$ 
2  while (true) {
3    if (Valid( $\mathcal{I} \Rightarrow \phi$ )) return ERROR_DISCHARGED
4    if ( $\exists \psi \in W. \text{UNSAT}(\mathcal{I} \wedge \psi \wedge \phi)$ ) return ERROR_VALIDATED
5     $V_1 = \text{ComputeMSA}(\mathcal{I} \Rightarrow \phi, W \cup \mathcal{I}, \Pi_p)$ 
6     $\Gamma = \text{ElimQuantifier}(\forall \overline{V_1}. (\mathcal{I} \Rightarrow \phi))$ 
7     $V_2 = \text{ComputeMSA}(\mathcal{I} \Rightarrow \neg \phi, W \cup \mathcal{I}, \Pi_w)$ 
8     $\Upsilon = \text{ElimQuantifier}(\forall \overline{V_2}. (\mathcal{I} \Rightarrow \neg \phi))$ 
9
10   if (Cost( $\Gamma$ ) < Cost( $\Upsilon$ )) {
11      $Q_1 = \text{FormInvariantQuery}(\Gamma)$ 
12     if (answer to  $Q_1 = \text{YES}$ ) return ERROR_DISCHARGED
13      $W := W \cup \neg \Gamma$ 
14   } else {
15      $Q_2 = \text{FormWitnessQuery}(\Upsilon)$ 
16     if (answer to  $Q_2 = \text{YES}$ ) return ERROR_VALIDATED
17      $\mathcal{I} := \mathcal{I} \wedge \neg \Upsilon$ 
18   }
19 }
```

- Implemented on top of Compass analysis framework for C programs
- Also reasons about heap objects, arrays and function calls
- Sources of imprecisions are loops, non-linear arithmetic, inline assembly, etc.
- Allow the user to answer *I don't know*
- Uses own Mistral SMT solver to compute minimum satisfying assignments.



Isill Dillig, Thomas Dillig and Alex Aiken.

Automated Error Diagnosis Using Abductive Inference

*Proceedings of the 33rd ACM SIGPLAN conference on Programming Language Design and Implementation (PLDI)*, 181–192, 2012.