

# Automated Error Diagnosis Using Abductive Inference

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## Motivation



- If we use sound program analysis tool to verify a property, answer is either **yes** or **no**
- If answer is **yes**, program is error-free
- If answer is **no**, there are two possibilities:
  - Either the program is indeed buggy
  - Or report is a false alarm

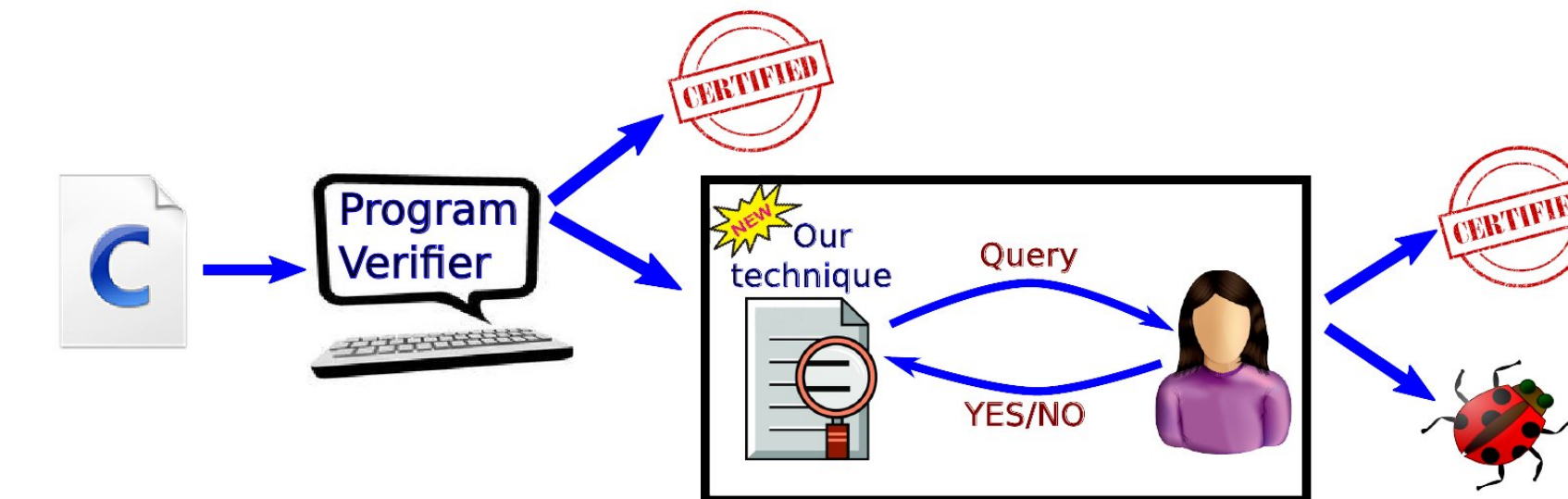
## When Verification Fails

- When verifier fails to prove property, user must decide whether report is real bug or false alarm.
- But manually classifying error reports is time-consuming and error-prone.
- Furthermore, user must redo all the reasoning the tool performed just to discover where it became stuck.
- Very painful process for most users of static analysis tools!



## Our Goal

A new technique for semi-automating error report classification when automated program verification fails



- Allows verifier to interact with user by asking small, relevant queries until report is classified as real bug or false positive
- Queries capture only the information verifier is missing  $\Rightarrow$  user contributes facts verifier could not decide on its own
- Answering queries much easier than classifying error report

## Key Ideas



**Key Idea #1:** Analysis makes explicit not only facts it knows, but also facts it does **not** know

- Sources of imprecision/incompleteness in static analysis represented using **abstraction variables**
- For example, if value of variable is unknown after a loop, represent this unknown value using abstraction variable
- This representation allows analysis to be "introspective" and reason about what facts it could be missing

## Key Ideas, cont.

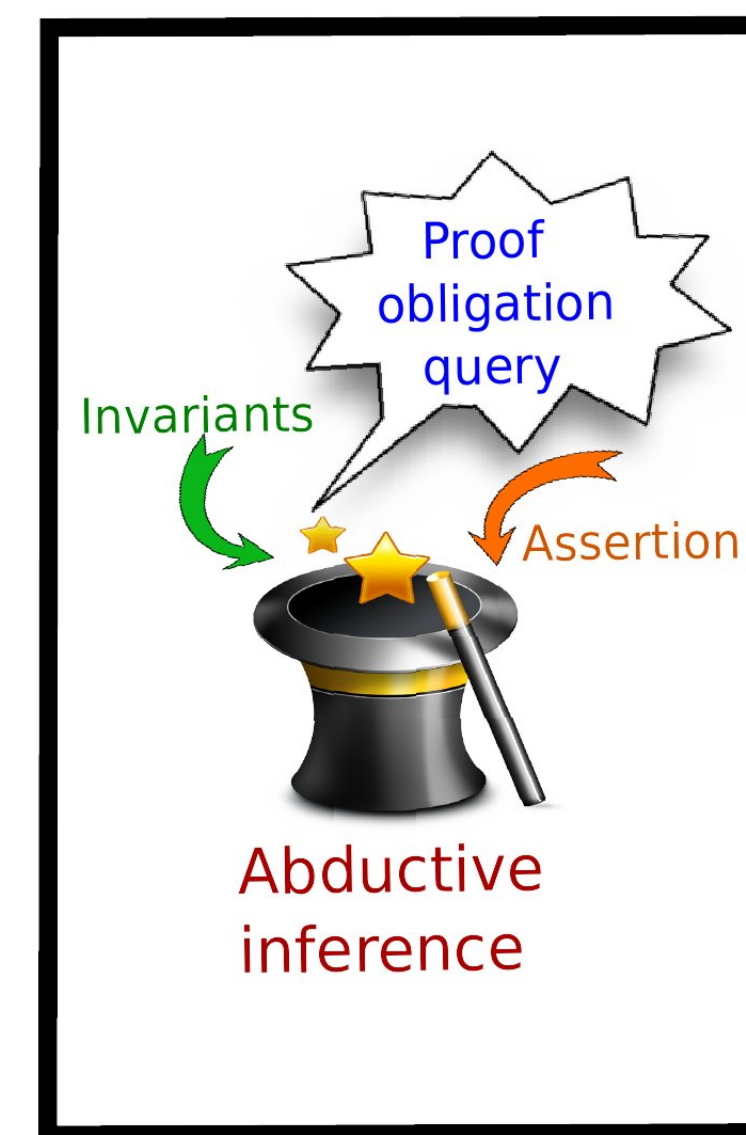


**Key Idea #2:** Abductive inference

- Given known facts  $F$  and desired outcome  $O$ , **abductive inference** finds simple explanatory hypothesis  $E$  such that
 
$$F \wedge E \models O \text{ and } \text{SAT}(F \wedge E)$$
- We use abductive inference to generate **simple explanations** that either guarantee that program is error-free or definitely buggy
- These abductive explanations are presented as queries to user

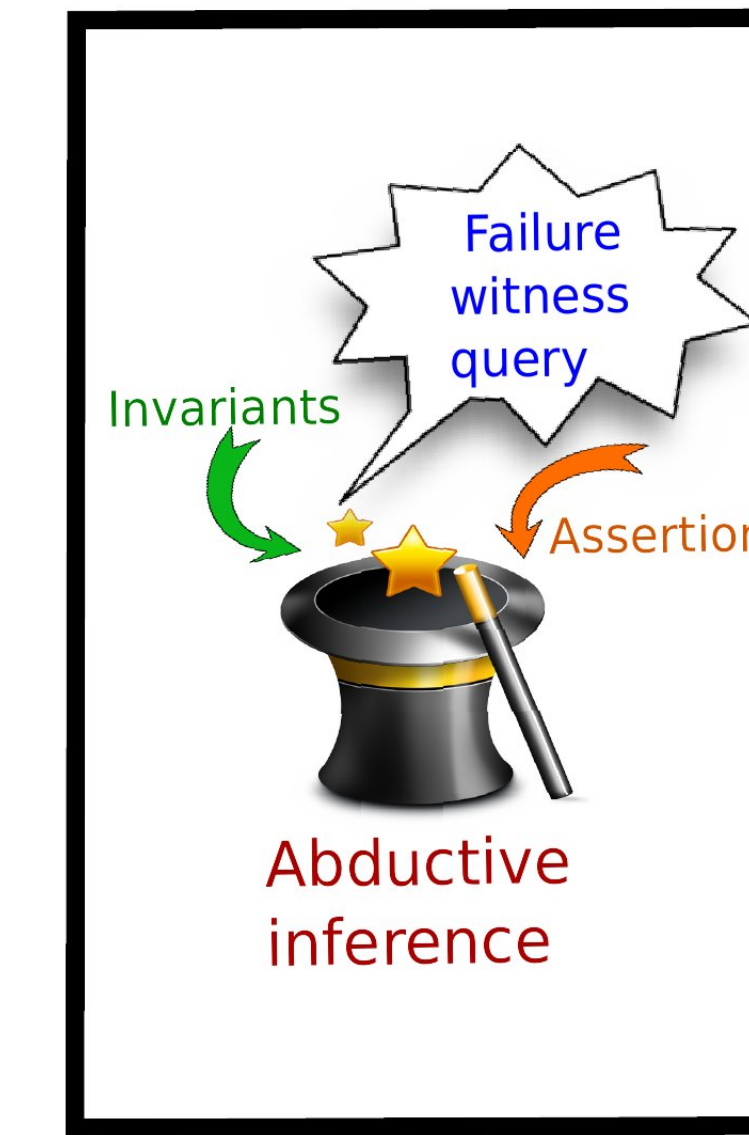
## Proof Obligation via Abductive Inference

- Input:** invariants computed by verifier and assertion to discharge
- Technique computes formulas  $I$  and  $\phi$  describing invariant and assertion in terms of abstraction variables
- Use abduction to compute simple and general explanation  $\Gamma$  s.t.:
 
$$\Gamma \wedge I \models \phi \text{ and } \text{SAT}(\Gamma \wedge I)$$
- Abductive explanation  $\Gamma$  is presented to user as **proof obligation query**
- If  $\Gamma$  is invariant, report is **false alarm**



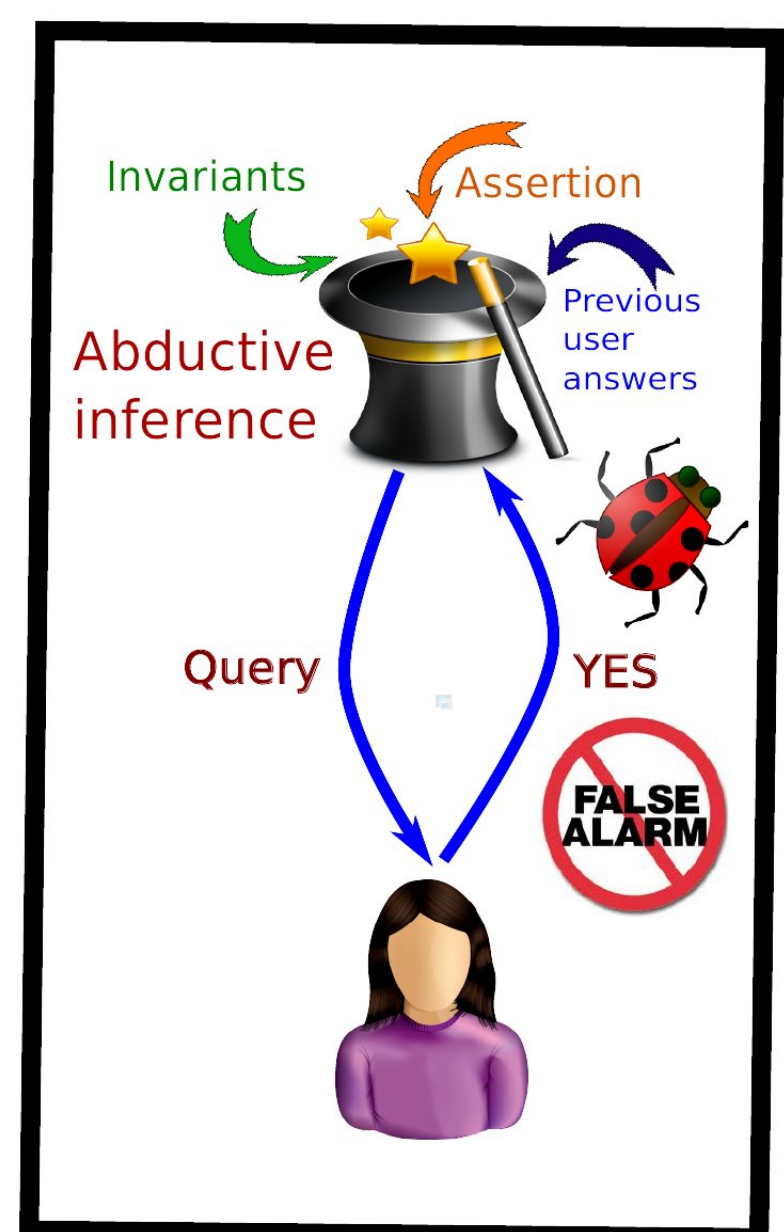
## Failure Witnesses

- Proof obligation query used to show report is false alarm
- We generate another query, called **failure witness query**, to show report is a real bug
- To generate failure witness query, solve a dual abductive inference problem:
 
$$\Delta \wedge I \models \neg \phi \text{ and } \text{SAT}(\Delta \wedge I)$$
- If  $\Delta$  can hold in **some** program execution, then report is **real bug!**



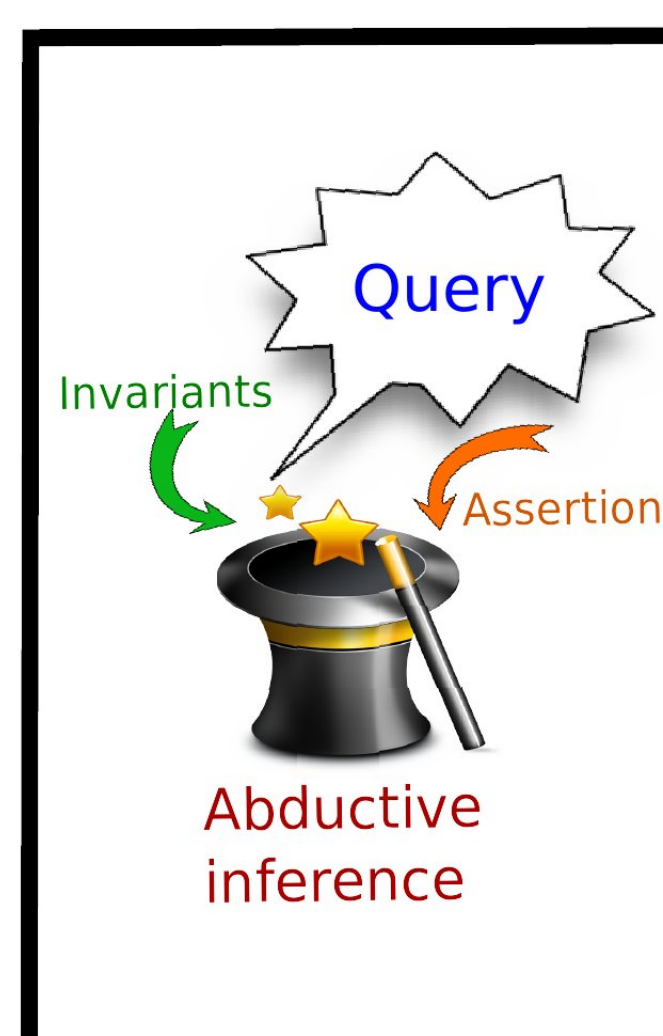
## Automated Error Diagnosis via Abductive Inference

- Our technique helps user classify error reports by generating simple queries
- If query is a proof obligation and user answers yes, report classified as false alarm
- If query is a failure witness and user answers yes, report classified as real bug
- If user answers "no" or "I don't know", technique computes new abductive explanation **distinct** from previous ones
- Interaction continues until report is classified as real bug or false alarm



## Computing Abductive Explanations

- Abduction is useful, but how do we compute these explanations?
- Given invariants  $I$  and desired outcome  $\phi$ , how to find explanation  $E$  s.t.:
 
$$I \wedge E \models \phi \wedge \text{SAT}(I \wedge E)$$
- Trivial solution is  $E = \phi$ , but useless b/c same as asking user to prove assertion!
- Want solutions that are as simple and as general as possible!



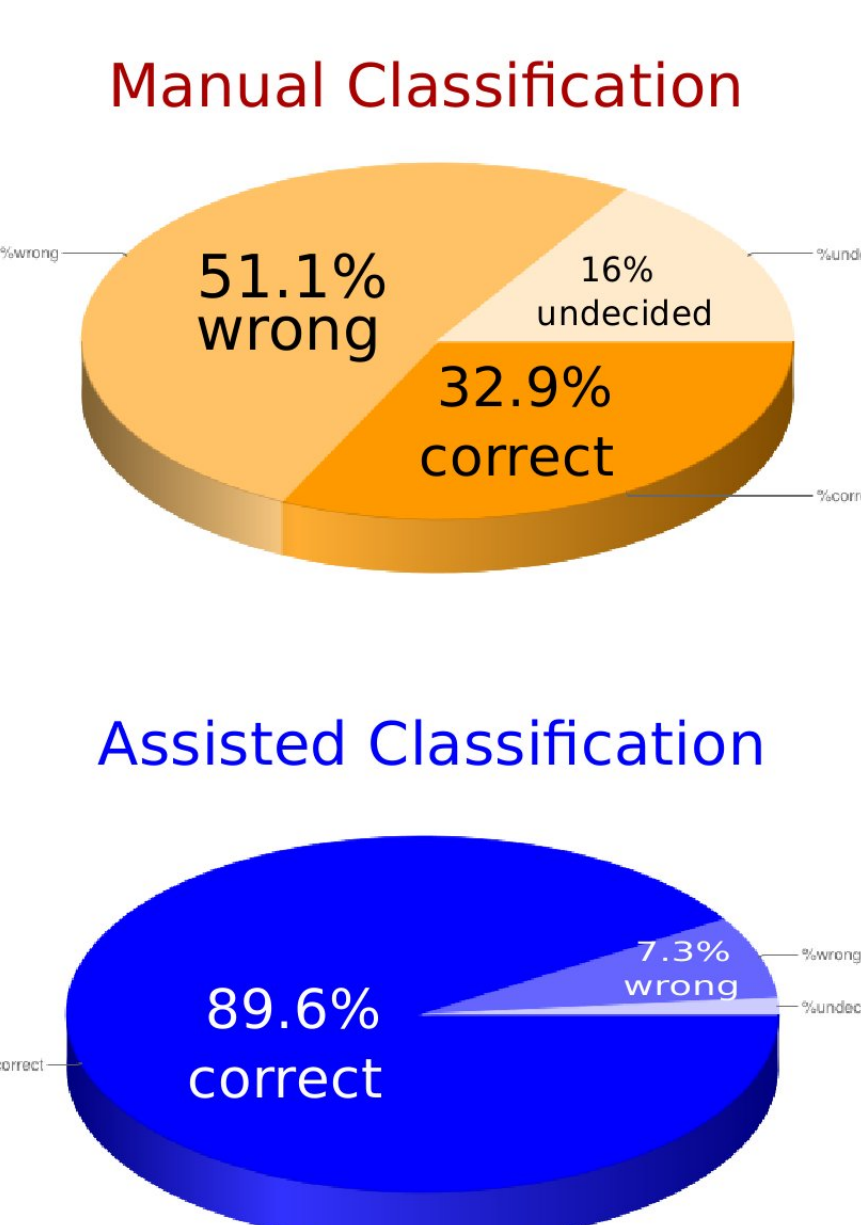
## Experimental Evaluation

- Performed user study to evaluate new technique
- Hired 56 programmers through ODesk and asked them to classify error reports
- Each programmer asked to classify (randomly selected) half of reports manually, and other half using our technique
- Manual classification:** Given code and error report, decide if bug, false alarm, or unknown
- Our technique:** Given code and series of queries, asked to answer "Yes", "No", or "Don't know"
- Based on answers to queries, report classified automatically

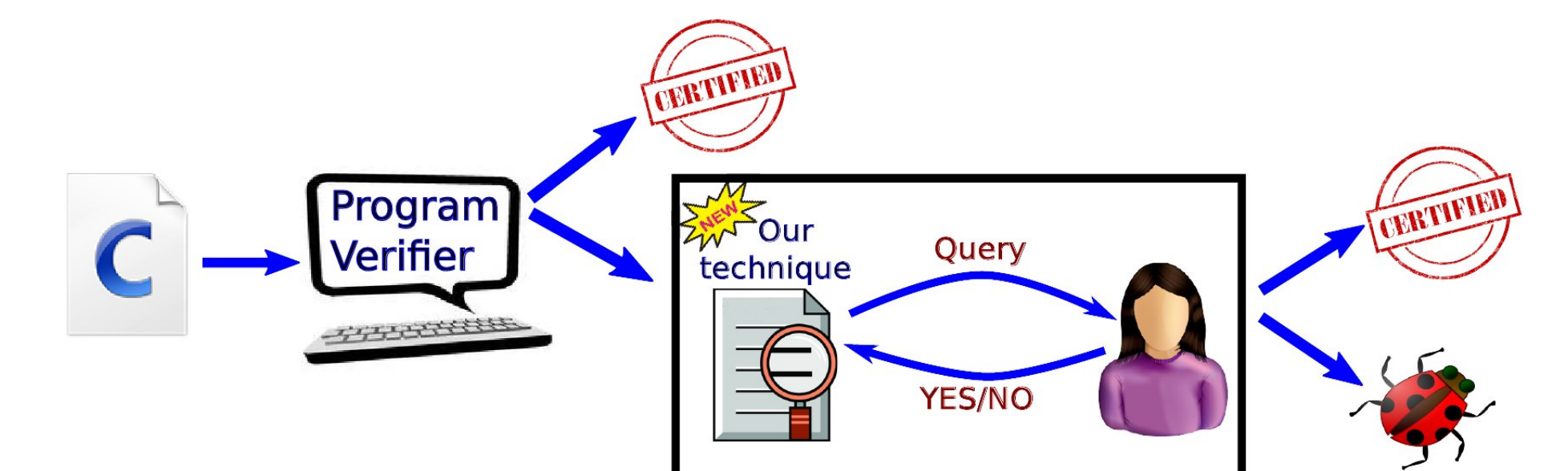


## Results of User Study

- With manual classification, programmers classified **51.1%** of reports **incorrectly**
- With assisted classification, programmers classified **only 7.3%** of reports incorrectly
- Our technique dramatically improves classification accuracy
- Also dramatically reduces time needed to classify report
- Using manual classification, programmers need **293** seconds on average
- Using new technique, programmers take **55** seconds on average



## Summary



- New technique to help programmers classify error reports as real bugs or false alarms
- Uses abductive inference to compute simple queries that capture what analysis is missing
- Interacts with user until report is classified as bug/false alarm