Advanced Programming **Methods**

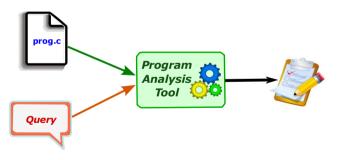
Seminar

Introduction in program analysis

Introduction in program analysis

What is Program Analysis?

- Very broad topic, but generally speaking, automated analysis of program behavior
- Program analysis is about developing algorithms and tools that can analyze other programs



Applications of Program Analysis

- Bug finding. e.g., expose as many assertion failures as possible
- Security. e.g., does an app leak private user data?
- Verification. e.g., does the program always behave according to its specification?
- Compiler optimizations. e.g., which variables should be kept in registers for fastest memory access?
- Automatic parallelization. e.g., is it safe to execute different loop iterations on parallel?

Dynamic vs. Static Program Analysis

Two flavors of program analysis:

- Dynamic analysis: Analyzes program while it is running
- Static analysis: Analyzes source code of the program

Static

- + reasons about all executions
- less precise



Dynamic

- + more precise
- results limited to
 - observed executions

Testing /Formal Verification

A very crude dichotomy:

Testing	Formal Verification
Correct with respect to the set of test inputs, and reference system	Correct with respect to all inputs, with respect to a formal specification
Easy to perform	Decidability problems, Computational problems,
Dynamic	Static

Static Program Analysis

Typical static analysis question: "Given source code of program P and desired property Q, does P exhibit Q in all possible executions?"

But this question is undecidable! This means

static analyses are either:

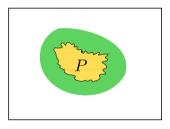
Unsound: May say program is safe even though it is unsafe

Sound, but incomplete: May say program is unsafe even though it is safe

Non-terminating: Always gives correct answer when it terminates, but may run forever

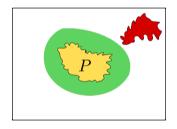
Many static analysis techniques are sound but incomplete.

Key idea: Overapproximate (i.e., abstract) program behavior

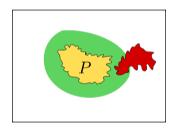


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Bad states outside over-approximation⇒ Program safe

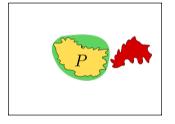


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- Bad states inside over-approximation, but outside P
 - ⇒ false alarm

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- Goal: Construct abstractions that are precise enough (i.e., few false alarms) and that scale to real programs

Examples of Abstractions

There is no "one size fits all" abstraction

 What information is useful depends on what you want to prove about the program!

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Application	Useful abstraction
No division-by-zero errors	zero vs. non-zero
Data structure verification	list, tree, graph,
No out-of-bounds array	ranges of integer
accesses	variables

How to create Sound Abstractions

- Useful theory for understanding how to design sound static analyses is abstract interpretation
 - Seminal '77 paper by Patrick & Radhia Cousot
- Not a specific analysis, but rather a framework for designing sound-by-construction static analyses
- Let's look at an example: A static analysis that tracks the sign of each integer variable (e.g., positive, non-negative, zero etc.)

First Step: Design An Abstract Domain

- An abstract domain is just a set of abstract values we want to track in our analysis
- For our example, let's fix the following abstract domain:

```
pos: \{x \mid x \in Z \land x > 0\}
zero: \{0\}
neg: \{x \mid x \in Z \land x < 0\}
non-neg: \{x \mid x \in Z \land x \geq 0\}
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In addition, every abstract domain contains:

⊤ (top): "Don't know", represents any value

⊥ (bottom): Represents empty-set

- Abstraction function (α) maps sets of concrete elements to the most precise value in the abstract domain

Second Step: Abstraction and Concretization Function

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$$\alpha(\{2, 10, 0\}) = \text{non-neg}$$
 $\alpha(\{3, 99\}) = \text{pos}$
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Concretization function (γ) maps each abstract value to sets of concrete elements

$$\gamma(pos) = \{ x \mid x \in Z \land x > 0 \}$$

Lattices and Abstract Domains

* Concretization function defines partial order on abstract values:

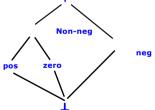
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$$A_1 \leq A_2 \text{ iff } \gamma(A_1) \subseteq \gamma(A_2)$$

Furthermore, in an abstract domain, every pair of elements has a lub and glb ⇒ mathematical lattice



 Least upper bound of two elements is called their join – useful for reasoning about control flow in programs

Almost Inverses

Important property of the abstraction and concretization function is that they are almost inverses:

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This is called a Galois insertion and captures the soundness of the abstraction

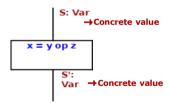
Step 3: Abstract Semantics

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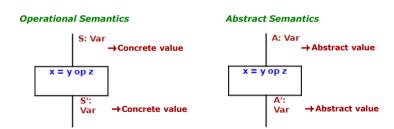
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 - Abstract counter-part of operational semantics rules

Operational Semantics



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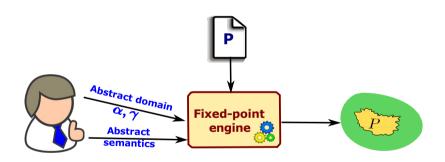
Back to Our Example

For our sign analysis, we can define abstract transformer for x = y + z as follows:

	pos	neg	zero	non-neg	Т	Τ.
pos	pos	Т	pos	pos	Т	\perp
neg	Т	neg	neg	Т	Т	\perp
zero	pos	neg	zero	non-neg	Т	\perp
non-neg	pos	Т	non-neg	non-neg	Т	\perp
Т	Т	Т	Т	Т	Т	T
Т	Т	Т	Т	Τ	Τ	Τ

To ensure soundness of static analysis, must prove that abstract semantics faithfully models concrete semantics

Putting It All Together



Fixed-point Computations

Fixed-point computation: Repeated symbolic execution of the program using abstract semantics until our approximation of the program reaches an equilibrium

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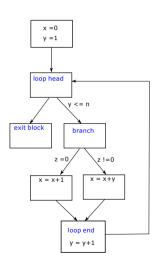
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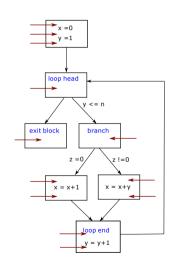


Assuming correctness of your abstract semantics, the least fixed point is an overapproximation of the program!

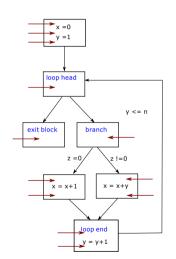
Represent program as a control-flow graph



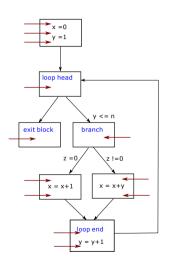
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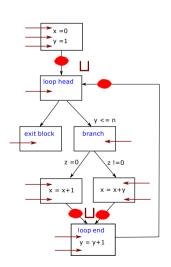


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Performing Least Fixed Point Computation

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 - Compute abstract state on entry to a basic block B by taking the join of B's predecessors

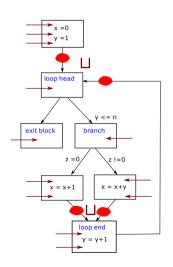


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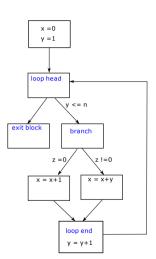
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- Repeat until no abstract state changes at any program point:
 - Compute abstract state on entry to a basic block B by taking the join of B's predecessors
 - Symbolically execute each basic block using abstract semantics



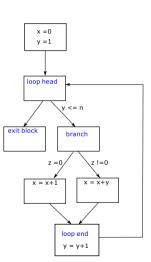
An Example

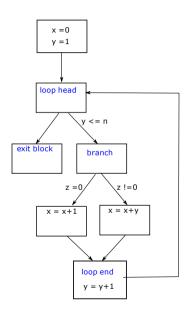
```
x = 0;
y = 0;
while(y <= n) {
  if (z == 0)
      {x = x+1;
  }
  else {
      x = x + y;
  }
  y = y+1
}
```

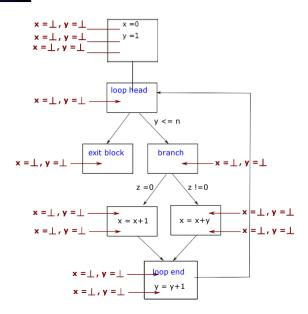


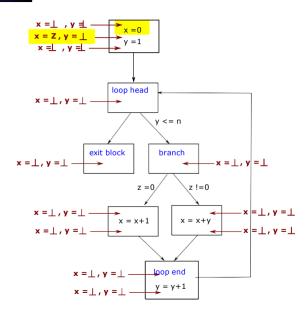
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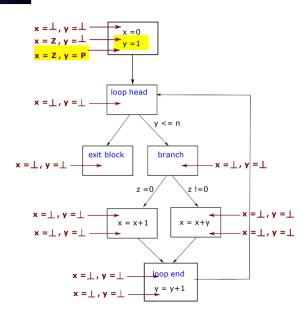
```
x = 0;
y =0;
                         Is x always
while(y \le n)
                        non-negative
 if (z == 0)
                       inside the loop?
   {x = x+1}
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 y = y+1
```

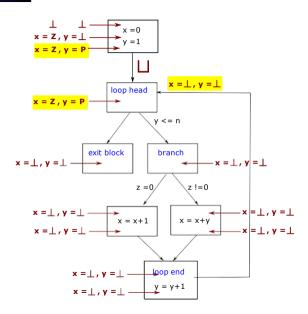


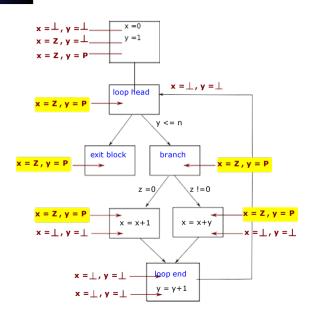


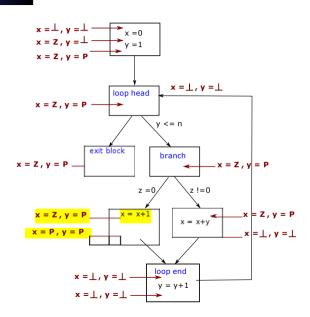


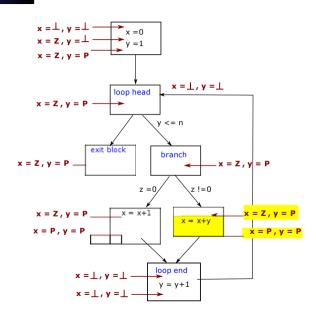


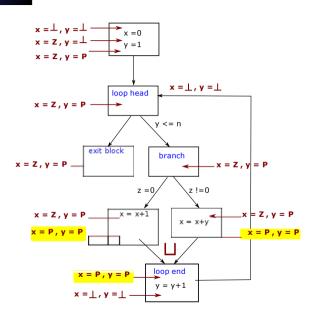


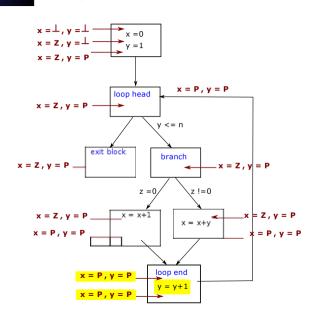


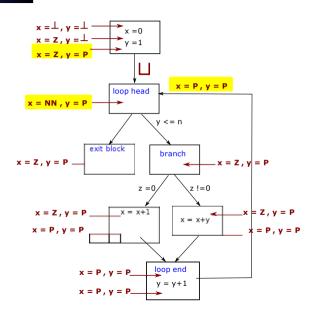


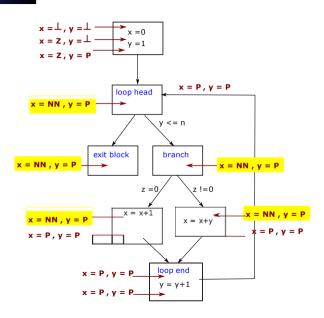


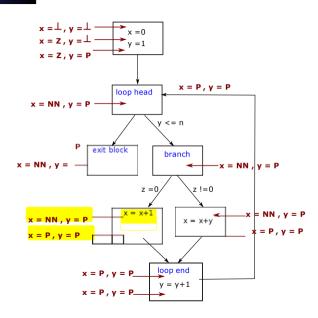


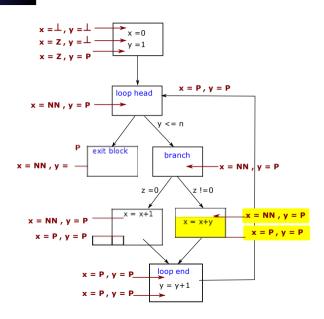


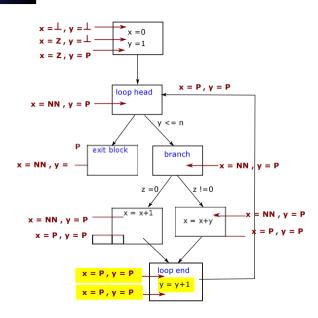


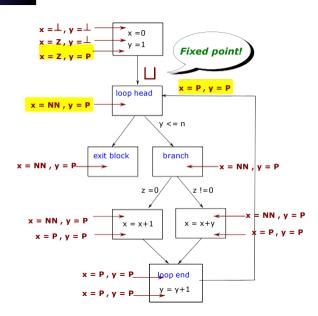












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 - Yes, assuming abstract domain forms complete lattice
 - This means every subset of elements (including infinite subsets) have a LUB
- Unfortunately, many interesting domains do not have this property, so we need widening operators for convergence.

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- But many static analyses also differ in several ways:
 - Flow (in)sensitivity: Some analyses only compute facts for the whole program, not for every program point
 - Path sensitivity: More precise analyses compute different facts for different program paths

Challenges and Open Problems

Many open problems

- Precise and scalable heap reasoning
- Concurrency
- Dealing with open programs
- Modular program analysis