

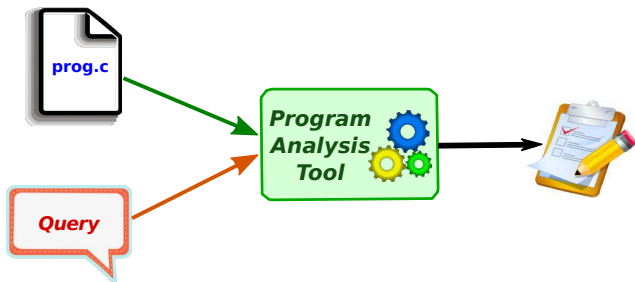
# **Advanced Programming Methods**

## **Seminar 13**

**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 <i>set of test inputs</i> , and reference system | Correct with respect to <i>all inputs</i> , with respect to a <i>formal specification</i> |
| Easy to perform  | Decidability problems,<br>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

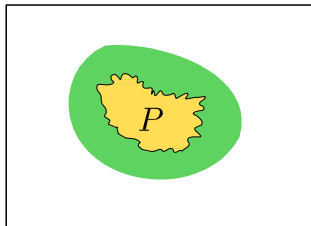
static analyses are either:

- 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.

## How to design Sound Static Analyses

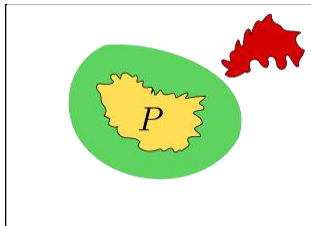
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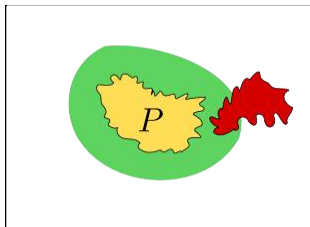
- Bad states outside over-approximation  
⇒ Program safe





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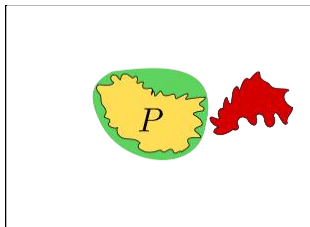
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- Bad states inside over-approximation, but outside  $P$

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



- Bad states outside over-approximation

⇒ Program safe

▪

- Bad states inside over-approximation, but outside  $P$

⇒ false alarm

⇒ **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 accesses | ranges of integer 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**:  $\{x \mid x \in \mathbb{Z} \wedge x > 0\}$

**zero**:  $\{0\}$

**neg**:  $\{x \mid x \in \mathbb{Z} \wedge x < 0\}$

**non-neg**:  $\{x \mid x \in \mathbb{Z} \wedge x \geq 0\}$

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- In addition, every abstract domain contains:
  - $\top$  (**top**): "Don't know", represents any value
  - $\perp$  (**bottom**): Represents empty-set

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



## Second Step: Abstraction and Concretization Function

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

$$\alpha(\{2, 10, 0\}) = \text{non-neg}$$

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Concretization function ( $\gamma$ ) maps each abstract value to sets of concrete elements

$$\gamma(\text{pos}) = \{x \mid x \in \mathbb{Z} \wedge x > 0\}$$

## Lattices and Abstract Domains

- Concretization function defines **partial order** on abstract values:

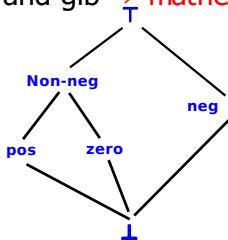
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- Concretization function defines **partial order** on abstract values:

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- Furthermore, in an abstract domain, every pair of elements has a lub and glb  $\Rightarrow$  **mathematical lattice**

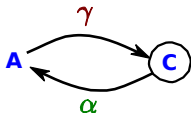


- Least upper bound of two elements is called their **join** – useful for reasoning about control flow in programs

## Almost Inverses

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

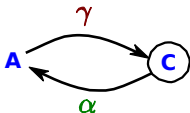
$$\alpha(\gamma(A)) = A$$



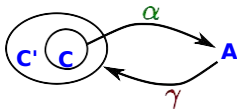
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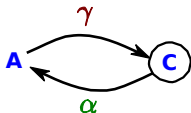
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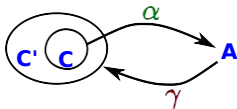
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- This is called a **Galois insertion** and captures the soundness of the abstraction

## Step 3: Abstract Semantics

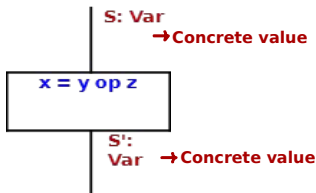
- Given abstract domain,  $\alpha$ ,  $\gamma$ , need to define **abstract transformers (i.e., semantics)** for each statement
  - Describes how statements affect our abstraction



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- Given abstract domain,  $\alpha$ ,  $\gamma$ , need to define **abstract transformers (i.e., semantics)** for each statement
  - Describes how statements affect our abstraction
  - Abstract counter-part of operational semantics rules

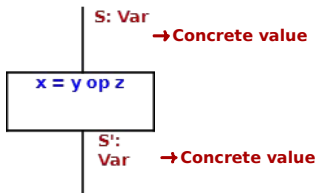
### *Operational Semantics*



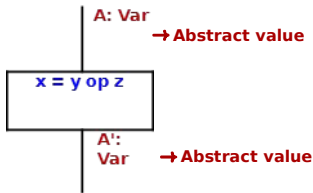
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### Operational Semantics



### Abstract Semantics



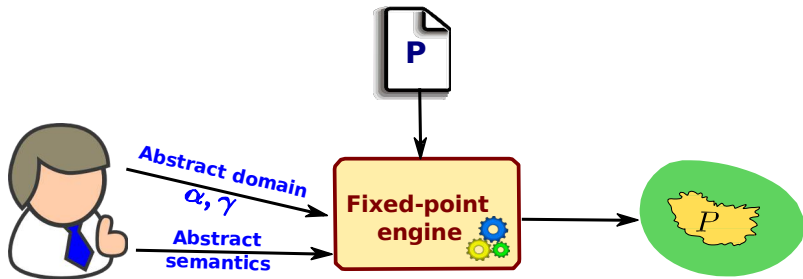
## Back to Our Example

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

|         | pos     | neg     | zero    | non-neg | $\top$  | $\perp$ |
|---------|---------|---------|---------|---------|---------|---------|
| pos     | pos     | $\top$  | pos     | pos     | $\top$  | $\perp$ |
| neg     | $\top$  | neg     | neg     | $\top$  | $\top$  | $\perp$ |
| zero    | pos     | neg     | zero    | non-neg | $\top$  | $\perp$ |
| non-neg | pos     | $\top$  | non-neg | non-neg | $\top$  | $\perp$ |
| $\top$  | $\top$  | $\top$  | $\top$  | $\top$  | $\top$  | $\perp$ |
| $\perp$ | $\perp$ | $\perp$ | $\perp$ | $\perp$ | $\perp$ | $\perp$ |

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

## Putting It All Together

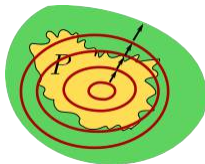


## Fixed-point Computations

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

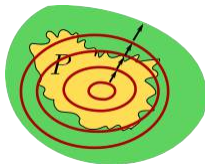
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## Fixed-point Computations

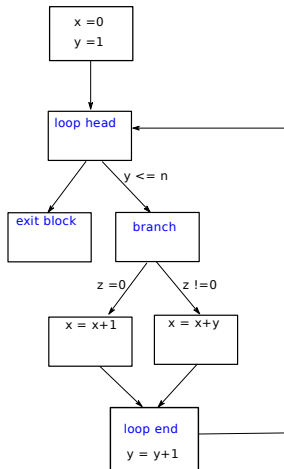
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- Assuming correctness of your abstract semantics, the least fixed point is an **overapproximation** of the program!

## Performing Least Fixed Point Computation

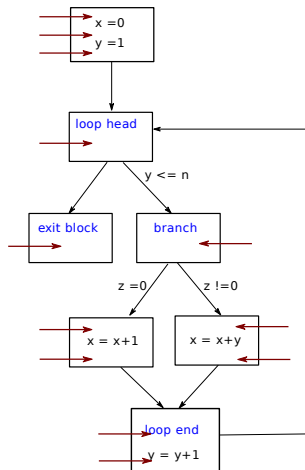
- Represent program as a control-flow graph





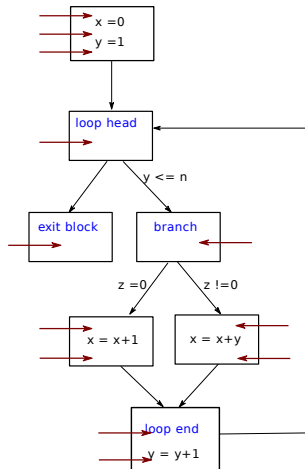
# Performing Least Fixed Point Computation

- Represent program as a **control-flow graph**
- Want to compute abstract values at every program point



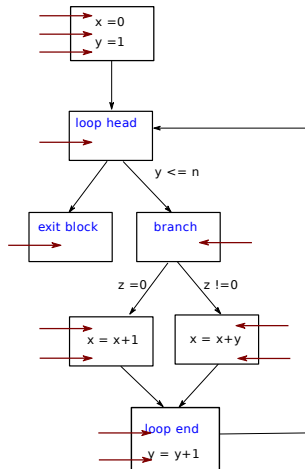
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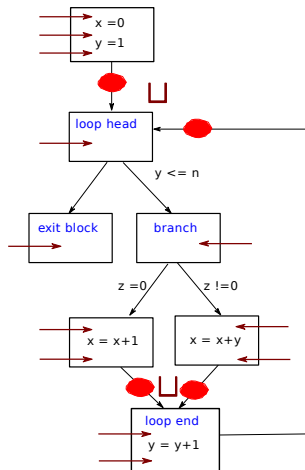
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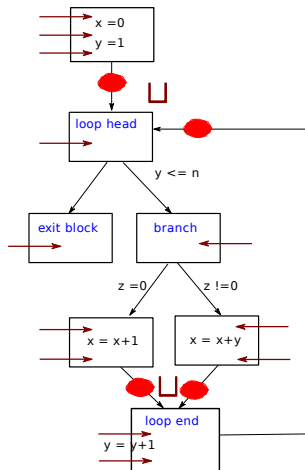
## Performing Least Fixed Point Computation

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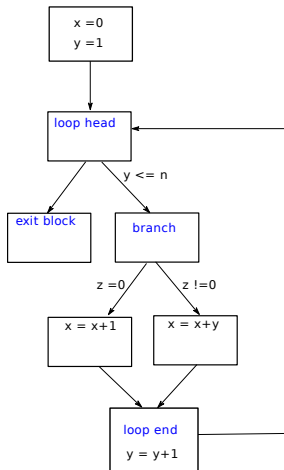
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  - Symbolically execute each basic block using abstract semantics



## An Example

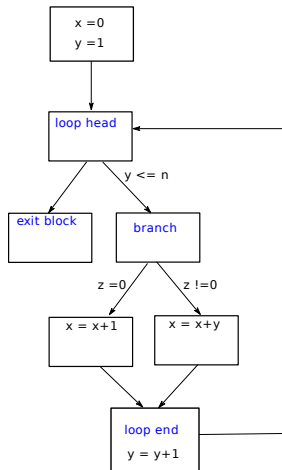
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x = 0;  
y = 0;  
  
while(y <= n)  
{  
  if (z == 0) {  
    x = x+1;  
  }  
  else {  
    x = x + y;  
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```



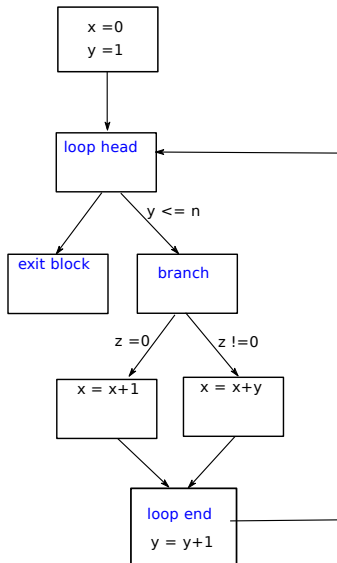
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*Is x always  
non-negative  
inside the loop?*

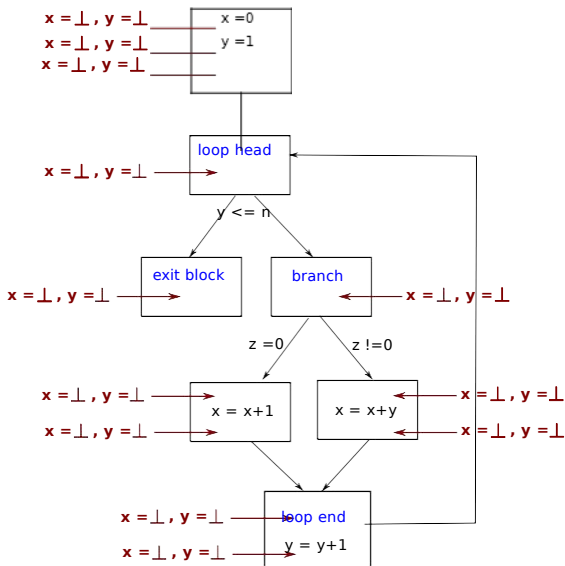


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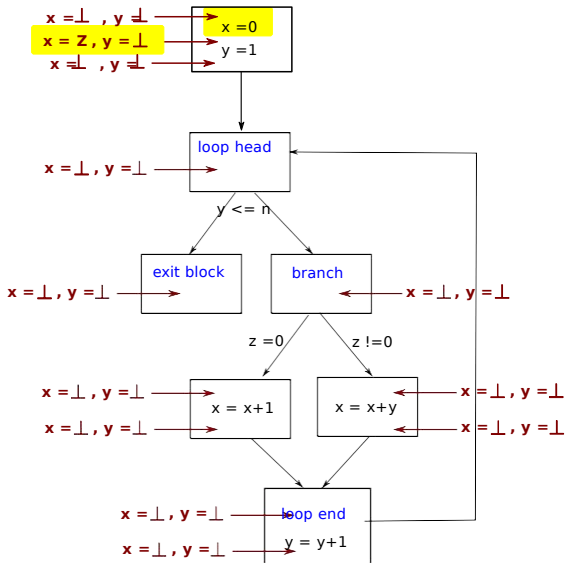




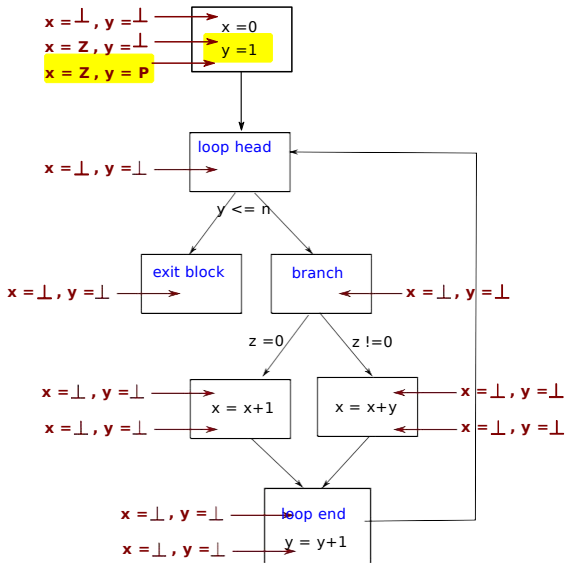
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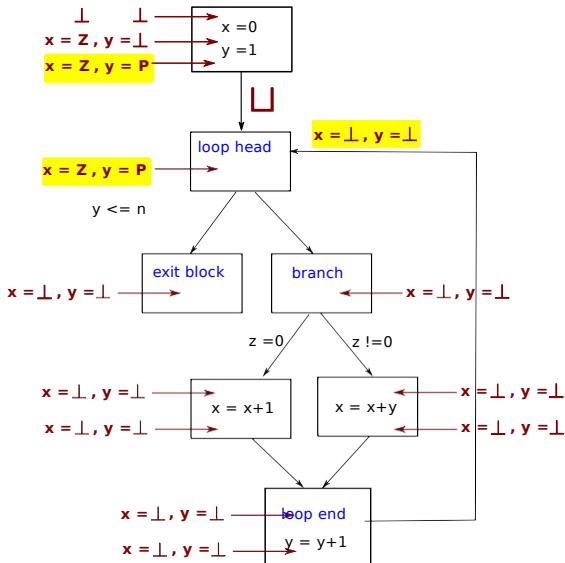
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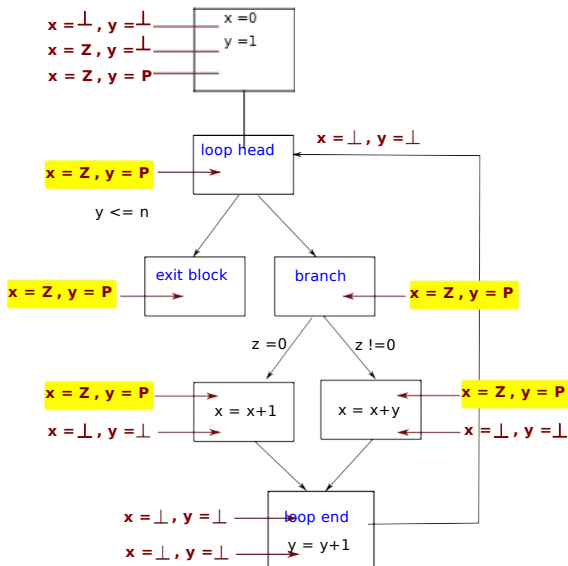
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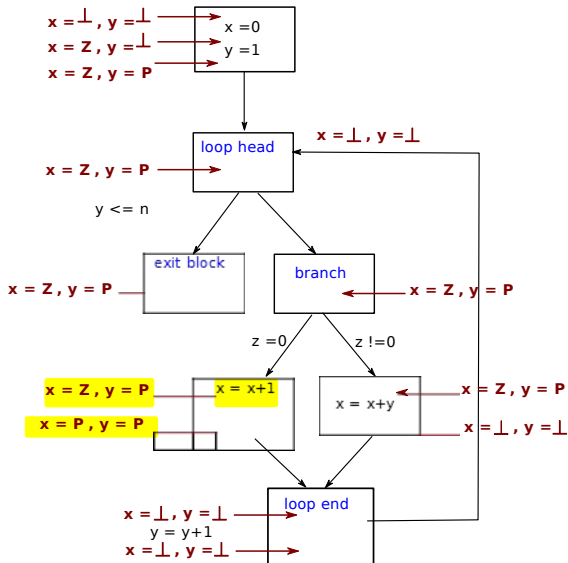
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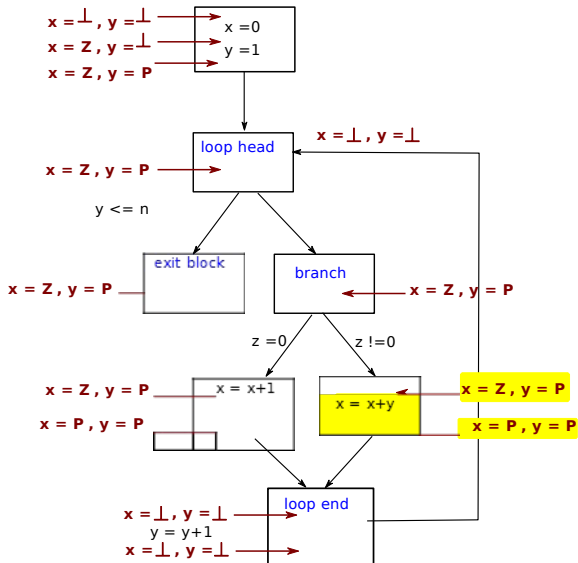
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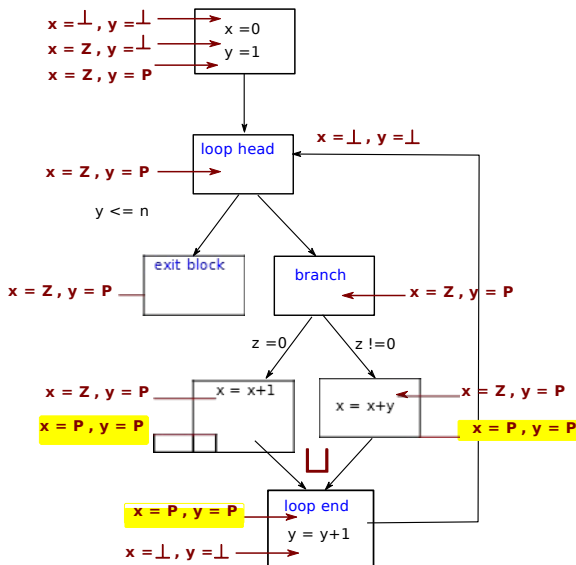
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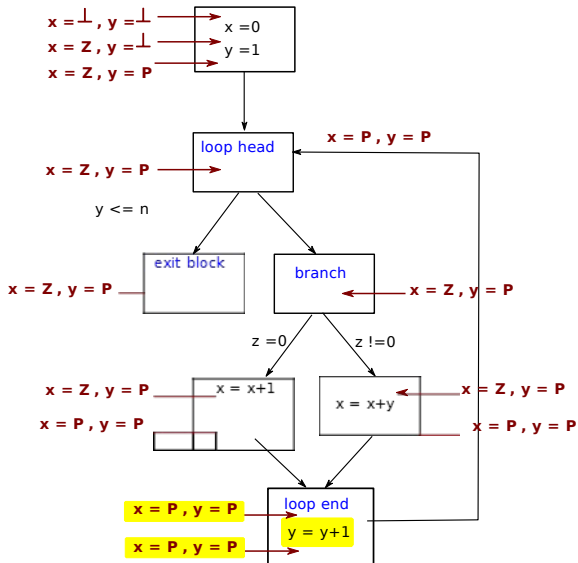


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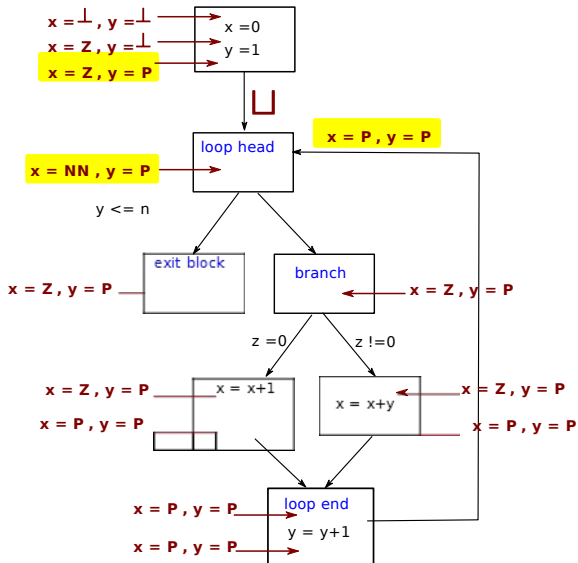




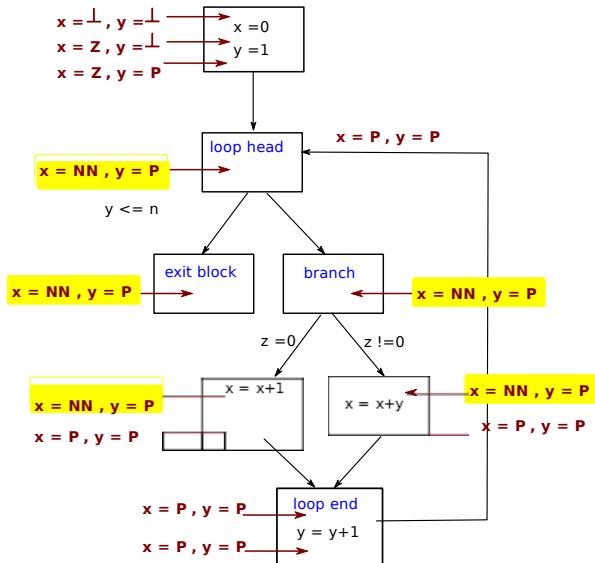
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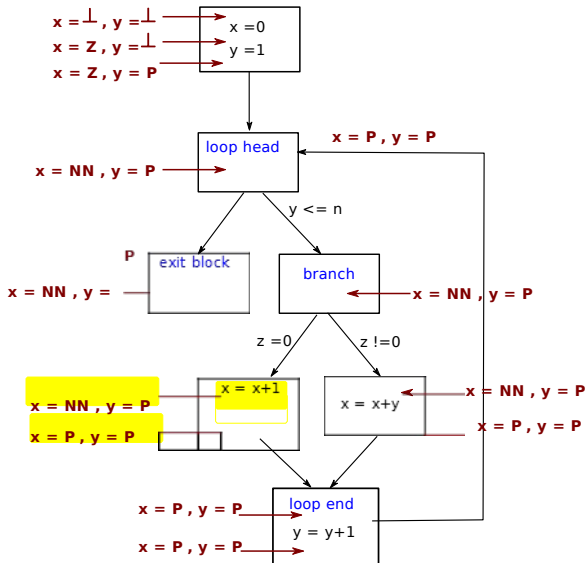
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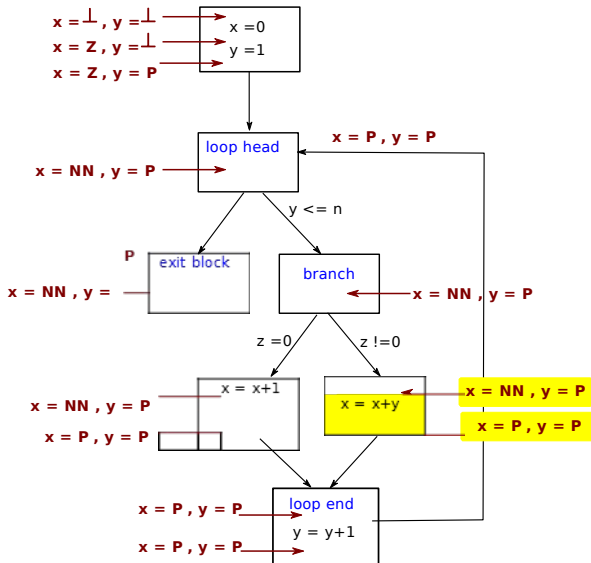
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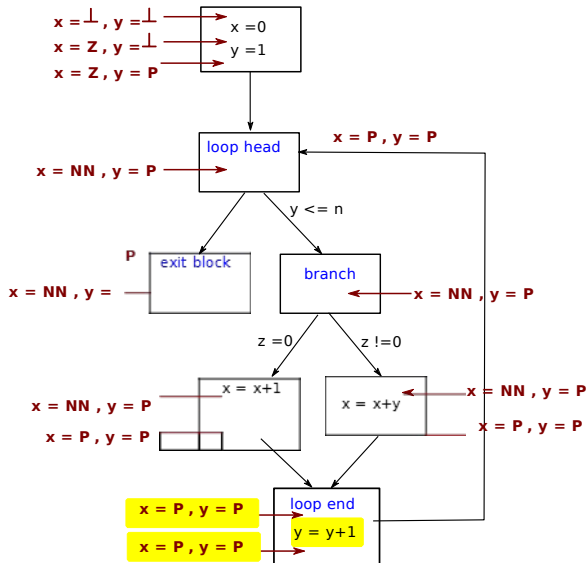
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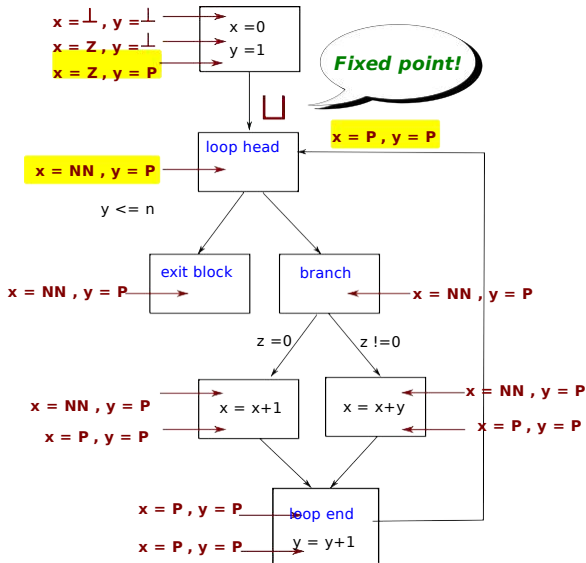
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## Termination of Fixed-Point Computation

- In this example, we quickly reached least fixed point – but does this computation always terminate?
  - 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.

## Lessons To Take Away

- Considered only one static analysis approach, but illustrates two key ideas underlying program analysis:
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- But many static analyses also differ in several ways:
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  - **Fixed-point computation:** Allows us to obtain sound overapproximation of the program
- 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