

# Software Verification

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# Part I

## Introduction

- 1 Software Verification: Why?
- 2 Software Verification: How?

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- 2 Software Verification: How?

# Ubiquity of Software in Modern Life



*Once upon a time, lecturers used hand-written transparencies with an overhead projector.*

- pens
- transparencies
- scissors
- sticky tape
- lamp
- lenses
- mirror
- screen

Nowadays **softwares** are used to design the slides and to project them

Similar evolution in many, many areas

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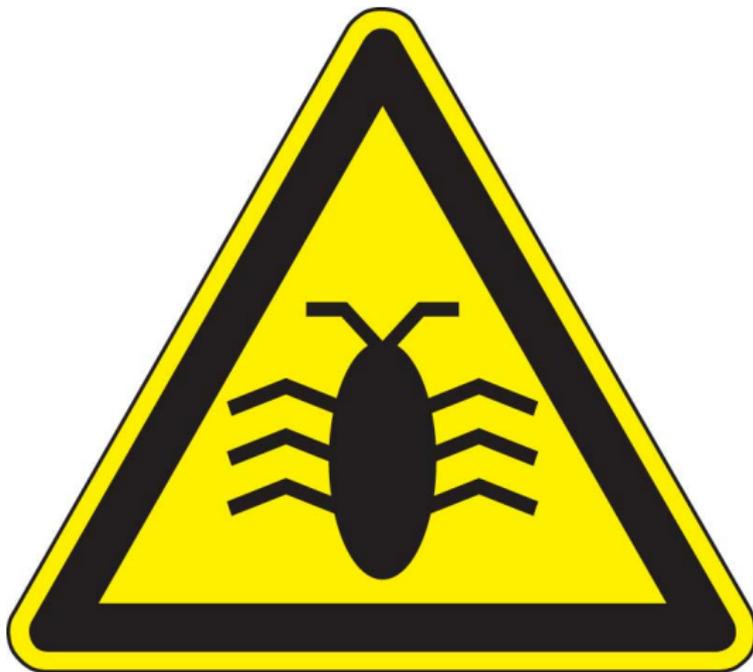
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## Some advantages of software over dedicated hardware components

- Reduce time to market
  - Less time to write the slides (really?)
  - Ability to re-organize the presentation
- Reduce costs
  - No pen, no transparencies
  - Re-usability of slides, ability to make minor modifications for free
- Increase functionality
  - Automatic generation of some slides (table of contents)
  - Nicer overlays (sticky tape is not required anymore!)
  - Ability to display videos

But software is not without risk. . .

# Bugs are Frequent in Software



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```
A problem has been detected and windows has been shut down to prevent damage
to your computer.
```

```
The problem seems to be caused by the following file: SPCMDCON.SYS
```

```
PAGE_FAULT_IN_NONPAGED_AREA
```

```
If this is the first time you've seen this Stop error screen,
restart your computer. If this screen appears again, follow
these steps:
```

```
Check to make sure any new hardware or software is properly installed.
If this is a new installation, ask your hardware or software manufacturer
for any windows updates you might need.
```

```
If problems continue, disable or remove any newly installed hardware
or software. Disable BIOS memory options such as caching or shadowing.
If you need to use Safe Mode to remove or disable components, restart
your computer, press F8 to select Advanced Startup Options, and then
select Safe Mode.
```

```
Technical information:
```

```
*** STOP: 0x00000050 (0xFD3094C2,0x00000001,0xFBFE7617,0x00000000)
```

```
*** SPCMDCON.SYS - Address FBFE7617 base at FBFE5000, DateStamp 3d6dd67c
```

# A Critical Software Bug: Ariane 5.01



*« On 4 June 1996, the maiden flight of the Ariane 5 launcher ended in a failure. Only about 40 seconds after initiation of the flight sequence, at an altitude of about 3700 m, the launcher veered off its flight path, broke up and exploded. »*

*« The failure of the Ariane 5.01 was caused by the complete loss of guidance and attitude information 37 seconds after start of the main engine ignition sequence (30 seconds after lift-off). This loss of information was due to **specification and design errors in the software of the inertial reference system.** »*

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# Software in Embedded Systems

Embedded systems in: cell phones, satellites, airplanes, cars, wireless routers, MP3 players, refrigerators, . . .

## Examples of Critical Systems

- attitude and orbit control systems in satellites
- *X-by-wire* control systems in airplanes and in cars (soon)

Increasing importance of software in embedded systems

- custom hardware replaced by processor + custom software
- software is a dominant factor in design time and cost (70 %)

Critical embedded systems require “exhaustive” validation

# Software Complexity Grows Exponentially

As computational power grows ...

Moore's law: « *the number of transistors on a chip doubles every two years* »

... software complexity grows ...

Wirth's Law: « *software gets slower faster than hardware gets faster* »

... and so does the number of bugs!

Watts S. Humphrey: « *5 – 10 bugs per 1000 lines of code after product test* »

Growing need for automatic validation techniques

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Growing need for automatic validation techniques

- 1 Software Verification: Why?
- 2 Software Verification: How?**

Running the executable (obtained by compilation)

- on multiple inputs
- usually on the target platform

Testing is a widespread validation approach in the software industry

- can be (partially) automated
- can detect a lot of bugs

But

Costly and time-consuming

Not exhaustive

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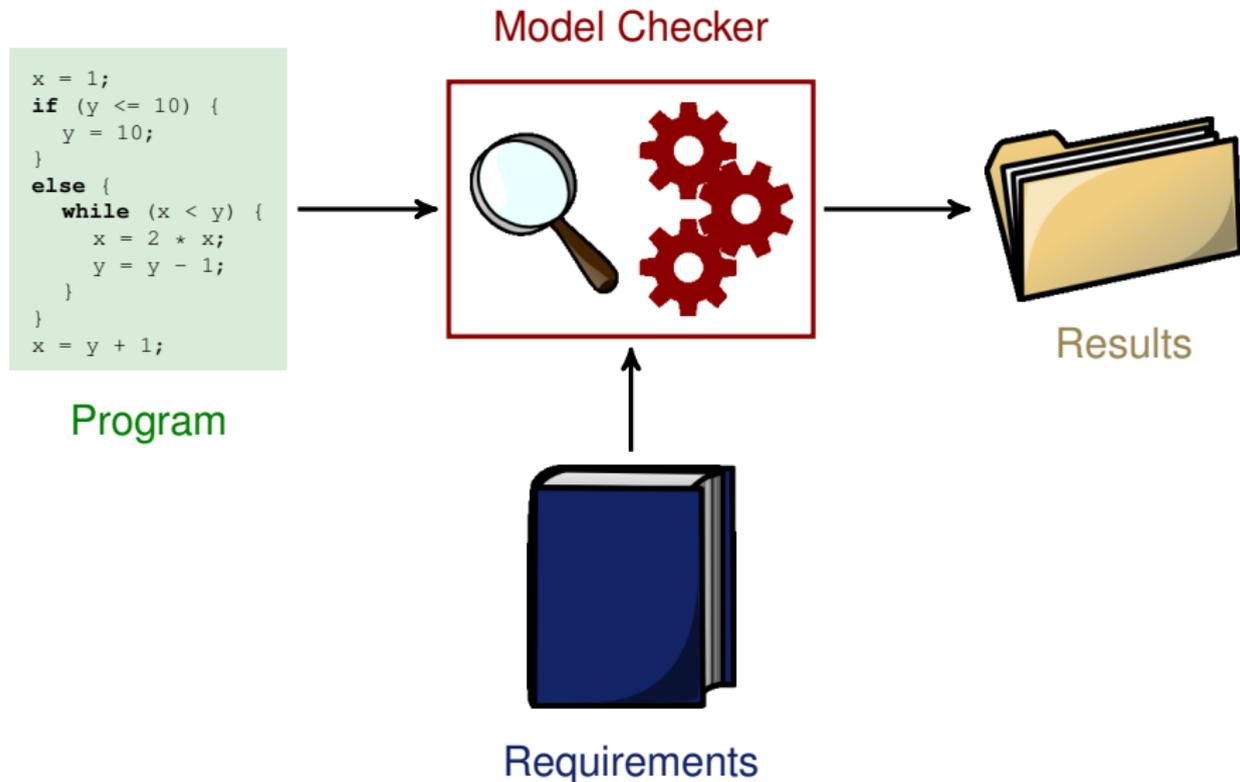
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# Dream of Software Model-Checking



## Rice's Theorem

Any non-trivial **semantic** property of programs is undecidable.

## Classical Example: Termination

There exists no algorithm which can solve the **halting problem**:

- given a description of a program as input,
- decide whether the program terminates or loops forever.

# Practical Limit: Combinatorial Explosion

Implicit in Rice's Theorem is an idealized program model, where programs have access to *unbounded memory*.

In reality programs are run on a computer with *bounded memory*.

Model-checking becomes decidable for finite-state systems.

But even with bounded memory, **complexity** in practice is **too high** for finite-state model-checking:

- 1 megabyte (1 000 000 bytes) of memory  $\approx 10^{2400000}$  states
- 1000 variables  $\times$  64 bits  $\approx 10^{19200}$  states
- optimistic limit for finite-state model checkers:  $10^{100}$  states

# More Realistic Objectives for Software Verification

## Incomplete Methods

### Approximate Algorithms

- 😊 Always terminate
- 😞 Indefinite answer (yes / no / ?)

### Exact Semi-Algorithms

- 😊 Definite answer (yes / no)
- 😞 May not terminate

## Topics of the lecture

Static Analysis

Abstraction Refinement

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## Tentative Definition

Compile-time techniques to gather run-time information about programs without actually running them

## Example

Detection of variables that are used before initialization

- ☺ Always terminates
- ☺ Applies to large programs
- ☹ Simple analyses (original goal was compilation)
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## In the Lecture

Data Flow Analysis

Abstract Interpretation

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Analysis-time techniques to verify programs by model-checking and refinement of finite-state approximate models

## Example

Verification of safety and fairness of a mutual exclusion algorithm

- ☺ Complex analyses (properties expressed in temporal logics)
- ☺ Definite answer (yes / no)
- ☹ May not terminate
- ☹ Modeling of the program into a finite-state transition system

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Abstract Model Refinement for Safety Properties

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Abstract Model Refinement for Safety Properties

# Common Ingredient: Property-Preserving Abstraction

## Abstraction Process

Interpret programs according to a simplified, “abstract” semantics.

## Property-Preserving Abstraction

Formally relate the “abstract” semantics with the “standard” semantics, so as to preserve relevant properties.

## Preservation of Properties

Program interpretation with this abstract semantics therefore gives “correct” information about properties of real runs.

# Abstract Interpretation Example: Sign Analysis

## Objective of Sign Analysis

Discover for each program point the sign of possible run-time values that numerical variables can have at that point.

The abstract semantics “tracks” the following information, for each variable  $x$ :

$$x < 0$$

$$x \leq 0$$

$$x = 0$$

$$x \geq 0$$

$$x > 0$$

# Abstract Interpretation Example: Sign Analysis

```
1 x = 1;
2 if (y ≤ 10) {
3     y = 10;
4 }
5 else {
6     while (x < y) {
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# Credits: Pioneers (1970's)

## Iterative Data Flow Analysis

Gary Kildall

John Kam & Jeffrey Ullman

Michael Karr

...

## Abstract Interpretation

Patrick Cousot & Radhia Cousot

Nicolas Halbwachs

...

And many, many more...

Apologies!

# Outline of the Lecture

- Control Flow Automata
- Data Flow Analysis
- Abstract Interpretation
- Abstract Model Refinement

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➤ Control Flow Automata

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## Part II

# Control Flow Automata

- 3 Syntax and Semantics
- 4 Verification of Control Flow Automata

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4 Verification of Control Flow Automata

# Short Introduction to Control Flow Automata

Requirement for verification: formal **semantics** of programs

## Formal Semantics

Formalization as a mathematical model of the meaning of programs

Denotational

Operational

Axiomatic

## Operational Semantics

Labeled transition system describing the possible computational steps

## First Step Towards an Operational Semantics

Program text  $\longrightarrow$  Graph-based representation

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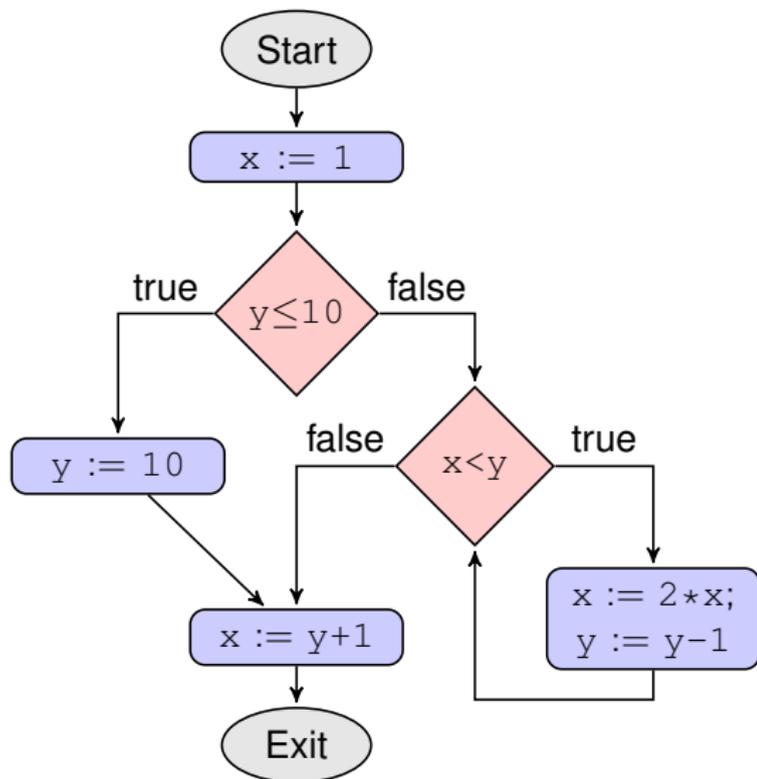
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Program text  $\longrightarrow$  Graph-based representation  
**Control flow automaton**

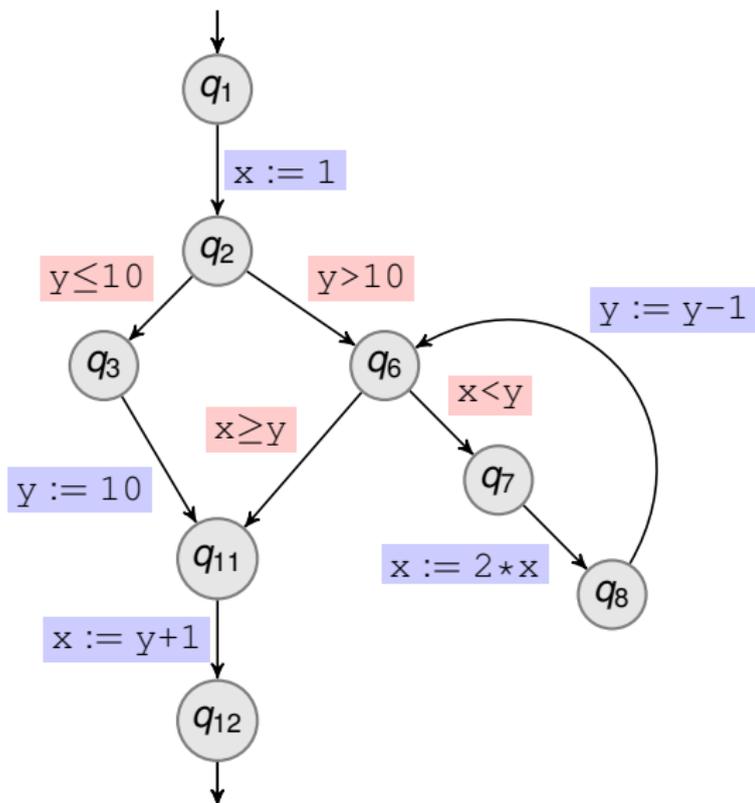
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## Definition

A **labeled directed graph** is a triple  $G = \langle V, \Sigma, \rightarrow \rangle$  where:

- $V$  is a finite set of *vertices*,
- $\Sigma$  is a finite set of *labels*,
- $\rightarrow \subseteq V \times \Sigma \times V$  is a finite set of *edges*.

Notation for edges:  $v \xrightarrow{\sigma} v'$  instead of  $(v, \sigma, v') \in \rightarrow$

A **path** in  $G$  is a finite sequence  $v_0 \xrightarrow{\sigma_0} v'_0, \dots, v_k \xrightarrow{\sigma_k} v'_k$  of edges such that  $v'_i = v_{i+1}$  for each  $0 \leq i < k$ .

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- $q_{in} \in Q$  is an *initial location* and  $q_{out} \in Q$  is an *exit location*,
- $X$  is a finite set of *variables*,
- $\rightarrow \subseteq Q \times \text{Op} \times Q$  is a finite set of *transitions*.

$\text{Op}$  is the set of operations defined by:

$$\begin{aligned} cst &::= c \in Q \\ var &::= x \in X \\ expr &::= cst \mid var \mid expr \bullet expr, \text{ with } \bullet \in \{+, -, *\} \\ guard &::= expr \blacktriangleleft expr, \text{ with } \blacktriangleleft \in \{<, \leq, =, \neq, \geq, >\} \\ \text{Op} &::= guard \mid var := expr \end{aligned}$$

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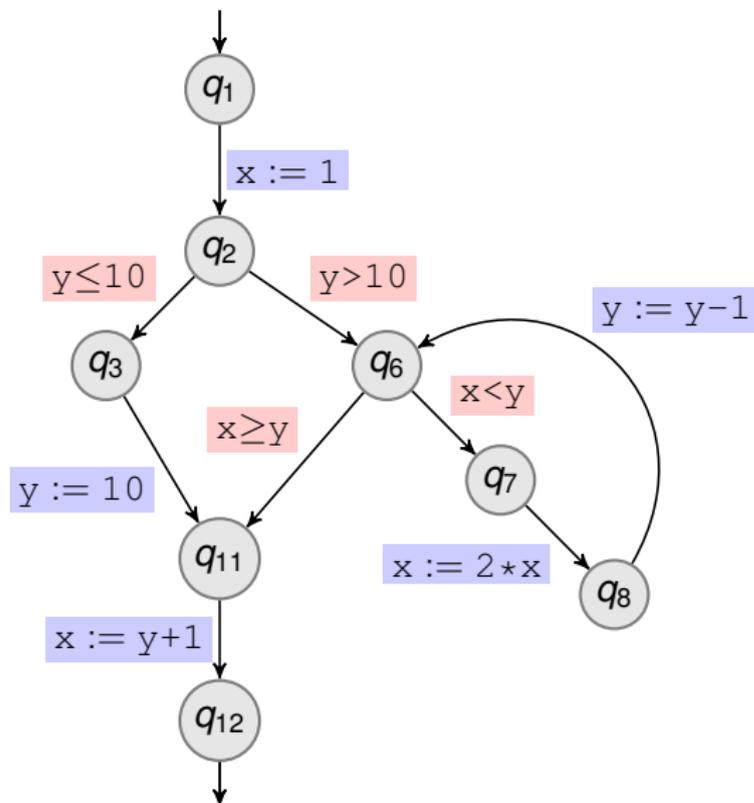
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# Control Flow Automata: Syntax



$$Q = \left\{ \begin{array}{l} q_1, q_2, q_3, q_6, \\ q_7, q_8, q_{11}, q_{12} \end{array} \right\}$$

$$q_{in} = q_1$$

$$q_{out} = q_{12}$$

$$X = \{x, y\}$$

$$\rightarrow = \left\{ \begin{array}{l} (q_1, x := 1, q_2), \\ (q_2, y \leq 10, q_3), \\ (q_2, y > 10, q_6), \\ (q_3, y := 10, q_{11}), \\ \dots \end{array} \right\}$$

# Programs as Control Flow Automata

Control flow automata **can model**:

- ☺ flow of control (program points),
- ☺ numerical variables and numerical operations,
- ☺ non-determinism (uninitialized variables, boolean inputs).

Control flow automata **cannot model**:

- ☹ pointers
- ☹ recursion
- ☹ threads
- ☹ ...

But they are **complex enough** for verification. . . . . and for **learning!**

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Forget about these...

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# Verification of Safety Properties

## Goal

Check that “nothing bad can happen”.

Bad behaviors specified e.g. as assertion violations in the original program

An assertion violation can be modeled as a location:

$$\text{assert}(x > 0) \quad \Longrightarrow \quad \text{if}(x > 0) \text{ then } \{ \text{BAD: } \}$$

## Goal (refined)

Check that there is no “run” that visits a location  $q$  contained in a given set  $Q_{\text{BAD}} \subseteq Q$  of bad locations.

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Check that “nothing bad can happen”.

Bad behaviors specified e.g. as assertion violations in the original program

An assertion violation can be modeled as a location:

$$\text{assert}(x > 0) \quad \Longrightarrow \quad \text{if}(x > 0) \text{ then } \{ \text{BAD: } \}$$

## Goal (refined)

Check that there is no “run” that visits a location  $q$  contained in a given set  $Q_{\text{BAD}} \subseteq Q$  of bad locations.

# Verification of Safety Properties

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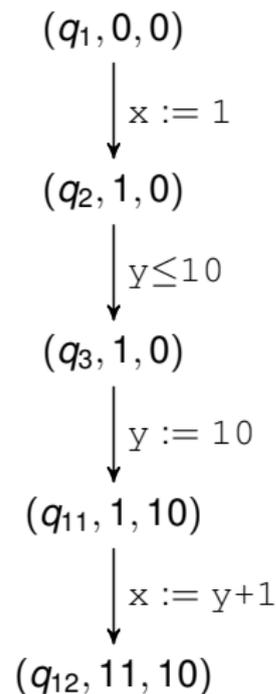
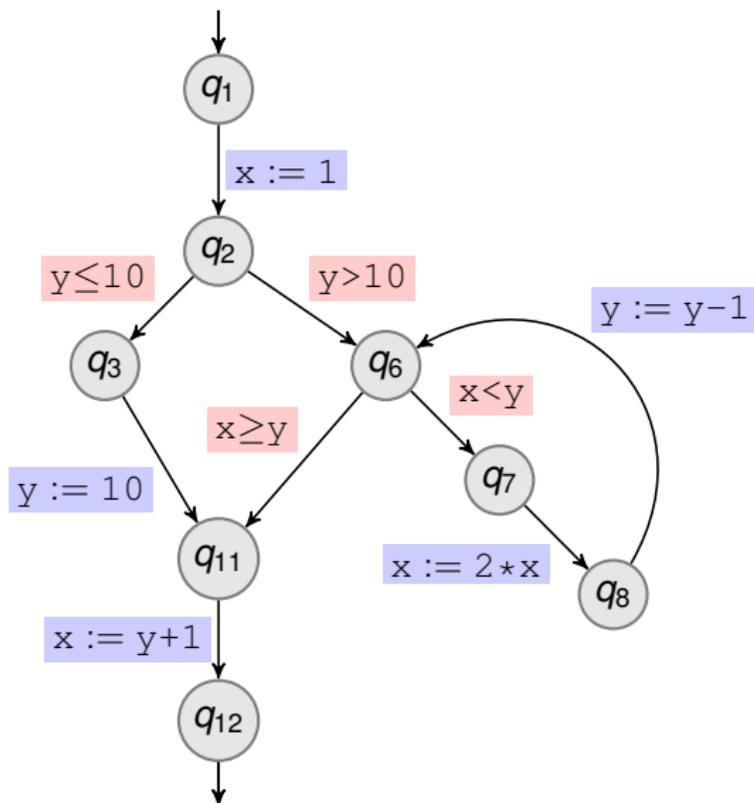
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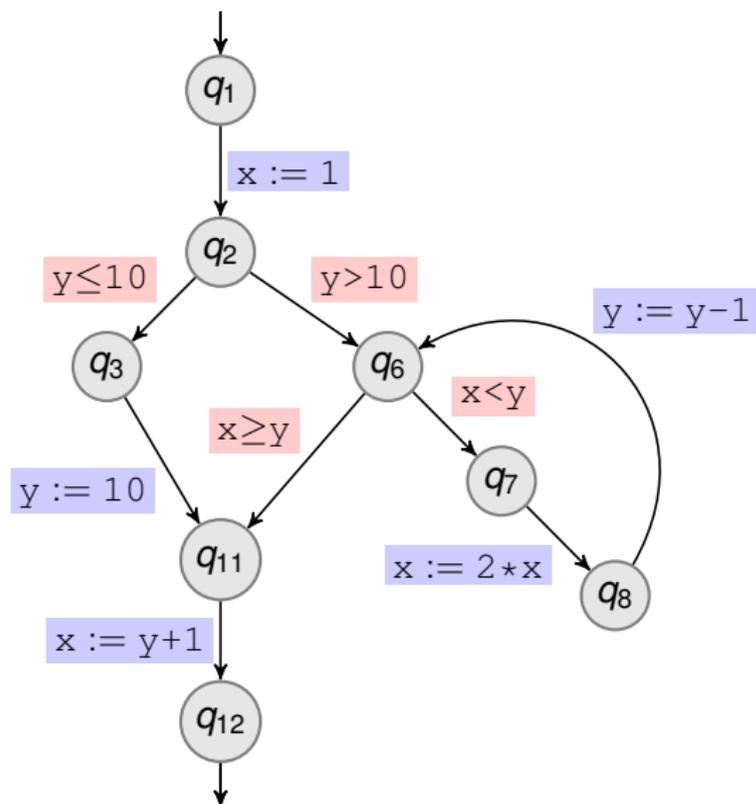
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# Runs: Examples



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$(q_6, 1, 27)$

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## Definition

A **labeled transition system** is a quintuple  $\langle C, Init, Out, \Sigma, \rightarrow \rangle$  where :

- $C$  is a set of *configurations*
- $Init \subseteq C$  and  $Out \subseteq C$  are sets of *initial* and *exit configurations*
- $\Sigma$  is a finite set of *actions*
- $\rightarrow \subseteq C \times \Sigma \times C$  is a set of *transitions*

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# Semantics of Expressions and Guards

Consider a finite set  $X$  of variables. A **valuation** is a function  $v : X \rightarrow \mathbb{R}$ .

Expressions:  $\llbracket e \rrbracket_v$

$$\llbracket c \rrbracket_v = c \quad [c \in \mathbb{Q}]$$

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$$\llbracket e_1 + e_2 \rrbracket_v = \llbracket e_1 \rrbracket_v + \llbracket e_2 \rrbracket_v$$

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$$\llbracket e_1 * e_2 \rrbracket_v = \llbracket e_1 \rrbracket_v \times \llbracket e_2 \rrbracket_v$$

Guards:  $v \models g$

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Examples with  $X = \{x, y\}$

$$\llbracket x * y \leq 10 \rrbracket = \{(v, v) \mid v(x) \times v(y) \leq 10\}$$

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# Operational Semantics of Control Flow Automata

## Definition

The **interpretation** of a control flow automaton  $\langle Q, q_{in}, q_{out}, X, \rightarrow \rangle$  is the labeled transition system  $\langle C, Init, Out, \text{op}, \rightarrow \rangle$  defined by:

- $C = Q \times (X \rightarrow \mathbb{R})$
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Two kinds of labeled directed graphs

### Control Flow Automata

Use: program source codes

- Syntactic objects
- Finite

### Interpretations (LTS)

Use: program behaviors

- Semantic objects
- Uncountably **infinite**

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# Control Paths, Execution Paths and Runs

A **control path** is a path in the control flow automaton:

$$q_0 \xrightarrow{\text{op}_0} q_1 \cdots q_{k-1} \xrightarrow{\text{op}_{k-1}} q_k$$

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A **run** is an execution path that starts with an initial configuration:

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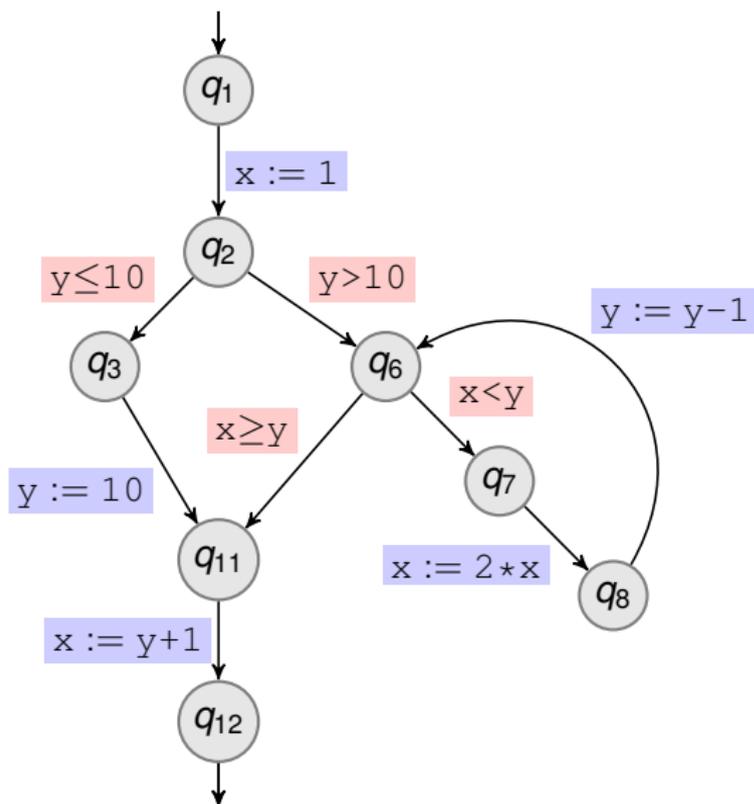
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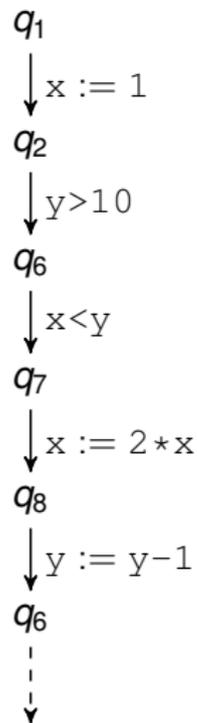
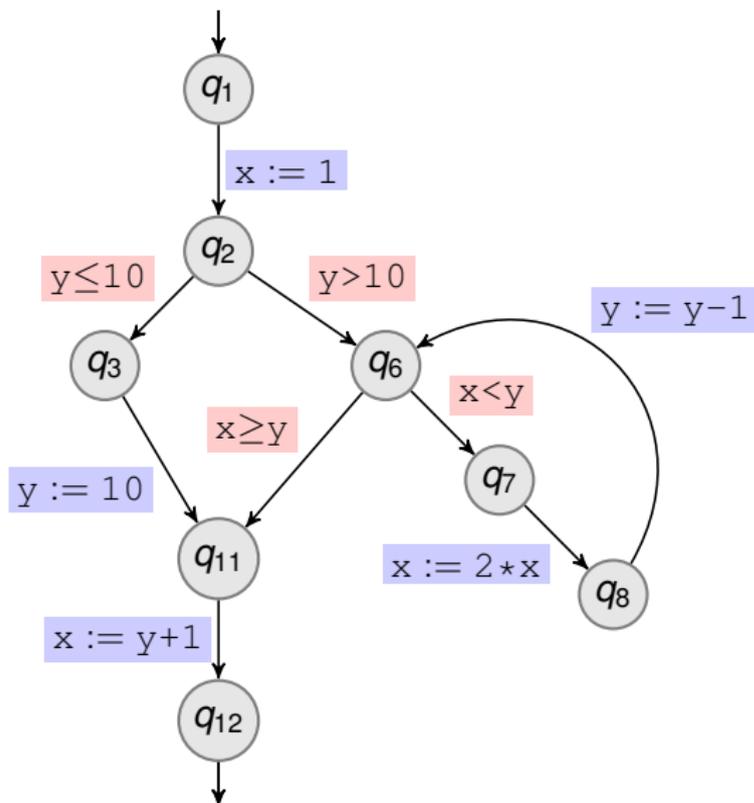
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# Execution Path: Example



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# Control Path: Example



3 Syntax and Semantics

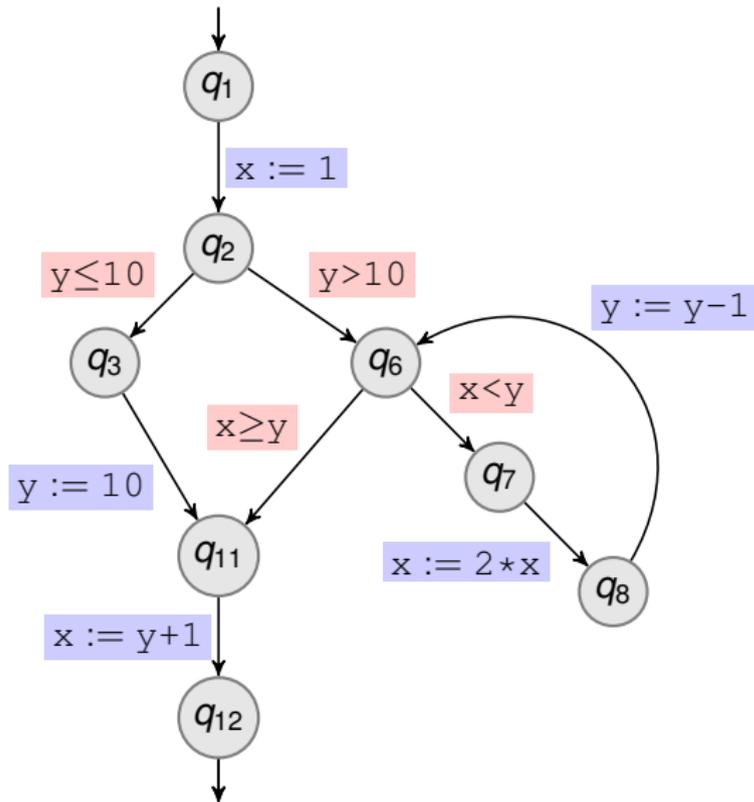
**4 Verification of Control Flow Automata**

# Forward Reachability Set Post\*

Set of all configurations that are reachable from an initial configuration

$$\begin{aligned}\text{Post}^* &= \bigcup_{\rho: \text{run}} \{(q, v) \mid (q, v) \text{ occurs on } \rho\} \\ &= \bigcup_{i \in \mathbb{N}} \text{Post}^i(\text{Init}) \\ &= \bigcup_{q_{in} \xrightarrow{\text{op}_0} \dots \xrightarrow{\text{op}_{k-1}} q} \{q\} \times ([\text{op}_{k-1}] \circ \dots \circ [\text{op}_0])[(X \rightarrow \mathbb{R})]\end{aligned}$$

# Forward Reachability Set Post\* on Running Example



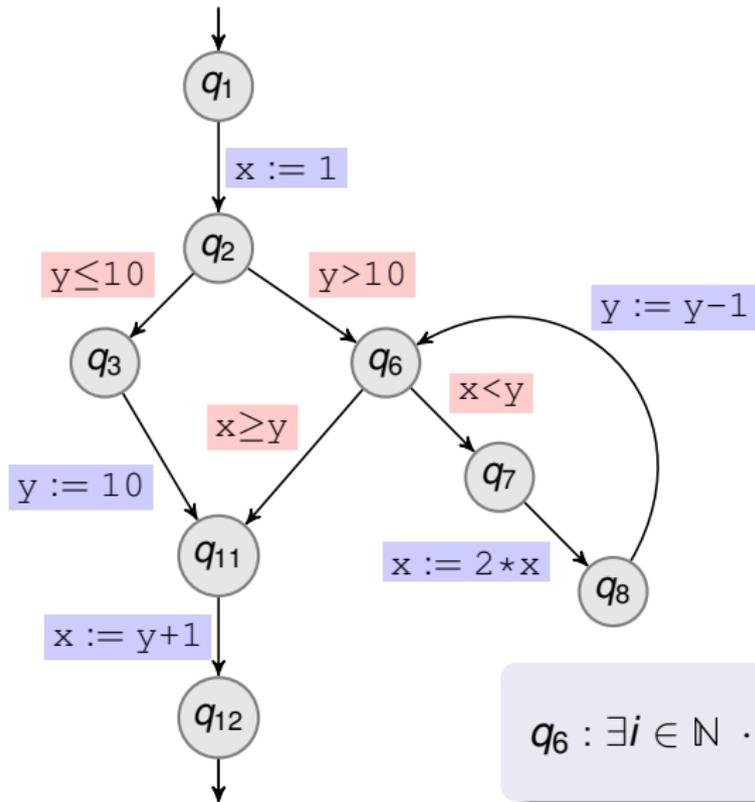
$q_1 : \mathbb{R} \times \mathbb{R}$

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$q_3 : \{1\} \times ]-\infty, 10]$

$q_6 : \{1\} \times ]10, +\infty[ \cup$   
 $\{2\} \times ]9, +\infty[ \cup$   
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...

# Forward Reachability Set Post\* on Running Example



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$$q_6 : \{1\} \times ]10, +\infty[ \cup \{2\} \times ]9, +\infty[ \cup \{4\} \times ]8, +\infty[ \cup \dots$$

$$q_6 : \exists i \in \mathbb{N} \cdot \begin{cases} x = 2^i \wedge y + i > 10 \wedge \\ i \geq 1 \implies 2^{i-1} < y + 1 \end{cases}$$

Set of all configurations that can reach an exit configuration

$$\begin{aligned} \text{Pre}^* &= \bigcup_{i \in \mathbb{N}} \text{Pre}^i(\text{Out}) \\ &= \bigcup_{q \xrightarrow{\text{op}_0} \dots \xrightarrow{\text{op}_{k-1}} q_{\text{out}}} \{q\} \times \left( \llbracket \text{op}_0 \rrbracket^{-1} \circ \dots \circ \llbracket \text{op}_{k-1} \rrbracket^{-1} \right) [(X \rightarrow \mathbb{R})] \\ &= \bigcup_{q \xrightarrow{\text{op}_0} \dots \xrightarrow{\text{op}_{k-1}} q_{\text{out}}} \{q\} \times \left( \left( \llbracket \text{op}_{k-1} \rrbracket \circ \dots \circ \llbracket \text{op}_0 \rrbracket \right)^{-1} \right) [(X \rightarrow \mathbb{R})] \end{aligned}$$

# Verification of Control Flow Automata

## Goal (Repetition)

Check that there is no run that visits a location  $q$  contained in a given set  $Q_{BAD} \subseteq Q$  of bad locations.

Define the set *Bad* of bad configurations by:  $Bad = Q_{BAD} \times (\mathbb{X} \rightarrow \mathbb{R})$ .

## Goal (Equivalent Formulation)

Check that  $Post^*$  is disjoint from *Bad*

## Undecidability

The *location reachability* and *configuration reachability* problems are both undecidable for control flow automata.

Proof by reduction to location reachability in two-counters machines.

## Two-Counters (Minsky) Machines

Finite-state automaton extended with:

- two counters over nonnegative integers
- test for zero, increment and guarded decrement

Reachability is undecidable for this class.

Any two-counters machine can (effectively) be represented as a control flow automaton in this **restricted class**:

- two variables:  $X = \{c_1, c_2\}$
- allowed guards:  $x = 0$  and  $x \neq 0$  for each  $x \in X$
- allowed assignments:  $x := x+1$  and  $x := x-1$  for each  $x \in X$

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## Definition

An **invariant** is any set  $Inv \subseteq C$  such that  $Post^* \subseteq Inv$ .

Idea:

- 1 Compute an invariant  $Inv$  (easier to compute than  $Post^*$ )
- 2 If  $Inv$  is disjoint from  $Bad$  then  $Post^*$  is also disjoint from  $Bad$

Rest of the lecture:

Computation of precise enough invariants

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Rest of the lecture:

Computation of precise enough invariants

- Computational model for programs: control flow automata
  - syntax
  - semantics
- Undecidability in general of model-checking for control flow automata
- Tentative solution: computation of invariants

## Part III

# Data Flow Analysis

- 5 Classical Data Flow Analyses
- 6 Basic Lattice Theory
- 7 Monotone Data Flow Analysis Frameworks

5 Classical Data Flow Analyses

6 Basic Lattice Theory

7 Monotone Data Flow Analysis Frameworks

# Short Introduction to Data Flow Analysis

## Tentative Definition

Compile-time techniques to gather **run-time** information about **data** in programs without actually running them

## Applications

### Code **optimization**

- Avoid *redundant* computations (e.g. reuse available results)
- Avoid *superfluous* computations (e.g. eliminate dead code)

### Code **validation**

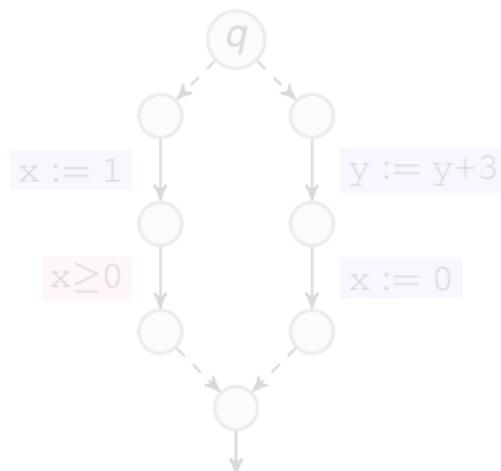
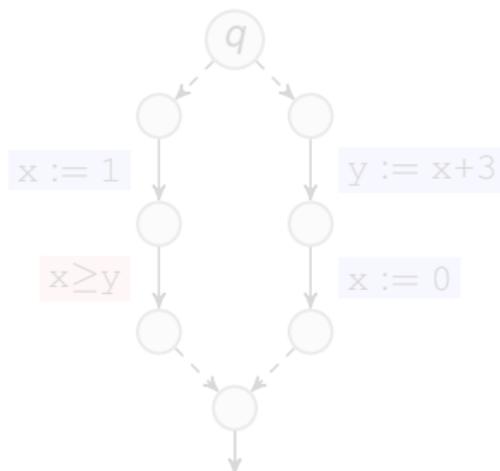
- Invariant generation

**Conservative approximations**

# Live Variables Analysis: Definition

## Definition

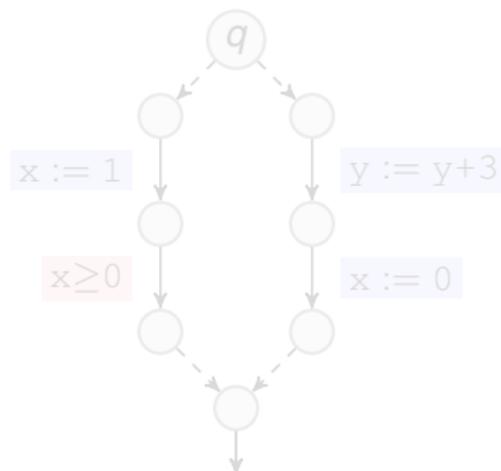
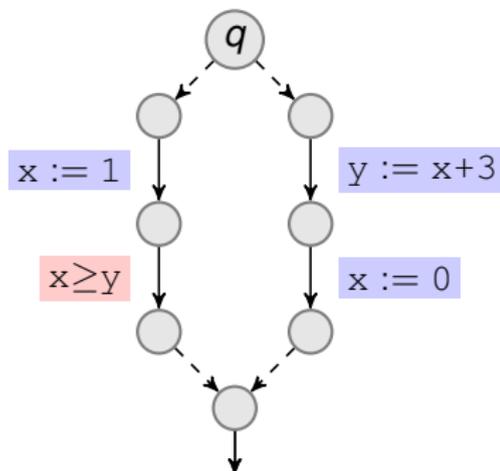
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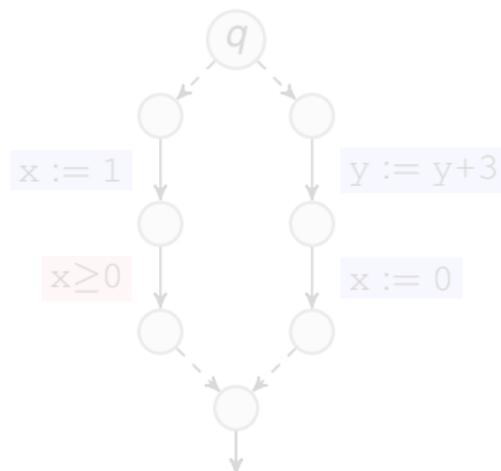
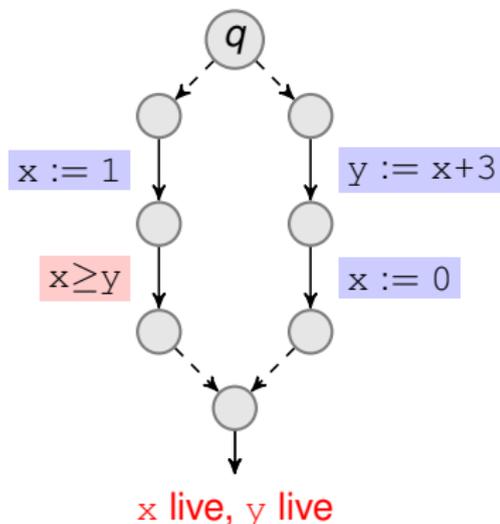
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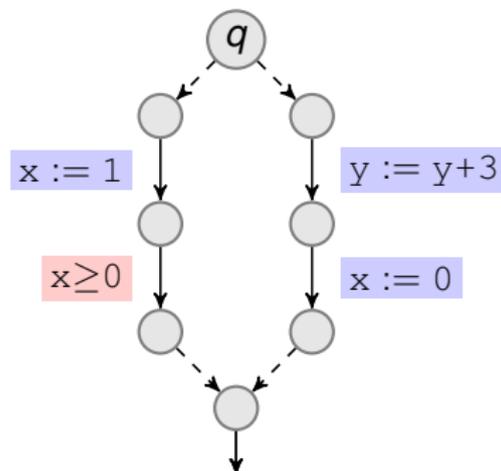
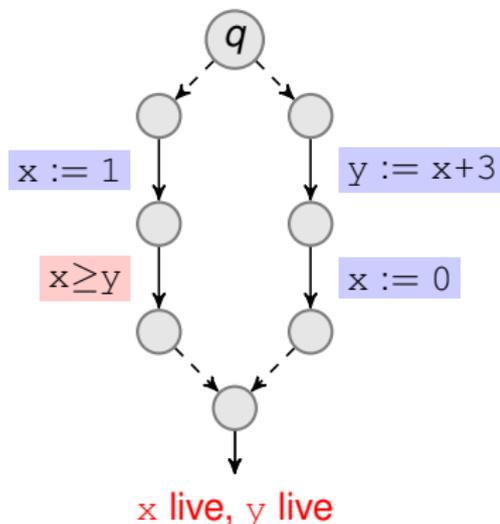
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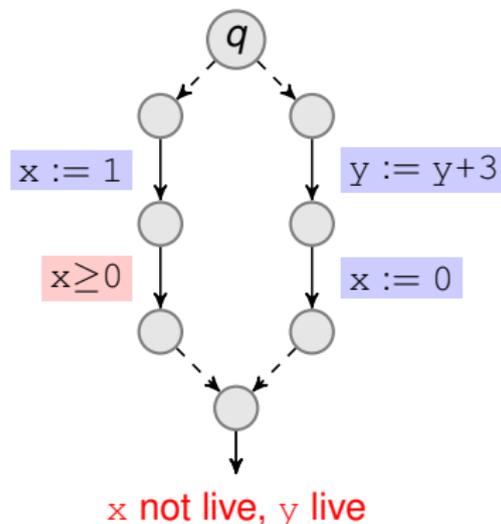
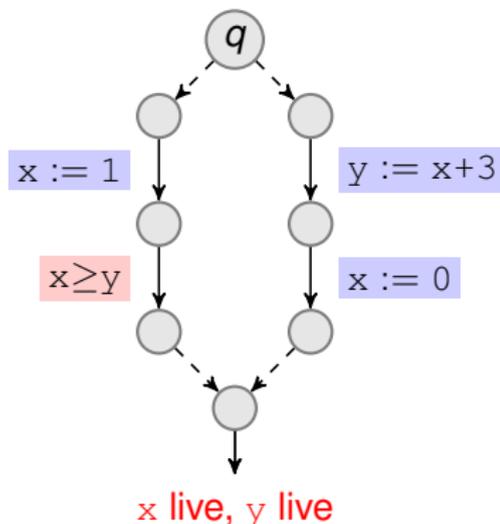
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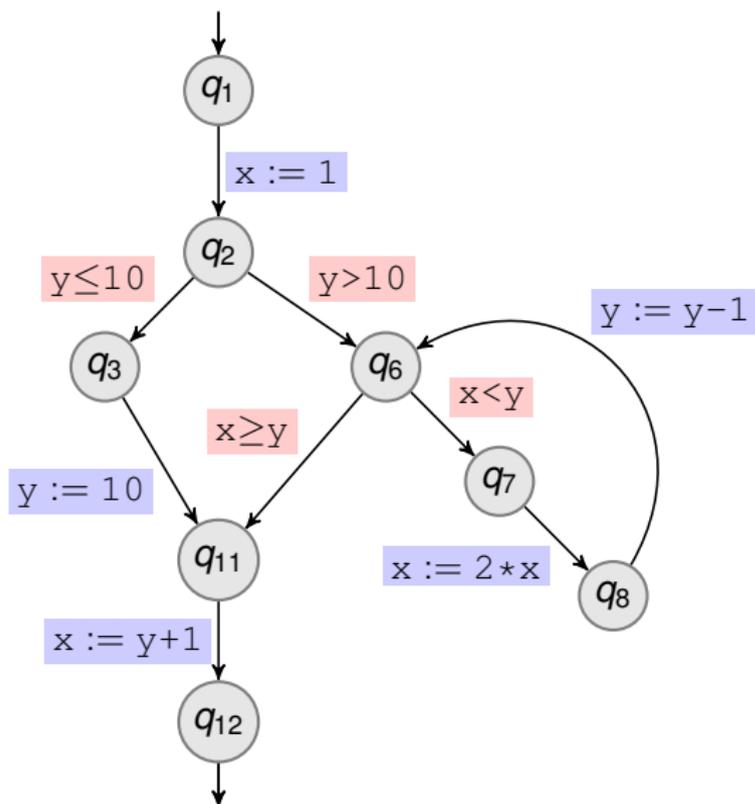
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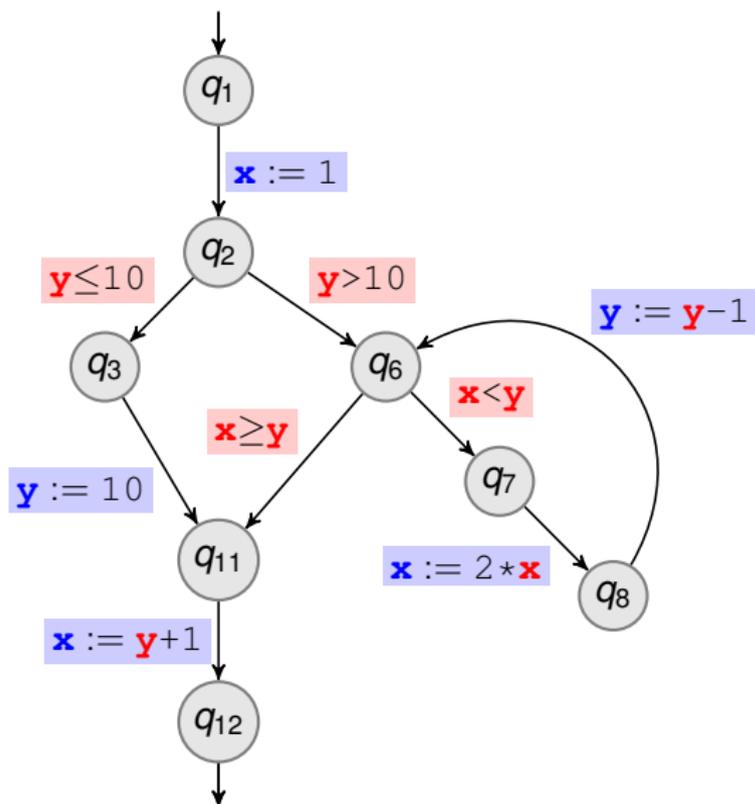
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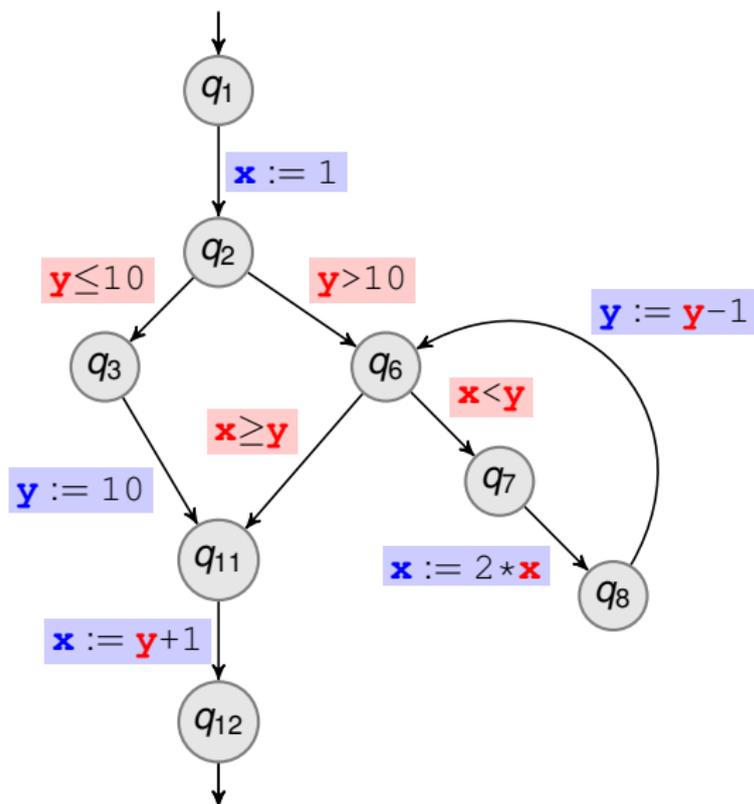
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# Live Variables Analysis: Running Example



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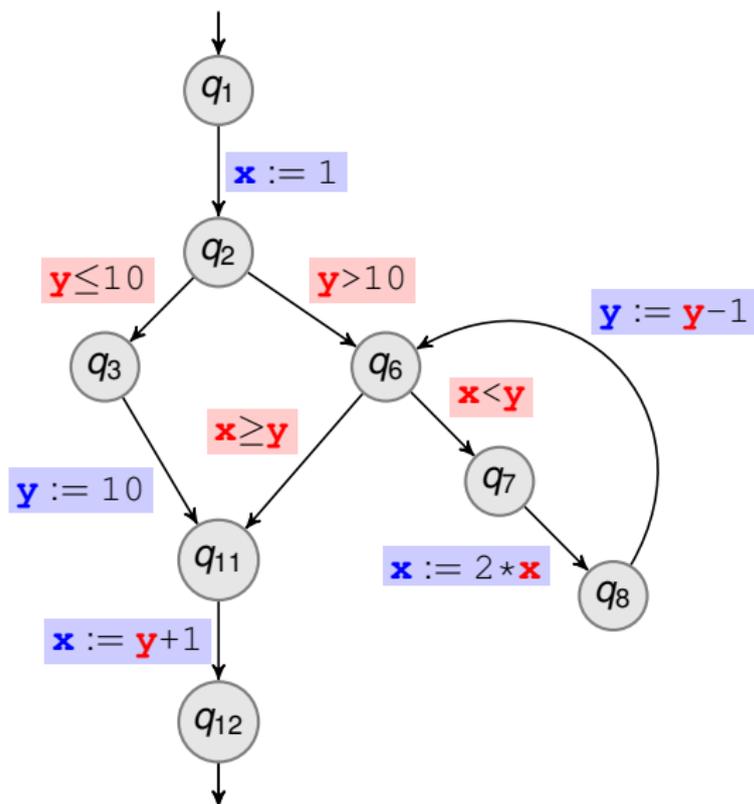
0 : Initialization

1 : Local information

2 : Propagation ( $\leftarrow$ )

	x	y
$q_1$		
$q_2$		
$q_3$		
$q_6$		
$q_7$		
$q_8$		
$q_{11}$		
$q_{12}$		

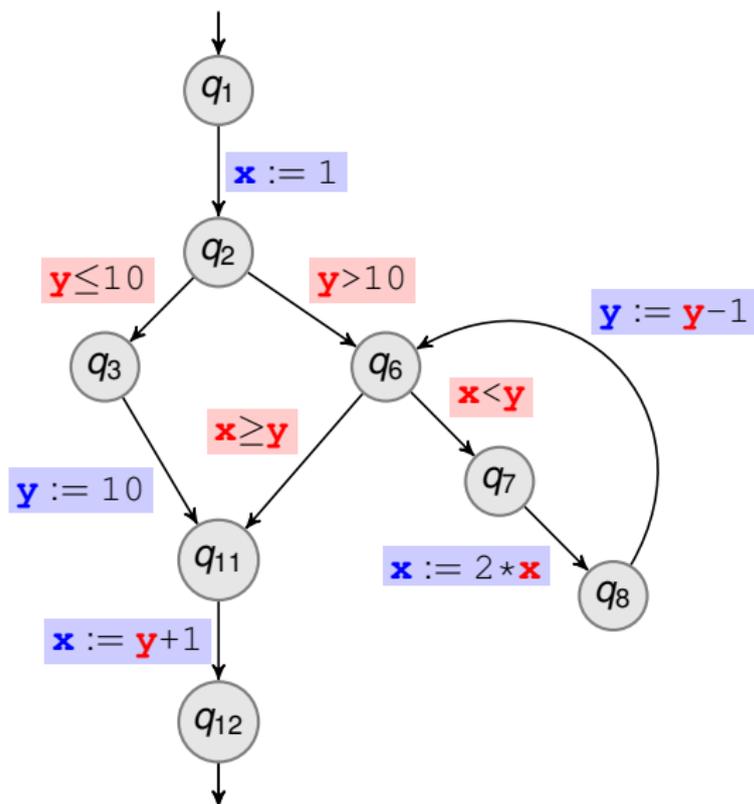
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$q_2$		•
$q_3$		
$q_6$	•	•
$q_7$	•	
$q_8$		•
$q_{11}$		•
$q_{12}$		

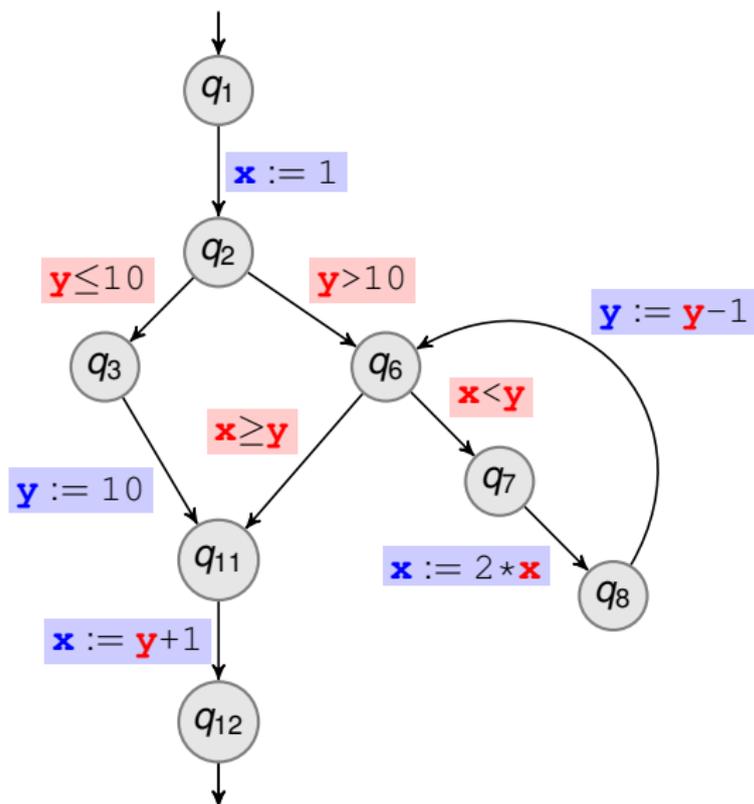
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$q_8$		•
$q_{11}$		•
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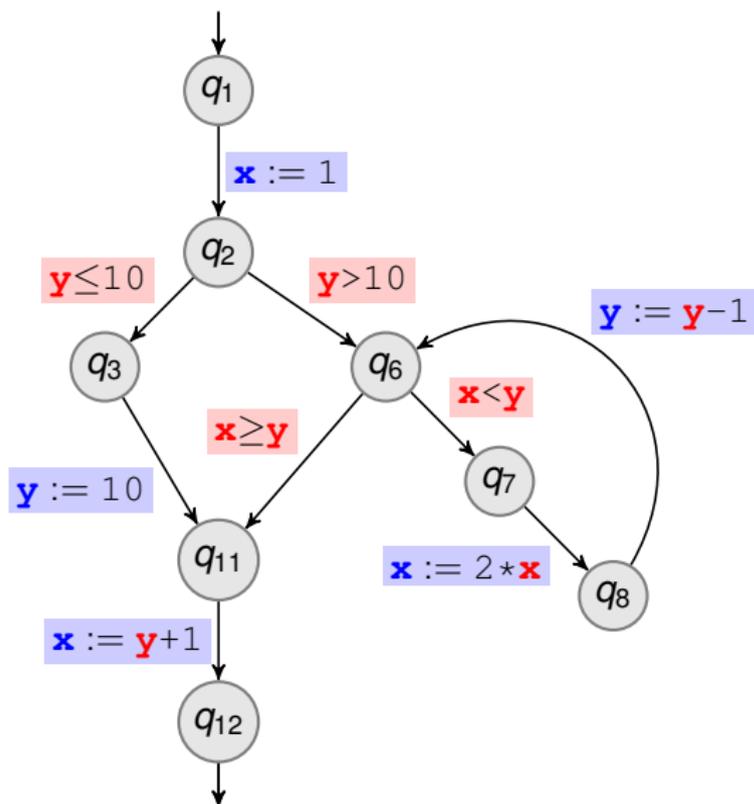
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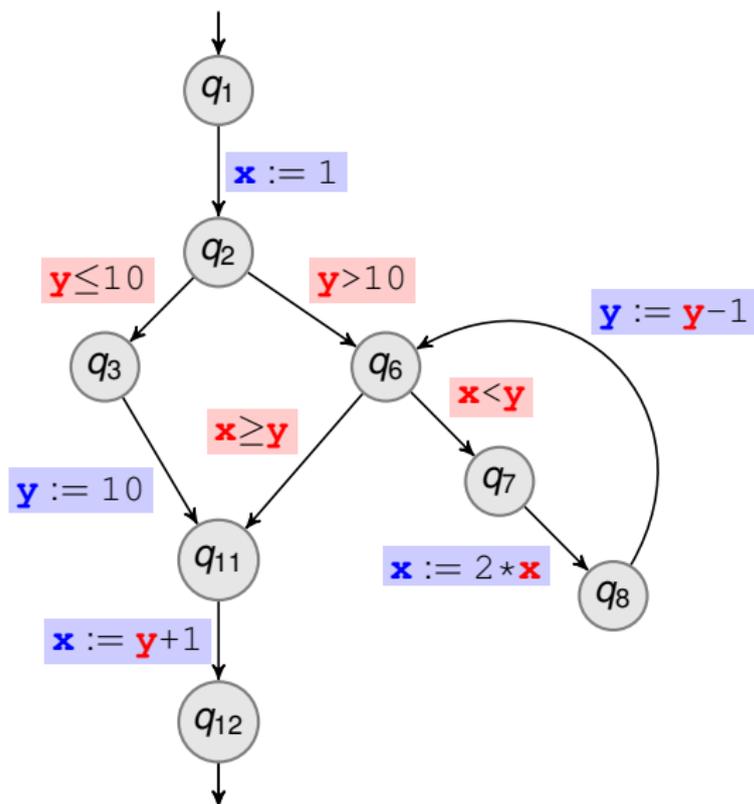
# Live Variables Analysis: Running Example



- 0 : Initialization
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	x	y
$q_1$		●
$q_2$	●	●
$q_3$		
$q_6$	●	●
$q_7$	●	
$q_8$		●
$q_{11}$		●
$q_{12}$		

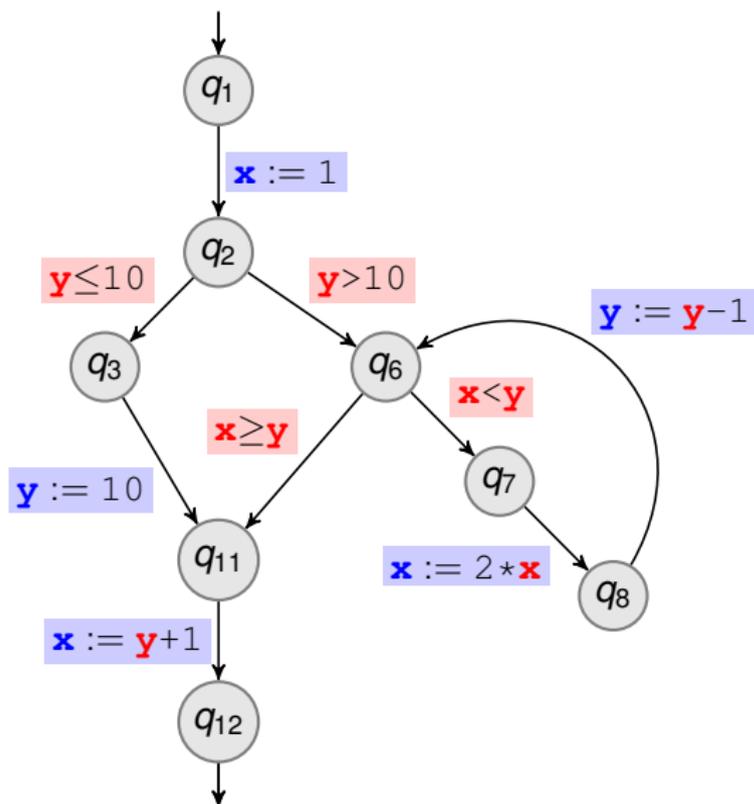
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$q_8$		•
$q_{11}$		•
$q_{12}$		

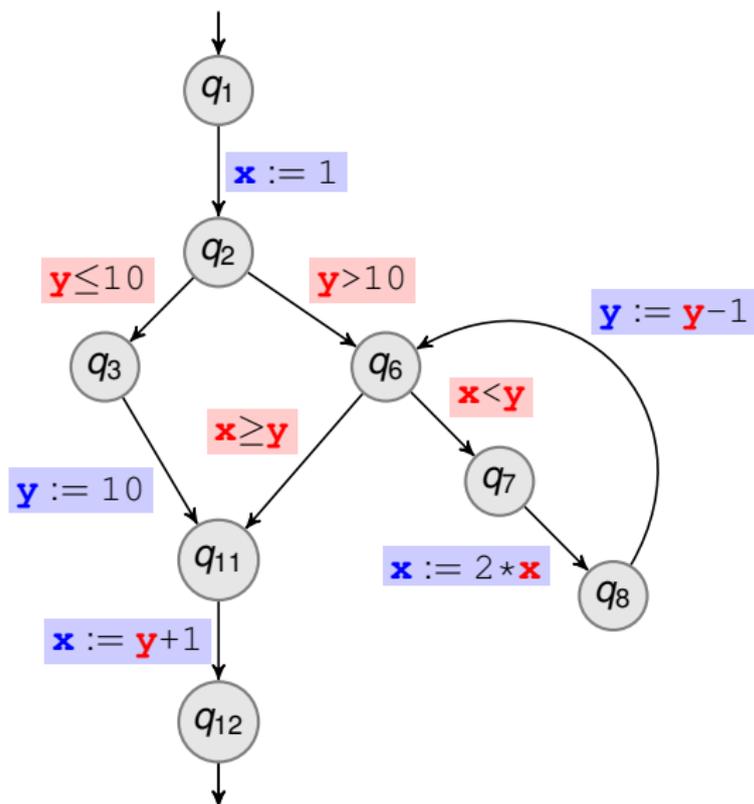
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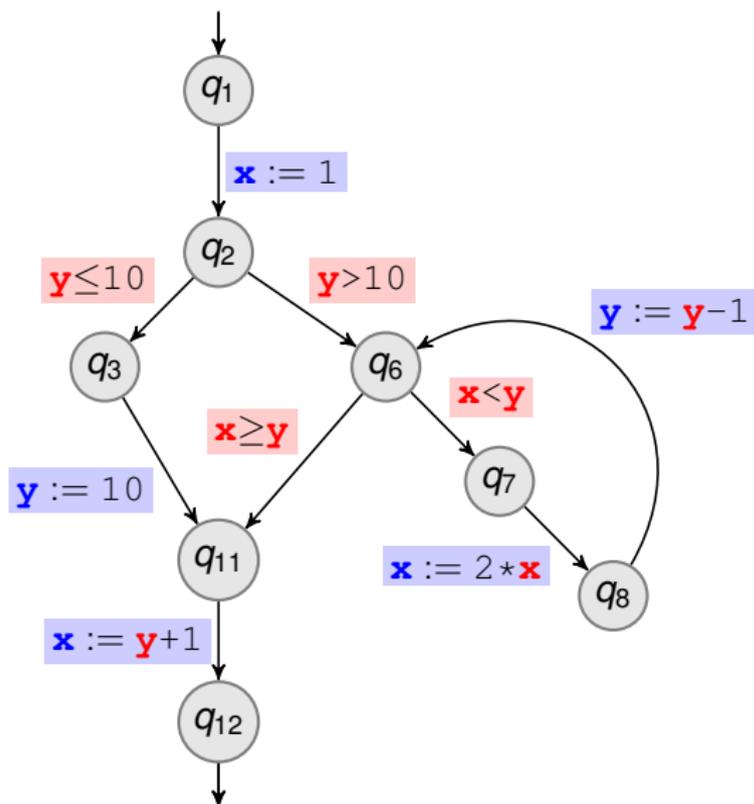
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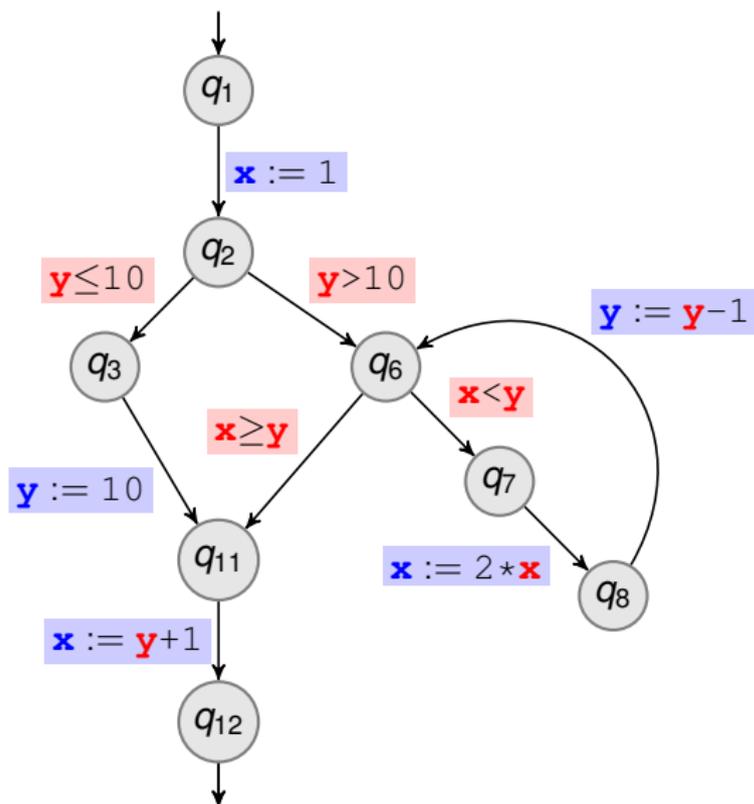
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	x	y
$q_1$		●
$q_2$	●	●
$q_3$		
$q_6$	●	●
$q_7$	●	●
$q_8$	●	●
$q_{11}$		●
$q_{12}$		

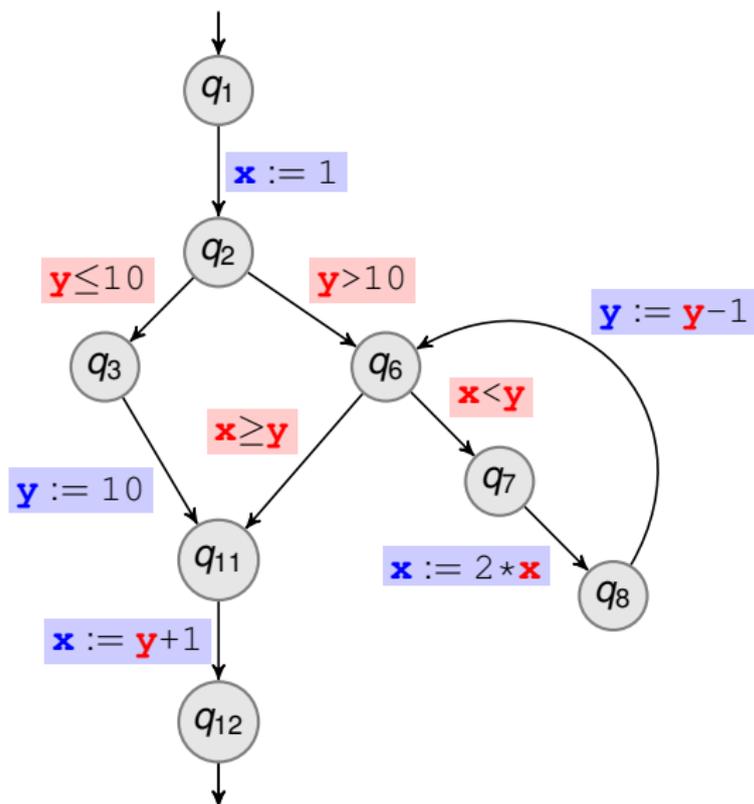
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$q_3$		
$q_6$	•	•
$q_7$	•	•
$q_8$	•	•
$q_{11}$		•
$q_{12}$		

# Live Variables Analysis: Formulation

Control Flow Automaton:  $\langle Q, q_{in}, q_{out}, X, \rightarrow \rangle$

System of equations: variables  $L_q$  for  $q \in Q$ , with  $L_q \subseteq X$

$$L_q = \bigcup_{q \xrightarrow{\text{op}} q'} \text{Gen}_{\text{op}} \cup (L_{q'} \setminus \text{Kill}_{\text{op}}) \qquad L(q_{out}) = \emptyset$$

$$\text{Gen}_{\text{op}} = \begin{cases} \text{Var}(g) & \text{if } \text{op} = g \\ \text{Var}(e) & \text{if } \text{op} = x := e \end{cases} \qquad \text{Kill}_{\text{op}} = \begin{cases} \emptyset & \text{if } \text{op} = g \\ \{x\} & \text{if } \text{op} = x := e \end{cases}$$

$$f_{\text{op}}(X) = \text{Gen}_{\text{op}} \cup (X \setminus \text{Kill}_{\text{op}})$$

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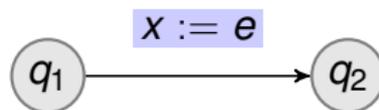
$$L_q = \bigcup_{q \xrightarrow{\text{op}} q'} f_{\text{op}}(L_{q'}) \qquad L(q_{out}) = \emptyset$$

$$Gen_{\text{op}} = \begin{cases} \text{Var}(g) & \text{if } \text{op} = g \\ \text{Var}(e) & \text{if } \text{op} = x := e \end{cases} \qquad Kill_{\text{op}} = \begin{cases} \emptyset & \text{if } \text{op} = g \\ \{x\} & \text{if } \text{op} = x := e \end{cases}$$

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## Code Optimization

### Dead code elimination



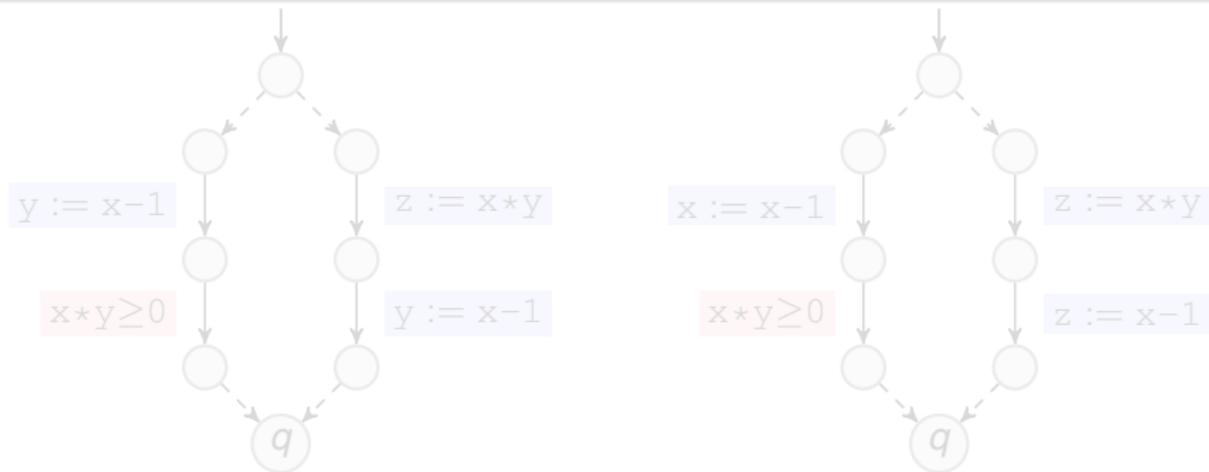
If  $x$  is not live at location  $q_2$  then we may remove the assignment  $x := e$  on the edge from  $q_1$  to  $q_2$ .

**This is sound since the analysis is conservative**

# Available Expressions Analysis: Definition

## Definition

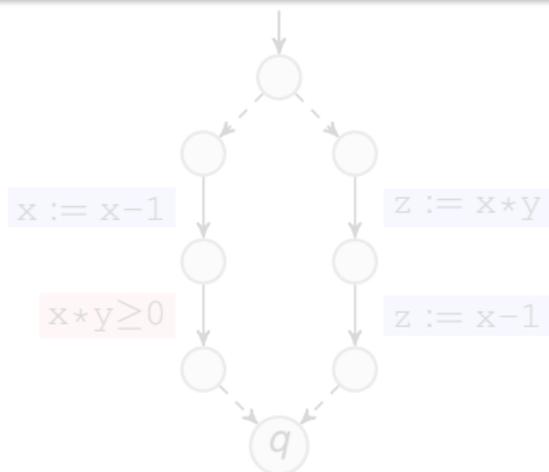
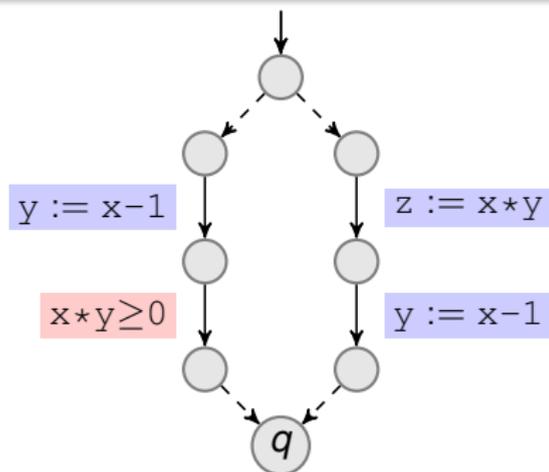
A expression  $e$  is **available** at location  $q$  if **every** control path from  $q_{in}$  to  $q$  contains an evaluation of  $e$  which is not followed by an assignment of any variable  $x$  occurring in  $e$ .



# Available Expressions Analysis: Definition

## Definition

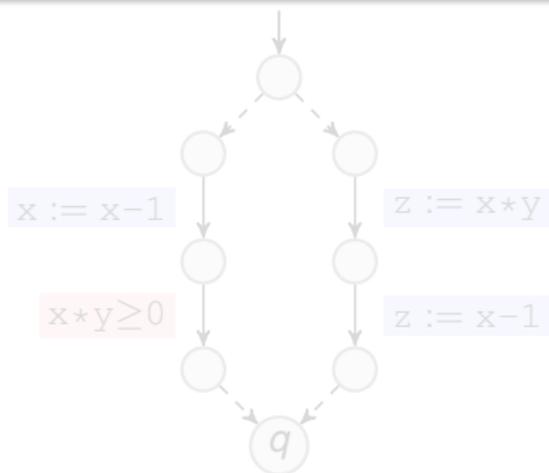
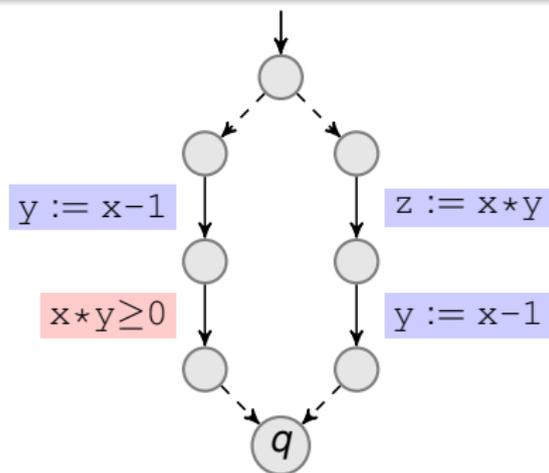
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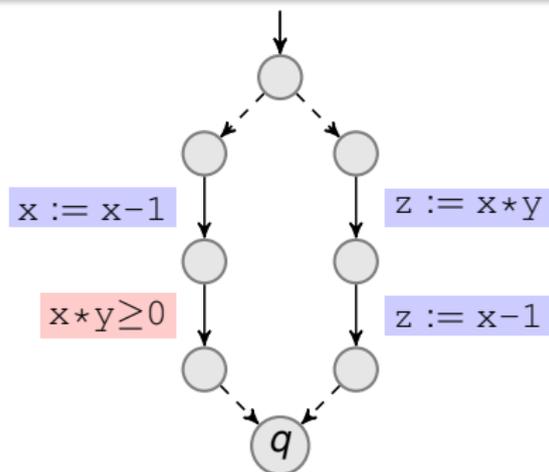
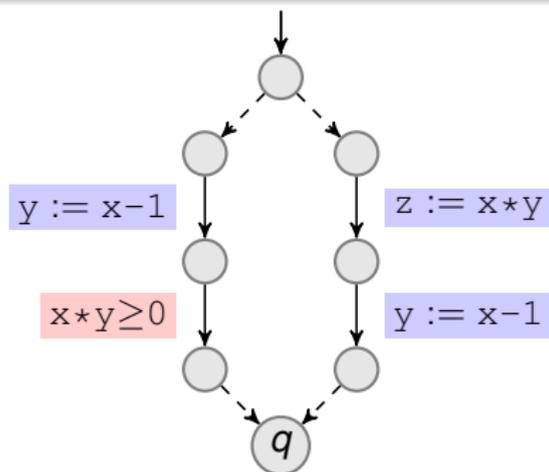


$x-1$  available,  $x*y$  not available

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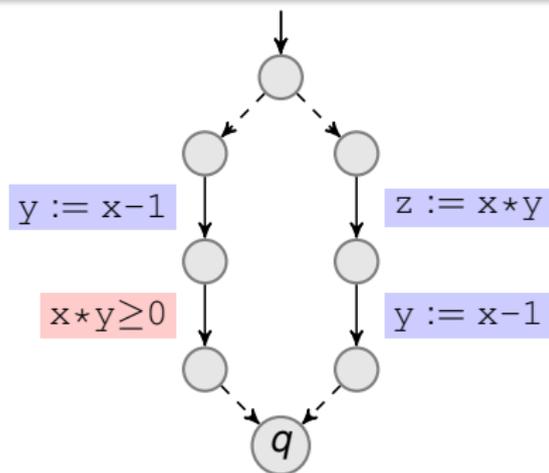


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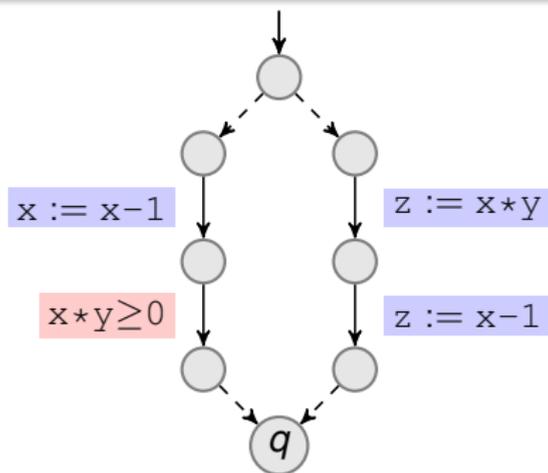
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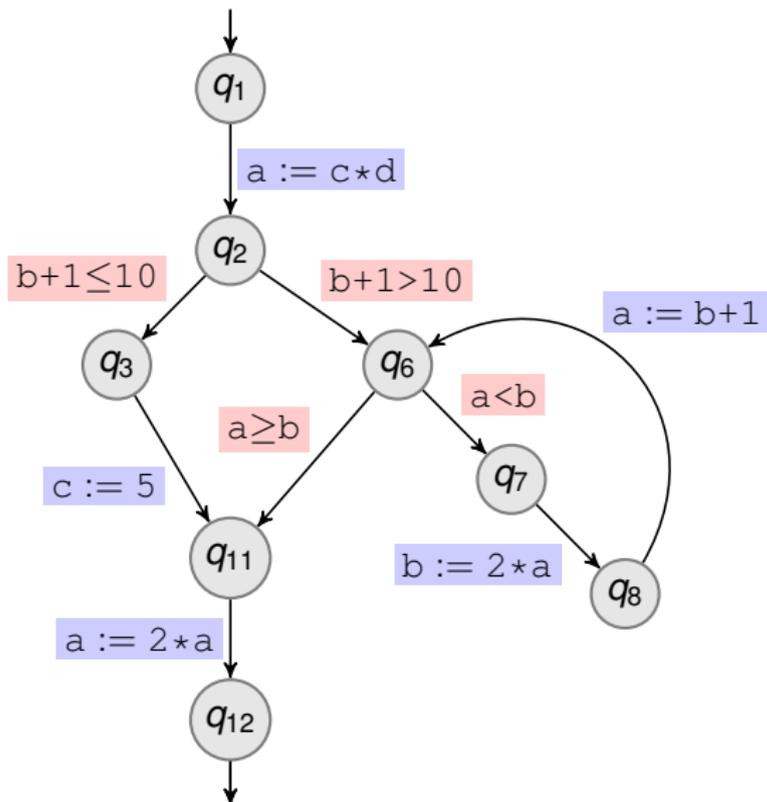


$x-1$  available,  $x*y$  not available

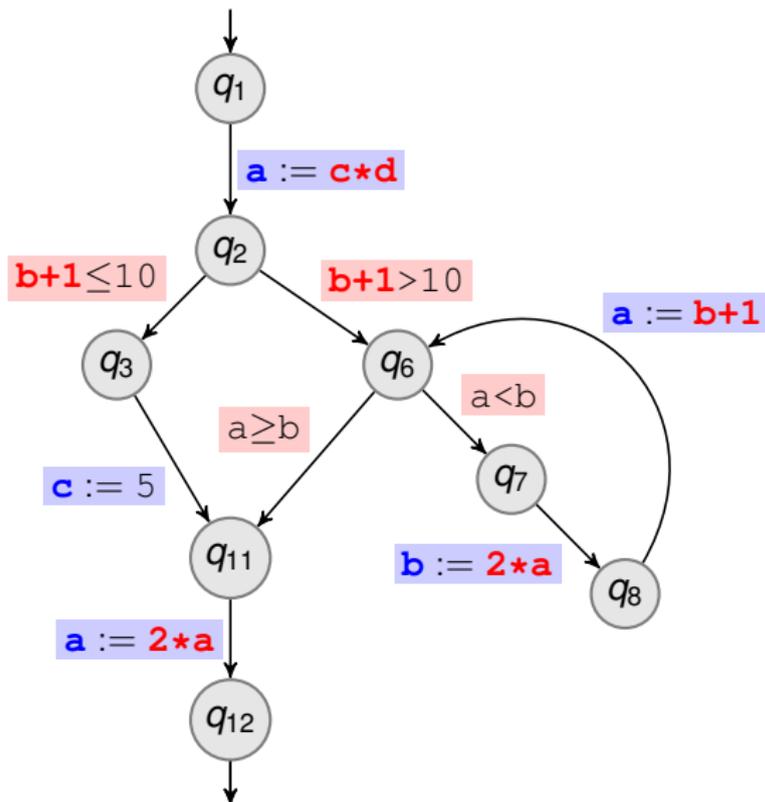


$x-1$  not available,  $x*y$  available

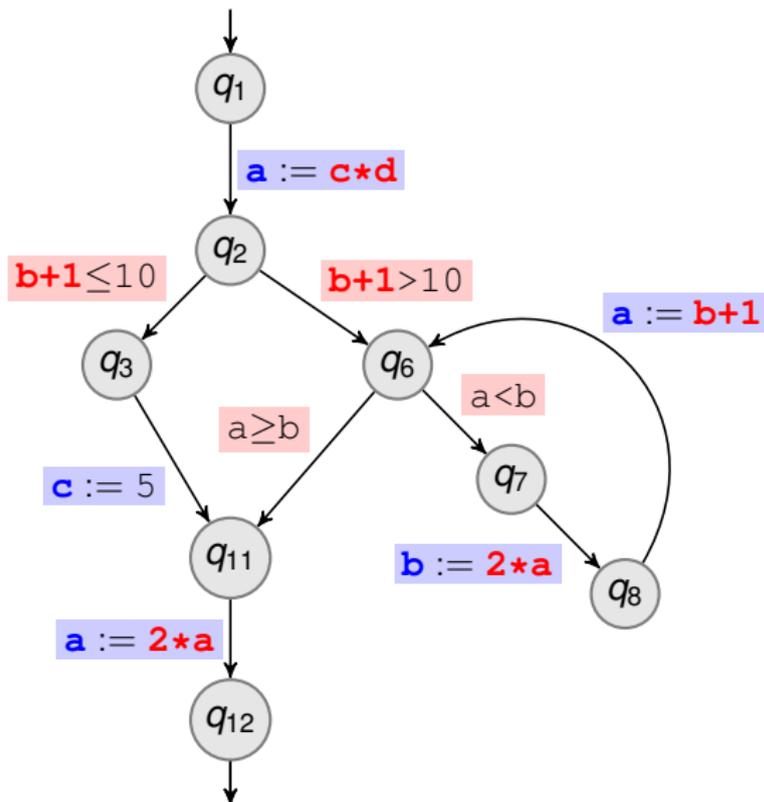
# Available Expressions Analysis: Other Example



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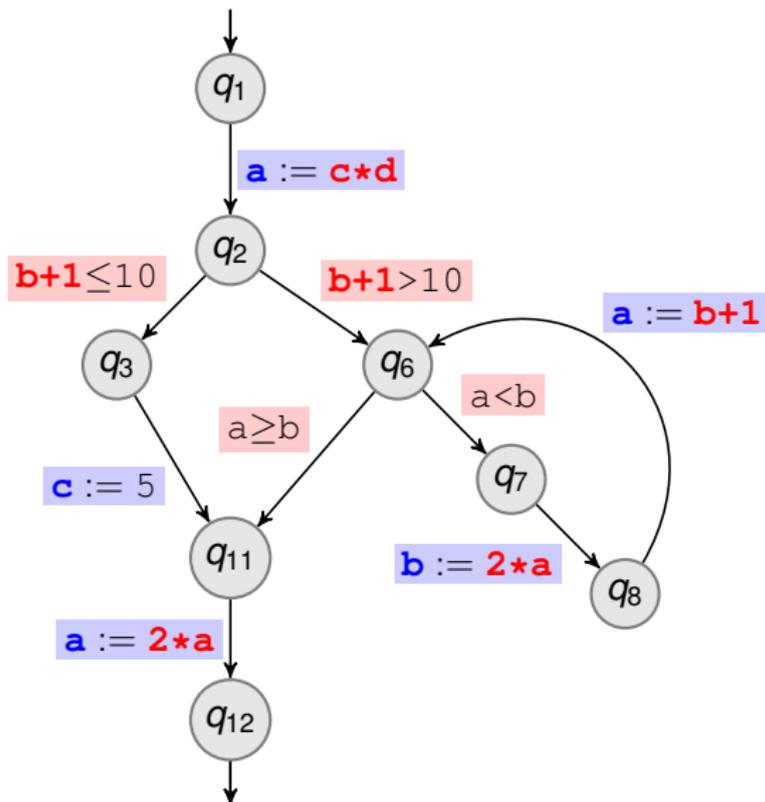
0 : Initialization

1 : Local information

2 : Propagation ( $\rightarrow$ )

	$c * d$	$b + 1$	$2 * a$
$q_1$			
$q_2$	•	•	•
$q_3$	•	•	•
$q_6$	•	•	•
$q_7$	•	•	•
$q_8$	•	•	•
$q_{11}$	•	•	•
$q_{12}$	•	•	•

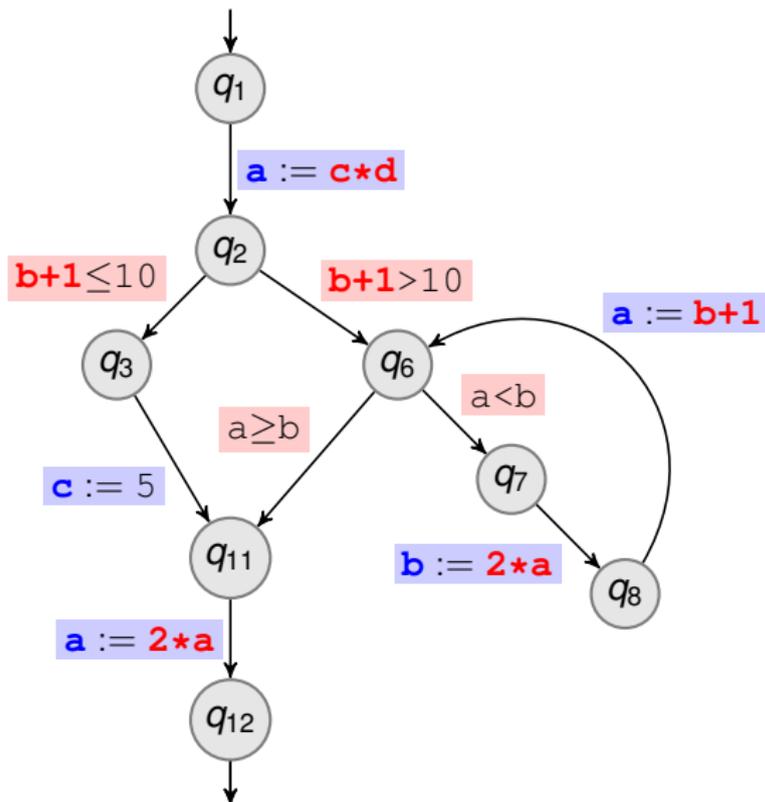
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$q_3$	•	•	•
$q_6$	•	•	
$q_7$	•	•	•
$q_8$	•		•
$q_{11}$		•	•
$q_{12}$	•	•	

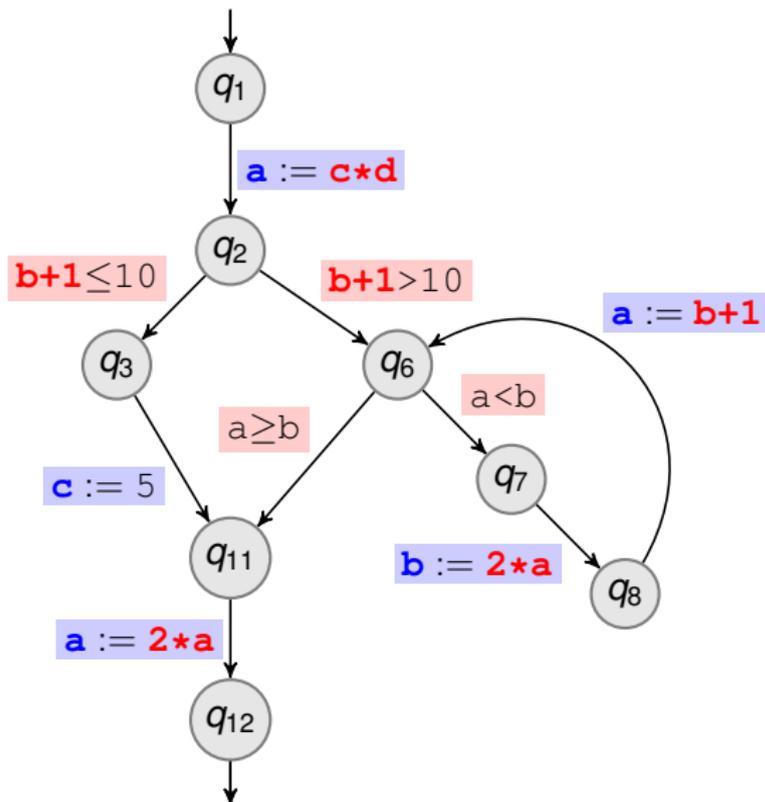
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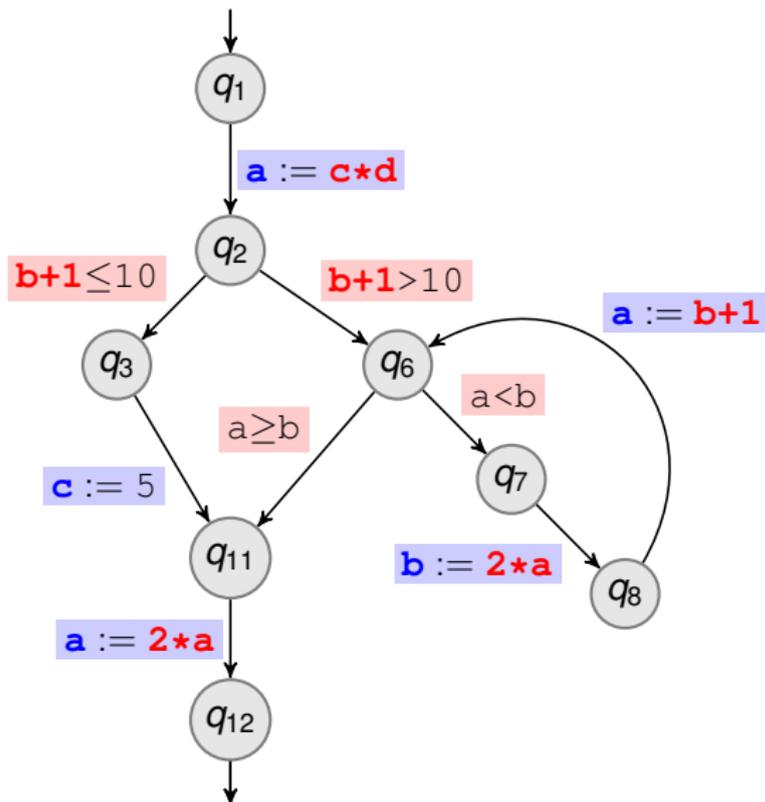
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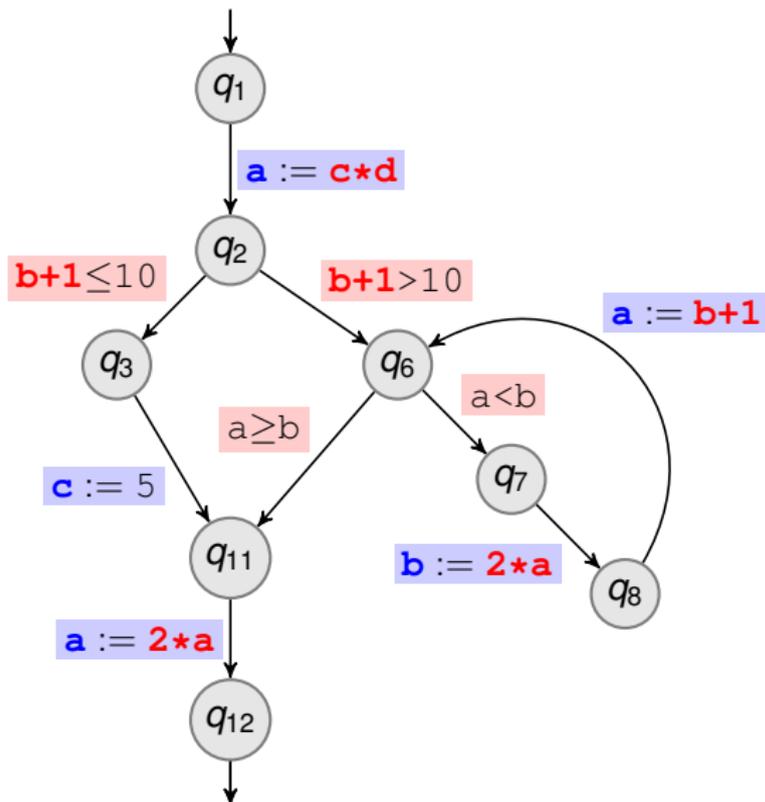
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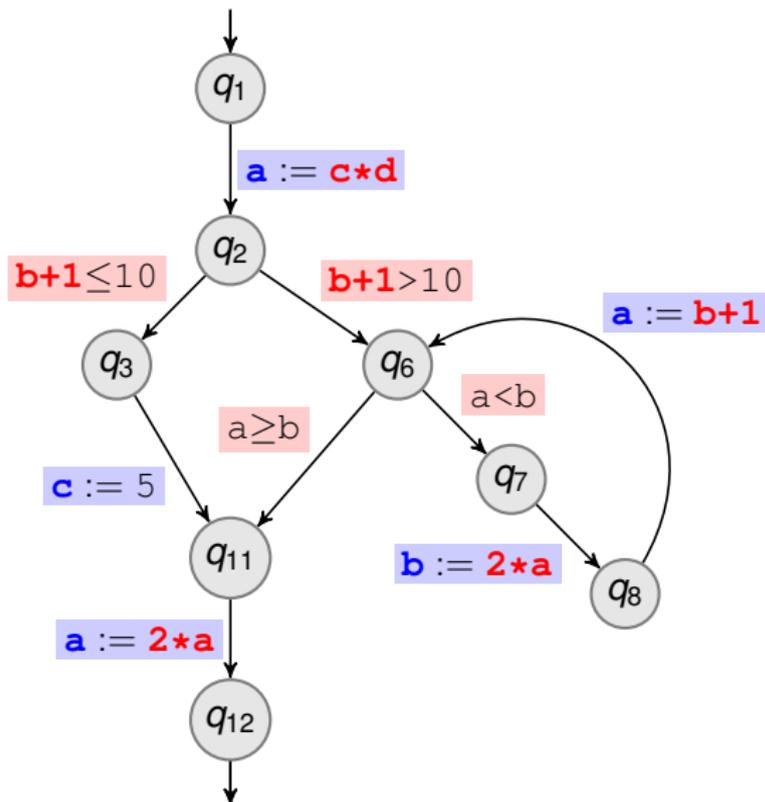
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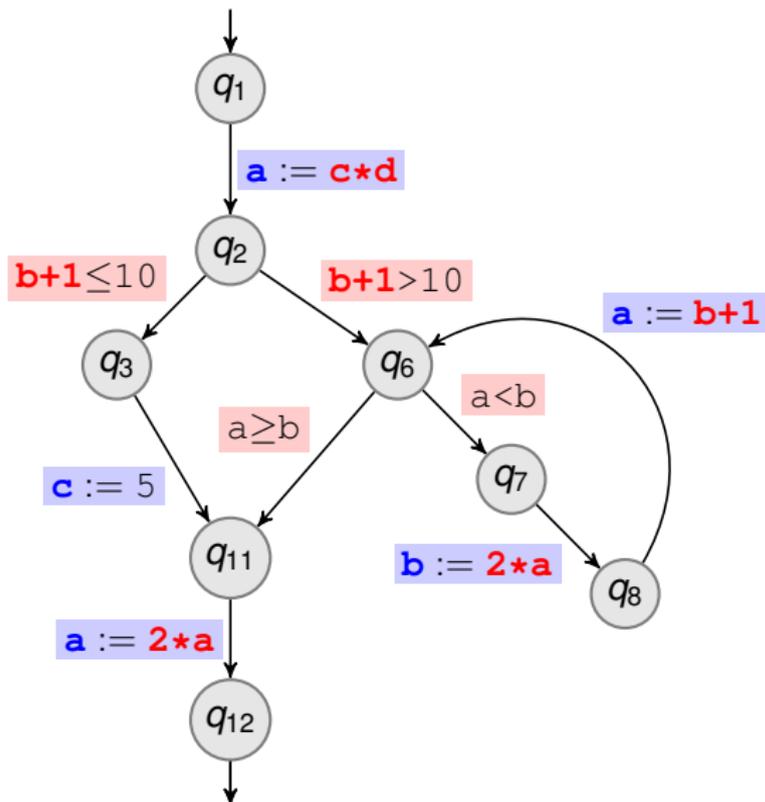
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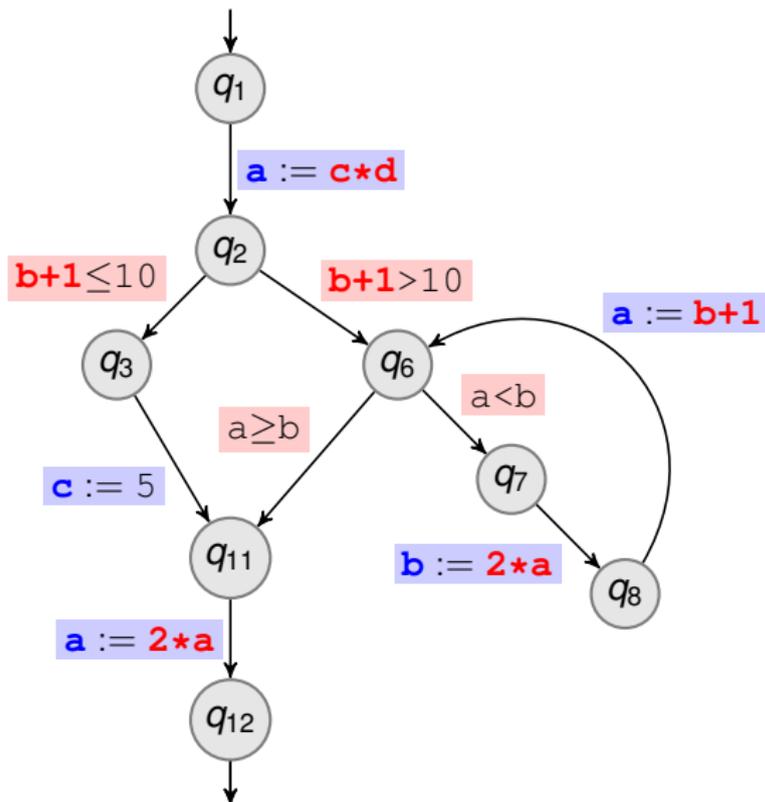
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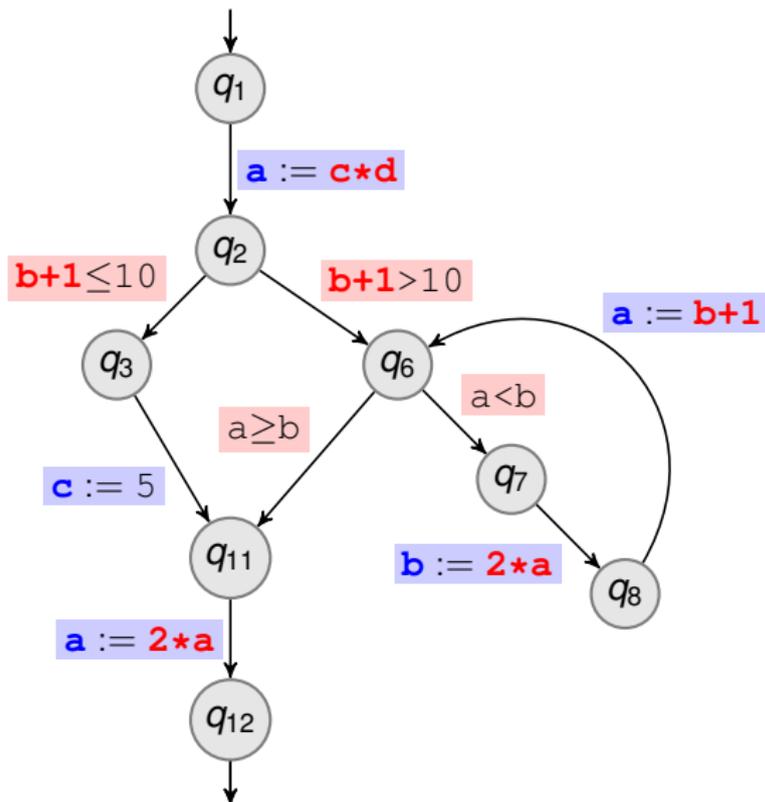
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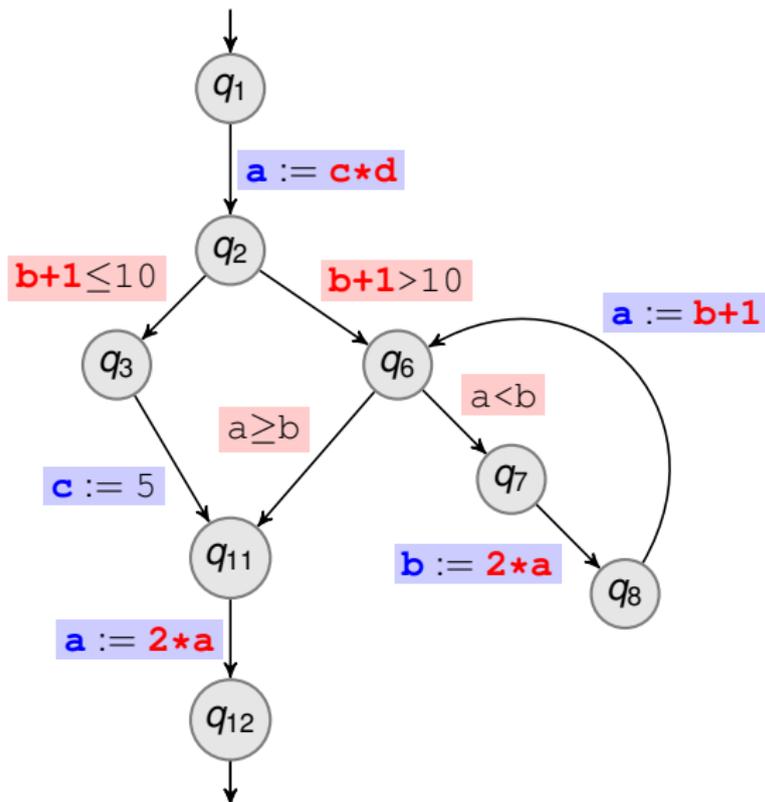
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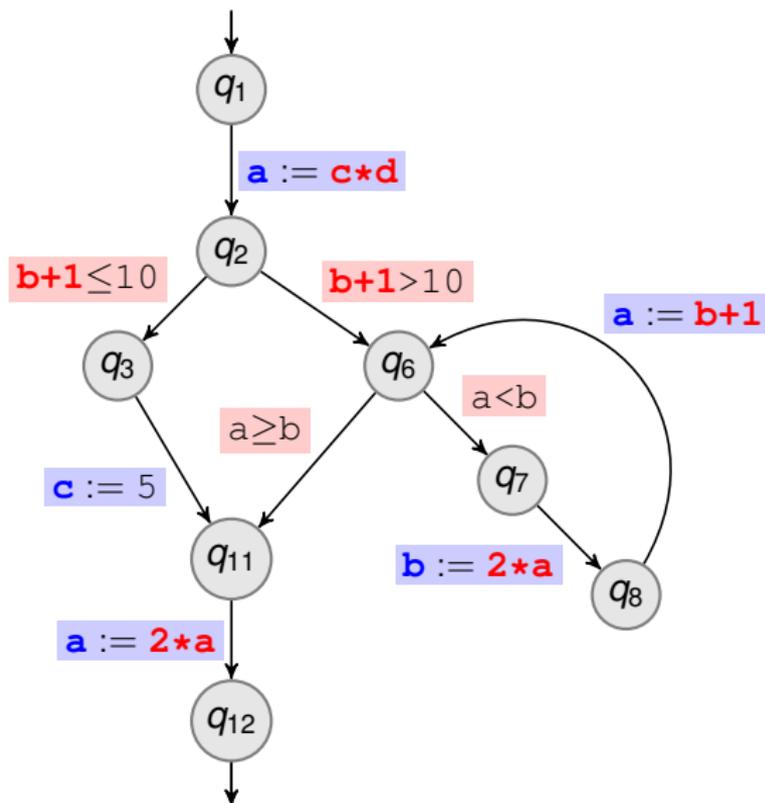
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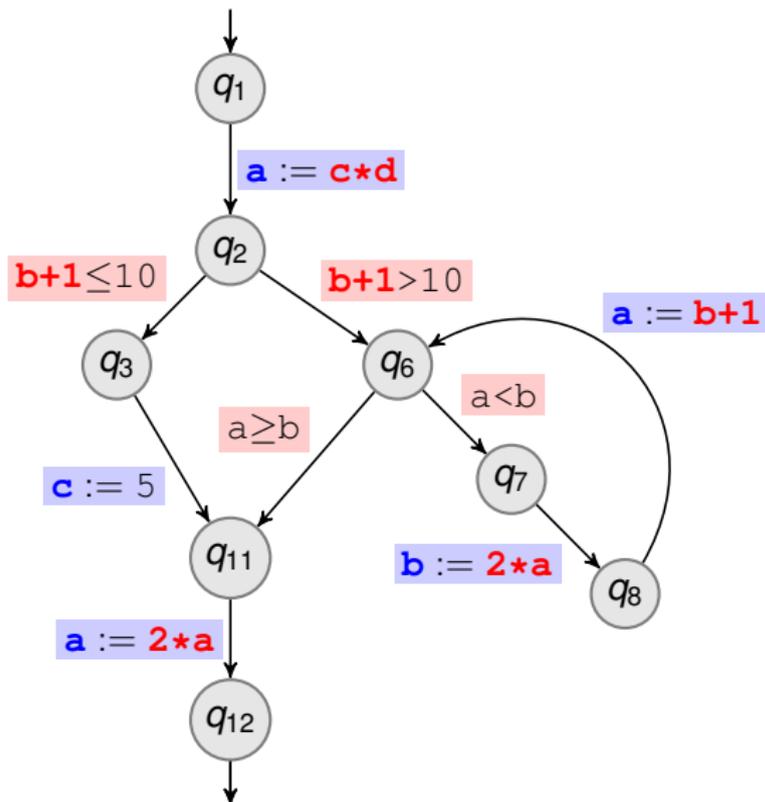
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# Available Expressions Analysis: Formulation

Control Flow Automaton:  $\langle Q, q_{in}, q_{out}, X, \rightarrow \rangle$

System of equations: variables  $A_q$ , with  $A_q \subseteq SubExp(\rightarrow)$

$$A_q = \bigcap_{q' \xrightarrow{op} q} Gen_{op} \cup (A_{q'} \setminus Kill_{op}) \qquad A(q_{in}) = \emptyset$$

$$Gen_{op} = \begin{cases} SubExp(g) & \text{if } op = g \\ \{f \in SubExp(e) \mid x \notin SubExp(e)\} & \text{if } op = x := e \end{cases}$$

$$Kill_{op} = \begin{cases} \emptyset & \text{if } op = g \\ \{e \in SubExp(\rightarrow) \mid x \in Var(e)\} & \text{if } op = x := e \end{cases}$$

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## Code Optimization

Avoid recomputation of an expression



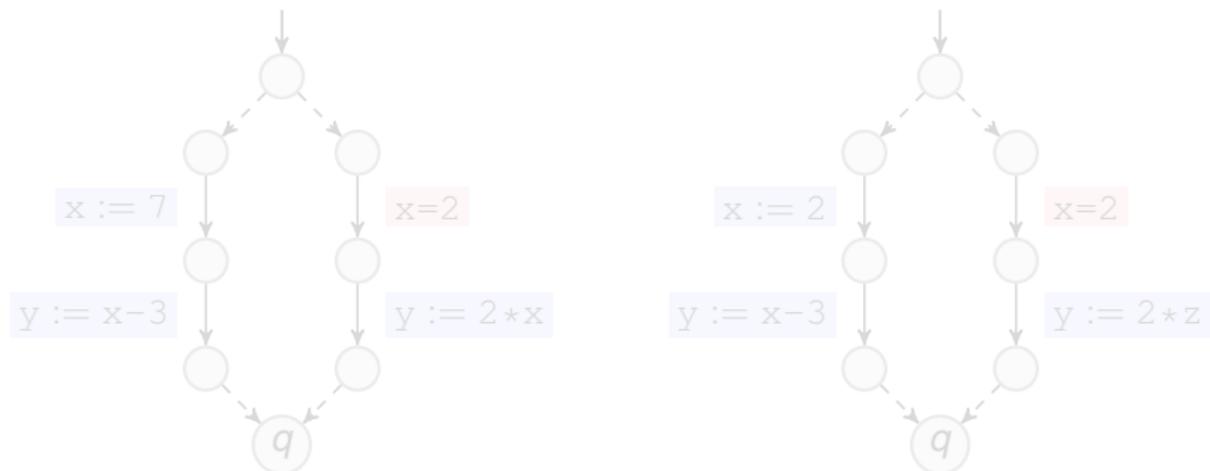
If  $e$  is available at location  $q_1$  then we may reuse its value to evaluate the operation on the edge from  $q_1$  to  $q_2$ .

This is sound since the analysis is conservative

# Constant Propagation Analysis: Definition

## Definition

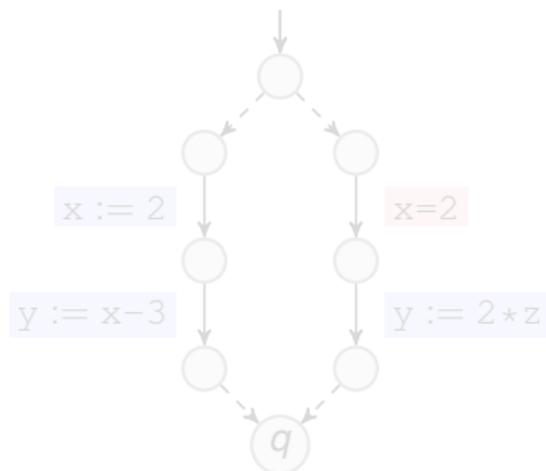
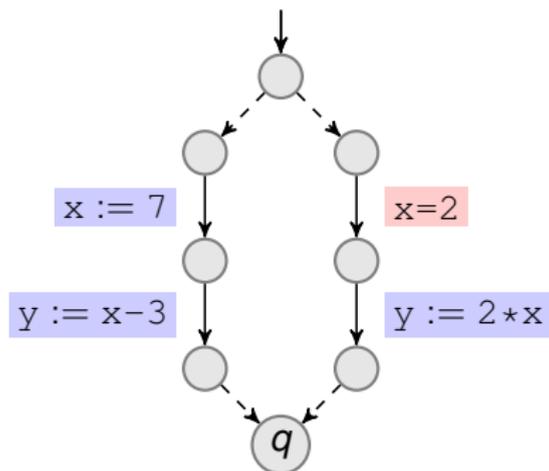
A variable  $x$  is **constant** at location  $q$  if we have  $v(x) = v'(x)$  for any two reachable configurations  $(q, v)$  and  $(q, v')$  in  $\text{Post}^*$ .



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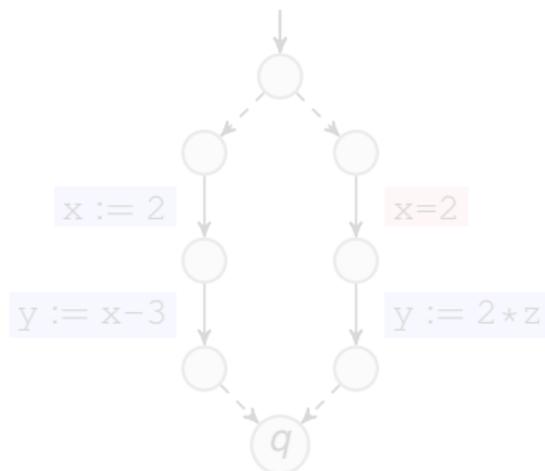
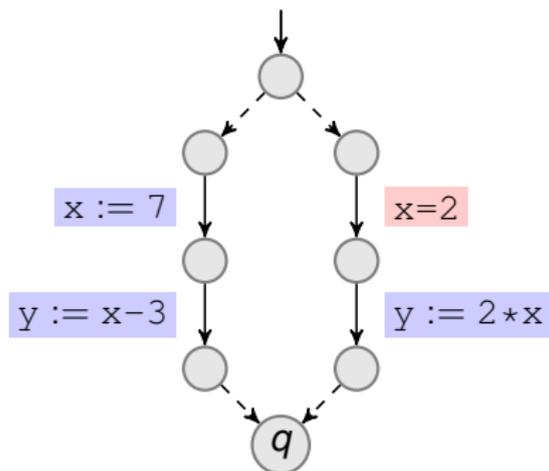
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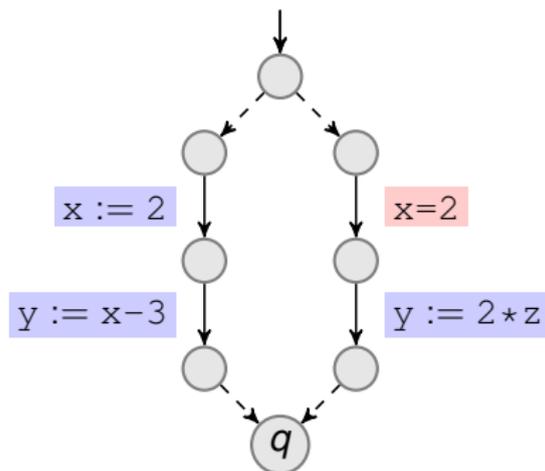
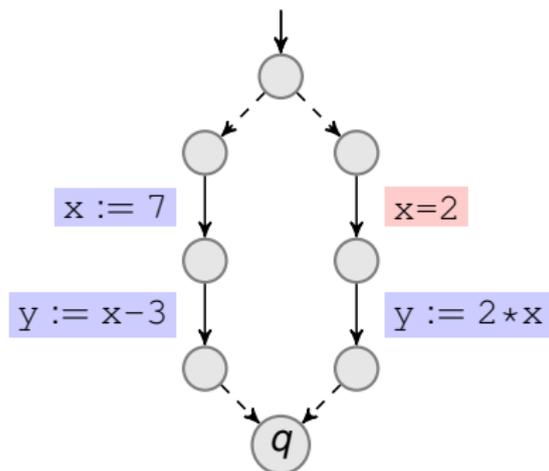


$x$  not constant,  $y$  constant

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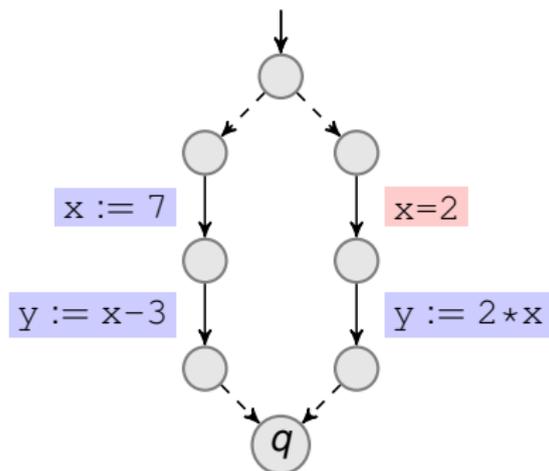


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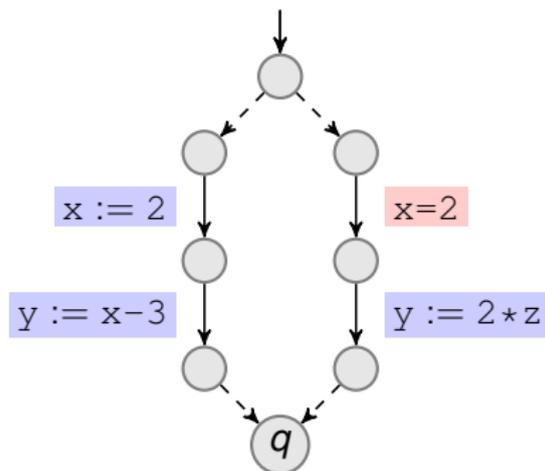
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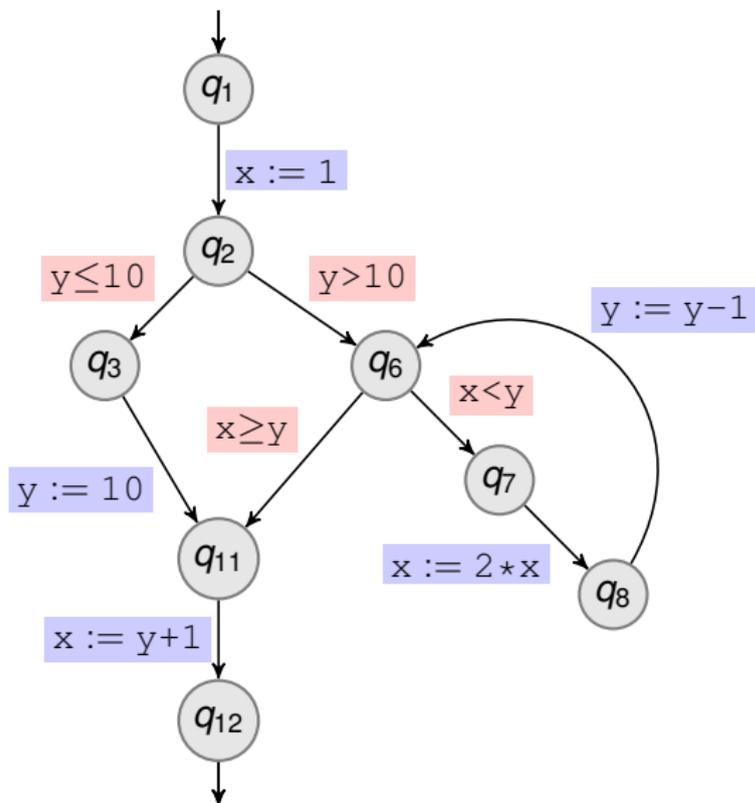


$x$  not constant,  $y$  constant

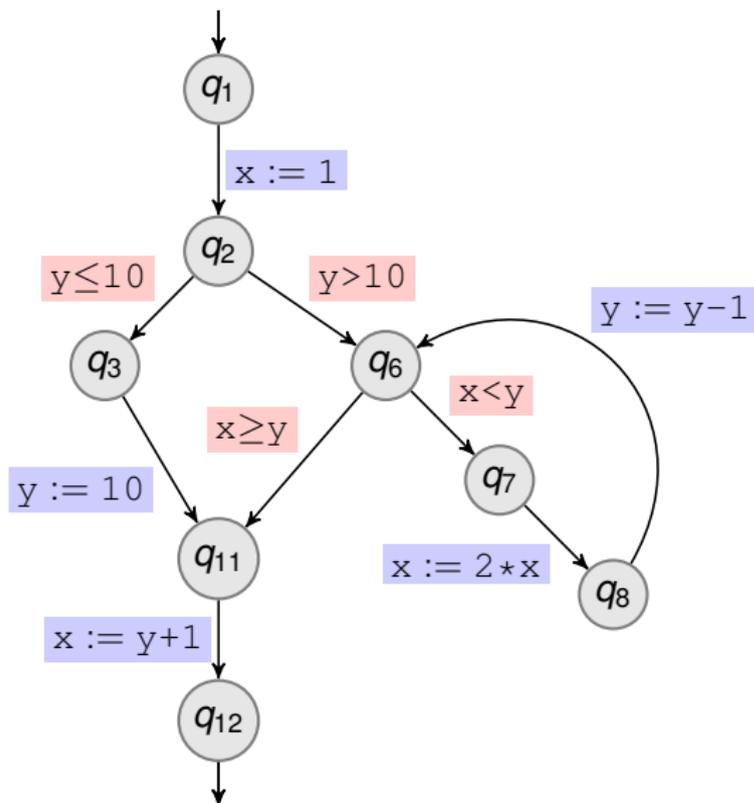


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# Constant Propagation Analysis: Running Example



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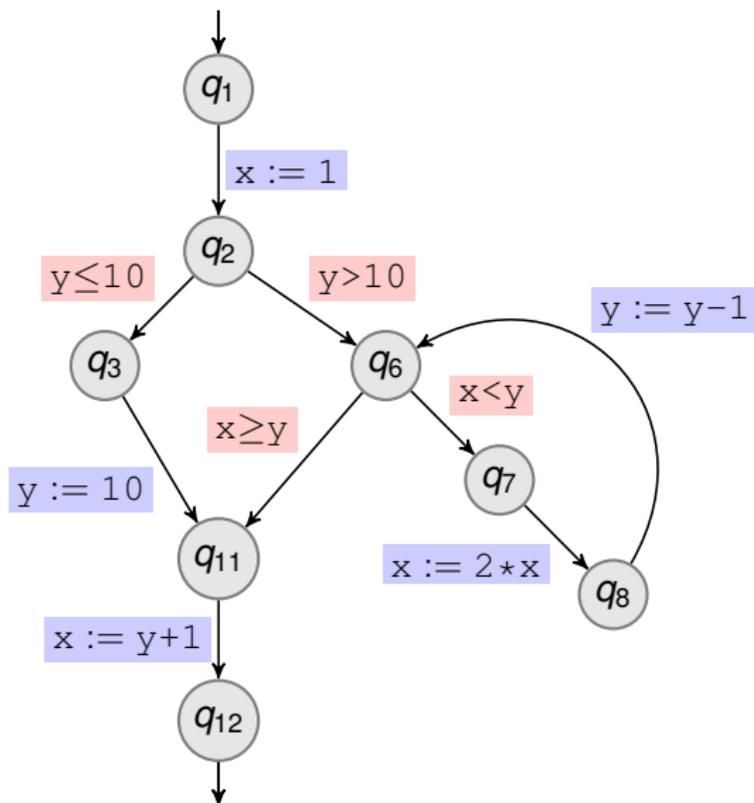


0 : Initialization

1 : Propagation ( $\rightarrow$ )

	x	y
$q_1$	$\top$	$\top$
$q_2$		
$q_3$		
$q_6$		
$q_7$		
$q_8$		
$q_{11}$		
$q_{12}$		

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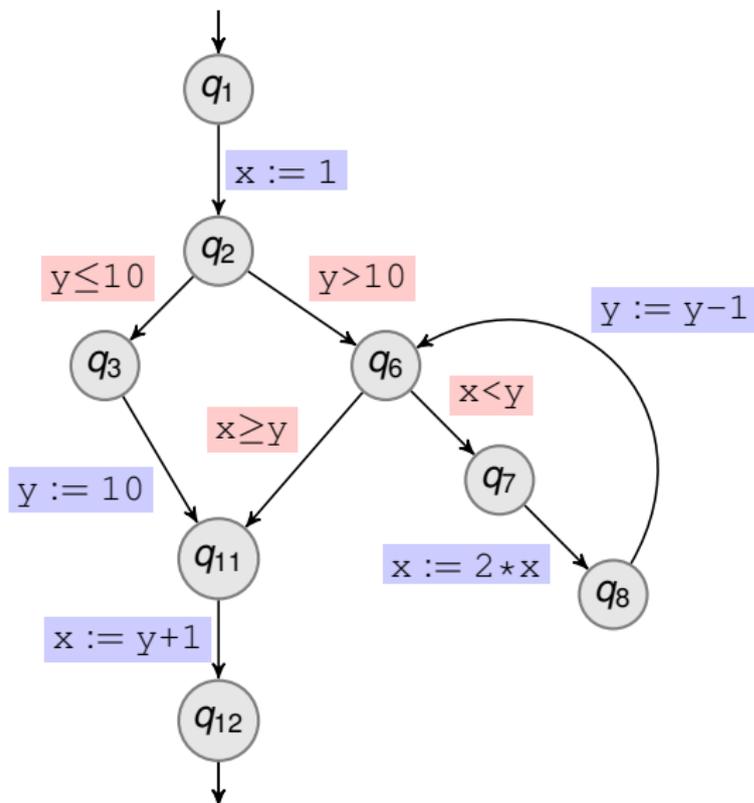


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$q_{11}$		
$q_{12}$		

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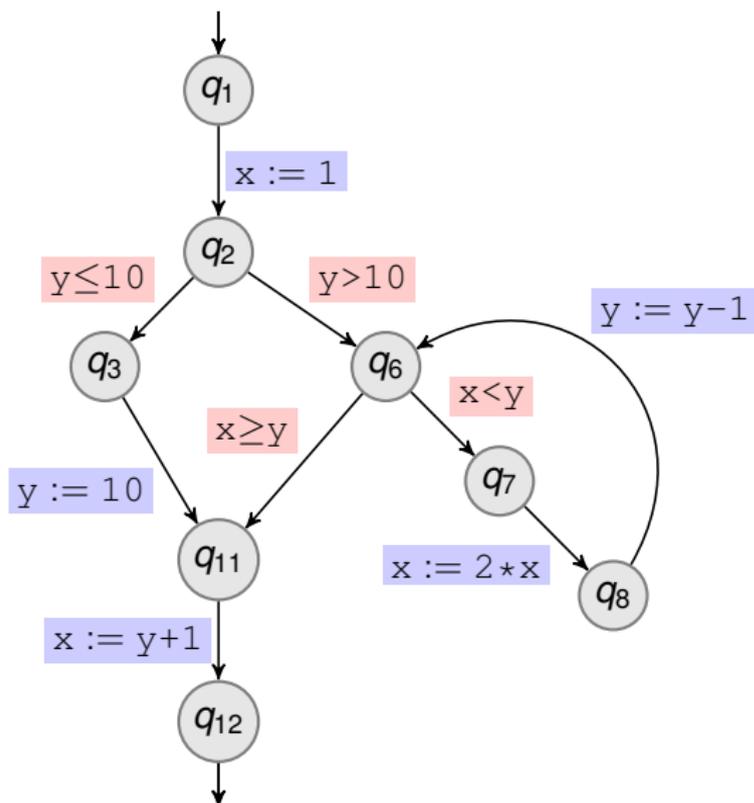


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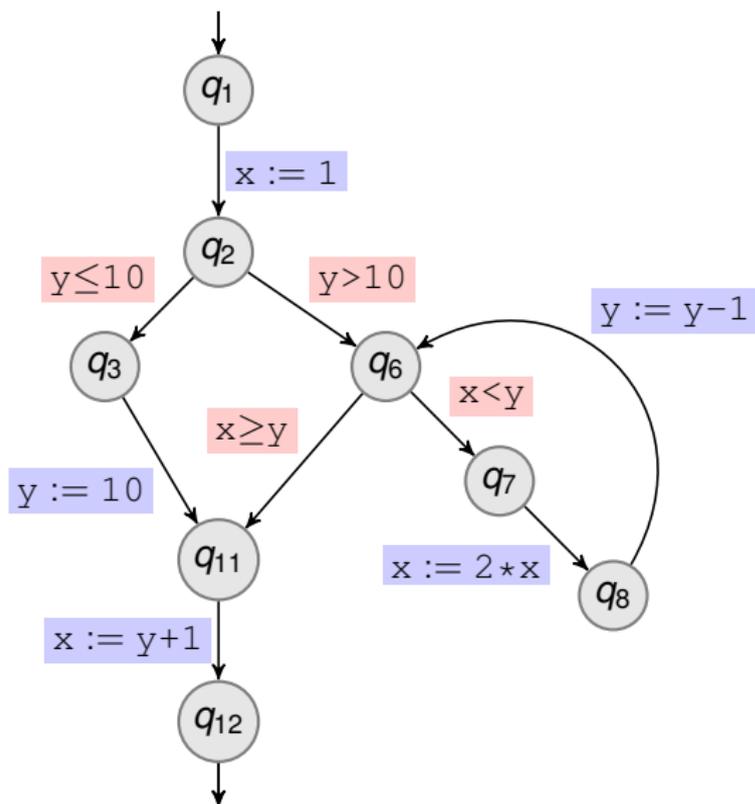


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$q_3$	1	$\top$
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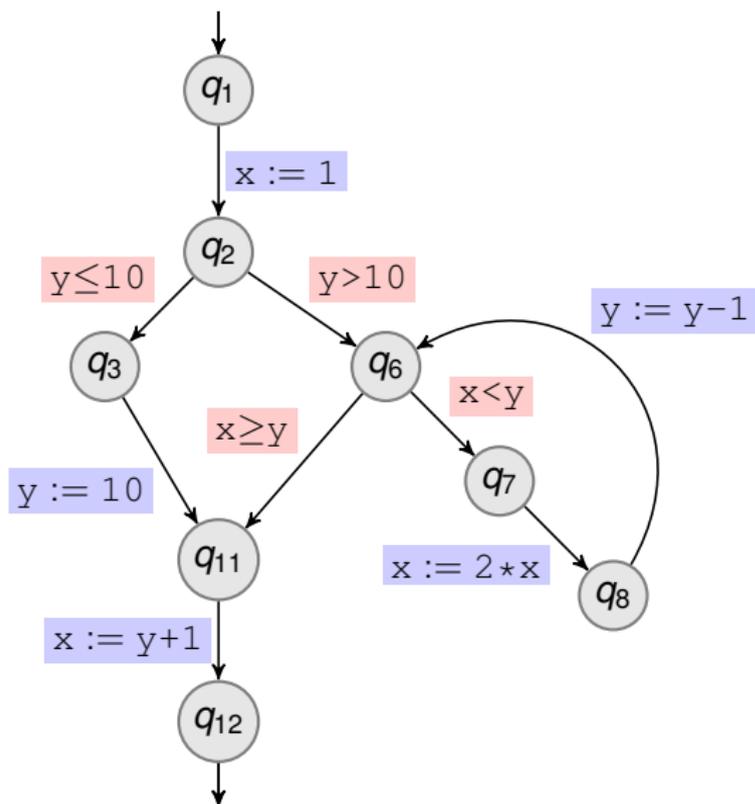


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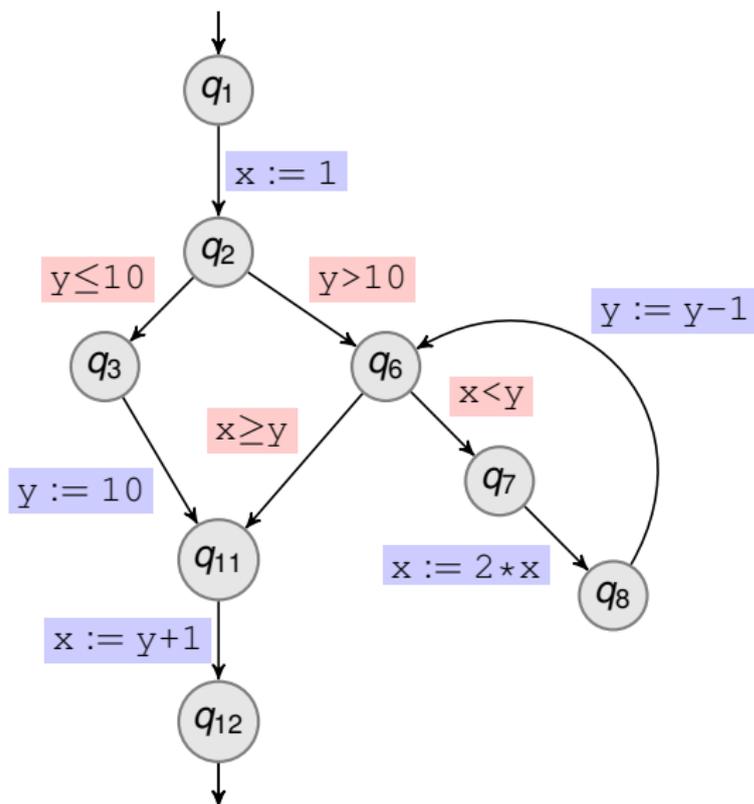
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$q_2$	1	$\top$
$q_3$	1	$\top$
$q_6$		
$q_7$		
$q_8$		
$q_{11}$	1	10
$q_{12}$		

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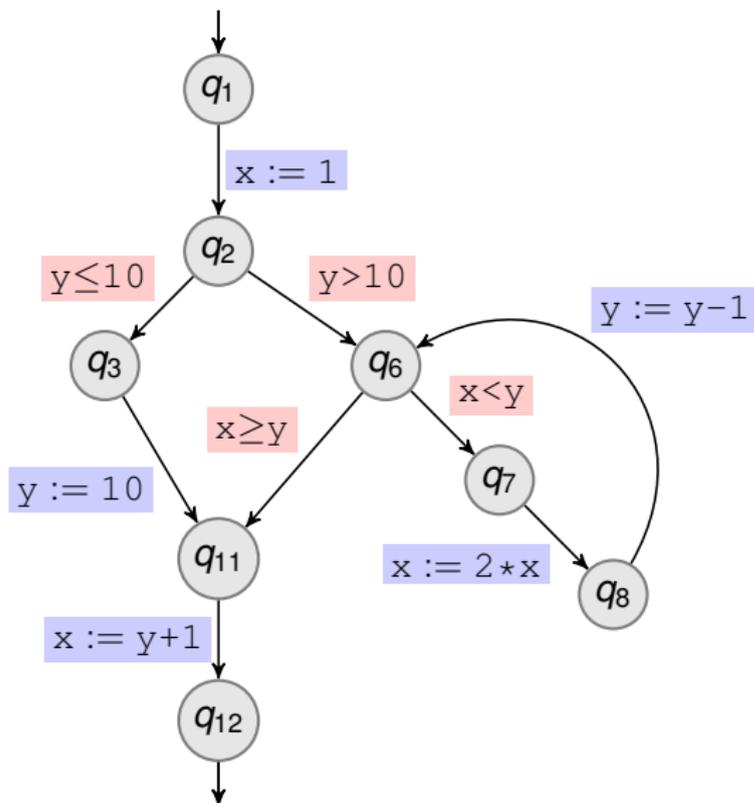


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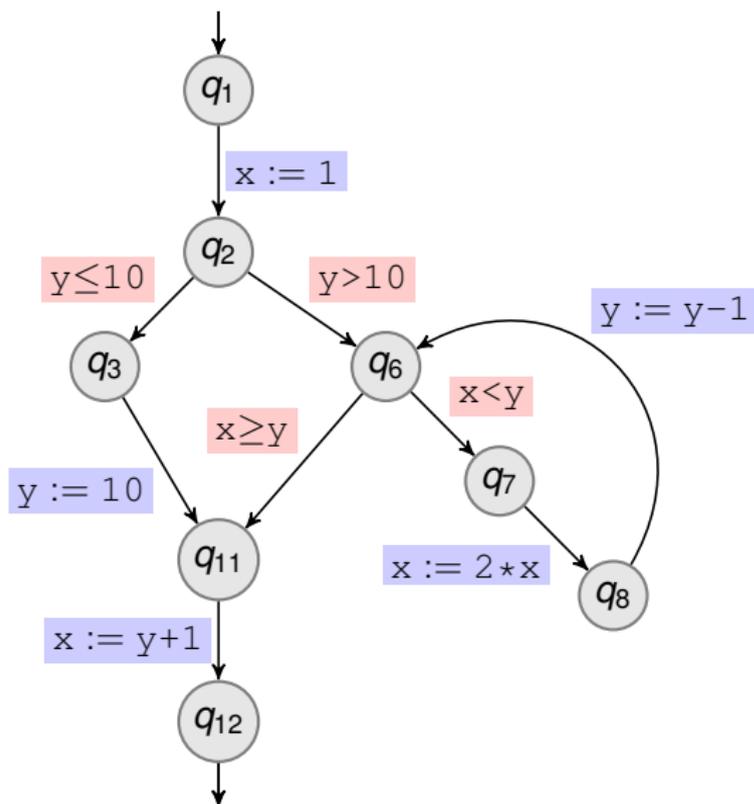
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$q_6$		
$q_7$		
$q_8$		
$q_{11}$	1	10
$q_{12}$	11	10

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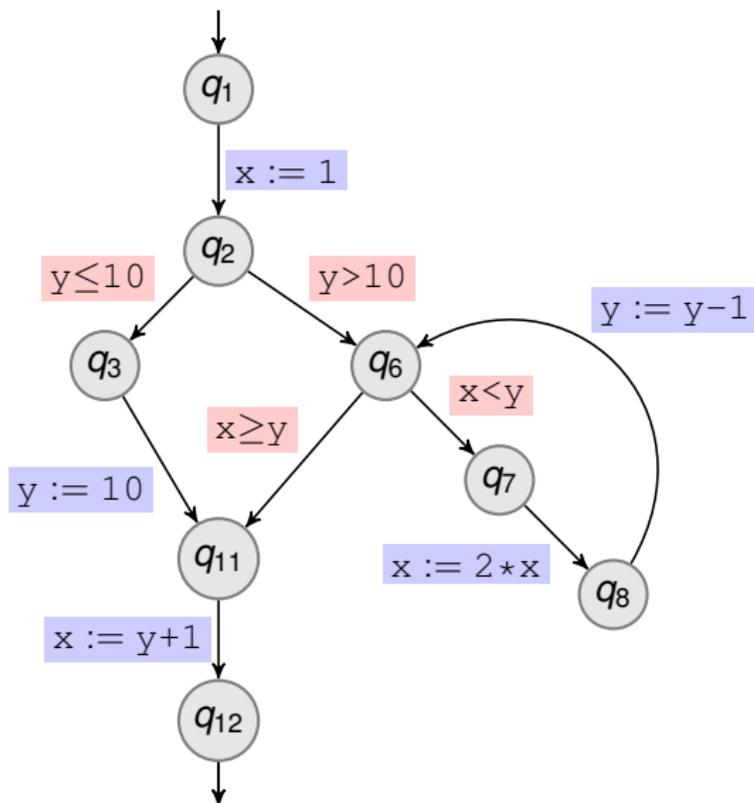


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1 : Propagation ( $\rightarrow$ )

	x	y
$q_1$	$\top$	$\top$
$q_2$	1	$\top$
$q_3$	1	$\top$
$q_6$		
$q_7$		
$q_8$		
$q_{11}$	1	10
$q_{12}$	11	10

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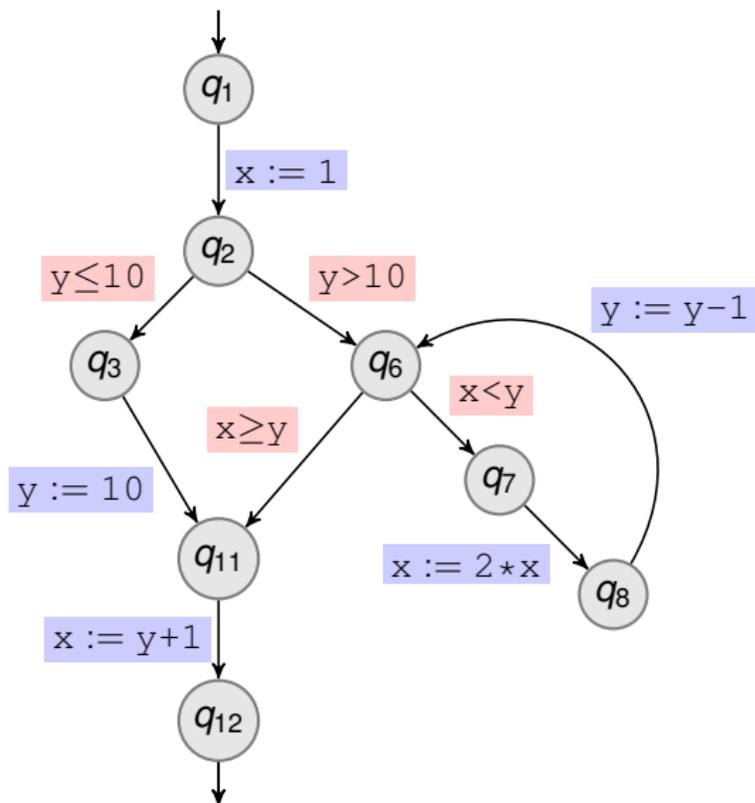


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$q_3$	1	$\top$
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$q_7$		
$q_8$		
$q_{11}$	1	10
$q_{12}$	11	10

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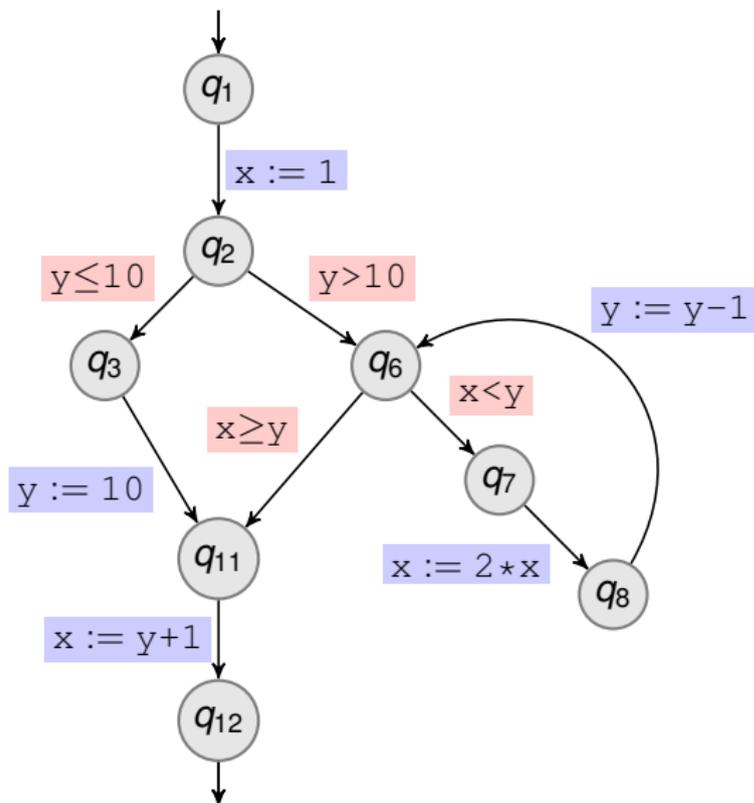


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$q_7$		
$q_8$		
$q_{11}$	1	10
$q_{12}$	11	10

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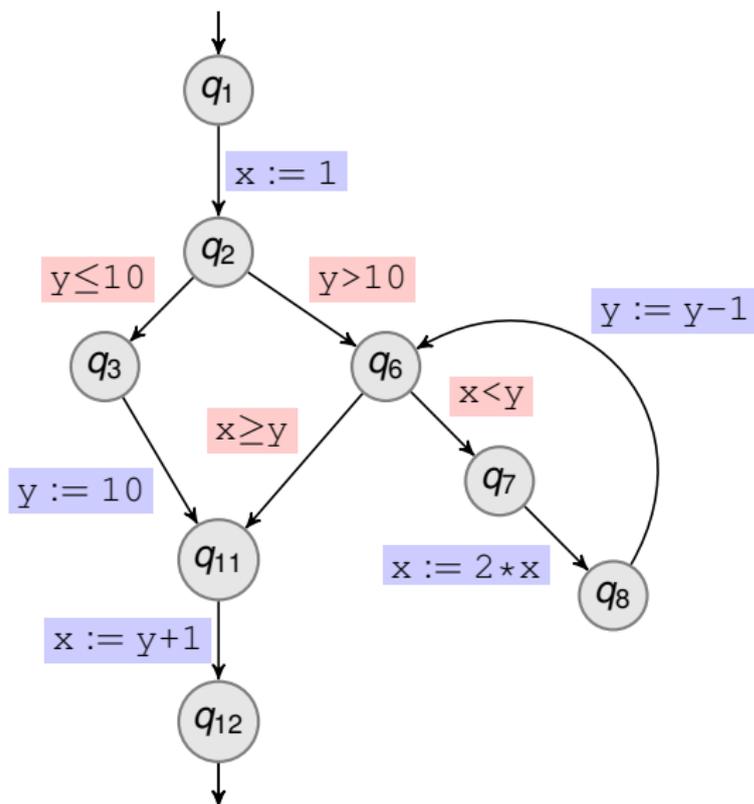


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1 : Propagation ( $\rightarrow$ )

	x	y
$q_1$	$\top$	$\top$
$q_2$	1	$\top$
$q_3$	1	$\top$
$q_6$	1	$\top$
$q_7$		
$q_8$		
$q_{11}$	1	10, $\top$
$q_{12}$	11	10

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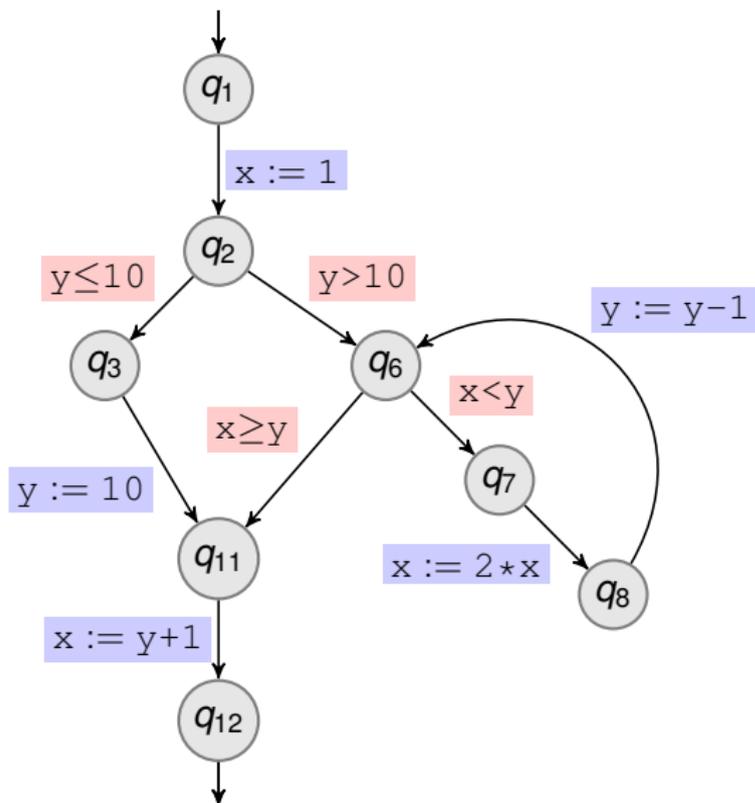


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	x	y
$q_1$	$\top$	$\top$
$q_2$	1	$\top$
$q_3$	1	$\top$
$q_6$	1	$\top$
$q_7$		
$q_8$		
$q_{11}$	1	$\top$
$q_{12}$	11	10

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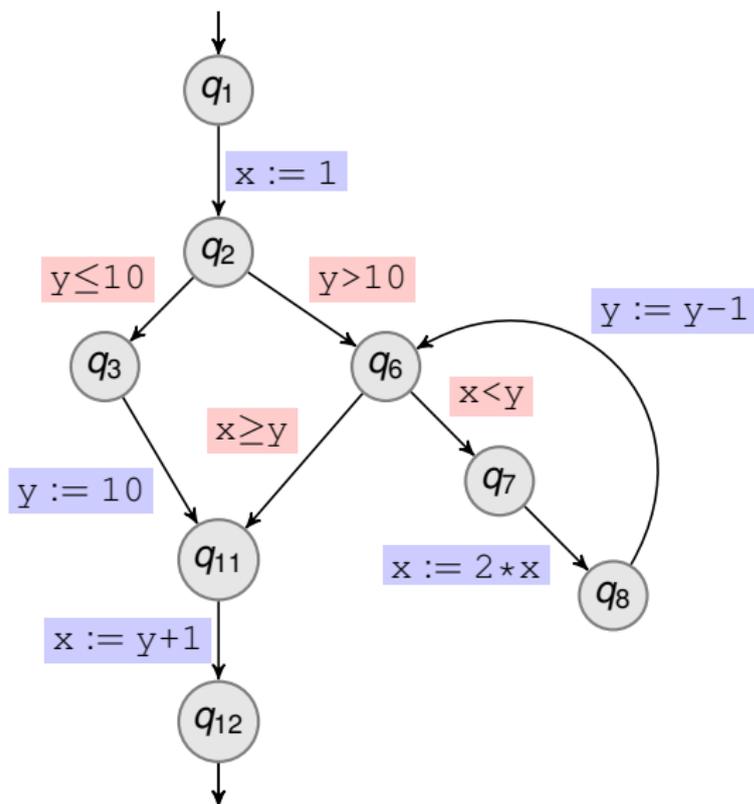


0 : Initialization

1 : Propagation ( $\rightarrow$ )

	x	y
$q_1$	$\top$	$\top$
$q_2$	1	$\top$
$q_3$	1	$\top$
$q_6$	1	$\top$
$q_7$		
$q_8$		
$q_{11}$	1	$\top$
$q_{12}$	11, 2	10, $\top$

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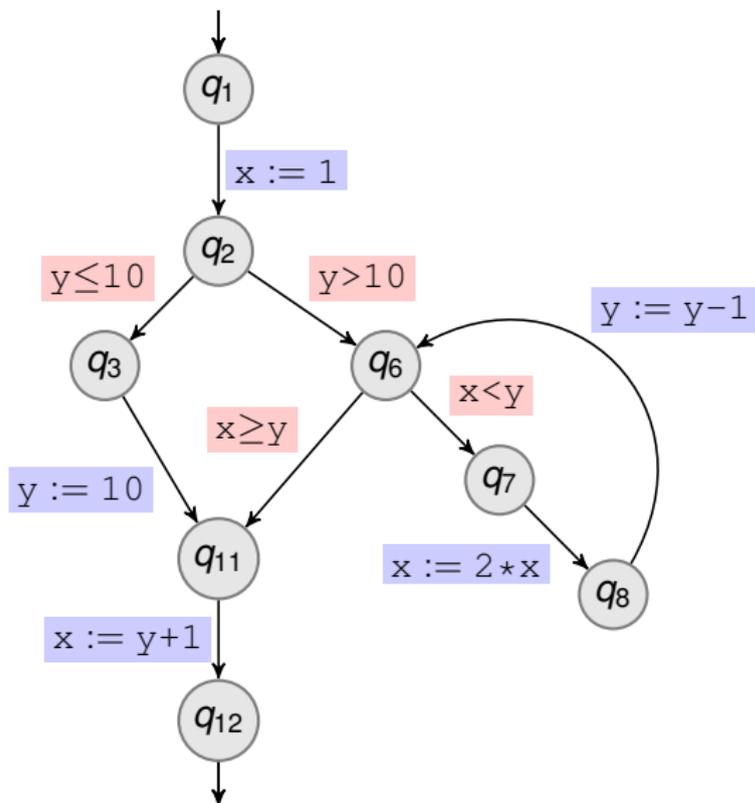


0 : Initialization

1 : Propagation ( $\rightarrow$ )

	x	y
$q_1$	⊥	⊥
$q_2$	1	⊥
$q_3$	1	⊥
$q_6$	1	⊥
$q_7$		
$q_8$		
$q_{11}$	1	⊥
$q_{12}$	⊥	⊥

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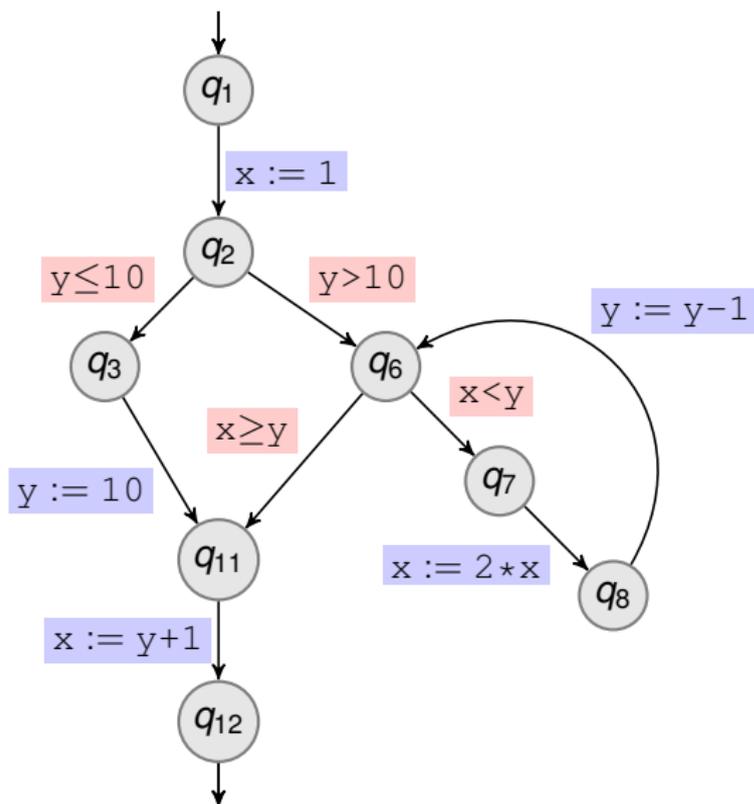


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	x	y
$q_1$	⊤	⊤
$q_2$	1	⊤
$q_3$	1	⊤
$q_6$	1	⊤
$q_7$	1	⊤
$q_8$		
$q_{11}$	1	⊤
$q_{12}$	⊤	⊤

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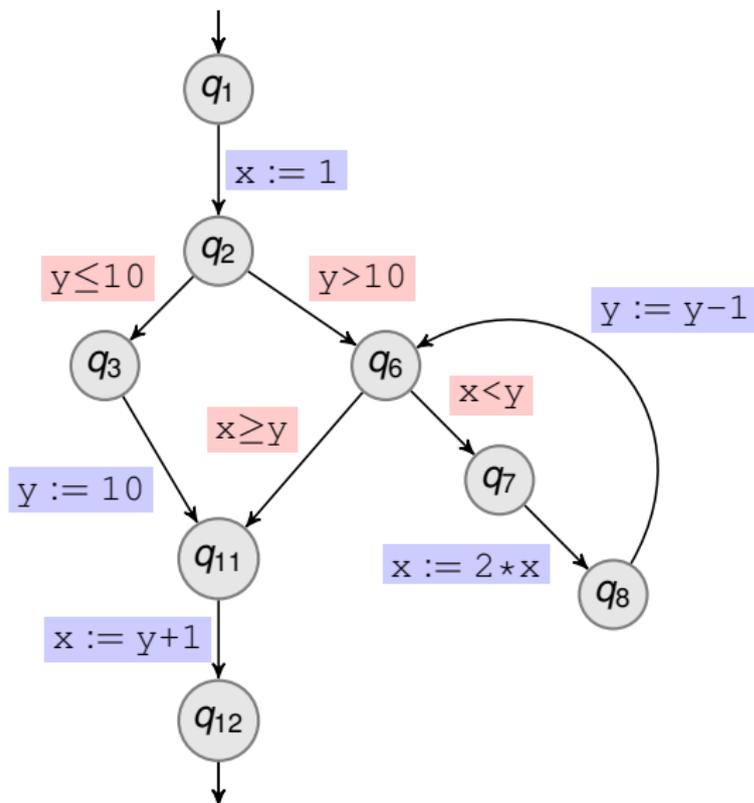


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	x	y
$q_1$	⊤	⊤
$q_2$	1	⊤
$q_3$	1	⊤
$q_6$	1	⊤
$q_7$	1	⊤
$q_8$		
$q_{11}$	1	⊤
$q_{12}$	⊤	⊤

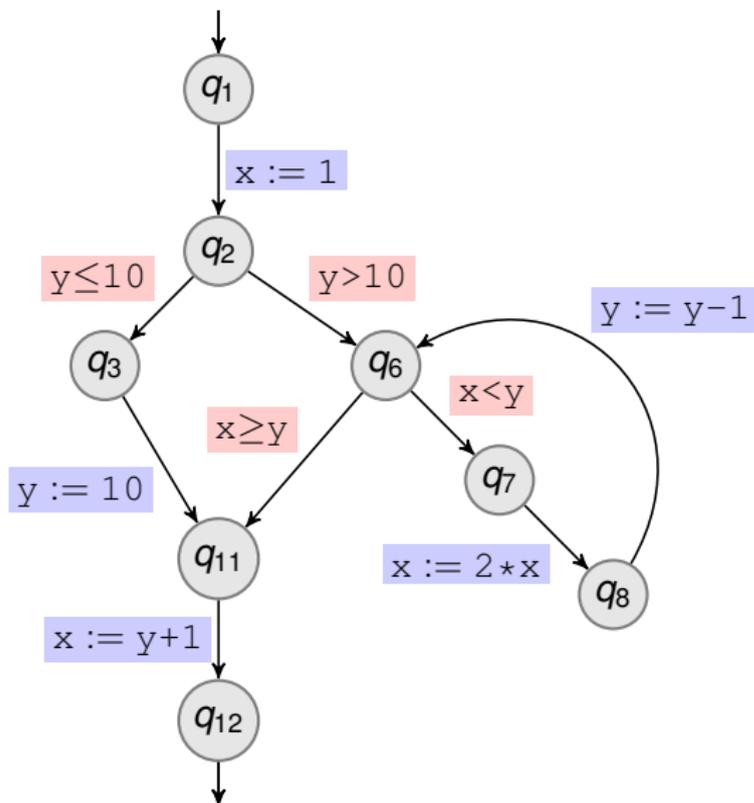
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1 : Propagation ( $\rightarrow$ )

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$q_1$	$\top$	$\top$
$q_2$	1	$\top$
$q_3$	1	$\top$
$q_6$	1	$\top$
$q_7$	1	$\top$
$q_8$	2	$\top$
$q_{11}$	1	$\top$
$q_{12}$	$\top$	$\top$

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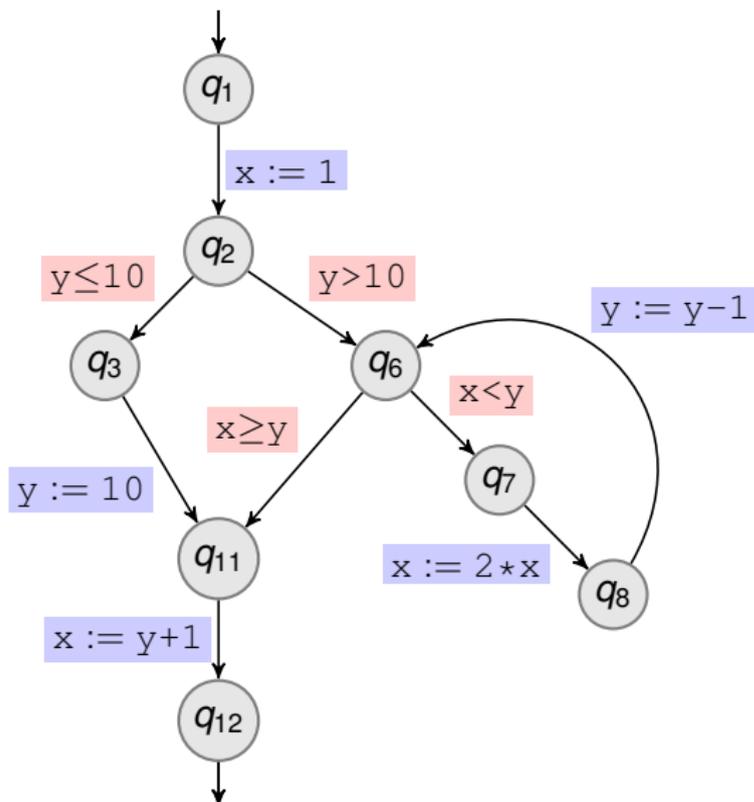


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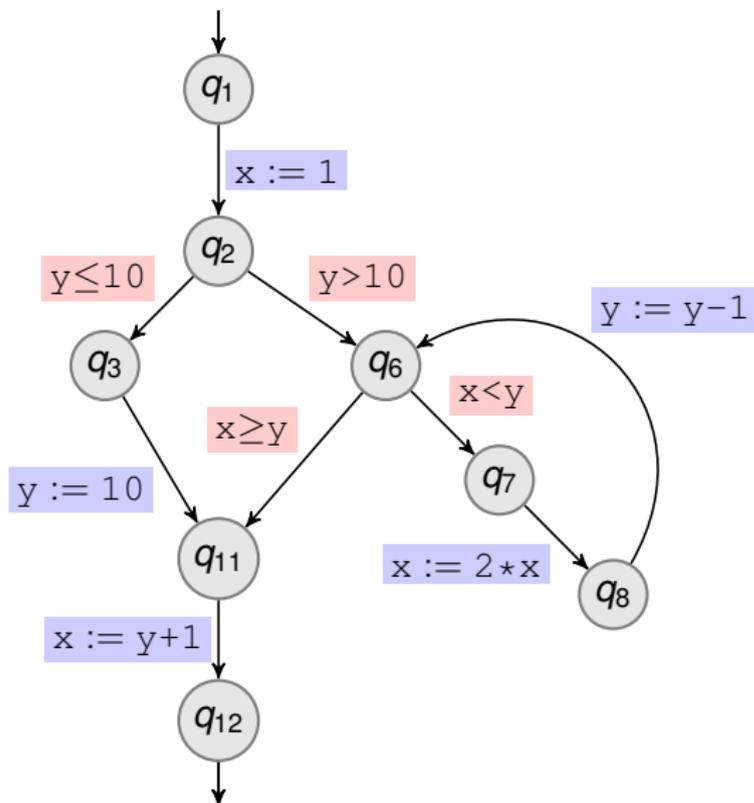


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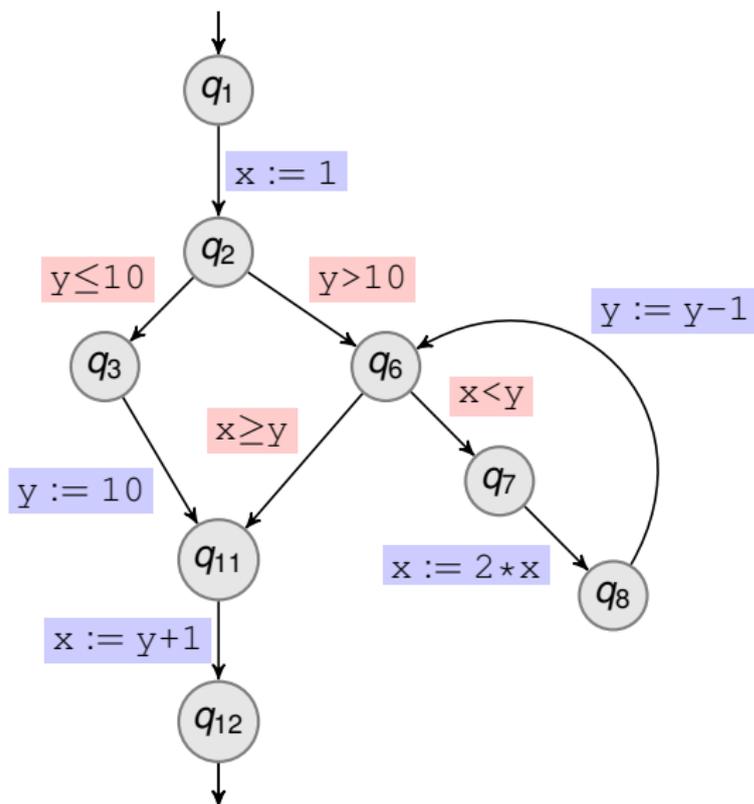


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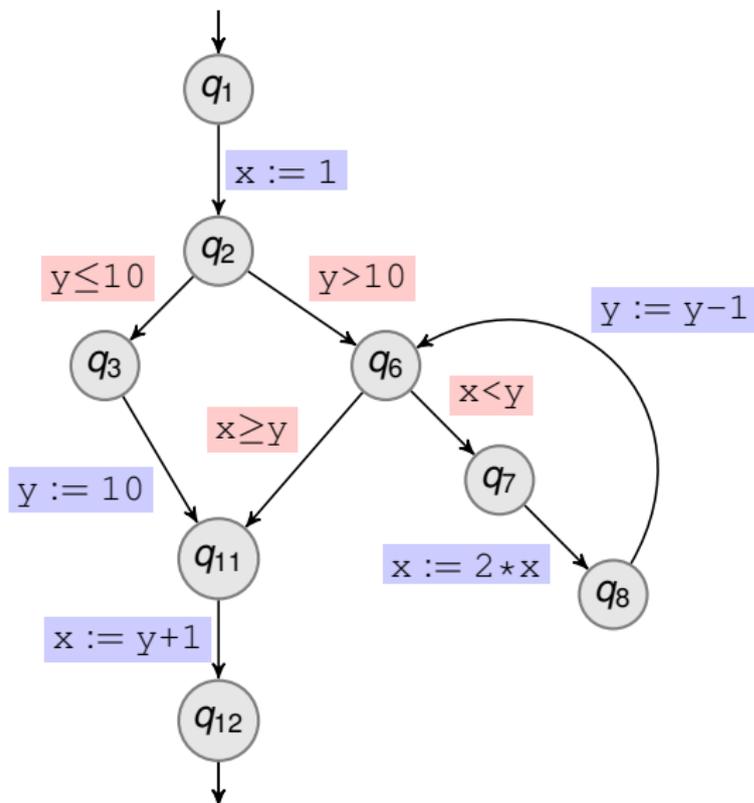


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$q_8$	2	$\top$
$q_{11}$	1, $\top$	$\top$
$q_{12}$	$\top$	$\top$

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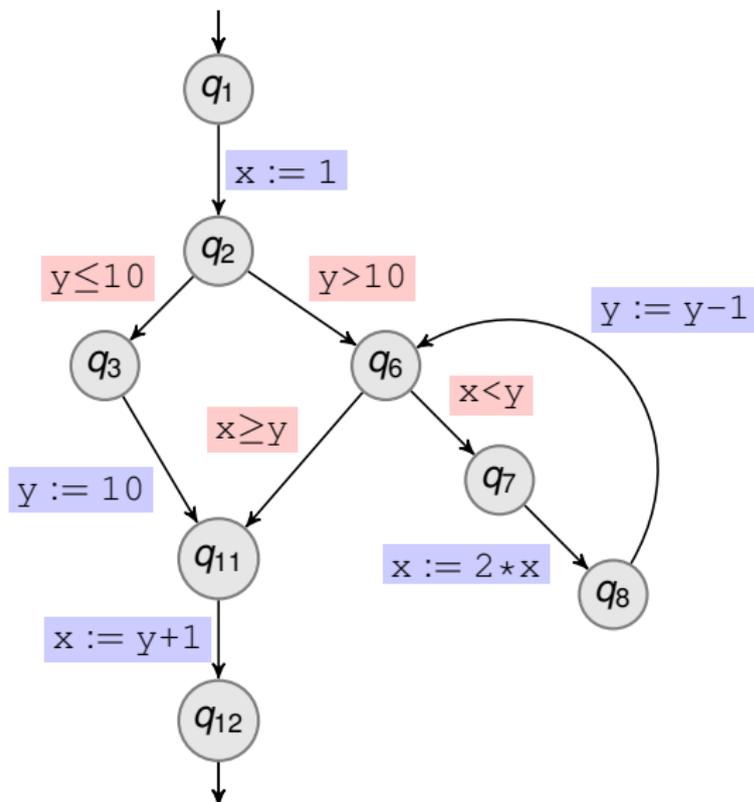


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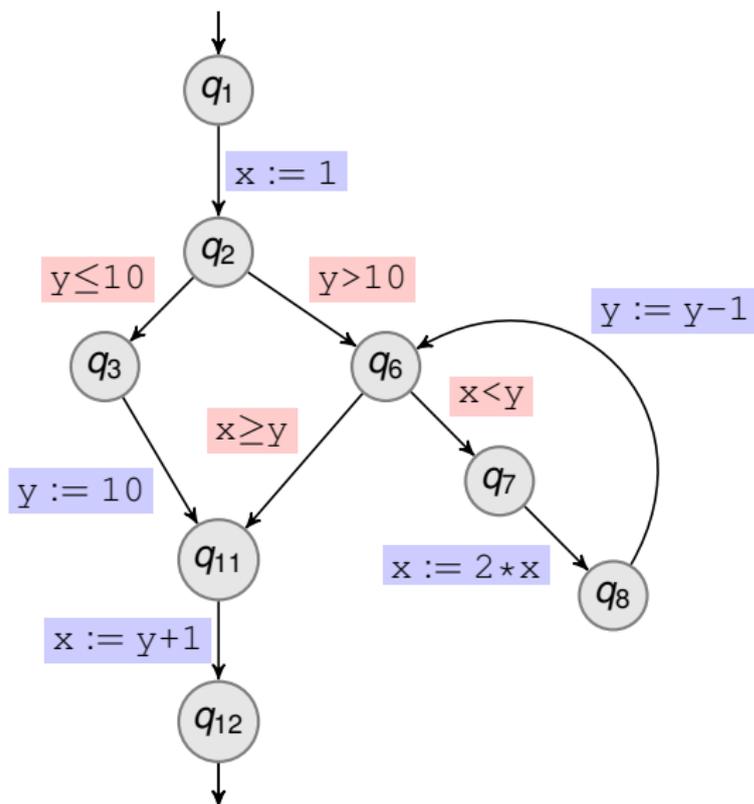


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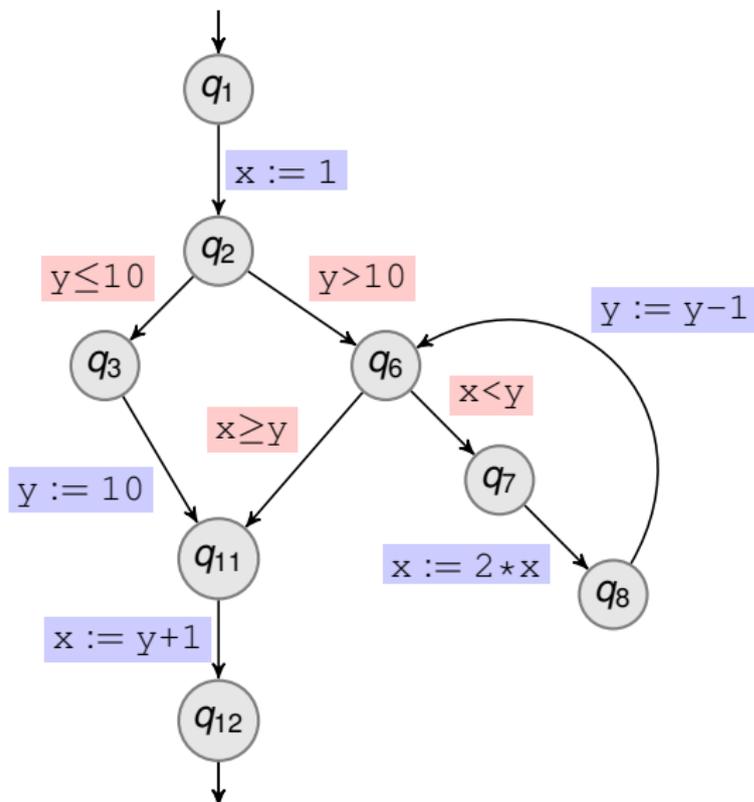


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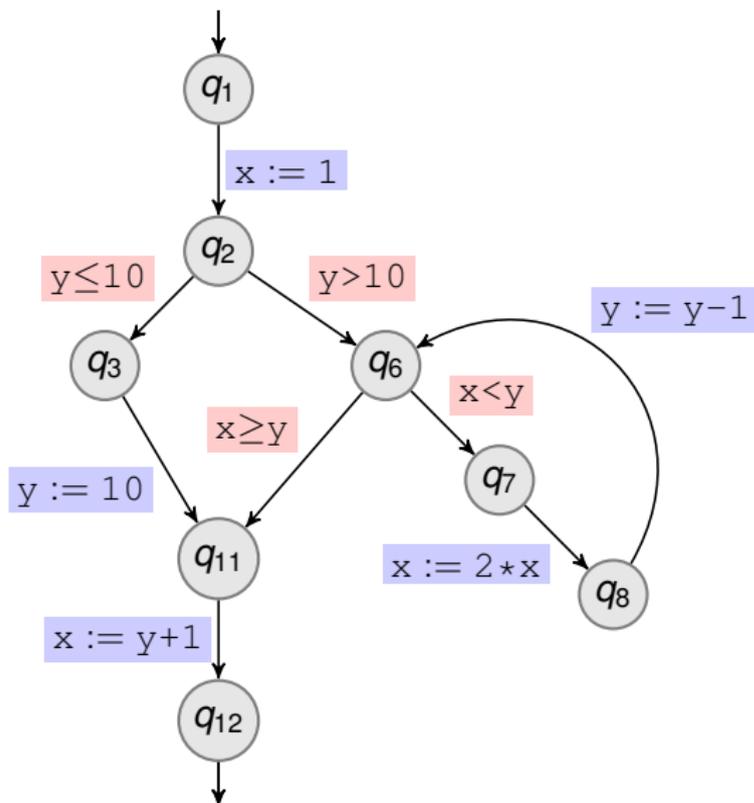


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$q_7$	$\top$	$\top$
$q_8$	2, $\top$	$\top$
$q_{11}$	$\top$	$\top$
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$q_{11}$	$\top$	$\top$
$q_{12}$	$\top$	$\top$

# Constant Propagation Analysis: Formulation

Extend  $\mathbb{R}$  with a new element  $\top$  to account for non-constant values

Extend  $+$ ,  $-$  and  $\times$  such that  $\top$  is absorbent

$$\begin{aligned} \top + r &= r + \top = \top \\ \top - r &= r - \top = \top \\ \top \times r &= r \times \top = \top \end{aligned} \quad \text{for } r \in \mathbb{R} \cup \{\top\}$$

Extend  $\llbracket e \rrbracket_v$  to valuations from  $X$  to  $\mathbb{R} \cup \{\top\}$

Domain of data flow “information”

$$\mathbb{D} = X \rightarrow (\mathbb{R} \cup \{\top\})$$

# Constant Propagation Analysis: Formulation

$$\mathbb{D} = X \rightarrow (\mathbb{R} \cup \{\top\})$$

System of equations: variables  $C_q$  for  $q \in Q$ , with  $C_q \in \mathbb{D}$

$$C_q = \bigotimes_{q' \xrightarrow{\text{op}} q} f_{\text{op}}(C_{q'}) \qquad C(q_{\text{in}}) = \lambda x. \top$$

$$v \otimes v' = \lambda y. \begin{cases} v(y) & \text{if } v(y) = v'(y) \\ \top & \text{otherwise} \end{cases}$$

Functions  $f_{\text{op}}$

$$f_{x:=e}(v) = \lambda y. \begin{cases} v(y) & \text{if } y \neq x \\ \llbracket e \rrbracket_v & \text{if } y = x \end{cases} \qquad f_g(v) = v$$

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## Code Optimization

### Constant folding



For each variable  $y$  occurring in  $e$ , if  $y$  is constant at location  $q_1$  then we may replace  $y$  with its constant value in  $e$ .

This is sound since the analysis is conservative

# Common Form of Data Flow Equations

- Domain  $\mathbb{D}$  of data flow “information”
  - sets of variables, sets of expressions, valuations, ...
- Variables  $D_q$  for  $q \in Q$ , with value in  $\mathbb{D}$ 
  - $D_q$  holds data-flow information for location  $q$

$$D_q = \mathbb{M} f(D_{q'})$$

- “Confluence” operator  $\mathbb{M}$  on  $\mathbb{D}$  to merge data flow information
  - $\cup, \cap, \otimes, \dots$
- Functions  $f : \mathbb{D} \rightarrow \mathbb{D}$  to model the effect of operations

# Common Form of Data Flow Equations

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$$D_q = \mathbb{M} f(D_{q'})$$

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- 5 Classical Data Flow Analyses
- 6 Basic Lattice Theory**
- 7 Monotone Data Flow Analysis Frameworks

# Partial Order

A **partial order** on a set  $L$  is any binary relation  $\sqsubseteq \subseteq L \times L$  satisfying for all  $x, y, z \in L$ :

$$x \sqsubseteq x \quad (\text{reflexivity})$$

$$x \sqsubseteq y \wedge y \sqsubseteq x \implies x = y \quad (\text{antisymmetry})$$

$$x \sqsubseteq y \wedge y \sqsubseteq z \implies x \sqsubseteq z \quad (\text{transitivity})$$

A **partially ordered set** is any pair  $(L, \sqsubseteq)$  where  $L$  is a set and  $\sqsubseteq$  is a **partial order** on  $L$ .

There can be  $x$  and  $y$  in  $L$  such that  $x \not\sqsubseteq y$  and  $y \not\sqsubseteq x$ .

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# Lower and Upper Bounds

Consider a partially ordered set  $(L, \sqsubseteq)$  and a subset  $X \subseteq L$ .

## Greatest Lower Bound

A **lower bound** of  $X$  is any  $b \in X$  such that  $b \sqsubseteq x$  for all  $x \in X$ .

A **greatest lower bound** of  $X$  is any  $glb \in X$  such that:

- 1  $glb$  is a lower bound of  $X$ ,
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## Least Upper Bound

An **upper bound** of  $X$  is any  $b \in X$  such that  $b \sqsupseteq x$  for all  $x \in X$ .

A **least upper bound** of  $X$  is any  $lub \in X$  such that:

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# Lower and Upper Bounds: Examples

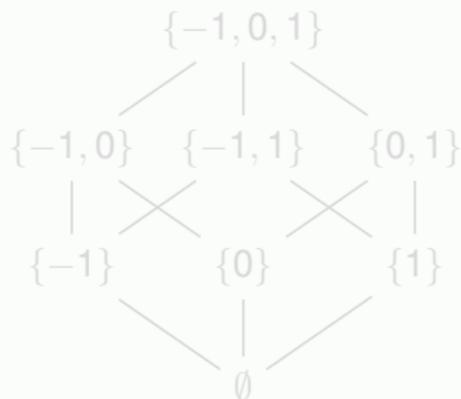
$(\mathbb{R}, \leq)$

$$\sqcup \{0, \sqrt{2}, 4\} = 4$$

$$\sqcap \left\{ \frac{1}{2^n} \mid n \in \mathbb{N} \right\} = 0$$

But  $\{\dots, -2, -1, 0, 1, 2, \dots\}$  has no upper bound and no lower bound.

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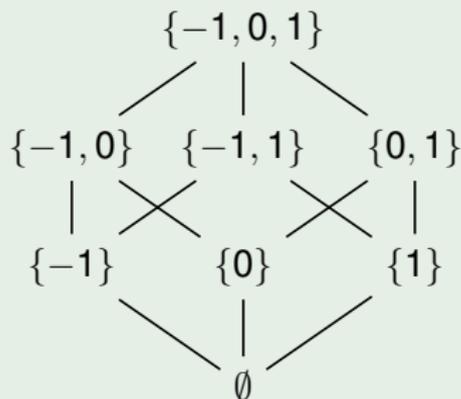
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A **fixpoint** of  $f$  is any  $x \in L$  such that  $f(x) = x$ .

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# Knaster-Tarski Fixpoint Theorem

A function  $f : L \rightarrow L$  on a partially ordered set  $(L, \sqsubseteq)$  is **monotonic** if for all  $x, y \in L$ :

$$x \sqsubseteq y \implies f(x) \sqsubseteq f(y)$$

## Theorem

*Every monotonic function  $f$  on a complete lattice  $(L, \sqsubseteq)$  has a least fixpoint  $\text{lfp}(f)$  and a greatest fixpoint  $\text{gfp}(f)$ . Moreover:*

$$\text{lfp}(f) = \bigsqcap \{x \in L \mid f(x) \sqsubseteq x\}$$

$$\text{gfp}(f) = \bigsqcup \{x \in L \mid f(x) \sqsupseteq x\}$$

# Order Duality

If  $(L, \sqsubseteq)$  is a partially ordered set then so is  $(L, \supseteq)$ .

If  $(L, \sqsubseteq)$  is a complete lattice then so is  $(L, \supseteq)$ .

$$\bigcap_{(L, \supseteq)} = \bigcup_{(L, \sqsubseteq)} \qquad \perp_{(L, \supseteq)} = \top_{(L, \sqsubseteq)}$$

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# Ascending Chain Condition

An **ascending chain** in a partially ordered set  $(L, \sqsubseteq)$  is any infinite sequence  $x_0, x_1, \dots$  of elements of  $L$  satisfying  $x_i \sqsubseteq x_{i+1}$  for all  $i \in \mathbb{N}$ .

A partially ordered set  $(L, \sqsubseteq)$  satisfies the **ascending chain condition** if every ascending chain  $x_0 \sqsubseteq x_1 \sqsubseteq \dots$  of elements of  $L$  is eventually stationary.

## Examples

$(\mathbb{R}, \leq)$  does not satisfy the ascending chain condition.

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Consider a partially ordered set  $(L, \sqsubseteq)$  and  $f : L \rightarrow L$  monotonic.

The **Kleene iteration**  $(f^i(\perp))_{i \in \mathbb{N}}$  is an ascending chain:

$$\perp \sqsubseteq f(\perp) \sqsubseteq \dots \sqsubseteq f^i(\perp) \sqsubseteq f^{i+1}(\perp) \sqsubseteq \dots$$

For every  $k \in \mathbb{N}$ , if  $f^k(\perp) = f^{k+1}(\perp)$  then  $f^k(\perp)$  is the least fixpoint of  $f$ .

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LFP( $f : L \rightarrow L$ )
 $x \leftarrow \perp$ 
repeat
   $t \leftarrow x$ 
   $x \leftarrow f(x)$ 
until  $t = x$ 
return  $x$ 
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## Correction and termination

- 1 For every monotonic  $f$ , if LFP( $f$ ) terminates then it returns  $\text{lfp}(f)$ .
- 2 If  $L$  satisfies the ascending chain condition then LFP( $f$ ) always terminates (on monotonic  $f$ ).

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# Constructing Complete Lattices: Power Set

For any set  $S$ , the pair  $(\mathcal{P}(S), \sqsubseteq)$  is a complete lattice, where  $\sqsubseteq = \subseteq$ .

$\sqcap$ ,  $\sqcup$ ,  $\perp$  and  $\top$  satisfy:

$$\sqcap = \cap \qquad \perp = \emptyset$$

$$\sqcup = \cup \qquad \top = S$$

If  $S$  is finite then  $(\mathcal{P}(S), \sqsubseteq)$  satisfies the ascending chain condition.

# Constructing Complete Lattices: Functions

For any set  $S$  and complete lattice  $(L, \sqsubseteq)$ , the pair  $(S \rightarrow L, \sqsubseteq)$  is a complete lattice, where  $\sqsubseteq$  is defined by:

$$f \sqsubseteq g \quad \text{if} \quad f(x) \sqsubseteq g(x) \quad \text{for all } x \in S$$

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If  $S$  is finite and  $(L, \sqsubseteq)$  satisfies the ascending chain condition then  $(S \rightarrow L, \sqsubseteq)$  satisfies the ascending chain condition.

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Partial order  $\sqsubseteq$  compares the precision of data flow facts:

- $\phi \sqsubseteq \psi$  means that  $\phi$  is **more precise** than  $\psi$ .
- $\bigsqcup X$  is the **most precise** fact consistent with all facts  $\phi \in X$ .

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## Semantic Definition of Liveness

A variable  $x$  is **live** at location  $q$  if there **exists** an execution path starting from  $q$  where  $x$  is used before it is modified.

Consider a control flow automaton with variables  $X = \{x, y, z\}$ .

Complete lattice  $(L, \sqsubseteq)$  of data flow facts:  $(\mathcal{P}(X), \subseteq)$

The fact  $\{x, z\}$  means: *the variables that are live are **among**  $\{x, z\}$ .*  
i.e. *the variable  $y$  is **not live**.*

The fact  $\{x\}$  is **more precise** than  $\{x, z\}$ , but incomparable with  $\{y\}$ .

The fact  $\{x, z\}$  **soundly approximates** the fact  $\{x\}$ .

## Data Flow Instance

- Monotone framework  $\langle (L, \sqsubseteq), \mathcal{F} \rangle$
- Control flow automaton  $\langle Q, q_{in}, q_{out}, X, \rightarrow \rangle$
- Transfer mapping  $f : \text{Op} \rightarrow \mathcal{F}$
- Initial data flow value  $\iota \in L$

Notation for transfer mapping:  $f_{\text{op}}$  instead of  $f(\text{op})$

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# Data Flow Equations

Consider a data flow instance  $\langle (L, \sqsubseteq), \mathcal{F}, Q, q_{in}, q_{out}, X, \rightarrow, f, \iota \rangle$ .

System of equations: variables  $A_q$  for  $q \in Q$ , with  $A_q \in L$

## Forward Analysis

$$A_q = I_q \sqcup \bigsqcup_{q' \xrightarrow{\text{op}} q} f_{\text{op}}(A_{q'}) \quad I_q = \begin{cases} \iota & \text{if } q = q_{in} \\ \perp & \text{otherwise} \end{cases}$$

## Backward Analysis

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The system of data flow equations may have several solutions. . .

We are interested in the “least solution” to the data flow equations.

Complete lattice  $(L, \sqsubseteq)$  extended to  $(Q \rightarrow L, \sqsubseteq)$

The **forward minimal fixpoint solution**  $\overrightarrow{\text{MFP}}$  of the data flow instance is the **least fixpoint** of the monotonic function  $\overrightarrow{\Delta}$  on  $(Q \rightarrow L)$ :

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The **backward minimal fixpoint solution**  $\overleftarrow{\text{MFP}}$  of the data flow instance is the **least fixpoint** of the monotonic function  $\overleftarrow{\Delta}$  on  $(Q \rightarrow L)$ :

$$\overleftarrow{\Delta}(a) = \lambda q. \begin{cases} \perp \sqcup \bigsqcup_{q \xrightarrow{\text{op}} q'} f_{\text{op}}(a(q')) & \text{if } q = q_{\text{out}} \\ \bigsqcup_{q \xrightarrow{\text{op}} q'} f_{\text{op}}(a(q')) & \text{otherwise} \end{cases}$$

# Constraint-Based Formulation

Consider a data flow instance  $\langle (L, \sqsubseteq), \mathcal{F}, Q, q_{in}, q_{out}, X, \rightarrow, f, \iota \rangle$ .

Constraint system: variables  $A_q$  for  $q \in Q$ , with  $A_q \in L$

## Forward Analysis

$$\overrightarrow{(CS)} \quad \begin{cases} A_{q_{in}} \sqsupseteq \iota \\ A_{q'} \sqsupseteq f_{\text{op}}(A_q) \text{ for each } q \xrightarrow{\text{op}} q' \end{cases}$$

By Knaster-Tarski Fixpoint Theorem,

$$\overrightarrow{\text{MFP}} = \bigcap \left\{ a \in Q \rightarrow L \mid a \models \overrightarrow{(CS)} \right\}$$

Any solution to  $\overrightarrow{(CS)}$  is a sound approximation of  $\overrightarrow{\text{MFP}}$ .

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Constraint system: variables  $A_q$  for  $q \in Q$ , with  $A_q \in L$

## Backward Analysis

$$\overleftarrow{(CS)} \quad \begin{cases} A_{q_{out}} \sqsupseteq \iota \\ A_{q'} \sqsupseteq f_{op}(A_q) \text{ for each } q' \xrightarrow{op} q \end{cases}$$

By Knaster-Tarski Fixpoint Theorem,

$$\overleftarrow{MFP} = \bigcap \left\{ a \in Q \rightarrow L \mid a \models \overleftarrow{(CS)} \right\}$$

Any solution to  $\overleftarrow{(CS)}$  is a sound approximation of  $\overleftarrow{MFP}$ .

# Live Variables Analysis (Revisited)

Control Flow Automaton:  $\langle Q, q_{in}, q_{out}, X, \rightarrow \rangle$

## Monotone Framework

- Complete lattice  $(L, \sqsubseteq)$  of data flow facts:  $(\mathcal{P}(X), \subseteq)$
- Set  $\mathcal{F}$  of monotonic transfer functions:

$$\mathcal{F} = \{ \lambda \phi. gen \cup (\phi \setminus kill) \mid gen, kill \in L \}$$

## Data Flow Instance

- Initial data flow value:  $\emptyset$
- Transfer mapping:  $f_{op}(\phi) = Gen_{op} \cup (\phi \setminus Kill_{op})$

Backward analysis

# Available Expressions Analysis (Revisited)

Control Flow Automaton:  $\langle Q, q_{in}, q_{out}, X, \rightarrow \rangle$

## Monotone Framework

- Complete lattice  $(L, \sqsubseteq)$  of data flow facts:  $(\mathcal{P}(\text{SubExp}(\rightarrow)), \supseteq)$
- Set  $\mathcal{F}$  of monotonic transfer functions:

$$\mathcal{F} = \{ \lambda \phi. \text{gen} \cup (\phi \setminus \text{kill}) \mid \text{gen}, \text{kill} \in L \}$$

## Data Flow Instance

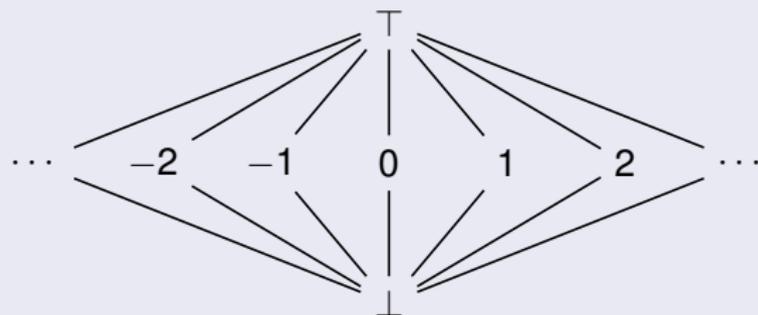
- Initial data flow value:  $\emptyset$
- Transfer mapping:  $f_{op}(\phi) = \text{Gen}_{op} \cup (\phi \setminus \text{Kill}_{op})$

Forward analysis

# Constant Propagation Analysis (Revisited)

Control Flow Automaton:  $\langle Q, q_{in}, q_{out}, X, \rightarrow \rangle$

## Constant Propagation Lattice for a Single Variable



$(\mathbb{R} \cup \{\perp, \top\}, \sqsubseteq)$

$\phi$	Meaning
$\top$	$\mathbb{R}$
$r \in \mathbb{R}$	$\{r\}$
$\perp$	$\emptyset$

## Monotone Framework

- Complete lattice  $(L, \sqsubseteq)$  of data flow facts:  $(X \rightarrow (\mathbb{R} \cup \{\perp, \top\}), \sqsubseteq)$
- Set  $\mathcal{F}$  defined as the set of all monotonic transfer functions on  $L$ .

# Constant Propagation Analysis (Revisited)

Control Flow Automaton:  $\langle Q, q_{in}, q_{out}, X, \rightarrow \rangle$

## Monotone Framework

- Complete lattice  $(L, \sqsubseteq)$  of data flow facts:  $(X \rightarrow (\mathbb{R} \cup \{\perp, \top\}), \sqsubseteq)$
- Set  $\mathcal{F}$  defined as the set of all monotonic transfer functions on  $L$ .

## Data Flow Instance

- Initial data flow value:  $\top$
- Transfer mapping:

$$f_{x:=e}(\phi) = \lambda y. \begin{cases} \phi(y) & \text{if } y \neq x \\ \llbracket e \rrbracket_{\phi} & \text{if } y = x \end{cases} \quad f_g(\phi) = \phi$$

Forward analysis

# Constant Propagation Analysis (Revisited)

Extension of  $\llbracket e \rrbracket$  to valuations in  $x \rightarrow (\mathbb{R} \cup \{\perp, T\})$

For  $r \in \mathbb{R} \cup \{T\}$

$$T + r = r + T = T$$

$$T - r = r - T = T$$

$$T \times r = r \times T = T$$

For  $r \in \mathbb{R} \cup \{\perp, T\}$

$$\perp + r = r + \perp = \perp$$

$$\perp - r = r - \perp = \perp$$

$$\perp \times r = r \times \perp = \perp$$

Expressions:  $\llbracket e \rrbracket_v$

$$\llbracket c \rrbracket_v = c \quad [c \in \mathbb{Q}]$$

$$\llbracket x \rrbracket_v = v(x) \quad [x \in X]$$

$$\llbracket e_1 + e_2 \rrbracket_v = \llbracket e_1 \rrbracket_v + \llbracket e_2 \rrbracket_v$$

$$\llbracket e_1 - e_2 \rrbracket_v = \llbracket e_1 \rrbracket_v - \llbracket e_2 \rrbracket_v$$

$$\llbracket e_1 * e_2 \rrbracket_v = \llbracket e_1 \rrbracket_v \times \llbracket e_2 \rrbracket_v$$

# Constant Propagation Analysis (Revisited)

Extension of  $\llbracket e \rrbracket$  to valuations in  $X \rightarrow (\mathbb{R} \cup \{\perp, T\})$

For  $r \in \mathbb{R} \cup \{T\}$

$$T + r = r + T = T$$

$$T - r = r - T = T$$

$$T \times r = r \times T = T$$

For  $r \in \mathbb{R} \cup \{\perp, T\}$

$$\perp + r = r + \perp = \perp$$

$$\perp - r = r - \perp = \perp$$

$$\perp \times r = r \times \perp = \perp$$

Expressions:  $\llbracket e \rrbracket_v$

$$\llbracket c \rrbracket_v = c \quad [c \in \mathbb{Q}]$$

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$$\llbracket e_1 + e_2 \rrbracket_v = \llbracket e_1 \rrbracket_v + \llbracket e_2 \rrbracket_v$$

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$$\llbracket e_1 * e_2 \rrbracket_v = \llbracket e_1 \rrbracket_v \times \llbracket e_2 \rrbracket_v$$

# (Forward) MFP Computation by Kleene Iteration

Consider a data flow instance  $\langle (L, \sqsubseteq), \mathcal{F}, Q, q_{in}, q_{out}, X, \rightarrow, f, \iota \rangle$ .

```
a ←  $\lambda q. \perp$ 
repeat
  b ← a
  a ←  $\overrightarrow{\Delta}(a)$ 
until b = a
return a
```

## Correction and termination

- 1 Returns  $\overrightarrow{\text{MFP}}$  when it terminates
- 2 Always terminates when  $(L, \sqsubseteq)$  satisfies the **ascending chain condition**

```
foreach  $q \in Q$ 
  a[q] ←  $\perp$ 
a[qin] ←  $\iota$ 
repeat
  foreach  $q \in Q$ 
    b[q] ← a[q]
  foreach  $q \in Q$ 
    a[q] ←  $\bigsqcup_{q' \xrightarrow{\text{op}} q} f_{\text{op}}(b[q'])$ 
until  $(\forall q \in Q \cdot b[q] = a[q])$ 
return a
```

# (Forward) MFP Computation by Kleene Iteration

Consider a data flow instance  $\langle (L, \sqsubseteq), \mathcal{F}, Q, q_{in}, q_{out}, X, \rightarrow, f, \iota \rangle$ .

```
a ← λq. ⊥  
repeat  
  b ← a  
  a ←  $\overrightarrow{\Delta}(a)$   
until b = a  
return a
```

## Correction and termination

- 1 Returns  $\overrightarrow{\text{MFP}}$  when it terminates
- 2 Always terminates when  $(L, \sqsubseteq)$  satisfies the **ascending chain condition**

```
foreach q ∈ Q  
  a[q] ← ⊥  
a[qin] ← ι  
repeat  
  foreach q ∈ Q  
    b[q] ← a[q]  
  foreach q ∈ Q  
    a[q] ←  $\bigsqcup_{q' \xrightarrow{\text{op}} q} f_{\text{op}}(b[q'])$   
until (∀ q ∈ Q · b[q] = a[q])  
return a
```

# (Forward) MFP Computation by Kleene Iteration

Consider a data flow instance  $\langle (L, \sqsubseteq), \mathcal{F}, Q, q_{in}, q_{out}, X, \rightarrow, f, \iota \rangle$ .

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  foreach q ∈ Q  
    a[q] ←  $\bigsqcup_{q' \xrightarrow{\text{op}} q} f_{\text{op}}(b[q'])$   
until (∀ q ∈ Q · b[q] = a[q])  
return a
```

We can improve!

# (Forward) MFP Computation by Round-Robin Iteration

Consider a data flow instance  $\langle (L, \sqsubseteq), \mathcal{F}, Q, q_{in}, q_{out}, X, \rightarrow, f, \iota \rangle$ .

```
foreach  $q \in Q$ 
   $a[q] \leftarrow \perp$ 
 $a[q_{in}] \leftarrow \iota$ 
do
   $change \leftarrow false$ 
  foreach  $q \xrightarrow{op} q'$ 
     $new \leftarrow f_{op}(a[q])$ 
    if  $new \not\sqsubseteq a[q']$ 
       $a[q'] \leftarrow a[q'] \sqcup new$ 
       $change \leftarrow true$ 
while  $change$ 
return  $a$ 
```

The foreach loop iterates over transitions in  $\rightarrow$ .

Propagation of facts

- benefits from previous propagations
- records whether there was a change

Correct and always faster than Kleene iteration

# (Forward) MFP Computation by Round-Robin Iteration

Consider a data flow instance  $\langle (L, \sqsubseteq), \mathcal{F}, Q, q_{in}, q_{out}, X, \rightarrow, f, \iota \rangle$ .

```
foreach  $q \in Q$ 
   $a[q] \leftarrow \perp$ 
 $a[q_{in}] \leftarrow \iota$ 
do
  change  $\leftarrow$  false
  foreach  $q \xrightarrow{op} q'$ 
    new  $\leftarrow f_{op}(a[q])$ 
    if new  $\not\sqsubseteq a[q']$ 
       $a[q'] \leftarrow a[q'] \sqcup$  new
      change  $\leftarrow$  true
while change
return a
```

The foreach loop iterates over transitions in  $\rightarrow$ .

Propagation of facts

- benefits from previous propagations
- records whether there was a change

**Correct** and always **faster** than Kleene iteration

# (Forward) MFP Computation by Worklist Iteration

```
wl ← nil
foreach  $q' \xrightarrow{\text{op}} q$ 
  wl ← cons((q, op, q'), wl)
foreach  $q \in Q$ 
  a[q] ←  $\perp$ 
a[qin] ←  $\iota$ 
while wl ≠ nil
  (q, op, q') ← head(wl)
  wl ← tail(wl)
  new ← fop(a[q])
  if new  $\not\sqsubseteq$  a[q']
    a[q'] ← a[q]  $\sqcup$  new
    foreach  $q' \xrightarrow{\text{op}'}$  q''
      wl ← cons((q', op', q''), wl)
return a
```

## Vs Round-Robin

☺ Less computations

☹ Overhead

## Worklist structures

• LIFO

• FIFO

• Set

• ...

# (Forward) MFP Computation by Worklist Iteration

```
wl ← nil
foreach  $q' \xrightarrow{\text{op}} q$ 
  wl ← cons((q, op, q'), wl)
foreach  $q \in Q$ 
  a[q] ←  $\perp$ 
a[qin] ←  $\iota$ 
while wl ≠ nil
  (q, op, q') ← head(wl)
  wl ← tail(wl)
  new ← fop(a[q])
  if new  $\not\sqsubseteq$  a[q']
    a[q'] ← a[q]  $\sqcup$  new
    foreach  $q' \xrightarrow{\text{op}'} q''$ 
      wl ← cons((q', op', q''), wl)
return a
```

## Vs Round-Robin

😊 Less computations

😞 Overhead

## Worklist structures

- LIFO
- FIFO
- Set
- ...

# Optimization of MFP Computation with SCCs

- 1 Decompose control flow automaton into strongly connected components
- 2 Transitions between SCCs induce a partial order between SCCs
- 3 Compute the MFP solution component after component, following the partial order between SCCs

This optimization often pays off in practice

Further optimizations are possible...

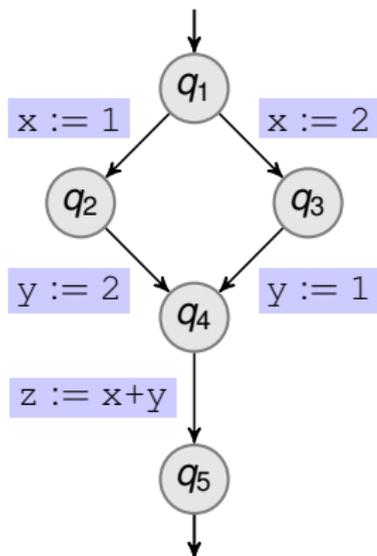
# Optimization of MFP Computation with SCCs

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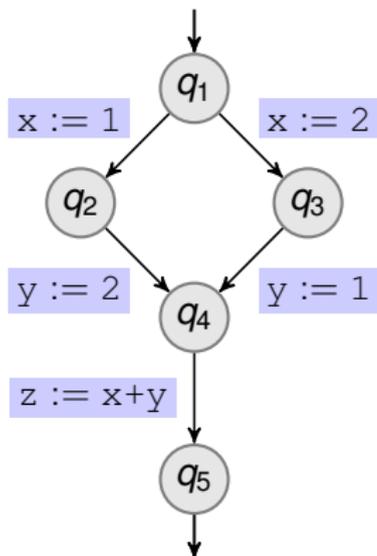
Further optimizations are possible...

# Loss of Precision with the MFP Solution



At  $q_5$ , we have  $z = 3$

# Loss of Precision with the MFP Solution



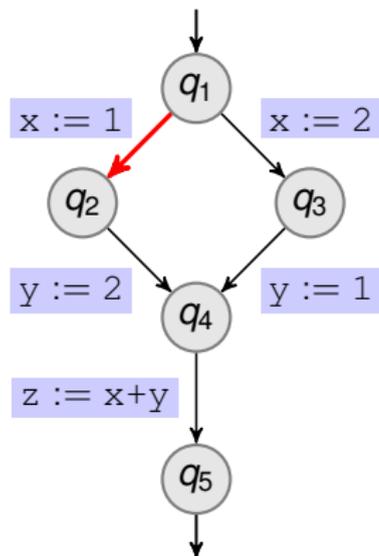
	x	y	z
$q_1$	$\top$	$\top$	$\top$
$q_2$	$\perp$	$\perp$	$\perp$
$q_3$	$\perp$	$\perp$	$\perp$
$q_4$	$\perp$	$\perp$	$\perp$
$q_5$	$\perp$	$\perp$	$\perp$

At  $q_5$ , we have  $z = 3$

## Loss of Precision

Cause: application of  $\sqcup$  at  $q_4$  to merge data flow information

# Loss of Precision with the MFP Solution



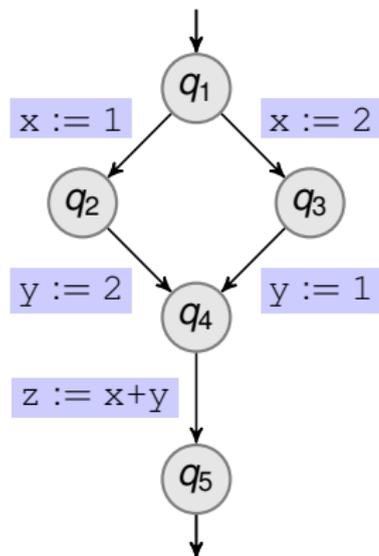
	x	y	z
$q_1$	$\top$	$\top$	$\top$
$q_2$	<b>1</b>	$\top$	$\top$
$q_3$	$\perp$	$\perp$	$\perp$
$q_4$	$\perp$	$\perp$	$\perp$
$q_5$	$\perp$	$\perp$	$\perp$

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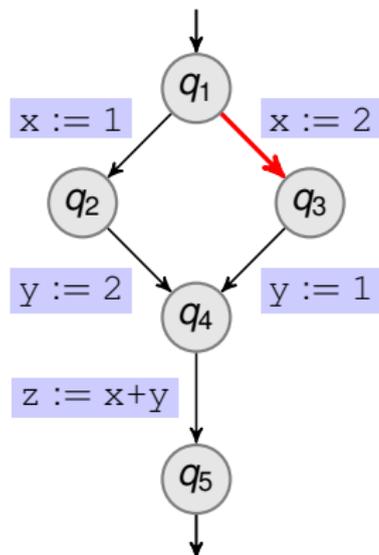
	x	y	z
$q_1$	$\top$	$\top$	$\top$
$q_2$	1	$\top$	$\top$
$q_3$	$\perp$	$\perp$	$\perp$
$q_4$	$\perp$	$\perp$	$\perp$
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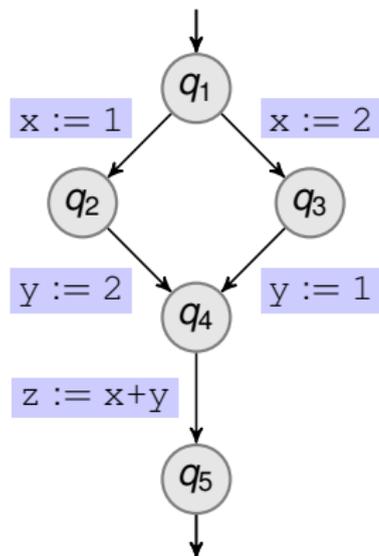
	x	y	z
$q_1$	$\top$	$\top$	$\top$
$q_2$	1	$\top$	$\top$
$q_3$	2	$\top$	$\top$
$q_4$	$\perp$	$\perp$	$\perp$
$q_5$	$\perp$	$\perp$	$\perp$

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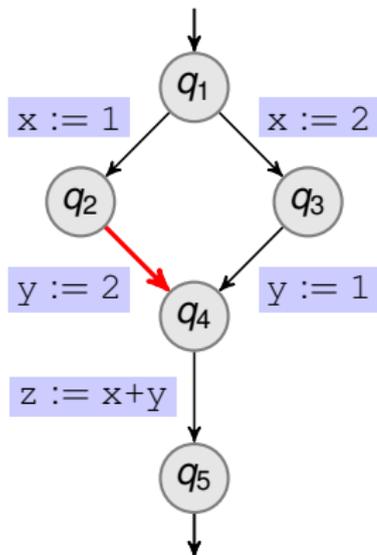
	x	y	z
$q_1$	$\top$	$\top$	$\top$
$q_2$	1	$\top$	$\top$
$q_3$	2	$\top$	$\top$
$q_4$	$\perp$	$\perp$	$\perp$
$q_5$	$\perp$	$\perp$	$\perp$

At  $q_5$ , we have  $z = 3$

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# Loss of Precision with the MFP Solution

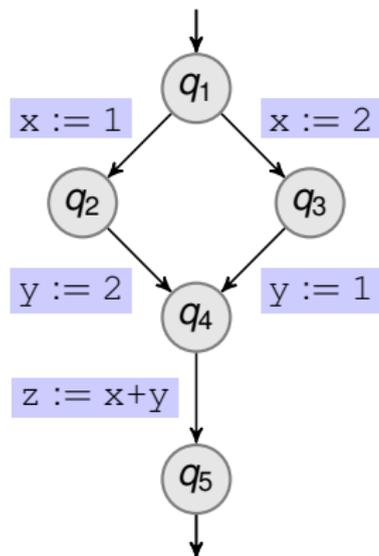


	x	y	z
$q_1$	$\top$	$\top$	$\top$
$q_2$	1	$\top$	$\top$
$q_3$	2	$\top$	$\top$
$q_4$	1	2	$\top$
$q_5$	$\perp$	$\perp$	$\perp$

At  $q_5$ , we have  $z = 3$

Loss of Precision  
Cause: application of  $\sqcup$  at  $q_4$  to merge data flow information

# Loss of Precision with the MFP Solution



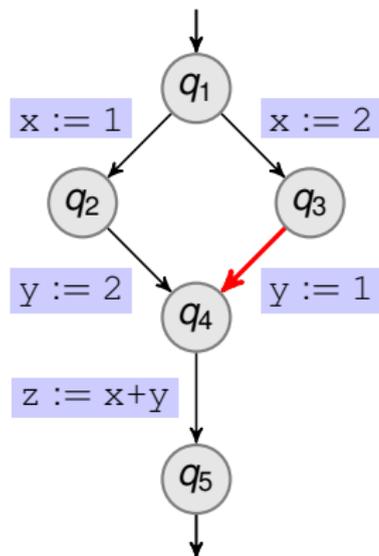
	x	y	z
$q_1$	$\top$	$\top$	$\top$
$q_2$	1	$\top$	$\top$
$q_3$	2	$\top$	$\top$
$q_4$	1	2	$\top$
$q_5$	$\perp$	$\perp$	$\perp$

At  $q_5$ , we have  $z = 3$

## Loss of Precision

Cause: application of  $\sqcup$  at  $q_4$  to merge data flow information

# Loss of Precision with the MFP Solution



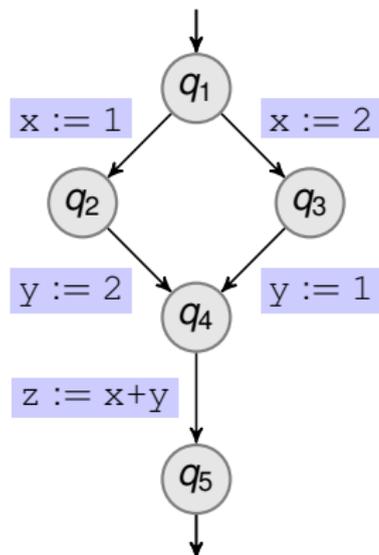
	x	y	z
$q_1$	$\top$	$\top$	$\top$
$q_2$	1	$\top$	$\top$
$q_3$	2	$\top$	$\top$
$q_4$	$1 \sqcup 2$	$2 \sqcup 1$	$\top$
$q_5$	$\perp$	$\perp$	$\perp$

At  $q_5$ , we have  $z = 3$

## Loss of Precision

Cause: application of  $\sqcup$  at  $q_4$  to merge data flow information

# Loss of Precision with the MFP Solution



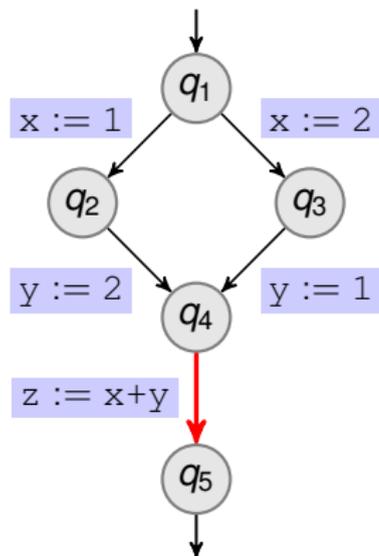
	x	y	z
$q_1$	$\top$	$\top$	$\top$
$q_2$	1	$\top$	$\top$
$q_3$	2	$\top$	$\top$
$q_4$	$\top$	$\top$	$\top$
$q_5$	$\perp$	$\perp$	$\perp$

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# Loss of Precision with the MFP Solution



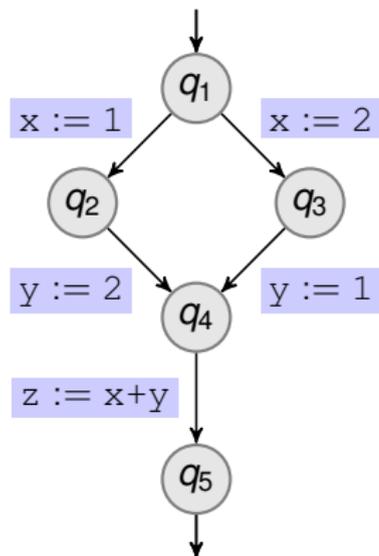
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$q_3$	2	$\top$	$\top$
$q_4$	$\top$	$\top$	$\top$
$q_5$	$\top$	$\top$	$\top$

At  $q_5$ , we have  $z = 3$

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# Loss of Precision with the MFP Solution



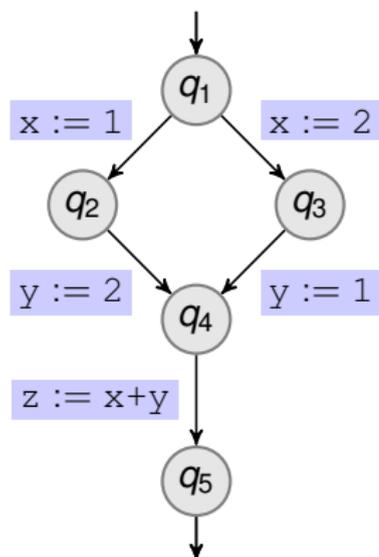
	x	y	z
$q_1$	$\top$	$\top$	$\top$
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$q_3$	2	$\top$	$\top$
$q_4$	$\top$	$\top$	$\top$
$q_5$	$\top$	$\top$	$\top$

At  $q_5$ , we have  $z = 3$

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Cause: application of  $\sqcup$  at  $q_4$  to merge data flow information

# Loss of Precision with the MFP Solution



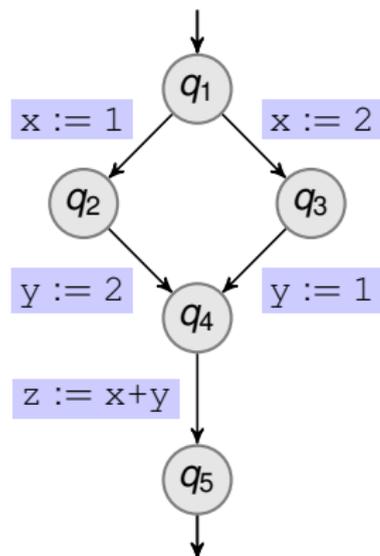
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$q_1$	$\top$	$\top$	$\top$
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$q_4$	$\top$	$\top$	$\top$
$q_5$	$\top$	$\top$	$\top$

At  $q_5$ , we have  $z = 3$

## Loss of Precision

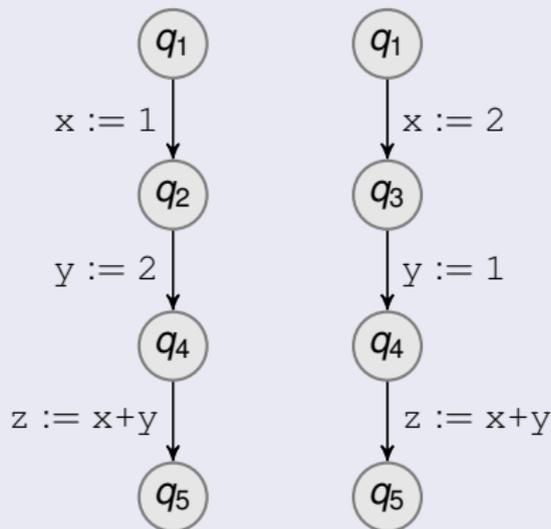
Cause: application of  $\sqcup$  at  $q_4$  to merge data flow information

# Alternative Approach for Better Precision

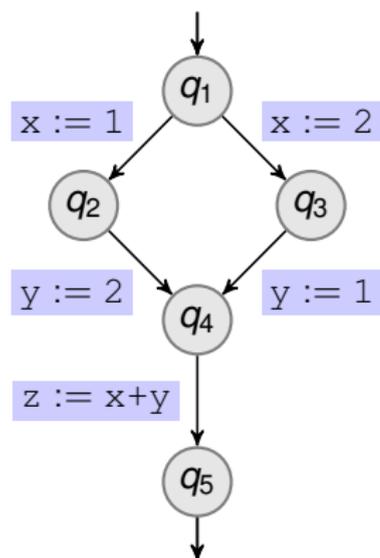


At  $q_5$ , we have  $z = 3$

## Control Paths from $q_1$ to $q_5$

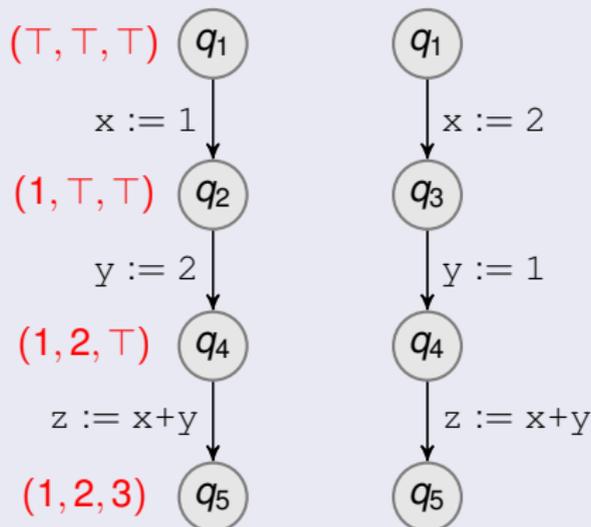


# Alternative Approach for Better Precision

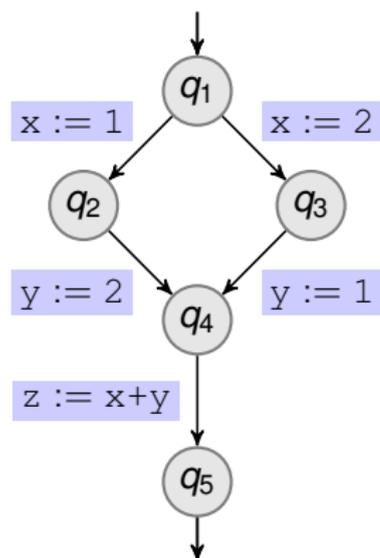


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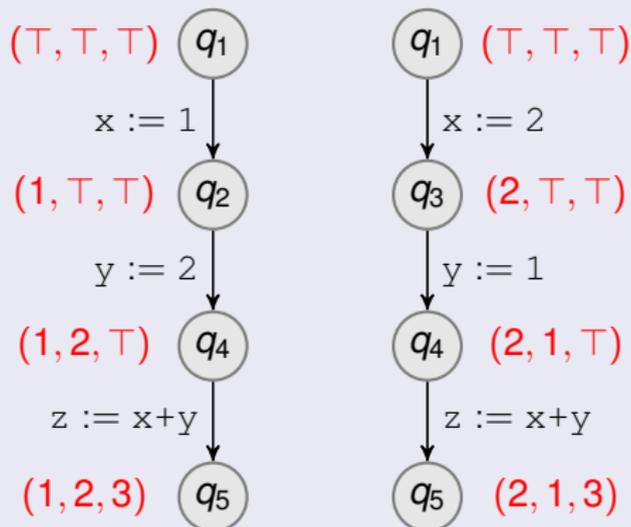


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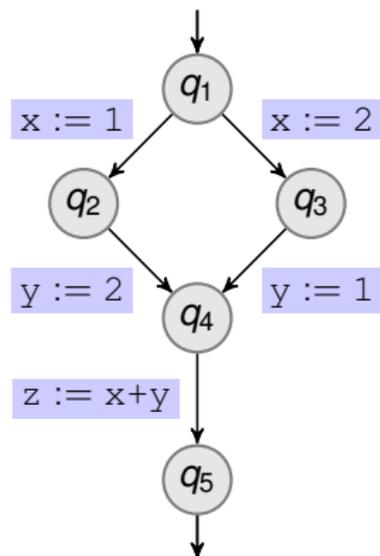


At  $q_5$ , we have  $z = 3$

## Control Paths from $q_1$ to $q_5$

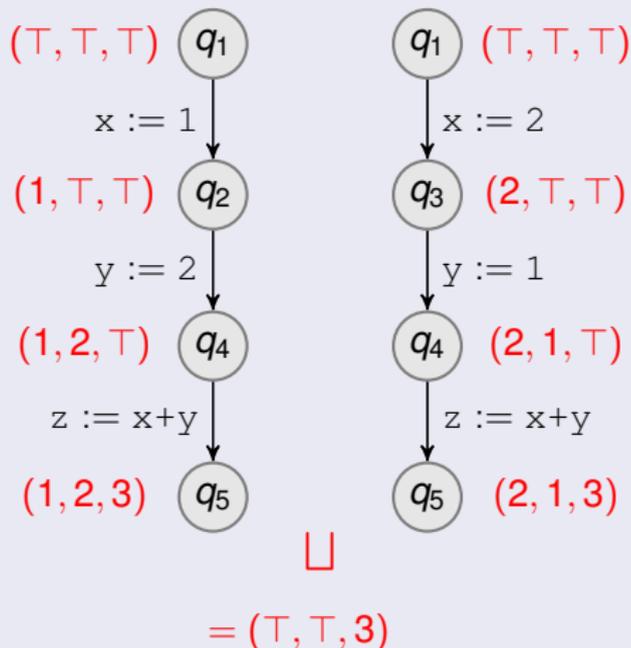


# Alternative Approach for Better Precision



At  $q_5$ , we have  $z = 3$

## Control Paths from $q_1$ to $q_5$



# Meet Over All Paths (MOP) Solution

Consider a data flow instance  $\langle (L, \sqsubseteq), \mathcal{F}, Q, q_{in}, q_{out}, X, \rightarrow, f, \iota \rangle$ .

## Forward Meet Over All Paths Solution

$$\overrightarrow{\text{MOP}} = \lambda q. \bigsqcup \left\{ f_{\text{op}_k} \circ \dots \circ f_{\text{op}_0}(\iota) \mid q_{in} \xrightarrow{\text{op}_0} q_1 \dots q_k \xrightarrow{\text{op}_k} q \right\}$$

## Backward Meet Over All Paths Solution

$$\overleftarrow{\text{MOP}} = \lambda q. \bigsqcup \left\{ f_{\text{op}_0} \circ \dots \circ f_{\text{op}_k}(\iota) \mid q \xrightarrow{\text{op}_0} q_1 \dots q_k \xrightarrow{\text{op}_k} q_{out} \right\}$$

More precise than MFP

$$\begin{aligned} \overrightarrow{\text{MOP}} &\sqsubseteq \overrightarrow{\text{MFP}} \\ \overleftarrow{\text{MOP}} &\sqsubseteq \overleftarrow{\text{MFP}} \end{aligned}$$

Not Computable in General

$\overrightarrow{\text{MOP}}(q) \stackrel{?}{=} 1$  is **undecidable** for constant propagation

# Meet Over All Paths (MOP) Solution

Consider a data flow instance  $\langle (L, \sqsubseteq), \mathcal{F}, Q, q_{in}, q_{out}, X, \rightarrow, f, \iota \rangle$ .

## Forward Meet Over All Paths Solution

$$\overrightarrow{\text{MOP}} = \lambda q. \bigsqcup \left\{ f_{\text{op}_k} \circ \dots \circ f_{\text{op}_0}(\iota) \mid q_{in} \xrightarrow{\text{op}_0} q_1 \dots q_k \xrightarrow{\text{op}_k} q \right\}$$

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# MOP = MFP in Distributive Frameworks

A monotone framework  $\langle (L, \sqsubseteq), \mathcal{F} \rangle$  is **distributive** if every  $f \in \mathcal{F}$  is **completely additive**:

$$f(\bigsqcup X) = \bigsqcup \{f(\phi) \mid \phi \in X\} \quad (\text{for all } X \subseteq L)$$

## Theorem

*For any data flow instance over a distributive monotone framework,*

$$\begin{aligned} \overrightarrow{\text{MOP}} &= \overrightarrow{\text{MFP}} \\ \overleftarrow{\text{MOP}} &= \overleftarrow{\text{MFP}} \end{aligned}$$

## Intuition

In a distributive framework, applying  $\bigsqcup$  “early” does not lose precision:

$$f_{\text{OP}_5} (f_{\text{OP}_2}(\phi) \bigsqcup f_{\text{OP}_3}(\psi)) = f_{\text{OP}_5} \circ f_{\text{OP}_2}(\phi) \bigsqcup f_{\text{OP}_5} \circ f_{\text{OP}_3}(\psi)$$

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# Examples of Distributive Monotone Frameworks

## Gen / Kill Monotone Frameworks

- Complete lattice  $(L, \sqsubseteq)$  of data flow facts:

$$L = \mathcal{P}(S) \text{ for some set } S \qquad \sqsubseteq \text{ is } \subseteq \text{ or } \supseteq$$

- Set  $\mathcal{F}$  of monotonic transfer functions:

$$\mathcal{F} = \{ \lambda \phi. \mathit{gen} \cup (\phi \setminus \mathit{kill}) \mid \mathit{gen}, \mathit{kill} \in L \}$$

All gen/kill monotone frameworks are distributive

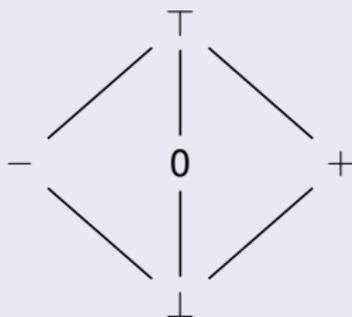
## Examples

- Live Variables
- Available Expressions
- Uninitialized Variables
- ...

# Sign Analysis: Monotone Framework

Control Flow Automaton:  $\langle Q, q_{in}, q_{out}, X, \rightarrow \rangle$

(Simplified) Sign Lattice for a Single Variable:  $(Sign, \sqsubseteq)$



$\phi$	Meaning
$\top$	$\mathbb{R}$
$-$	$\{r \in \mathbb{R} \mid r < 0\}$
$+$	$\{r \in \mathbb{R} \mid r > 0\}$
$0$	$\{0\}$
$\perp$	$\emptyset$

## Monotone Framework

- Complete lattice  $(L, \sqsubseteq)$  of data flow facts:  $(X \rightarrow Sign, \sqsubseteq)$
- Set  $\mathcal{F}$  defined as the set of all monotonic transfer functions on  $L$ .

# Sign Analysis: Data Flow Instance

Control Flow Automaton:  $\langle Q, q_{in}, q_{out}, X, \rightarrow \rangle$

## Monotone Framework

- Complete lattice  $(L, \sqsubseteq)$  of data flow facts:  $(x \rightarrow \text{Sign}, \sqsubseteq)$
- Set  $\mathcal{F}$  defined as the set of all monotonic transfer functions on  $L$ .

## Data Flow Instance

- Initial data flow value:  $\top$
- Transfer mapping:

$$f_{x:=e}(\phi) = \lambda y. \begin{cases} \phi(y) & \text{if } y \neq x \\ \llbracket e \rrbracket_{\phi} & \text{if } y = x \end{cases} \quad f_g(\phi) = \phi$$

Forward analysis

# Sign Analysis: Transfer Mapping

Need to define  $\llbracket e \rrbracket$  for valuations  $v$  in  $X \rightarrow \{-, 0, +, \perp, \top\}$

Expressions:  $\llbracket e \rrbracket_v$

$$\llbracket c \rrbracket_v = \text{sign}(c) \quad [c \in \mathbb{Q}]$$

$$\llbracket x \rrbracket_v = v(x) \quad [x \in X]$$

$$\llbracket e_1 + e_2 \rrbracket_v = \llbracket e_1 \rrbracket_v \oplus \llbracket e_2 \rrbracket_v$$

$$\llbracket e_1 - e_2 \rrbracket_v = \llbracket e_1 \rrbracket_v \ominus \llbracket e_2 \rrbracket_v$$

$$\llbracket e_1 * e_2 \rrbracket_v = \llbracket e_1 \rrbracket_v \otimes \llbracket e_2 \rrbracket_v$$

$$\text{sign}(c) = \begin{cases} - & \text{if } c < 0 \\ 0 & \text{if } c = 0 \\ + & \text{if } c > 0 \end{cases}$$

“Abstract” Addition

$\oplus$	$\perp$	$-$	$0$	$+$	$\top$
$\perp$	$\perp$	$\perp$	$\perp$	$\perp$	$\perp$
$-$	$\perp$	$-$	$-$	$\top$	$\top$
$0$	$\perp$	$-$	$0$	$+$	$\top$
$+$	$\perp$	$\top$	$+$	$+$	$\top$
$\top$	$\perp$	$\top$	$\top$	$\top$	$\top$

Tables also required for:

- “abstract” subtraction
- “abstract” multiplication

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“Abstract” Addition

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$\perp$	$\perp$	$\perp$	$\perp$	$\perp$	$\perp$
$-$	$\perp$	$-$	$-$	$\top$	$\top$
$0$	$\perp$	$-$	$0$	$+$	$\top$
$+$	$\perp$	$\top$	$+$	$+$	$\top$
$\top$	$\perp$	$\top$	$\top$	$\top$	$\top$

Tables also required for:

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# Sign Analysis: Transfer Mapping

Need to define  $\llbracket e \rrbracket_v$  for evaluations  $v$  in  $X \rightarrow \{-, 0, +, \perp, T\}$

Expressions:  $\llbracket e \rrbracket_v$

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Are these tables correct?

“abstract” Addition

		0	+	T
$\perp$				$\perp$
-	$\perp$			
0	$\perp$	-	0	
+	$\perp$	T	+	+
T	$\perp$	T	T	T

Tables also required for:

- “abstract” subtraction
- “abstract” multiplication

# Sign Analysis: Transfer Mapping

Need to define  $\llbracket e \rrbracket_v$  for evaluations  $v$  in  $X \rightarrow \{-, 0, +, \perp, T\}$

Expressions:  $\llbracket e \rrbracket_v$  "subtract" Addition

$$\llbracket c \rrbracket_v = \text{sign}(c) \quad [c \in \mathbb{Q}]$$

$$\llbracket x \rrbracket_v =$$

	0	+	T
			$\perp$

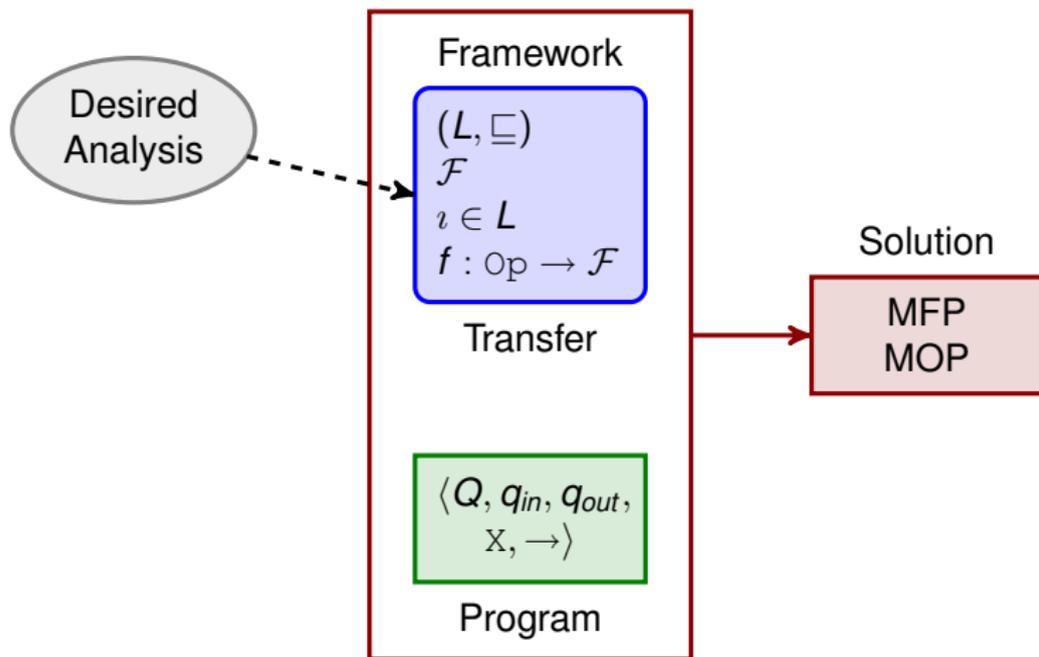
Are these tables correct?

*Does this data flow instance really perform **sign analysis**?*

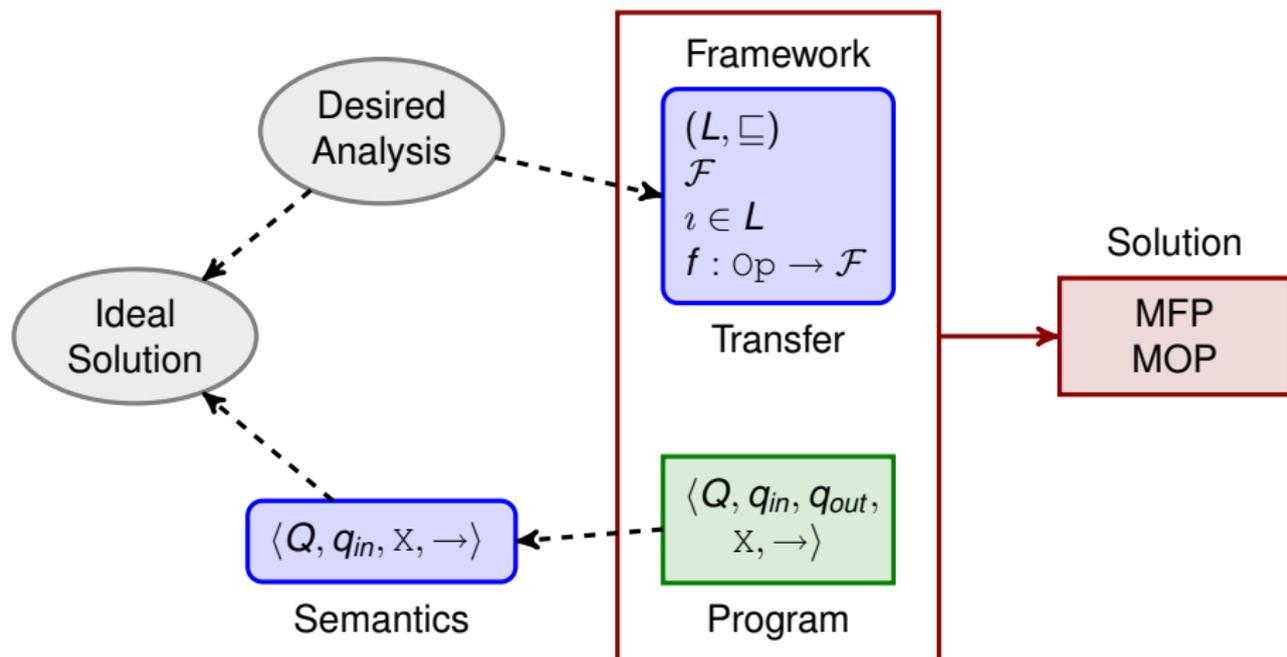
*Is the analysis **correct**?*

*Is it **precise**?*

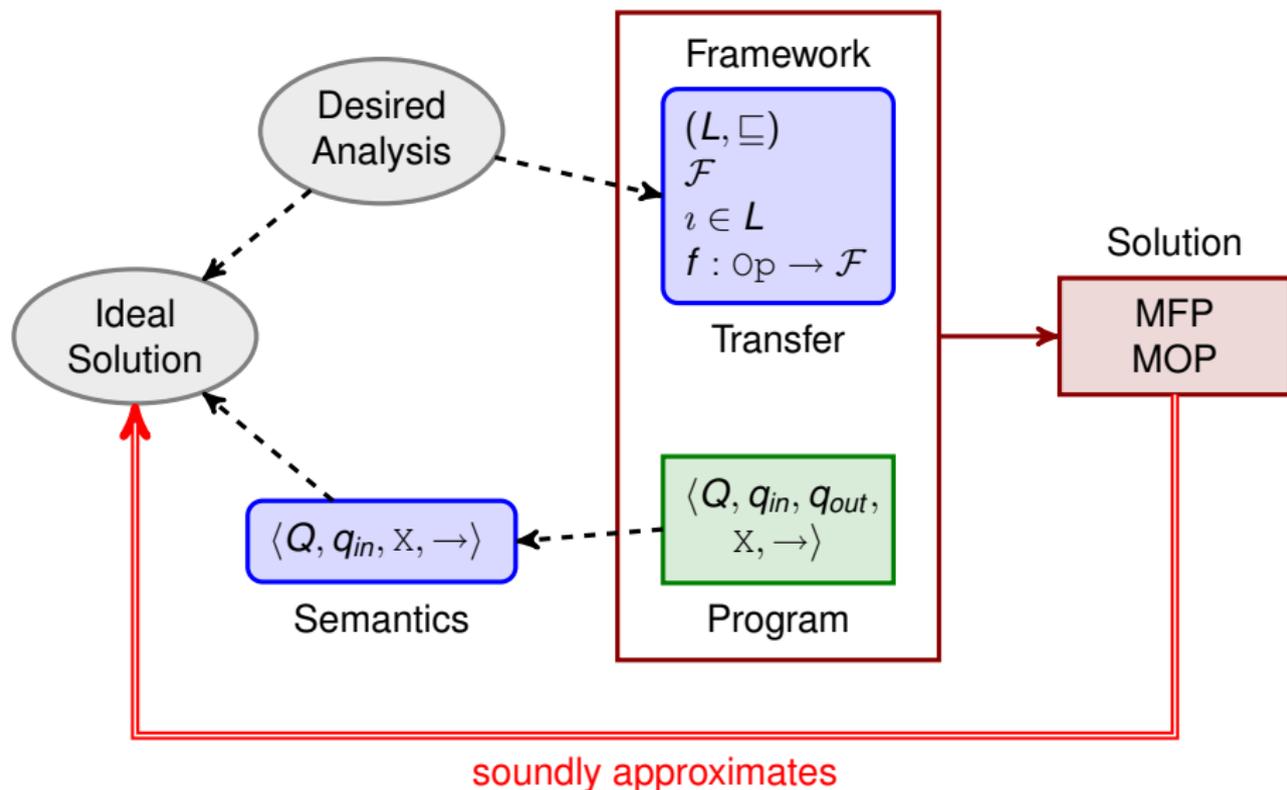
# What About Correctness of Data Flow Analyses?



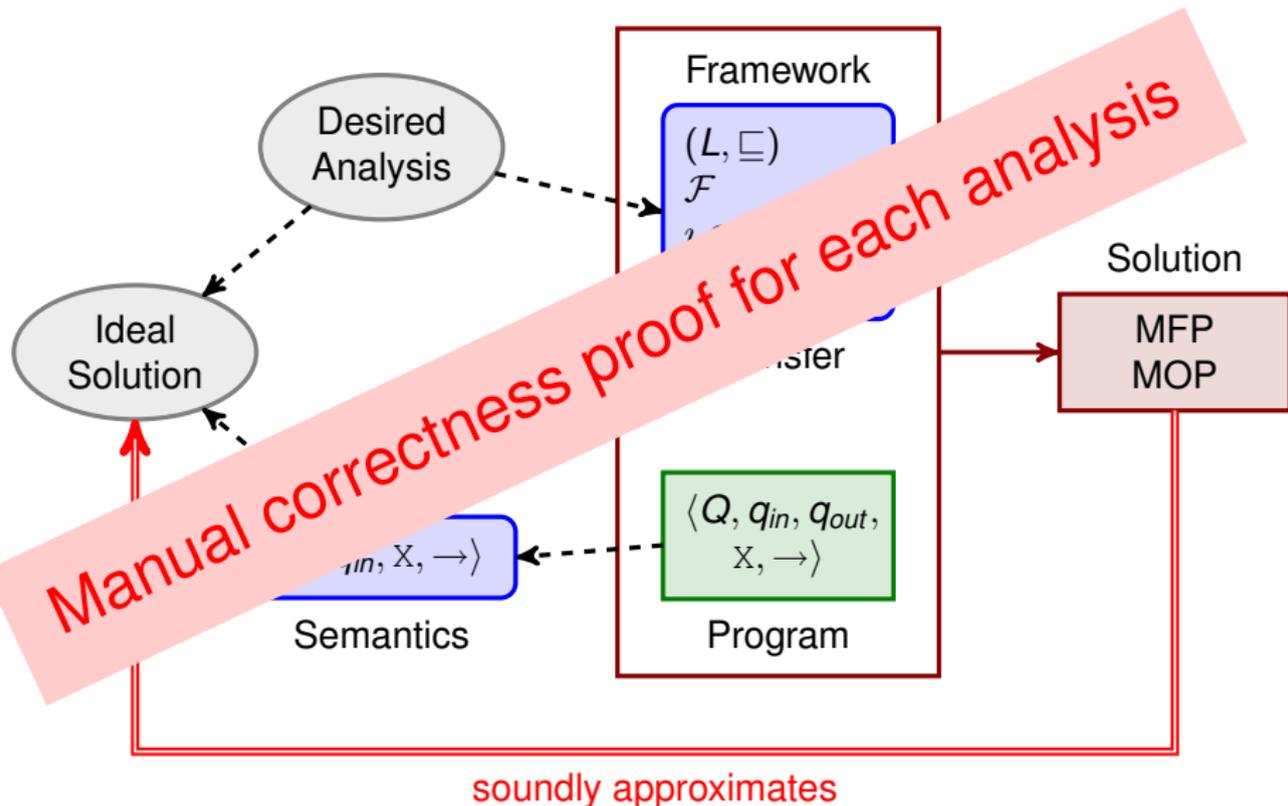
# What About Correctness of Data Flow Analyses?



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# How to Systematically Ensure Correctness?

Data flow facts have an **intended meaning**.

The transfer mapping is designed according to this **intended meaning**.

We need a formal link to relate data flow facts and transfer functions with the **formal semantics**.

Solution: Abstract Interpretation

« *This paper is devoted to the **systematic** and **correct** design of program analysis frameworks with respect to a **formal semantics**.* »

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## Part IV

# Abstract Interpretation

- 8 Some More Lattice Theory: Galois Connections
- 9 Abstract Interpretation-Based Data Flow Analysis
- 10 Convergence Acceleration with Widening and Narrowing

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# Concrete Lattice & Abstract Lattice: Notations

Concrete lattice

$(L, \sqsubseteq)$

Abstract lattice

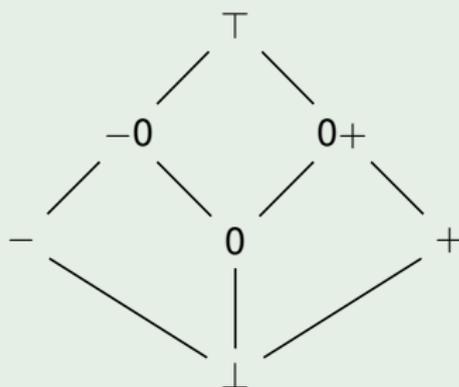
$(\bar{L}, \bar{\sqsubseteq})$

Example (Sets of Values)

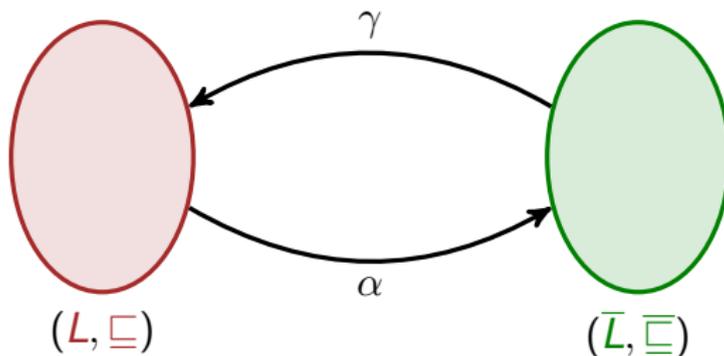
For a variable ranging over a domain  $\mathbb{D}$ :

$(\mathcal{P}(\mathbb{D}), \subseteq)$

Example (Sign Lattice)



# Galois Connections: Definition



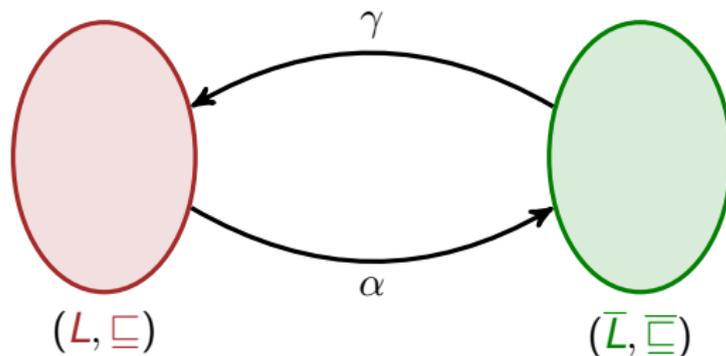
## Definition

A **Galois connection** between a lattice  $(L, \subseteq)$  and a lattice  $(\bar{L}, \bar{\subseteq})$  is a pair of functions  $(\alpha, \gamma)$ , with  $\alpha : L \rightarrow \bar{L}$  and  $\gamma : \bar{L} \rightarrow L$ , satisfying:

$$\alpha(x) \bar{\subseteq} \bar{y} \quad \text{iff} \quad x \subseteq \gamma(\bar{y}) \quad (\text{for all } x \in L, \bar{y} \in \bar{L})$$

Notation for Galois connections:  $(L, \subseteq) \xleftrightarrow[\alpha]{\gamma} (\bar{L}, \bar{\subseteq})$

# Galois Connections: Definition



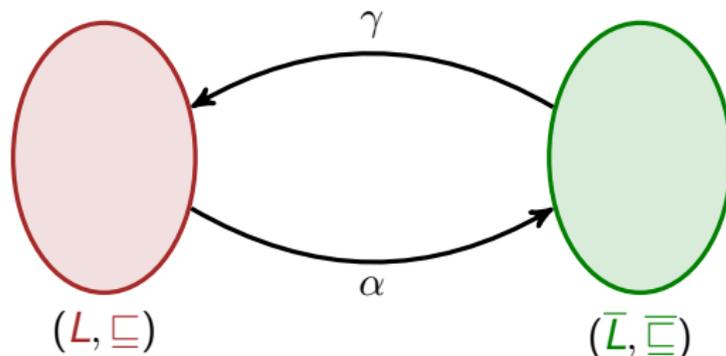
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# Galois Connections: Intuition



## Concretization

$\gamma$  is the **concretization** function.

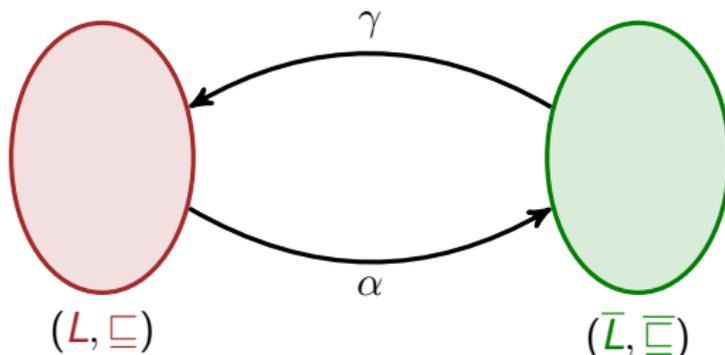
$\gamma(\bar{y})$  is the concrete value in  $L$  that is **represented** by  $\bar{y}$ .

## Abstraction

$\alpha$  is the **abstraction** function.

$\alpha(x)$  is the **most precise** abstract value in  $\bar{L}$  whose concretization **approximates**  $x$ .

# Galois Connections: Intuition



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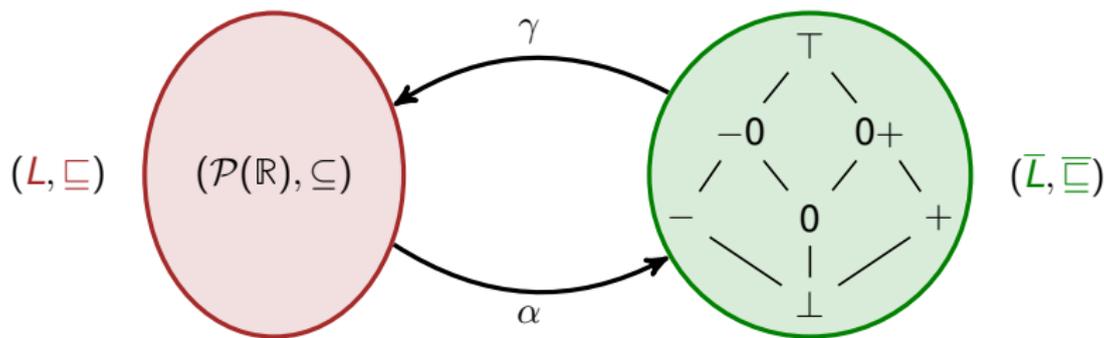
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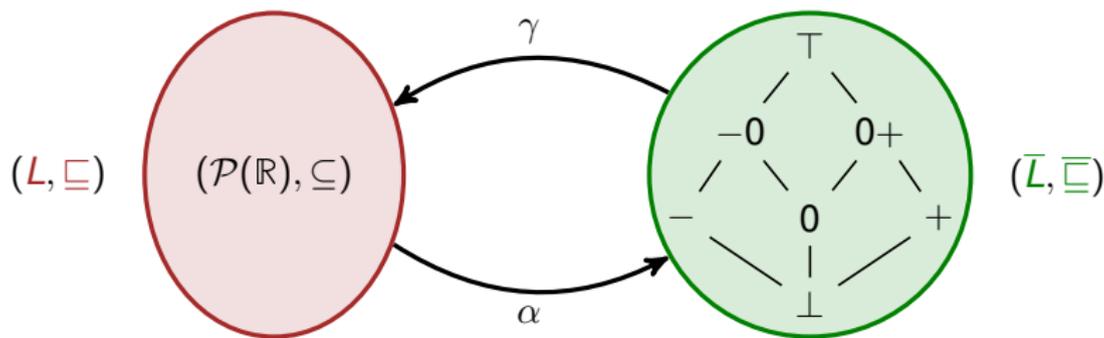
# Galois Connections: Example



$$\alpha(x) = \begin{cases} \perp & \text{if } x = \emptyset \\ - & \text{if } x \subseteq \{r \in \mathbb{R} \mid r < 0\} \\ 0 & \text{if } x = \{0\} \\ + & \text{if } x \subseteq \{r \in \mathbb{R} \mid r > 0\} \\ -0 & \text{if } \{0\} \subset x \subseteq \{r \in \mathbb{R} \mid r \leq 0\} \\ 0+ & \text{if } \{0\} \subset x \subseteq \{r \in \mathbb{R} \mid r \geq 0\} \\ \top & \text{otherwise} \end{cases}$$

$\bar{y}$	$\gamma(\bar{y})$
$\perp$	$\emptyset$
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$\top$	$\mathbb{R}$

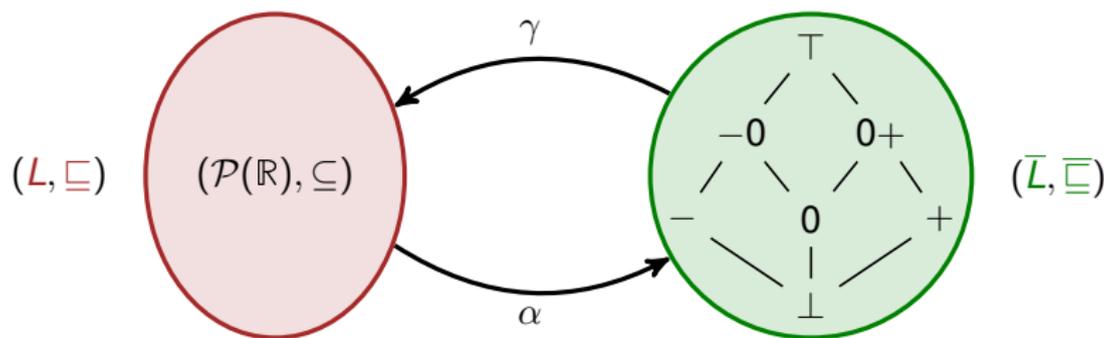
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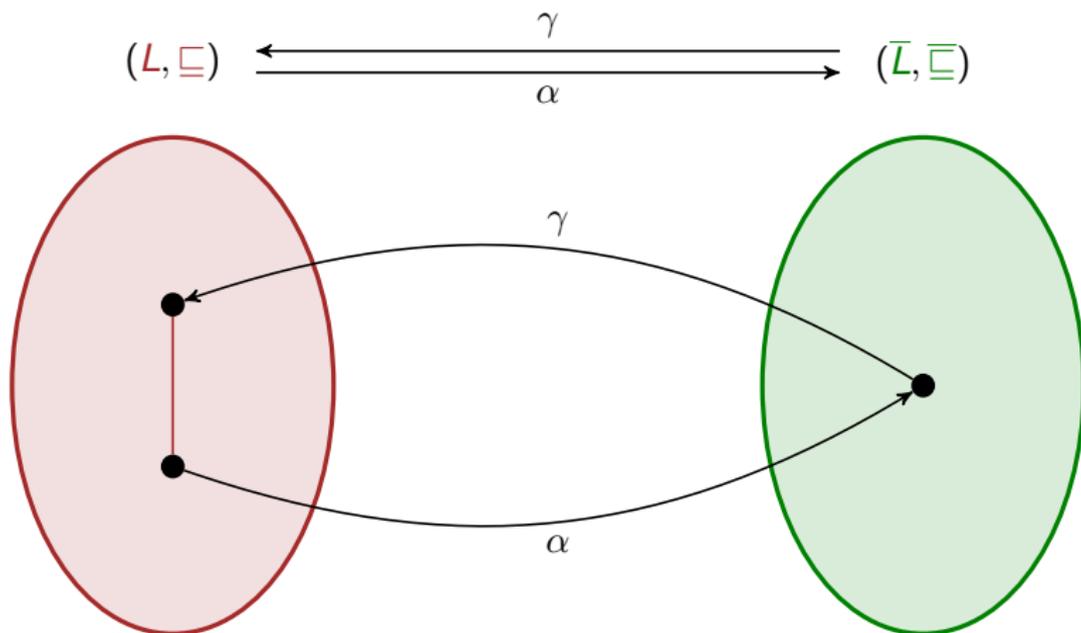
# Galois Connections: Characterization

Consider two lattices  $(L, \sqsubseteq)$  and  $(\bar{L}, \bar{\sqsubseteq})$ .

For any two functions  $\alpha : L \rightarrow \bar{L}$  et  $\gamma : \bar{L} \rightarrow L$ , we have

$$(L, \sqsubseteq) \begin{array}{c} \xleftarrow{\gamma} \\ \xrightarrow{\alpha} \end{array} (\bar{L}, \bar{\sqsubseteq}) \quad \text{iff} \quad \left\{ \begin{array}{ll} x \sqsubseteq \gamma \circ \alpha(x) & \text{(for all } x \in L) \\ \alpha \circ \gamma(\bar{y}) \bar{\sqsubseteq} \bar{y} & \text{(for all } \bar{y} \in \bar{L}) \\ \alpha \text{ is monotonic} & \\ \gamma \text{ is monotonic} & \end{array} \right.$$

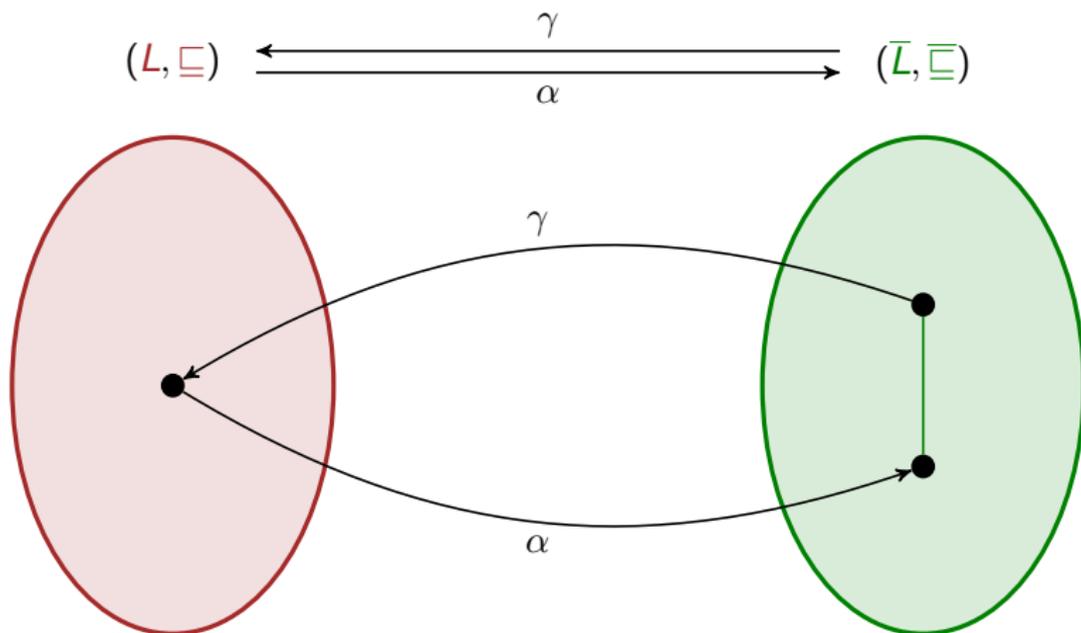
# Galois Connections: Characterization



$$x \sqsubseteq \gamma \circ \alpha(x)$$

$(\gamma \circ \alpha \text{ extensive})$

# Galois Connections: Characterization

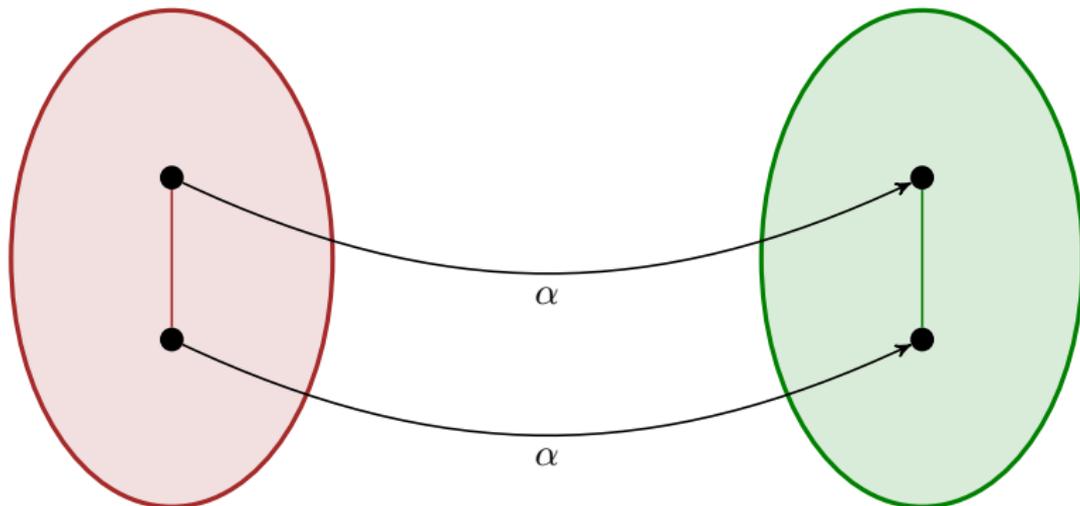


$$\alpha \circ \gamma(\bar{y}) \bar{\sqsubseteq} \bar{y}$$

$(\alpha \circ \gamma \text{ reductive})$

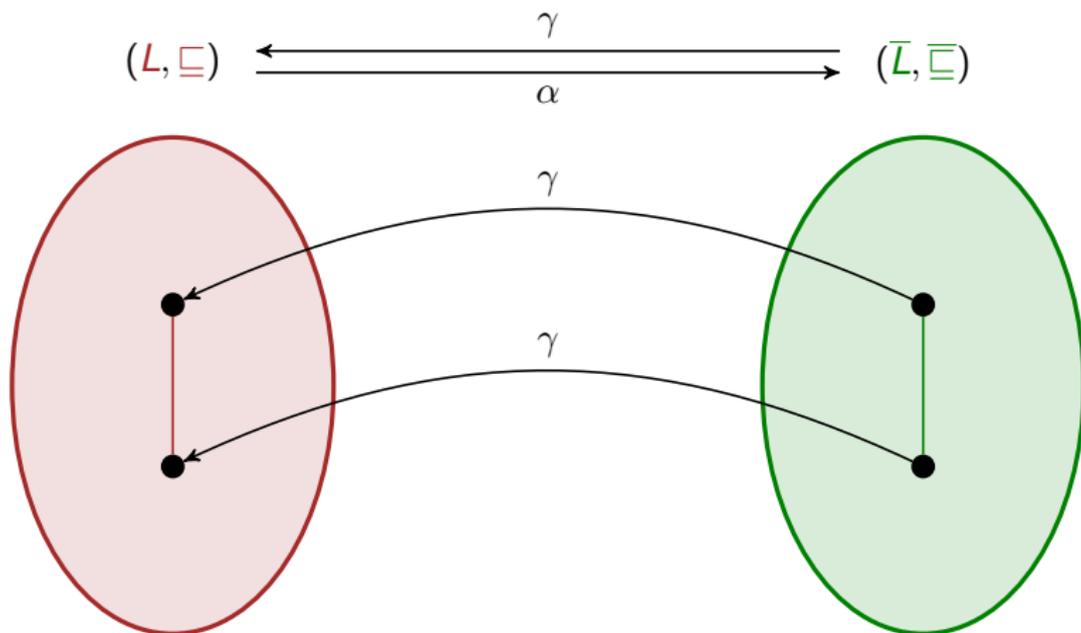
# Galois Connections: Characterization

$$(L, \sqsubseteq) \begin{array}{c} \xleftarrow{\gamma} \\ \xrightarrow{\alpha} \end{array} (\bar{L}, \bar{\sqsubseteq})$$



$\alpha$  is monotonic

# Galois Connections: Characterization



$\gamma$  is monotonic

# Galois Connections: Properties

For any Galois connection  $(L, \sqsubseteq) \xleftrightarrow[\alpha]{\gamma} (\bar{L}, \bar{\sqsubseteq})$ , we have

$$\alpha = \alpha \circ \gamma \circ \alpha$$

$$\gamma = \gamma \circ \alpha \circ \gamma$$

$\alpha$  is surjective    iff     $\gamma$  is injective    iff     $\alpha \circ \gamma = \lambda \bar{\gamma} . \bar{y}$

## Definition

A **Galois insertion** between a lattice  $(L, \sqsubseteq)$  and a lattice  $(\bar{L}, \bar{\sqsubseteq})$  is any Galois connection  $(L, \sqsubseteq) \xleftrightarrow[\alpha]{\gamma} (\bar{L}, \bar{\sqsubseteq})$  where  $\alpha$  is surjective.

Notation for Galois insertions:  $(L, \sqsubseteq) \xleftrightarrow[\alpha]{\gamma} (\bar{L}, \bar{\sqsubseteq})$

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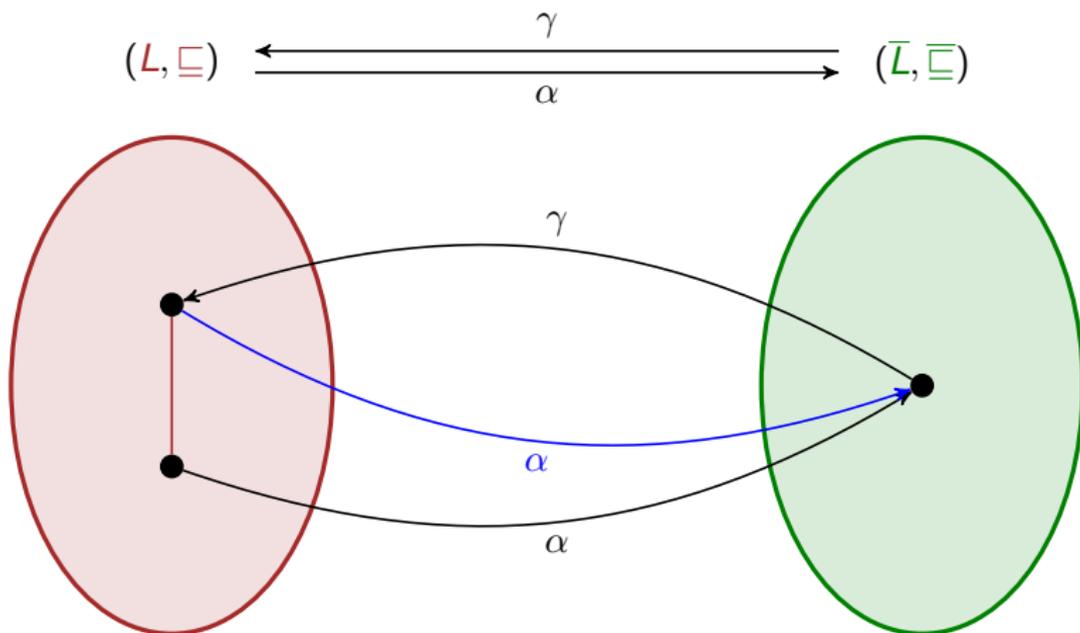
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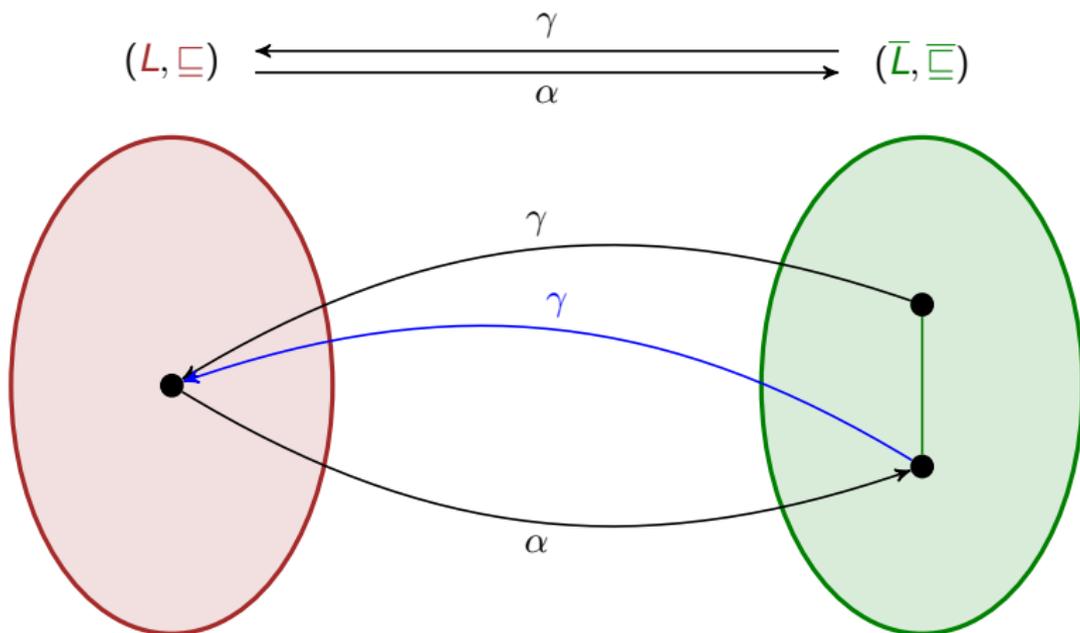
Notation for Galois insertions:  $(L, \sqsubseteq) \xleftrightarrow[\alpha]{\gamma} \Rightarrow (\bar{L}, \bar{\sqsubseteq})$

# Galois Connections: Properties



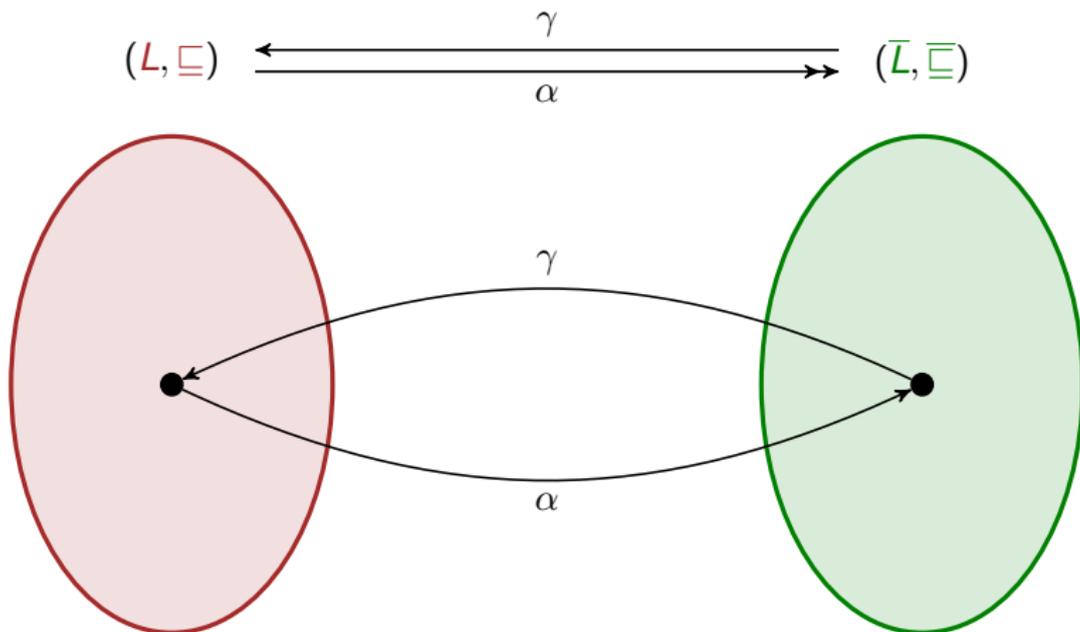
$$\alpha = \alpha \circ \gamma \circ \alpha$$

# Galois Connections: Properties



$$\gamma = \gamma \circ \alpha \circ \gamma$$

# Galois Connections: Properties



$$\alpha \circ \gamma = \lambda \bar{y} . \bar{y}$$

(Galois insertion)

# Galois Connections: Properties for Complete Lattices

For any Galois connection  $(L, \sqsubseteq) \xleftrightarrow[\alpha]{\gamma} (\bar{L}, \bar{\sqsubseteq})$  on **complete** lattices,

$$\alpha(x) = \bar{\bigcap} \{ \bar{y} \in \bar{L} \mid x \sqsubseteq \gamma(\bar{y}) \} \quad (\text{for all } x \in L)$$

$$\gamma(\bar{y}) = \bigcup \{ x \in L \mid \alpha(x) \bar{\sqsubseteq} \bar{y} \} \quad (\text{for all } \bar{y} \in \bar{L})$$

$$\alpha(\bigcup X) = \bar{\bigcap} \{ \alpha(x) \mid x \in X \} \quad (\text{for all } X \subseteq L)$$

$$\gamma(\bar{\bigcap} \bar{Y}) = \bigcap \{ \gamma(\bar{y}) \mid \bar{y} \in \bar{Y} \} \quad (\text{for all } \bar{Y} \subseteq \bar{L})$$

## Informally

$\alpha$  uniquely determines  $\gamma$  and  $\gamma$  uniquely determines  $\alpha$ .

$\alpha$  preserves least upper bounds,  $\gamma$  preserves greatest lower bounds.

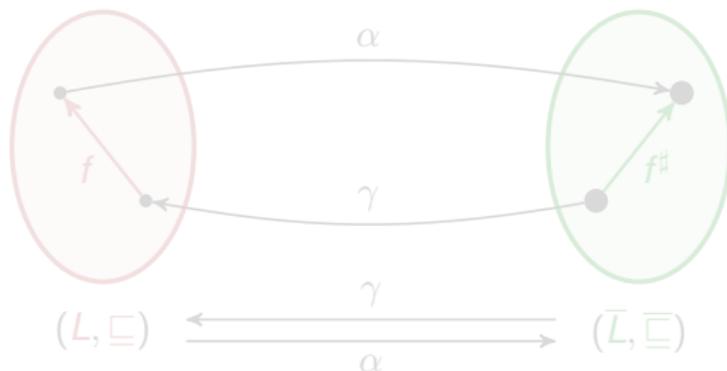
# Best Abstraction of a Monotonic Concrete Function

Consider a Galois connection  $(L, \sqsubseteq) \xleftrightarrow[\alpha]{\gamma} (\bar{L}, \bar{\sqsubseteq})$  on **complete** lattices.

## Definition

For any *monotonic* function  $f : L \rightarrow L$ , the **best abstraction** of  $f$  is the *monotonic* function  $f^\# : \bar{L} \rightarrow \bar{L}$  defined by:

$$f^\# = \alpha \circ f \circ \gamma$$



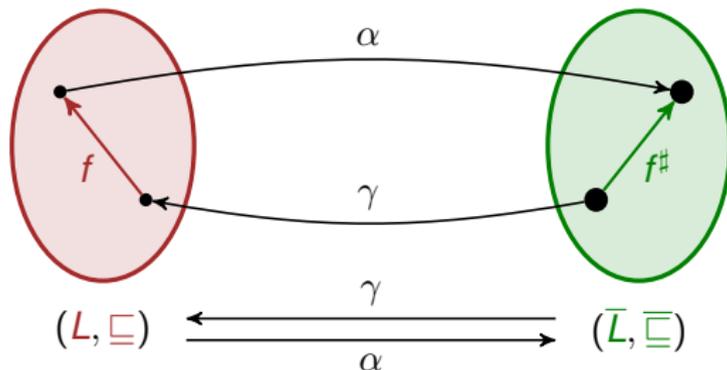
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$$f^\# = \alpha \circ f \circ \gamma$$



# Best Abstraction: Justification

Given a **monotonic** function  $f : L \rightarrow L$ , we look for a **monotonic** function  $\bar{g} : \bar{L} \rightarrow \bar{L}$  that is a **sound approximation** of  $f$ :

$$f(x) \sqsubseteq \gamma \circ \bar{g} \circ \alpha(x)$$

or equivalently (when  $\bar{g}$  is monotonic):

$$\alpha(x) \sqsubseteq \bar{y} \implies \bar{g}(\bar{y}) \sqsupseteq \alpha \circ f(x)$$

The **most precise** function satisfying the above condition is defined by:

$$\bar{g}(\bar{y}) = \bigsqcap \{ \alpha \circ f(x) \mid \alpha(x) \sqsubseteq \bar{y} \}$$

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# Best Abstraction: Justification

$$\bar{g}(\bar{y}) = \overline{\bigsqcup} \{ \alpha \circ f(x) \mid \alpha(x) \sqsubseteq \bar{y} \}$$

Recall that  $\alpha$  preserves least upper bounds, hence:

$$\bar{g}(\bar{y}) = \alpha \left( \bigsqcup \{ f(x) \mid x \in X \} \right)$$

where  $X = \{ x \in L \mid \alpha(x) \sqsubseteq \bar{y} \}$ . Since  $f$  is monotonic,

$$\bigsqcup \{ f(x) \mid x \in X \} \sqsubseteq f \left( \bigsqcup X \right)$$

Recall that  $\alpha \circ \gamma(\bar{y}) \sqsubseteq \bar{y}$ , hence  $\gamma(\bar{y}) \in X$  and we come to:

$$\bigsqcup \{ f(x) \mid x \in X \} \sqsupseteq f(\gamma(\bar{y}))$$

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# Best Abstraction: Justification

$$\bar{g}(\bar{y}) = \alpha \left( \bigsqcup \{f(x) \mid x \in X\} \right)$$

where  $X = \{x \in L \mid \alpha(x) \sqsubseteq \bar{y}\}$ .

$$f(\gamma(\bar{y})) \sqsubseteq \bigsqcup \{f(x) \mid x \in X\} \sqsubseteq f \left( \bigsqcup X \right)$$

Recall that  $\gamma(\bar{y}) = \bigsqcup X$ , hence:

$$\bigsqcup \{f(x) \mid x \in X\} = f \left( \bigsqcup X \right) = f(\gamma(\bar{y}))$$

We obtain that:

$$\bar{g}(\bar{y}) = \alpha \circ f \circ \gamma(\bar{y})$$

And... all is well, since  $\alpha \circ f \circ \gamma$  is **monotonic**!

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We obtain that:

$$\bar{g}(\bar{y}) = \alpha \circ f \circ \gamma(\bar{y})$$

And... all is well, since  $\alpha \circ f \circ \gamma$  is **monotonic**!

# Galois Connections: Fixpoint Abstraction

Consider a Galois connection  $(L, \sqsubseteq) \xleftrightarrow[\alpha]{\gamma} (\bar{L}, \bar{\sqsubseteq})$  on **complete** lattices.

Recall that for any monotonic function  $f : L \rightarrow L$ , we denote by  $f^\#$  the monotonic function:

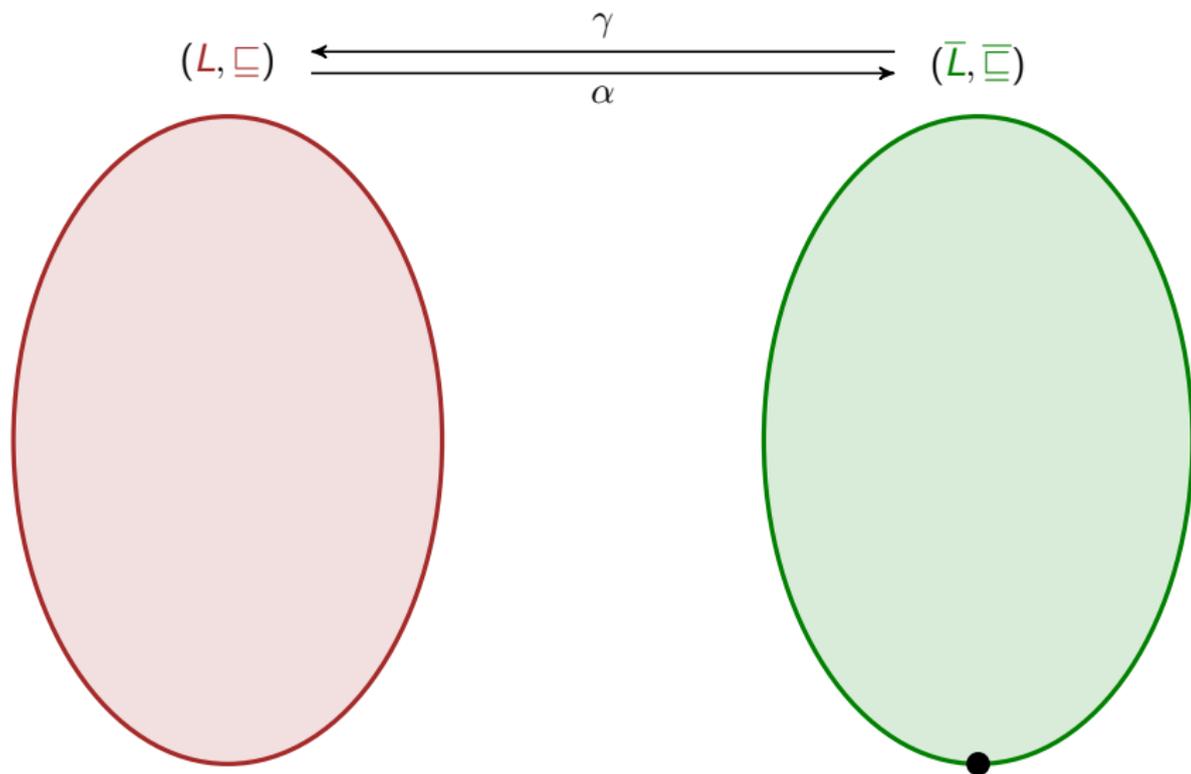
$$f^\# = \alpha \circ f \circ \gamma$$

## Theorem

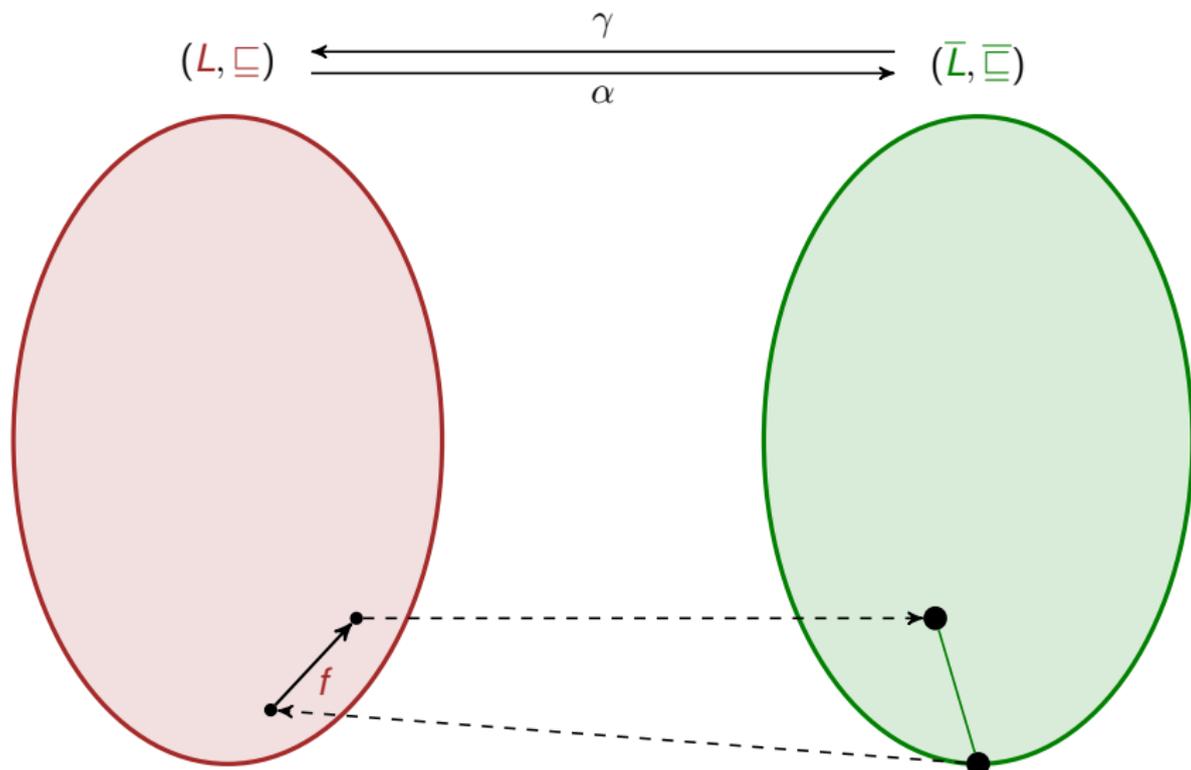
*For any monotonic function  $f : L \rightarrow L$ , the least fixpoints of  $f$  and  $f^\#$  satisfy:*

$$\text{lfp}(f) \sqsubseteq \gamma \left( \text{lfp}(f^\#) \right)$$

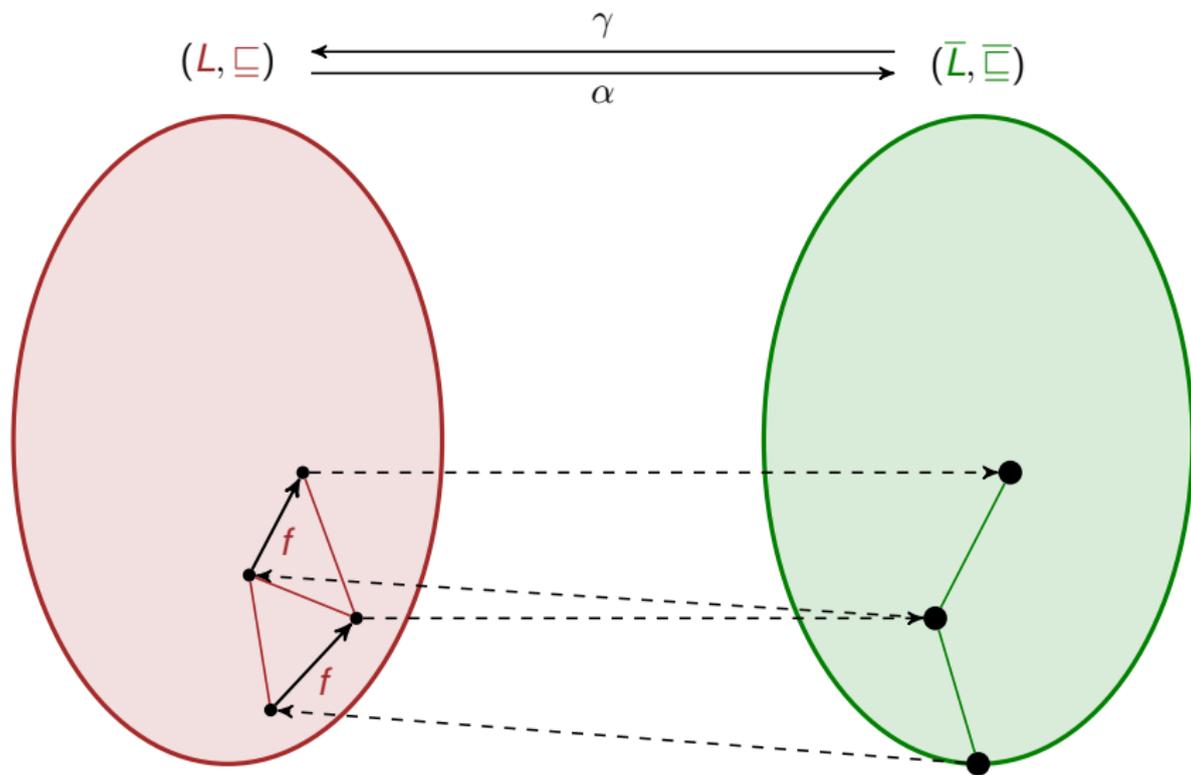
# Galois Connections: Fixpoint Abstraction



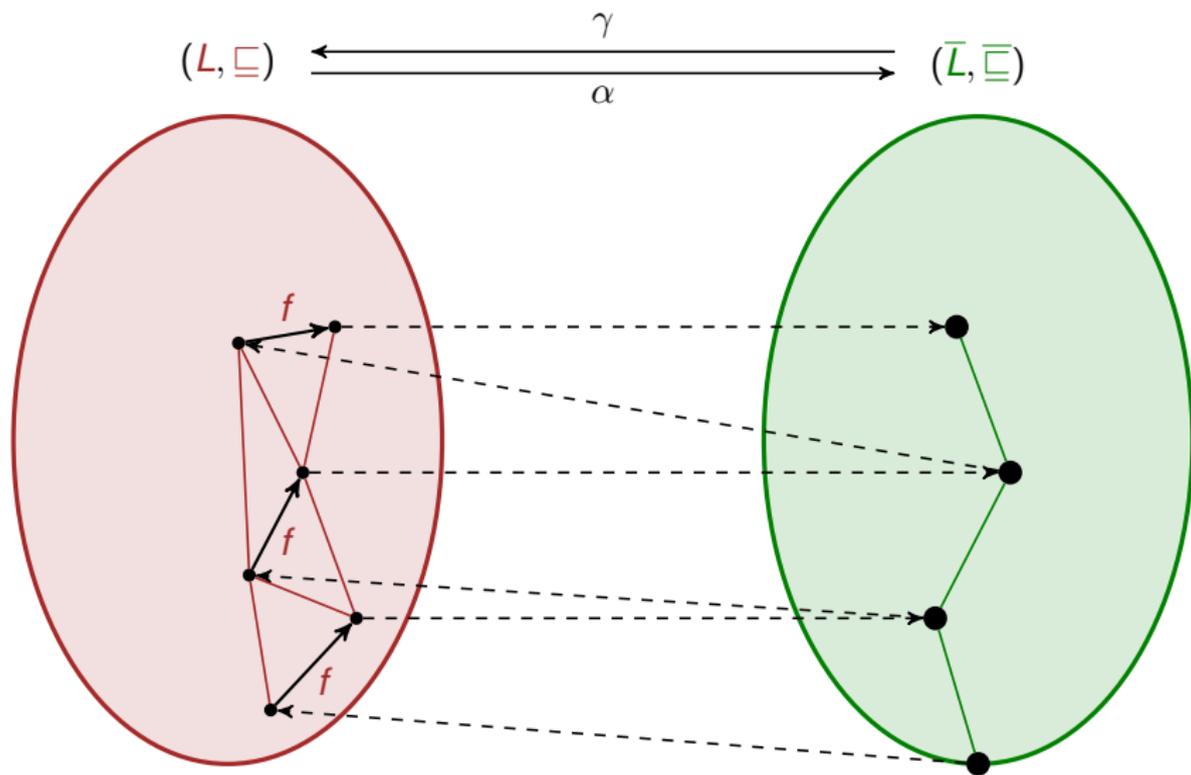
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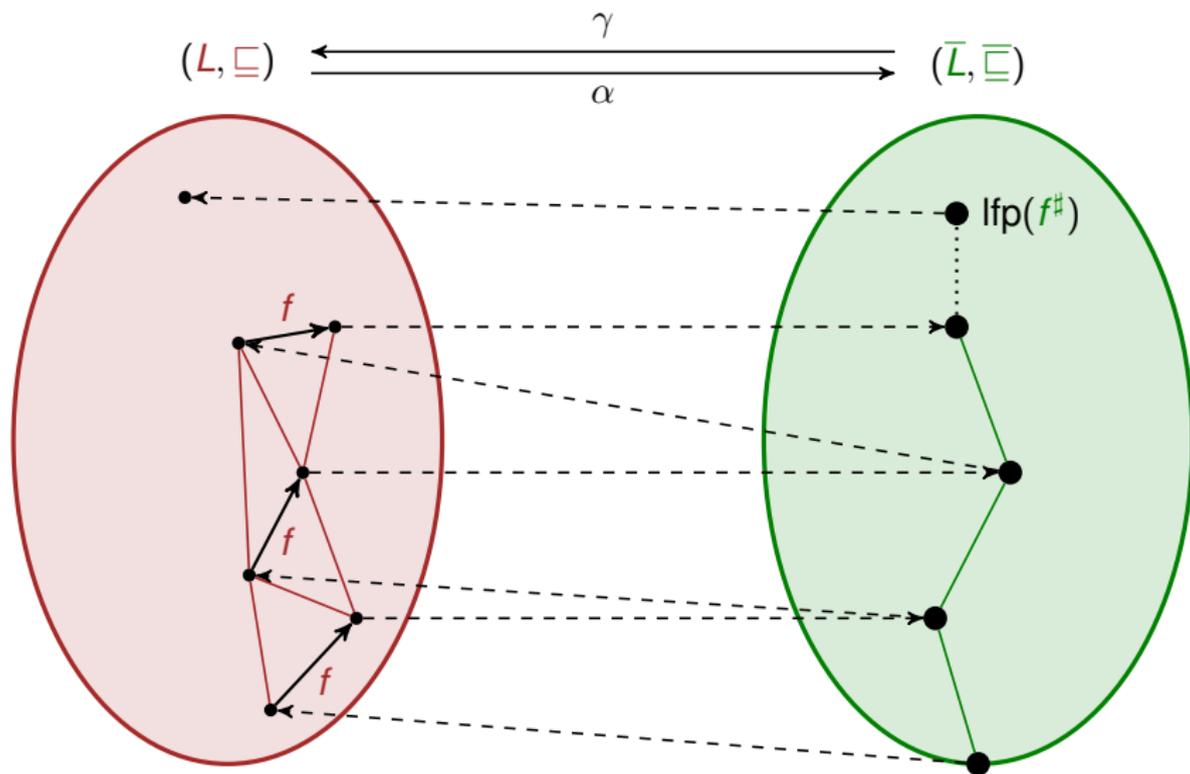
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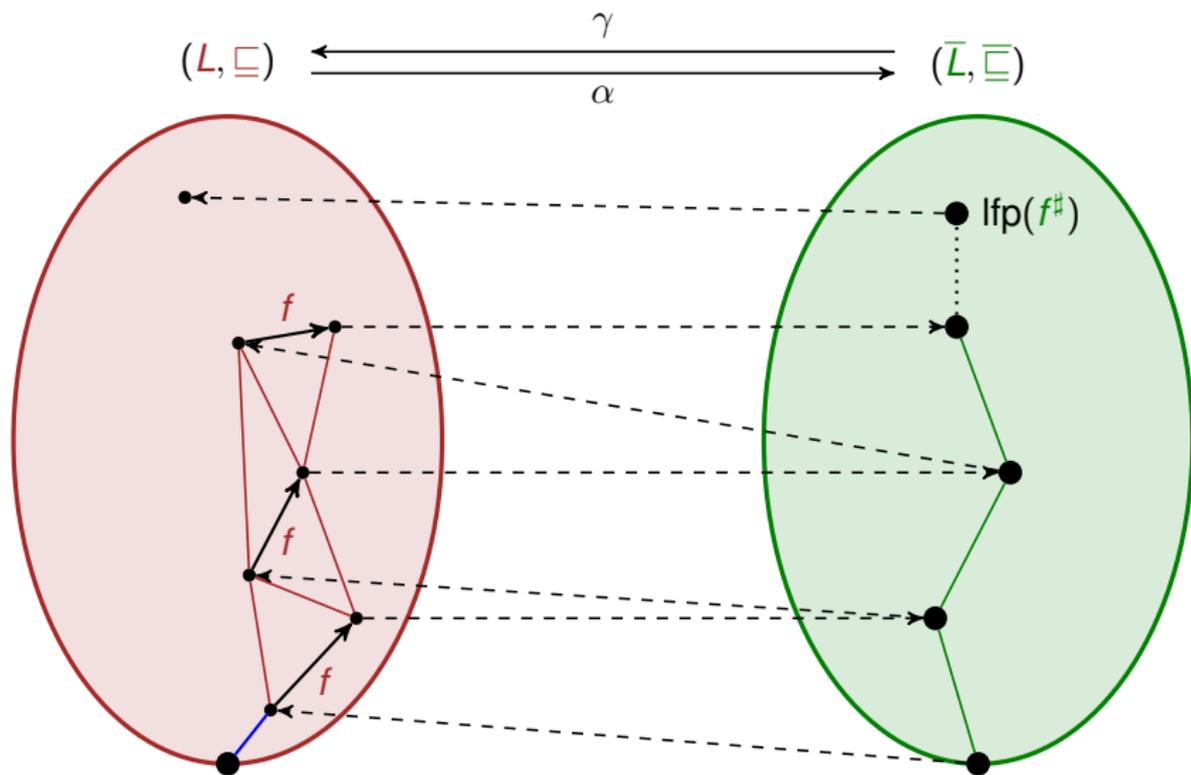
# Galois Connections: Fixpoint Abstraction



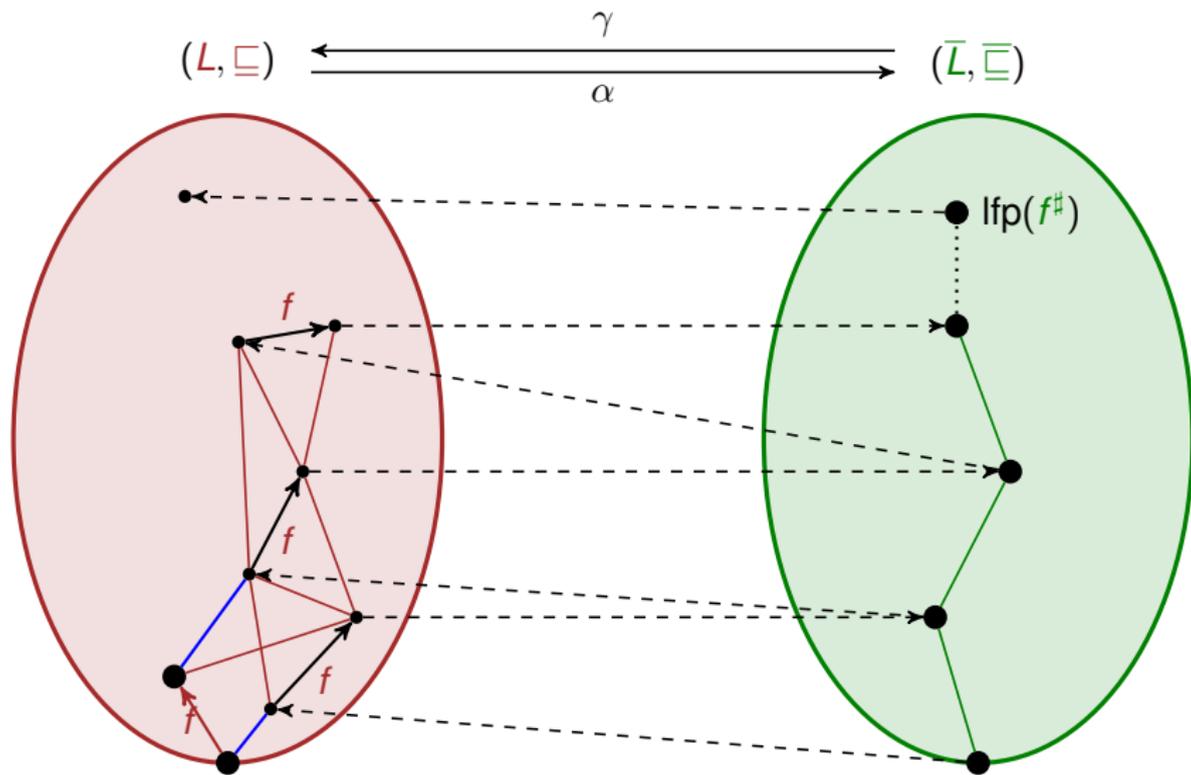
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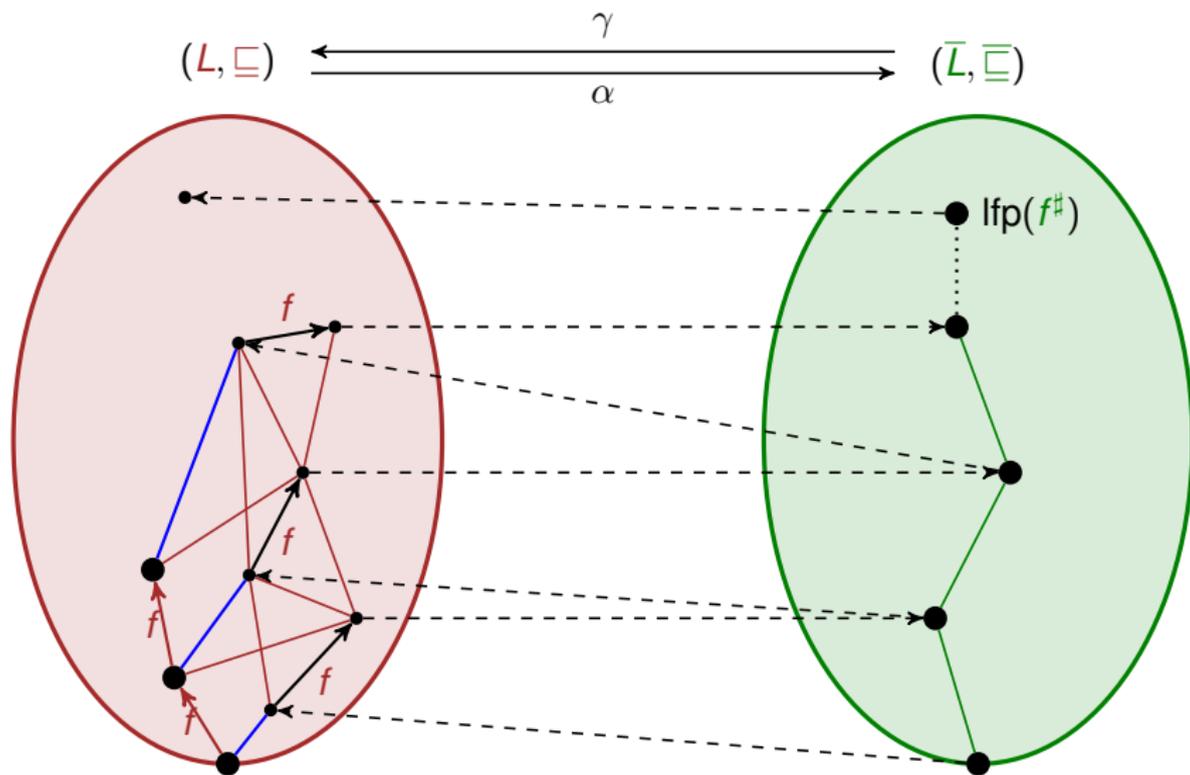
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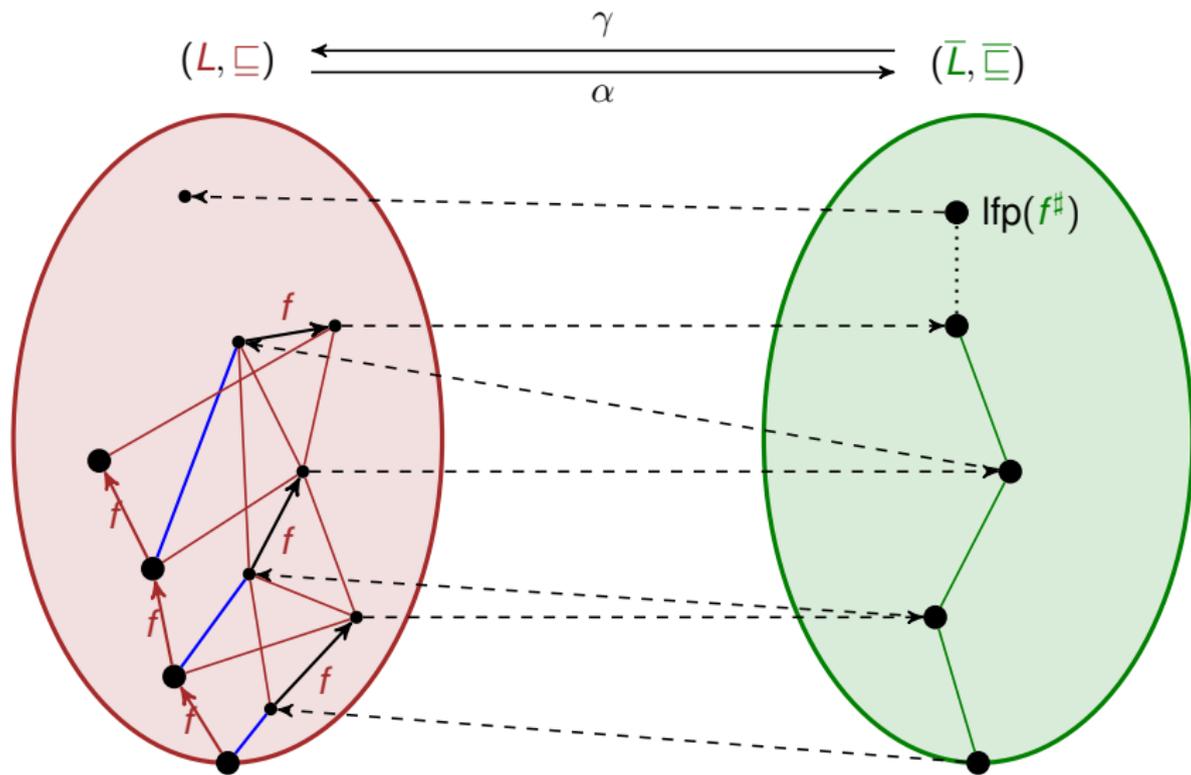
# Galois Connections: Fixpoint Abstraction



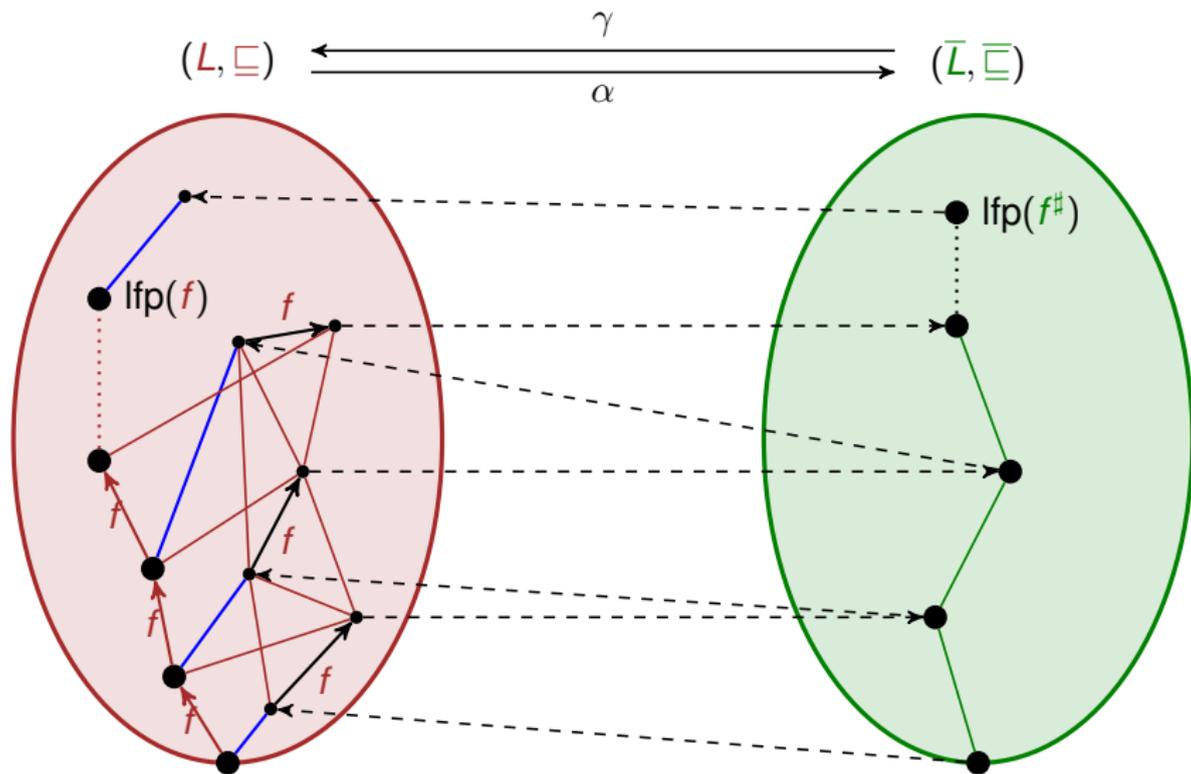
# Galois Connections: Fixpoint Abstraction



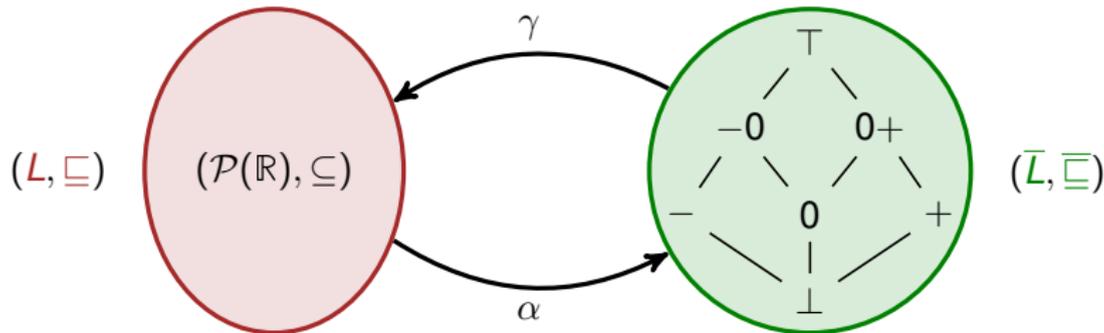
# Galois Connections: Fixpoint Abstraction



# Galois Connections: Fixpoint Abstraction



# Best Abstraction & Fixpoint Abstraction: Example



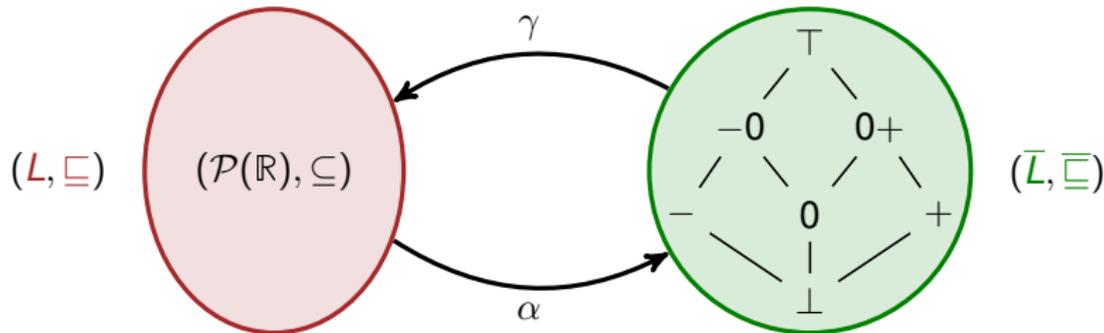
$f$

$\mathcal{P}(\mathbb{R})$	$\rightarrow$	$\mathcal{P}(\mathbb{R})$
$x$	$\mapsto$	$\{r+2 \mid r \in x\} \cup \{5\}$

$f^\#$

$\bar{y}$	$\perp$	$-$	$0$	$+$	$-0$	$0+$	$\top$
$f^\#(\bar{y})$							

# Best Abstraction & Fixpoint Abstraction: Example



$f$

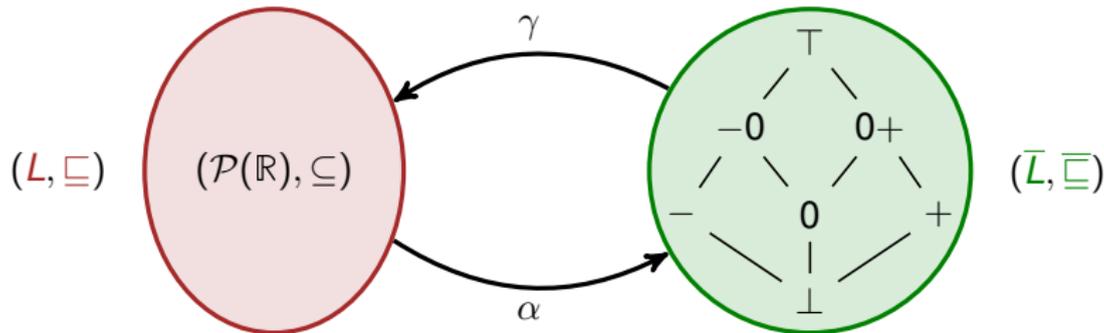
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$f^\#(\bar{y})$	$+$						

$$\begin{aligned}
 f^\#(\perp) &= \alpha \circ f \circ \gamma(\perp) \\
 &= \alpha \circ f(\emptyset) \\
 &= \alpha(\{5\}) \\
 &= +
 \end{aligned}$$

# Best Abstraction & Fixpoint Abstraction: Example



$f$

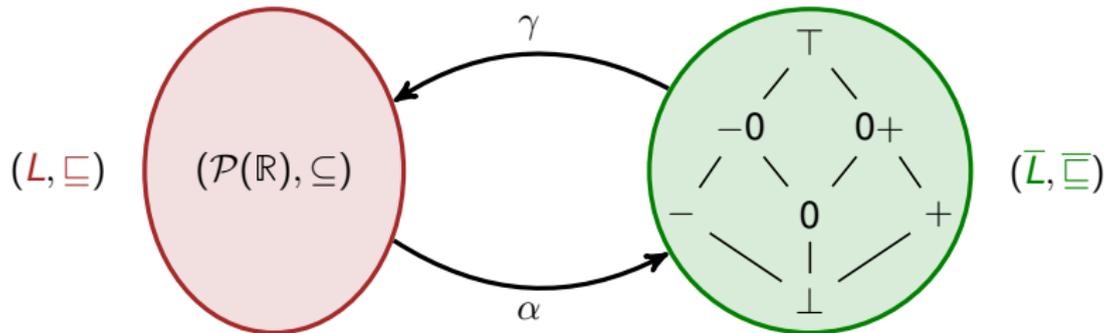
$\mathcal{P}(\mathbb{R})$	$\rightarrow$	$\mathcal{P}(\mathbb{R})$
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$f^\#$

$\bar{y}$	$\perp$	$-$	$0$	$+$	$-0$	$0+$	$\top$
$f^\#(\bar{y})$	$+$	$\top$					

$$\begin{aligned}
 f^\#(-) &= \alpha \circ f \circ \gamma(-) \\
 &= \alpha \circ f(\{r \in \mathbb{R} \mid r < 0\}) \\
 &= \alpha(\{r+2 \mid r < 0\} \cup \{5\}) \\
 &= \top
 \end{aligned}$$

# Best Abstraction & Fixpoint Abstraction: Example



$f$

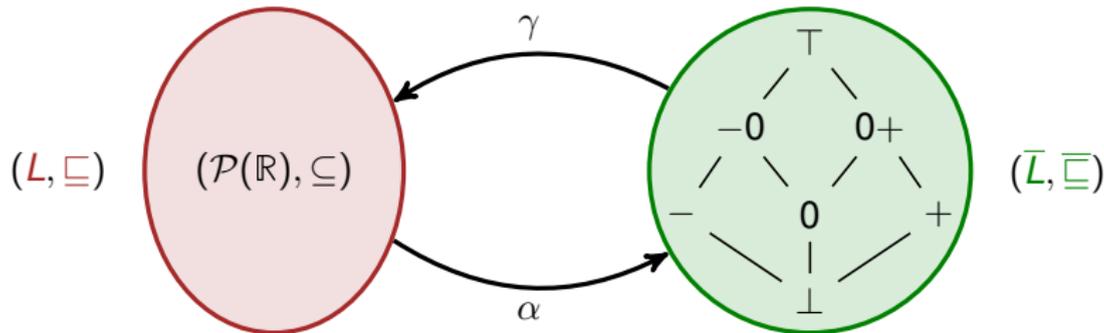
$\mathcal{P}(\mathbb{R})$	$\rightarrow$	$\mathcal{P}(\mathbb{R})$
$x$	$\mapsto$	$\{r+2 \mid r \in x\} \cup \{5\}$

$f^\#$

$\bar{y}$	$\perp$	$-$	$0$	$+$	$-0$	$0+$	$\top$
$f^\#(\bar{y})$	$+$	$\top$	$+$				

$$\begin{aligned}
 f^\#(0) &= \alpha \circ f \circ \gamma(0) \\
 &= \alpha \circ f(\{0\}) \\
 &= \alpha(\{2\} \cup \{5\}) \\
 &= +
 \end{aligned}$$

# Best Abstraction & Fixpoint Abstraction: Example



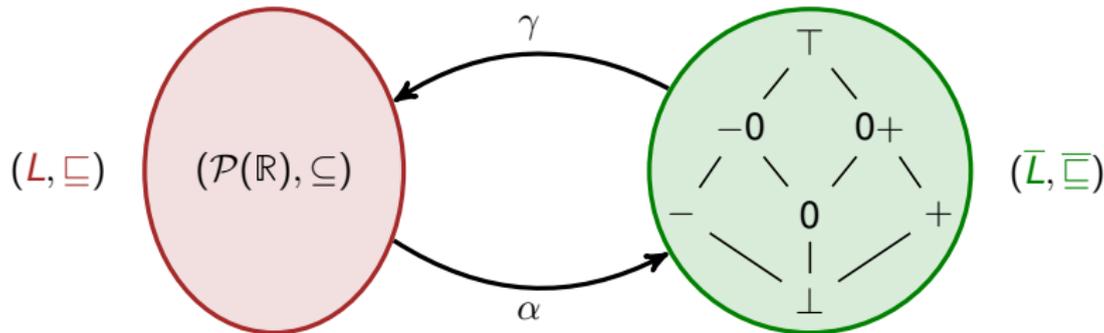
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$\mathcal{P}(\mathbb{R})$	$\rightarrow$	$\mathcal{P}(\mathbb{R})$
$x$	$\mapsto$	$\{r+2 \mid r \in x\} \cup \{5\}$

$f^\#$

$\bar{y}$	$\perp$	$-$	$0$	$+$	$-0$	$0+$	$\top$
$f^\#(\bar{y})$	$+$	$\top$	$+$	$+$	$\top$	$+$	$\top$

# Best Abstraction & Fixpoint Abstraction: Example



$f$	
$\mathcal{P}(\mathbb{R})$	$\rightarrow \mathcal{P}(\mathbb{R})$
$x$	$\mapsto \{r + 2 \mid r \in x\} \cup \{5\}$

$$f(\emptyset) = \{5\}$$

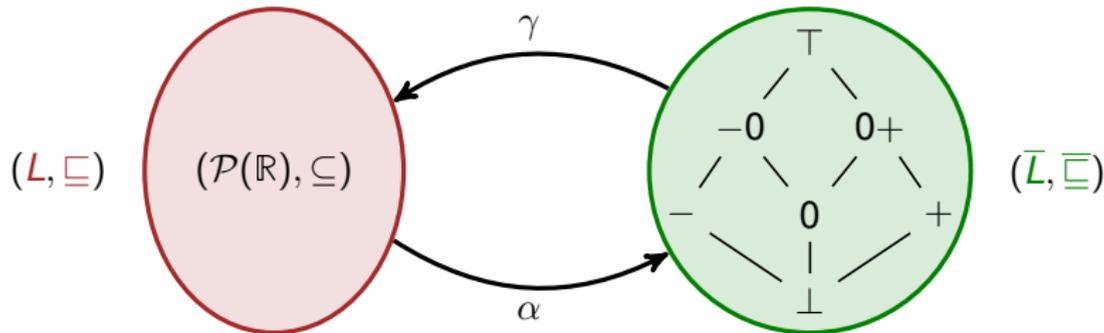
$$f^2(\emptyset) = \{5, 7\}$$

$$f^3(\emptyset) = \{5, 7, 9\}$$

$$\text{lfp } f = \{5 + 2k \mid k \in \mathbb{N}\}$$

$f^\#$							
$\bar{y}$	$\perp$	$-$	$0$	$+$	$-0$	$0+$	$\top$
$f^\#(\bar{y})$	$+$	$\top$	$+$	$+$	$\top$	$+$	$\top$

# Best Abstraction & Fixpoint Abstraction: Example



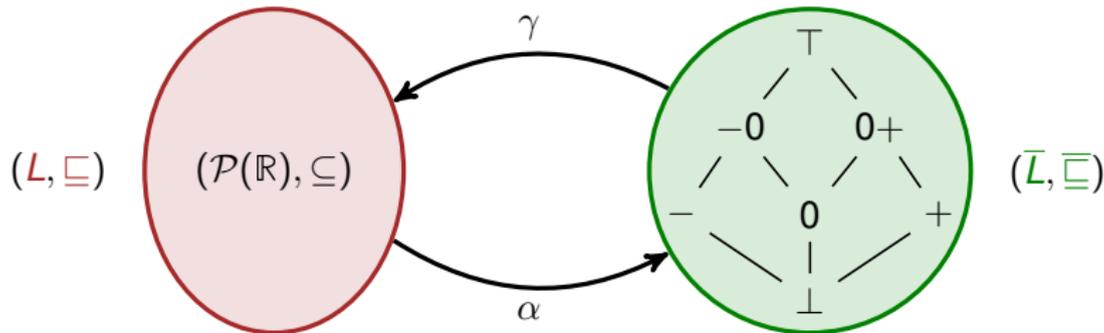
$f$	
$\mathcal{P}(\mathbb{R})$	$\rightarrow \mathcal{P}(\mathbb{R})$
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 f(\emptyset) &= \{5\} \\
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$f^\#$							
$\bar{y}$	$\perp$	$-$	$0$	$+$	$-0$	$0+$	$\top$
$f^\#(\bar{y})$	$+$	$\top$	$+$	$+$	$\top$	$+$	$\top$

$$\begin{aligned}
 f^\#(\perp) &= + \\
 f^{\#2}(\perp) &= + \\
 \text{lfp } f^\# &= +
 \end{aligned}$$

# Best Abstraction & Fixpoint Abstraction: Example



$f$	
$\mathcal{P}(\mathbb{R})$	$\rightarrow \mathcal{P}(\mathbb{R})$
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 \end{aligned}$$

$f^\#$							
$\bar{y}$	$\perp$	$-$	$0$	$+$	$-0$	$0+$	$\top$
$f^\#(\bar{y})$	$+$	$\top$	$+$	$+$	$\top$	$+$	$\top$

$$\text{lfp } f^\# = +$$

$$\gamma(\text{lfp}(f^\#)) = \{r \in \mathbb{R} \mid r > 0\}$$

# Galois Connections: Summary & Application

We want to “compute” the least fixpoint  $\text{lfp}(f)$  of monotonic function  $f : L \rightarrow L$  on a complete lattice  $(L, \sqsubseteq)$ .

If Kleene iteration  $\perp \sqsubseteq f(\perp) \sqsubseteq \dots \sqsubseteq f^i(\perp) \sqsubseteq \dots$  diverges then:

- 1 design an abstract complete lattice  $(\bar{L}, \bar{\sqsubseteq})$ , simpler than  $(L, \sqsubseteq)$ , and formalize the “meaning” of abstract values by a Galois connection

$$(L, \sqsubseteq) \begin{matrix} \xleftarrow{\gamma} \\ \xrightarrow{\alpha} \end{matrix} (\bar{L}, \bar{\sqsubseteq})$$

- 2 compute  $\text{lfp}(f^\#)$ , where  $f^\# = \alpha \circ f \circ \gamma$  is the best abstraction of  $f$ .

By Fixpoint Abstraction Theorem,  $\gamma(\text{lfp}(f^\#))$  **soundly approximates**  $\text{lfp}(f)$

$$\text{lfp}(f) \sqsubseteq \gamma(\text{lfp}(f^\#)) \quad \text{or equivalently} \quad \alpha(\text{lfp}(f)) \bar{\sqsubseteq} \text{lfp}(f^\#)$$

- 8 Some More Lattice Theory: Galois Connections
- 9 Abstract Interpretation-Based Data Flow Analysis**
  - Design of Approximate Transfer Mappings for Sign Analysis
- 10 Convergence Acceleration with Widening and Narrowing

# Short Introduction to Abstract Interpretation

Recall that to ensure correctness of data flow analyses. . .

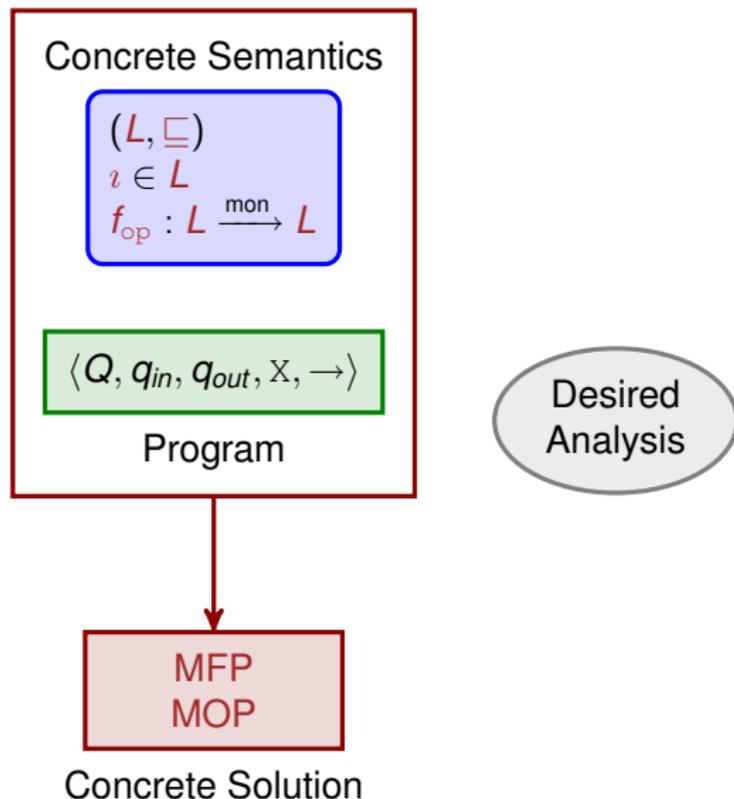
. . . we need a formal link to relate data flow facts and transfer functions with the formal semantics.

Abstract interpretation relies on **Galois connections** to formally express these relationships.

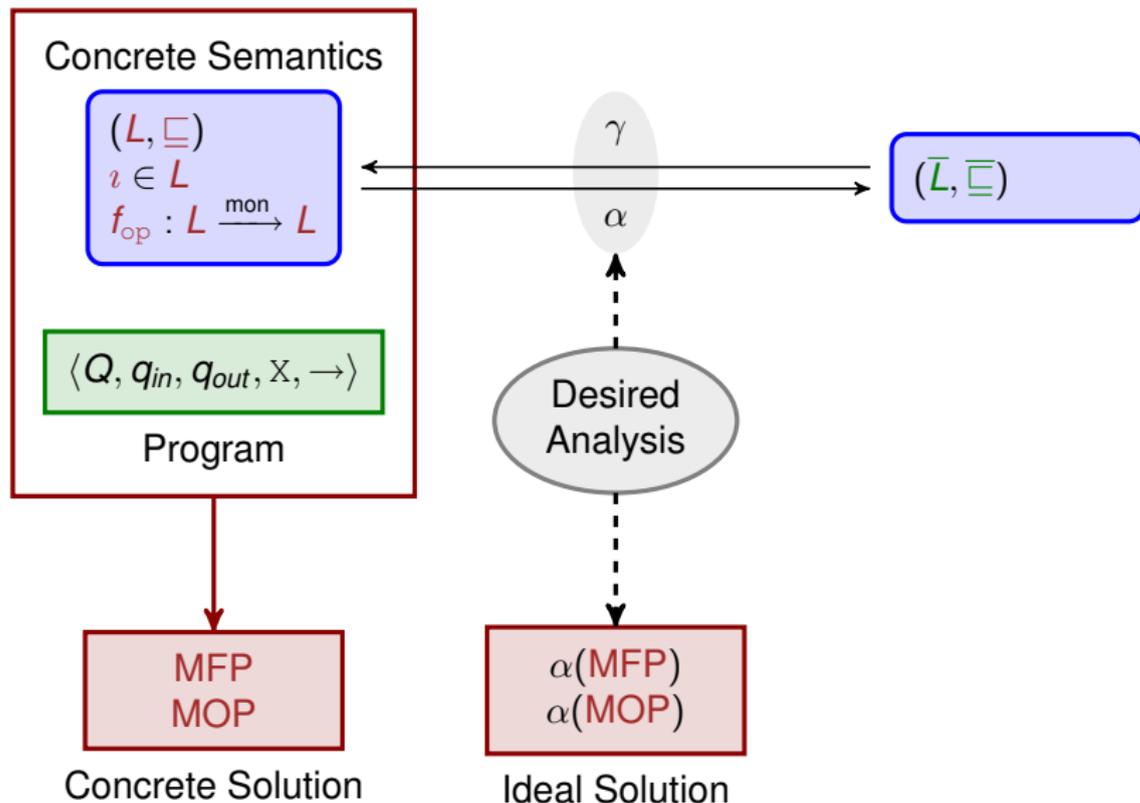
- **Formal meaning of data flow facts** by a concretization function
- Transfer mapping that **soundly approximates** the formal **semantics**
- **Sound** fixpoint **approximation**

Data flow analyses that are correct by design: **crucial for verification!**

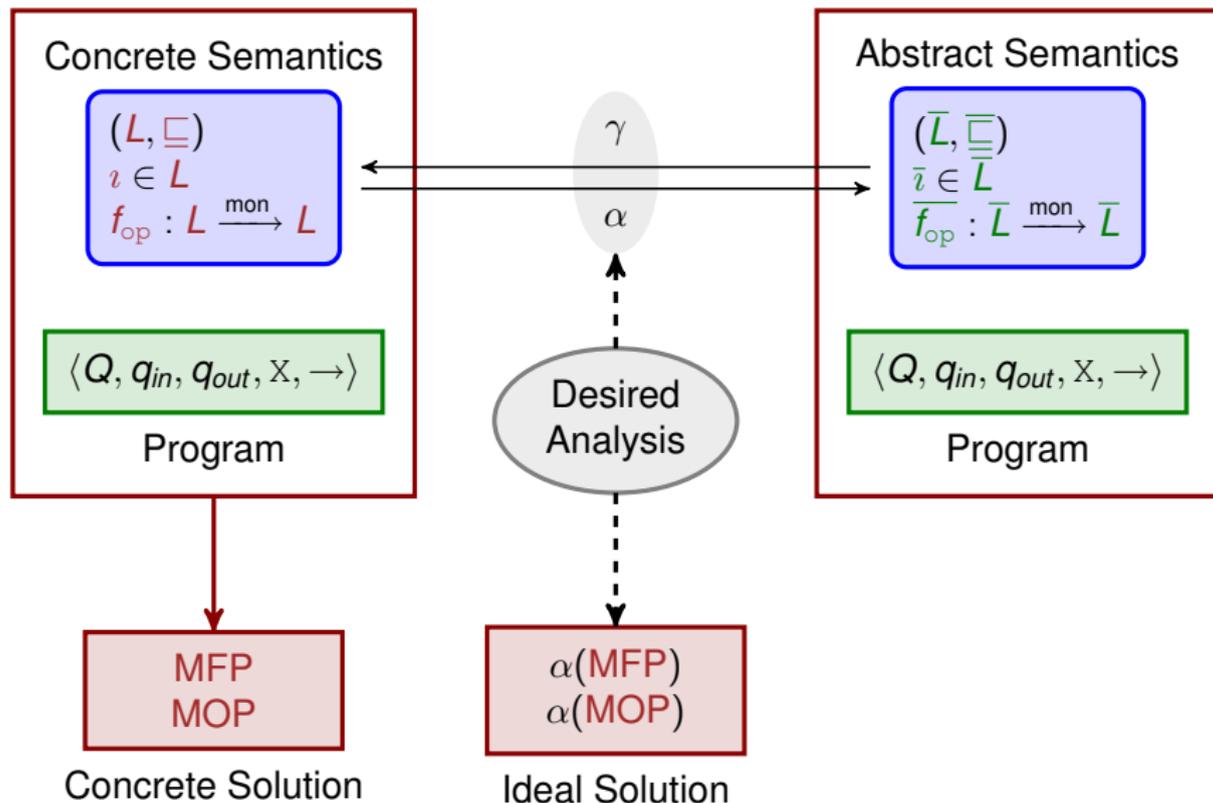
# Systematic Design of Correct of Data Flow Analyses



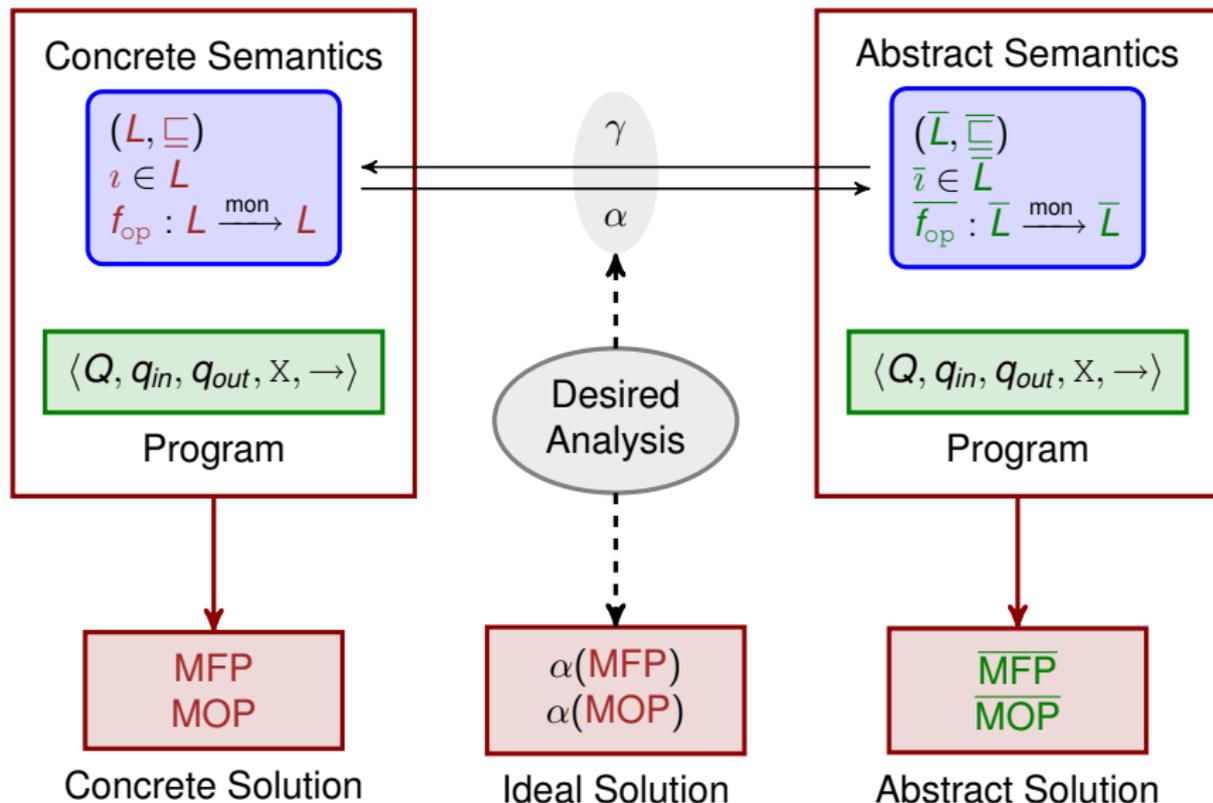
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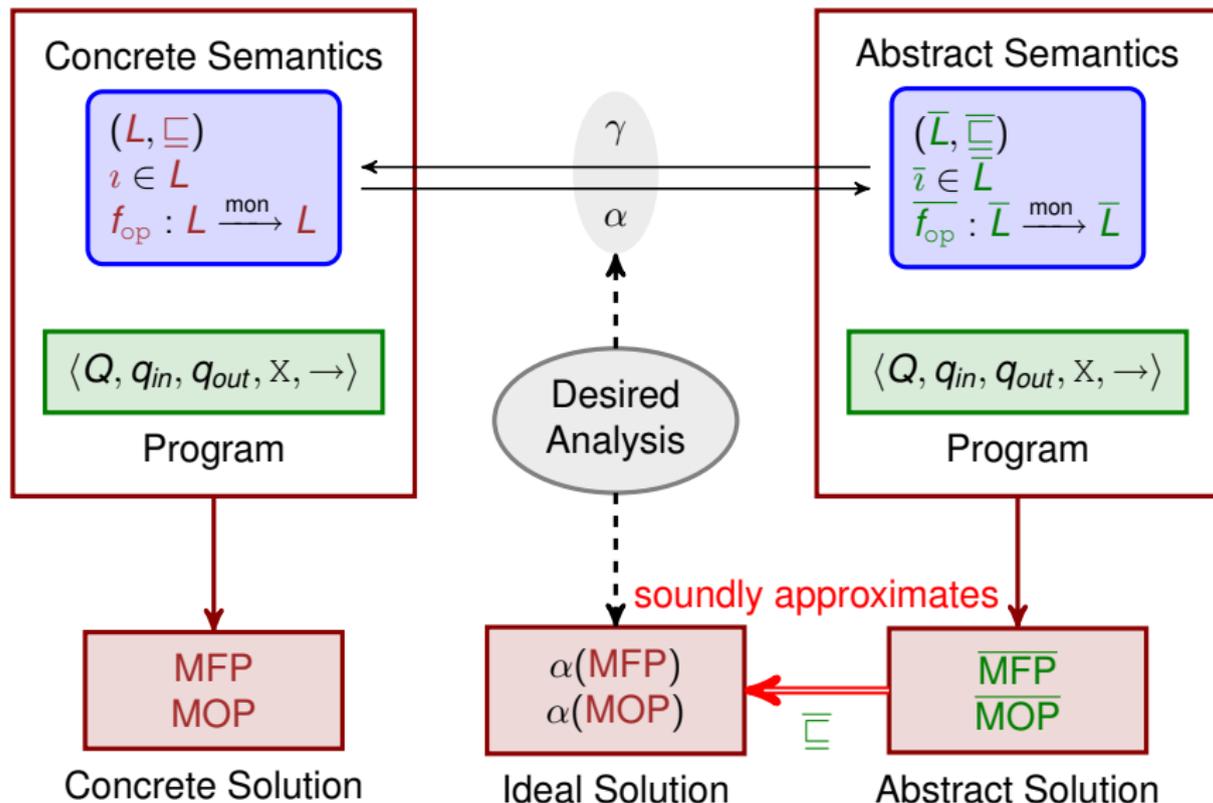
# Systematic Design of Correct of Data Flow Analyses



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# Systematic Design of Correct of Data Flow Analyses



# Systematic Design of Correct of Data Flow Analyses

The **MOP** solution of the concrete semantics is the strongest property (i.e. the most precise fact) that is satisfied by all runs of the program.

The ideal solution to a given analysis is an approximation of the concrete **MOP** solution.

## Natural Limitation

The class of possible analyses depends on the **choice** of “standard” **concrete semantics**.

Abstract data flow facts and transfer functions cannot be more precise than concrete ones.

Our operational semantics:  $\langle Q \times (X \rightarrow \mathbb{R}), \text{Init}, \text{Out}, \text{Op}, \rightarrow \rangle$

Focus on numerical analyses

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Focus on numerical analyses

# Standard Concrete Semantics

Control Flow Automaton:  $\langle Q, q_{in}, q_{out}, X, \rightarrow \rangle$

Recall: semantics  $\llbracket op \rrbracket$  of operations  $op \in Op$

$$\llbracket op \rrbracket \subseteq (X \rightarrow \mathbb{R}) \times (X \rightarrow \mathbb{R})$$

## Monotone Framework

- Complete lattice  $(L, \sqsubseteq)$  of data flow facts:  $(\mathcal{P}(X \rightarrow \mathbb{R}), \subseteq)$
- Set  $\mathcal{F}$  of monotonic transfer functions:

$$\mathcal{F} = \{ \lambda \phi. R[\phi] \mid R \subseteq (X \rightarrow \mathbb{R}) \times (X \rightarrow \mathbb{R}) \}$$

## Data Flow Instance $\vec{s}$ for Forward Analysis

- Initial data flow value:  $\top = X \rightarrow \mathbb{R}$
- Transfer mapping:  $f_{op}(\phi) = \llbracket op \rrbracket[\phi]$

# Standard Concrete Semantics

Control Flow Automaton:  $\langle Q, q_{in}, q_{out}, X, \rightarrow \rangle$

Recall: semantics  $\llbracket op \rrbracket$  of operations  $op \in Op$

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## Data Flow Instance $\overleftarrow{\mathcal{S}}$ for Backward Analysis

- Initial data flow value:  $\top = X \rightarrow \mathbb{R}$
- Transfer mapping:  $f_{op}(\phi) = \llbracket op \rrbracket^{-1}[\phi]$

# Standard Concrete Semantics

Control Flow Automaton:  $\langle Q, q_{in}, q_{out}, X, \rightarrow \rangle$

Recall: semantics  $[[op]]$  of operations  $op \in Op$

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Monotone Framework: **Distributive**

- Complete lattice  $(L, \sqsubseteq)$  of data flow facts:  $(\mathcal{P}(X \rightarrow \mathbb{R}), \subseteq)$
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Data Flow Instance  $\overleftarrow{\mathcal{S}}$  for **Backward** Analysis

- Initial data flow value:  $\top = X \rightarrow \mathbb{R}$
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# Post\* and Pre\* as Data Flow Analysis Solutions

Consider a control flow automaton:  $\langle Q, q_{in}, q_{out}, X, \rightarrow \rangle$ . Recall that:

$$\text{Post}^* = \bigcup_{q_{in} \xrightarrow{\text{op}_0} \dots \xrightarrow{\text{op}_k} q} \{q\} \times ([\text{op}_k] \circ \dots \circ [\text{op}_0]) [(X \rightarrow \mathbb{R})]$$

$$\text{Pre}^* = \bigcup_{q \xrightarrow{\text{op}_0} \dots \xrightarrow{\text{op}_k} q_{out}} \{q\} \times \left( ([\text{op}_k] \circ \dots \circ [\text{op}_0])^{-1} \right) [(X \rightarrow \mathbb{R})]$$

$$\text{Post}^* = \overrightarrow{\text{MOP}} \left( \overrightarrow{S} \right) = \overrightarrow{\text{MFP}} \left( \overrightarrow{S} \right)$$

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# Abstraction of the Concrete Semantics: Intuition

## Concrete Semantics

$\langle (\mathcal{P}(X \rightarrow \mathbb{R}), \subseteq), \mathcal{F}, Q, q_{in}, q_{out}, X, \rightarrow, f, v \rangle$

## Galois connection

$(\mathcal{P}(X \rightarrow \mathbb{R}), \subseteq) \xrightleftharpoons[\alpha]{\gamma} (\bar{\mathcal{L}}, \bar{\subseteq})$

$\bar{\mathcal{L}}$  is a set of machine-representable “properties” of the variables.

## Example

$\bar{\mathcal{L}} = \{x \text{ is even, } y \text{ is odd or negative, } x \geq y \Rightarrow x = 2^i\}$

$\gamma(\bar{\psi})$  is the **meaning** of an abstract “property”  $\bar{\psi}$ .

$\alpha(\phi)$  encodes a **sound approximation** of  $\phi$ , the most precise one.

$\bar{\subseteq}$  corresponds to entailment between “properties”, and abstracts  $\subseteq$ .

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# Abstract Semantics Induced by a Galois Connection

Consider a data flow instance  $\mathcal{A} = \langle (L, \sqsubseteq), \mathcal{F}, Q, q_{in}, q_{out}, X, \rightarrow, f, \iota \rangle$   
and a Galois connection  $(L, \sqsubseteq) \xrightleftharpoons[\alpha]{\gamma} (\bar{L}, \bar{\sqsubseteq})$ .

## Definition

The **abstract data flow instance**  $\bar{\mathcal{A}}$  induced by  $\mathcal{A}$  and  $(L, \sqsubseteq) \xrightleftharpoons[\alpha]{\gamma} (\bar{L}, \bar{\sqsubseteq})$  is  $\bar{\mathcal{A}} = \langle (\bar{L}, \bar{\sqsubseteq}), \bar{\mathcal{F}}, Q, q_{in}, q_{out}, X, \rightarrow, \bar{f}, \bar{\iota} \rangle$  where:

$$\begin{aligned}\bar{\mathcal{F}} &= \bar{L} \xrightarrow{\text{mon}} \bar{L} \\ \bar{f} &= \lambda \text{op}. f_{\text{op}}^{\#} \\ \bar{\iota} &= \alpha(\iota)\end{aligned}$$

Recall that  $f_{\text{op}}^{\#} = \alpha \circ f_{\text{op}} \circ \gamma$  is the *best abstraction* of  $f_{\text{op}}$ .

# Correctness of Induced Abstract Data Flow Analysis

## Extension of Galois Connections to Functions

For any set  $Q$  and Galois connection  $(L, \sqsubseteq) \xleftrightarrow[\alpha]{\gamma} (\bar{L}, \bar{\sqsubseteq})$ , we have  $(Q \rightarrow L, \sqsubseteq) \xleftrightarrow[\alpha]{\gamma} (Q \rightarrow \bar{L}, \bar{\sqsubseteq})$  where:

$$\alpha(\mathbf{a}) = \lambda q. \alpha(\mathbf{a}(q))$$
$$\gamma(\bar{\mathbf{b}}) = \lambda q. \gamma(\bar{\mathbf{b}}(q))$$

## Theorem (Correctness of Induced Abstract Forward Analysis)

For any data flow instance  $\mathcal{A}$  and Galois connection  $(L, \sqsubseteq) \xleftrightarrow[\alpha]{\gamma} (\bar{L}, \bar{\sqsubseteq})$ , the induced abstract data flow instance  $\bar{\mathcal{A}}$  satisfies:

$$\begin{array}{ll} \overrightarrow{\text{MFP}}(\mathcal{A}) \sqsubseteq \gamma(\overrightarrow{\text{MFP}}(\bar{\mathcal{A}})) & \alpha(\overrightarrow{\text{MFP}}(\mathcal{A})) \sqsubseteq \overrightarrow{\text{MFP}}(\bar{\mathcal{A}}) \\ \overrightarrow{\text{MOP}}(\mathcal{A}) \sqsubseteq \gamma(\overrightarrow{\text{MOP}}(\bar{\mathcal{A}})) & \alpha(\overrightarrow{\text{MOP}}(\mathcal{A})) \sqsubseteq \overrightarrow{\text{MOP}}(\bar{\mathcal{A}}) \end{array}$$

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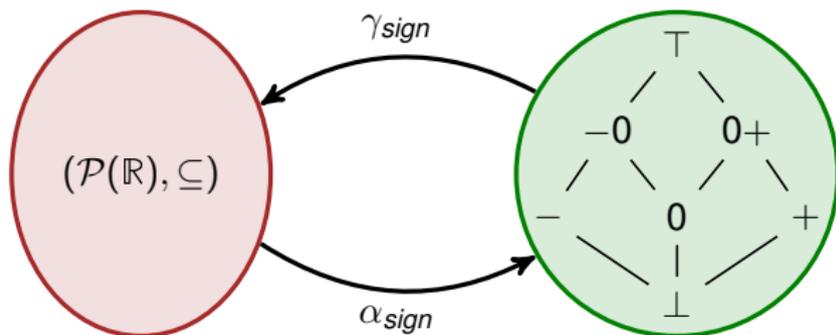
$$\alpha(\mathbf{a}) = \lambda q. \alpha(\mathbf{a}(q))$$
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## Theorem (Correctness of Induced Abstract Backward Analysis)

For any data flow instance  $\mathcal{A}$  and Galois connection  $(L, \sqsubseteq) \xleftrightarrow[\alpha]{\gamma} (\bar{L}, \bar{\sqsubseteq})$ , the induced abstract data flow instance  $\bar{\mathcal{A}}$  satisfies:

$$\overleftarrow{\text{MFP}}(\mathcal{A}) \sqsubseteq \gamma(\overleftarrow{\text{MFP}}(\bar{\mathcal{A}})) \quad \alpha(\overleftarrow{\text{MFP}}(\mathcal{A})) \sqsubseteq \overleftarrow{\text{MFP}}(\bar{\mathcal{A}})$$
$$\overleftarrow{\text{MOP}}(\mathcal{A}) \sqsubseteq \gamma(\overleftarrow{\text{MOP}}(\bar{\mathcal{A}})) \quad \alpha(\overleftarrow{\text{MOP}}(\mathcal{A})) \sqsubseteq \overleftarrow{\text{MOP}}(\bar{\mathcal{A}})$$

# Back Again to Sign Analysis: Galois Connection



$\gamma$							
$\bar{y}$	$\perp$	$-$	$0$	$+$	$-0$	$0+$	$\top$
$\gamma(\bar{y})$	$\emptyset$	$\{r \mid r < 0\}$	$\{0\}$	$\{r \mid r > 0\}$	$\{r \mid r \leq 0\}$	$\{r \mid r \geq 0\}$	$\mathbb{R}$

## Objective

Design a Galois Connection between:

- $(\mathcal{P}(x \rightarrow \mathbb{R}), \subseteq)$ , concrete data flow facts from standard semantics
- $(x \rightarrow \text{Sign}, \bar{\subseteq})$ , abstract data flow facts

# Intermediate Galois Connection: Projection

Convenient intermediate step for **non-relational analyses**

## Objective of Projection

Design a Galois Connection between:

- $(\mathcal{P}(X \rightarrow \mathbb{R}), \subseteq)$ , concrete data flow facts from standard semantics
- $(X \rightarrow \mathcal{P}(\mathbb{R}), \bar{\subseteq})$ , projected data flow facts

where  $\bar{\subseteq}$  is as expected:  $\bar{\psi} \bar{\subseteq} \bar{\psi}'$  if  $\bar{\psi}(x) \subseteq \bar{\psi}'(x)$  for all  $x \in X$ .

## Fact

$$(\mathcal{P}(X \rightarrow \mathbb{R}), \subseteq) \begin{matrix} \xleftarrow{\gamma_\pi} \\ \xrightarrow{\alpha_\pi} \end{matrix} (X \rightarrow \mathcal{P}(\mathbb{R}), \bar{\subseteq})$$

where:  $\alpha_\pi(\phi) = \lambda x. \{v(x) \mid v \in \phi\}$

$$\gamma_\pi(\bar{\psi}) = \{v \in X \rightarrow \mathbb{R} \mid v(x) \in \bar{\psi}(x) \text{ for all } x \in X\}$$

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# Back Again to Sign Analysis: Galois Connection

## Projection

$$(\mathcal{P}(X \rightarrow \mathbb{R}), \sqsubseteq) \begin{array}{c} \xleftarrow{\gamma_\pi} \\ \xrightarrow{\alpha_\pi} \end{array} (X \rightarrow \mathcal{P}(\mathbb{R}), \sqsubseteq)$$

## Sign

$$(\mathcal{P}(\mathbb{R}), \sqsubseteq) \begin{array}{c} \xleftarrow{\gamma_{\text{sign}}} \\ \xrightarrow{\alpha_{\text{sign}}} \end{array} (\text{Sign}, \bar{\sqsubseteq})$$

## Extension of Sign to Functions

$$(X \rightarrow \mathcal{P}(\mathbb{R}), \sqsubseteq) \begin{array}{c} \xleftarrow{\gamma_{\text{sign}}} \\ \xrightarrow{\alpha_{\text{sign}}} \end{array} (X \rightarrow \text{Sign}, \bar{\sqsubseteq})$$

$$(\mathcal{P}(X \rightarrow \mathbb{R}), \sqsubseteq) \begin{array}{c} \xleftarrow{\gamma_\pi \circ \gamma_{\text{sign}}} \\ \xrightarrow{\alpha_{\text{sign}} \circ \alpha_\pi} \end{array} (X \rightarrow \text{Sign}, \bar{\sqsubseteq})$$

The composition of Galois connections is a Galois connection.

# Back Again to Sign Analysis: Induced Instance

$$(\mathcal{P}(X \rightarrow \mathbb{R}), \subseteq) \xrightleftharpoons[\alpha_\pi]{\gamma_\pi} (X \rightarrow \mathcal{P}(\mathbb{R}), \subseteq) \xrightleftharpoons[\alpha_{\text{sign}}]{\gamma_{\text{sign}}} (X \rightarrow \text{Sign}, \bar{\subseteq})$$

$$\begin{aligned}\bar{\mathcal{F}} &= (X \rightarrow \text{Sign}) \xrightarrow{\text{mon}} (X \rightarrow \text{Sign}) \\ \bar{f} &= \lambda \text{op} . f_{\text{op}}^\# \\ \bar{i} &= \alpha_{\text{sign}} \circ \alpha_\pi(\top) = \lambda x . \bar{\top}\end{aligned}$$

But this data flow instance looks similar to what we did previously (less painfully) without Galois connections. . .

What do we get?

The most precise transfer mapping that soundly approximates the standard semantics

# Back Again to Sign Analysis: Induced Instance

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# Forward Sign Analysis: Transfer Mapping

$$(\mathcal{P}(X \rightarrow \mathbb{R}), \subseteq) \xrightleftharpoons[\alpha_\pi]{\gamma_\pi} (X \rightarrow \mathcal{P}(\mathbb{R}), \subseteq) \xrightleftharpoons[\alpha_{sign}]{\gamma_{sign}} (X \rightarrow \mathbf{Sign}, \bar{\subseteq})$$

$$f_{\text{op}}^\#(\bar{\psi}) = \alpha_{sign} \circ \alpha_\pi (\llbracket \text{op} \rrbracket [\gamma_\pi \circ \gamma_{sign}(\bar{\psi})])$$

Extensions of  $\llbracket e \rrbracket$  and  $\llbracket g \rrbracket$  to subsets of  $\mathbb{R}$

$\llbracket e \rrbracket$

$$\begin{array}{l} \mathcal{P}(X \rightarrow \mathbb{R}) \rightarrow \mathcal{P}(\mathbb{R}) \\ \phi \mapsto \{\llbracket e \rrbracket_v \mid v \in \phi\} \end{array}$$

$\llbracket g \rrbracket$

$$\begin{array}{l} \mathcal{P}(X \rightarrow \mathbb{R}) \rightarrow \mathcal{P}(X \rightarrow \mathbb{R}) \\ \phi \mapsto \{v \in \phi \mid v \models g\} \end{array}$$

$$f_{x:=e}^\#(\bar{v}) = \lambda y. \begin{cases} \bar{v}(y) & \text{if } y \neq x \\ \alpha_{sign} \circ \llbracket e \rrbracket \circ \gamma(\bar{v}) & \text{if } y = x \end{cases}$$

$$f_g^\# = \alpha_{sign} \circ \alpha_\pi \circ \llbracket g \rrbracket \circ \gamma_\pi \circ \gamma_{sign}$$

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Not easy to compute!

## Example

Consider a (non-constant) multivariate polynomial expression  $e$  and the operation  $\text{op} = x := e * e$ .

$$f_{\text{op}}^\#(\bar{\top}) = \lambda y. \begin{cases} \bar{\top} & \text{if } y \neq x \\ 0+ & \text{if } y = x \text{ and } e \text{ has a root} \\ + & \text{if } y = x \text{ and } e \text{ has no root} \end{cases}$$

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# Forward Sign Analysis: Transfer Mapping

What can be done?

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Approximate!

But soundly 😊

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But soundly 😊

## Approximate Transfer Mapping

Replace each  $f_{op}^\#$  with an **approximate transfer function**  $\overline{h_{op}}$  that

- exploits the structure of operations to obtain
- better performance at the expense of precision.

## Theorem

For any two monotonic functions  $f, g$  on a complete lattice  $(L, \sqsubseteq)$ ,

if  $f(x) \sqsubseteq g(x)$  for all  $x \in L$  then  $\text{lfp}(f) \sqsubseteq \text{lfp}(g)$

## Proof.

$$\{x \in L \mid f(x) \sqsubseteq x\} \supseteq \{x \in L \mid g(x) \sqsubseteq x\}$$

Hence

$$\text{lfp}(f) = \bigcap \{x \in L \mid f(x) \sqsubseteq x\} \sqsubseteq \bigcap \{x \in L \mid g(x) \sqsubseteq x\} = \text{lfp}(g)$$



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# Correctness of Approximate Transfer Mapping

Consider a data flow instance  $\mathcal{A}$  with a set  $\mathcal{F}$  of transfer functions and a transfer mapping  $f : \text{Op} \rightarrow \mathcal{F}$ .

For any monotonic function  $h : \text{Op} \rightarrow \mathcal{F}$  verifying

$$f_{\text{op}}(x) \sqsubseteq h_{\text{op}}(x) \quad (\text{for all } \text{op} \in \text{Op}, x \in L)$$

the data flow instance  $\mathcal{B}$  obtained from  $\mathcal{A}$  by replacing  $f$  with  $h$  satisfies:

$$\overrightarrow{\text{MFP}}(\mathcal{A}) \sqsubseteq \overrightarrow{\text{MFP}}(\mathcal{B}) \quad \overrightarrow{\text{MOP}}(\mathcal{A}) \sqsubseteq \overrightarrow{\text{MOP}}(\mathcal{B})$$

## Application to Induced Abstract Data Flow Instances

Replace  $f_{\text{op}}^\#$  with a simpler monotonic  $\overline{h_{\text{op}}}$  verifying

$$f_{\text{op}}^\#(\bar{x}) \sqsubseteq \overline{h_{\text{op}}}(\bar{x}) \quad (\text{for all } \text{op} \in \text{Op}, \bar{x} \in L)$$

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the data flow instance  $\mathcal{B}$  obtained from  $\mathcal{A}$  by replacing  $f$  with  $h$  satisfies:

$$\overleftarrow{\text{MFP}}(\mathcal{A}) \sqsubseteq \overleftarrow{\text{MFP}}(\mathcal{B}) \quad \overleftarrow{\text{MOP}}(\mathcal{A}) \sqsubseteq \overleftarrow{\text{MOP}}(\mathcal{B})$$

## Application to Induced Abstract Data Flow Instances

Replace  $f_{\text{op}}^\#$  with a simpler monotonic  $\overline{h_{\text{op}}}$  verifying

$$f_{\text{op}}^\#(\overline{x}) \sqsupseteq \overline{h_{\text{op}}}(\overline{x}) \quad (\text{for all } \text{op} \in \text{Op}, \overline{x} \in L)$$

# Correctness of Approximate Transfer Mapping

Consider a data flow instance  $\mathcal{A}$  with a set  $\mathcal{F}$  of transfer functions and a transfer mapping  $f : \text{Op} \rightarrow \mathcal{F}$ .

For any monotonic function  $h : \text{Op} \rightarrow \mathcal{F}$  verifying

$$f_{\text{op}}(x) \sqsubseteq h_{\text{op}}(x) \quad (\text{for all } \text{op} \in \text{Op}, x \in L)$$

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# Design of Approximate Transfer Mapping

Given a Galois connection  $(\mathcal{P}(X \rightarrow \mathbb{R}), \subseteq) \xrightleftharpoons[\alpha]{\gamma} (\bar{L}, \bar{E})$  the resulting abstract data flow instance is obtained systematically.

But in practice,  $f^\#$  is rarely used: an approximate transfer mapping is required.

Tradeoff between computational cost and precision: many possibilities!

General principle: exploit the structure operations

- 1 define an abstract conservative semantics for arithmetic operators and comparators, ideally the most precise one
- 2 derive inductively an abstract semantics for operations, as usual

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# Sign Analysis: Abstract Arithmetic Operators

$$(\mathcal{P}(\mathbb{R}), \subseteq) \begin{array}{c} \xleftarrow{\gamma_{\text{sign}}} \\ \xrightarrow{\alpha_{\text{sign}}} \end{array} (\text{Sign}, \bar{\subseteq})$$

## Extension of Arithmetic Operators to Subsets of $\mathbb{R}$

For each function  $*$   $\in$   $\{+, -, \times, \dots\}$  from  $\mathbb{R} \times \mathbb{R}$  to  $\mathbb{R}$ , define the function  $*$  :  $(\mathcal{P}(\mathbb{R}) \times \mathcal{P}(\mathbb{R})) \rightarrow \mathcal{P}(\mathbb{R})$  by:

$$U * V = \{u * v \mid u \in U, v \in V\}$$

## Abstract Arithmetic Operators

Define the best abstraction  $*^{\#}$  :  $(\text{Sign} \times \text{Sign}) \rightarrow \text{Sign}$  of each function  $*$   $\in$   $\{+, -, \times, \dots\}$  by:

$$\bar{x} *^{\#} \bar{y} = \alpha_{\text{sign}} (\gamma_{\text{sign}}(\bar{x}) * \gamma_{\text{sign}}(\bar{y}))$$

# Abstract Arithmetic Operators: Table for $+^\#$

$$\bar{x} +^\# \bar{y} = \alpha_{sign} (\gamma_{sign}(\bar{x}) + \gamma_{sign}(\bar{y}))$$

$+^\#$	$\perp$	$-$	$0$	$+$	$-0$	$0+$	$\top$
$\perp$							
$-$	$\perp$	$-$	$-$	$\top$	$-$	$\top$	$\top$
$0$	$\perp$	$-$	$0$	$+$	$-0$	$0+$	$\top$
$+$	$\perp$	$\top$	$+$	$+$	$\top$	$+$	$\top$
$-0$	$\perp$	$-$	$-0$	$\top$	$-0$	$\top$	$\top$
$0+$	$\perp$	$\top$	$0+$	$+$	$\top$	$0+$	$\top$
$\top$	$\perp$	$\top$	$\top$	$\top$	$\top$	$\top$	$\top$

After mechanical inspection of all cases, we derive the above table.

We can derive similar tables for  $-^\#$  and  $\times^\#$ .

# Sign Analysis: Abstract Semantics of Expressions

$$(\mathcal{P}(X \rightarrow \mathbb{R}), \subseteq) \xrightleftharpoons[\alpha_\pi]{\gamma_\pi} (X \rightarrow \mathcal{P}(\mathbb{R}), \subseteq) \xrightleftharpoons[\alpha_{sign}]{\gamma_{sign}} (X \rightarrow \mathbf{Sign}, \bar{\subseteq})$$

For any abstract valuation  $\bar{v} : X \rightarrow \mathbf{Sign}$ , define  $\bar{\llbracket} e \rrbracket_{\bar{v}}$  inductively:

$$\begin{aligned}\bar{\llbracket} c \rrbracket_{\bar{v}} &= \alpha_{sign}(\{c\}) & [c \in \mathbb{Q}] \\ \bar{\llbracket} x \rrbracket_{\bar{v}} &= \bar{v}(x) & [x \in X] \\ \bar{\llbracket} e_1 + e_2 \rrbracket_{\bar{v}} &= \bar{\llbracket} e_1 \rrbracket_{\bar{v}} +^\# \bar{\llbracket} e_2 \rrbracket_{\bar{v}} \\ \bar{\llbracket} e_1 - e_2 \rrbracket_{\bar{v}} &= \bar{\llbracket} e_1 \rrbracket_{\bar{v}} -^\# \bar{\llbracket} e_2 \rrbracket_{\bar{v}} \\ \bar{\llbracket} e_1 * e_2 \rrbracket_{\bar{v}} &= \bar{\llbracket} e_1 \rrbracket_{\bar{v}} \times^\# \bar{\llbracket} e_2 \rrbracket_{\bar{v}}\end{aligned}$$

Fact (Conservative Approximation)

$$\bar{\llbracket} e \rrbracket(\bar{v}) \bar{\subseteq} \alpha_{sign} \circ \llbracket e \rrbracket \circ \gamma(\bar{v})$$

# Sign Analysis: Abstract Semantics of Expressions

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# Sign Analysis: Abstract Arithmetic Comparators

$$(\mathcal{P}(\mathbb{R}), \subseteq) \begin{array}{c} \xleftarrow{\gamma_{\text{sign}}} \\ \xrightarrow{\alpha_{\text{sign}}} \end{array} (\text{Sign}, \bar{\subseteq})$$

## Extension of Arithmetic Comparators to Subsets of $\mathbb{R}$

For each binary relation  $\bowtie \in \{<, \leq, =, \neq, >, \geq, \dots\}$  on  $\mathbb{R}$ , define the binary relation  $\bowtie$  on  $\mathcal{P}(\mathbb{R})$  by:

$$U \bowtie V \quad \text{if} \quad u \bowtie v \quad \text{for some } u \in U \text{ and } v \in V$$

## Abstract Arithmetic Comparators

Define the best abstraction  $\bowtie^\# \subseteq \text{Sign} \times \text{Sign}$  of each binary relation  $\bowtie \in \{<, \leq, =, \neq, >, \geq, \dots\}$  by:

$$\bar{x} \bowtie^\# \bar{y} \quad \text{if} \quad \gamma_{\text{sign}}(\bar{x}) \bowtie \gamma_{\text{sign}}(\bar{y})$$

# Abstract Arithmetic Comparators: Table for $<^\#$

$$\bar{x} <^\# \bar{y} \text{ if } \gamma_{\text{sign}}(\bar{x}) < \gamma_{\text{sign}}(\bar{y})$$

$<^\#$	$\perp$	$-$	$0$	$+$	$-0$	$0+$	$\top$
$\perp$							
$-$		•	•	•	•	•	•
$0$				•		•	•
$+$				•		•	•
$-0$		•	•	•	•	•	•
$0+$				•		•	•
$\top$		•	•	•	•	•	•

After mechanical inspection of all cases, we derive the above table.

We can derive similar tables for  $\leq^\#$ ,  $=^\#$ ,  $\neq^\#$ ,  $>^\#$ , and  $\geq^\#$ .

# Sign Analysis: Abstract Semantics of Guards

$$(\mathcal{P}(X \rightarrow \mathbb{R}), \sqsubseteq) \xrightleftharpoons[\alpha_\pi]{\gamma_\pi} (X \rightarrow \mathcal{P}(\mathbb{R}), \sqsubseteq) \xrightleftharpoons[\alpha_{\text{sign}}]{\gamma_{\text{sign}}} (X \rightarrow \text{Sign}, \bar{\sqsubseteq})$$

For any abstract valuation  $\bar{v} : X \rightarrow \text{Sign}$ , define  $\bar{v} \models g$  inductively:

$\bar{v} \models e_1 < e_2$	<i>if</i>	$\llbracket e_1 \rrbracket_{\bar{v}} <^\# \llbracket e_2 \rrbracket_{\bar{v}}$
$\bar{v} \models e_1 \leq e_2$	<i>if</i>	$\llbracket e_1 \rrbracket_{\bar{v}} \leq^\# \llbracket e_2 \rrbracket_{\bar{v}}$
$\bar{v} \models e_1 = e_2$	<i>if</i>	$\llbracket e_1 \rrbracket_{\bar{v}} =^\# \llbracket e_2 \rrbracket_{\bar{v}}$
$\bar{v} \models e_1 \neq e_2$	<i>if</i>	$\llbracket e_1 \rrbracket_{\bar{v}} \neq^\# \llbracket e_2 \rrbracket_{\bar{v}}$
$\bar{v} \models e_1 \geq e_2$	<i>if</i>	$\llbracket e_1 \rrbracket_{\bar{v}} \geq^\# \llbracket e_2 \rrbracket_{\bar{v}}$
$\bar{v} \models e_1 > e_2$	<i>if</i>	$\llbracket e_1 \rrbracket_{\bar{v}} >^\# \llbracket e_2 \rrbracket_{\bar{v}}$

Fact (Conservative Approximation)

$$\text{if } v \models g \text{ for some } v \in \gamma(\bar{v}) \text{ then } \bar{v} \models g$$

# Sign Analysis: Abstract Semantics of Guards

$$(\mathcal{P}(X \rightarrow \mathbb{R}), \sqsubseteq) \xrightleftharpoons[\alpha_\pi]{\gamma_\pi} (X \rightarrow \mathcal{P}(\mathbb{R}), \sqsubseteq) \xrightleftharpoons[\alpha_{\text{sign}}]{\gamma_{\text{sign}}} (X \rightarrow \text{Sign}, \bar{\sqsubseteq})$$

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$\bar{v} \models e_1 = e_2$	<i>if</i>	$\llbracket e_1 \rrbracket_{\bar{v}} =^\# \llbracket e_2 \rrbracket_{\bar{v}}$
$\bar{v} \models e_1 \neq e_2$	<i>if</i>	$\llbracket e_1 \rrbracket_{\bar{v}} \neq^\# \llbracket e_2 \rrbracket_{\bar{v}}$
$\bar{v} \models e_1 \geq e_2$	<i>if</i>	$\llbracket e_1 \rrbracket_{\bar{v}} \geq^\# \llbracket e_2 \rrbracket_{\bar{v}}$
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## Fact (Conservative Approximation)

*if*  $v \models g$  *for some*  $v \in \gamma(\bar{v})$  *then*  $\bar{v} \models g$

# Forward Sign Analysis: Approximate Transfer Mapping

$$(\mathcal{P}(X \rightarrow \mathbb{R}), \subseteq) \xrightleftharpoons[\alpha_\pi]{\gamma_\pi} (X \rightarrow \mathcal{P}(\mathbb{R}), \subseteq) \xrightleftharpoons[\alpha_{\text{sign}}]{\gamma_{\text{sign}}} (X \rightarrow \text{Sign}, \bar{\subseteq})$$

$$\overline{h_{x:=e}}(\bar{v}) = \lambda y. \begin{cases} \bar{v}(y) & \text{if } y \neq x \\ \llbracket e \rrbracket_{\bar{v}} & \text{if } y = x \end{cases} \quad \overline{h_g}(\bar{v}) = \begin{cases} \perp & \text{if } \bar{v} \not\models g \\ \bar{v} & \text{if } \bar{v} \models g \end{cases}$$

Fact (Conservative Approximation)

$$f_{\text{op}}^\#(\bar{x}) \bar{\subseteq} \overline{h_{\text{op}}}(\bar{x}) \quad (\text{for all } \text{op} \in \text{Op}, \bar{x} \in X \rightarrow \text{Sign})$$

Vs Transfer Mapping Previously Designed by Hand

- ☺ guaranteed to lead to a correct data flow analysis
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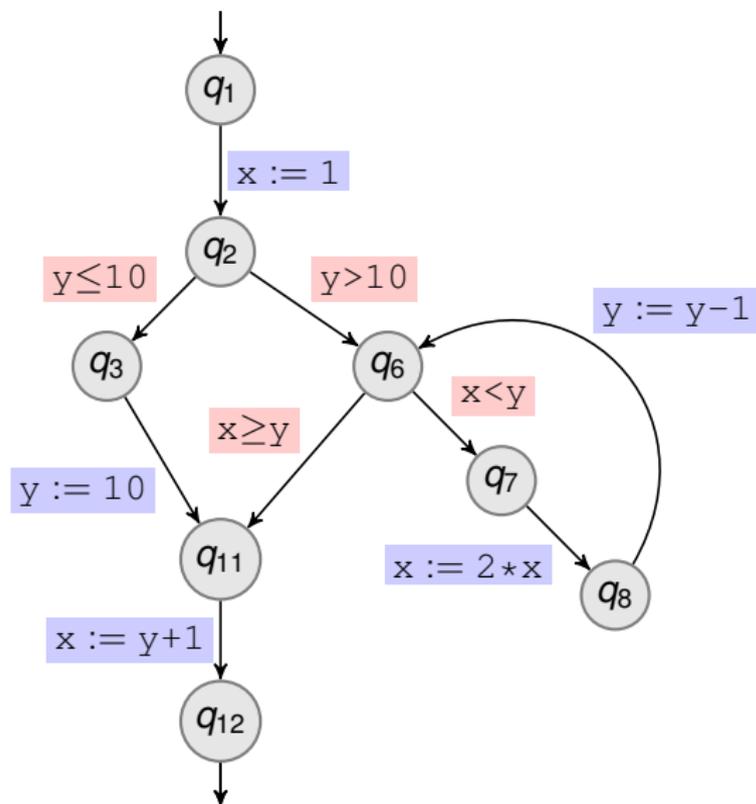
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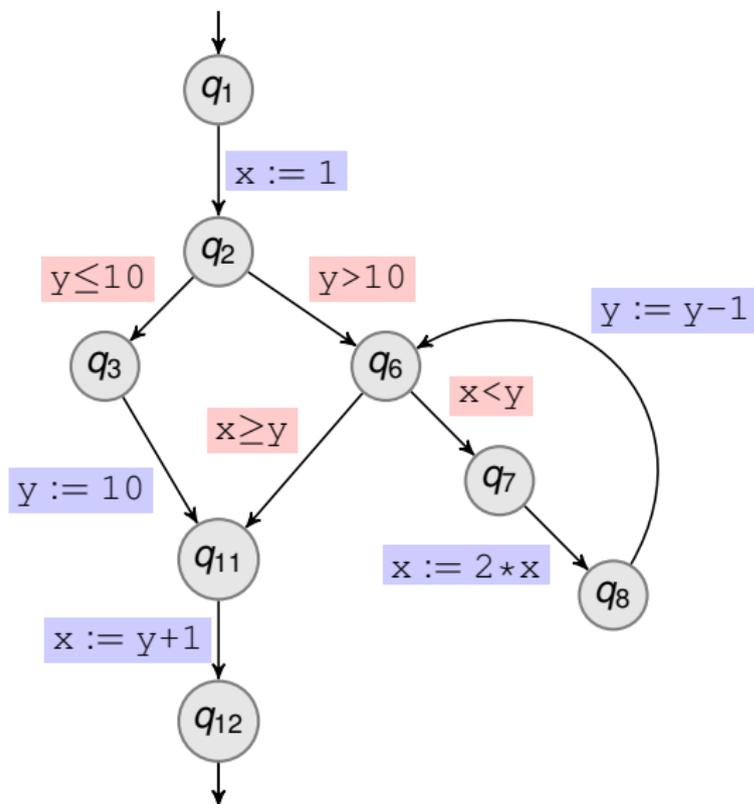
# Forward Sign Analysis on Running Example



Goal

Show that  $x > 0$  at  $q_{12}$

# Forward Sign Analysis on Running Example

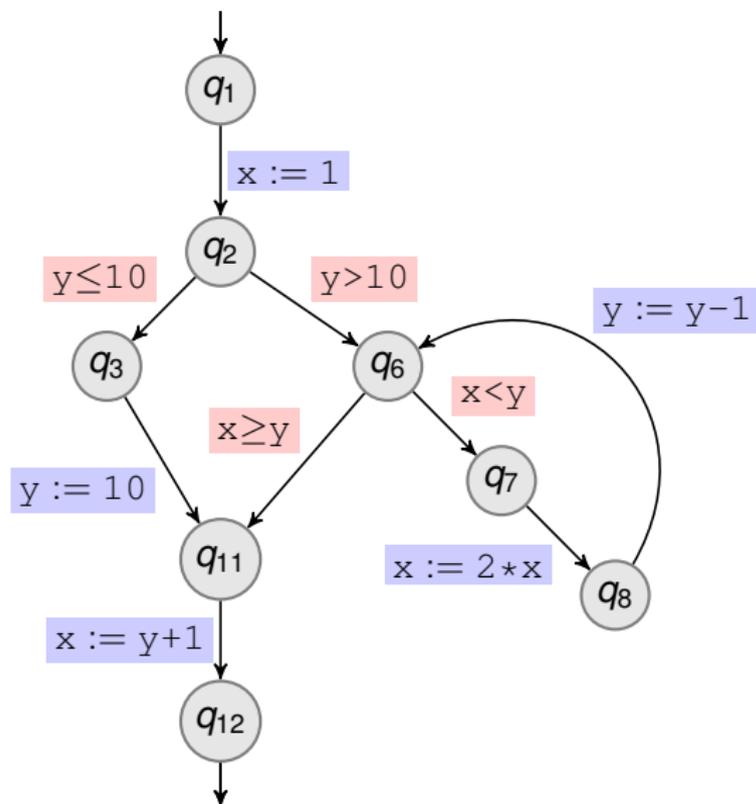


Goal

Show that  $x > 0$  at  $q_{12}$

	$x$	$y$
$q_1$	$\top$	$\top$
$q_2$	$\perp$	$\perp$
$q_3$	$\perp$	$\perp$
$q_6$	$\perp$	$\perp$
$q_7$	$\perp$	$\perp$
$q_8$	$\perp$	$\perp$
$q_{11}$	$\perp$	$\perp$
$q_{12}$	$\perp$	$\perp$

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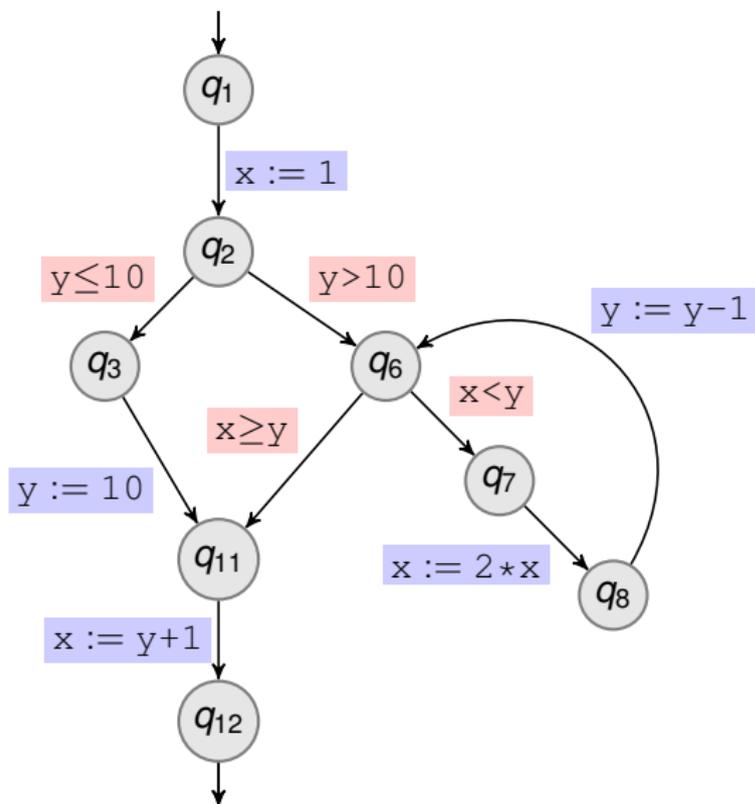


Goal

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	$x$	$y$
$q_1$	$\top$	$\top$
$q_2$	$+$	$\top$
$q_3$	$\perp$	$\perp$
$q_6$	$\perp$	$\perp$
$q_7$	$\perp$	$\perp$
$q_8$	$\perp$	$\perp$
$q_{11}$	$\perp$	$\perp$
$q_{12}$	$\perp$	$\perp$

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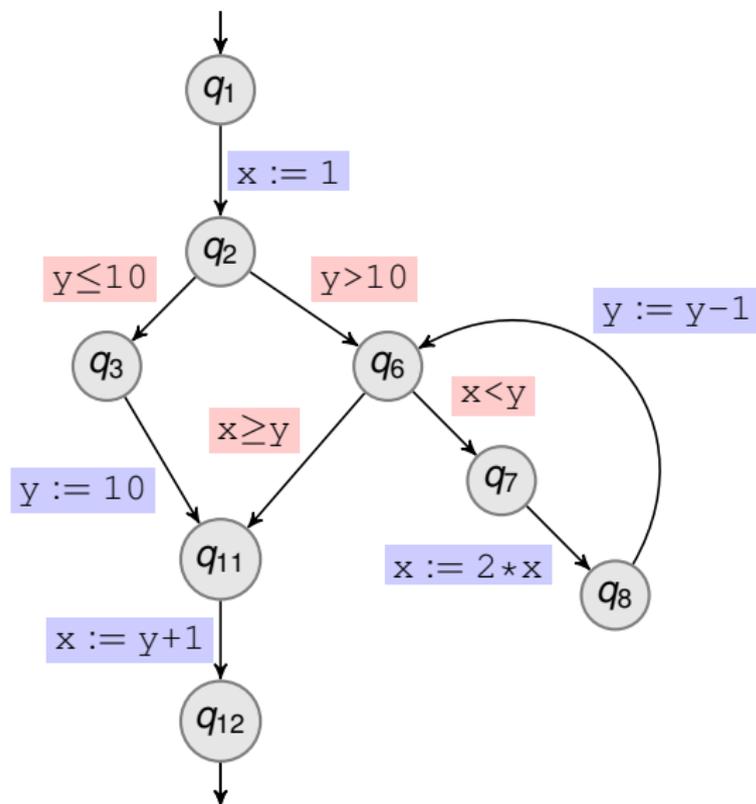


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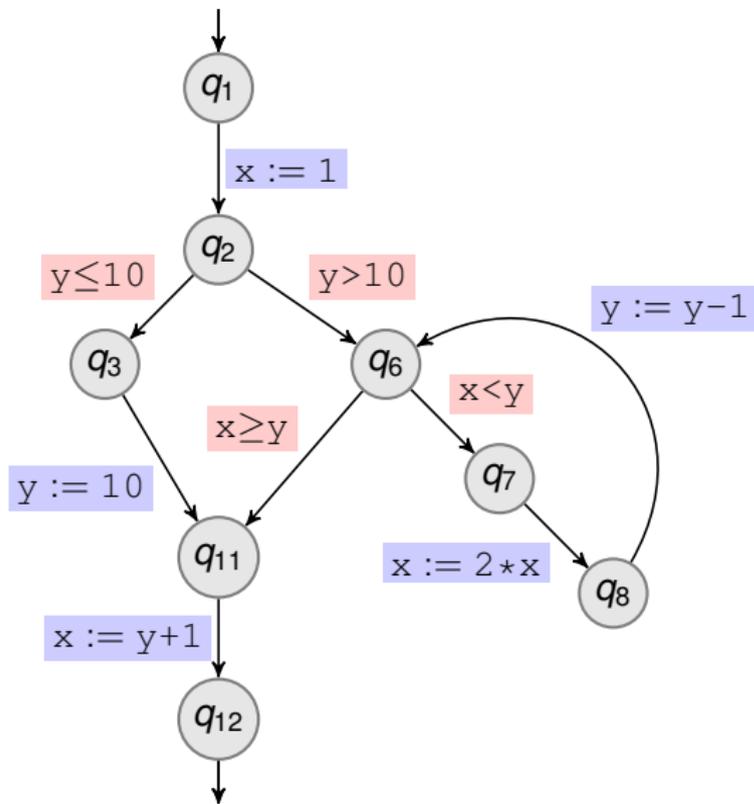


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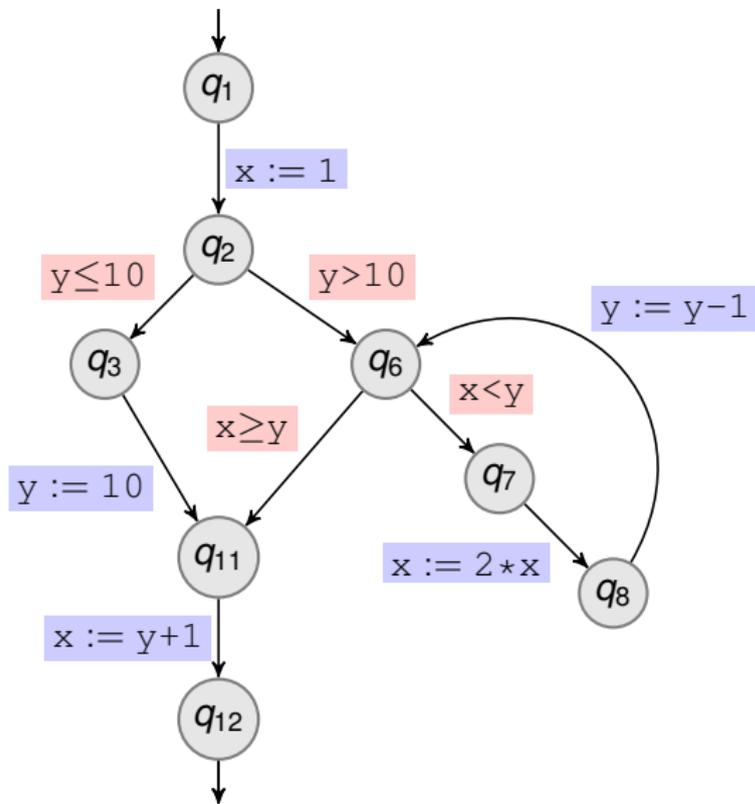


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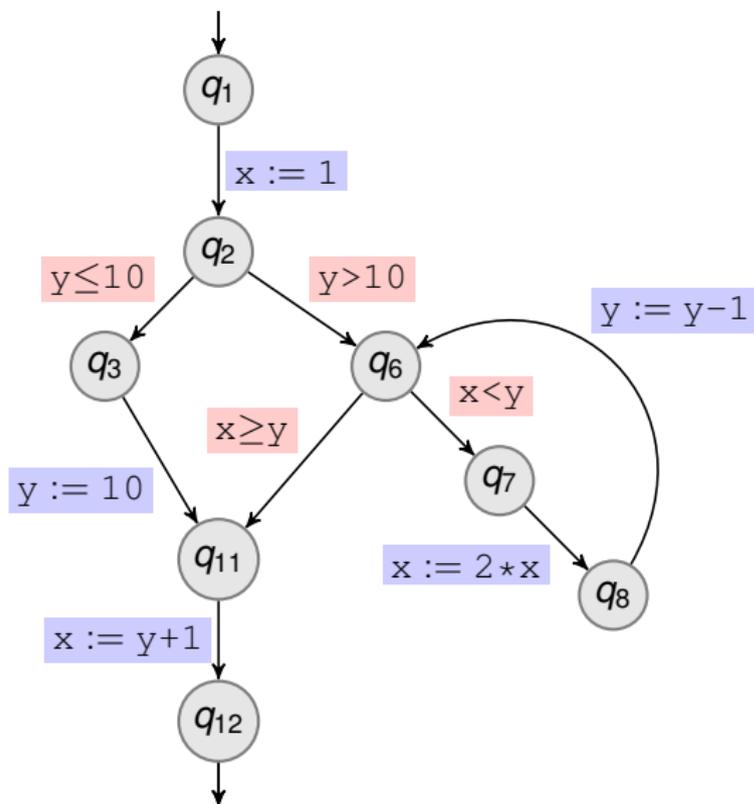


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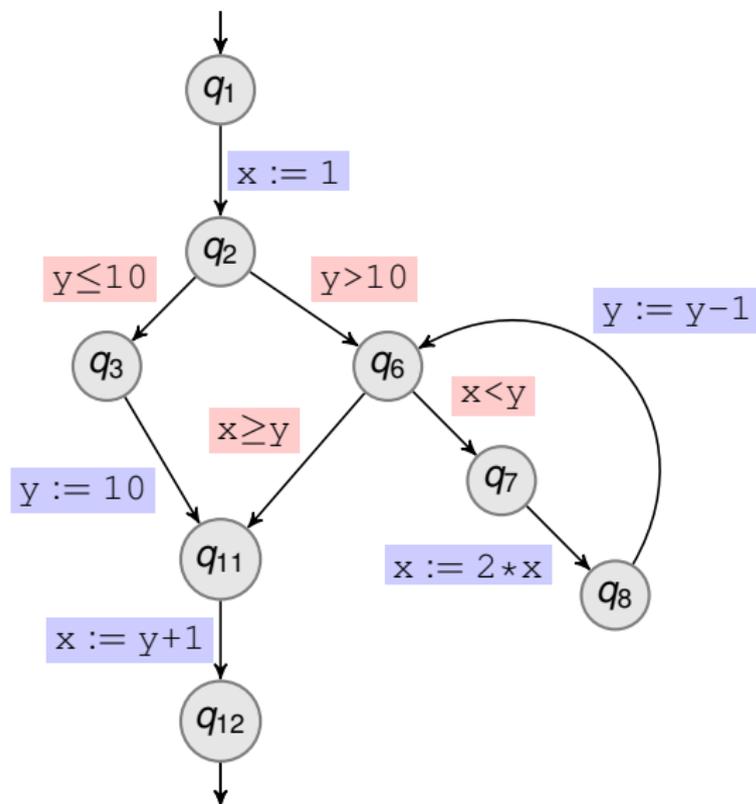


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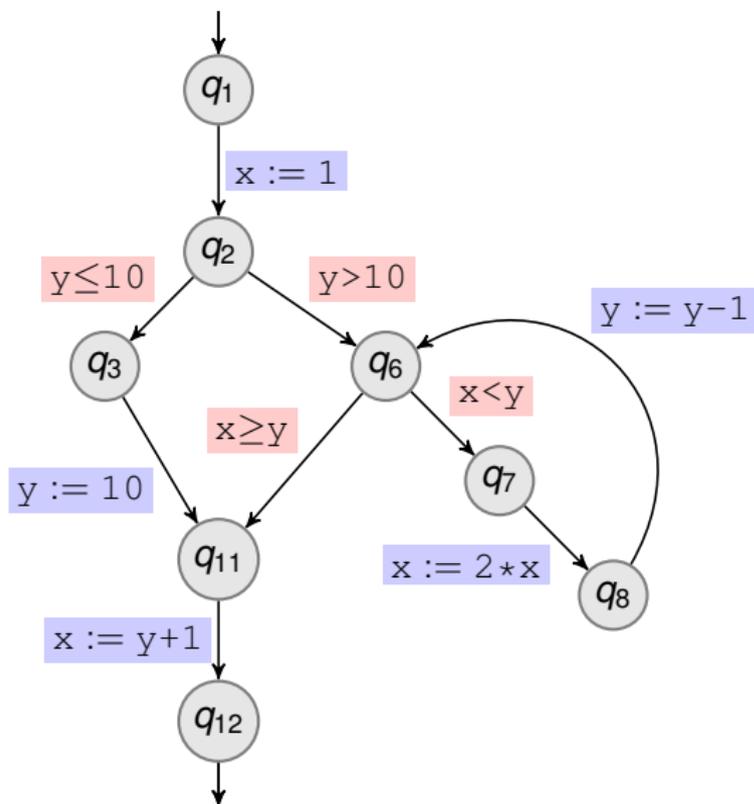


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$q_7$	$\perp$	$\perp$
$q_8$	$\perp$	$\perp$
$q_{11}$	$+$	$+$
$q_{12}$	$+$	$+$

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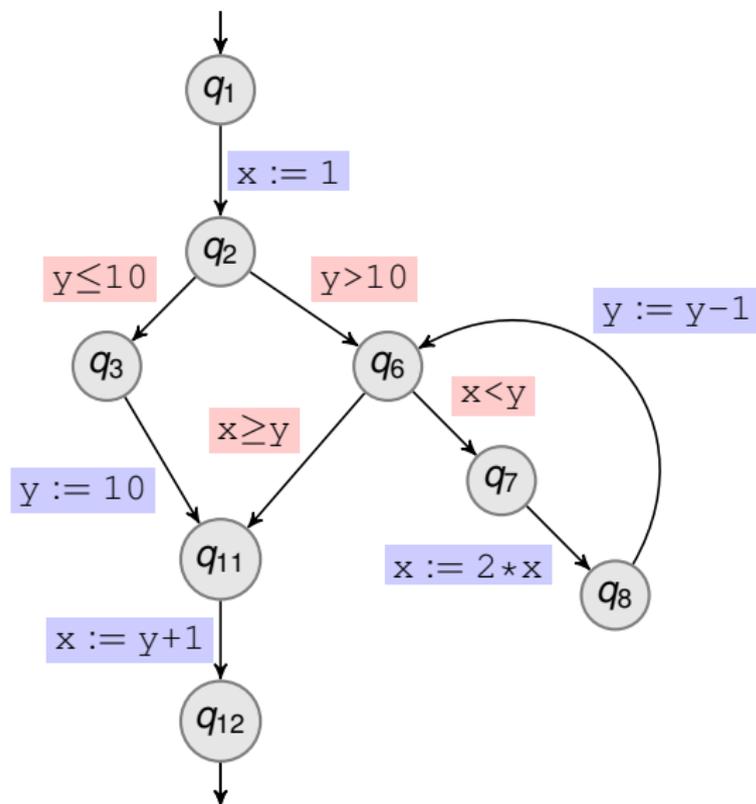


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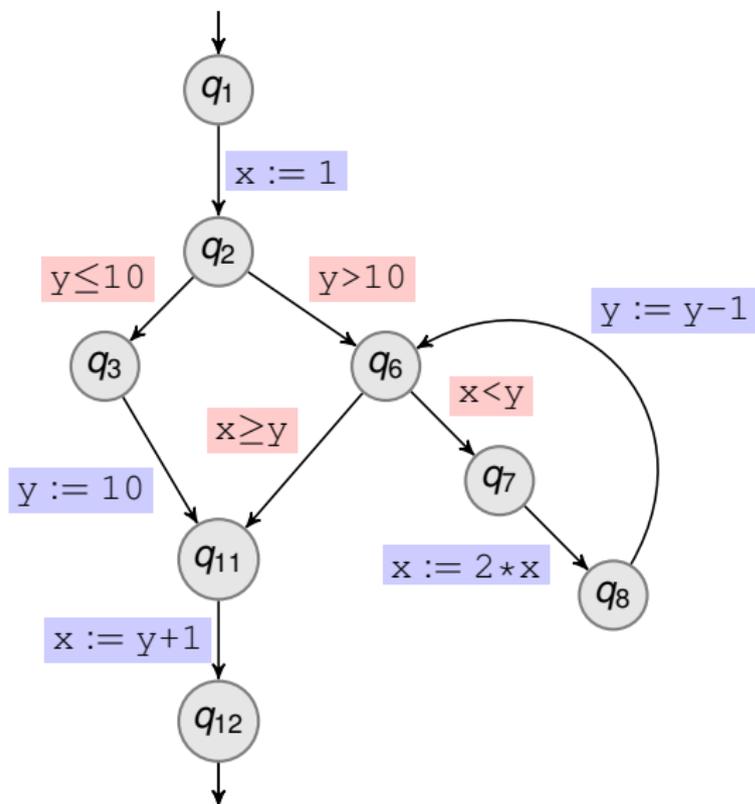


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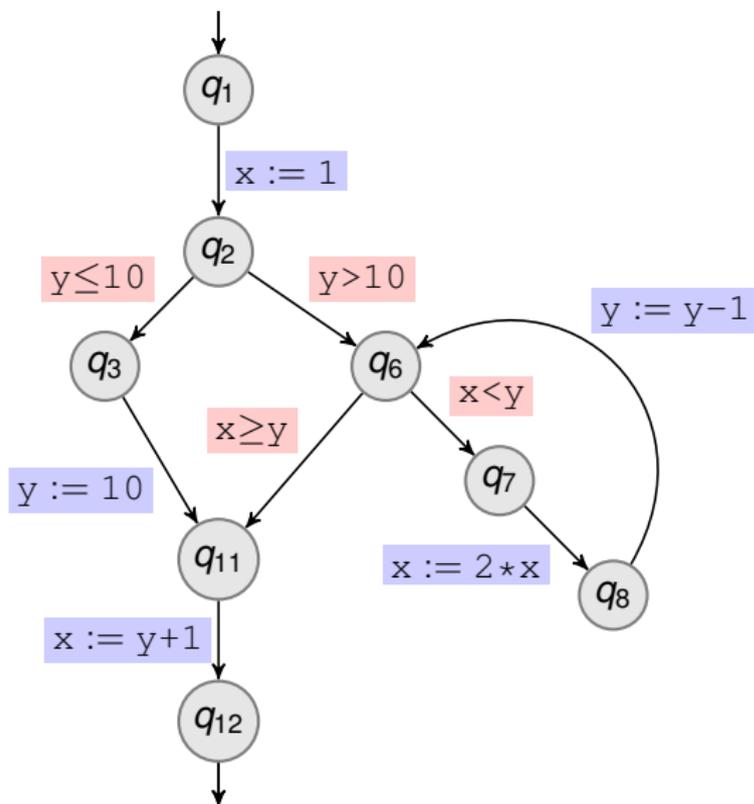


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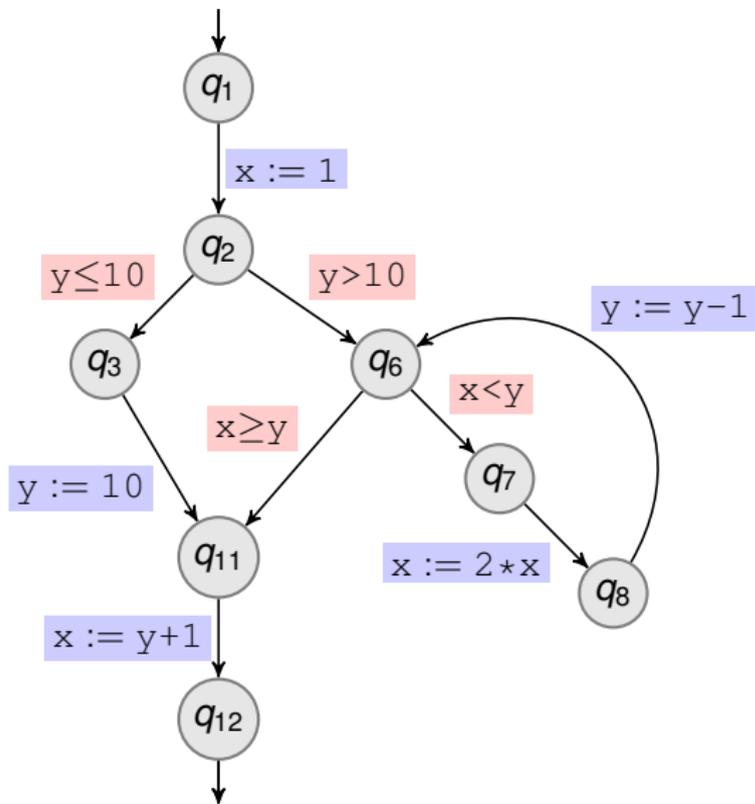


Goal

Show that  $x > 0$  at  $q_{12}$

	$x$	$y$
$q_1$	$\top$	$\top$
$q_2$	$+$	$\top$
$q_3$	$+$	$\top$
$q_6$	$+$	$\top$
$q_7$	$+$	$\top$
$q_8$	$\perp$	$\perp$
$q_{11}$	$+$	$+$
$q_{12}$	$+$	$+$

# Forward Sign Analysis on Running Example

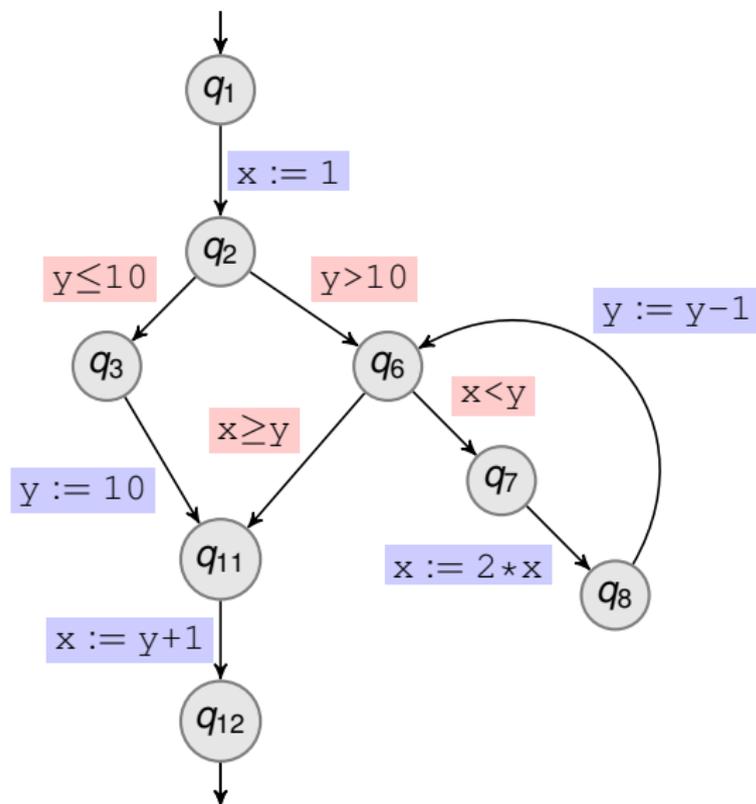


Goal

Show that  $x > 0$  at  $q_{12}$

	$x$	$y$
$q_1$	$\top$	$\top$
$q_2$	$+$	$\top$
$q_3$	$+$	$\top$
$q_6$	$+$	$\top$
$q_7$	$+$	$\top$
$q_8$	$\perp$	$\perp$
$q_{11}$	$+$	$+$
$q_{12}$	$+$	$+$

# Forward Sign Analysis on Running Example

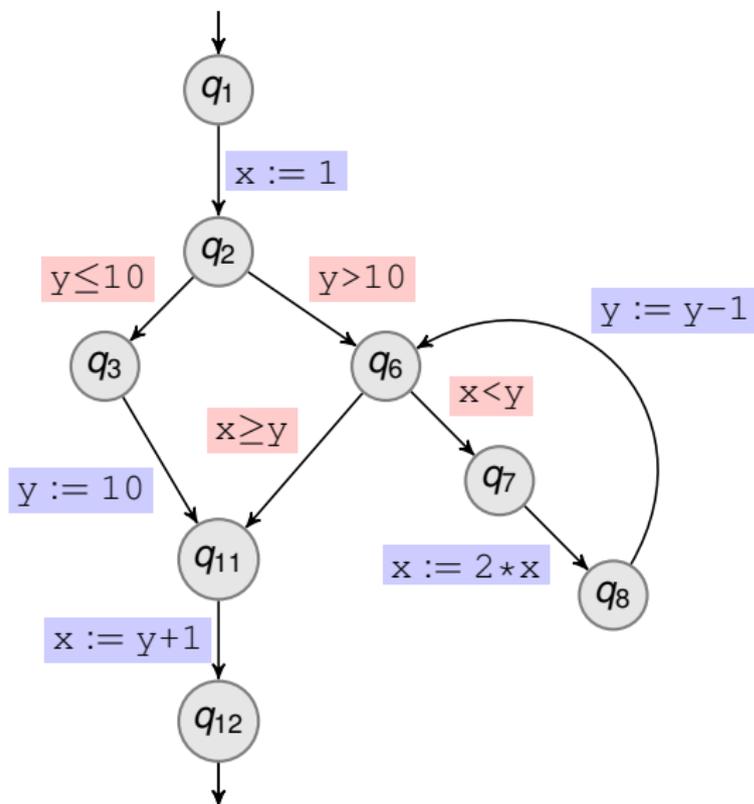


Goal

Show that  $x > 0$  at  $q_{12}$

	$x$	$y$
$q_1$	$\top$	$\top$
$q_2$	$+$	$\top$
$q_3$	$+$	$\top$
$q_6$	$+$	$\top$
$q_7$	$+$	$\top$
$q_8$	$+$	$\top$
$q_{11}$	$+$	$+$
$q_{12}$	$+$	$+$

# Forward Sign Analysis on Running Example

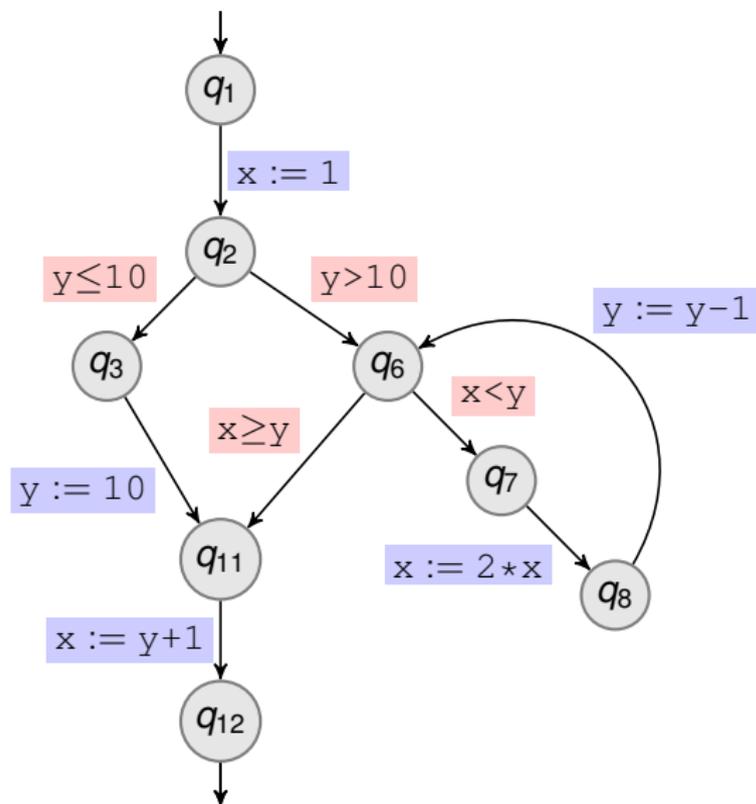


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$q_1$	$\top$	$\top$
$q_2$	$+$	$\top$
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$q_6$	$+$	$\top$
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$q_8$	$+$	$\top$
$q_{11}$	$+$	$+$
$q_{12}$	$+$	$+$

# Forward Sign Analysis on Running Example

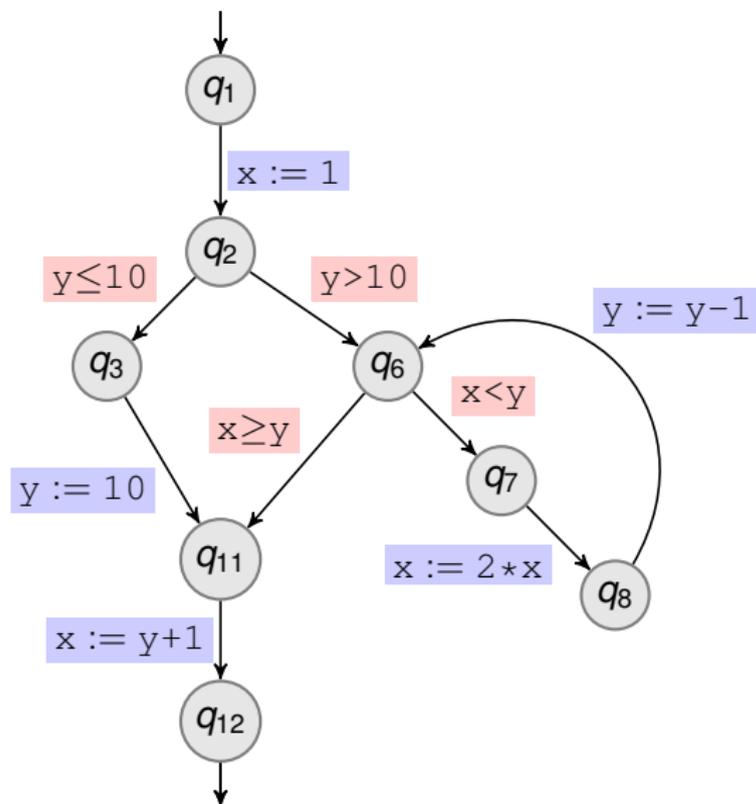


Goal

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$q_2$	$+$	$\top$
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$q_6$	$+$	$\top$
$q_7$	$+$	$\top$
$q_8$	$+$	$\top$
$q_{11}$	$+$	$\top$
$q_{12}$	$+$	$+$

# Forward Sign Analysis on Running Example

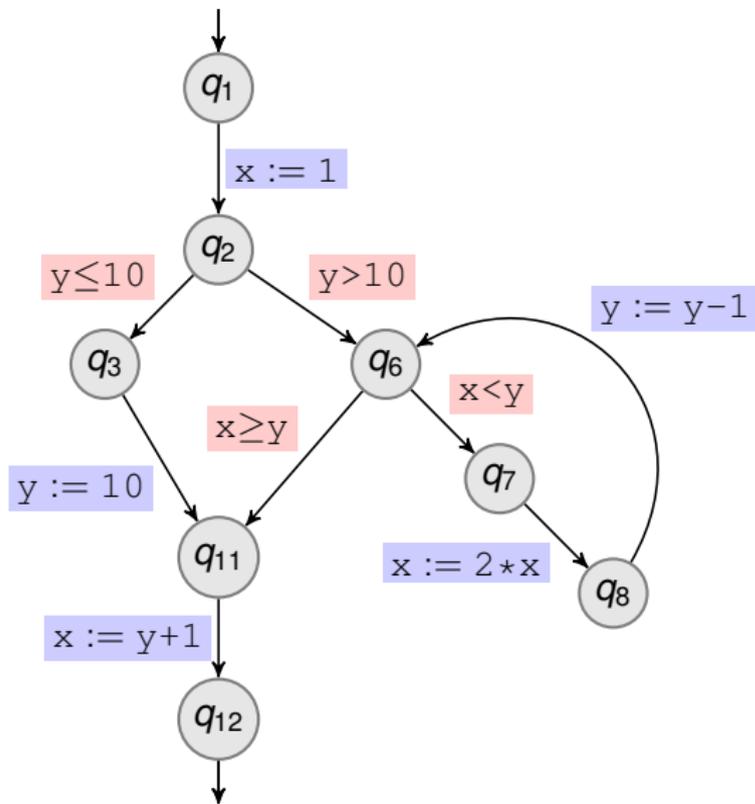


Goal

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	$x$	$y$
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$q_2$	$+$	$\top$
$q_3$	$+$	$\top$
$q_6$	$+$	$\top$
$q_7$	$+$	$\top$
$q_8$	$+$	$\top$
$q_{11}$	$+$	$\top$
$q_{12}$	$+$	$+$

# Forward Sign Analysis on Running Example

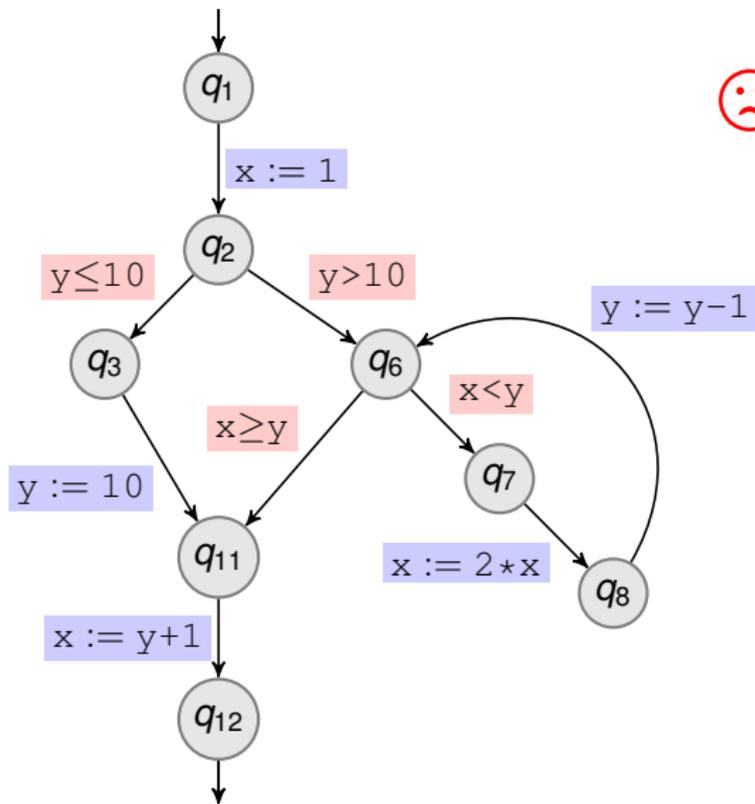


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Show that  $x > 0$  at  $q_{12}$

	$x$	$y$
$q_1$	$\top$	$\top$
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$q_3$	$+$	$\top$
$q_6$	$+$	$\top$
$q_7$	$+$	$\top$
$q_8$	$+$	$\top$
$q_{11}$	$+$	$\top$
$q_{12}$	$\top$	$\top$

# Forward Sign Analysis on Running Example



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$q_7$	$+$	$\top$
$q_8$	$+$	$\top$
$q_{11}$	$+$	$\top$
$q_{12}$	$\top$	$\top$

# Loss of Precision with Approximate Transfer Mapping

Example (Assignment  $op = x := z * z$ )

$$f_{op}^{\#}(\bar{T}) = \lambda y. \begin{cases} \bar{T} & \text{if } y \neq x \\ 0+ & \text{if } y = x \end{cases} \quad \overline{h_{op}}(\bar{T}) = \lambda y. \begin{cases} \bar{T} & \text{if } y \neq x \\ \bar{T} & \text{if } y = x \end{cases}$$

Indeed with  $\overline{h_{op}}$  the new value for  $x$  is:  $\llbracket z * z \rrbracket_{\bar{T}} = \llbracket z \rrbracket_{\bar{T}} \times^{\#} \llbracket z \rrbracket_{\bar{T}} = \bar{T}$ .

Example (Guard  $op = x = 0$ )

$$f_{op}^{\#}(\bar{T}) = \lambda y. \begin{cases} \bar{T} & \text{if } y \neq x \\ 0 & \text{if } y = x \end{cases} \quad \overline{h_{op}}(\bar{T}) = \lambda y. \begin{cases} \bar{T} & \text{if } y \neq x \\ \bar{T} & \text{if } y = x \end{cases}$$

Indeed  $\overline{h_{op}}(\bar{v})$  is either  $\bar{\perp}$  (if  $\bar{v} \not\models g$ ) or  $\bar{v}$ .

Example (Guard  $op = x > x$ )

$$f_{op}^{\#}(\bar{T}) = \bar{\perp} \quad \overline{h_{op}}(\bar{T}) = \bar{T}$$

# Loss of Precision with Approximate Transfer Mapping

Example (Assignment  $op = x := z * z$ )

$$f_{op}^{\#}(\bar{T}) = \lambda y. \begin{cases} \bar{T} & \text{if } y \neq x \\ 0_+ & \text{if } y = x \end{cases} \quad \overline{h_{op}}(\bar{T}) = \lambda y. \begin{cases} \bar{T} & \text{if } y \neq x \\ \bar{T} & \text{if } y = x \end{cases}$$

Indeed with  $\overline{h_{op}}$  the new value for  $x$  is:  $\llbracket z * z \rrbracket_{\bar{T}} = \llbracket z \rrbracket_{\bar{T}} \times^{\#} \llbracket z \rrbracket_{\bar{T}} = \bar{T}$ .

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$$f_{op}^{\#}(\bar{T}) = \lambda y. \begin{cases} \bar{T} & \text{if } y \neq x \\ 0 & \text{if } y = x \end{cases} \quad \overline{h_{op}}(\bar{T}) = \lambda y. \begin{cases} \bar{T} & \text{if } y \neq x \\ \bar{T} & \text{if } y = x \end{cases}$$

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# Loss of Precision with Approximate Transfer Mapping

Example (Assignment  $op = x := z * z$ )

$$f_{op}^{\#}(\bar{T}) = \lambda y. \begin{cases} \bar{T} & \text{if } y \neq x \\ 0_+ & \text{if } y = x \end{cases} \quad \overline{h_{op}}(\bar{T}) = \lambda y. \begin{cases} \bar{T} & \text{if } y \neq x \\ \bar{T} & \text{if } y = x \end{cases}$$

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Example (Guard  $op = x > x$ )

$$f_{op}^{\#}(\bar{T}) = \bar{\perp} \quad \overline{h_{op}}(\bar{T}) = \bar{T}$$

# Enhanced Precision with Functional Comparators

## Gain information from guards

### Functional Extension of Arithmetic Comparators to Subsets of $\mathbb{R}$

For each binary relation  $\bowtie \in \{<, \leq, =, \neq, >, \geq, \dots\}$  on  $\mathbb{R}$ , define the function  $\bowtie : (\mathcal{P}(\mathbb{R}) \times \mathcal{P}(\mathbb{R})) \rightarrow (\mathcal{P}(\mathbb{R}) \times \mathcal{P}(\mathbb{R}))$  by:

$$U \bowtie V = (\{u \in U \mid \exists v \in V, u \bowtie v\}, \{v \in V \mid \exists u \in U, u \bowtie v\})$$

### Functional Abstract Arithmetic Comparators

Define the best abstraction  $\bowtie^\# : (\text{Sign} \times \text{Sign}) \rightarrow (\text{Sign} \times \text{Sign})$  of each function  $\bowtie \in \{<, \leq, =, \neq, >, \geq, \dots\}$  by:

$$\bar{x} \bowtie^\# \bar{y} = (\alpha_{\text{sign}}(U), \alpha_{\text{sign}}(V))$$

where  $(U, V) = \gamma_{\text{sign}}(\bar{x}) \bowtie \gamma_{\text{sign}}(\bar{y})$

# Enhanced Precision with Functional Comparators

Gain information from guards

## Functional Extension of Arithmetic Comparators to Subsets of $\mathbb{R}$

For each binary relation  $\bowtie \in \{<, \leq, =, \neq, >, \geq, \dots\}$  on  $\mathbb{R}$ , define the function  $\bowtie : (\mathcal{P}(\mathbb{R}) \times \mathcal{P}(\mathbb{R})) \rightarrow (\mathcal{P}(\mathbb{R}) \times \mathcal{P}(\mathbb{R}))$  by:

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# Enhanced Precision with Functional Comparators

Gain information from guards

## Functional Extension of Arithmetic Comparators to Subsets of $\mathbb{R}$

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## Functional Abstract Arithmetic Comparators

Define the best abstraction  $\bowtie^\# : (\mathit{Sign} \times \mathit{Sign}) \rightarrow (\mathit{Sign} \times \mathit{Sign})$  of each function  $\bowtie \in \{<, \leq, =, \neq, >, \geq, \dots\}$  by:

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where  $(U, V) = \gamma_{\mathit{sign}}(\bar{x}) \bowtie \gamma_{\mathit{sign}}(\bar{y})$

# Functional Abstract Comparators: Table for $\leq^\#$

$$\bar{x} \bowtie^\# \bar{y} = (\alpha_{\text{sign}}(U), \alpha_{\text{sign}}(V))$$

where  $(U, V) = \gamma_{\text{sign}}(\bar{x}) \bowtie \gamma_{\text{sign}}(\bar{y})$

$\leq^\#$	$\perp$	$-$	$0$	$+$	$-0$	$0+$	$\top$
$\perp$	$(\perp, \perp)$						
$-$	$(\perp, \perp)$	$(-, -)$	$(-, 0)$	$(-, +)$	$(-, -0)$	$(-, 0+)$	$(-, \top)$
$0$	$(\perp, \perp)$	$(\perp, \perp)$	$(0, 0)$	$(0, +)$	$(0, 0)$	$(0, 0+)$	$(0, \top)$
$+$	$(\perp, \perp)$	$(\perp, \perp)$	$(\perp, \perp)$	$(+, +)$	$(\perp, \perp)$	$(+, 0+)$	$(+, \top)$
$-0$	$(\perp, \perp)$	$(-, -)$	$(-0, 0)$	$(-0, +)$	$(-0, -0)$	$(-0, 0+)$	$(-0, \top)$
$0+$	$(\perp, \perp)$	$(\perp, \perp)$	$(0, 0)$	$(0+, +)$	$(0, 0)$	$(0+, 0+)$	$(0+, \top)$
$\top$	$(\perp, \perp)$	$(-, -)$	$(-0, 0)$	$(\top, +)$	$(-0, -0)$	$(\top, 0+)$	$(\top, \top)$

After mechanical inspection of all cases, we derive the above table.

We can derive similar tables for  $\leq^\#, =^\#, \neq^\#, >^\#,$  and  $\geq^\#$ .

# Enhanced Approximate Transfer Mapping

$$\overline{h_{x:=e}}(\bar{v}) = \lambda y. \begin{cases} \bar{v}(y) & \text{if } y \neq x \\ \llbracket e \rrbracket_{\bar{v}} & \text{if } y = x \end{cases} \quad \overline{h_g}(\bar{v}) = \begin{cases} \bar{\perp} & \text{if } \bar{v} \not\models g \\ \theta_g(\bar{v}) & \text{if } \bar{v} \models g \end{cases}$$

$$g = x \bowtie x$$

$$\overline{\theta_g}(\bar{v}) = \lambda y. \begin{cases} \bar{v}(y) & \text{if } \bowtie \in \{=, \leq, \geq\} \\ \bar{\perp} & \text{if } \bowtie \in \{\neq, <, >\} \end{cases}$$

$$g = x_1 \bowtie x_2 \quad \text{with } x_1 \neq x_2$$

$$\overline{\theta_g}(\bar{v}) = \lambda y. \begin{cases} \bar{t}_1 & \text{if } y = x_1 \\ \bar{t}_2 & \text{if } y = x_2 \\ \bar{v}(y) & \text{otherwise} \end{cases} \quad \text{where } (\bar{t}_1, \bar{t}_2) = \llbracket x_1 \rrbracket_{\bar{v}} \bowtie^{\sharp} \llbracket x_2 \rrbracket_{\bar{v}}$$

# Enhanced Approximate Transfer Mapping

$$\overline{h_{x:=e}}(\bar{v}) = \lambda y. \begin{cases} \bar{v}(y) & \text{if } y \neq x \\ \llbracket e \rrbracket_{\bar{v}} & \text{if } y = x \end{cases} \quad \overline{h_g}(\bar{v}) = \begin{cases} \bar{\perp} & \text{if } \bar{v} \not\models g \\ \overline{\theta_g}(\bar{v}) & \text{if } \bar{v} \models g \end{cases}$$

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$$g = x_1 \bowtie x_2 \quad \text{with } x_1 \neq x_2$$

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# Enhanced Approximate Transfer Mapping

$$\overline{h_{x:=e}}(\bar{v}) = \lambda y. \begin{cases} \bar{v}(y) & \text{if } y \neq x \\ \overline{[e]}_{\bar{v}} & \text{if } y = x \end{cases} \quad \overline{h_g}(\bar{v}) = \begin{cases} \overline{\perp} & \text{if } \bar{v} \not\models g \\ \overline{\theta_g}(\bar{v}) & \text{if } \bar{v} \models g \end{cases}$$

$g = x \bowtie e$  with  $e$  not reduced to a variable

$$\overline{\theta_g}(\bar{v}) = \lambda y. \begin{cases} \bar{t} & \text{if } y = x \\ \bar{v}(y) & \text{otherwise} \end{cases} \quad \text{where } (\bar{t}, \_) = \overline{[x]}_{\bar{v}} \bowtie^{\#} \overline{[e]}_{\bar{v}}$$

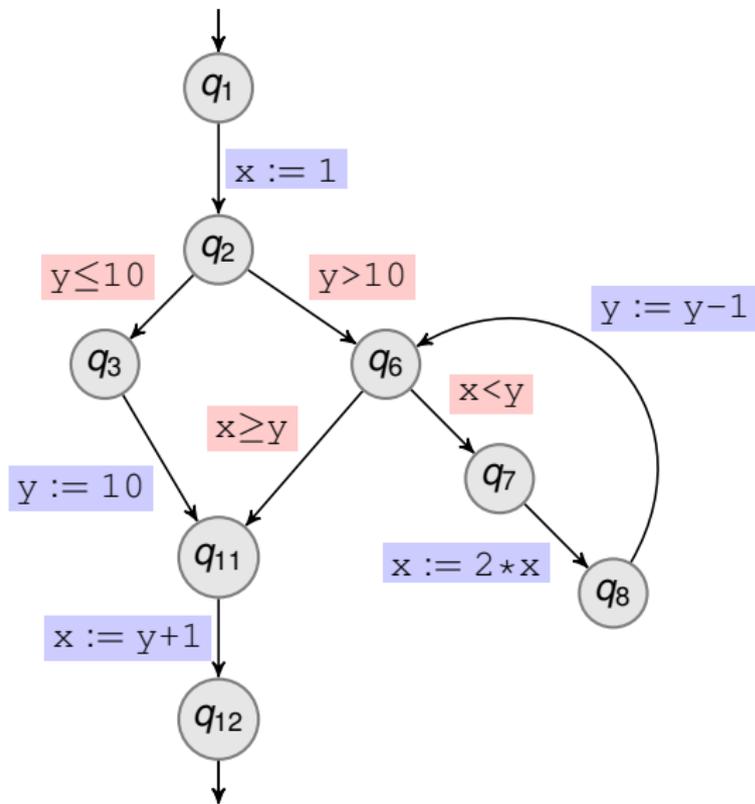
$g = e \bowtie x$  with  $e$  not reduced to a variable

$$\overline{\theta_g}(\bar{v}) = \lambda y. \begin{cases} \bar{t} & \text{if } y = x \\ \bar{v}(y) & \text{otherwise} \end{cases} \quad \text{where } (\_, \bar{t}) = \overline{[e]}_{\bar{v}} \bowtie^{\#} \overline{[x]}_{\bar{v}}$$

$g = e_1 \bowtie e_2$  with  $e_1, e_2$  not reduced to a variable

$$\overline{\theta_g}(\bar{v}) = \bar{v}$$

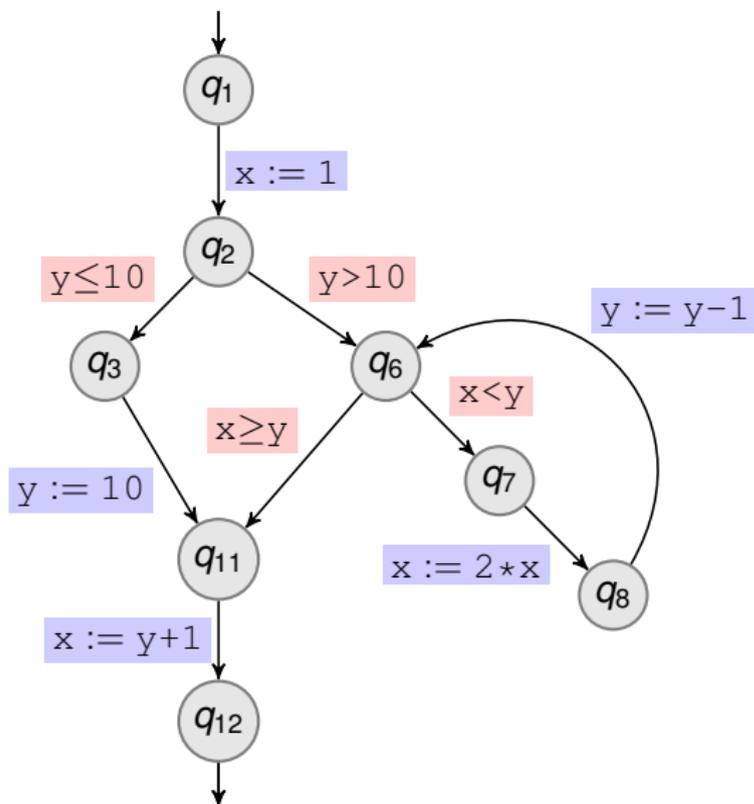
# Forward Sign Analysis on Example with Enhanced $\overline{h_{op}}$



Goal

Show that  $x > 0$  at  $q_{12}$

# Forward Sign Analysis on Example with Enhanced $\overline{h}_{op}$

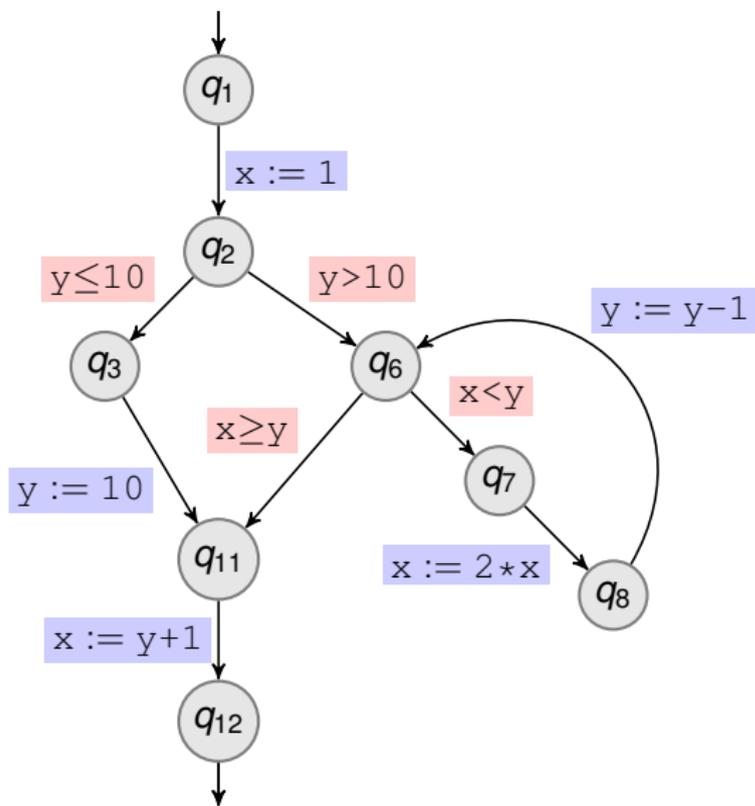


Goal

Show that  $x > 0$  at  $q_{12}$

	x	y
$q_1$	$\top$	$\top$
$q_2$	$\perp$	$\perp$
$q_3$	$\perp$	$\perp$
$q_6$	$\perp$	$\perp$
$q_7$	$\perp$	$\perp$
$q_8$	$\perp$	$\perp$
$q_{11}$	$\perp$	$\perp$
$q_{12}$	$\perp$	$\perp$

# Forward Sign Analysis on Example with Enhanced $\overline{h}_{op}$

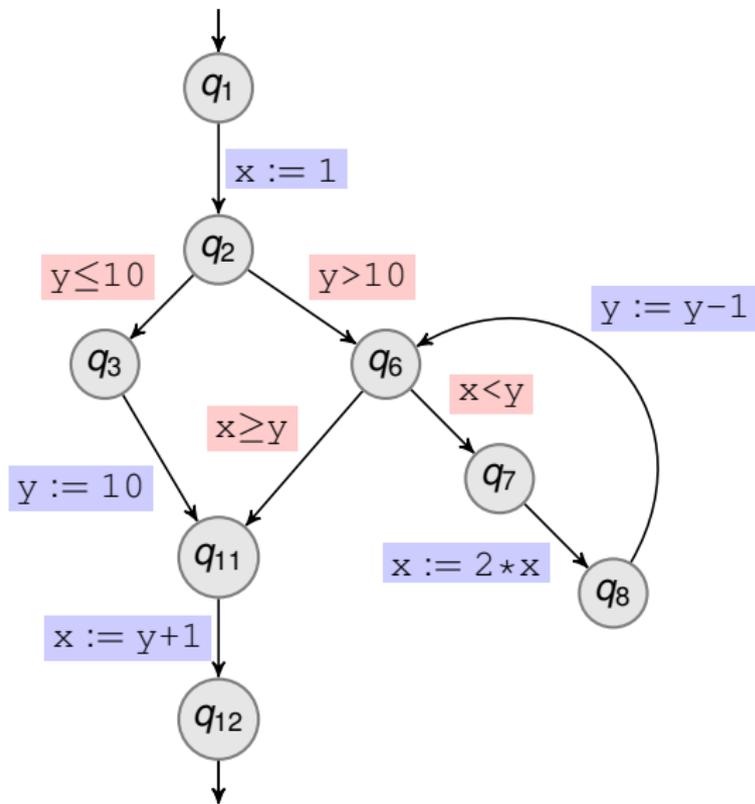


Goal

Show that  $x > 0$  at  $q_{12}$

	x	y
$q_1$	$\top$	$\top$
$q_2$	$+$	$\top$
$q_3$	$\perp$	$\perp$
$q_6$	$\perp$	$\perp$
$q_7$	$\perp$	$\perp$
$q_8$	$\perp$	$\perp$
$q_{11}$	$\perp$	$\perp$
$q_{12}$	$\perp$	$\perp$

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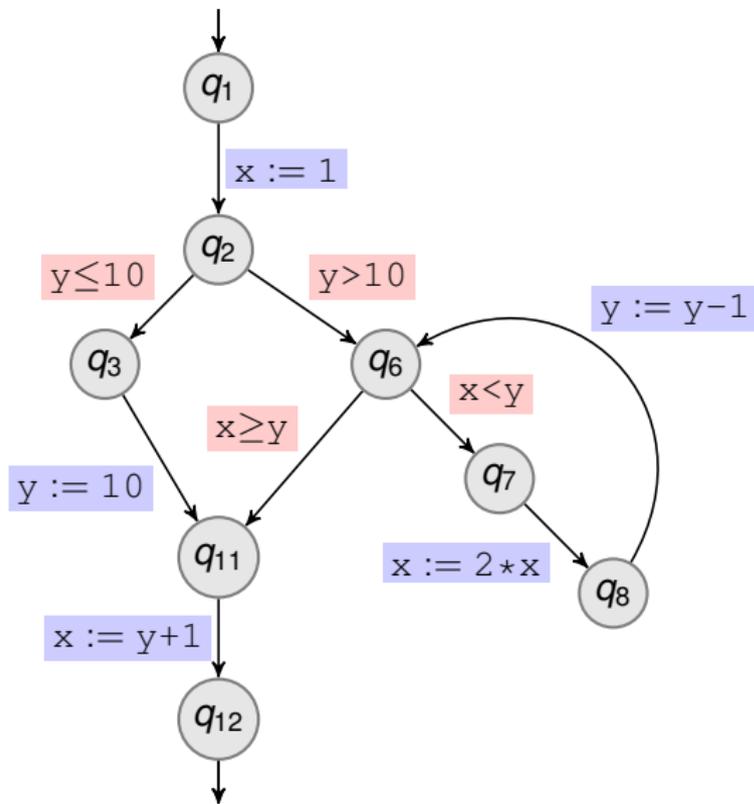


Goal

Show that  $x > 0$  at  $q_{12}$

	x	y
$q_1$	⊤	⊤
$q_2$	+	⊤
$q_3$	⊥	⊥
$q_6$	⊥	⊥
$q_7$	⊥	⊥
$q_8$	⊥	⊥
$q_{11}$	⊥	⊥
$q_{12}$	⊥	⊥

# Forward Sign Analysis on Example with Enhanced $\overline{h}_{op}$

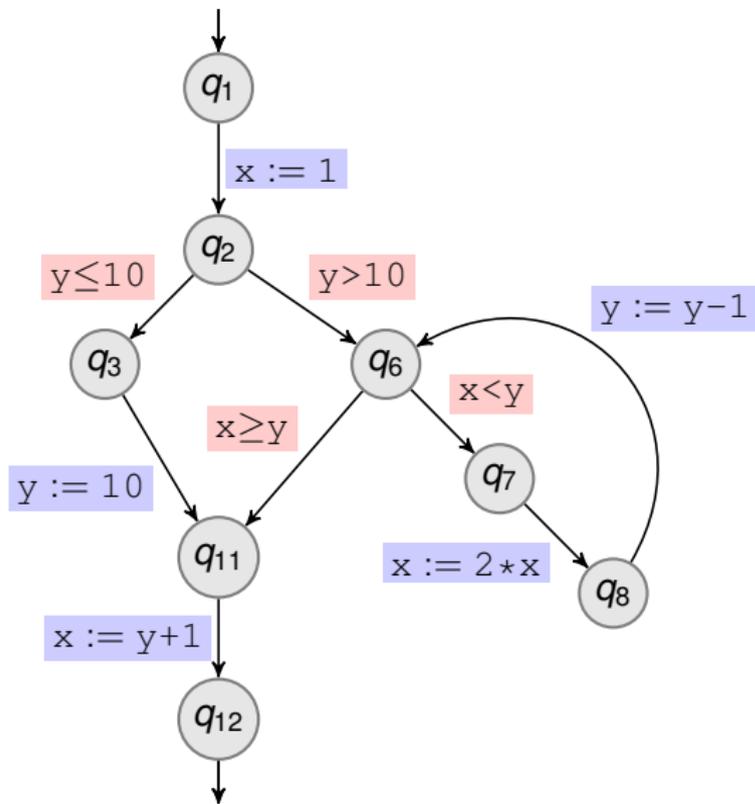


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$q_2$	+	⊥
$q_3$	+	⊥
$q_6$	⊥	⊥
$q_7$	⊥	⊥
$q_8$	⊥	⊥
$q_{11}$	⊥	⊥
$q_{12}$	⊥	⊥

# Forward Sign Analysis on Example with Enhanced $\overline{h}_{op}$

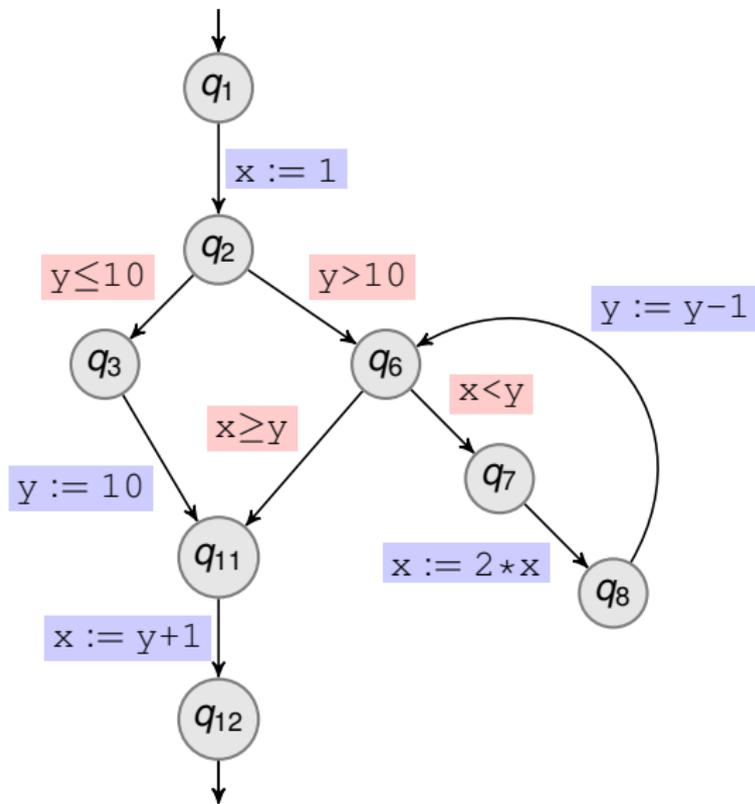


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$q_{11}$	$\perp$	$\perp$
$q_{12}$	$\perp$	$\perp$

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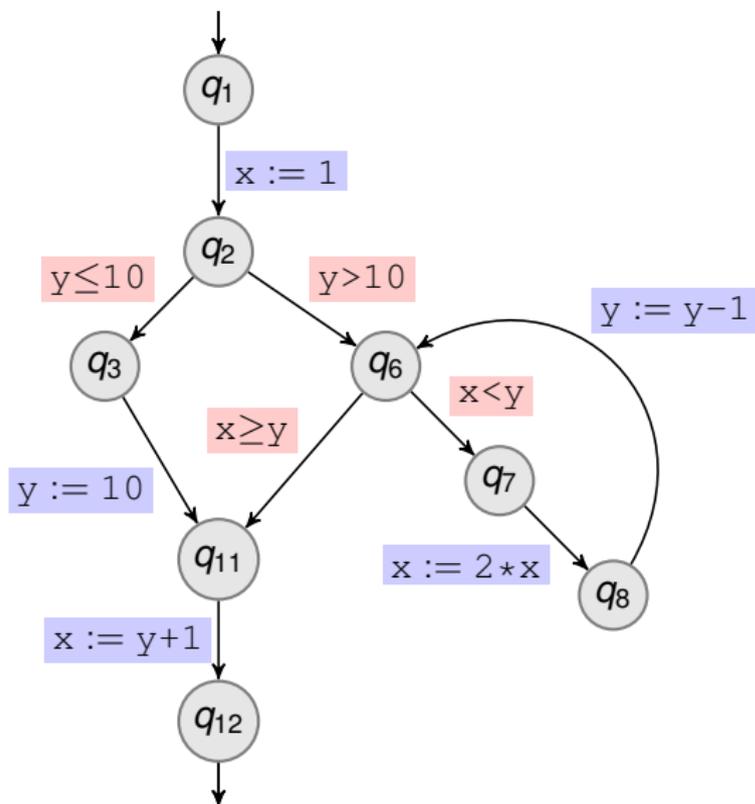


Goal

Show that  $x > 0$  at  $q_{12}$

	x	y
$q_1$	⊥	⊥
$q_2$	+	⊥
$q_3$	+	⊥
$q_6$	⊥	⊥
$q_7$	⊥	⊥
$q_8$	⊥	⊥
$q_{11}$	+	+
$q_{12}$	⊥	⊥

# Forward Sign Analysis on Example with Enhanced $\overline{h}_{op}$

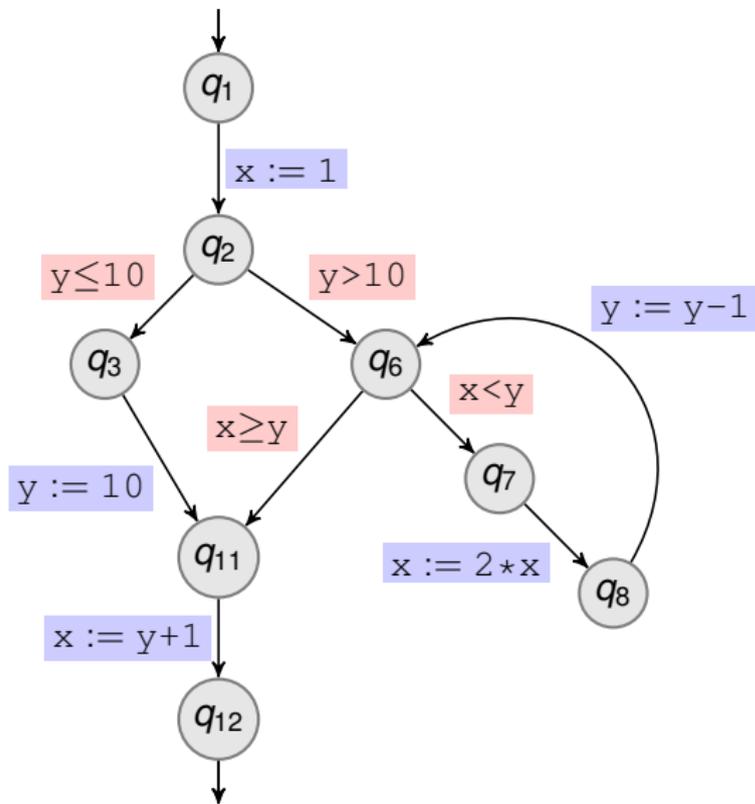


Goal

Show that  $x > 0$  at  $q_{12}$

	x	y
$q_1$	⊤	⊤
$q_2$	+	⊤
$q_3$	+	⊤
$q_6$	⊥	⊥
$q_7$	⊥	⊥
$q_8$	⊥	⊥
$q_{11}$	+	+
$q_{12}$	⊥	⊥

# Forward Sign Analysis on Example with Enhanced $\overline{h}_{op}$

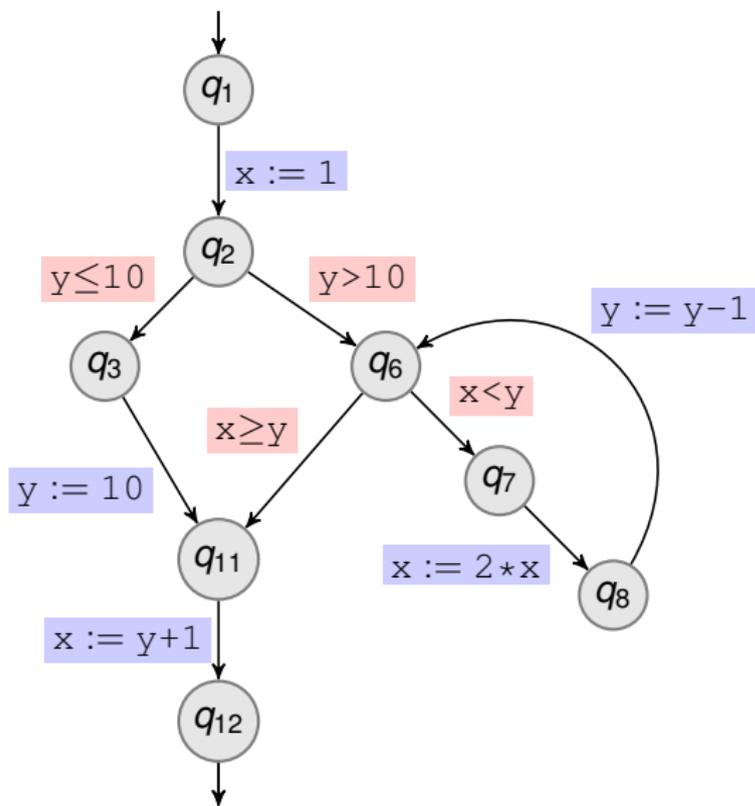


Goal

Show that  $x > 0$  at  $q_{12}$

	x	y
$q_1$	⊥	⊥
$q_2$	+	⊥
$q_3$	+	⊥
$q_6$	⊥	⊥
$q_7$	⊥	⊥
$q_8$	⊥	⊥
$q_{11}$	+	+
$q_{12}$	+	+

# Forward Sign Analysis on Example with Enhanced $\overline{h}_{op}$

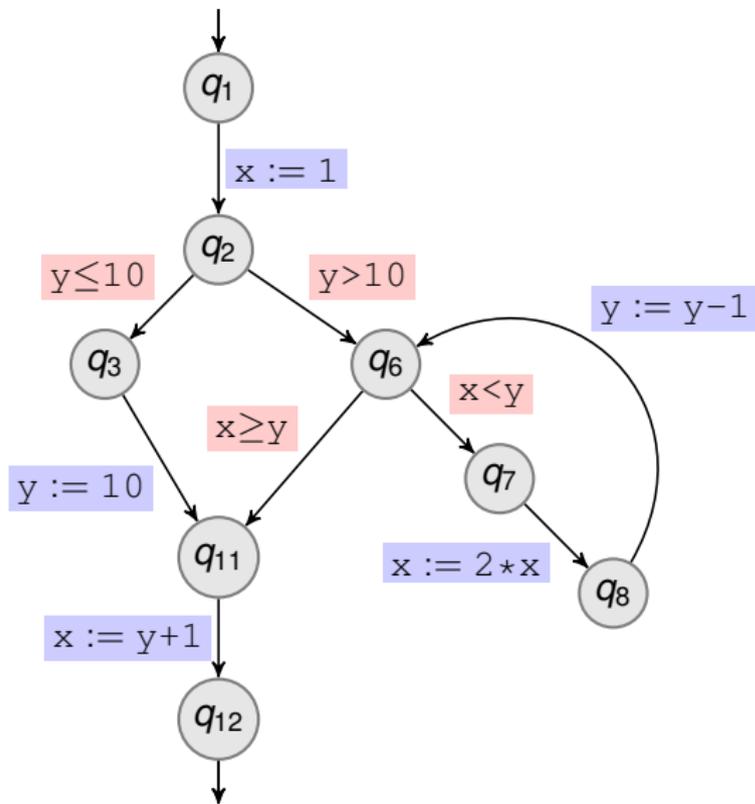


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Show that  $x > 0$  at  $q_{12}$

	$x$	$y$
$q_1$	$\top$	$\top$
$q_2$	$+$	$\top$
$q_3$	$+$	$\top$
$q_6$	$\perp$	$\perp$
$q_7$	$\perp$	$\perp$
$q_8$	$\perp$	$\perp$
$q_{11}$	$+$	$+$
$q_{12}$	$+$	$+$

# Forward Sign Analysis on Example with Enhanced $\overline{h}_{op}$

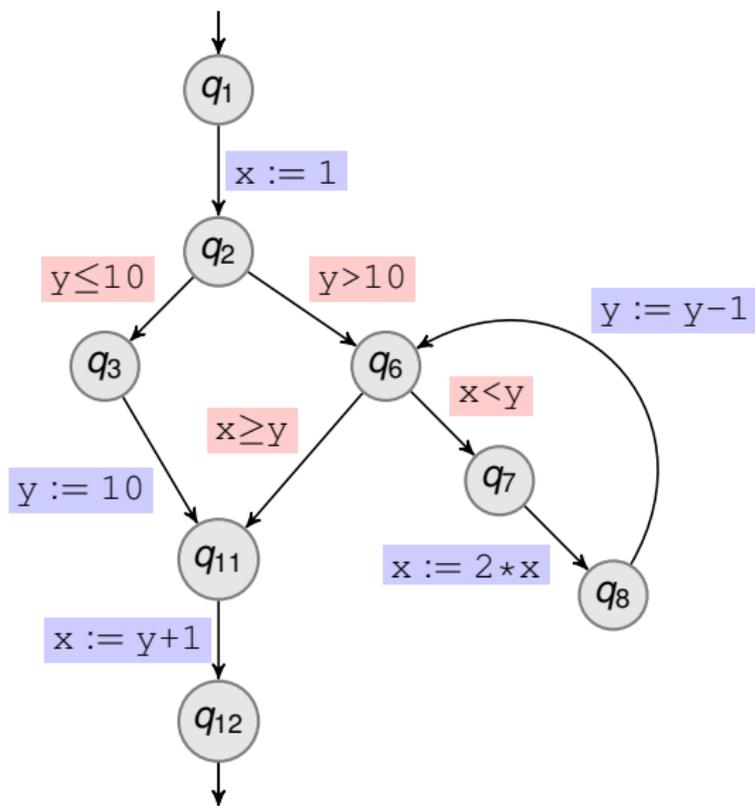


Goal

Show that  $x > 0$  at  $q_{12}$

	$x$	$y$
$q_1$	$\top$	$\top$
$q_2$	$+$	$\top$
$q_3$	$+$	$\top$
$q_6$	$+$	$+$
$q_7$	$\perp$	$\perp$
$q_8$	$\perp$	$\perp$
$q_{11}$	$+$	$+$
$q_{12}$	$+$	$+$

# Forward Sign Analysis on Example with Enhanced $\overline{h}_{op}$

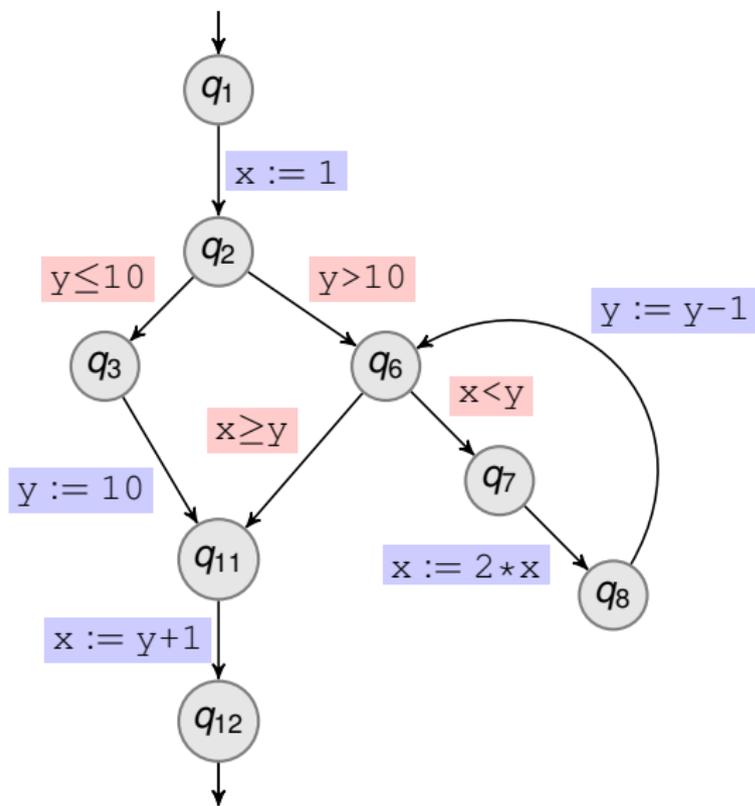


Goal

Show that  $x > 0$  at  $q_{12}$

	x	y
$q_1$	⊤	⊤
$q_2$	+	⊤
$q_3$	+	⊤
$q_6$	+	+
$q_7$	⊥	⊥
$q_8$	⊥	⊥
$q_{11}$	+	+
$q_{12}$	+	+

# Forward Sign Analysis on Example with Enhanced $\overline{h}_{op}$

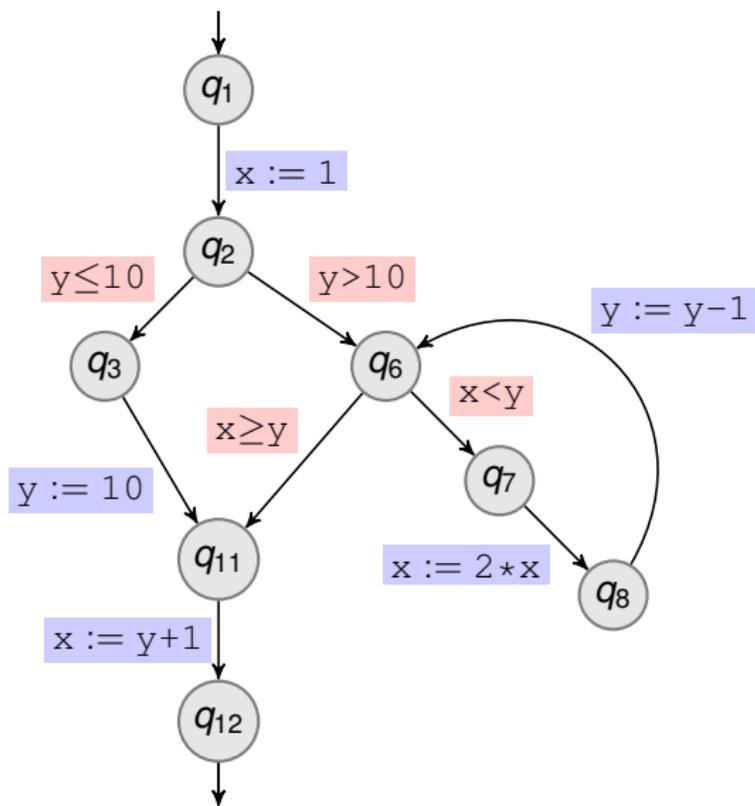


Goal

Show that  $x > 0$  at  $q_{12}$

	x	y
$q_1$	⊥	⊥
$q_2$	+	⊥
$q_3$	+	⊥
$q_6$	+	+
$q_7$	+	+
$q_8$	⊥	⊥
$q_{11}$	+	+
$q_{12}$	+	+

# Forward Sign Analysis on Example with Enhanced $\overline{h}_{op}$

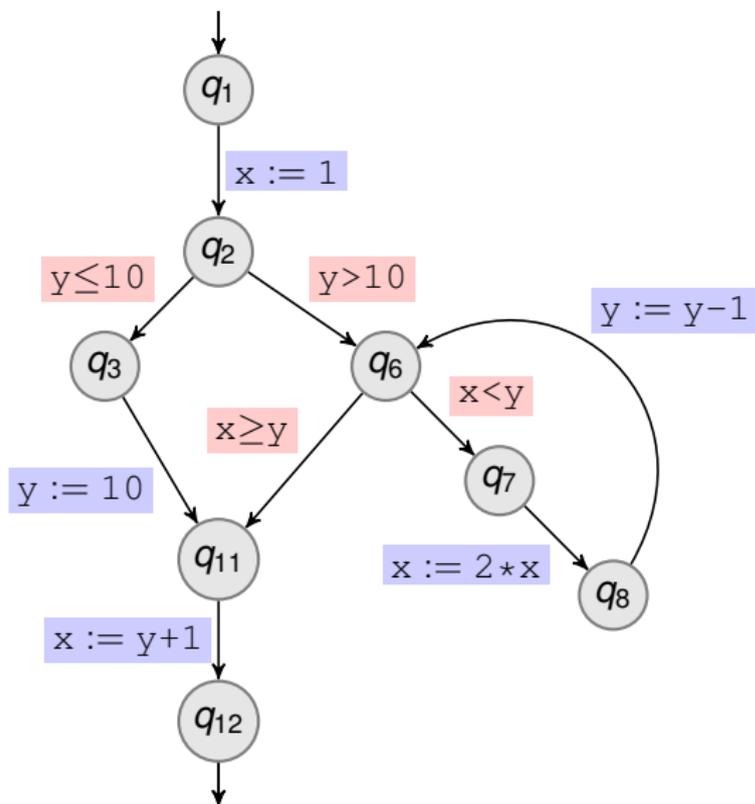


Goal

Show that  $x > 0$  at  $q_{12}$

	x	y
$q_1$	⊥	⊥
$q_2$	+	⊥
$q_3$	+	⊥
$q_6$	+	+
$q_7$	+	+
$q_8$	⊥	⊥
$q_{11}$	+	+
$q_{12}$	+	+

# Forward Sign Analysis on Example with Enhanced $\overline{h}_{op}$

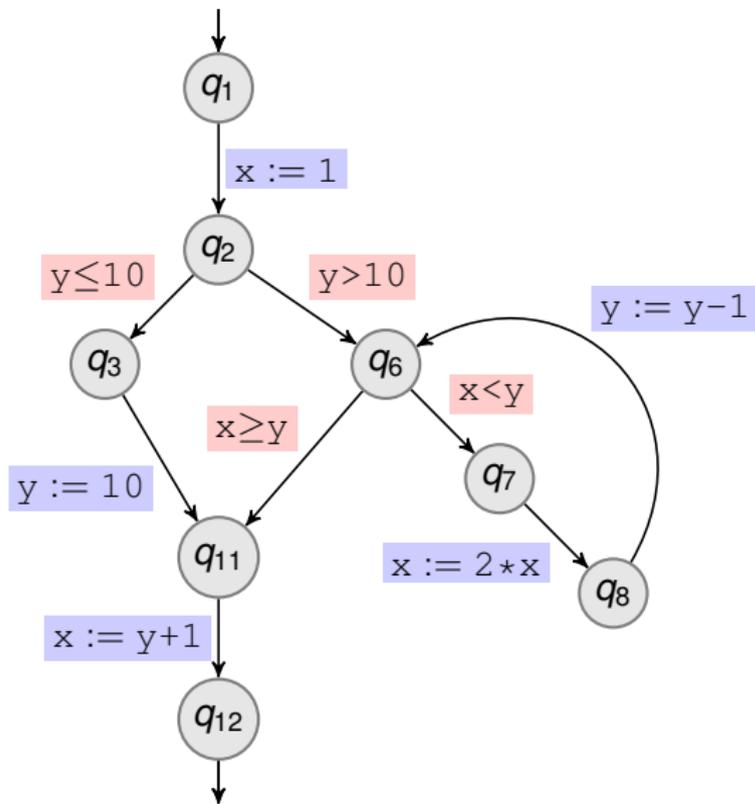


Goal

Show that  $x > 0$  at  $q_{12}$

	x	y
$q_1$	⊤	⊤
$q_2$	+	⊤
$q_3$	+	⊤
$q_6$	+	+
$q_7$	+	+
$q_8$	+	+
$q_{11}$	+	+
$q_{12}$	+	+

# Forward Sign Analysis on Example with Enhanced $\overline{h}_{op}$

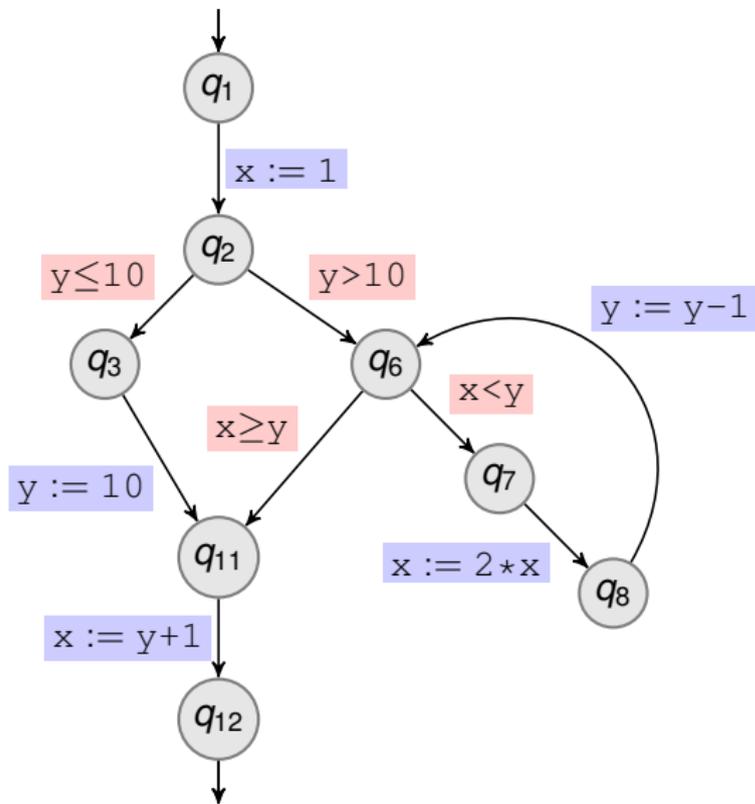


Goal

Show that  $x > 0$  at  $q_{12}$

	x	y
$q_1$	⊤	⊤
$q_2$	+	⊤
$q_3$	+	⊤
$q_6$	+	+
$q_7$	+	+
$q_8$	+	+
$q_{11}$	+	+
$q_{12}$	+	+

# Forward Sign Analysis on Example with Enhanced $\overline{h}_{op}$

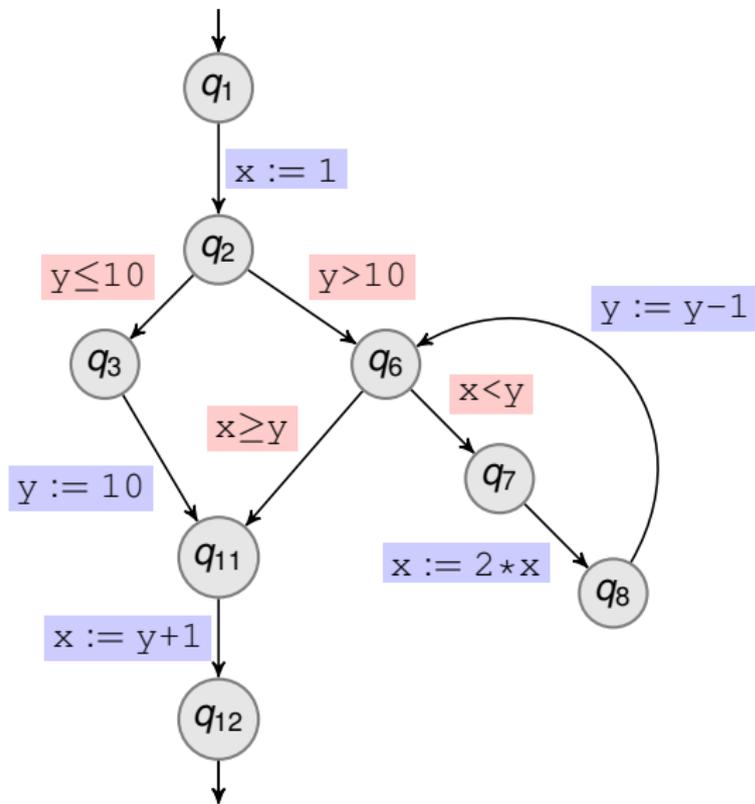


Goal

Show that  $x > 0$  at  $q_{12}$

	x	y
$q_1$	T	T
$q_2$	+	T
$q_3$	+	T
$q_6$	+	T
$q_7$	+	+
$q_8$	+	+
$q_{11}$	+	+
$q_{12}$	+	+

# Forward Sign Analysis on Example with Enhanced $\overline{h}_{op}$

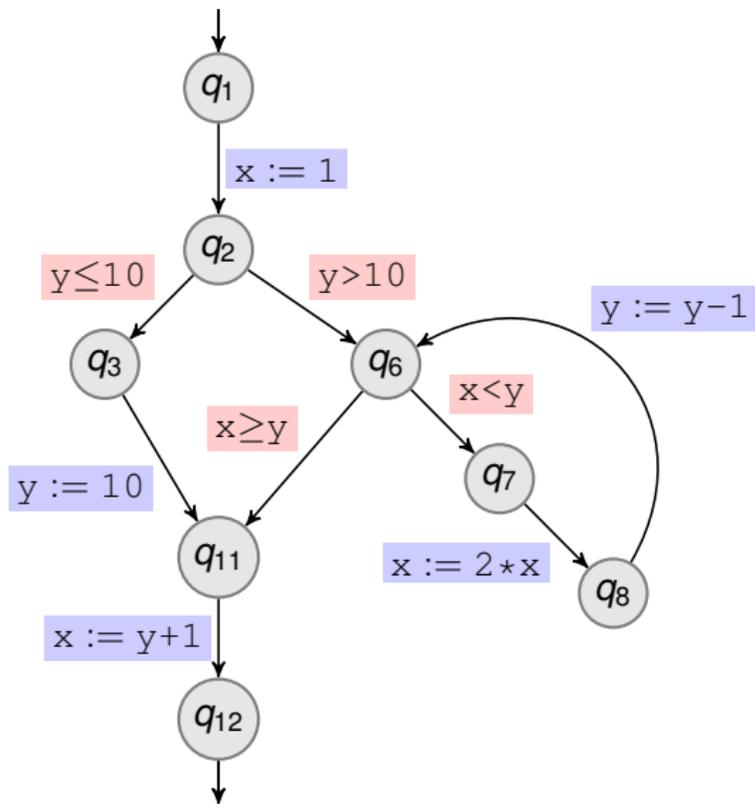


Goal

Show that  $x > 0$  at  $q_{12}$

	x	y
$q_1$	⊥	⊥
$q_2$	+	⊥
$q_3$	+	⊥
$q_6$	+	⊥
$q_7$	+	+
$q_8$	+	+
$q_{11}$	+	+
$q_{12}$	+	+

# Forward Sign Analysis on Example with Enhanced $\overline{h}_{op}$

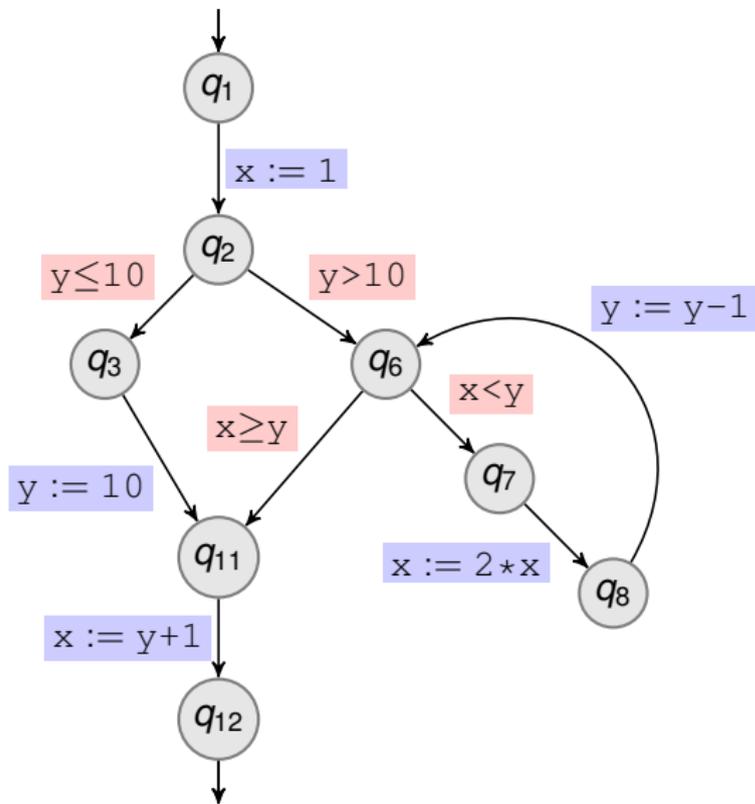


Goal

Show that  $x > 0$  at  $q_{12}$

	x	y
$q_1$	⊤	⊤
$q_2$	+	⊤
$q_3$	+	⊤
$q_6$	+	⊤
$q_7$	+	+
$q_8$	+	+
$q_{11}$	+	+
$q_{12}$	+	+

# Forward Sign Analysis on Example with Enhanced $\overline{h}_{op}$

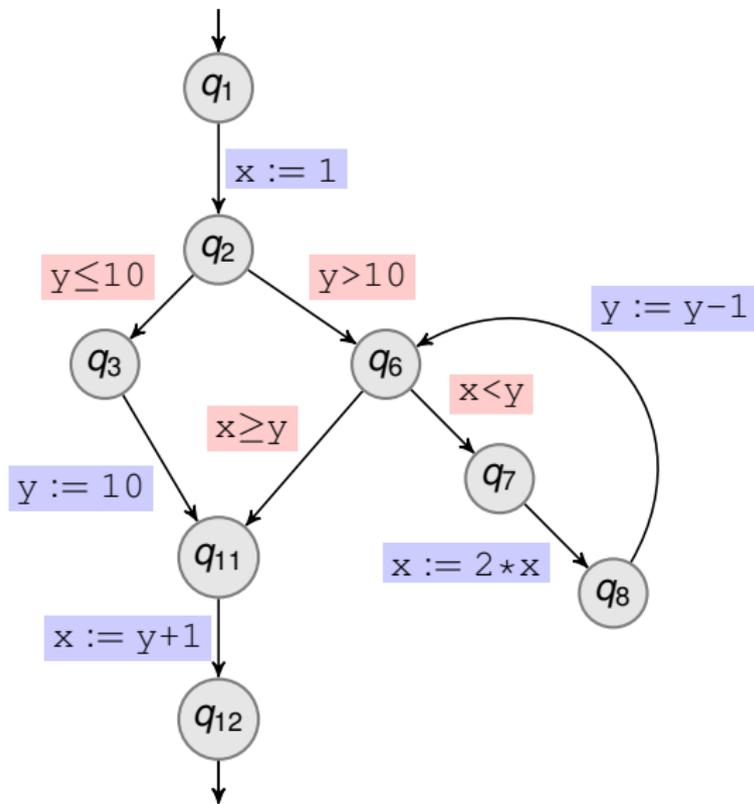


Goal

Show that  $x > 0$  at  $q_{12}$

	x	y
$q_1$	⊤	⊤
$q_2$	+	⊤
$q_3$	+	⊤
$q_6$	+	⊤
$q_7$	+	+
$q_8$	+	+
$q_{11}$	+	+
$q_{12}$	+	+

# Forward Sign Analysis on Example with Enhanced $\overline{h}_{op}$

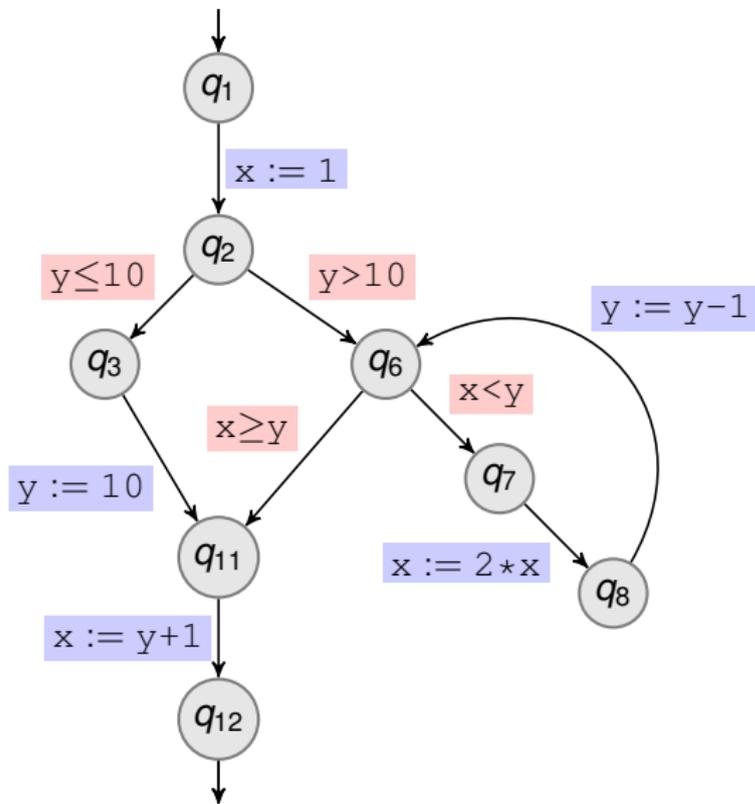


Goal

Show that  $x > 0$  at  $q_{12}$

	x	y
$q_1$	T	T
$q_2$	+	T
$q_3$	+	T
$q_6$	+	T
$q_7$	+	+
$q_8$	+	+
$q_{11}$	+	T
$q_{12}$	+	+

# Forward Sign Analysis on Example with Enhanced $\overline{h}_{op}$

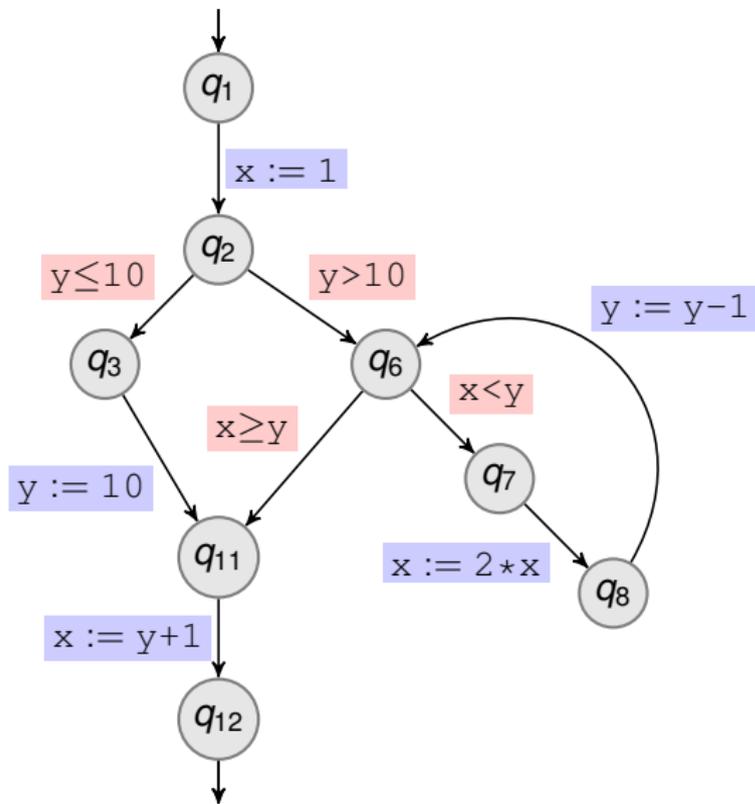


Goal

Show that  $x > 0$  at  $q_{12}$

	x	y
$q_1$	⊤	⊤
$q_2$	+	⊤
$q_3$	+	⊤
$q_6$	+	⊤
$q_7$	+	+
$q_8$	+	+
$q_{11}$	+	⊤
$q_{12}$	+	+

# Forward Sign Analysis on Example with Enhanced $\overline{h}_{op}$

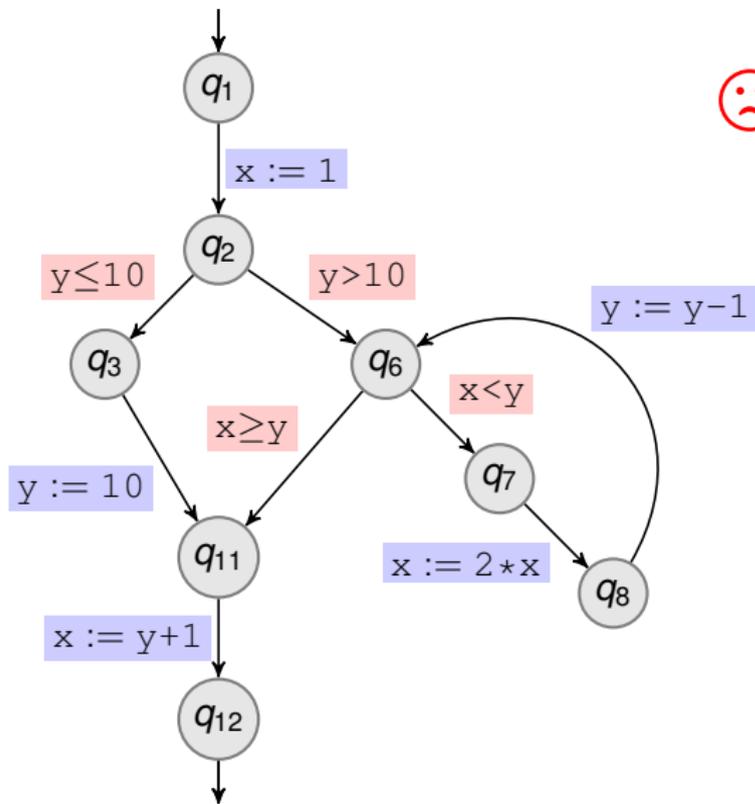


Goal

Show that  $x > 0$  at  $q_{12}$

	x	y
$q_1$	$\top$	$\top$
$q_2$	+	$\top$
$q_3$	+	$\top$
$q_6$	+	$\top$
$q_7$	+	+
$q_8$	+	+
$q_{11}$	+	$\top$
$q_{12}$	$\top$	$\top$

# Forward Sign Analysis on Example with Enhanced $\overline{h}_{op}$

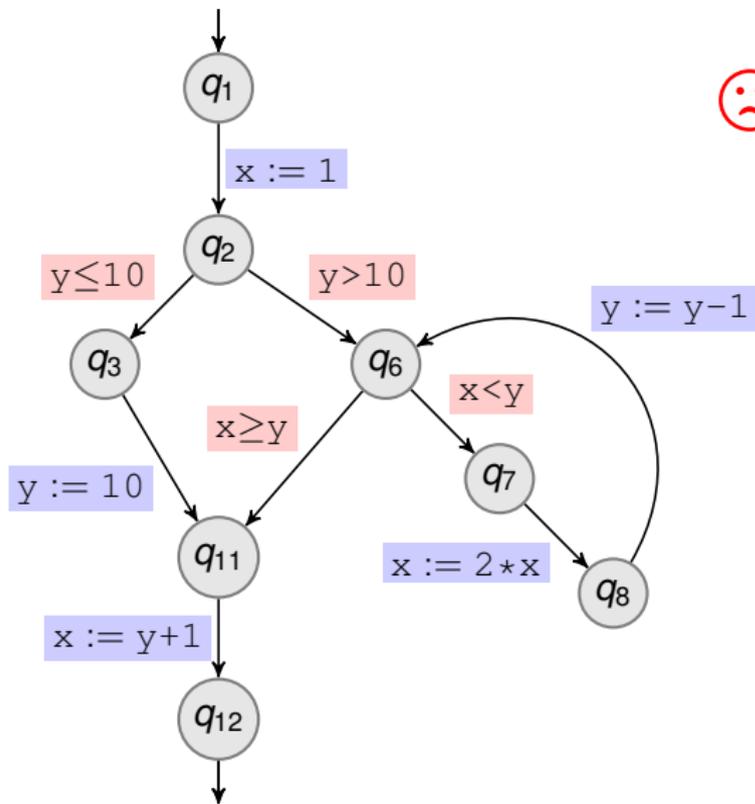


Goal

Show that  $x > 0$  at  $q_{12}$

	x	y
$q_1$	⊤	⊤
$q_2$	+	⊤
$q_3$	+	⊤
$q_6$	+	⊤
$q_7$	+	+
$q_8$	+	+
$q_{11}$	+	⊤
$q_{12}$	⊤	⊤

# Forward Sign Analysis on Example with Enhanced $\overline{h}_{op}$



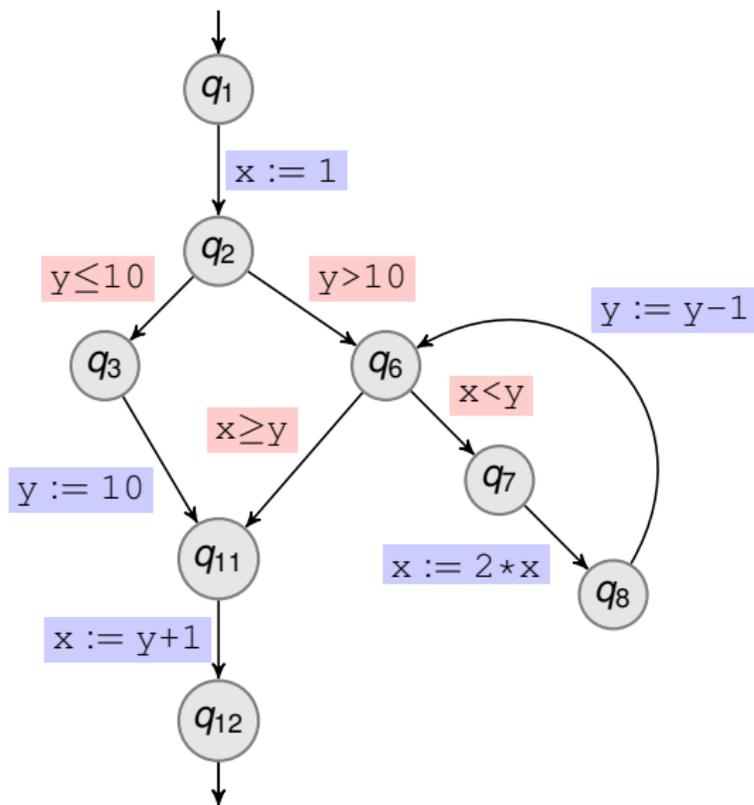
Goal

Show that  $x > 0$  at  $q_{12}$

	x	y
$q_1$	⊥	⊥
$q_2$	+	⊥
$q_3$	+	⊥
$q_6$	+	⊥
$q_7$	+	+
$q_8$	+	+
$q_{11}$	+	⊥
$q_{12}$	⊥	⊥

Getting closer...

# Forward Sign Analysis on Example with Enhanced $\overline{h}_{op}$

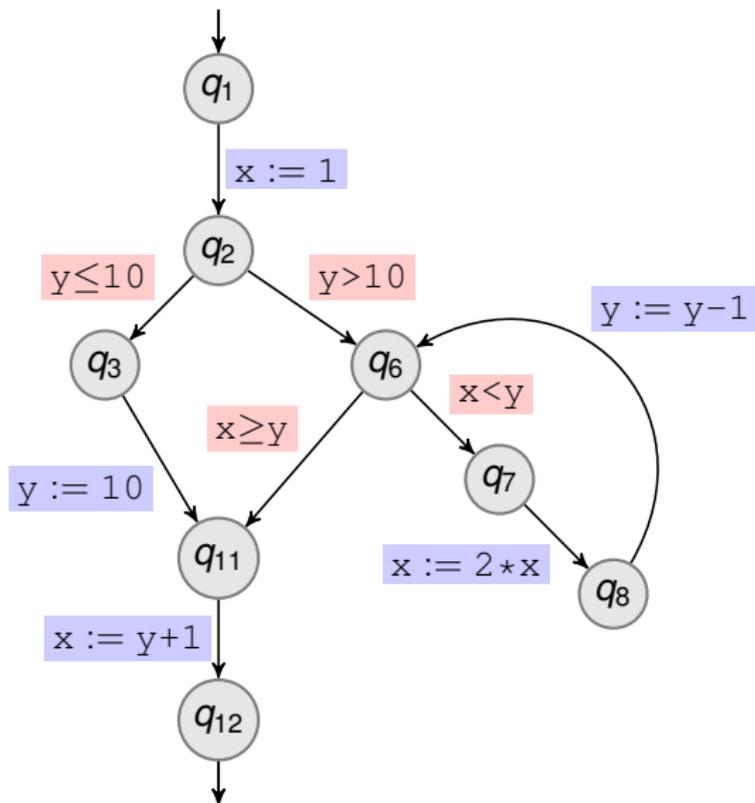


Show that  $x > 0$  at  $q_{12}$

**Assumption**

Variables range over  $\mathbb{Z}$

# Forward Sign Analysis on Example with Enhanced $\overline{h_{op}}$



Show that  $x > 0$  at  $q_{12}$

**Assumption**

Variables range over  $\mathbb{Z}$

$\gamma_{sign}(+) = \{1, 2, \dots\}$

**Tuned Semantics**

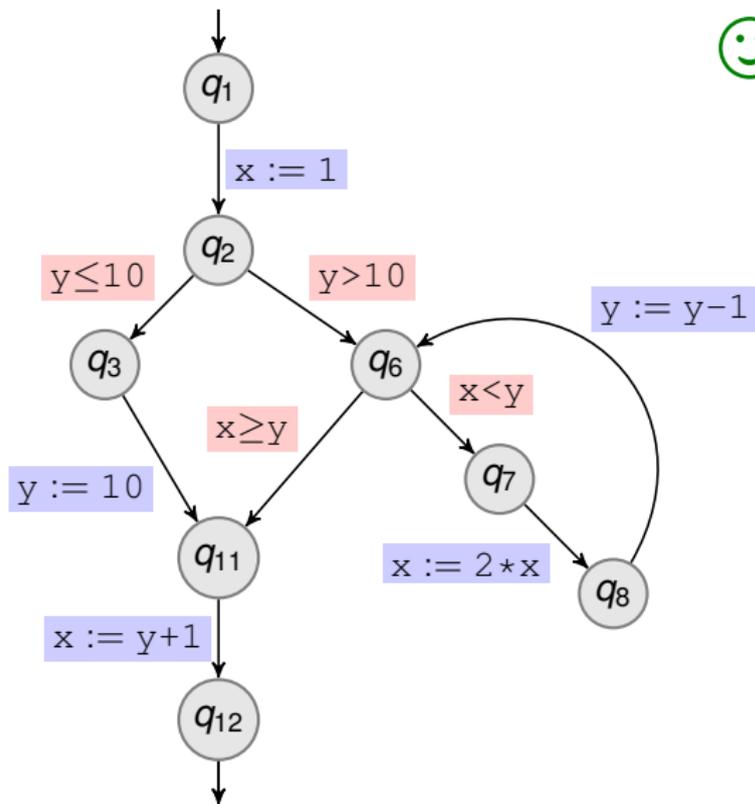
To exploit this new  $\gamma_{sign}$

$op = x := e - 1$

If  $\llbracket e \rrbracket_{\bar{v}} = +$  then

$\overline{h_{op}}(\bar{v})(x) = 0+$

# Forward Sign Analysis on Example with Enhanced $\overline{h}_{op}$



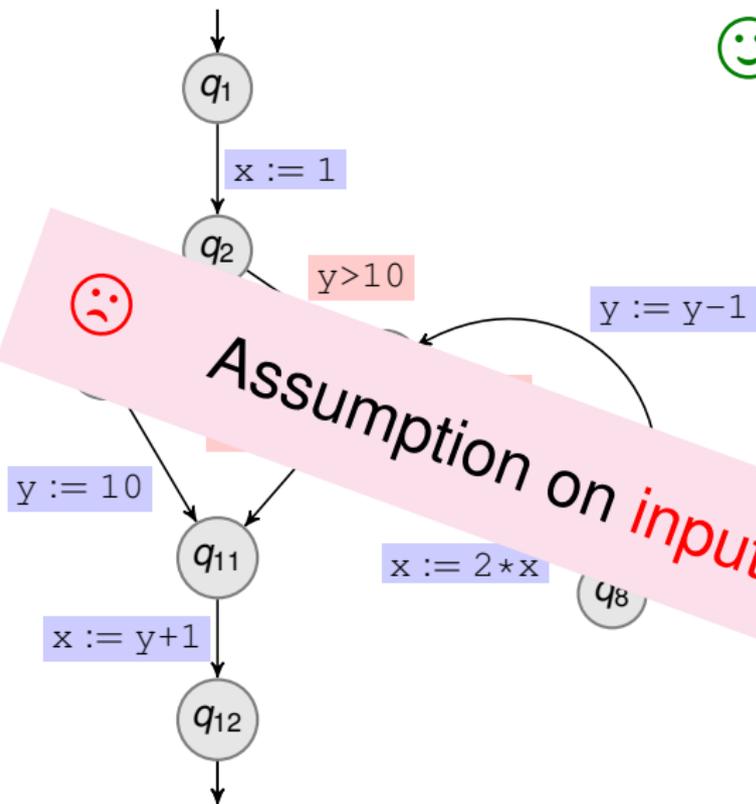
Show that  $x > 0$  at  $q_{12}$

**Assumption**

Variables range over  $\mathbb{Z}$

	x	y
$q_1$	⊤	⊤
$q_2$	+	⊤
$q_3$	+	⊤
$q_6$	+	0+
$q_7$	+	+
$q_8$	+	+
$q_{11}$	+	0+
$q_{12}$	+	+

# Forward Sign Analysis on Example with Enhanced $\overline{h}_{op}$



Show that  $x > 0$  at  $q_{12}$

**Assumption**

Variables range over  $\mathbb{Z}$

	x	y
$q_1$	$\top$	$\top$
$q_2$	+	$\top$
$q_3$	+	$\top$
	+	0+
		+
$q_{11}$		
$q_{12}$	+	

Assumption on **input variable**  $y$



## Beyond Sign Analysis...

A careful inspection of the control flow automaton shows that the desired property  $x > 0$  at  $q_{12}$  holds for any **real** initial value of  $y$ .

This property cannot be obtained even with best abstract transfer mapping  $f_{op}^\#$ .

The *Sign* abstract lattice is not sufficient!

Solution

Try with a **finer** abstract lattice!

## Beyond Sign Analysis...

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### Solution

Try with a **finer** abstract lattice!

- 8 Some More Lattice Theory: Galois Connections
- 9 Abstract Interpretation-Based Data Flow Analysis
- 10 Convergence Acceleration with Widening and Narrowing**

# Short Introduction to Widening and Narrowing

So far: **finite height** lattices. Iterative computation of MOP converges.

Finite height lattices not sufficient for (precise) numerical analysis

## Software Verification by Invariant Generation

Good precision is required for generation of useful invariants

**Infinite height** abstract lattices required to obtain good precision

But iterative computation of MOP may **diverge** in infinite height lattices.

## Solution

- 1 Use **widening** to compute a sound **approximation** of MOP.
- 2 Use **narrowing** to improve the **precision** of the approximation.

# Short Introduction to Widening and Narrowing

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- 1 Use **widening** to compute a sound **approximation** of MOP.
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# Illustration with Range Analysis

## Objective of Range Analysis

Discover for each location the range of possible run-time values that variables can have at that location.

Generalizes both sign analysis and constant propagation analysis.

We will first design a Galois Connection between:

- $(\mathcal{P}(x \rightarrow \mathbb{R}), \subseteq)$ , concrete data flow facts from standard semantics
- $(x \rightarrow \bar{L}, \bar{\sqsubseteq})$ , where  $(\bar{L}, \bar{\sqsubseteq})$  is an abstract lattice to represent range information.

Forward analysis

# Complete Lattice $\mathcal{Z}$ : Extension of $\mathbb{Z}$ with $-\infty$ and $+\infty$

Let  $\mathcal{Z} = \mathbb{Z} \cup \{-\infty, +\infty\}$  and define the partial order  $\leq$  on  $\mathcal{Z}$  with:

$$-\infty < \dots < -2 < -1 < 0 < 1 < 2 < \dots < +\infty$$

$(\mathcal{Z}, \leq)$  is a **complete lattice**

- Least element is  $-\infty$  and greatest element is  $+\infty$ .
- Least upper bound **sup**  $X$  is either  $\max(X)$  if it exists, or  $+\infty$ .
- Greatest lower bound **inf**  $X$  is either  $\min(X)$  if it exists, or  $-\infty$ .

# Abstract Lattice of Intervals

$$-\infty < \dots < -2 < -1 < 0 < 1 < 2 < \dots < +\infty$$

$$\text{Int} = \{\perp\} \cup \{(l, u) \in (\mathbb{Z} \cup \{-\infty\}) \times (\mathbb{Z} \cup \{+\infty\}) \mid l \leq u\}$$

Define the binary relation  $\sqsubseteq$  on  $\text{Int}$  as follows:

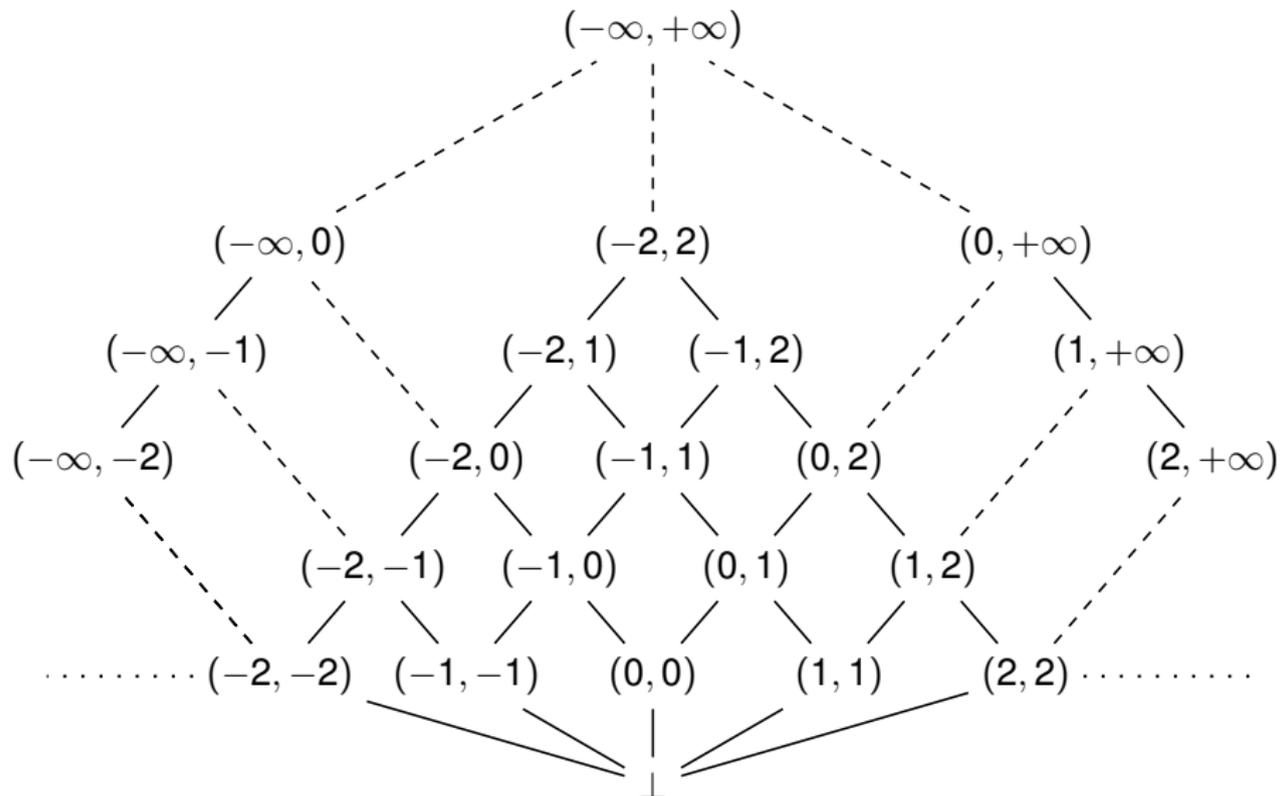
$$\perp \sqsubseteq \perp \qquad \perp \sqsubseteq (l, u) \not\sqsubseteq \perp$$

$$(l_1, u_1) \sqsubseteq (l_2, u_2) \text{ if } l_1 \leq l_2 \leq u_2 \leq u_1$$

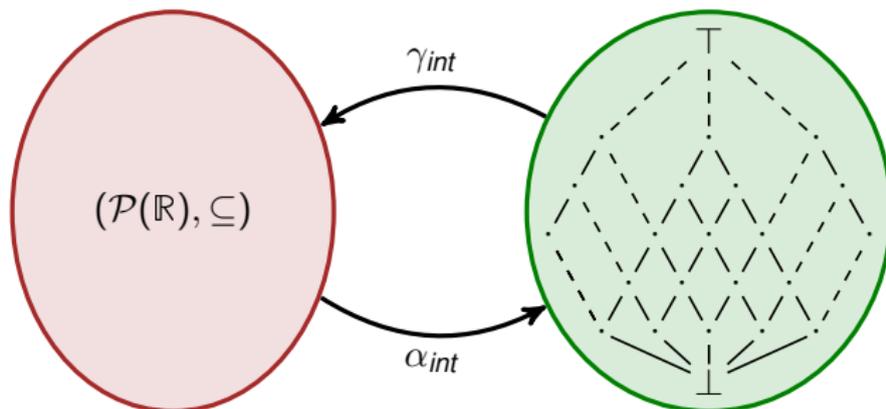
$(\text{Int}, \sqsubseteq)$  is a **complete lattice**

- Least element is  $\perp$  and greatest element is  $(-\infty, +\infty)$ .
- $\bigsqcup X$  is either  $\perp$  or  $(\inf \{l \mid (l, u) \in X\}, \sup \{u \mid (l, u) \in X\})$ .
- $\bigsqcap X$  is either  $\perp$  or  $(\sup \{l \mid (l, u) \in X\}, \inf \{u \mid (l, u) \in X\})$ .

# Abstract Lattice of Intervals



# Interpretation of Abstract Intervals: Galois Connection



$$\alpha_{int}(\phi) = \begin{cases} \overline{\perp} & \text{if } \phi = \emptyset \\ (\inf \{ \lfloor r \rfloor \mid r \in \phi \}, \sup \{ \lceil r \rceil \mid r \in \phi \}) & \text{otherwise} \end{cases}$$

$$\gamma_{int}(\overline{\perp}) = \emptyset$$

$$\gamma_{int}(\overline{(l, u)}) = \{ r \in \mathbb{R} \mid l \leq r \leq u \}$$

# Range Analysis with Intervals: Galois Connection

## Projection

$$(\mathcal{P}(X \rightarrow \mathbb{R}), \sqsubseteq) \begin{matrix} \xleftarrow{\gamma_\pi} \\ \xrightarrow{\alpha_\pi} \end{matrix} (X \rightarrow \mathcal{P}(\mathbb{R}), \sqsubseteq)$$

## Intervals

$$(\mathcal{P}(\mathbb{R}), \sqsubseteq) \begin{matrix} \xleftarrow{\gamma_{int}} \\ \xrightarrow{\alpha_{int}} \end{matrix} (Int, \bar{\sqsubseteq})$$

## Extension of Intervals to Functions

$$(X \rightarrow \mathcal{P}(\mathbb{R}), \sqsubseteq) \begin{matrix} \xleftarrow{\gamma_{int}} \\ \xrightarrow{\alpha_{int}} \end{matrix} (X \rightarrow Int, \bar{\sqsubseteq})$$

$$(\mathcal{P}(X \rightarrow \mathbb{R}), \sqsubseteq) \begin{matrix} \xleftarrow{\gamma_\pi \circ \gamma_{int}} \\ \xrightarrow{\alpha_{int} \circ \alpha_\pi} \end{matrix} (X \rightarrow Int, \bar{\sqsubseteq})$$

# Range Analysis with Intervals: Galois Connection

## Projection

$$(\mathcal{P}(X \rightarrow \mathbb{R}), \subseteq) \begin{array}{c} \xleftarrow{\gamma_\pi} \\ \xrightarrow{\alpha_\pi} \end{array} (X \rightarrow \mathcal{P}(\mathbb{R}), \subseteq)$$

## Intervals

$$(\mathcal{P}(\mathbb{R}), \subseteq) \begin{array}{c} \xleftarrow{\gamma_{int}} \\ \xrightarrow{\alpha_{int}} \end{array} (X \rightarrow \mathcal{P}(\mathbb{R}), \subseteq) \begin{array}{c} \xleftarrow{\gamma_{int}} \\ \xrightarrow{\alpha_{int}} \end{array} (X \rightarrow Int, \bar{\subseteq})$$



Same as sign analysis



$$(\mathcal{P}(X \rightarrow \mathbb{R}), \subseteq) \begin{array}{c} \xleftarrow{\gamma_\pi \circ \gamma_{int}} \\ \xrightarrow{\alpha_{int} \circ \alpha_\pi} \end{array} (X \rightarrow Int, \bar{\subseteq})$$

# Range Analysis: Induced Abstract Data Flow Instance

$$(\mathcal{P}(X \rightarrow \mathbb{R}), \subseteq) \xrightleftharpoons[\alpha_\pi]{\gamma_\pi} (X \rightarrow \mathcal{P}(\mathbb{R}), \subseteq) \xrightleftharpoons[\alpha_{int}]{\gamma_{int}} (X \rightarrow \mathit{Int}, \bar{\subseteq})$$

$$\bar{\mathcal{F}} = (X \rightarrow \mathit{Int}) \xrightarrow{\text{mon}} (X \rightarrow \mathit{Int})$$

$$\bar{f} = \lambda \text{op} . f_{\text{op}}^\#$$

$$\bar{i} = \alpha_{int} \circ \alpha_\pi(\top) = \lambda x . \bar{\top}$$

$$f_{x:=e}^\#(\bar{v}) = \lambda y . \begin{cases} \bar{v}(y) & \text{if } y \neq x \\ \alpha_{int} \circ \llbracket e \rrbracket \circ \gamma(\bar{v}) & \text{if } y = x \end{cases}$$

$$f_g^\# = \alpha_{int} \circ \alpha_\pi \circ \llbracket g \rrbracket \circ \gamma_\pi \circ \gamma_{int}$$

# Range Analysis: Induced Abstract Data Flow Instance

$$(\mathcal{P}(X \rightarrow \mathbb{R}), \subseteq) \xleftrightarrow[\alpha_\pi]{\gamma_\pi} (X \rightarrow \mathcal{P}(\mathbb{R}), \subseteq) \xleftrightarrow[\alpha_{int}]{\gamma_{int}} (X \rightarrow Int, \bar{\subseteq})$$

$$\begin{aligned} \bar{\mathcal{F}} &= (X \rightarrow Int) \xrightarrow{\text{mon}} (X \rightarrow Int) \\ \bar{f} &= \lambda \text{op}. f_{\text{op}}^\# \\ \bar{i} &= \alpha_{int} \circ \gamma_{int} \end{aligned}$$



Same as sign analysis



$$f_{\text{op}}^\# = \begin{cases} \gamma_{int} \circ \alpha_{int} \circ \gamma_{int} & \text{if } y \neq x \\ \alpha_{int} \circ \llbracket e \rrbracket \circ \gamma_{int} & \text{if } y = x \end{cases}$$

$$f_g^\# = \alpha_{int} \circ \alpha_\pi \circ \llbracket g \rrbracket \circ \gamma_\pi \circ \gamma_{int}$$

# Range Analysis: Approximate Transfer Mapping

$$\overline{h_{x:=e}}(\bar{v}) = \lambda y. \begin{cases} \bar{v}(y) & \text{if } y \neq x \\ \llbracket e \rrbracket_{\bar{v}} & \text{if } y = x \end{cases} \quad \overline{h_g}(\bar{v}) = \begin{cases} \perp & \text{if } \bar{v} \not\models g \\ \bar{v} & \text{if } \bar{v} \models g \end{cases}$$

## Similar to Sign Analysis

- Definition of abstract arithmetic operators and comparators
- Definition of  $\llbracket e \rrbracket_{\bar{v}}$  and of  $\bar{v} \models g$
- Precision enhancement by gaining information from guards (“refinement” of  $\overline{h_g}(\bar{v})$ )

Care about effective computability of  $f_{op}^\#$ ? Not in this section...

We will use  $f^\#$  for range analysis

# Range Analysis: Approximate Transfer Mapping

$$\overline{h_{x:=e}}(\bar{v}) = \lambda y. \begin{cases} \bar{v}(y) & \text{if } y \neq x \\ \llbracket e \rrbracket_{\bar{v}} & \text{if } y = x \end{cases} \quad \overline{h_g}(\bar{v}) = \begin{cases} \perp & \text{if } \bar{v} \not\models g \\ \bar{v} & \text{if } \bar{v} \models g \end{cases}$$

## Similar to Sign Analysis

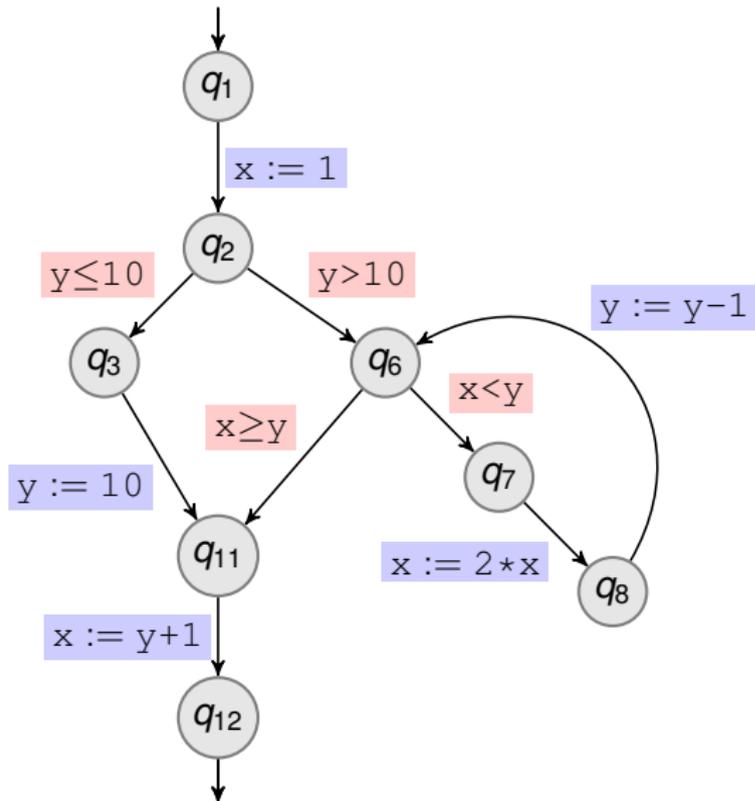
- Definition of abstract arithmetic operators and comparators
- Definition of  $\llbracket e \rrbracket_{\bar{v}}$  and of  $\bar{v} \models g$
- Precision enhancement by gaining information from guards (“refinement” of  $\overline{h_g}(\bar{v})$ )

Care about effective computability of  $f^\#$ ? Not in this section...

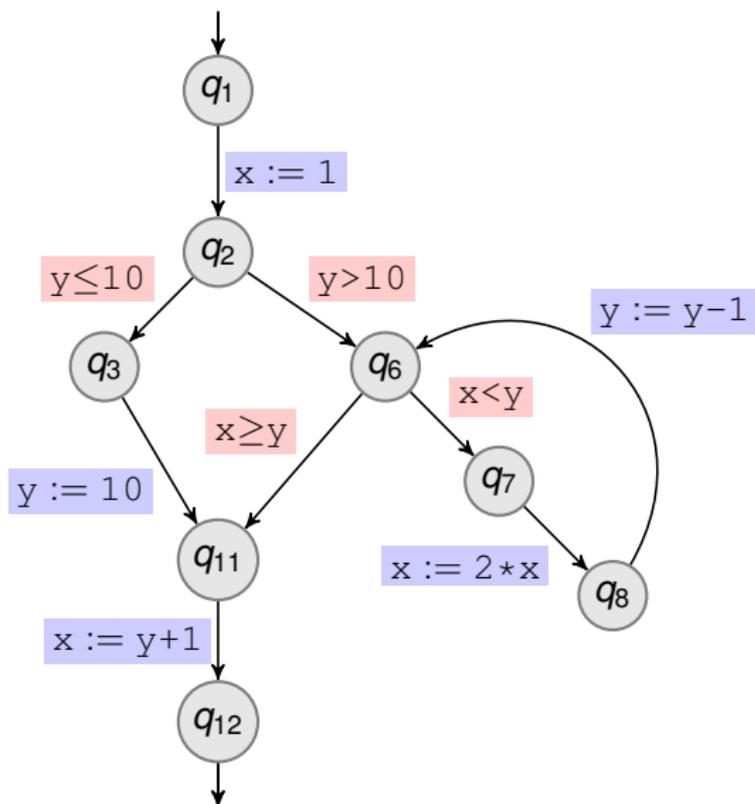
We will use  $f^\#$  for range analysis

# Range Analysis on Running Example

Recall that  $T = (-\infty, +\infty)$



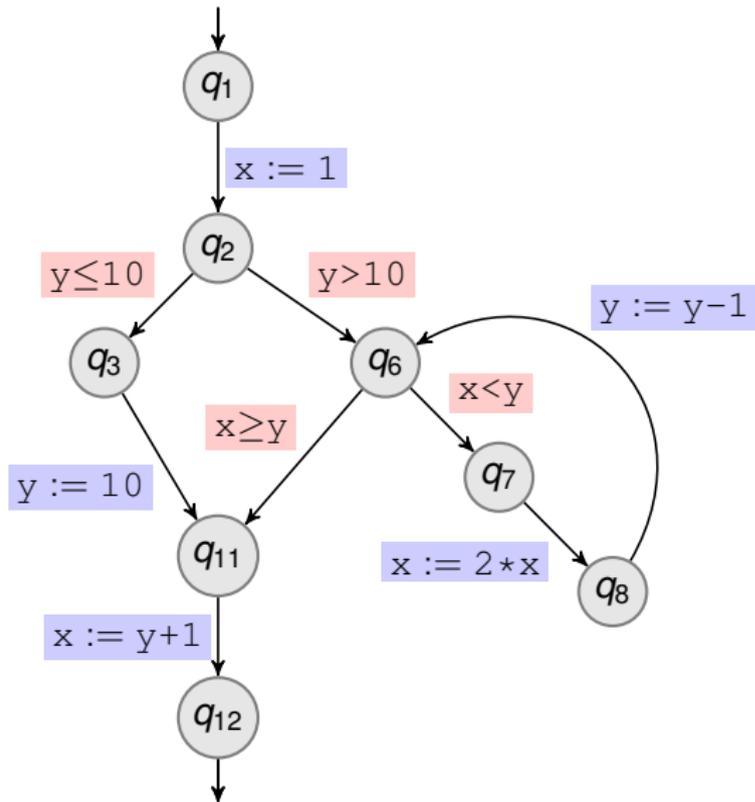
# Range Analysis on Running Example



Recall that  $T = (-\infty, +\infty)$

	x		y	
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	$\perp$	$\perp$	$\perp$	$\perp$
$q_3$	$\perp$	$\perp$	$\perp$	$\perp$
$q_6$	$\perp$	$\perp$	$\perp$	$\perp$
$q_7$	$\perp$	$\perp$	$\perp$	$\perp$
$q_8$	$\perp$	$\perp$	$\perp$	$\perp$
$q_{11}$	$\perp$	$\perp$	$\perp$	$\perp$
$q_{12}$	$\perp$	$\perp$	$\perp$	$\perp$

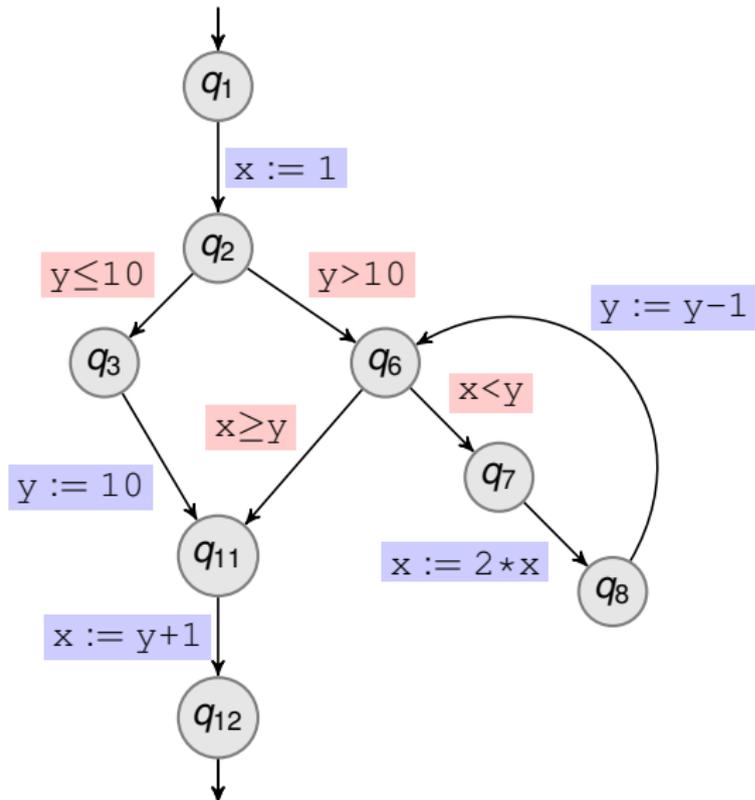
# Range Analysis on Running Example



Recall that  $T = (-\infty, +\infty)$

	x		y	
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	<del><math>+\infty</math></del>	<del><math>+\infty</math></del>
$q_3$	$\perp$	$\perp$	$\perp$	$\perp$
$q_6$	$\perp$	$\perp$	$\perp$	$\perp$
$q_7$	$\perp$	$\perp$	$\perp$	$\perp$
$q_8$	$\perp$	$\perp$	$\perp$	$\perp$
$q_{11}$	$\perp$	$\perp$	$\perp$	$\perp$
$q_{12}$	$\perp$	$\perp$	$\perp$	$\perp$

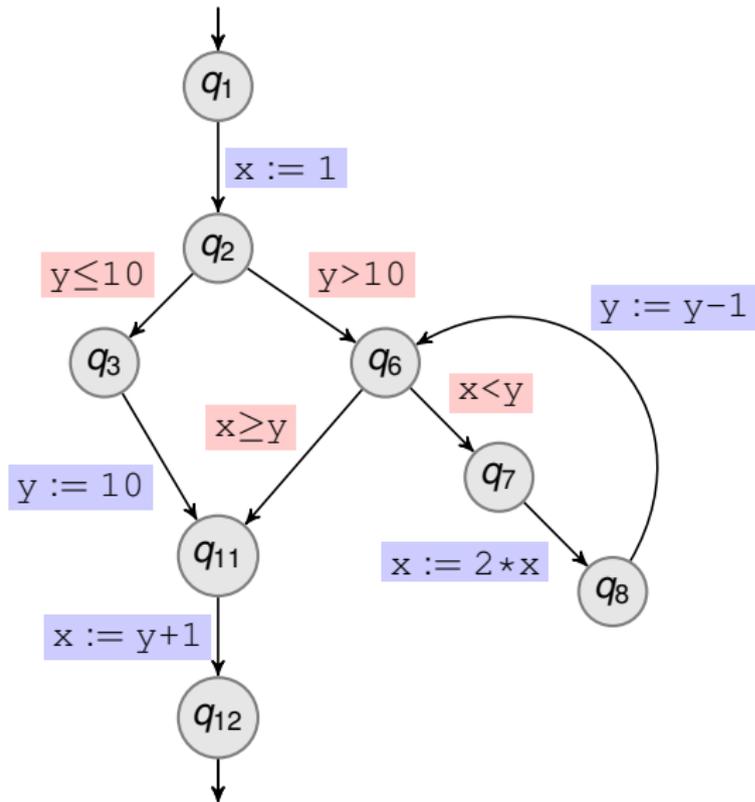
# Range Analysis on Running Example



Recall that  $T = (-\infty, +\infty)$

	x		y	
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	$\perp$	$\perp$	$\perp$	$\perp$
$q_6$	$\perp$	$\perp$	$\perp$	$\perp$
$q_7$	$\perp$	$\perp$	$\perp$	$\perp$
$q_8$	$\perp$	$\perp$	$\perp$	$\perp$
$q_{11}$	$\perp$	$\perp$	$\perp$	$\perp$
$q_{12}$	$\perp$	$\perp$	$\perp$	$\perp$

# Range Analysis on Running Example

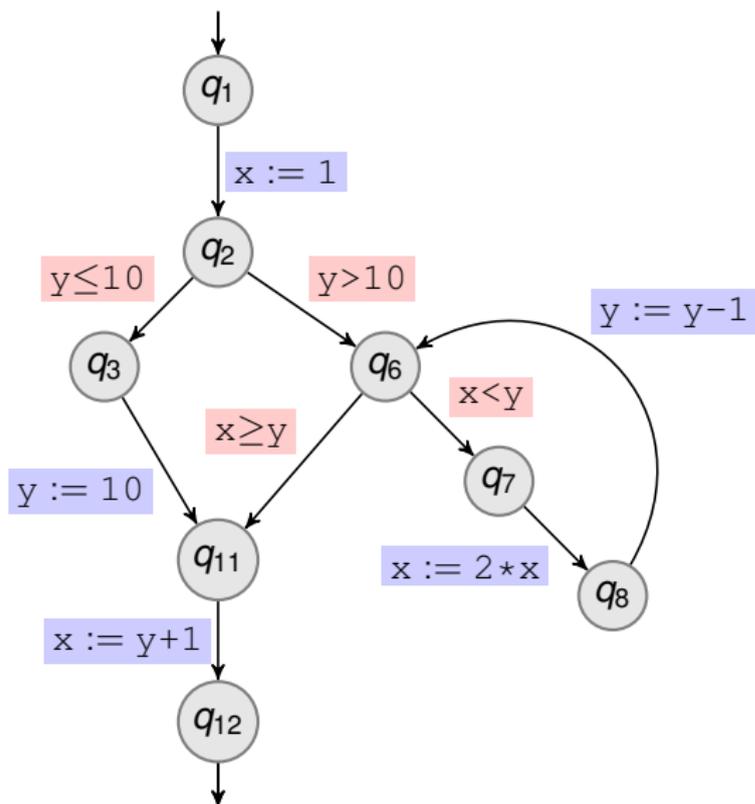


Recall that  $T = (-\infty, +\infty)$

	x		y	
	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	$\perp$	$\perp$	$\perp$	$\perp$
$q_7$	$\perp$	$\perp$	$\perp$	$\perp$
$q_8$	$\perp$	$\perp$	$\perp$	$\perp$
$q_{11}$	$\perp$	$\perp$	$\perp$	$\perp$
$q_{12}$	$\perp$	$\perp$	$\perp$	$\perp$

Gained from guard  $y \leq 10$

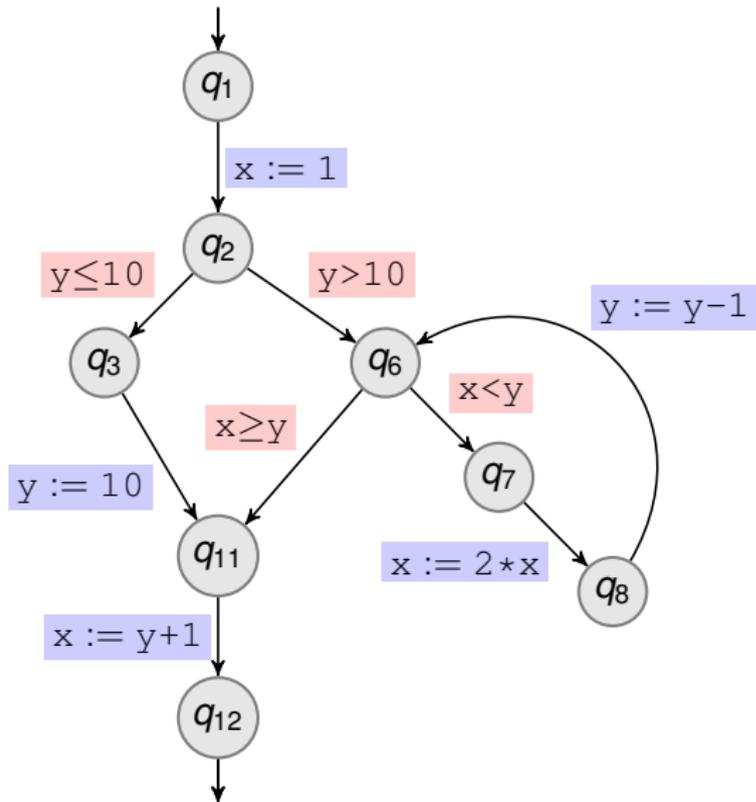
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Recall that  $T = (-\infty, +\infty)$

	x		y	
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	$\perp$	$\perp$	$\perp$	$\perp$
$q_7$	$\perp$	$\perp$	$\perp$	$\perp$
$q_8$	$\perp$	$\perp$	$\perp$	$\perp$
$q_{11}$	$\perp$	$\perp$	$\perp$	$\perp$
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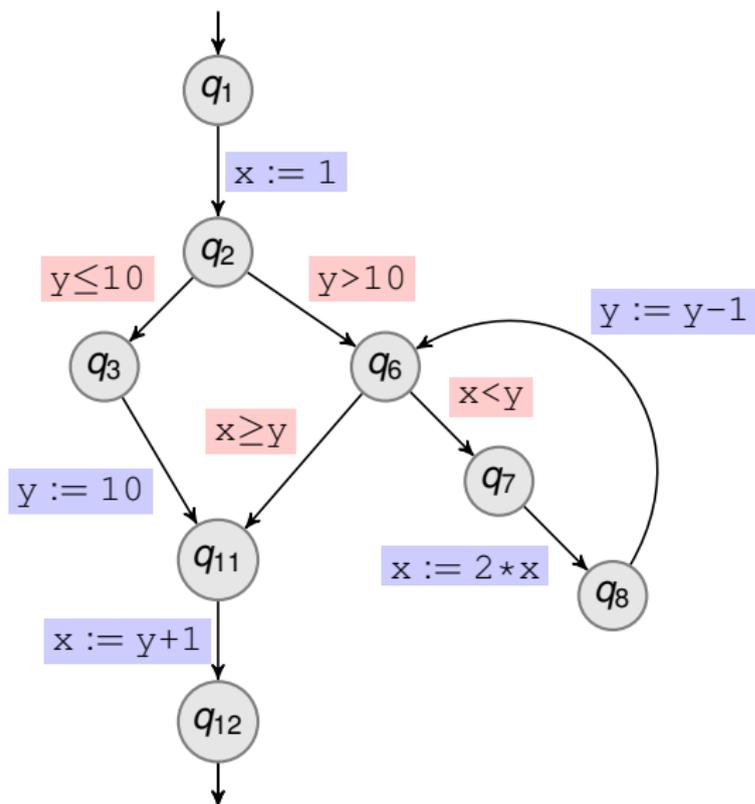
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Recall that  $T = (-\infty, +\infty)$

	x		y	
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	$\perp$	$\perp$	$\perp$	$\perp$
$q_7$	$\perp$	$\perp$	$\perp$	$\perp$
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$q_{11}$	1	1	10	10
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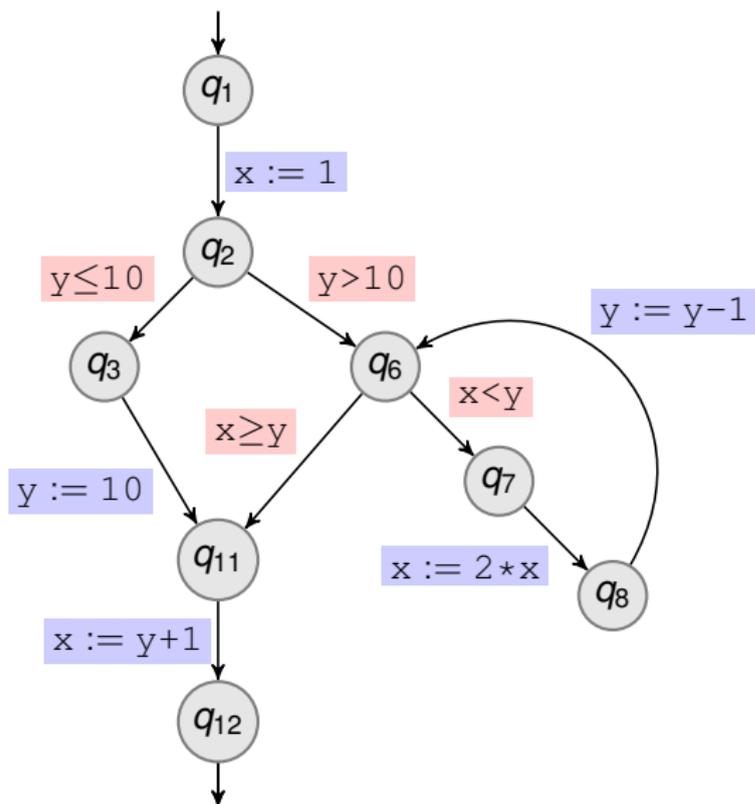
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	x		y	
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	$\perp$	$\perp$	$\perp$	$\perp$
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$q_{11}$	1	1	10	10
$q_{12}$	$\perp$	$\perp$	$\perp$	$\perp$

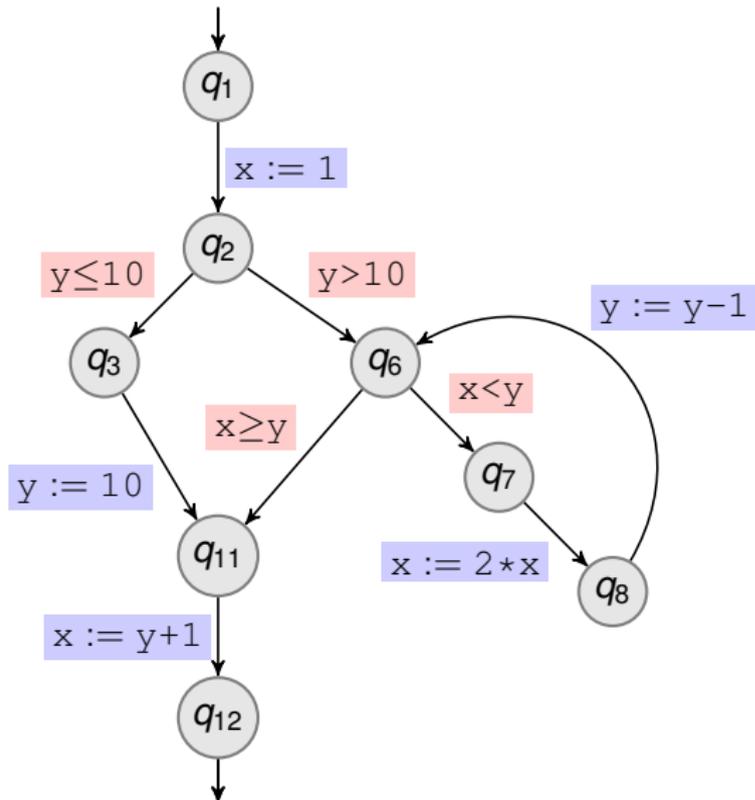
# Range Analysis on Running Example



Recall that  $T = (-\infty, +\infty)$

	x		y	
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	$\perp$	$\perp$	$\perp$	$\perp$
$q_7$	$\perp$	$\perp$	$\perp$	$\perp$
$q_8$	$\perp$	$\perp$	$\perp$	$\perp$
$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10

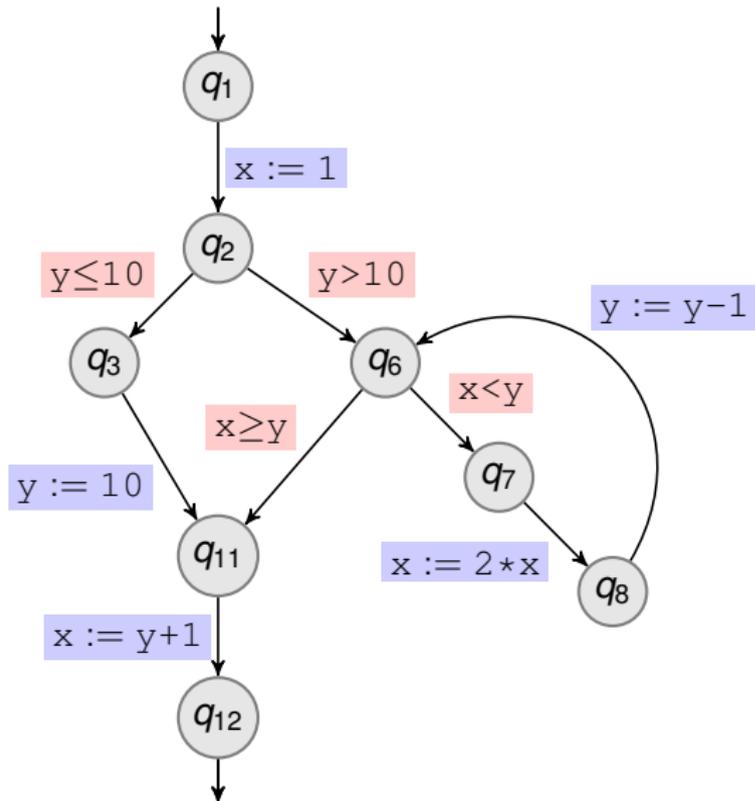
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Recall that  $T = (-\infty, +\infty)$

	x		y	
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	$\perp$	$\perp$	$\perp$	$\perp$
$q_7$	$\perp$	$\perp$	$\perp$	$\perp$
$q_8$	$\perp$	$\perp$	$\perp$	$\perp$
$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10

# Range Analysis on Running Example

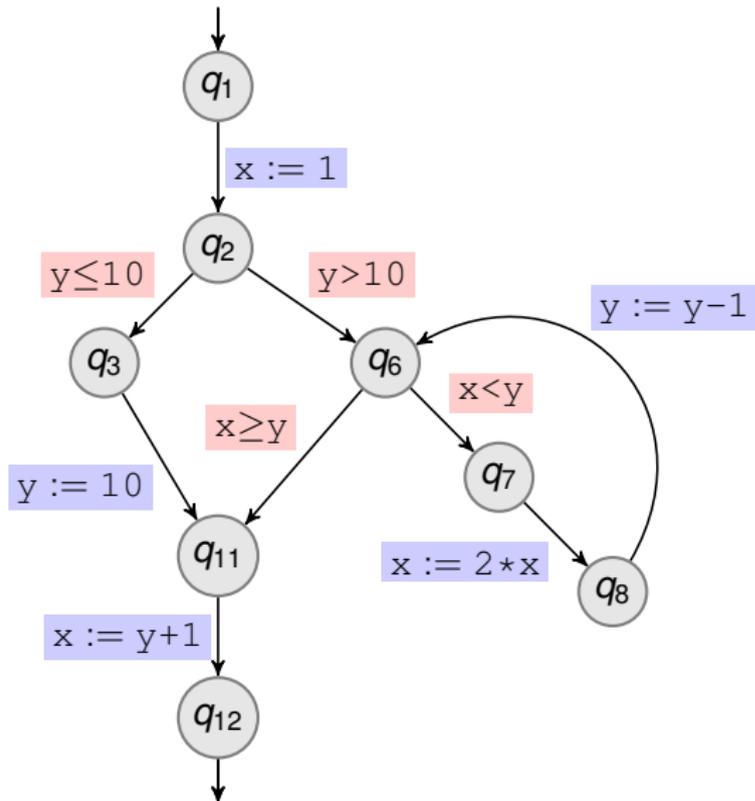


Recall that  $T = (-\infty, +\infty)$

	x		y	
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	1	10	$+\infty$
$q_7$	$\perp$	$\perp$	$\perp$	$\perp$
$q_8$	$\perp$	$\perp$	$\perp$	$\perp$
$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10

Gained from guard  $y > 10$

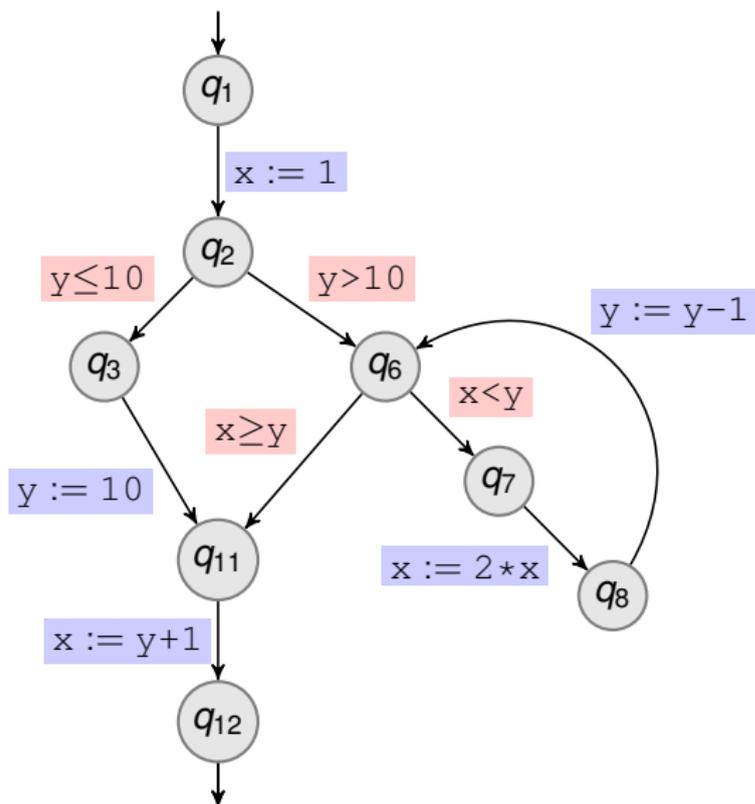
# Range Analysis on Running Example



Recall that  $T = (-\infty, +\infty)$

	x		y	
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	1	10	$+\infty$
$q_7$	$\perp$	$\perp$	$\perp$	$\perp$
$q_8$	$\perp$	$\perp$	$\perp$	$\perp$
$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10

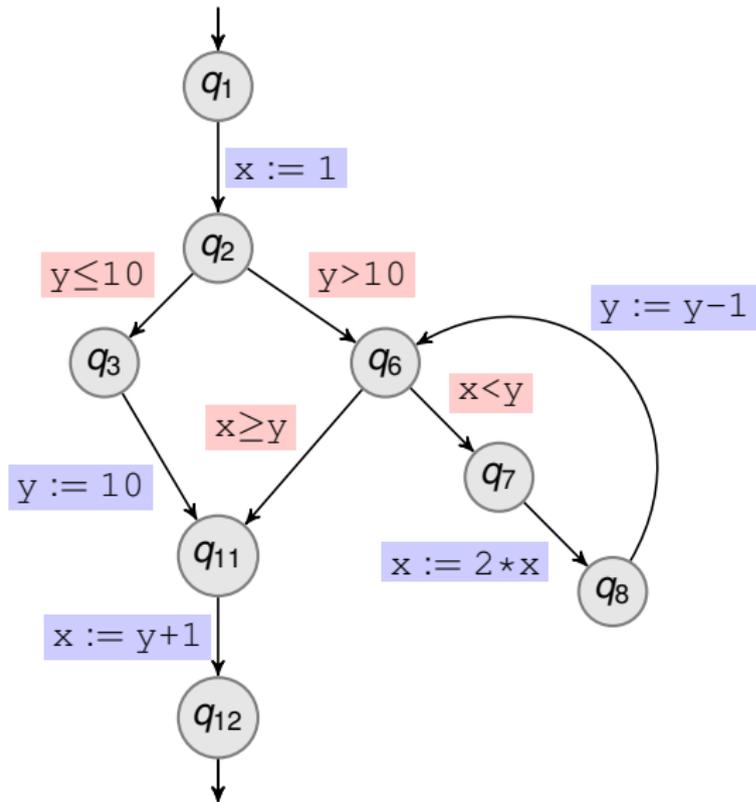
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	x		y	
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	1	10	$+\infty$
$q_7$	1	1	10	$+\infty$
$q_8$	$\perp$	$\perp$	$\perp$	$\perp$
$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10

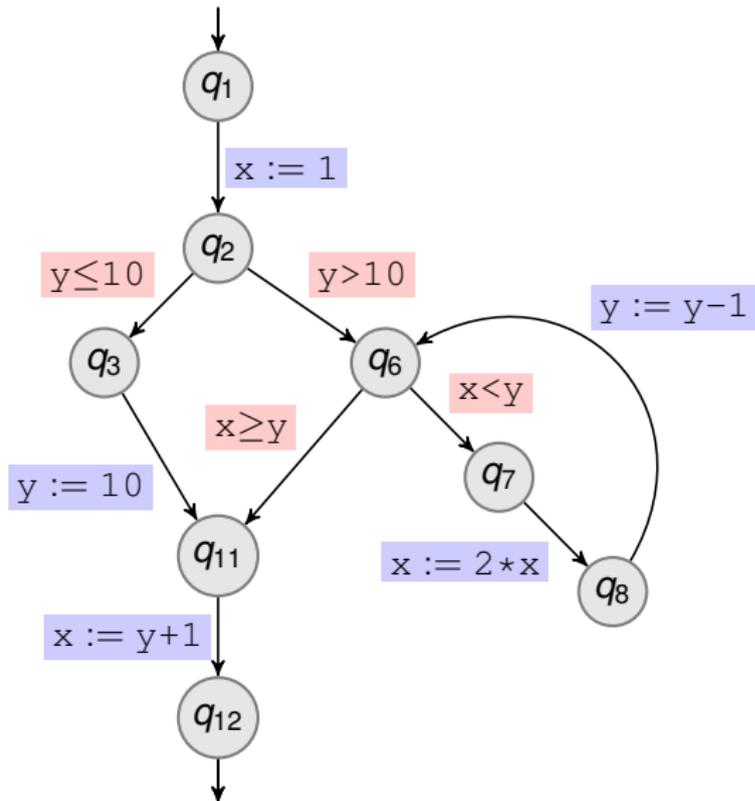
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	x		y	
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	1	10	$+\infty$
$q_7$	1	1	10	$+\infty$
$q_8$	$\perp$	$\perp$	$\perp$	$\perp$
$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10

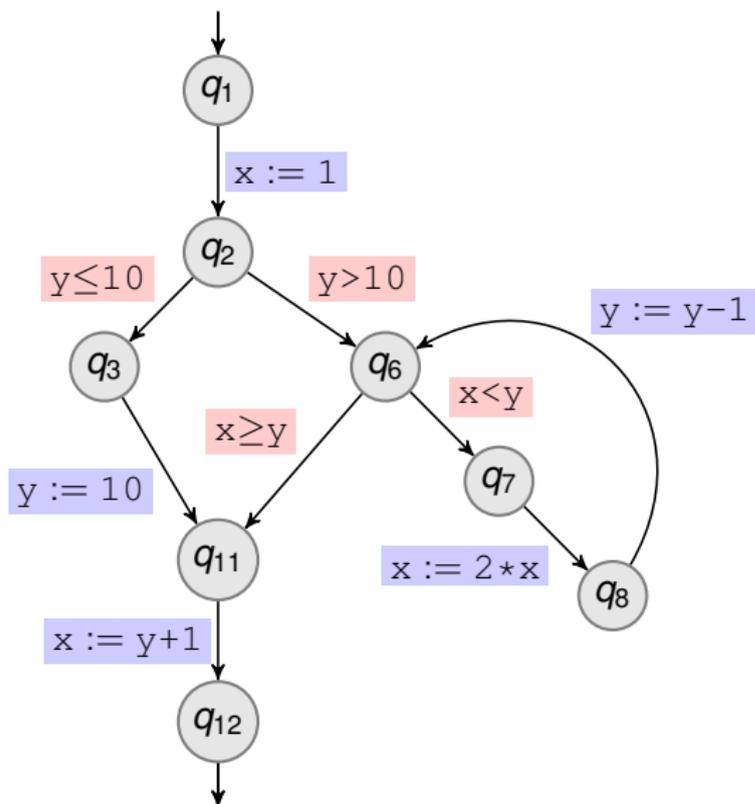
# Range Analysis on Running Example



Recall that  $T = (-\infty, +\infty)$

	x		y	
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	1	10	$+\infty$
$q_7$	1	1	10	$+\infty$
$q_8$	2	2	10	$+\infty$
$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10

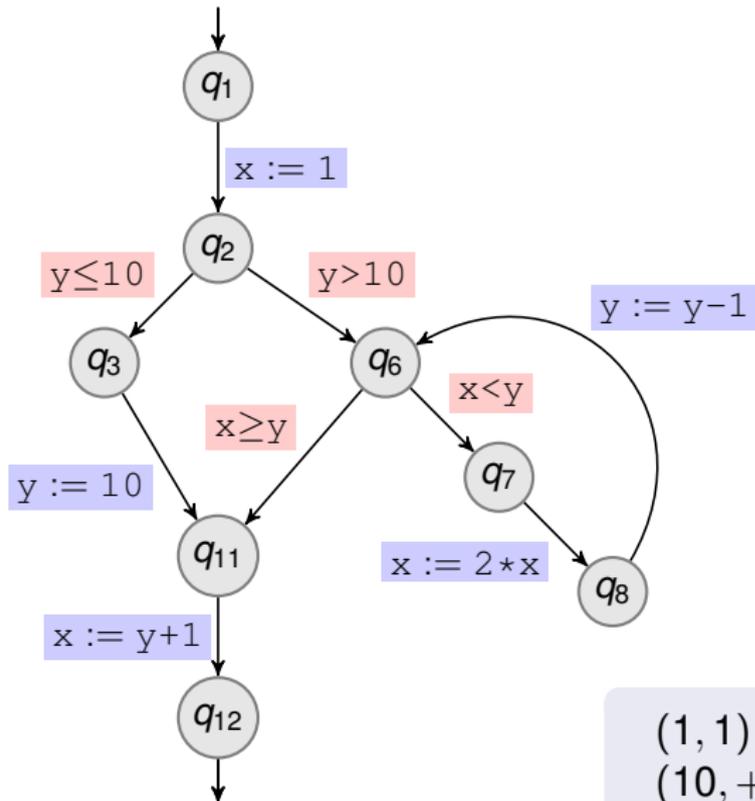
# Range Analysis on Running Example



Recall that  $T = (-\infty, +\infty)$

	x		y	
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	1	10	$+\infty$
$q_7$	1	1	10	$+\infty$
$q_8$	2	2	10	$+\infty$
$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10

# Range Analysis on Running Example



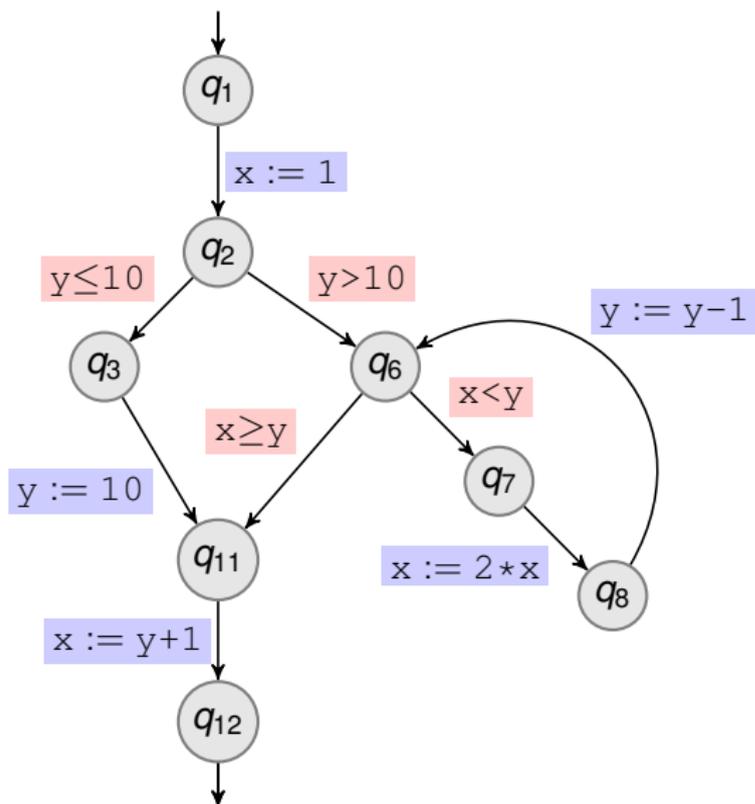
Recall that  $T = (-\infty, +\infty)$

	x		y	
	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	2	9	$+\infty$
$q_7$	1	1	10	$+\infty$
$q_8$	2	2	10	$+\infty$
$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10

$$(1, 1) \sqcup (2, 2) = (1, 2)$$

$$(10, +\infty) \sqcup (9, +\infty) = (9, +\infty)$$

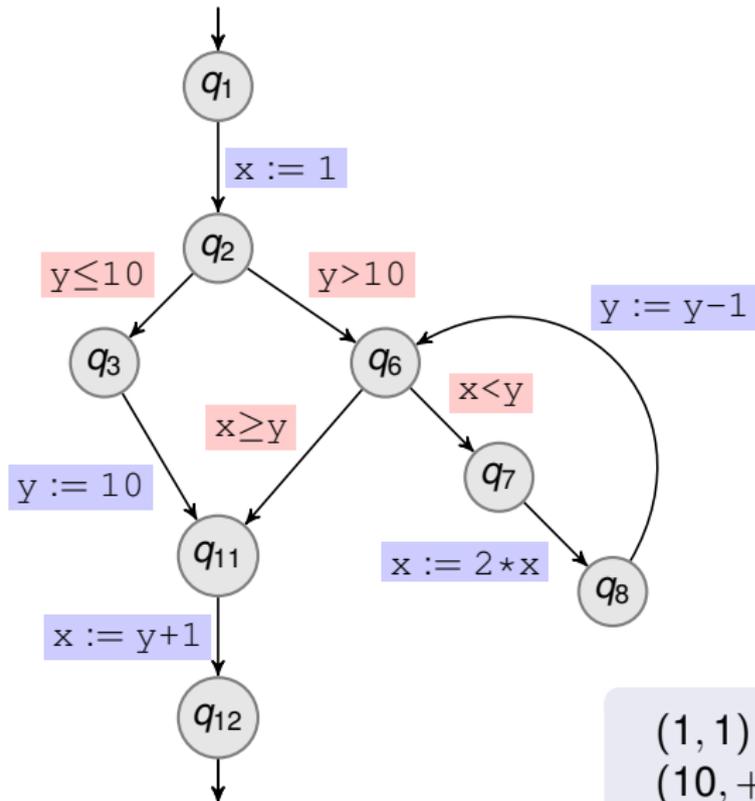
# Range Analysis on Running Example



Recall that  $T = (-\infty, +\infty)$

	x		y	
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	2	9	$+\infty$
$q_7$	1	1	10	$+\infty$
$q_8$	2	2	10	$+\infty$
$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10

# Range Analysis on Running Example

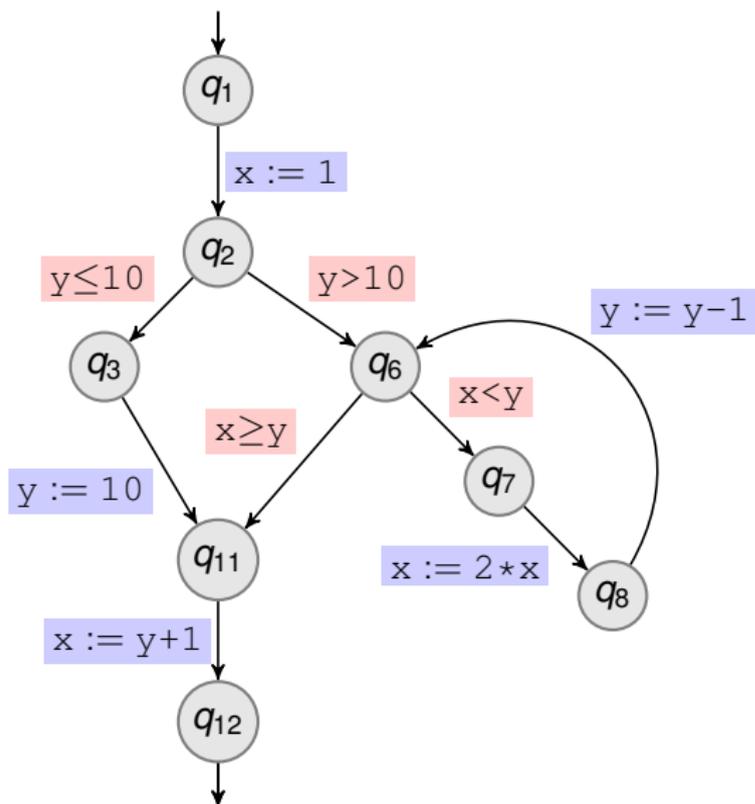


Recall that  $T = (-\infty, +\infty)$

	x		y	
	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	2	9	$+\infty$
$q_7$	1	2	9	$+\infty$
$q_8$	2	2	10	$+\infty$
$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10

$$\begin{aligned}
 (1, 1) &\sqcup (1, 2) = (1, 2) \\
 (10, +\infty) &\sqcup (9, +\infty) = (9, +\infty)
 \end{aligned}$$

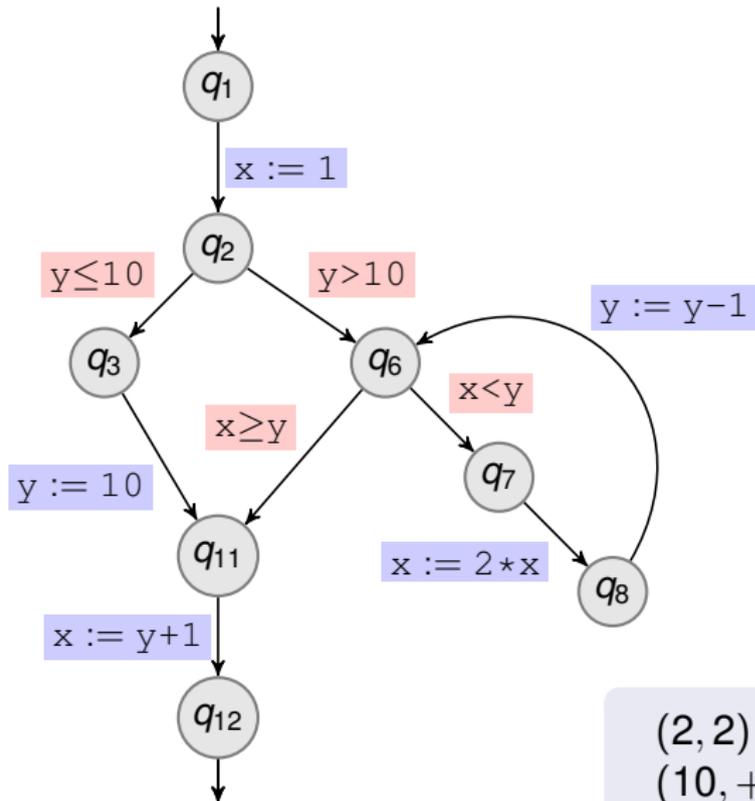
# Range Analysis on Running Example



Recall that  $T = (-\infty, +\infty)$

	x		y	
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	2	9	$+\infty$
$q_7$	1	2	9	$+\infty$
$q_8$	2	2	10	$+\infty$
$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10

# Range Analysis on Running Example

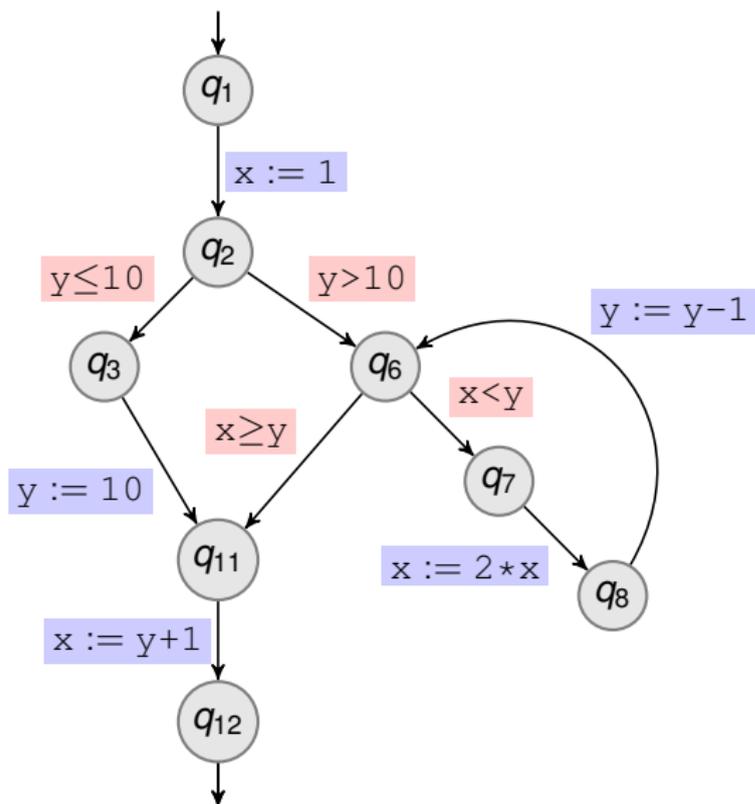


Recall that  $T = (-\infty, +\infty)$

	x		y	
	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	2	9	$+\infty$
$q_7$	1	2	9	$+\infty$
$q_8$	2	4	9	$+\infty$
$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10

$$\begin{aligned}
 (2, 2) &\sqcup (2, 4) = (2, 4) \\
 (10, +\infty) &\sqcup (9, +\infty) = (9, +\infty)
 \end{aligned}$$

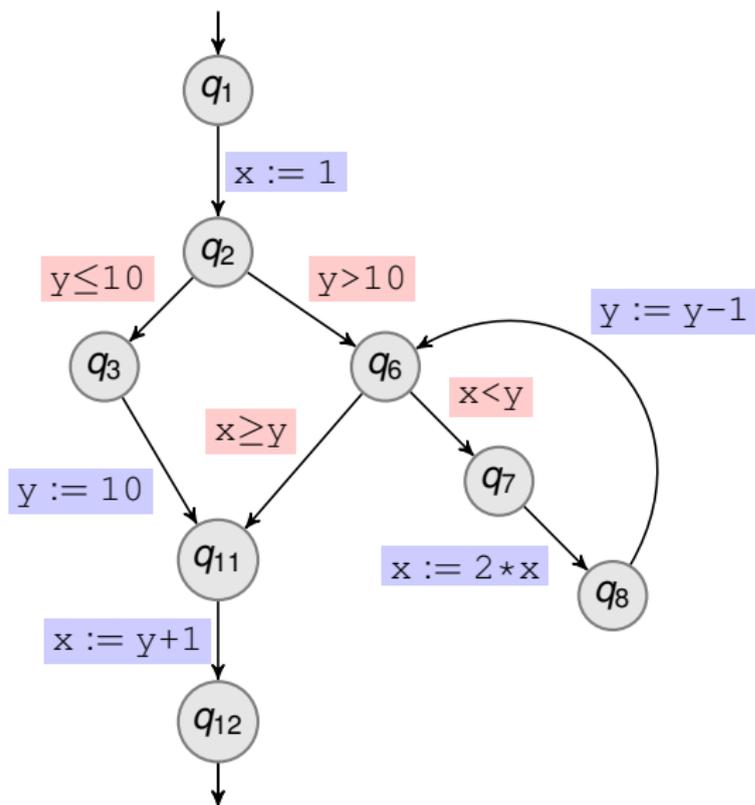
# Range Analysis on Running Example



Recall that  $T = (-\infty, +\infty)$

	x		y	
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	2	9	$+\infty$
$q_7$	1	2	9	$+\infty$
$q_8$	2	4	9	$+\infty$
$q_{11}$	1	1	10	10
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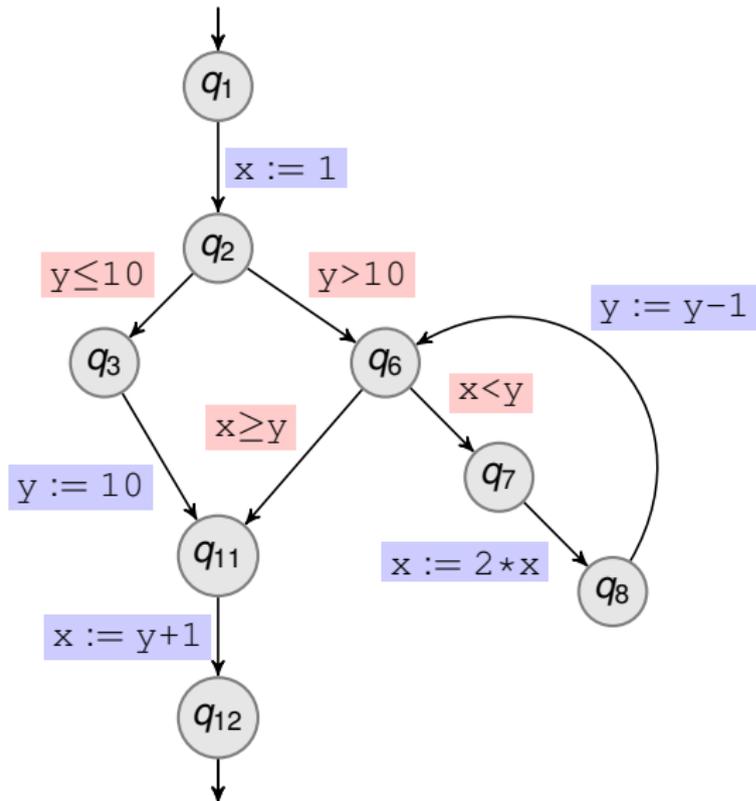
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Recall that  $T = (-\infty, +\infty)$

	x		y	
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	4	8	$+\infty$
$q_7$	1	2	9	$+\infty$
$q_8$	2	4	9	$+\infty$
$q_{11}$	1	1	10	10
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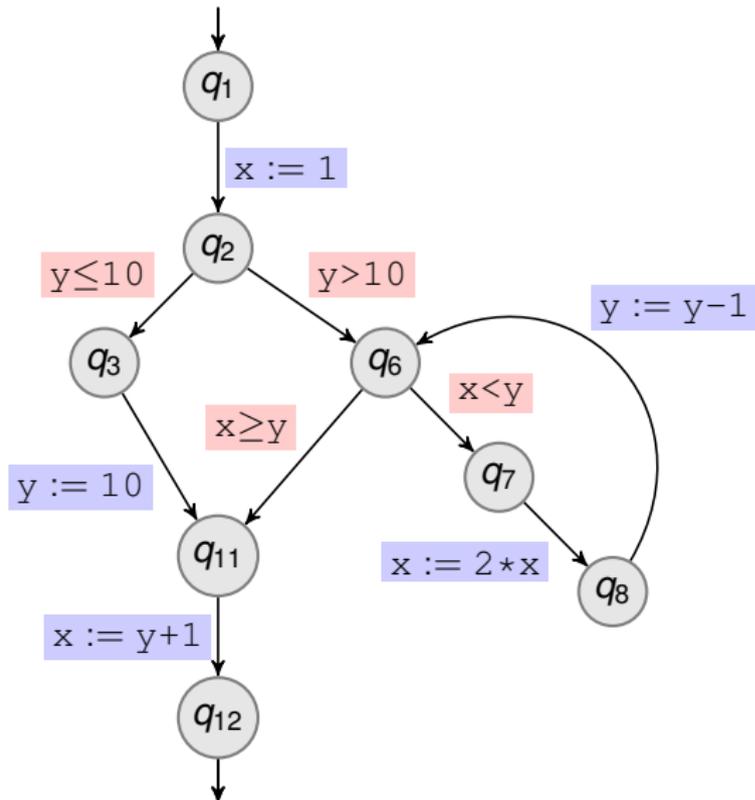
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Recall that  $T = (-\infty, +\infty)$

	x		y	
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	4	8	$+\infty$
$q_7$	1	2	9	$+\infty$
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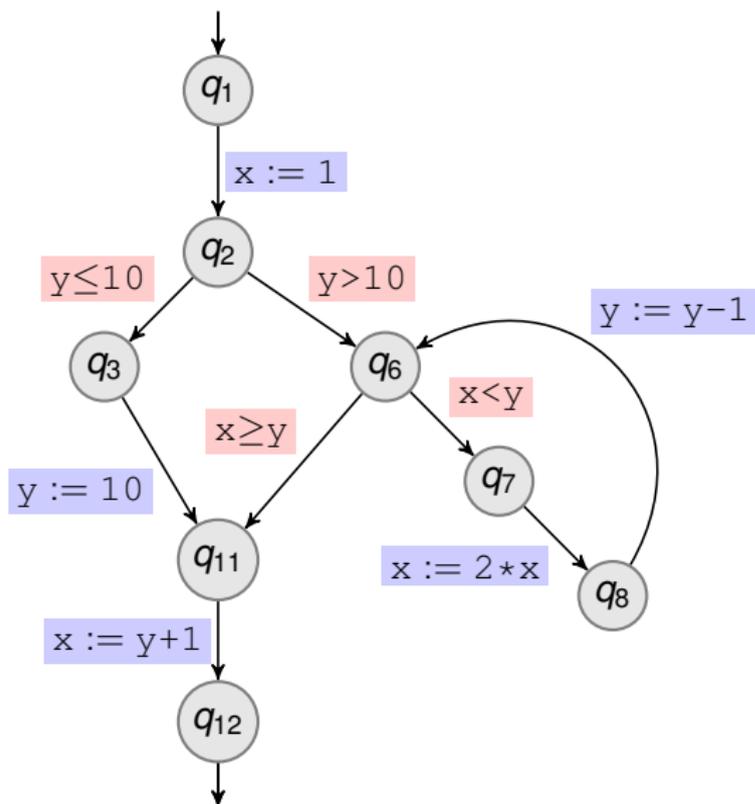
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Recall that  $T = (-\infty, +\infty)$

	x		y	
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	4	8	$+\infty$
$q_7$	1	4	8	$+\infty$
$q_8$	2	4	9	$+\infty$
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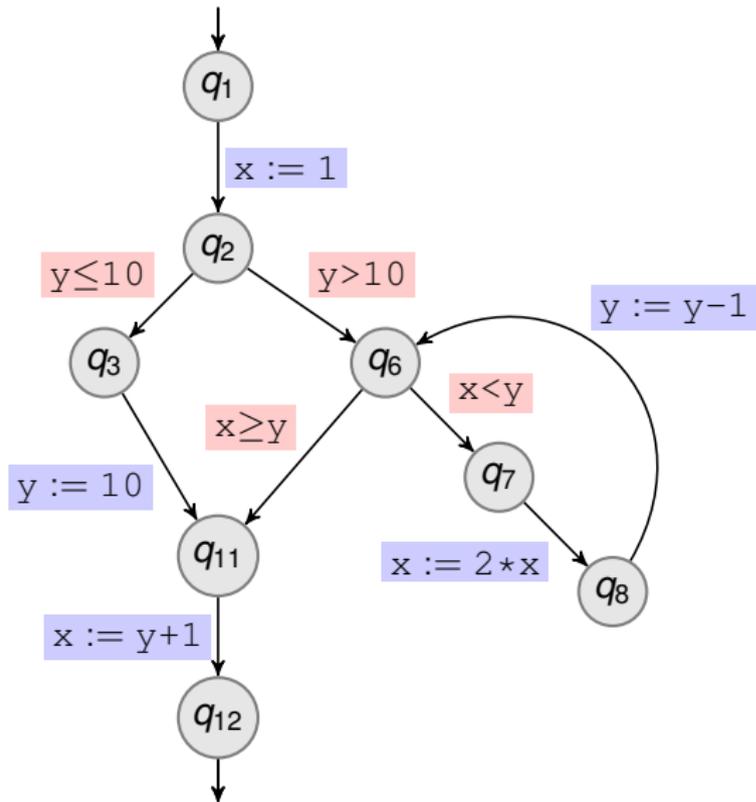
# Range Analysis on Running Example



Recall that  $T = (-\infty, +\infty)$

	x		y	
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	4	8	$+\infty$
$q_7$	1	4	8	$+\infty$
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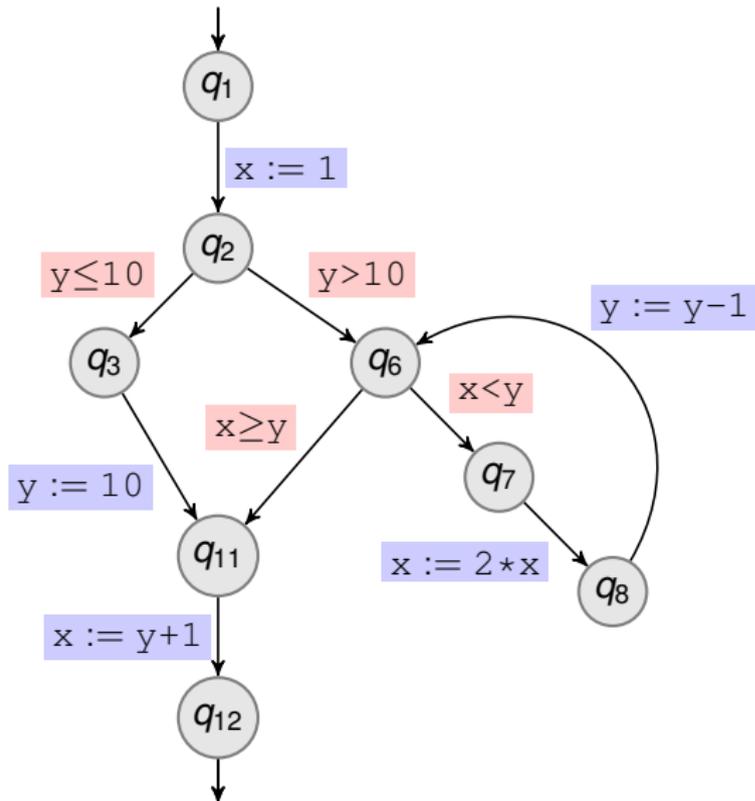
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Recall that  $T = (-\infty, +\infty)$

	x		y	
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	4	8	$+\infty$
$q_7$	1	4	8	$+\infty$
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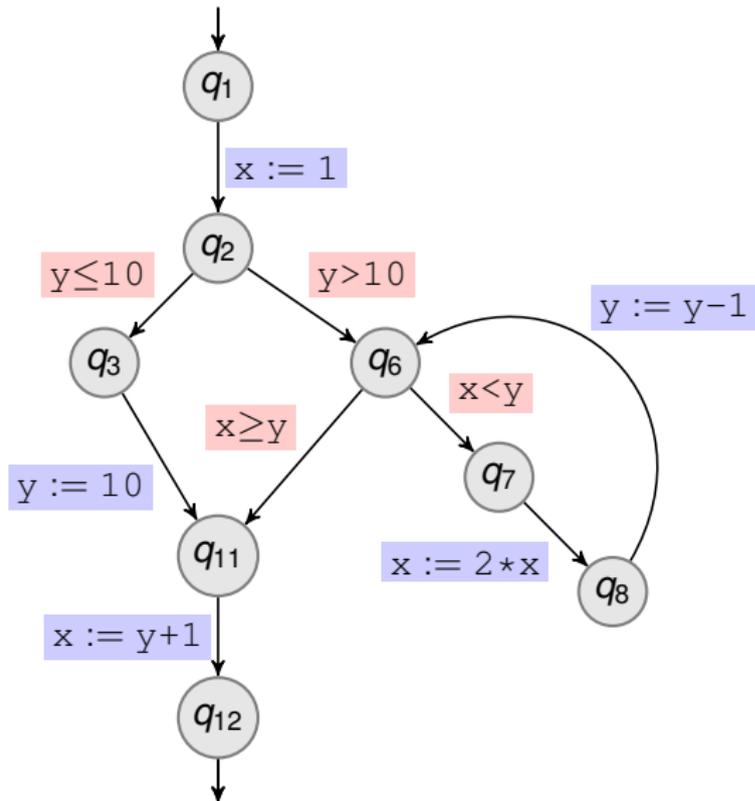
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$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
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$q_3$	1	1	$-\infty$	10
$q_6$	1	4	8	$+\infty$
$q_7$	1	4	8	$+\infty$
$q_8$	2	8	8	$+\infty$
$q_{11}$	1	1	10	10
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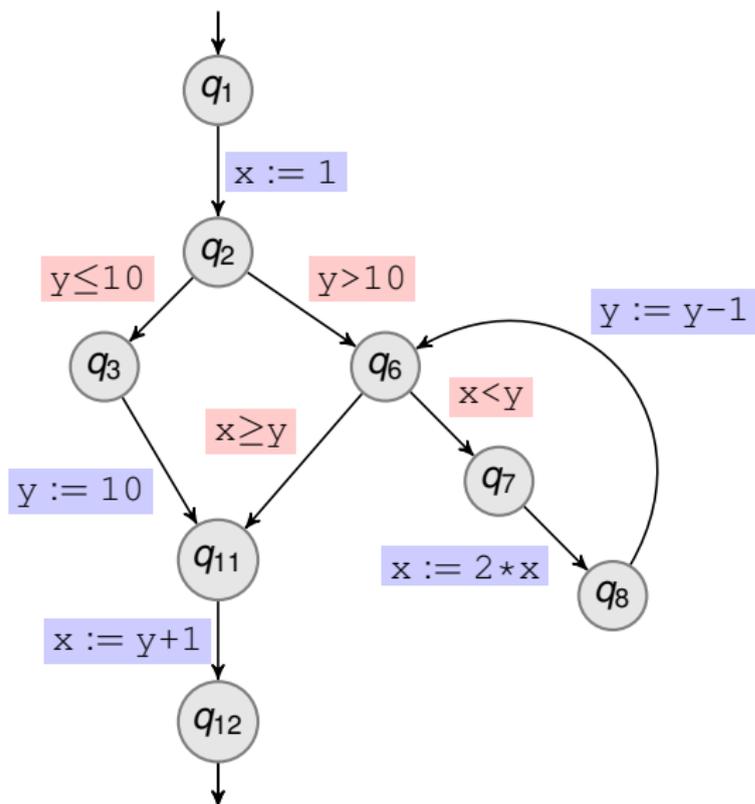
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$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
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$q_3$	1	1	$-\infty$	10
$q_6$	1	8	7	$+\infty$
$q_7$	1	4	8	$+\infty$
$q_8$	2	8	8	$+\infty$
$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10

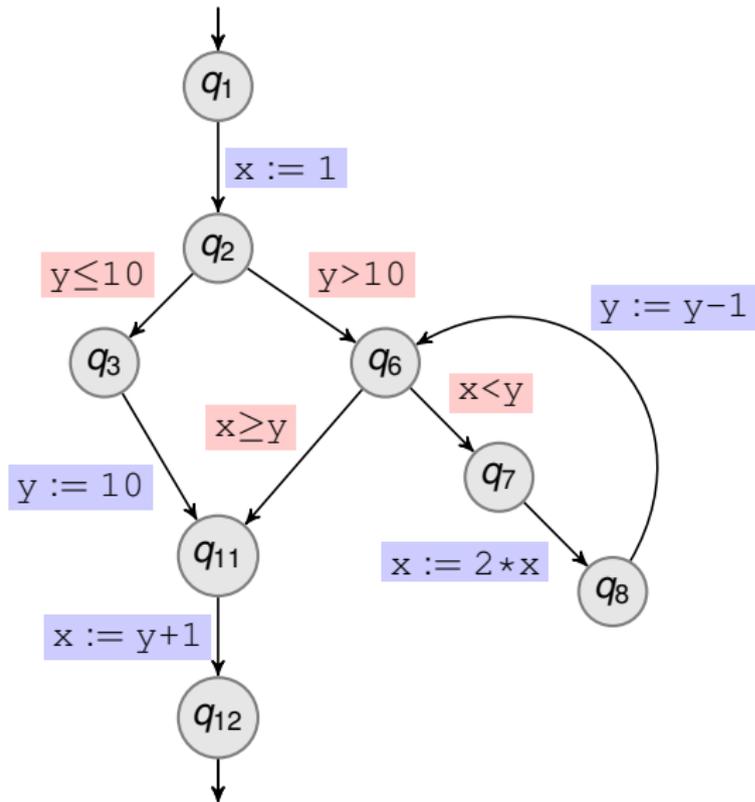
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$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10

# Range Analysis on Running Example

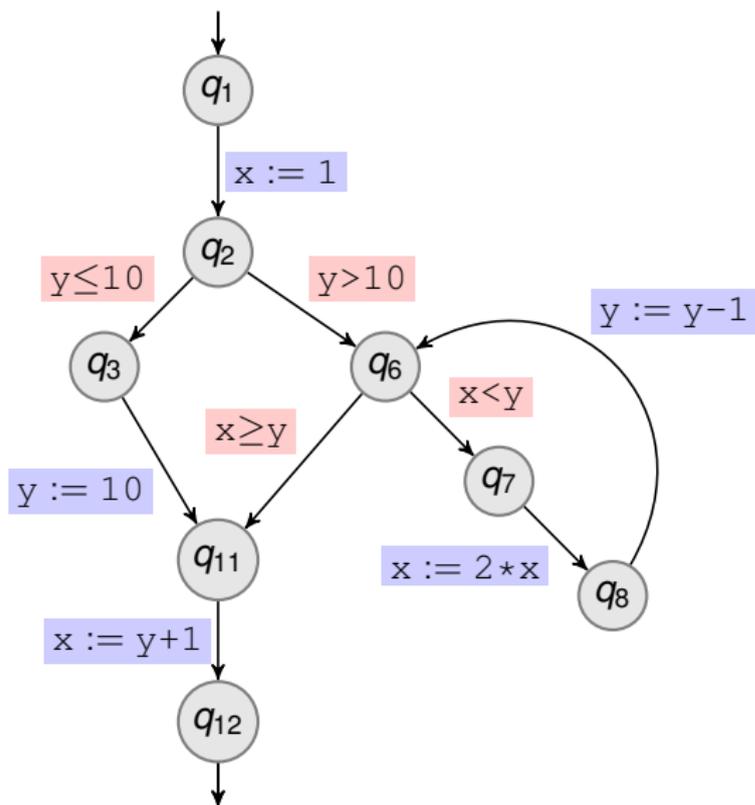


Recall that  $\top = (-\infty, +\infty)$

	x		y	
	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	8	7	$+\infty$
$q_7$	1	8	7	$+\infty$
$q_8$	2	8	8	$+\infty$
$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10

Nothing to be gained from guard  $x < y$

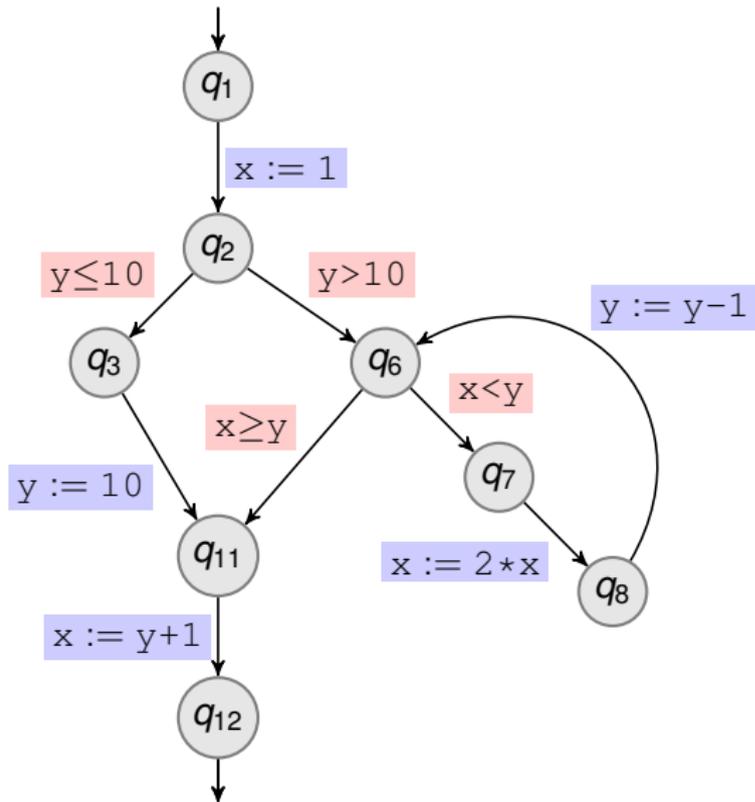
# Range Analysis on Running Example



Recall that  $T = (-\infty, +\infty)$

	x		y	
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	8	7	$+\infty$
$q_7$	1	8	7	$+\infty$
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$q_{11}$	1	1	10	10
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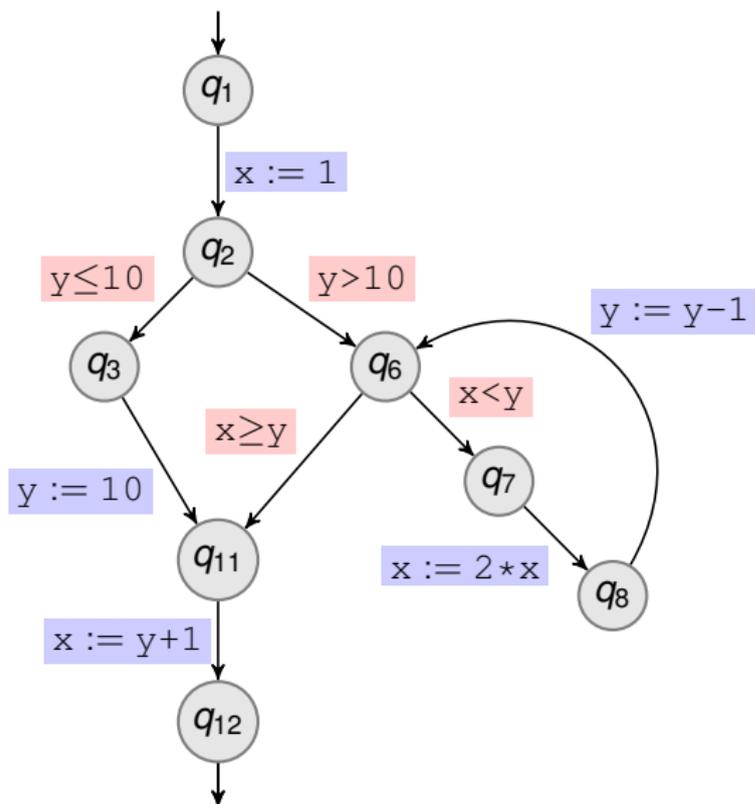
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Recall that  $T = (-\infty, +\infty)$

	x		y	
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	8	7	$+\infty$
$q_7$	1	8	7	$+\infty$
$q_8$	2	16	7	$+\infty$
$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10

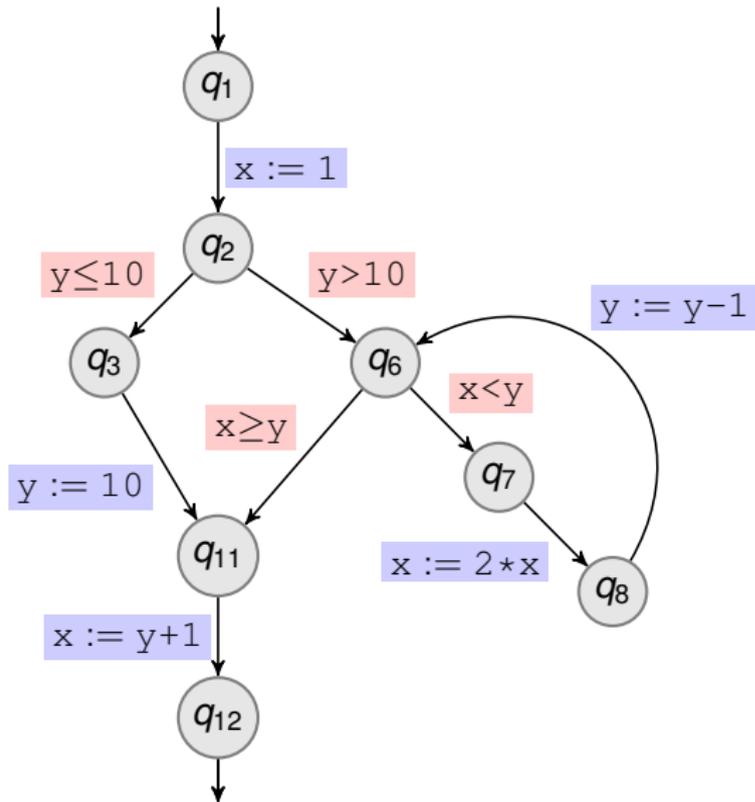
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Recall that  $T = (-\infty, +\infty)$

	x		y	
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	8	7	$+\infty$
$q_7$	1	8	7	$+\infty$
$q_8$	2	16	7	$+\infty$
$q_{11}$	1	1	10	10
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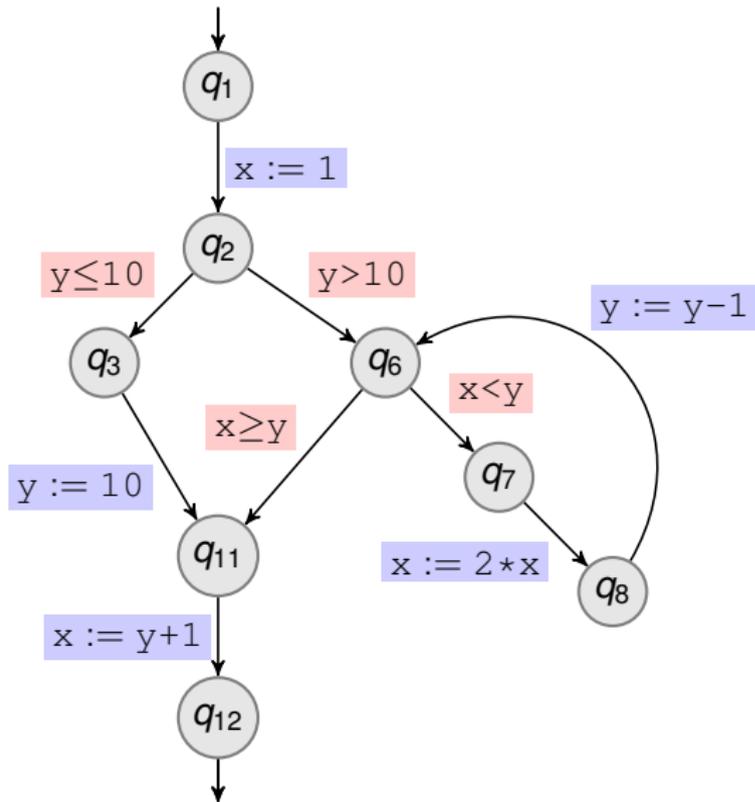
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Recall that  $T = (-\infty, +\infty)$

	x		y	
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	16	6	$+\infty$
$q_7$	1	16	6	$+\infty$
$q_8$	2	32	6	$+\infty$
$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10

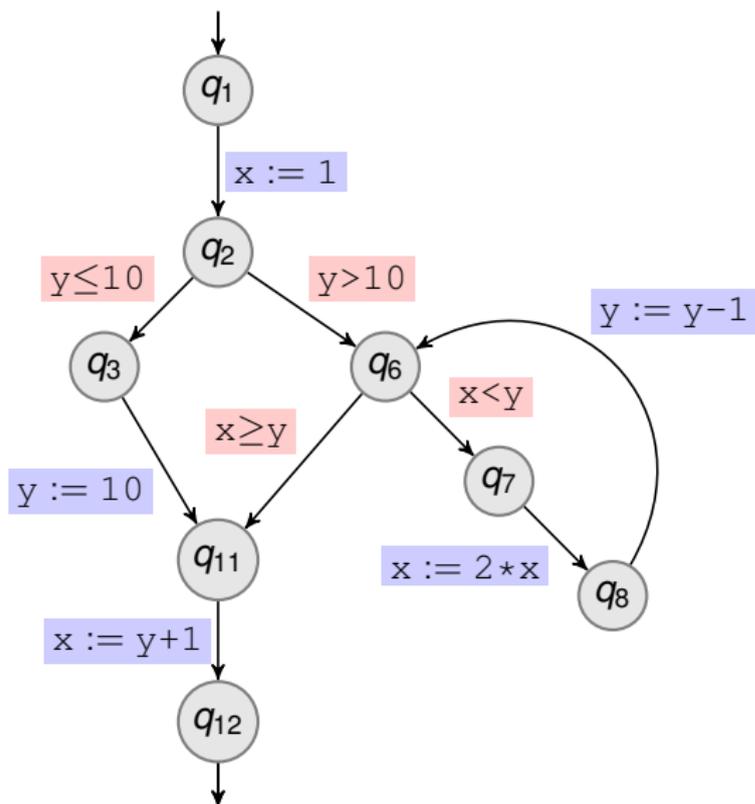
# Range Analysis on Running Example



Recall that  $T = (-\infty, +\infty)$

	x		y	
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	32	5	$+\infty$
$q_7$	1	32	5	$+\infty$
$q_8$	2	$2^6$	5	$+\infty$
$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10

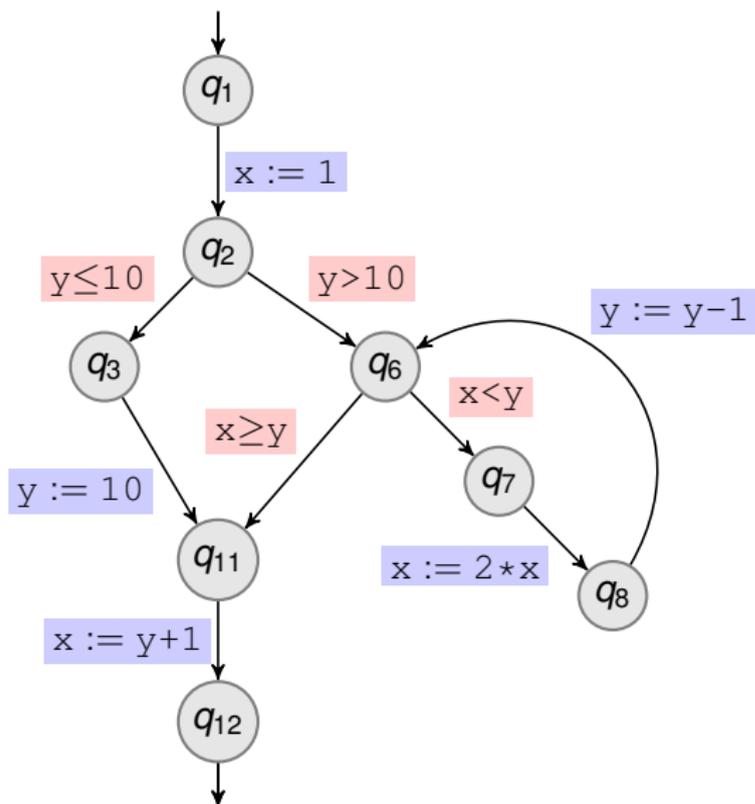
# Range Analysis on Running Example



Recall that  $T = (-\infty, +\infty)$

	x		y	
	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	$2^6$	4	$+\infty$
$q_7$	1	$2^6$	4	$+\infty$
$q_8$	2	$2^7$	4	$+\infty$
$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10

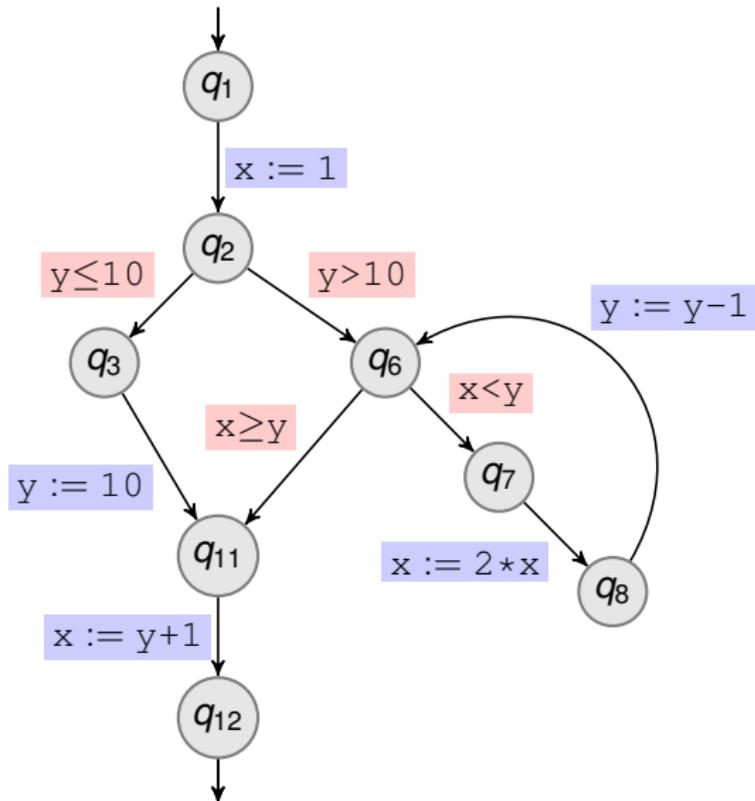
# Range Analysis on Running Example



Recall that  $T = (-\infty, +\infty)$

	x		y	
	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	$2^7$	3	$+\infty$
$q_7$	1	$2^7$	3	$+\infty$
$q_8$	2	$2^8$	3	$+\infty$
$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10

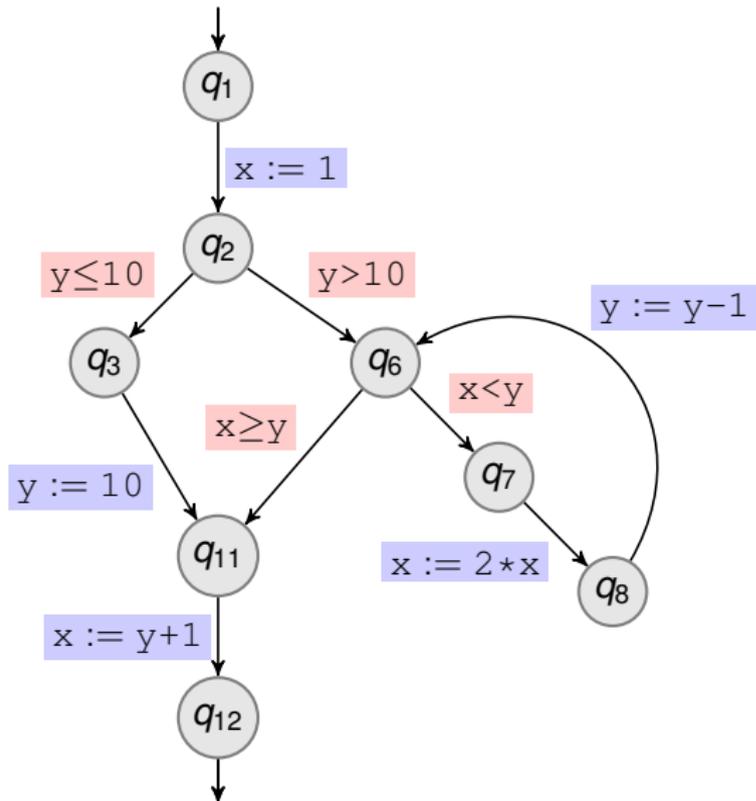
# Range Analysis on Running Example



Recall that  $T = (-\infty, +\infty)$

	x		y	
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	$2^8$	2	$+\infty$
$q_7$	1	$2^8$	2	$+\infty$
$q_8$	2	$2^9$	2	$+\infty$
$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10

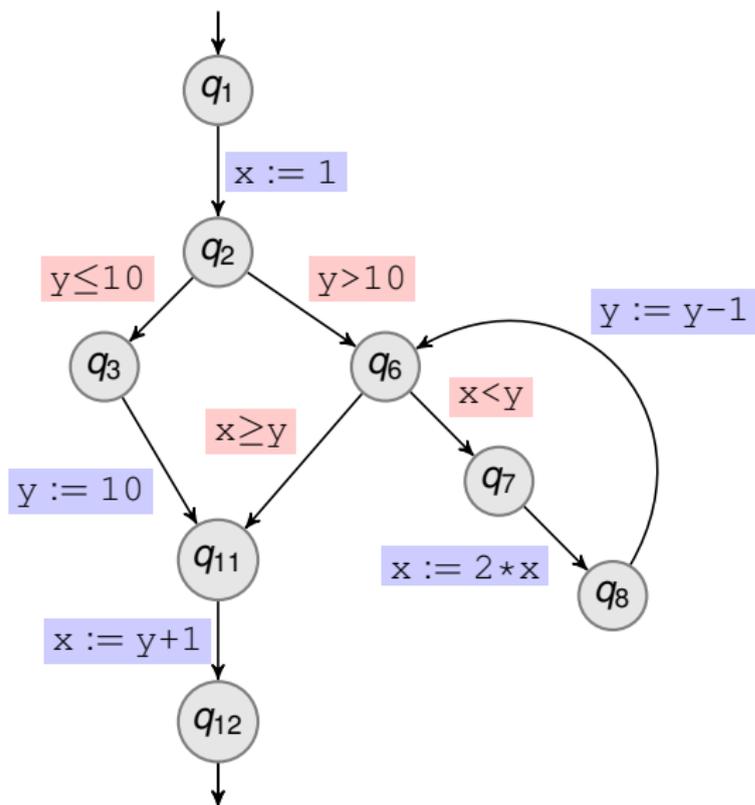
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Recall that  $T = (-\infty, +\infty)$

	x		y	
	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	$2^9$	1	$+\infty$
$q_7$	1	$2^9$	1	$+\infty$
$q_8$	2	$2^{10}$	1	$+\infty$
$q_{11}$	1	1	10	10
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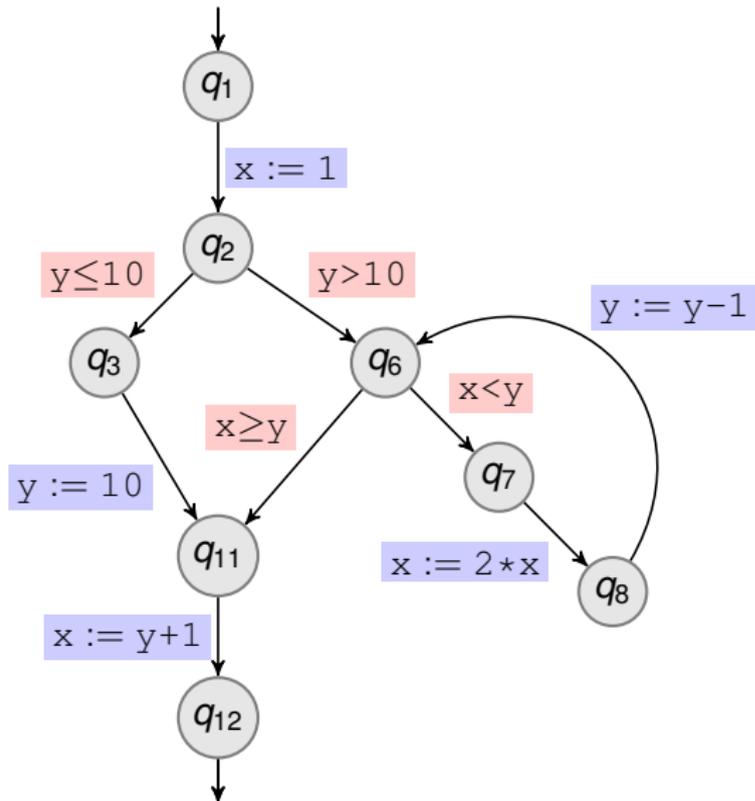
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Recall that  $T = (-\infty, +\infty)$

	x		y	
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	$2^9$	1	$+\infty$
$q_7$	1	$2^9$	1	$+\infty$
$q_8$	2	$2^{10}$	1	$+\infty$
$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10

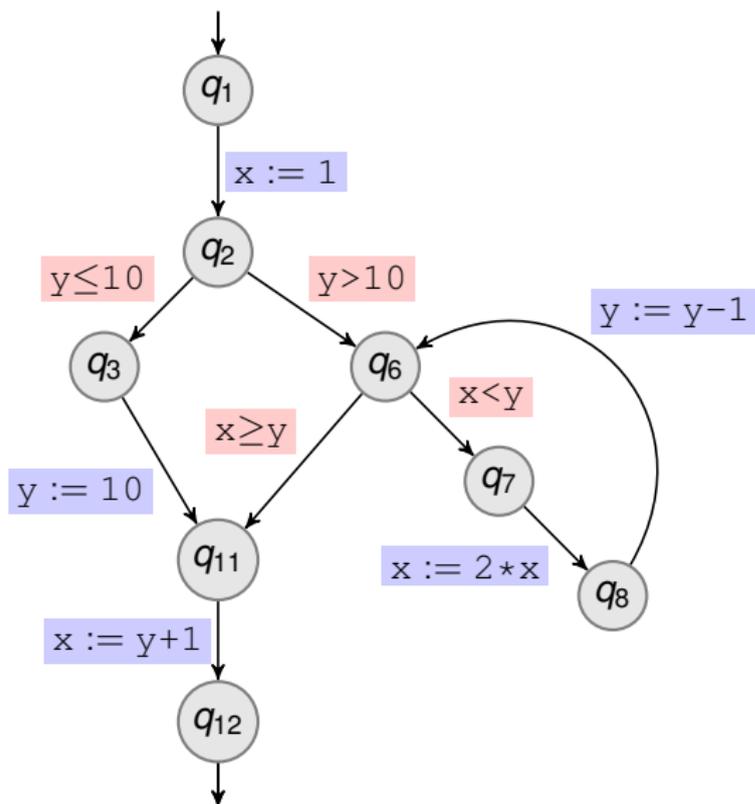
# Range Analysis on Running Example



Recall that  $T = (-\infty, +\infty)$

	x		y	
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	$2^{10}$	0	$+\infty$
$q_7$	1	$2^9$	1	$+\infty$
$q_8$	2	$2^{10}$	1	$+\infty$
$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10

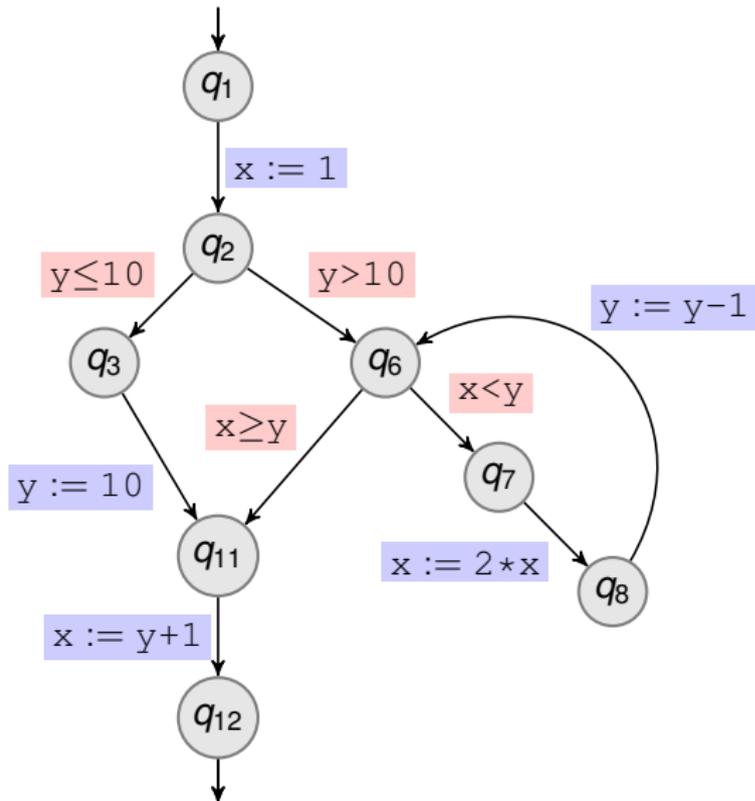
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Recall that  $T = (-\infty, +\infty)$

	x		y	
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	$2^{10}$	0	$+\infty$
$q_7$	1	$2^9$	1	$+\infty$
$q_8$	2	$2^{10}$	1	$+\infty$
$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10

# Range Analysis on Running Example

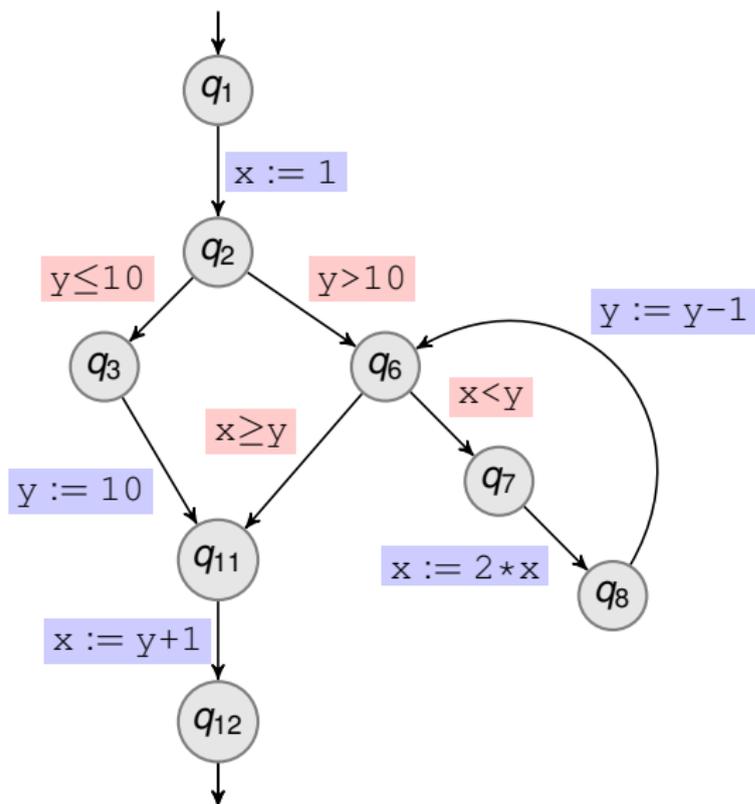


Recall that  $T = (-\infty, +\infty)$

	x		y	
	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	$2^{10}$	0	$+\infty$
$q_7$	1	$2^{10}$	1	$+\infty$
$q_8$	2	$2^{10}$	1	$+\infty$
$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10

Gained from guard  $x < y$

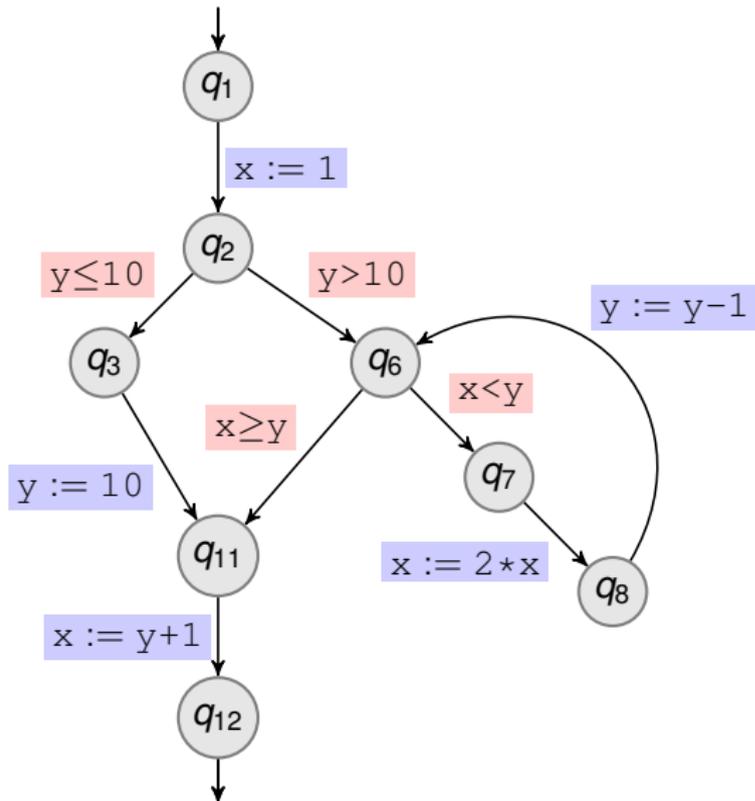
# Range Analysis on Running Example



Recall that  $T = (-\infty, +\infty)$

	x		y	
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	$2^{10}$	0	$+\infty$
$q_7$	1	$2^{10}$	1	$+\infty$
$q_8$	2	$2^{10}$	1	$+\infty$
$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10

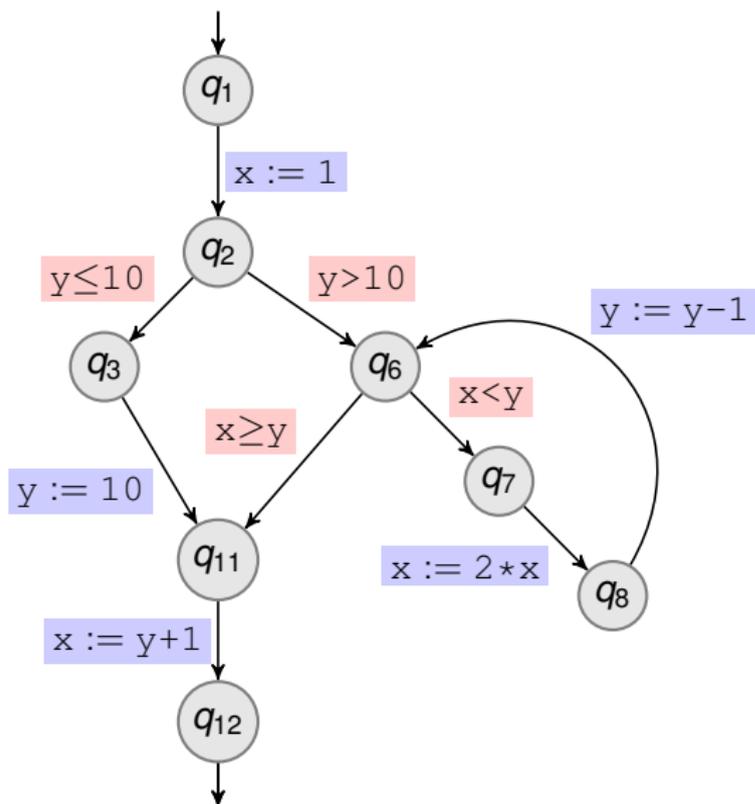
# Range Analysis on Running Example



Recall that  $T = (-\infty, +\infty)$

	x		y	
	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	$2^{10}$	0	$+\infty$
$q_7$	1	$2^{10}$	1	$+\infty$
$q_8$	2	$2^{11}$	1	$+\infty$
$q_{11}$	1	1	10	10
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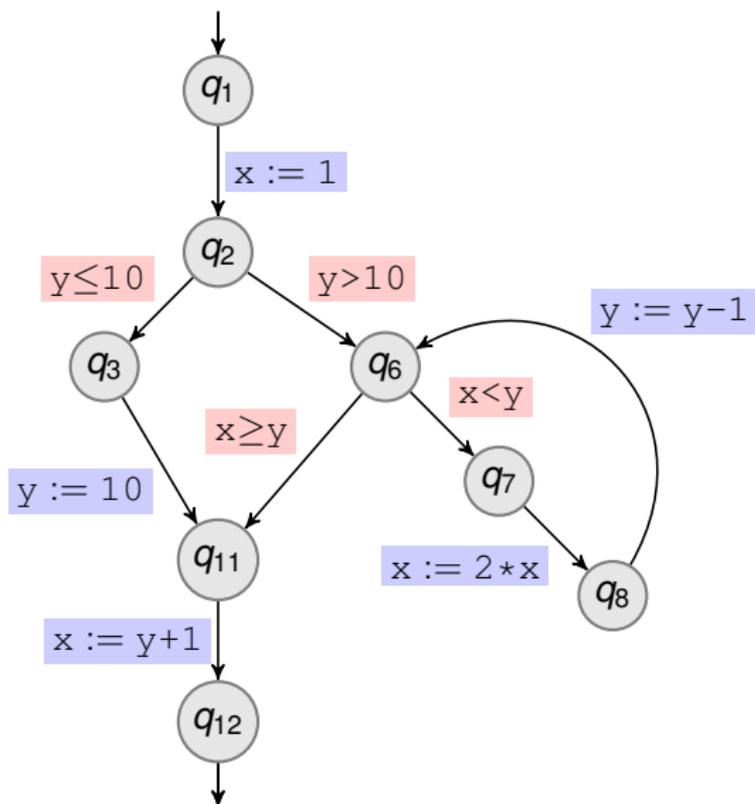
# Range Analysis on Running Example



Recall that  $T = (-\infty, +\infty)$

	x		y	
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	$2^{10}$	0	$+\infty$
$q_7$	1	$2^{10}$	1	$+\infty$
$q_8$	2	$2^{11}$	1	$+\infty$
$q_{11}$	1	1	10	10
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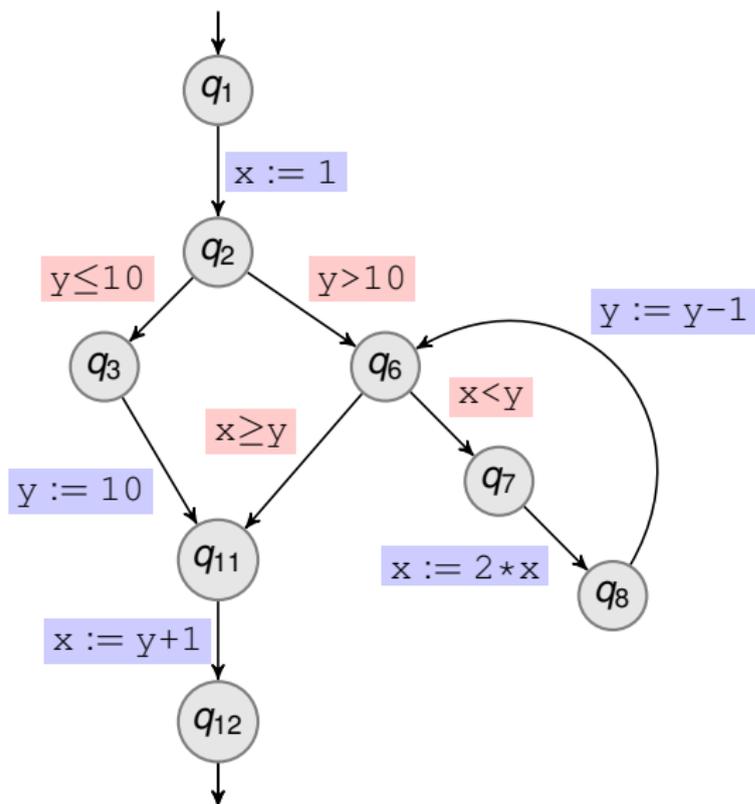
# Range Analysis on Running Example



Recall that  $T = (-\infty, +\infty)$

	x		y	
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	$2^{11}$	0	$+\infty$
$q_7$	1	$2^{11}$	1	$+\infty$
$q_8$	2	$2^{12}$	1	$+\infty$
$q_{11}$	1	1	10	10
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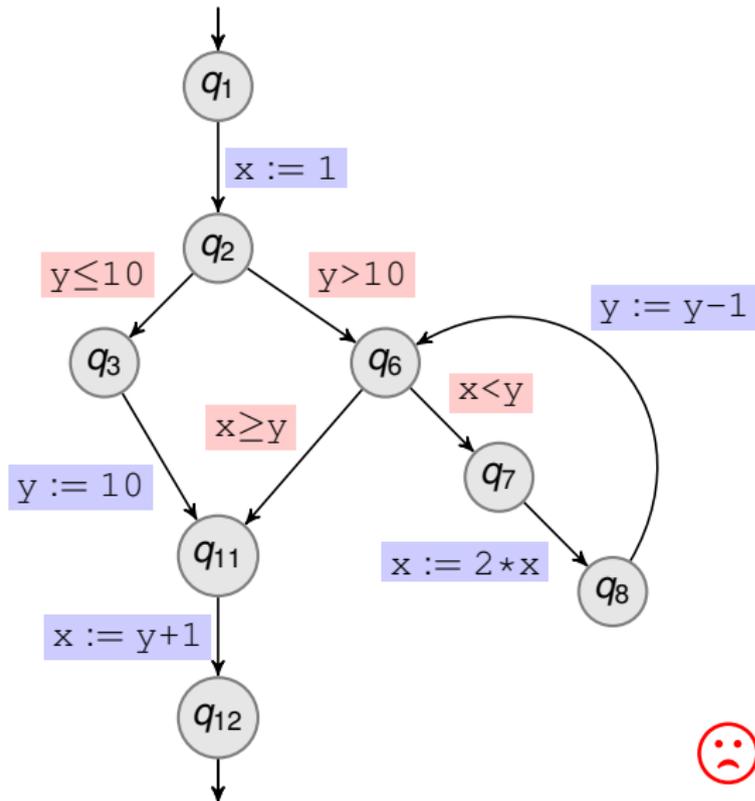
# Range Analysis on Running Example



Recall that  $T = (-\infty, +\infty)$

	x		y	
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	$2^{12}$	0	$+\infty$
$q_7$	1	$2^{12}$	1	$+\infty$
$q_8$	2	$2^{13}$	1	$+\infty$
$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10

# Range Analysis on Running Example



Recall that  $T = (-\infty, +\infty)$

	x		y	
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	...	0	$+\infty$
$q_7$	1	...	1	$+\infty$
$q_8$	2	...	1	$+\infty$
$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10



Does not converge!

# Dynamic Approximation: Widening Operators

Consider a complete lattice  $(L, \sqsubseteq)$ .

## Objective of Widening Operators

**Soundly** extrapolate “limits” of ascending chains

## Definition

A **widening** operator for  $(L, \sqsubseteq)$  is a function  $\nabla : (L \times L) \rightarrow L$  such that:

- 1  $x \sqcup y \sqsubseteq x \nabla y$  (for all  $x, y \in L$ )
- 2 for any ascending chain  $x_0 \sqsubseteq x_1 \sqsubseteq \dots$  of elements of  $L$ , the ascending chain  $y_0 \sqsubseteq y_1 \sqsubseteq \dots$  defined by

$$\begin{cases} y_0 = x_0 \\ y_{i+1} = y_i \nabla x_{i+1} \quad \text{for all } i \in \mathbb{N} \end{cases}$$

is not strictly increasing (i.e.  $y_{i+1} = y_i$  for some  $i \in \mathbb{N}$ ).

# Correctness of Kleene Iteration with Widening

Consider a complete lattice  $(L, \sqsubseteq)$  and a monotonic function  $f : L \rightarrow L$ .

## Theorem

If  $\nabla : (L \times L) \rightarrow L$  is a widening operator then the ascending chain  $x_0 \sqsubseteq x_1 \sqsubseteq \dots$  defined by

$$x_0 = \perp$$
$$x_{i+1} = \begin{cases} x_i & \text{if } f(x_i) \sqsubseteq x_i \\ x_i \nabla f(x_{i+1}) & \text{otherwise} \end{cases}$$

is eventually stationary, and its limit satisfies  $\bigsqcup \{x_i \mid i \in \mathbb{N}\} \sqsupseteq \text{lfp}(f)$ .

## Application to MFP Approximation in Data Flow Analysis

Replacing  $\sqcup$  with  $\nabla$  in Kleene / round-robin / worklist algorithms

- guarantees **termination**, but
- at the expense of **precision**.

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## Application to MFP Approximation in Data Flow Analysis

Replacing  $\sqcup$  with  $\nabla$  in Kleene / round-robin / worklist algorithms

- guarantees **termination**, but
- at the expense of **precision**.

# (Forward) Round-Robin Iteration with Widening

Consider a data flow instance  $\langle (L, \sqsubseteq), \mathcal{F}, Q, q_{in}, q_{out}, X, \rightarrow, f, \iota \rangle$ .

```
foreach  $q \in Q$ 
   $a[q] \leftarrow \perp$ 
 $a[q_{in}] \leftarrow \iota$ 
do
   $change \leftarrow false$ 
  foreach  $q \xrightarrow{op} q'$ 
     $new \leftarrow f_{op}(a[q])$ 
    if  $new \not\sqsubseteq a[q']$ 
       $a[q'] \leftarrow a[q'] \nabla new$ 
       $change \leftarrow true$ 
while  $change$ 
return  $a$ 
```

If  $\nabla$  is a widening operator on  $(L, \sqsubseteq)$  then:

- this algorithm **terminates** for any data flow instance on  $(L, \sqsubseteq)$ .
- the returned  $a \in Q \rightarrow L$  satisfies:

$$\overrightarrow{MFP}(q) \sqsubseteq a(q)$$

for every  $q \in Q$ .

# Widening Operator for Range Analysis: Intuition

## Objective of Widening Operators

**Soundly** extrapolate “limits” of ascending chains

Put  $\infty$  when the bound is moving towards  $\infty$

## Examples

$\dots, (1, 2), (1, 3), (1, 4) \longrightarrow (1, +\infty)$

$\dots, (1, 2), (-1, 2), (-6, 2) \longrightarrow (-\infty, 2)$

$\dots, (1, 2), (-9, 3), (-19, 4) \longrightarrow (-\infty, +\infty)$

$\dots, (1, +\infty), (-9, +\infty), (-19, +\infty) \longrightarrow (-\infty, +\infty)$

$\nabla$  only looks at the last two elements of the sequence

# Widening Operator for Range Analysis

## Widening Operator on the Complete Lattice $(Int, \sqsubseteq)$ of Intervals

$$\perp \nabla \perp = \perp \quad \perp \nabla (l, u) = (l, u) \nabla \perp = (l, u)$$

$$(l_1, u_1) \nabla (l_2, u_2) = (l_{\nabla}, u_{\nabla}) \quad \text{where} \quad \begin{cases} l_{\nabla} = \begin{cases} -\infty & \text{if } l_2 < l_1 \\ l_1 & \text{otherwise} \end{cases} \\ u_{\nabla} = \begin{cases} +\infty & \text{if } u_2 > u_1 \\ u_1 & \text{otherwise} \end{cases} \end{cases}$$

## Widening Operator on the Complete Lattice $(x \rightarrow Int, \sqsubseteq)$

Extension  $\nabla$  of the widening  $\nabla$  on  $(Int, \sqsubseteq)$  to  $(x \rightarrow Int, \sqsubseteq)$ , defined by:

$$\overline{v_1} \nabla \overline{v_2} = \lambda q. \overline{v_1}(q) \nabla \overline{v_2}(q)$$

# Widening Operator for Range Analysis

## Widening Operator on the Complete Lattice $(Int, \sqsubseteq)$ of Intervals

$$\perp \nabla \perp = \perp \quad \perp \nabla (l, u) = (l, u) \nabla \perp = (l, u)$$

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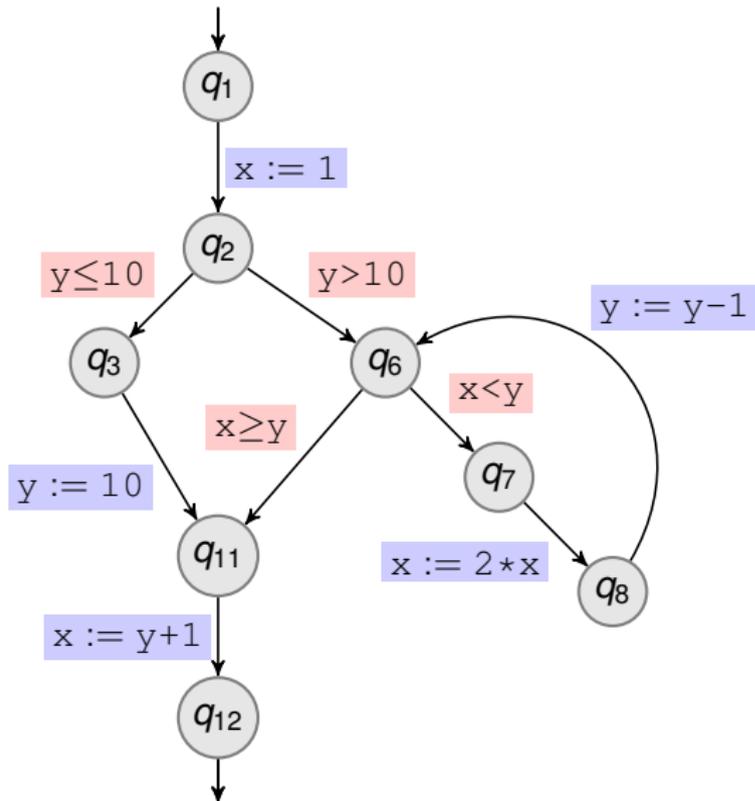
## Widening Operator on the Complete Lattice $(x \rightarrow Int, \overline{\sqsubseteq})$

Extension  $\nabla$  of the widening  $\nabla$  on  $(Int, \sqsubseteq)$  to  $(x \rightarrow Int, \overline{\sqsubseteq})$ , defined by:

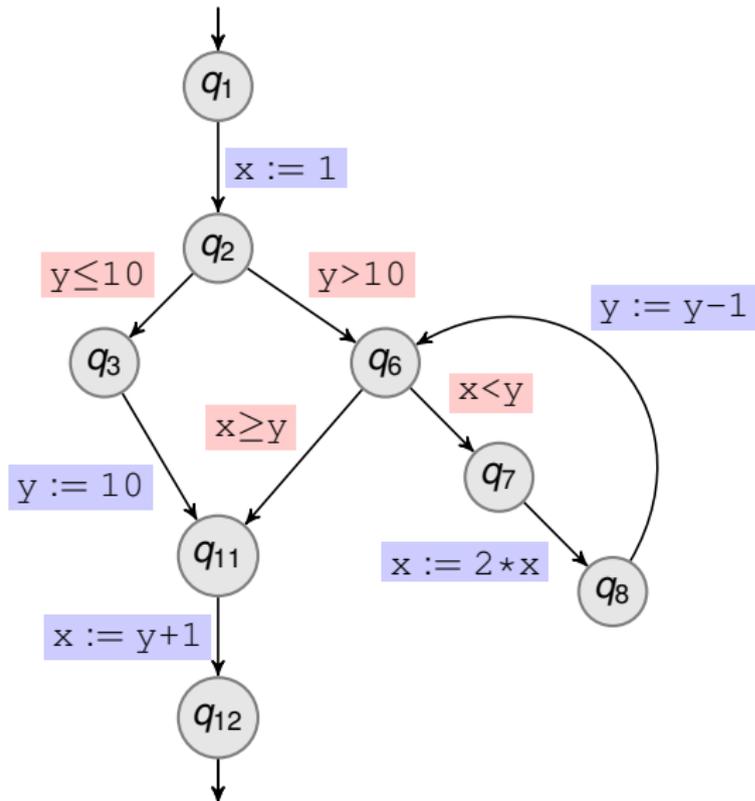
$$\overline{v_1} \nabla \overline{v_2} = \lambda q. \overline{v_1}(q) \nabla \overline{v_2}(q)$$

# Range Analysis on Example with Widening

Show that  $x > 0$  at  $q_{12}$



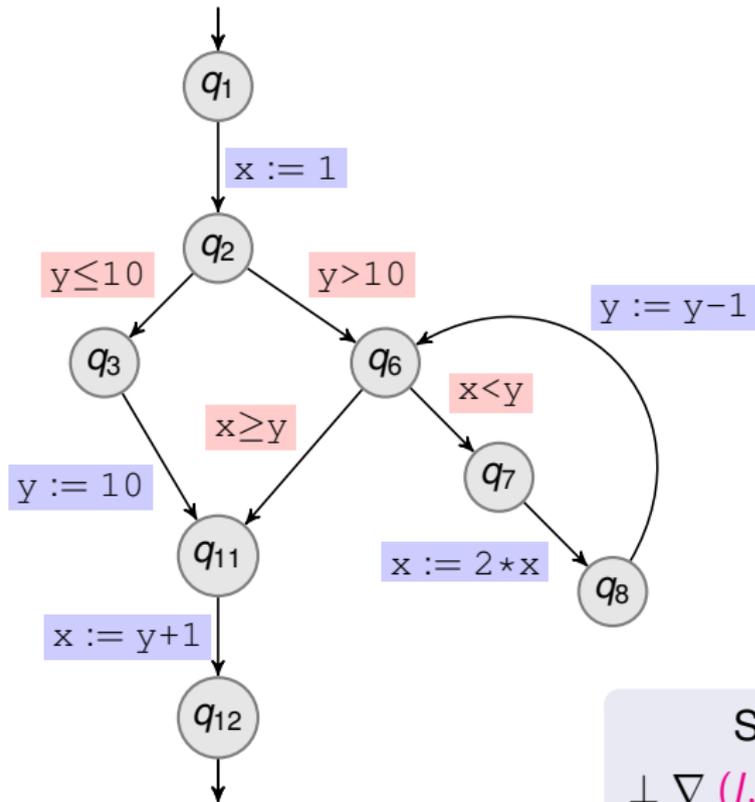
# Range Analysis on Example with Widening



Show that  $x > 0$  at  $q_{12}$

	x		y	
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	$\perp$	$\perp$	$\perp$	$\perp$
$q_3$	$\perp$	$\perp$	$\perp$	$\perp$
$q_6$	$\perp$	$\perp$	$\perp$	$\perp$
$q_7$	$\perp$	$\perp$	$\perp$	$\perp$
$q_8$	$\perp$	$\perp$	$\perp$	$\perp$
$q_{11}$	$\perp$	$\perp$	$\perp$	$\perp$
$q_{12}$	$\perp$	$\perp$	$\perp$	$\perp$

# Range Analysis on Example with Widening



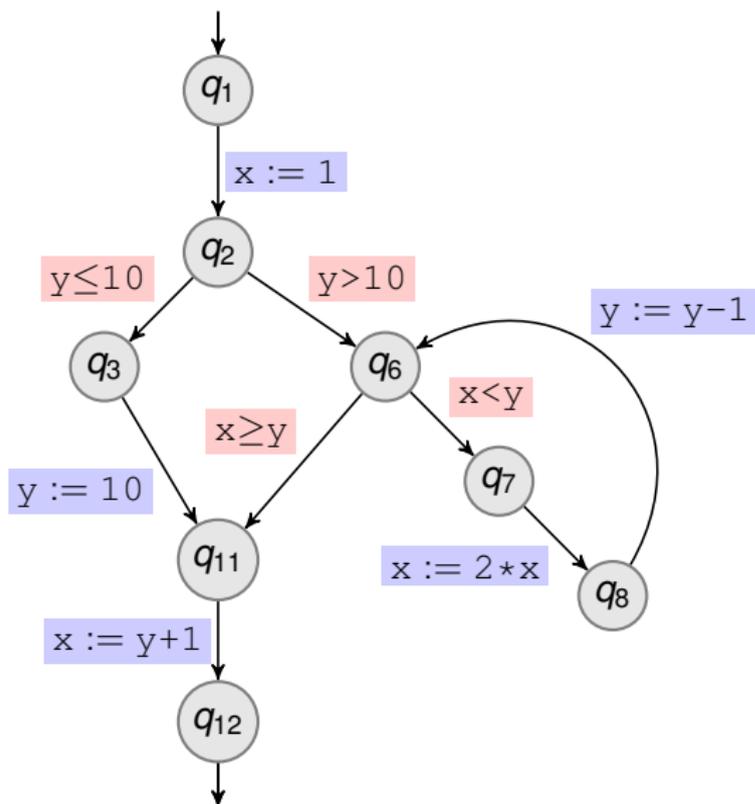
Show that  $x > 0$  at  $q_{12}$

	x		y	
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	1	10	$+\infty$
$q_7$	1	1	10	$+\infty$
$q_8$	2	2	10	$+\infty$
$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10

Same as without  $\nabla$  since

$$\perp \nabla (l, u) = \perp \sqcup (l, u) = (l, u)$$

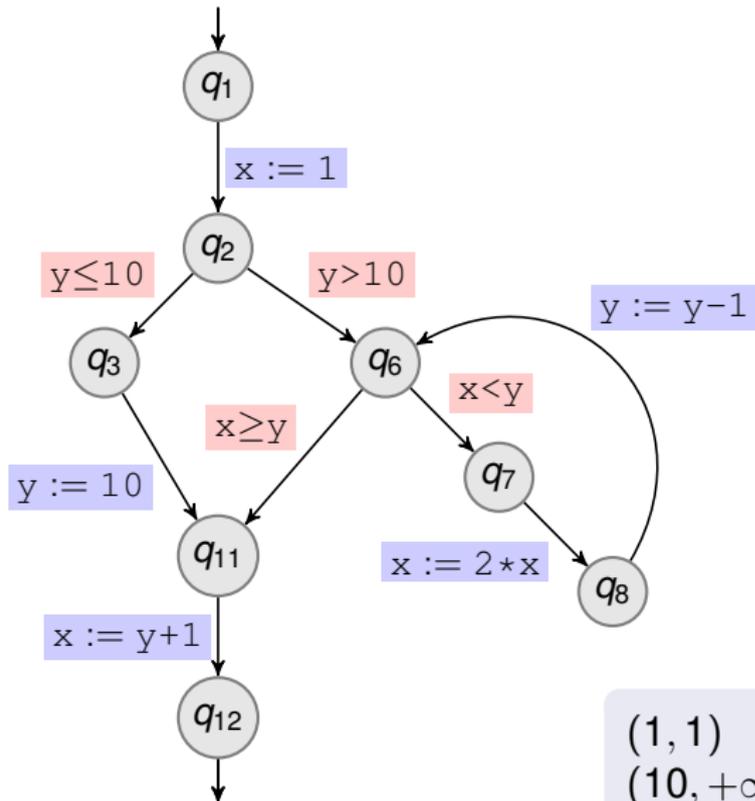
# Range Analysis on Example with Widening



Show that  $x > 0$  at  $q_{12}$

	x		y	
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	1	10	$+\infty$
$q_7$	1	1	10	$+\infty$
$q_8$	2	2	10	$+\infty$
$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10

# Range Analysis on Example with Widening

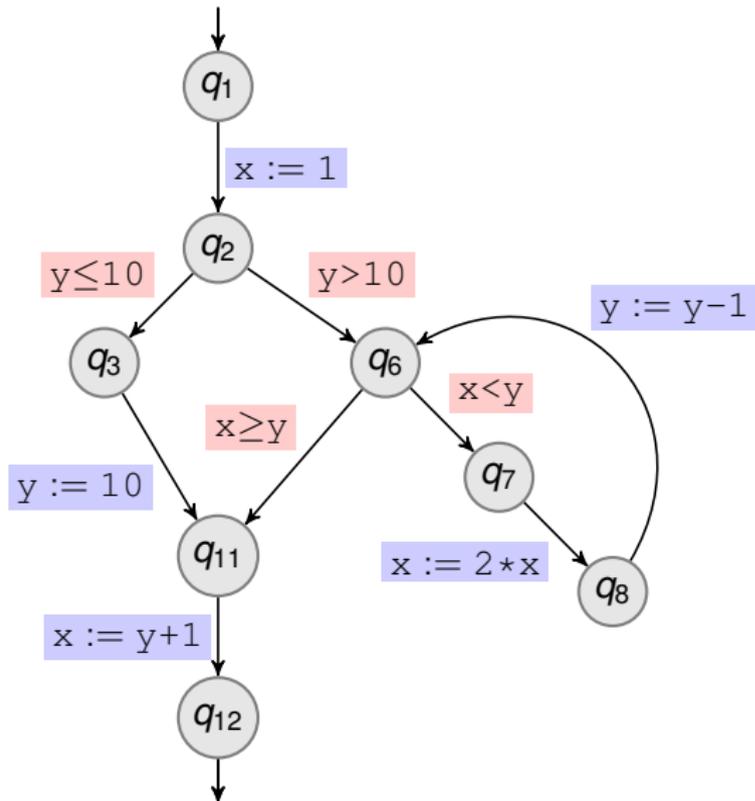


Show that  $x > 0$  at  $q_{12}$

	x		y	
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	$+\infty$	$-\infty$	$+\infty$
$q_7$	1	1	10	$+\infty$
$q_8$	2	2	10	$+\infty$
$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10

$$\begin{array}{l}
 (1, 1) \quad \nabla \quad (2, 2) \quad = \quad (1, +\infty) \\
 (10, +\infty) \quad \nabla \quad (9, +\infty) \quad = \quad (-\infty, +\infty)
 \end{array}$$

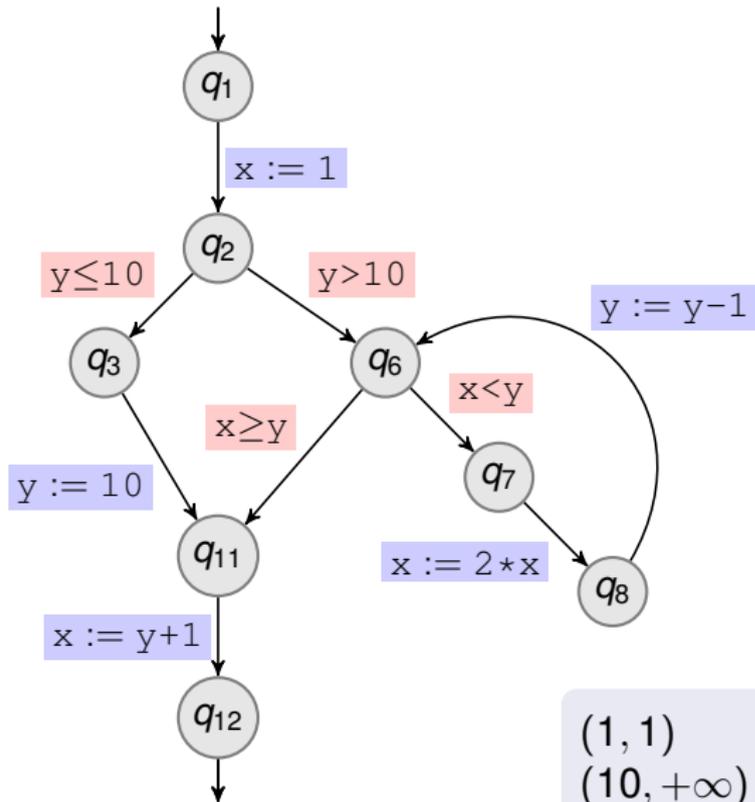
# Range Analysis on Example with Widening



Show that  $x > 0$  at  $q_{12}$

	x		y	
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	$+\infty$	$-\infty$	$+\infty$
$q_7$	1	1	10	$+\infty$
$q_8$	2	2	10	$+\infty$
$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10

# Range Analysis on Example with Widening

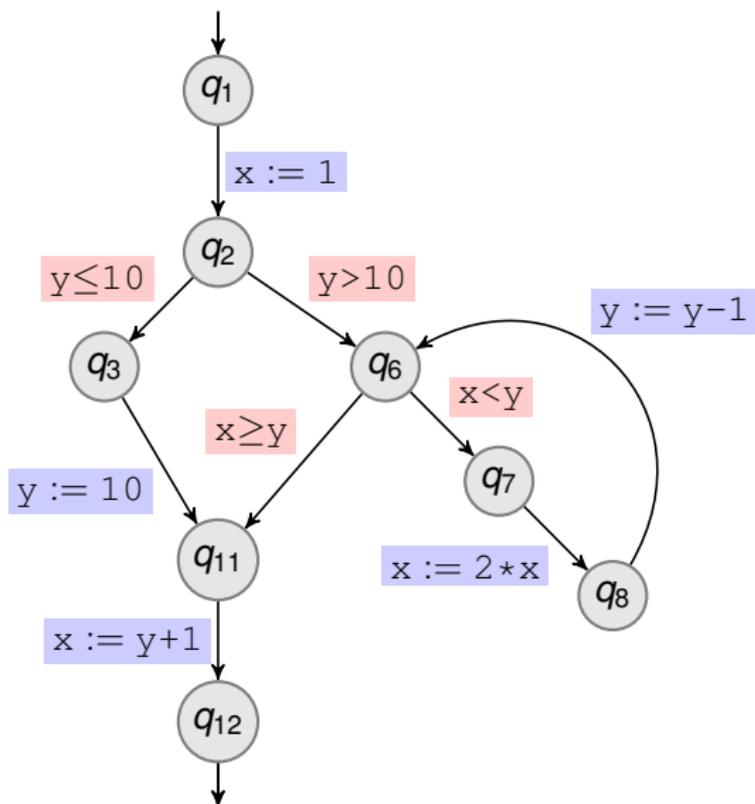


Show that  $x > 0$  at  $q_{12}$

	x		y	
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	$+\infty$	$-\infty$	$+\infty$
$q_7$	1	$+\infty$	$-\infty$	$+\infty$
$q_8$	2	2	10	$+\infty$
$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10

$$\begin{aligned}
 (1, 1) &\quad \nabla (1, +\infty) = (1, +\infty) \\
 (10, +\infty) &\quad \nabla (-\infty, +\infty) = (-\infty, +\infty)
 \end{aligned}$$

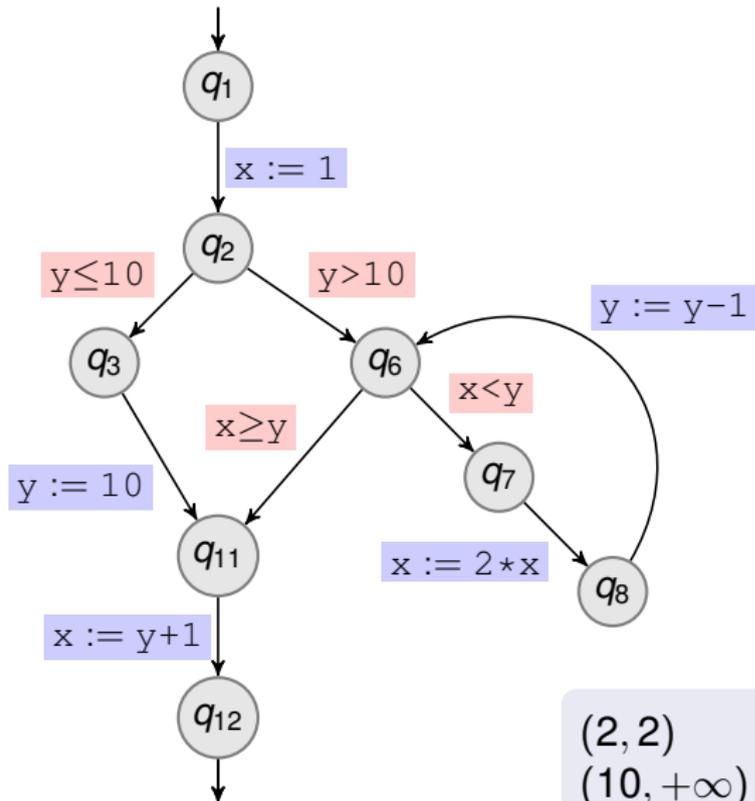
# Range Analysis on Example with Widening



Show that  $x > 0$  at  $q_{12}$

	x		y	
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	$+\infty$	$-\infty$	$+\infty$
$q_7$	1	$+\infty$	$-\infty$	$+\infty$
$q_8$	2	2	10	$+\infty$
$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10

# Range Analysis on Example with Widening



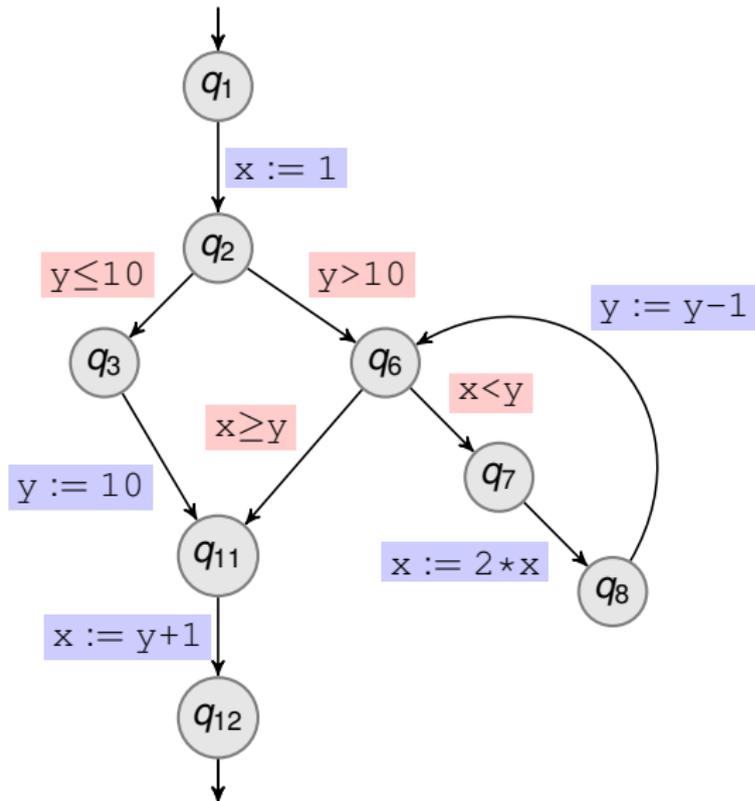
Show that  $x > 0$  at  $q_{12}$

	x		y	
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	$+\infty$	$-\infty$	$+\infty$
$q_7$	1	$+\infty$	$-\infty$	$+\infty$
$q_8$	2	$+\infty$	$-\infty$	$+\infty$
$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10

$$(2, 2) \quad \nabla \quad (2, +\infty) = (2, +\infty)$$

$$(10, +\infty) \quad \nabla \quad (-\infty, +\infty) = (-\infty, +\infty)$$

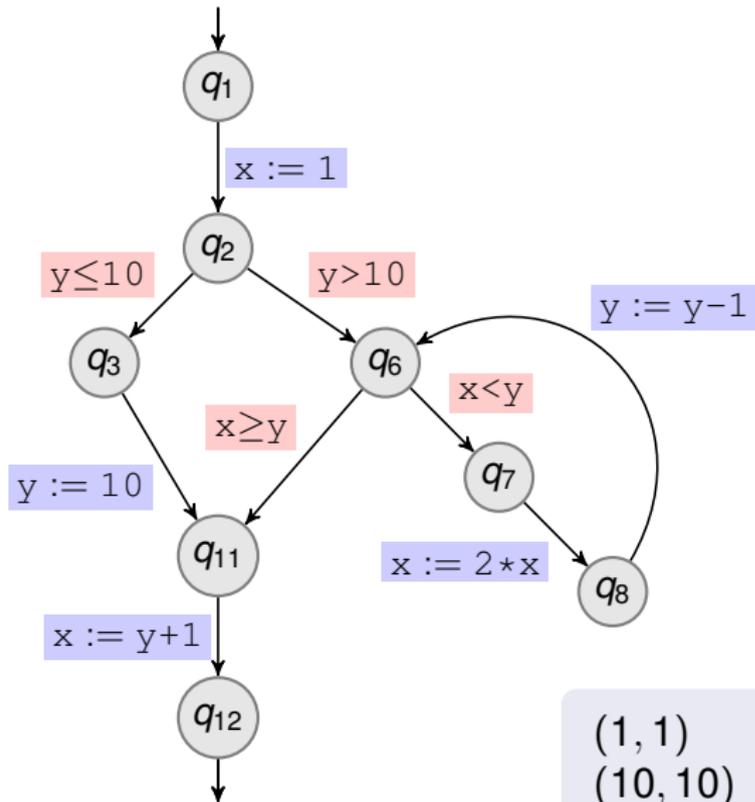
# Range Analysis on Example with Widening



Show that  $x > 0$  at  $q_{12}$

	x		y	
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	$+\infty$	$-\infty$	$+\infty$
$q_7$	1	$+\infty$	$-\infty$	$+\infty$
$q_8$	2	$+\infty$	$-\infty$	$+\infty$
$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10

# Range Analysis on Example with Widening

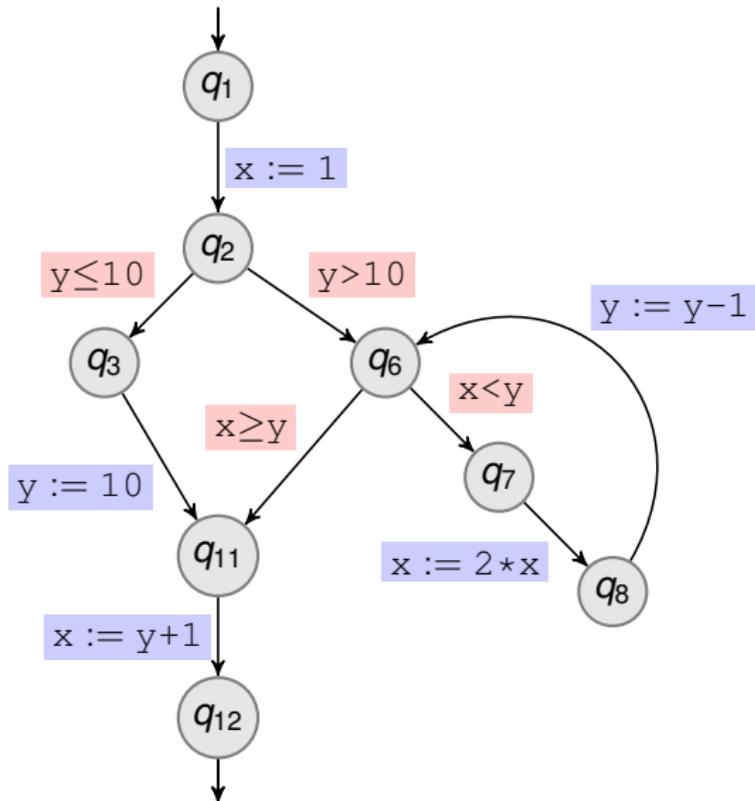


Show that  $x > 0$  at  $q_{12}$

	x		y	
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	$+\infty$	$-\infty$	$+\infty$
$q_7$	1	$+\infty$	$-\infty$	$+\infty$
$q_8$	2	$+\infty$	$-\infty$	$+\infty$
$q_{11}$	1	$+\infty$	$-\infty$	$+\infty$
$q_{12}$	11	11	10	10

$$\begin{aligned}
 (1, 1) \quad \nabla \quad (1, +\infty) &= (1, +\infty) \\
 (10, 10) \quad \nabla \quad (-\infty, +\infty) &= (-\infty, +\infty)
 \end{aligned}$$

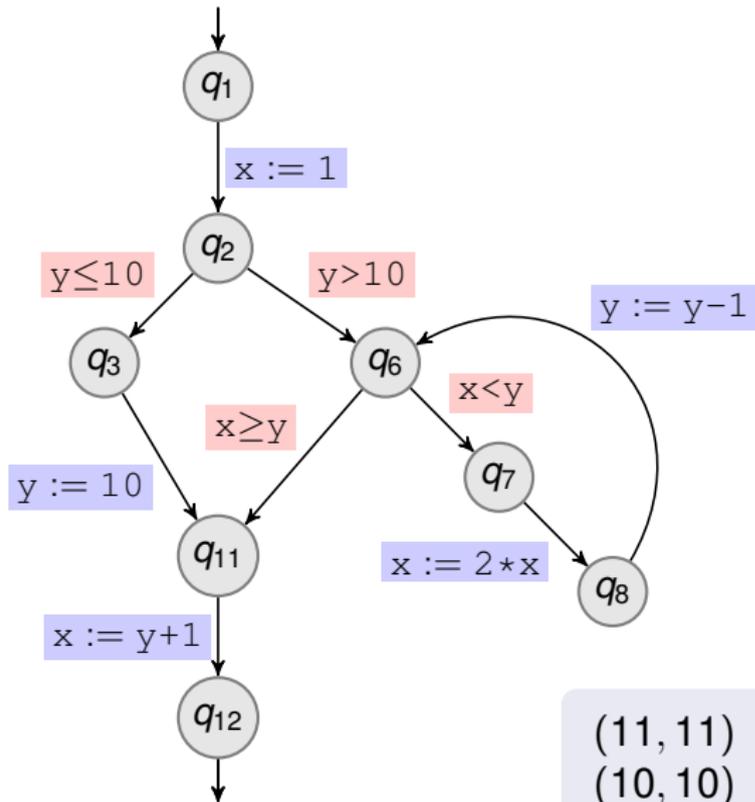
# Range Analysis on Example with Widening



Show that  $x > 0$  at  $q_{12}$

	x		y	
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	$+\infty$	$-\infty$	$+\infty$
$q_7$	1	$+\infty$	$-\infty$	$+\infty$
$q_8$	2	$+\infty$	$-\infty$	$+\infty$
$q_{11}$	1	$+\infty$	$-\infty$	$+\infty$
$q_{12}$	11	11	10	10

# Range Analysis on Example with Widening



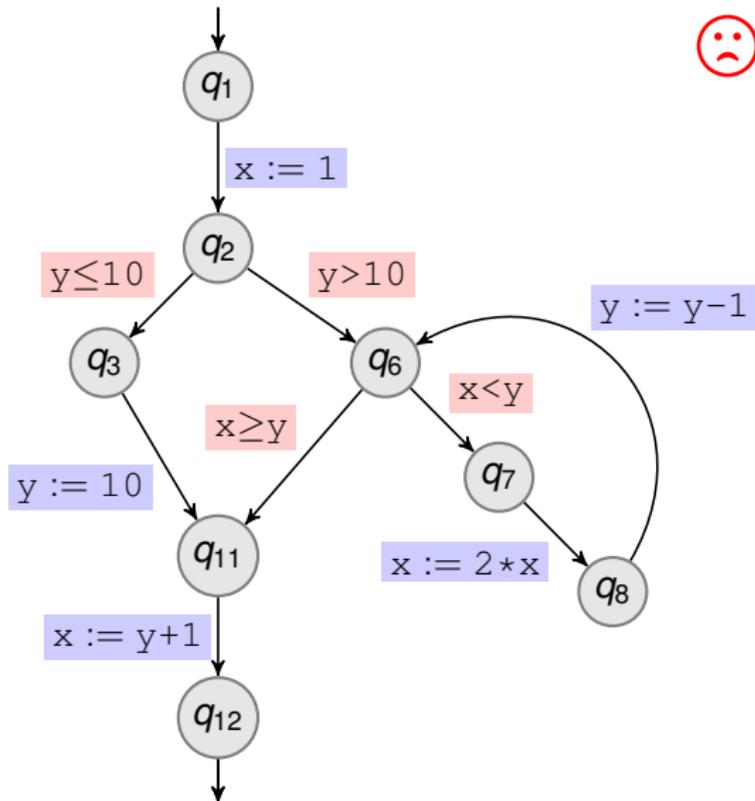
Show that  $x > 0$  at  $q_{12}$

	x		y	
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	$+\infty$	$-\infty$	$+\infty$
$q_7$	1	$+\infty$	$-\infty$	$+\infty$
$q_8$	2	$+\infty$	$-\infty$	$+\infty$
$q_{11}$	1	$+\infty$	$-\infty$	$+\infty$
$q_{12}$	$-\infty$	$+\infty$	$-\infty$	$+\infty$

$$(11, 11) \nabla (-\infty, +\infty) = (-\infty, +\infty)$$

$$(10, 10) \nabla (-\infty, +\infty) = (-\infty, +\infty)$$

# Range Analysis on Example with Widening

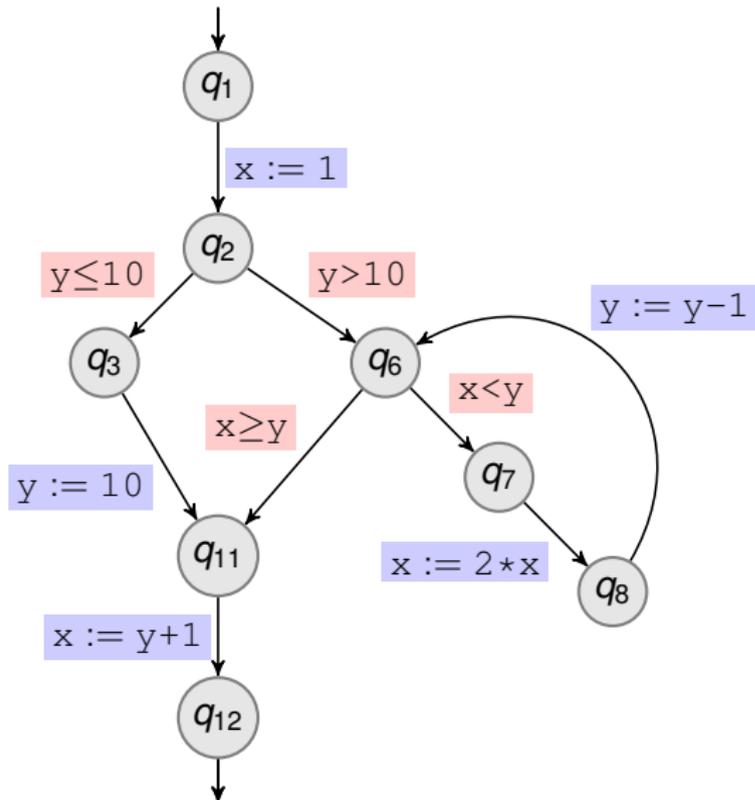


Show that  $x > 0$  at  $q_{12}$

	x		y	
	min	max	min	max
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	$+\infty$	$-\infty$	$+\infty$
$q_7$	1	$+\infty$	$-\infty$	$+\infty$
$q_8$	2	$+\infty$	$-\infty$	$+\infty$
$q_{11}$	1	$+\infty$	$-\infty$	$+\infty$
$q_{12}$	$-\infty$	$+\infty$	$-\infty$	$+\infty$

Too coarse!

# Range Analysis on Example with Delayed Widening

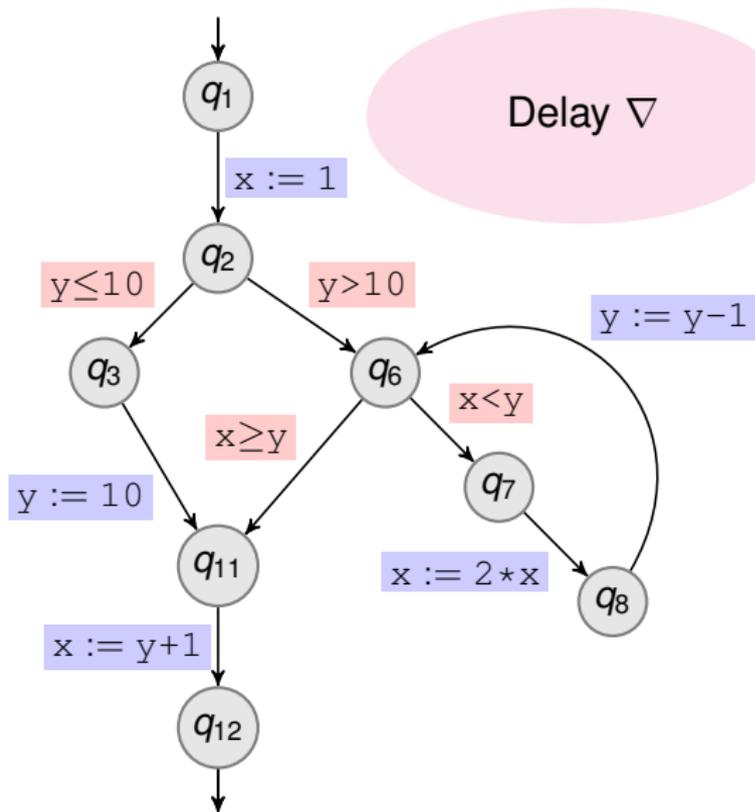


Show that  $x > 0$  at  $q_{12}$

## Delayed Widening

- 1 Keep  $\sqcup$  for the first iterations
- 2 Track number of “updates” for each location
- 3 Switch to  $\nabla$  after a **suitable** “delay”

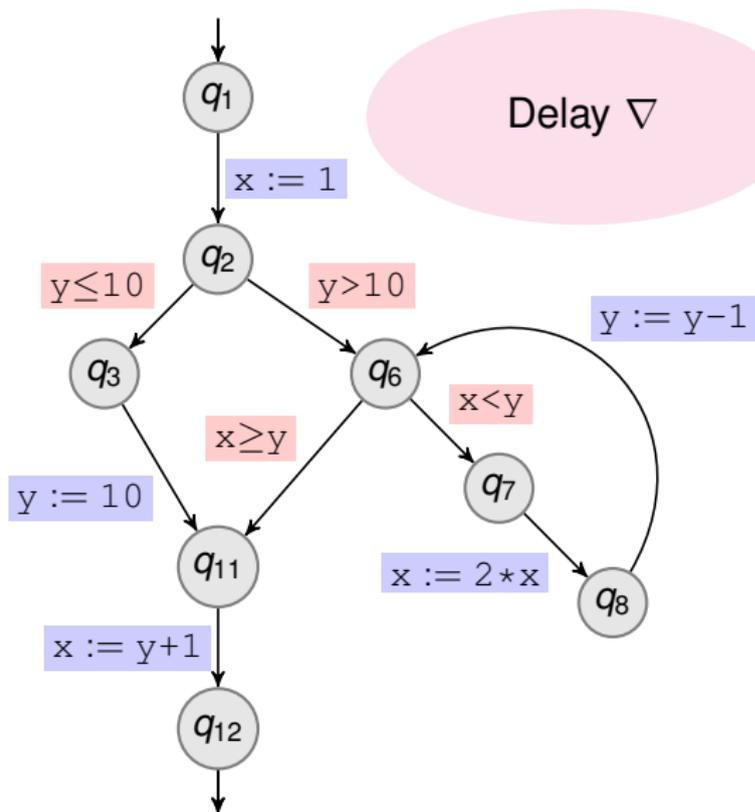
# Range Analysis on Example with Delayed Widening



Show that  $x > 0$  at  $q_{12}$

	x		y	
	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	$\perp$	$\perp$	$\perp$	$\perp$
$q_3$	$\perp$	$\perp$	$\perp$	$\perp$
$q_6$	$\perp$	$\perp$	$\perp$	$\perp$
$q_7$	$\perp$	$\perp$	$\perp$	$\perp$
$q_8$	$\perp$	$\perp$	$\perp$	$\perp$
$q_{11}$	$\perp$	$\perp$	$\perp$	$\perp$
$q_{12}$	$\perp$	$\perp$	$\perp$	$\perp$

# Range Analysis on Example with Delayed Widening

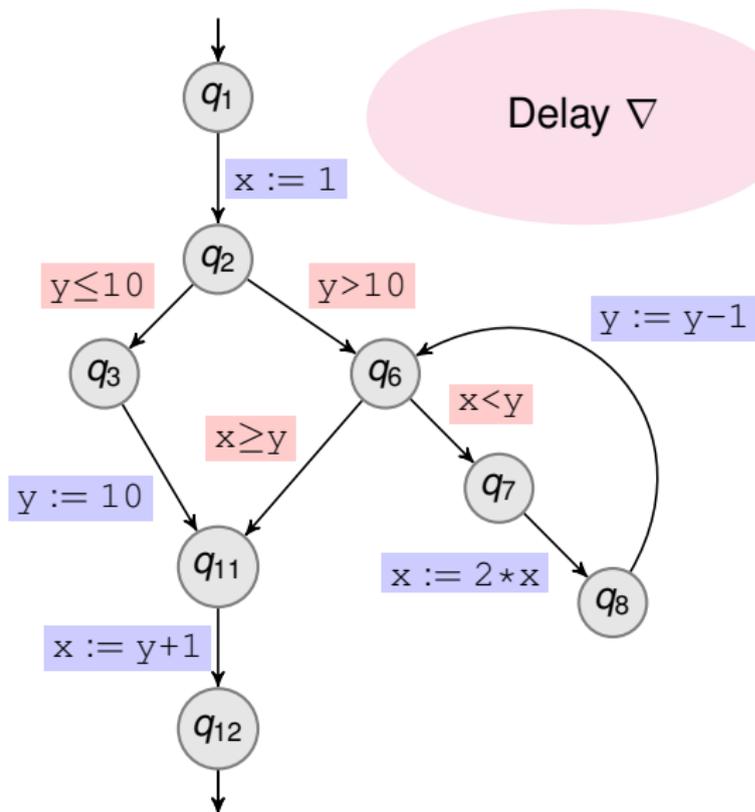


Show that  $x > 0$  at  $q_{12}$

	x		y	
	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	1	10	$+\infty$
$q_7$	1	1	10	$+\infty$
$q_8$	2	2	10	$+\infty$
$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10

Same as without  $\nabla$

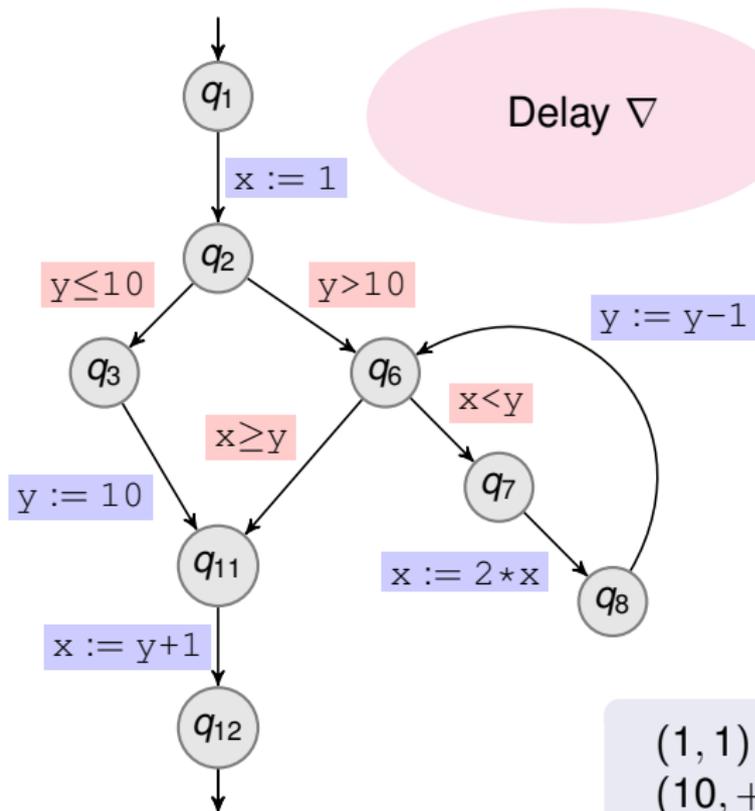
# Range Analysis on Example with Delayed Widening



Show that  $x > 0$  at  $q_{12}$

	x		y	
	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	1	10	$+\infty$
$q_7$	1	1	10	$+\infty$
$q_8$	2	2	10	$+\infty$
$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10

# Range Analysis on Example with Delayed Widening

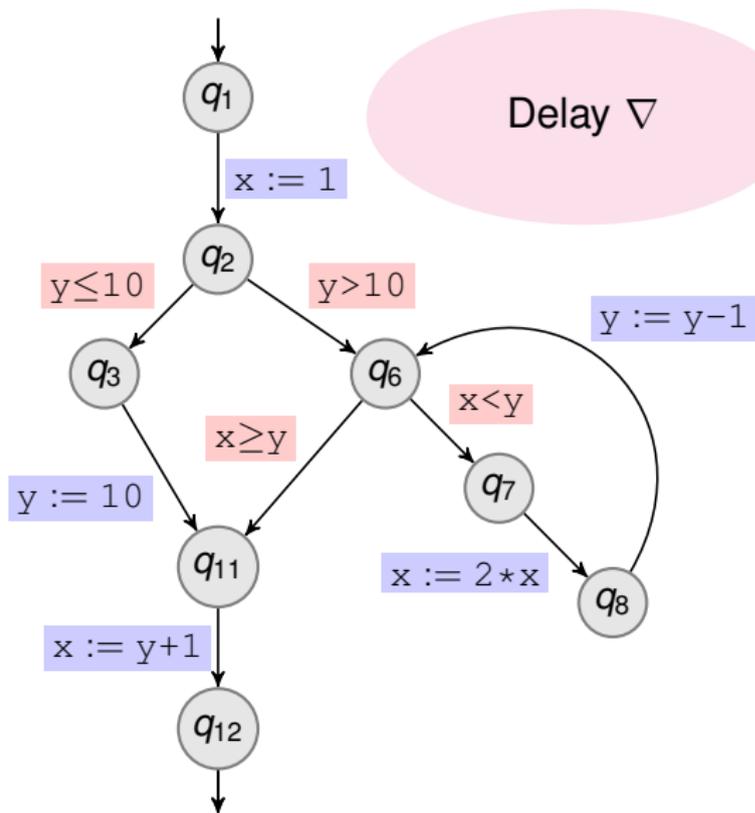


Show that  $x > 0$  at  $q_{12}$

	x		y	
	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	2	9	$+\infty$
$q_7$	1	1	10	$+\infty$
$q_8$	2	2	10	$+\infty$
$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10

$$\begin{aligned}
 (1, 1) &\sqcup (2, 2) = (1, 2) \\
 (10, +\infty) &\sqcup (9, +\infty) = (9, +\infty)
 \end{aligned}$$

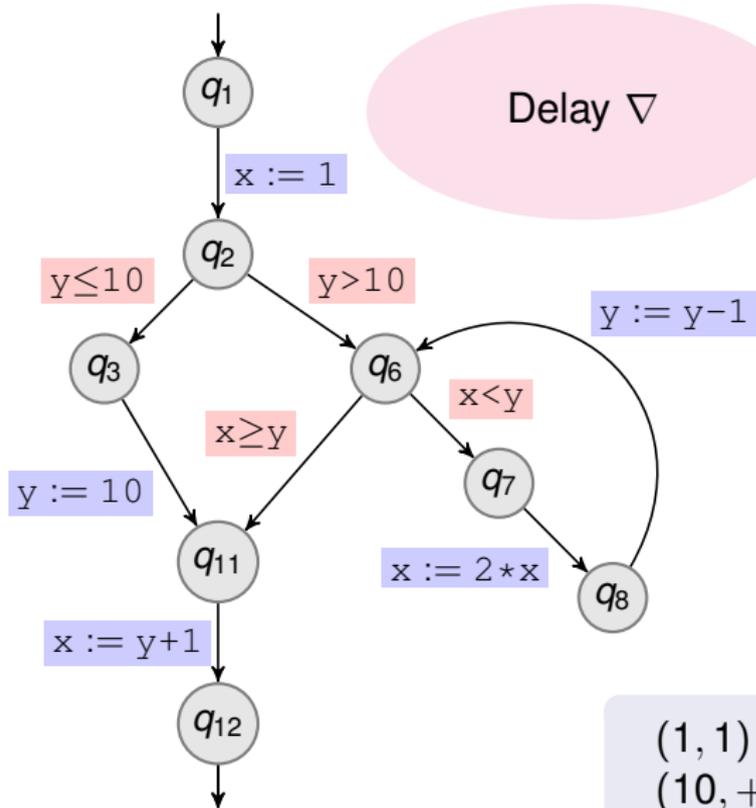
# Range Analysis on Example with Delayed Widening



Show that  $x > 0$  at  $q_{12}$

	x		y	
	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	2	9	$+\infty$
$q_7$	1	1	10	$+\infty$
$q_8$	2	2	10	$+\infty$
$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10

# Range Analysis on Example with Delayed Widening

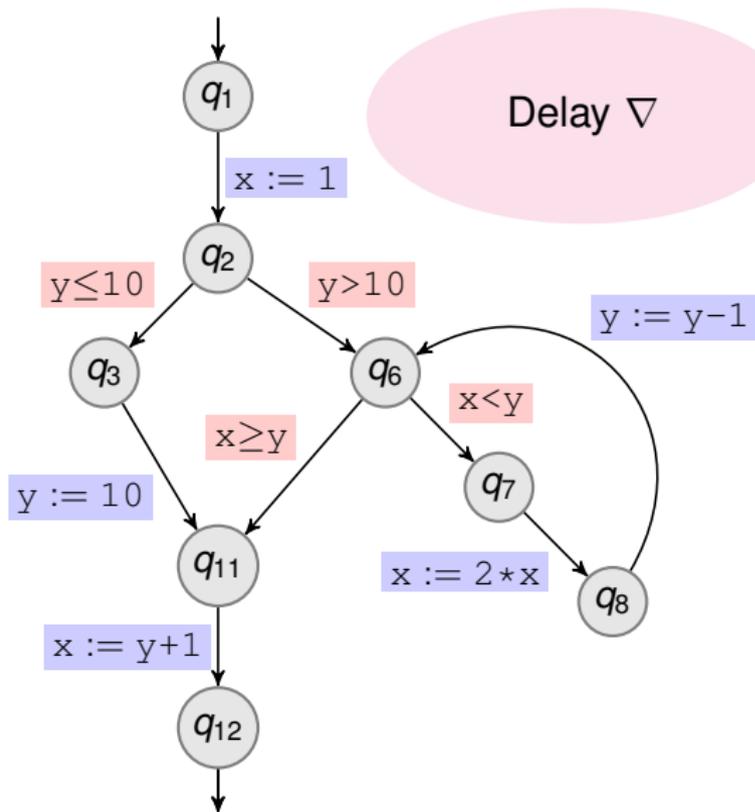


Show that  $x > 0$  at  $q_{12}$

	x		y	
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	2	9	$+\infty$
$q_7$	1	2	9	$+\infty$
$q_8$	2	2	10	$+\infty$
$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10

$$\begin{aligned}
 (1, 1) &\sqcup (1, 2) = (1, 2) \\
 (10, +\infty) &\sqcup (9, +\infty) = (9, +\infty)
 \end{aligned}$$

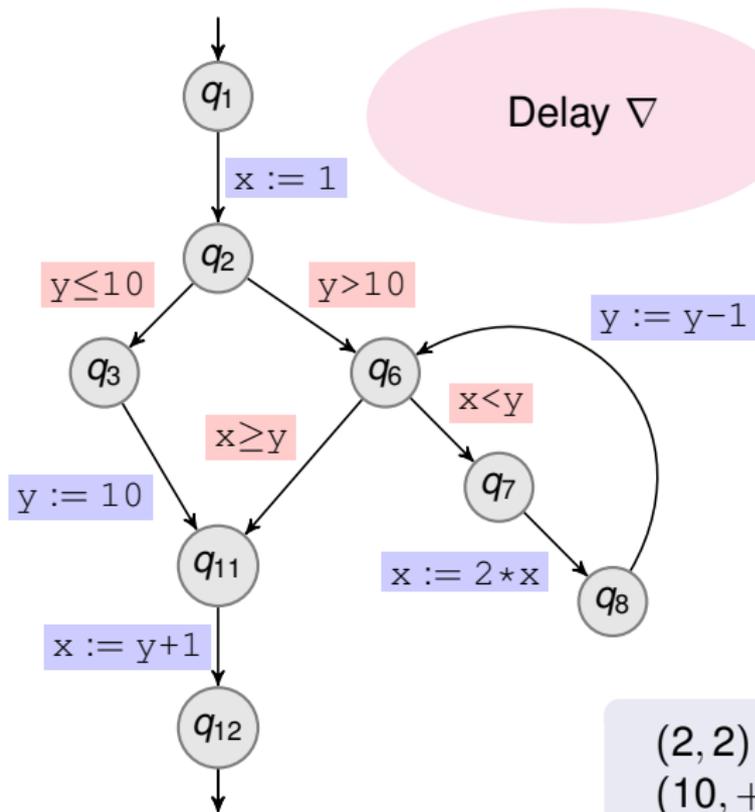
# Range Analysis on Example with Delayed Widening



Show that  $x > 0$  at  $q_{12}$

	x		y	
	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	2	9	$+\infty$
$q_7$	1	2	9	$+\infty$
$q_8$	2	2	10	$+\infty$
$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10

# Range Analysis on Example with Delayed Widening

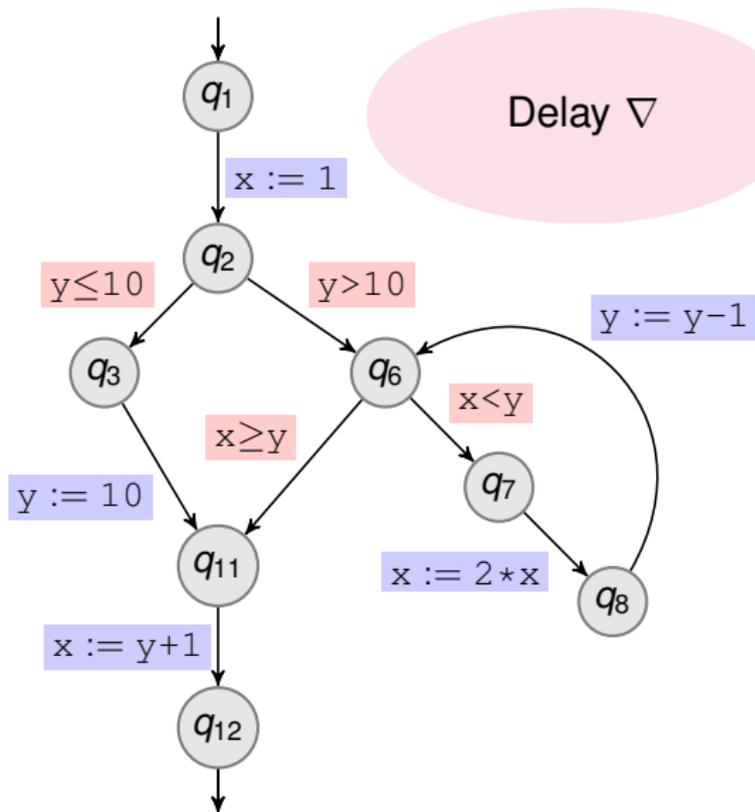


Show that  $x > 0$  at  $q_{12}$

	x		y	
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	2	9	$+\infty$
$q_7$	1	2	9	$+\infty$
$q_8$	2	4	9	$+\infty$
$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10

$$\begin{aligned}
 (2, 2) &\sqcup (2, 4) = (2, 4) \\
 (10, +\infty) &\sqcup (9, +\infty) = (9, +\infty)
 \end{aligned}$$

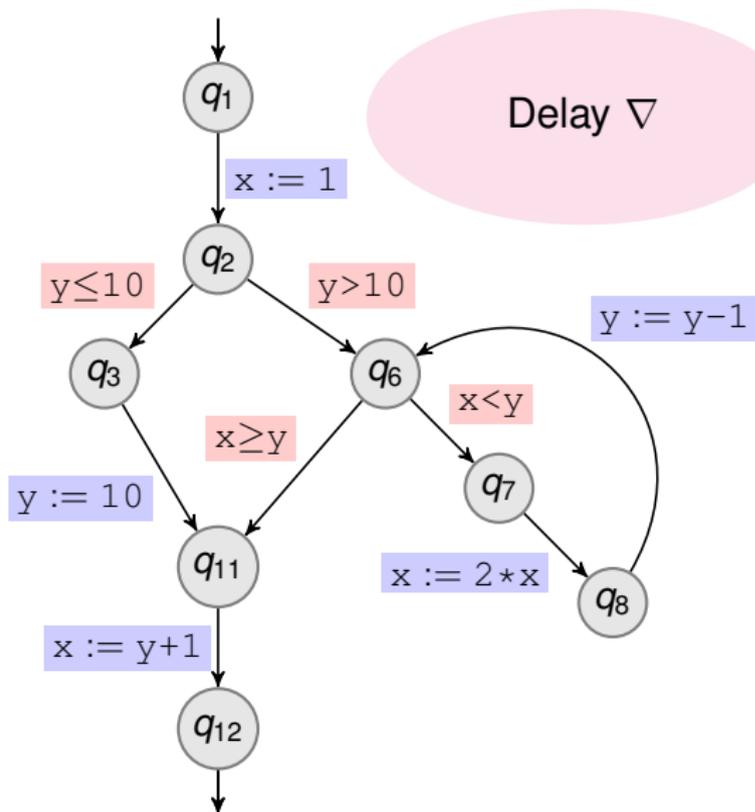
# Range Analysis on Example with Delayed Widening



Show that  $x > 0$  at  $q_{12}$

	x		y	
	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	2	9	$+\infty$
$q_7$	1	2	9	$+\infty$
$q_8$	2	4	9	$+\infty$
$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10

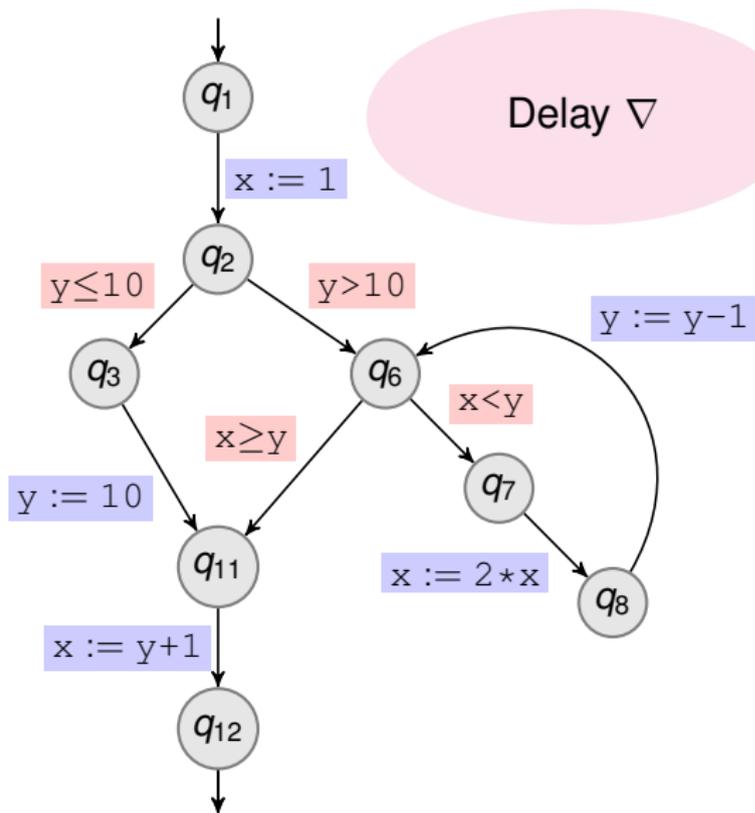
# Range Analysis on Example with Delayed Widening



Show that  $x > 0$  at  $q_{12}$

	x		y	
	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	4	8	$+\infty$
$q_7$	1	4	8	$+\infty$
$q_8$	2	8	8	$+\infty$
$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10

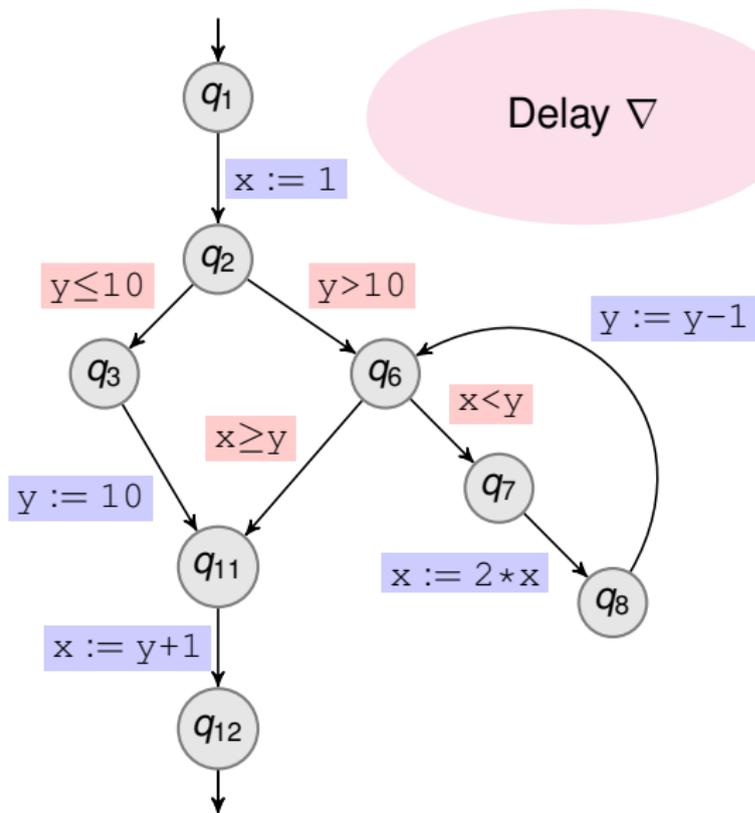
# Range Analysis on Example with Delayed Widening



Show that  $x > 0$  at  $q_{12}$

	x		y	
	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	8	7	$+\infty$
$q_7$	1	8	7	$+\infty$
$q_8$	2	16	7	$+\infty$
$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10

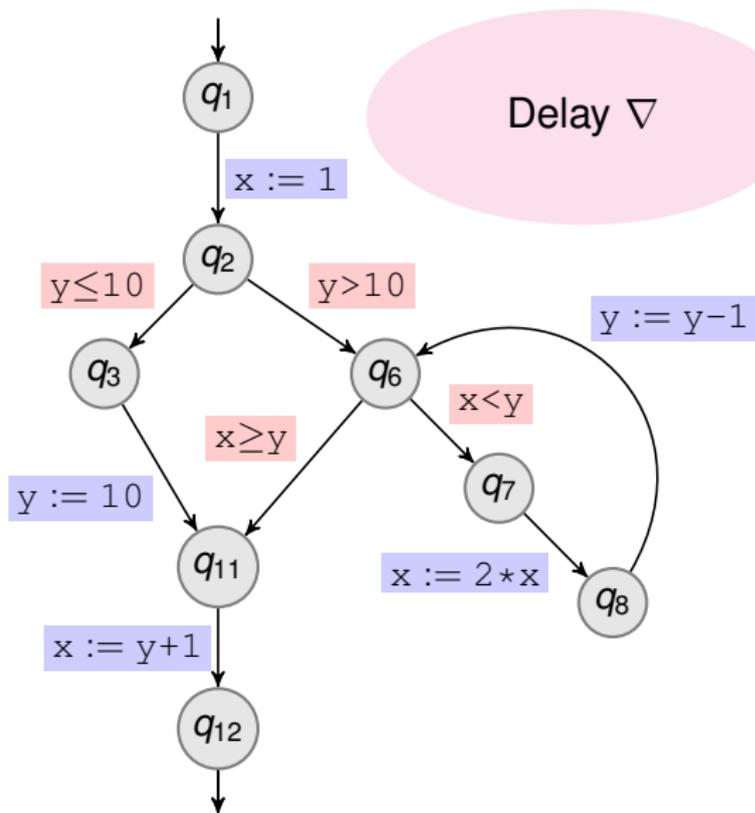
# Range Analysis on Example with Delayed Widening



Show that  $x > 0$  at  $q_{12}$

	x		y	
	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	16	6	$+\infty$
$q_7$	1	16	6	$+\infty$
$q_8$	2	32	6	$+\infty$
$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10

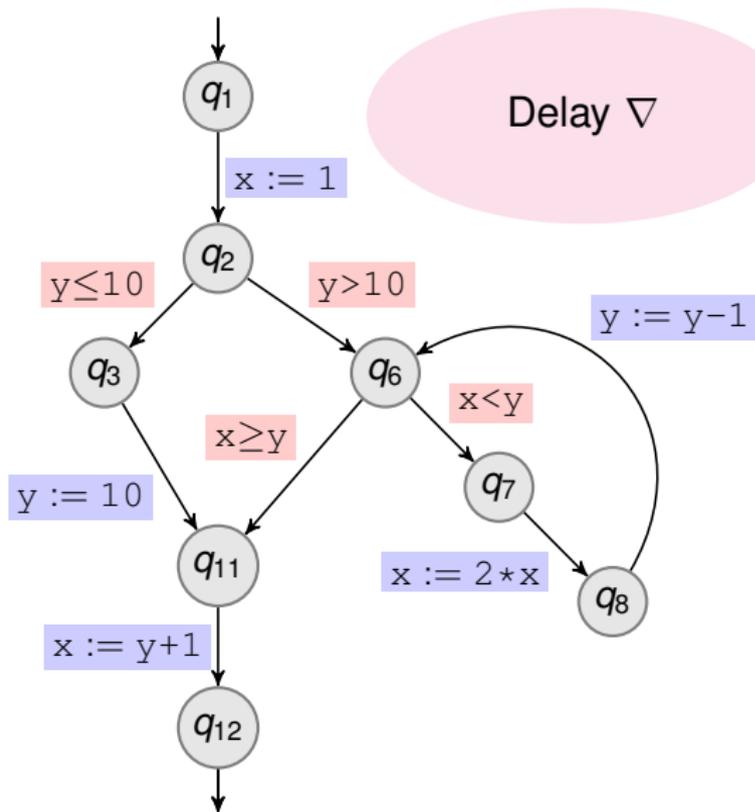
# Range Analysis on Example with Delayed Widening



Show that  $x > 0$  at  $q_{12}$

	x		y	
	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	32	5	$+\infty$
$q_7$	1	32	5	$+\infty$
$q_8$	2	$2^6$	5	$+\infty$
$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10

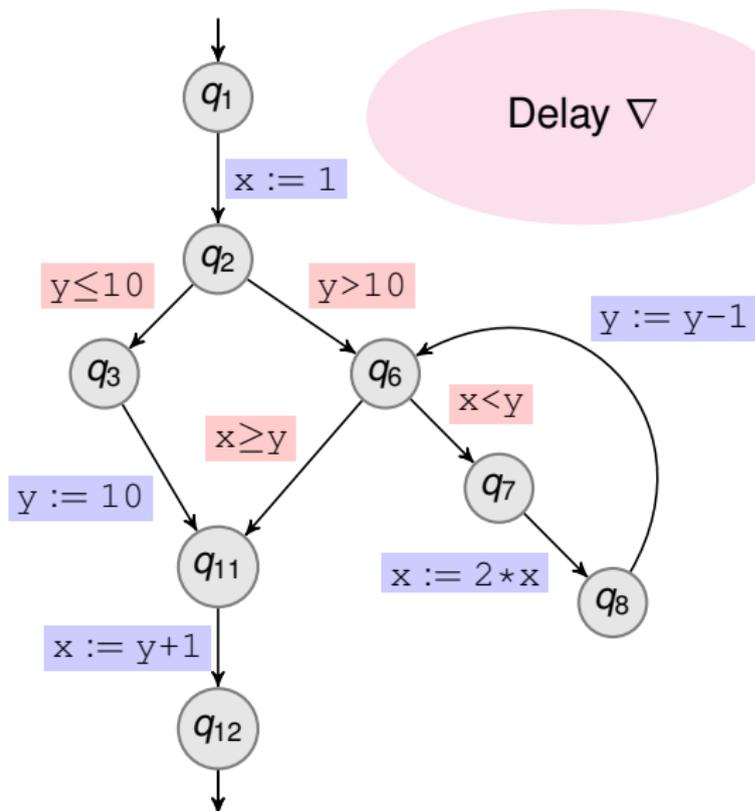
# Range Analysis on Example with Delayed Widening



Show that  $x > 0$  at  $q_{12}$

	x		y	
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	$2^6$	4	$+\infty$
$q_7$	1	$2^6$	4	$+\infty$
$q_8$	2	$2^7$	4	$+\infty$
$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10

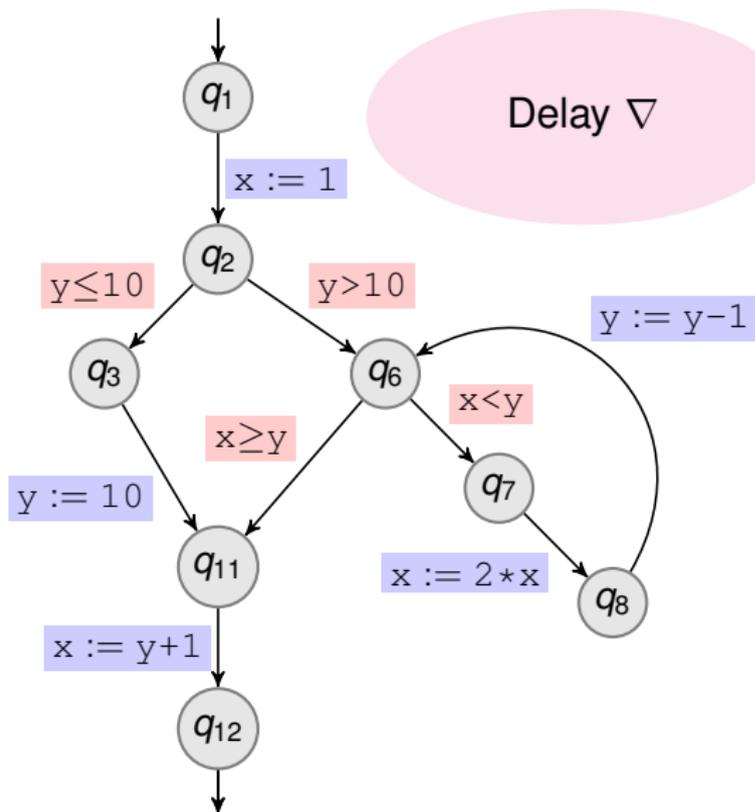
# Range Analysis on Example with Delayed Widening



Show that  $x > 0$  at  $q_{12}$

	x		y	
	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	$2^7$	3	$+\infty$
$q_7$	1	$2^7$	3	$+\infty$
$q_8$	2	$2^8$	3	$+\infty$
$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10

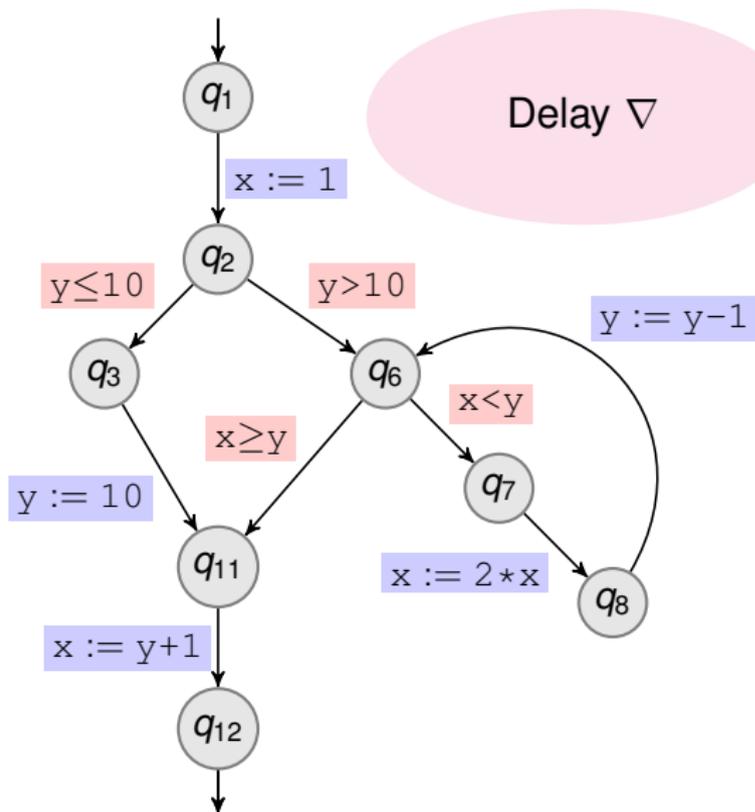
# Range Analysis on Example with Delayed Widening



Show that  $x > 0$  at  $q_{12}$

	x		y	
	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	$2^8$	2	$+\infty$
$q_7$	1	$2^8$	2	$+\infty$
$q_8$	2	$2^9$	2	$+\infty$
$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10

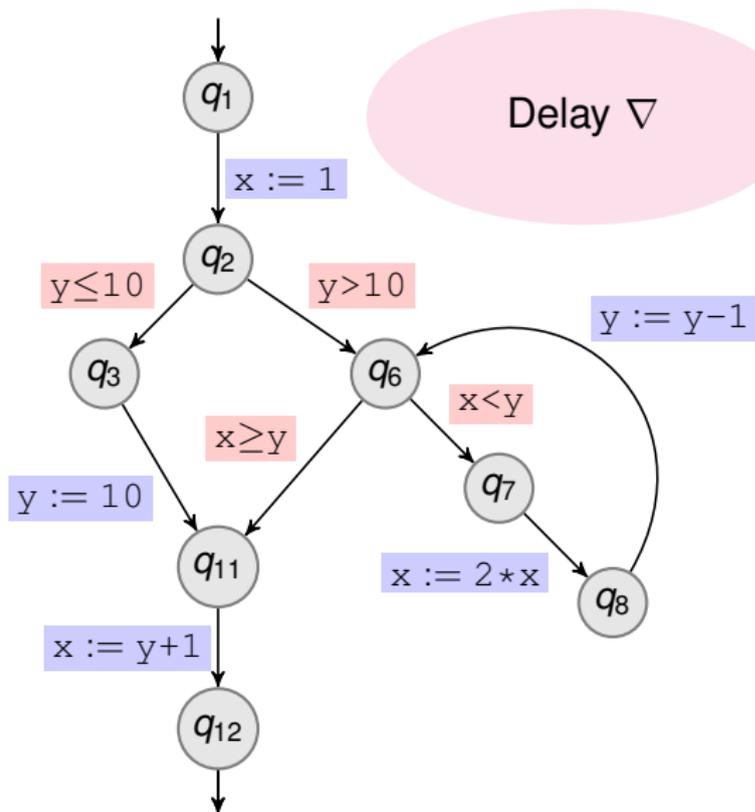
# Range Analysis on Example with Delayed Widening



Show that  $x > 0$  at  $q_{12}$

	x		y	
	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	$2^9$	1	$+\infty$
$q_7$	1	$2^9$	1	$+\infty$
$q_8$	2	$2^{10}$	1	$+\infty$
$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10

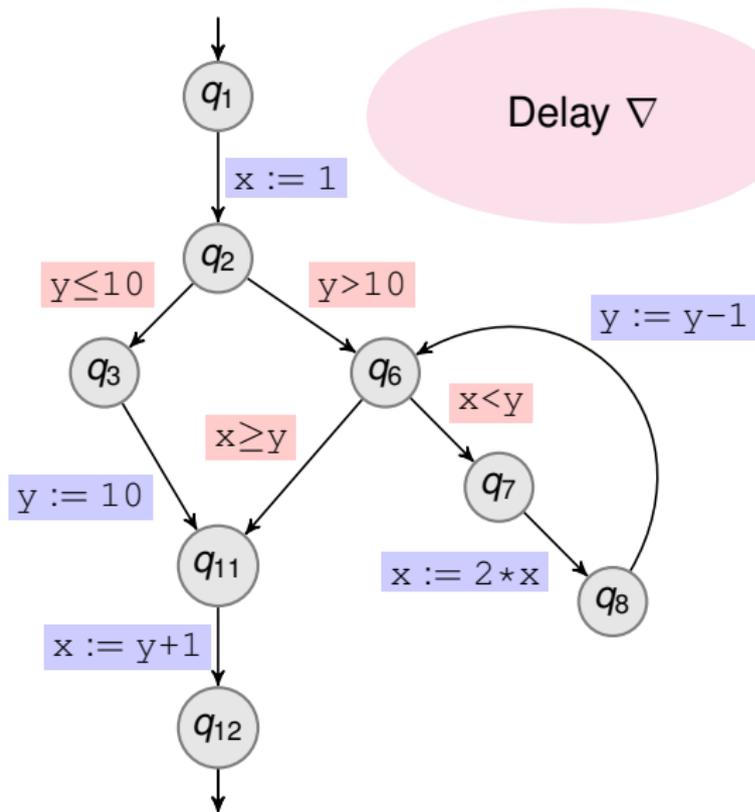
# Range Analysis on Example with Delayed Widening



Show that  $x > 0$  at  $q_{12}$

	x		y	
	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	$2^9$	1	$+\infty$
$q_7$	1	$2^9$	1	$+\infty$
$q_8$	2	$2^{10}$	1	$+\infty$
$q_{11}$	1	1	10	10
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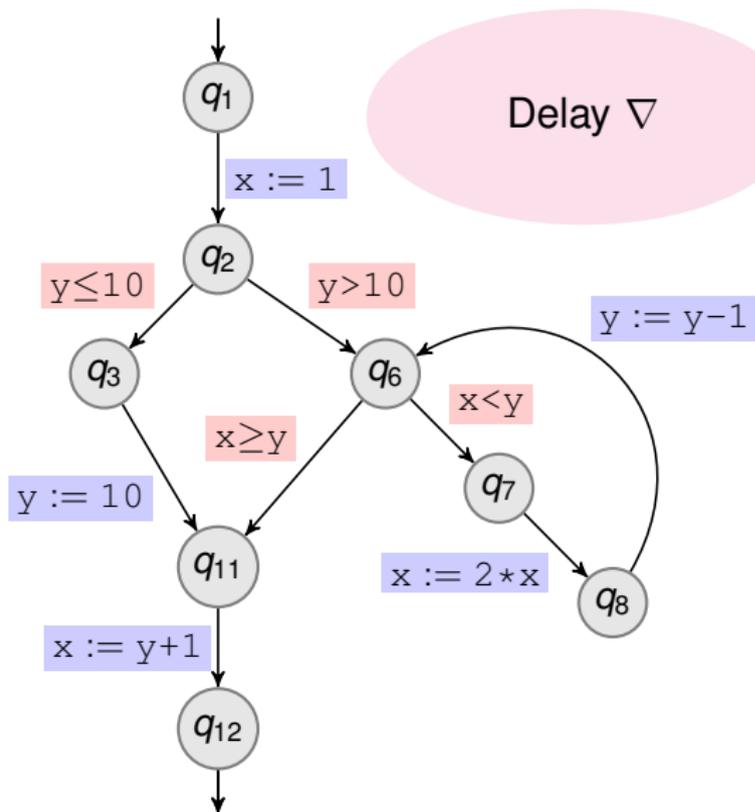
# Range Analysis on Example with Delayed Widening



Show that  $x > 0$  at  $q_{12}$

	x		y	
	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	$2^{10}$	0	$+\infty$
$q_7$	1	$2^9$	1	$+\infty$
$q_8$	2	$2^{10}$	1	$+\infty$
$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10

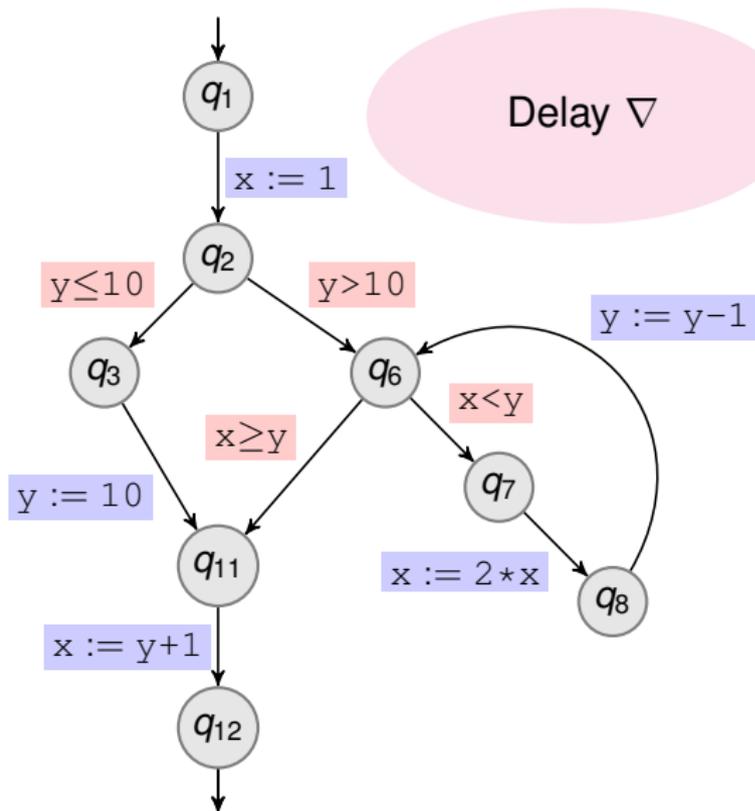
# Range Analysis on Example with Delayed Widening



Show that  $x > 0$  at  $q_{12}$

	x		y	
	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	$2^{10}$	0	$+\infty$
$q_7$	1	$2^9$	1	$+\infty$
$q_8$	2	$2^{10}$	1	$+\infty$
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$q_{12}$	11	11	10	10

# Range Analysis on Example with Delayed Widening

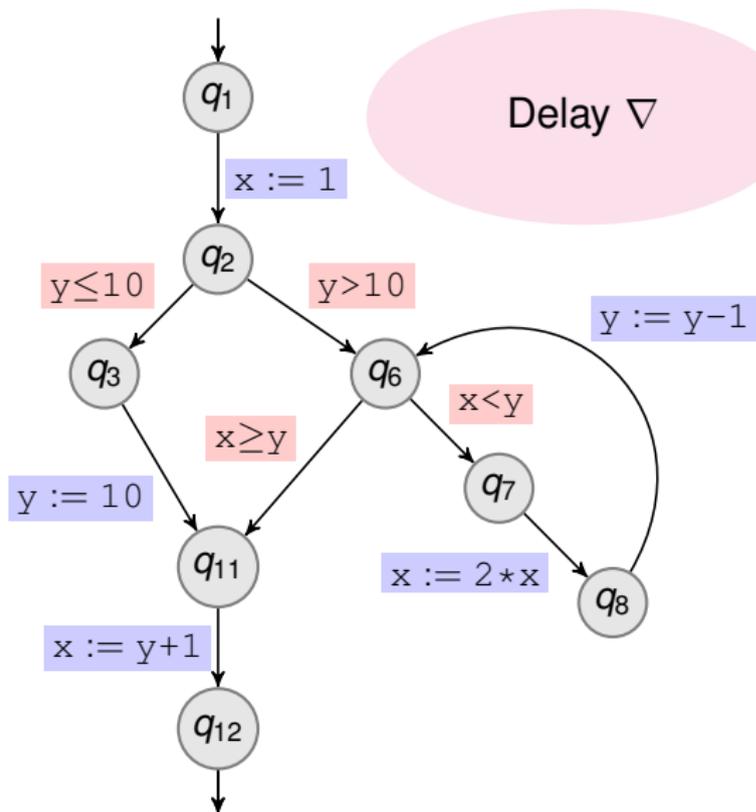


Show that  $x > 0$  at  $q_{12}$

	x		y	
	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	$2^{10}$	0	$+\infty$
$q_7$	1	$2^{10}$	1	$+\infty$
$q_8$	2	$2^{10}$	1	$+\infty$
$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10

Gained from guard  $x < y$

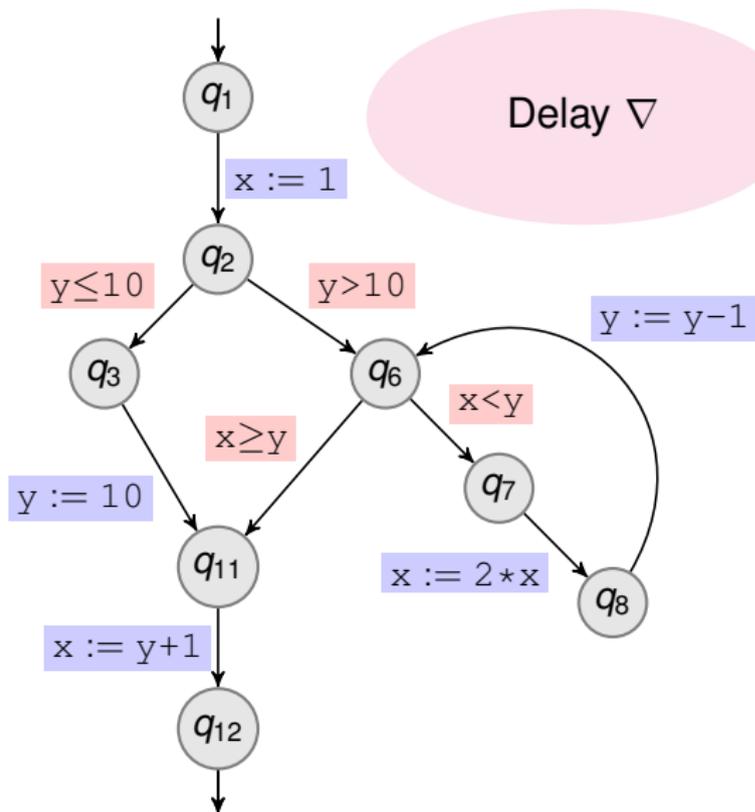
# Range Analysis on Example with Delayed Widening



Show that  $x > 0$  at  $q_{12}$

	x		y	
	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	$2^{10}$	0	$+\infty$
$q_7$	1	$2^{10}$	1	$+\infty$
$q_8$	2	$2^{10}$	1	$+\infty$
$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10

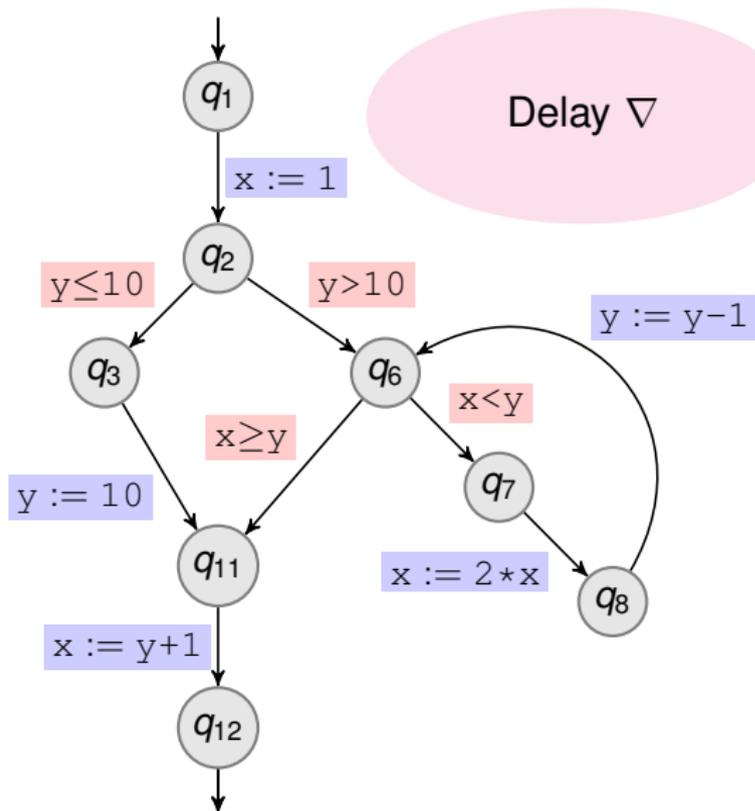
# Range Analysis on Example with Delayed Widening



Show that  $x > 0$  at  $q_{12}$

	x		y	
	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	$2^{10}$	0	$+\infty$
$q_7$	1	$2^{10}$	1	$+\infty$
$q_8$	2	$2^{11}$	1	$+\infty$
$q_{11}$	1	1	10	10
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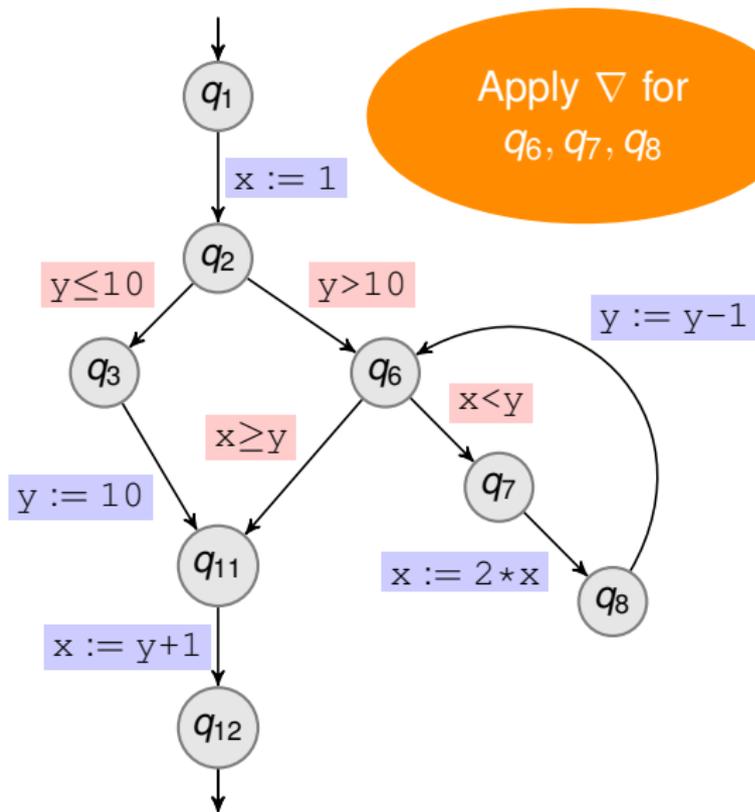
# Range Analysis on Example with Delayed Widening



Show that  $x > 0$  at  $q_{12}$

	x		y	
	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	$2^{10}$	0	$+\infty$
$q_7$	1	$2^{10}$	1	$+\infty$
$q_8$	2	$2^{11}$	1	$+\infty$
$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10

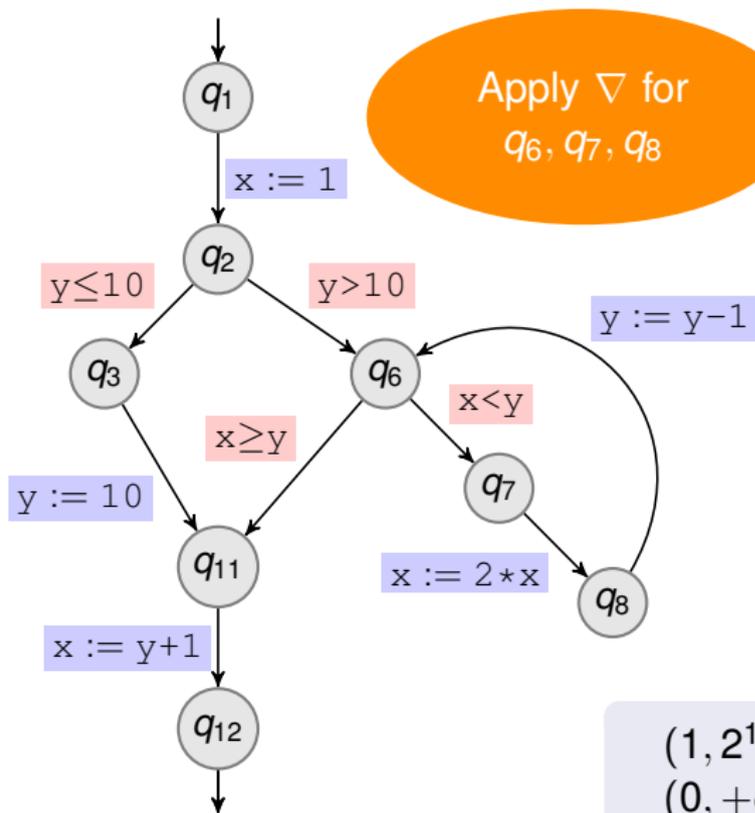
# Range Analysis on Example with Delayed Widening



Show that  $x > 0$  at  $q_{12}$

	x		y	
	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	$2^{10}$	0	$+\infty$
$q_7$	1	$2^{10}$	1	$+\infty$
$q_8$	2	$2^{11}$	1	$+\infty$
$q_{11}$	1	1	10	10
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# Range Analysis on Example with Delayed Widening



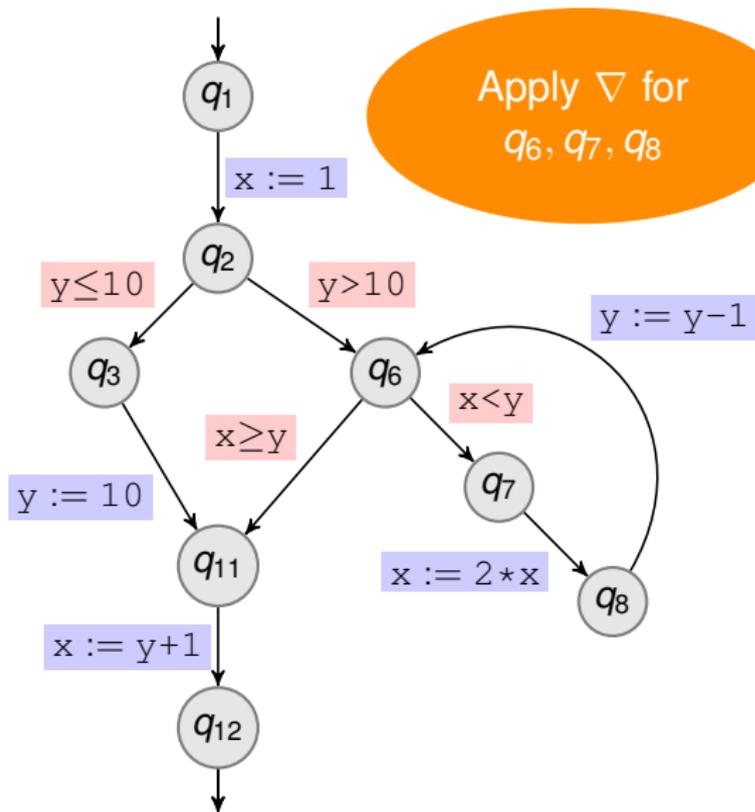
Show that  $x > 0$  at  $q_{12}$

	x		y	
	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	$+\infty$	0	$+\infty$
$q_7$	1	$2^{10}$	1	$+\infty$
$q_8$	2	$2^{11}$	1	$+\infty$
$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10

$$(1, 2^{10}) \nabla (2, 2^{11}) = (1, +\infty)$$

$$(0, +\infty) \nabla (0, +\infty) = (0, +\infty)$$

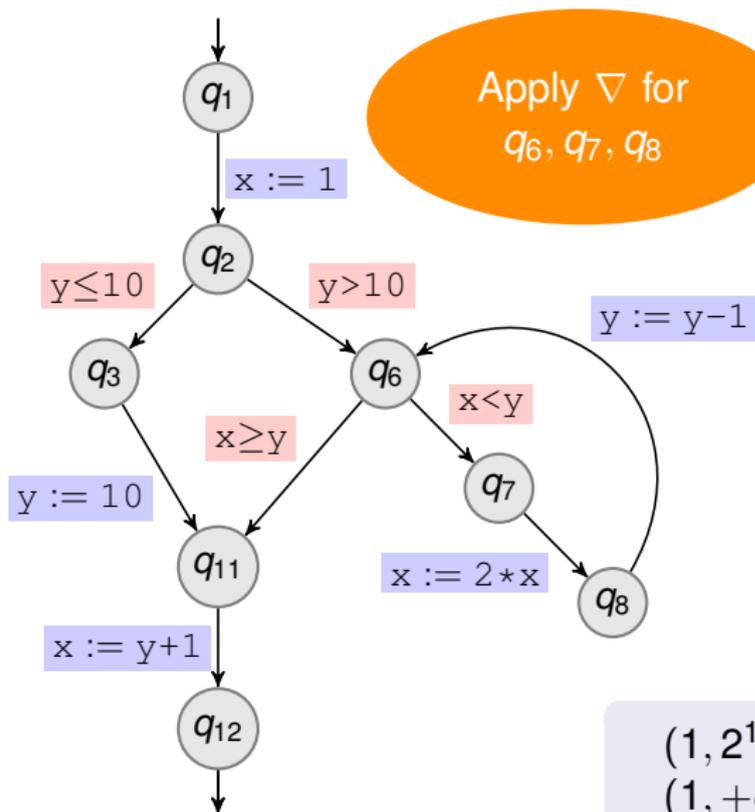
# Range Analysis on Example with Delayed Widening



Show that  $x > 0$  at  $q_{12}$

	x		y	
	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	$+\infty$	0	$+\infty$
$q_7$	1	$2^{10}$	1	$+\infty$
$q_8$	2	$2^{11}$	1	$+\infty$
$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10

# Range Analysis on Example with Delayed Widening



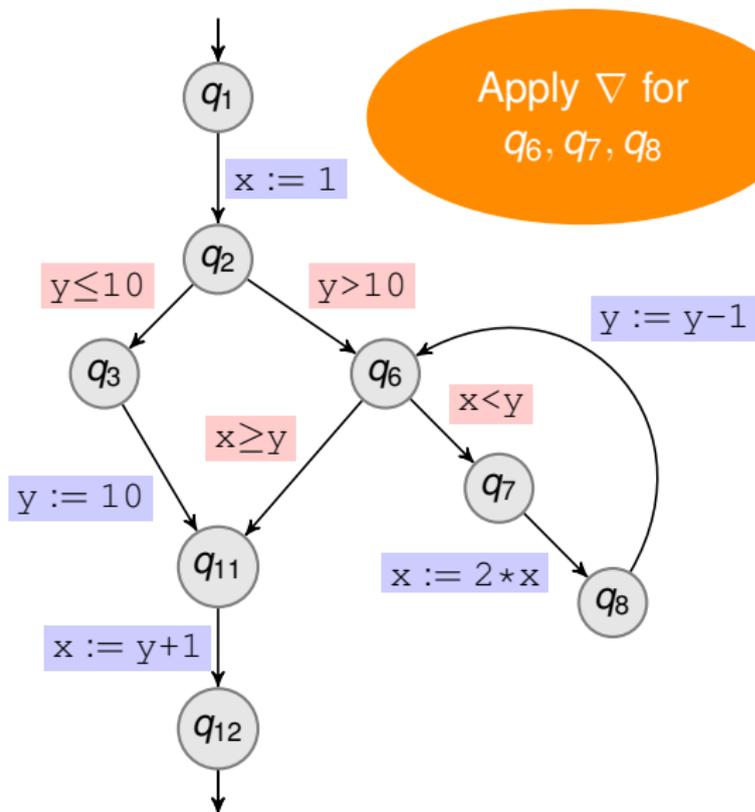
Show that  $x > 0$  at  $q_{12}$

	x		y	
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	$+\infty$	0	$+\infty$
$q_7$	1	$+\infty$	1	$+\infty$
$q_8$	2	$2^{11}$	1	$+\infty$
$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10

$$(1, 2^{10}) \nabla (1, +\infty) = (1, +\infty)$$

$$(1, +\infty) \nabla (1, +\infty) = (1, +\infty)$$

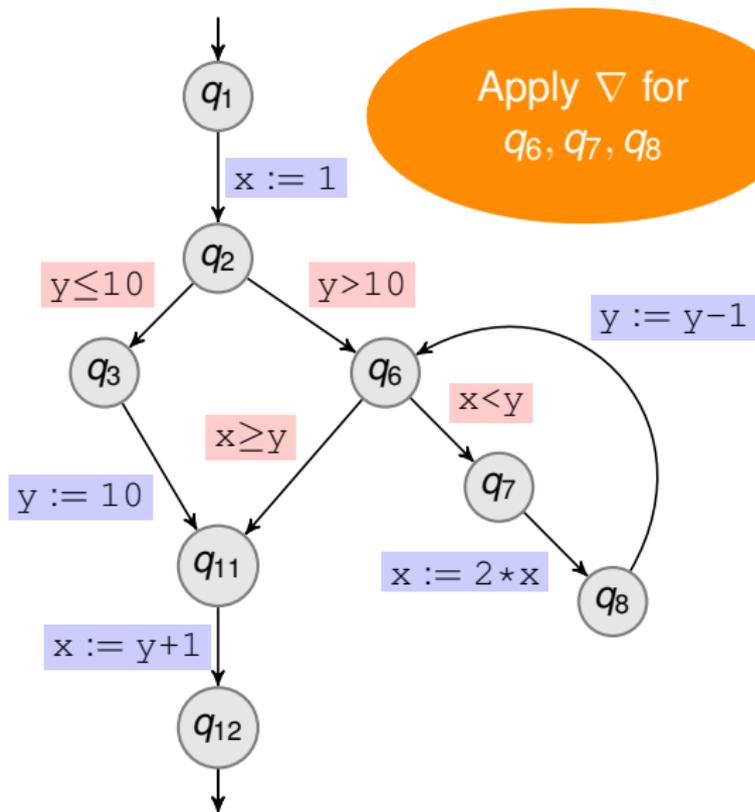
# Range Analysis on Example with Delayed Widening



Show that  $x > 0$  at  $q_{12}$

	x		y	
	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	$+\infty$	0	$+\infty$
$q_7$	1	$+\infty$	1	$+\infty$
$q_8$	2	$2^{11}$	1	$+\infty$
$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10

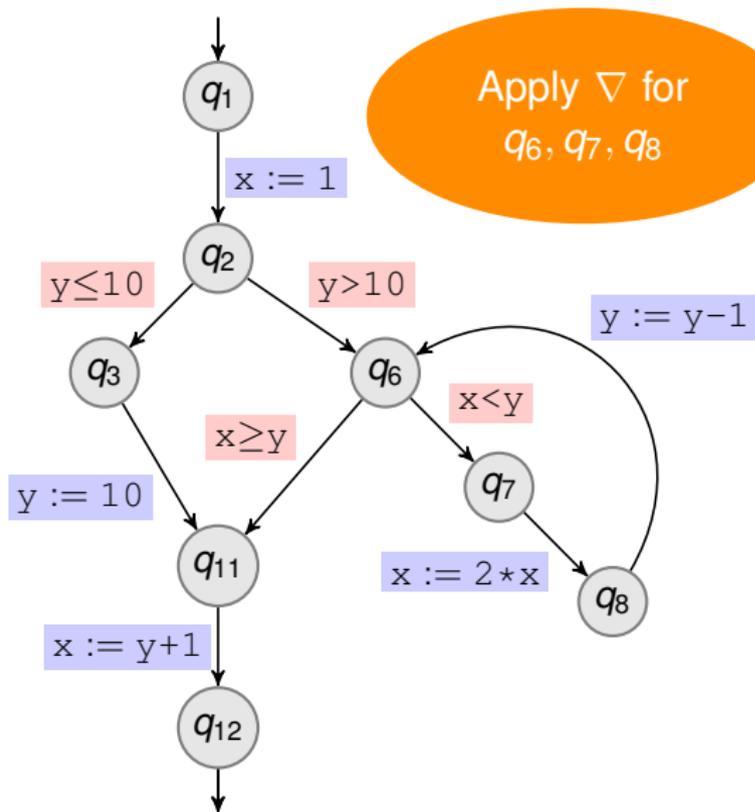
# Range Analysis on Example with Delayed Widening



Show that  $x > 0$  at  $q_{12}$

	x		y	
	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	$+\infty$	0	$+\infty$
$q_7$	1	$+\infty$	1	$+\infty$
$q_8$	2	$+\infty$	1	$+\infty$
$q_{11}$	1	1	10	10
$q_{12}$	11	11	10	10

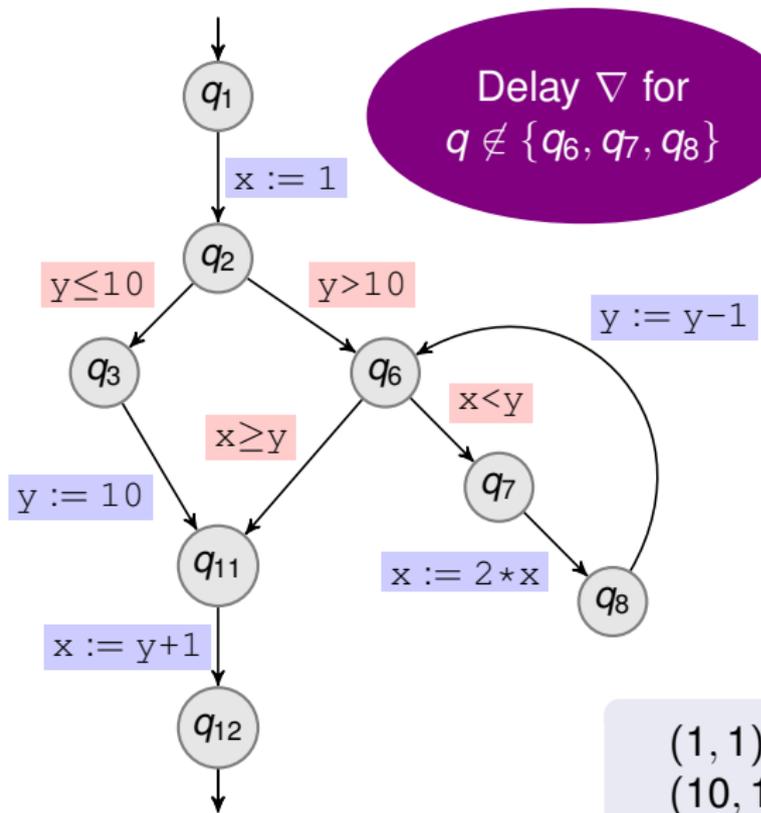
# Range Analysis on Example with Delayed Widening



Show that  $x > 0$  at  $q_{12}$

	x		y	
	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	$+\infty$	0	$+\infty$
$q_7$	1	$+\infty$	1	$+\infty$
$q_8$	2	$+\infty$	1	$+\infty$
$q_{11}$	1	1	10	10
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# Range Analysis on Example with Delayed Widening



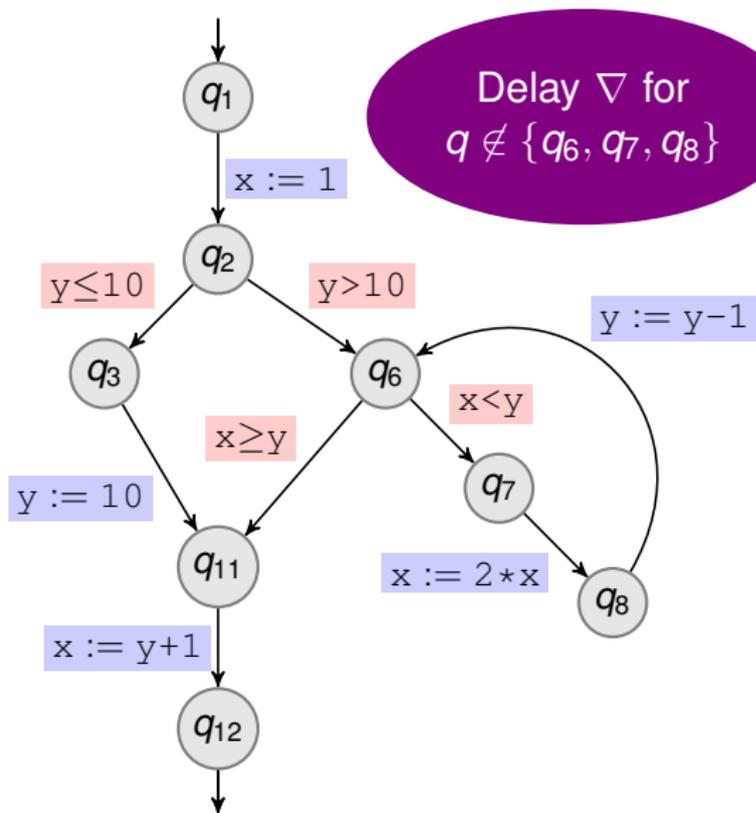
Show that  $x > 0$  at  $q_{12}$

	x		y	
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	$+\infty$	0	$+\infty$
$q_7$	1	$+\infty$	1	$+\infty$
$q_8$	2	$+\infty$	1	$+\infty$
$q_{11}$	1	$+\infty$	0	$+\infty$
$q_{12}$	11	11	10	10

$$(1, 1) \sqcup (1, +\infty) = (1, +\infty)$$

$$(10, 10) \sqcup (0, +\infty) = (0, +\infty)$$

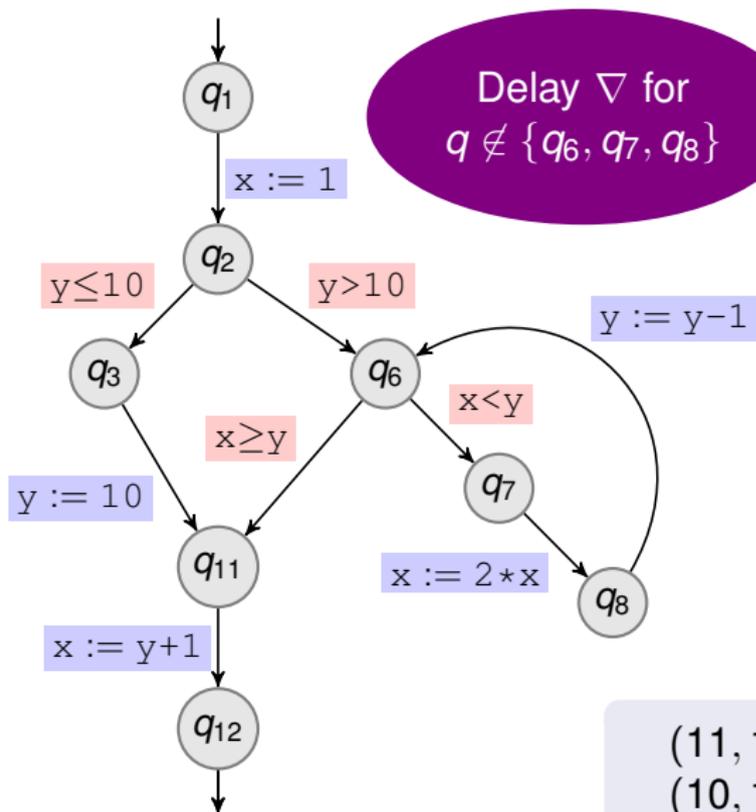
# Range Analysis on Example with Delayed Widening



Show that  $x > 0$  at  $q_{12}$

	x		y	
	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	$+\infty$	0	$+\infty$
$q_7$	1	$+\infty$	1	$+\infty$
$q_8$	2	$+\infty$	1	$+\infty$
$q_{11}$	1	$+\infty$	0	$+\infty$
$q_{12}$	11	11	10	10

# Range Analysis on Example with Delayed Widening



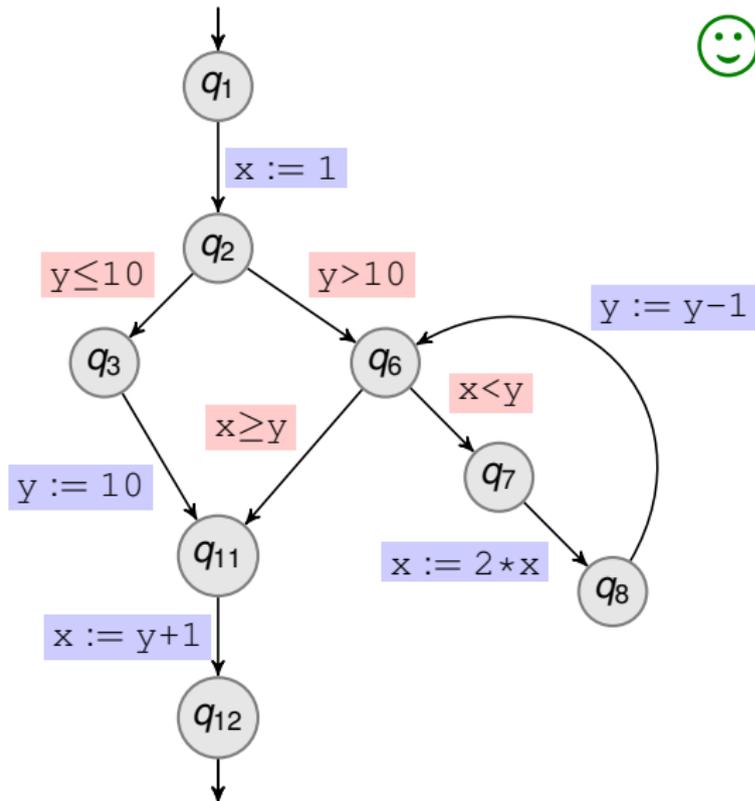
Show that  $x > 0$  at  $q_{12}$

	x		y	
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	$+\infty$	0	$+\infty$
$q_7$	1	$+\infty$	1	$+\infty$
$q_8$	2	$+\infty$	1	$+\infty$
$q_{11}$	1	$+\infty$	0	$+\infty$
$q_{12}$	0	$+\infty$	0	$+\infty$

$$(11, 11) \sqcup (1, +\infty) = (1, +\infty)$$

$$(10, 10) \sqcup (0, +\infty) = (0, +\infty)$$

# Range Analysis on Example with Delayed Widening



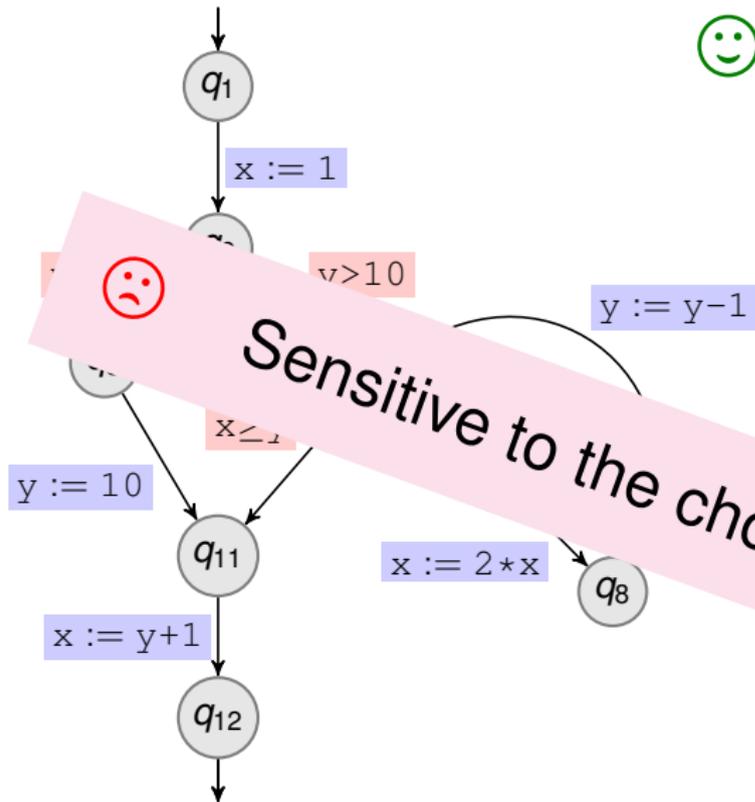
Show that  $x > 0$  at  $q_{12}$

	x		y	
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	$+\infty$	0	$+\infty$
$q_7$	1	$+\infty$	1	$+\infty$
$q_8$	2	$+\infty$	1	$+\infty$
$q_{11}$	1	$+\infty$	0	$+\infty$
$q_{12}$	0	$+\infty$	0	$+\infty$

# Range Analysis on Example with Delayed Widening



Show that  $x > 0$  at  $q_{12}$



	x		y	
	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	$+\infty$	0	$+\infty$
$q_7$	1	$+\infty$	1	$+\infty$
	2	$+\infty$	1	$+\infty$
	$+\infty$	$+\infty$	0	$+\infty$
	$+\infty$	$+\infty$	$+\infty$	$+\infty$

Sensitive to the choice of delay

# Precision Improvement with Narrowing

Consider a complete lattice  $(L, \sqsubseteq)$ .

## Objective of Narrowing Operators

**Soundly** improve the precision of an approximation obtained with  $\nabla$

## Definition

A **narrowing** operator for  $(L, \sqsubseteq)$  is a function  $\Delta : (L \times L) \rightarrow L$  such that:

- 1  $y \sqsubseteq x \implies y \sqsubseteq (x \Delta y) \sqsubseteq x$  (for all  $x, y \in L$ )
- 2 for any descending chain  $x_0 \sqsupseteq x_1 \sqsupseteq \dots$  of elements of  $L$ , the descending chain  $y_0 \sqsupseteq y_1 \sqsupseteq \dots$  defined by

$$\begin{cases} y_0 = x_0 \\ y_{i+1} = y_i \Delta x_{i+1} \quad \text{for all } i \in \mathbb{N} \end{cases}$$

is not strictly decreasing (i.e.  $y_{i+1} = y_i$  for some  $i \in \mathbb{N}$ ).

# Correctness of Decreasing Iteration with Narrowing

Consider a complete lattice  $(L, \sqsubseteq)$  and a monotonic function  $f : L \rightarrow L$ .

A **post-fixpoint** of  $f$  is any element  $a \in L$  satisfying  $a \sqsupseteq f(a)$ .

## Theorem

If  $\Delta : (L \times L) \rightarrow L$  is a narrowing operator then for any post-fixpoint  $a$  of  $f$ , the descending chain  $x_0 \sqsupseteq x_1 \sqsupseteq \dots$  defined by

$$\begin{aligned}x_0 &= a \\x_{i+1} &= x_i \Delta f(x_{i+1})\end{aligned}$$

is eventually stationary, and its limit satisfies  $\bigcap \{x_i \mid i \in \mathbb{N}\} \sqsupseteq \text{lfp}(f)$ .

## Application to Precision Improvement of MFP Approximations

- 1 Compute an approximation of *MFP* by Kleene iteration with  $\nabla$ .
- 2 Then perform a decreasing iteration with  $\Delta$  to regain precision.

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# Narrowing Operator for Range Analysis: Intuition

## Objective of Narrowing Operators

**Soundly** improve the precision of an approximation obtained with  $\nabla$

$\nabla$  may have introduced infinite bounds to accelerate convergence.

Improve infinite bounds when possible (leave the non-infinite ones)

## Examples

$$(1, +\infty) \Delta (1, 4) = (1, 4)$$

$$(1, 10) \Delta (1, 4) = (1, 10)$$

$$(-\infty, 10) \Delta (1, 4) = (1, 10)$$

# Narrowing Operator for Range Analysis

## Narrowing Operator on the Complete Lattice $(Int, \sqsubseteq)$ of Intervals

$$\perp \Delta \perp = \perp$$

$$\perp \Delta (l, u) = (l, u) \Delta \perp = \perp$$

$$(l_1, u_1) \Delta (l_2, u_2) = (l_\Delta, u_\Delta) \quad \text{where} \quad \begin{cases} l_\Delta = \begin{cases} l_2 & \text{if } l_1 = -\infty \\ l_1 & \text{otherwise} \end{cases} \\ u_\Delta = \begin{cases} u_2 & \text{if } u_1 = +\infty \\ u_1 & \text{otherwise} \end{cases} \end{cases}$$

## Narrowing Operator on the Complete Lattice $(x \rightarrow Int, \sqsubseteq)$

Extension  $\bar{\Delta}$  of the narrowing  $\Delta$  on  $(Int, \sqsubseteq)$  to  $(x \rightarrow Int, \sqsubseteq)$ , defined by:

$$\bar{v}_1 \bar{\Delta} \bar{v}_2 = \lambda q. \bar{v}_1(q) \Delta \bar{v}_2(q)$$

# Narrowing Operator for Range Analysis

## Narrowing Operator on the Complete Lattice $(Int, \sqsubseteq)$ of Intervals

$$\perp \Delta \perp = \perp \quad \perp \Delta (l, u) = (l, u) \Delta \perp = \perp$$

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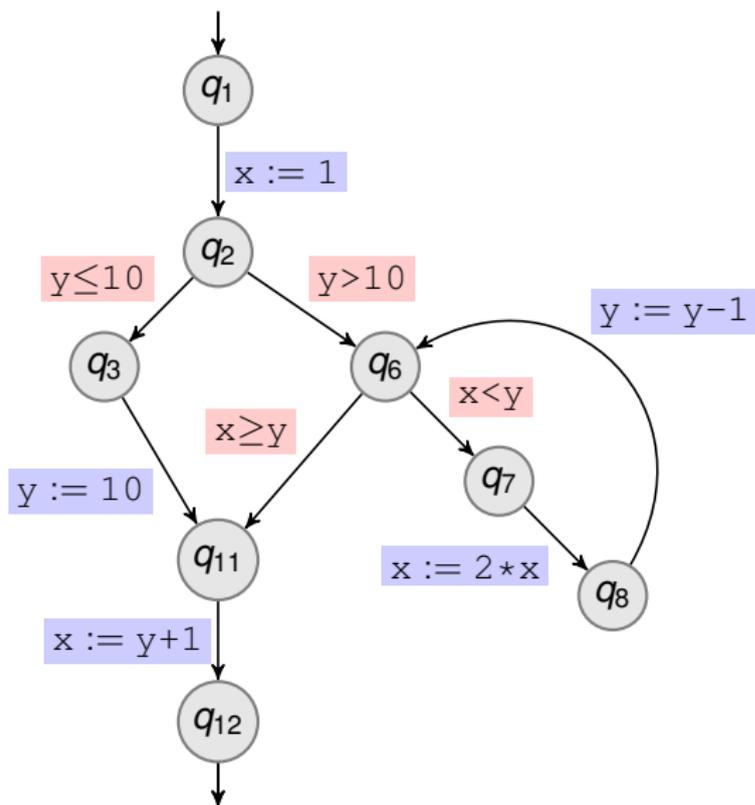
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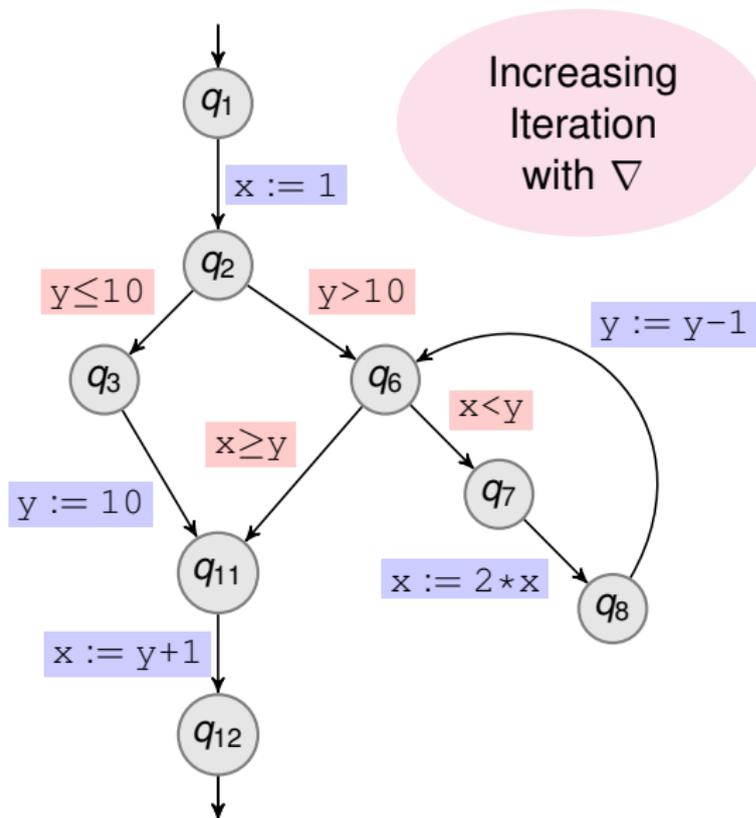
$$\bar{v}_1 \bar{\Delta} \bar{v}_2 = \lambda q. \bar{v}_1(q) \Delta \bar{v}_2(q)$$

# Range Analysis on Example with $\nabla$ and $\Delta$

Show that  $x > 0$  at  $q_{12}$



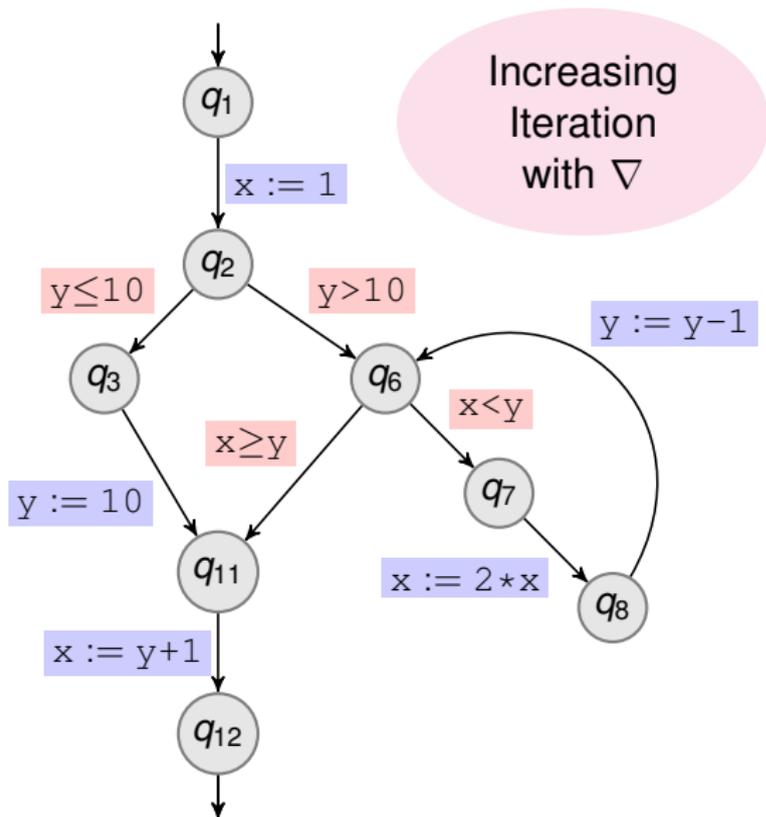
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Show that  $x > 0$  at  $q_{12}$

	x		y	
	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	$\perp$	$\perp$	$\perp$	$\perp$
$q_3$	$\perp$	$\perp$	$\perp$	$\perp$
$q_6$	$\perp$	$\perp$	$\perp$	$\perp$
$q_7$	$\perp$	$\perp$	$\perp$	$\perp$
$q_8$	$\perp$	$\perp$	$\perp$	$\perp$
$q_{11}$	$\perp$	$\perp$	$\perp$	$\perp$
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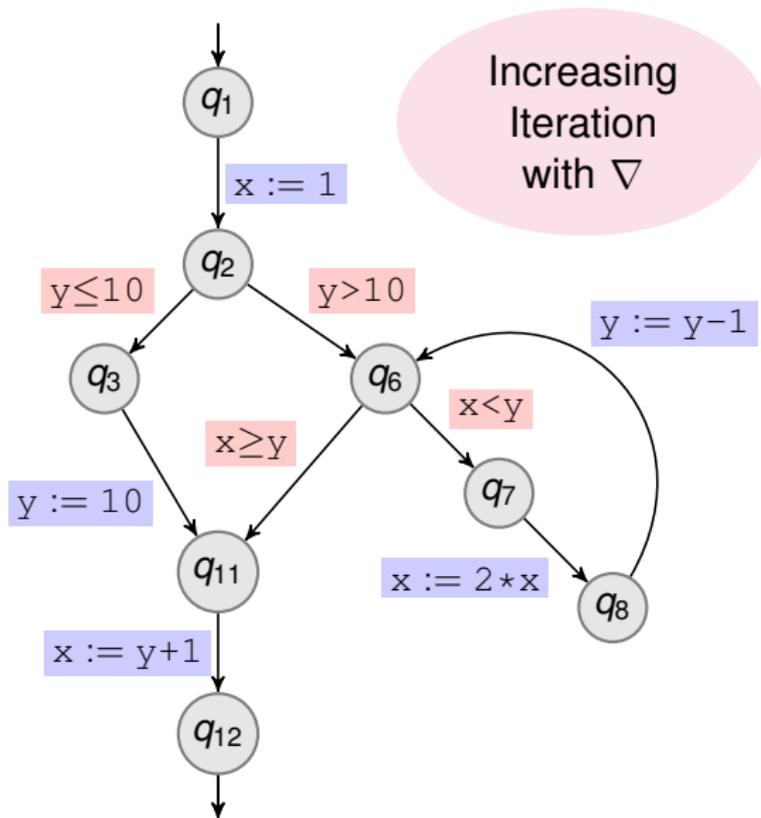


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	x		y	
	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_1$	$-\infty$	$+\infty$	$-\infty$	$+\infty$
$q_2$	1	1	$+\infty$	$+\infty$
$q_3$	1	1	$-\infty$	10
$q_6$	1	1	10	$+\infty$
$q_7$	1	1	10	$+\infty$
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$q_{11}$	1	1	10	10
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Same as with non-delayed widening

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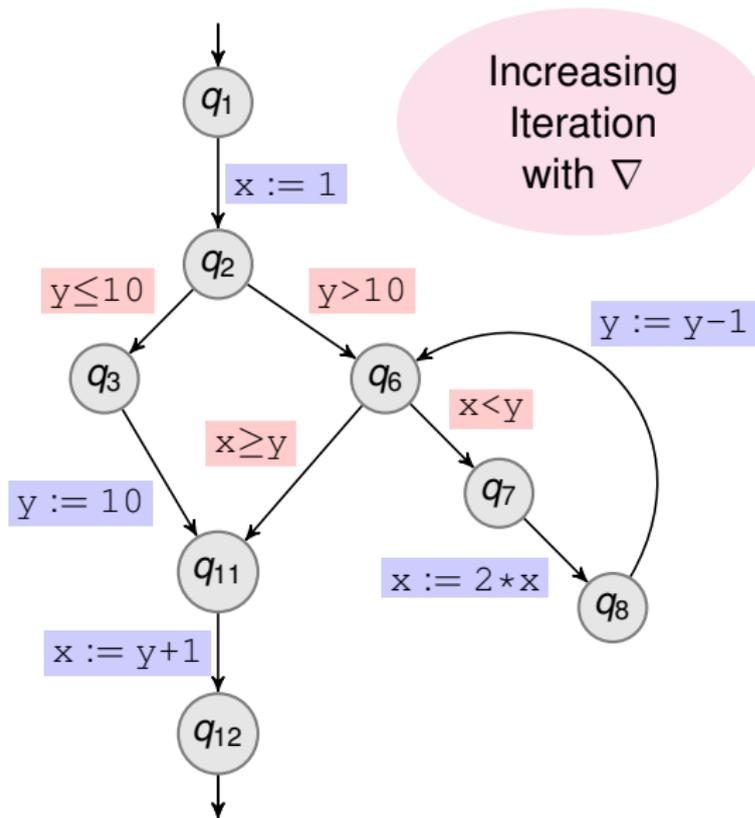


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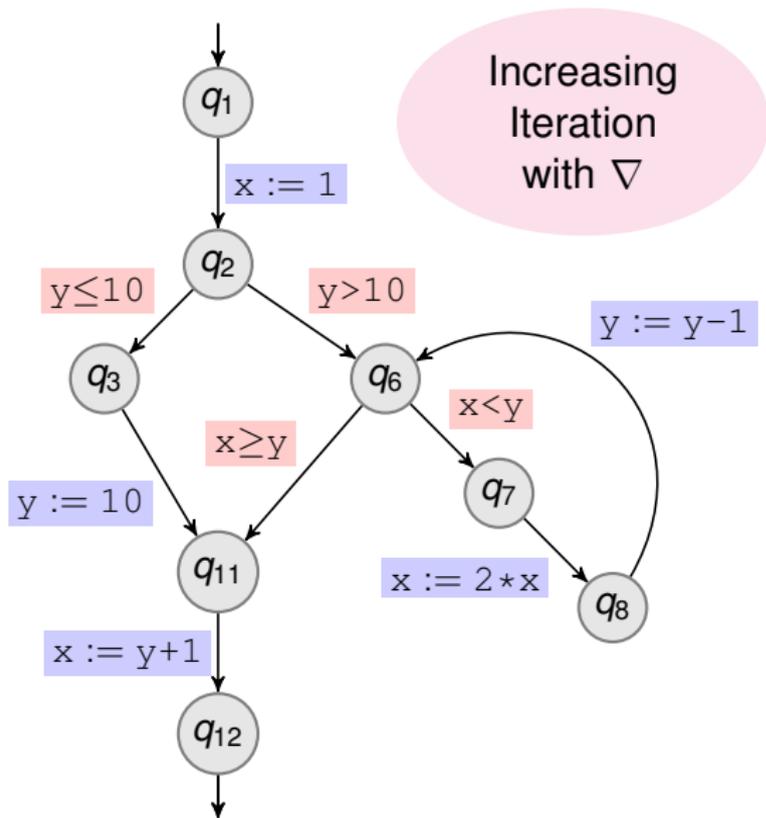


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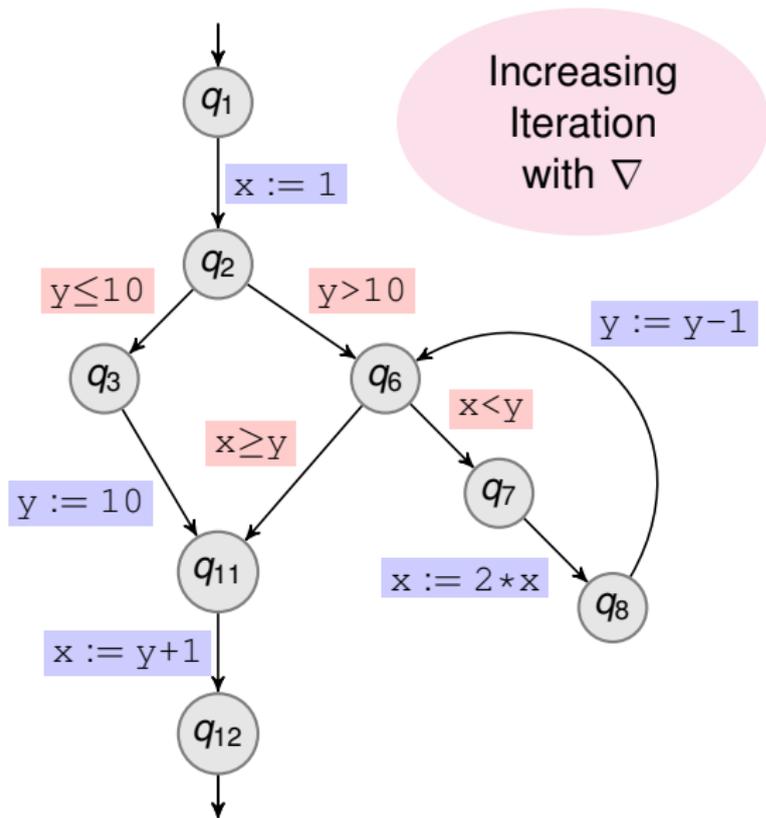


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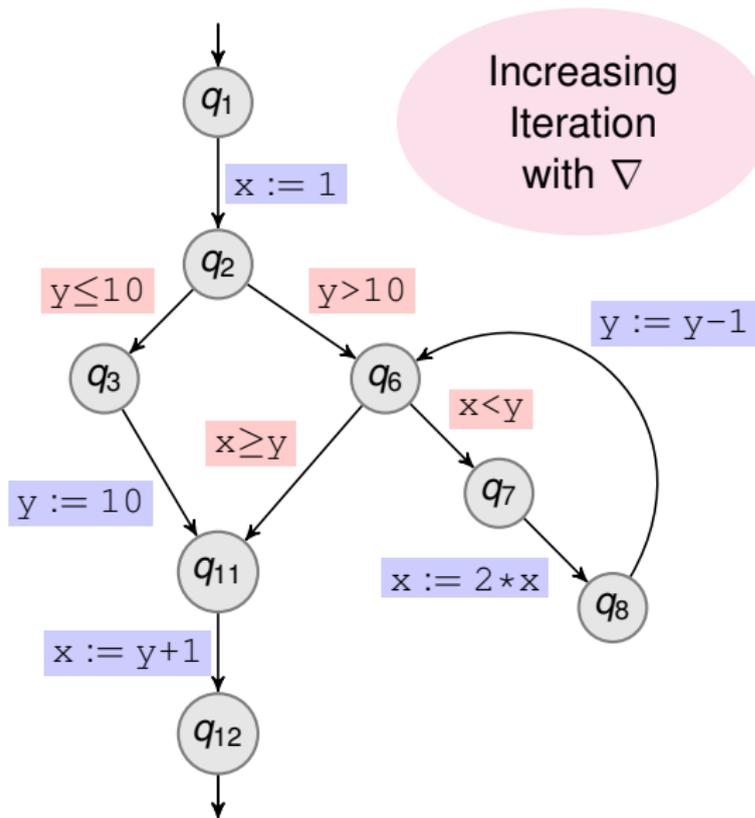


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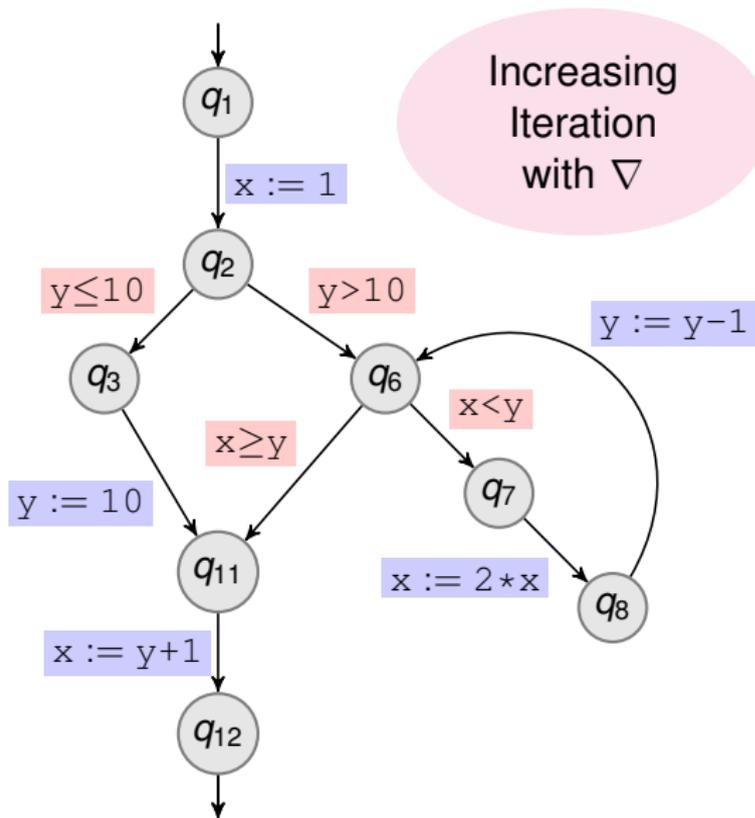


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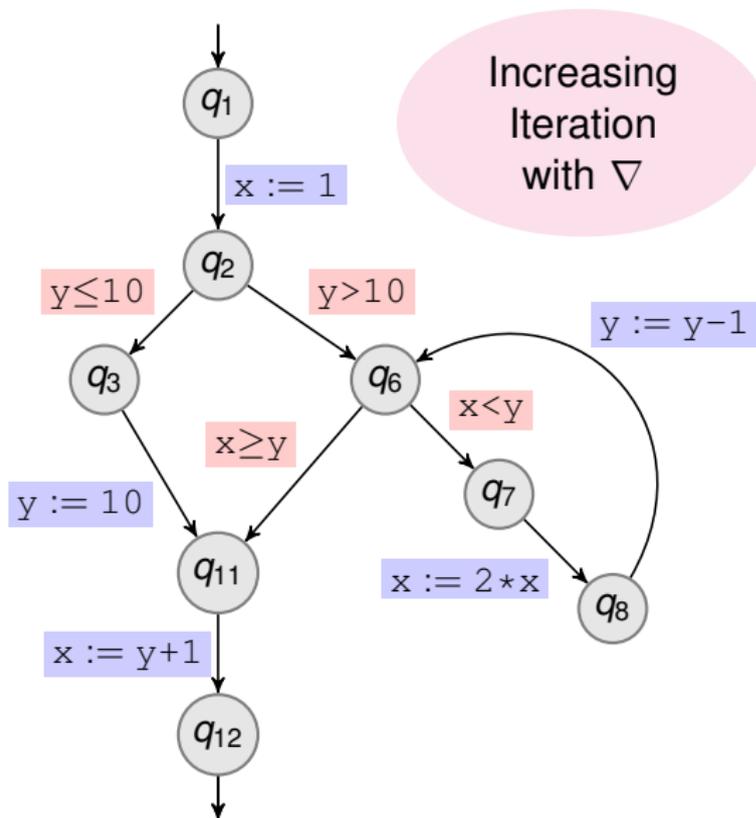


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$q_6$	1	$+\infty$	$-\infty$	$+\infty$
$q_7$	1	$+\infty$	$-\infty$	$+\infty$
$q_8$	2	$+\infty$	$-\infty$	$+\infty$
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$q_{12}$	$-\infty$	$+\infty$	$-\infty$	$+\infty$

Same as with non-delayed widening

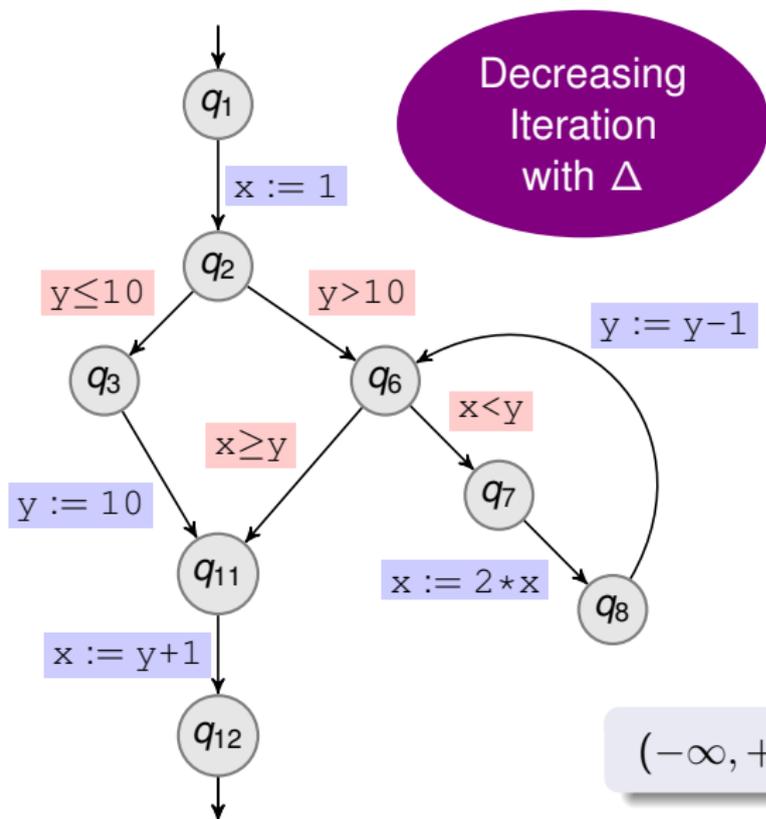
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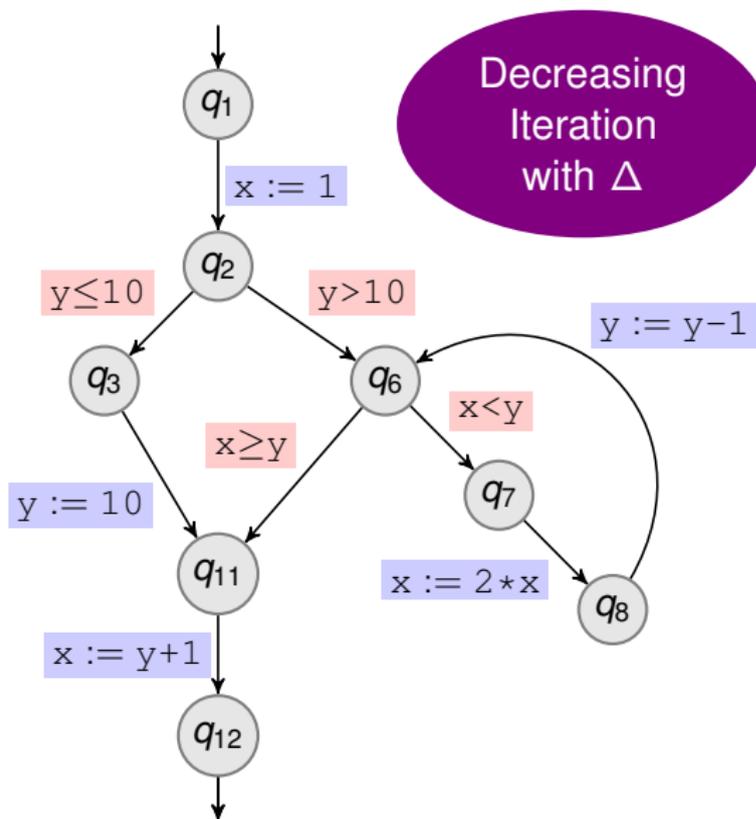


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$$(-\infty, +\infty) \Delta (1, +\infty) = (1, +\infty)$$

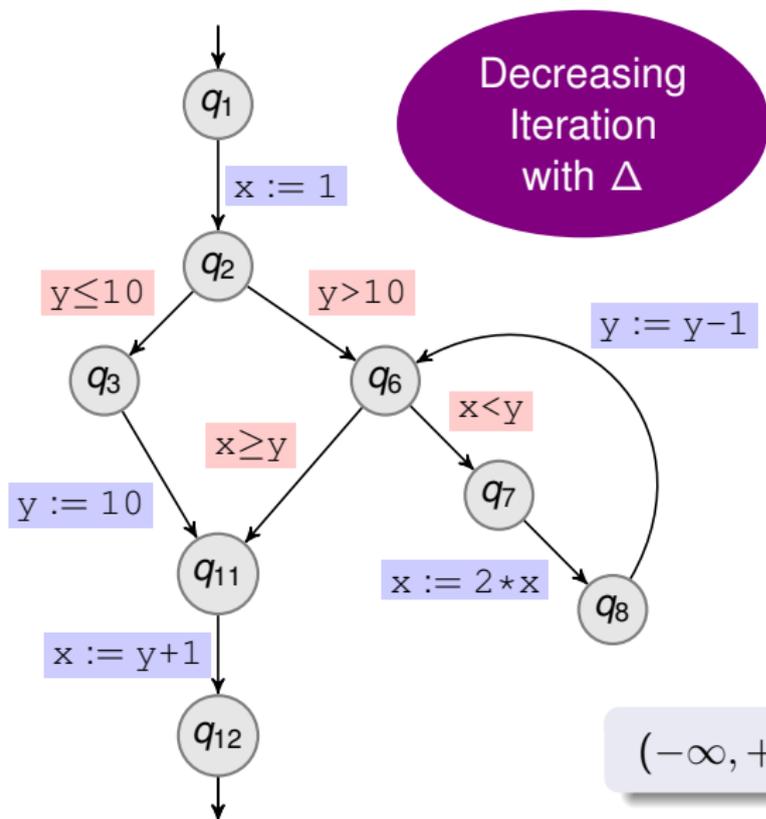
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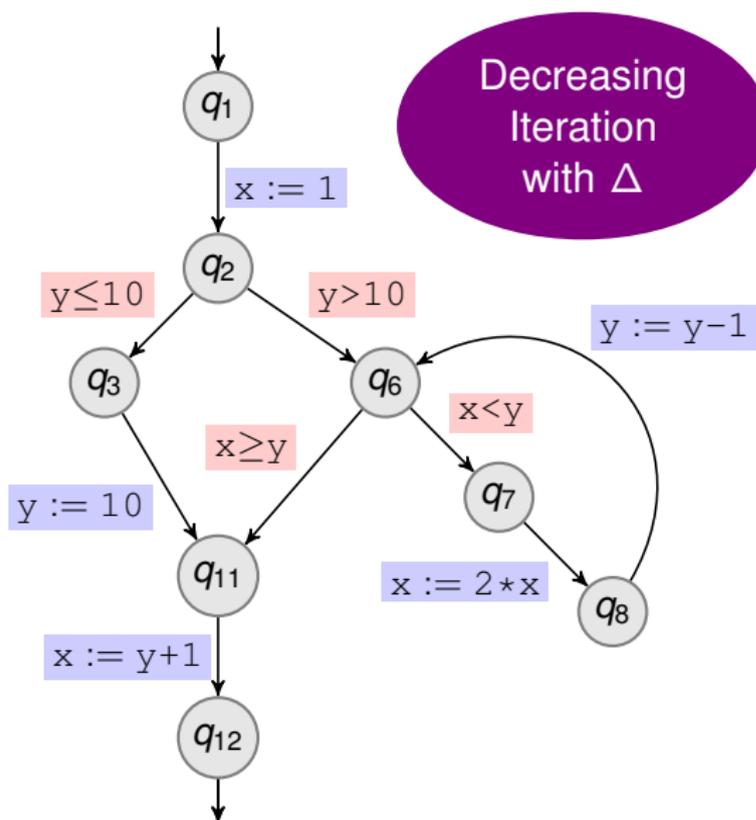


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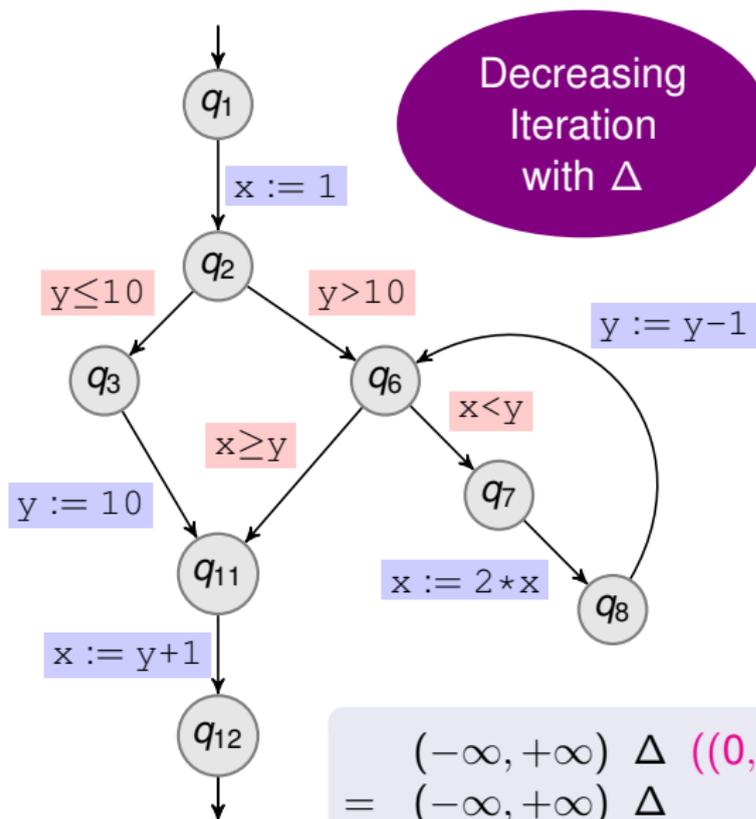
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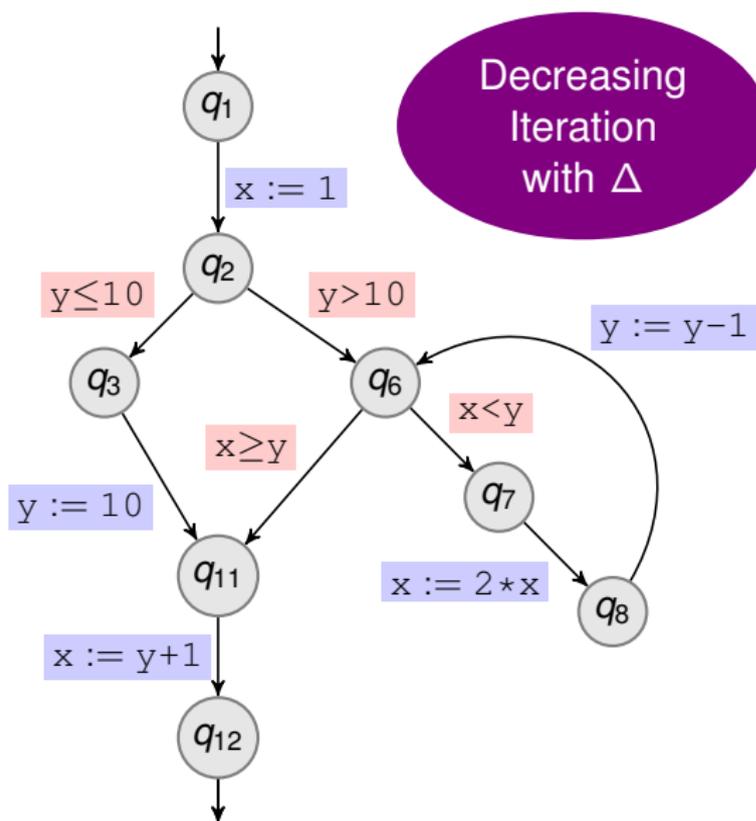


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$$\begin{aligned}
 & (-\infty, +\infty) \Delta ((0, +\infty) \sqcup (10, +\infty)) \\
 = & (-\infty, +\infty) \Delta (0, +\infty) = (0, +\infty)
 \end{aligned}$$

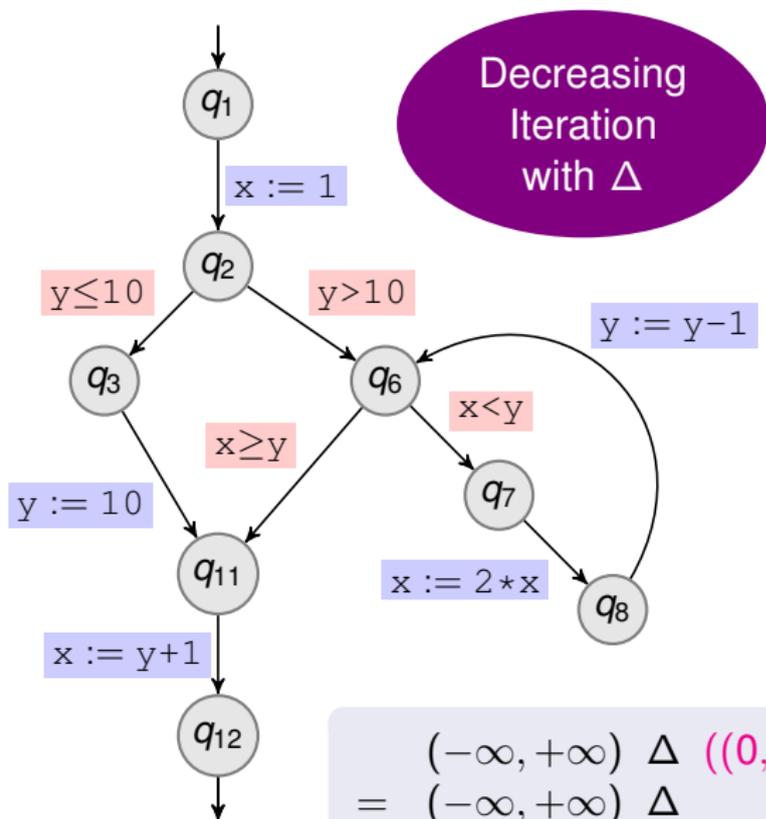
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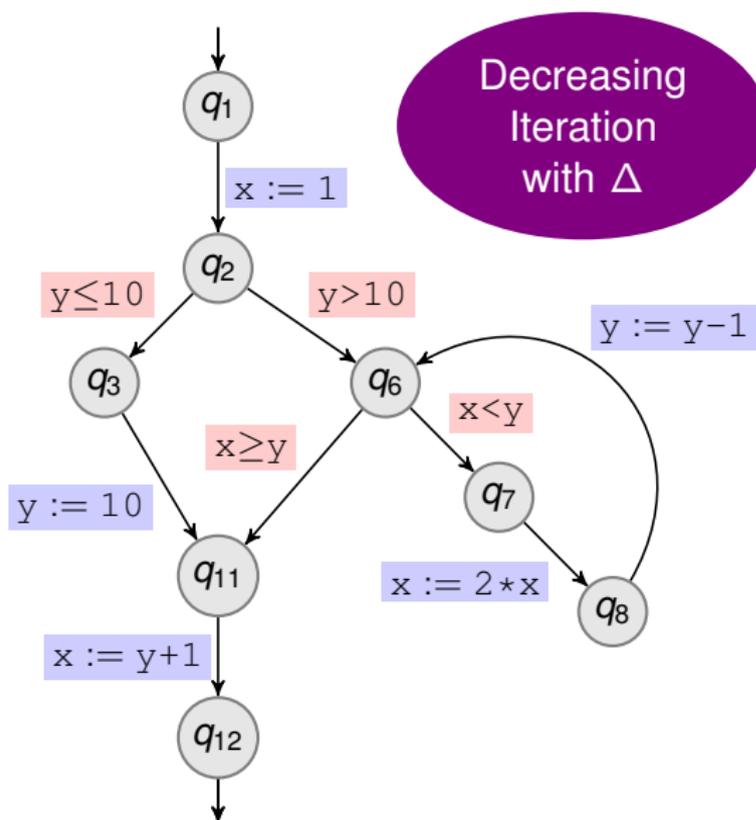


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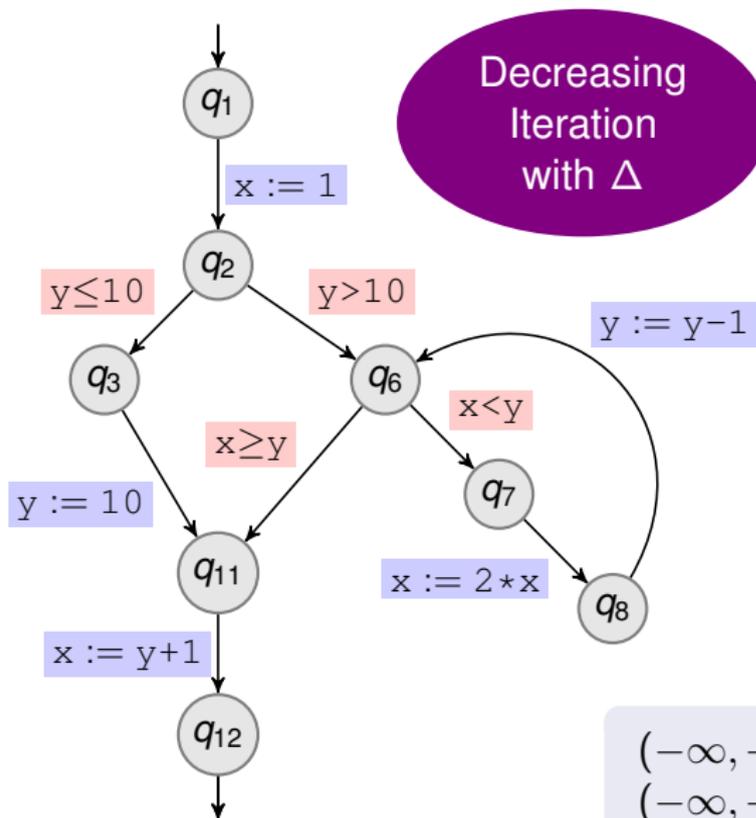
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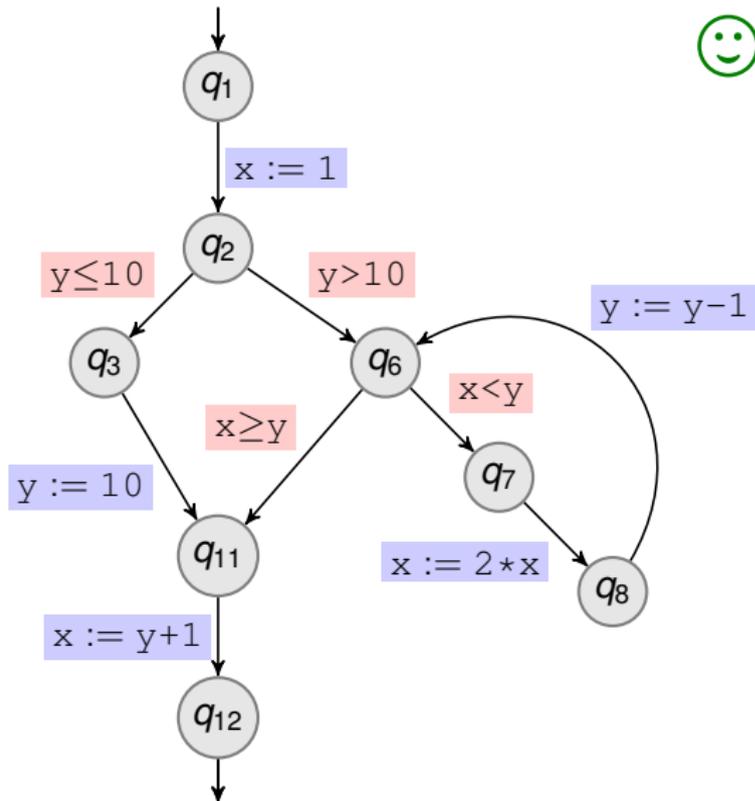
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# Selective Application of Widening

Widening introduces imprecision that often cannot be regained by narrowing.

To ensure convergence it is **enough** to only apply widening at **cut points**

Cut points: set of locations that cut each loop (in the control flow automaton's graph)

## Other Methods to Reduce Precision Loss of Widening

- Delayed widening
- Widening “up to”

Given a finite set  $M \subseteq L$ , use  $(x \nabla y) \sqcap \bigsqcap \{m \in M \mid a \sqcup b \sqsubseteq m\}$ .

- Look-ahead widening
- ...

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# Part V

## Software Verification by Static Analysis

- 11 Summary
- 12 Applications of Static Analysis to Software Verification
- 13 Limitations of Static Analysis for Software Verification
- 14 Some References

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# Summary: Data Flow Analysis

Compile-time techniques to gather **run-time** information about **data** in programs without actually running them

- Live Variables
- Available Expressions
- Uninitialized Variables
- Constant Propagation

## Monotone Data Flow Analysis Frameworks

### Minimal Fixpoint

- 😊 Computable in finite-height lattices
- 😞 Loss of Precision

### Meet Over All Paths

- 😊 Most precise solution
- 😞 Undecidable (constant propagation)

# Summary: Abstract Interpretation

**Semantics**-based systematic design of **correct** data flow analyses

Galois connections to formally relate abstract and concrete semantics

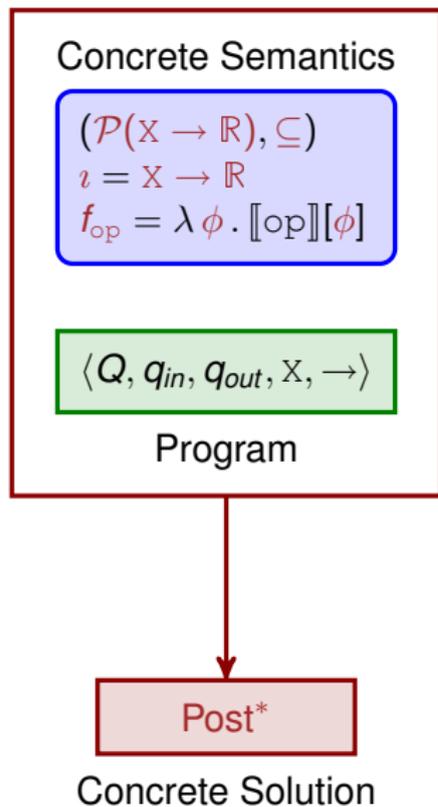
Safe approximations of the “best” abstract semantics

Convergence acceleration with widening and narrowing

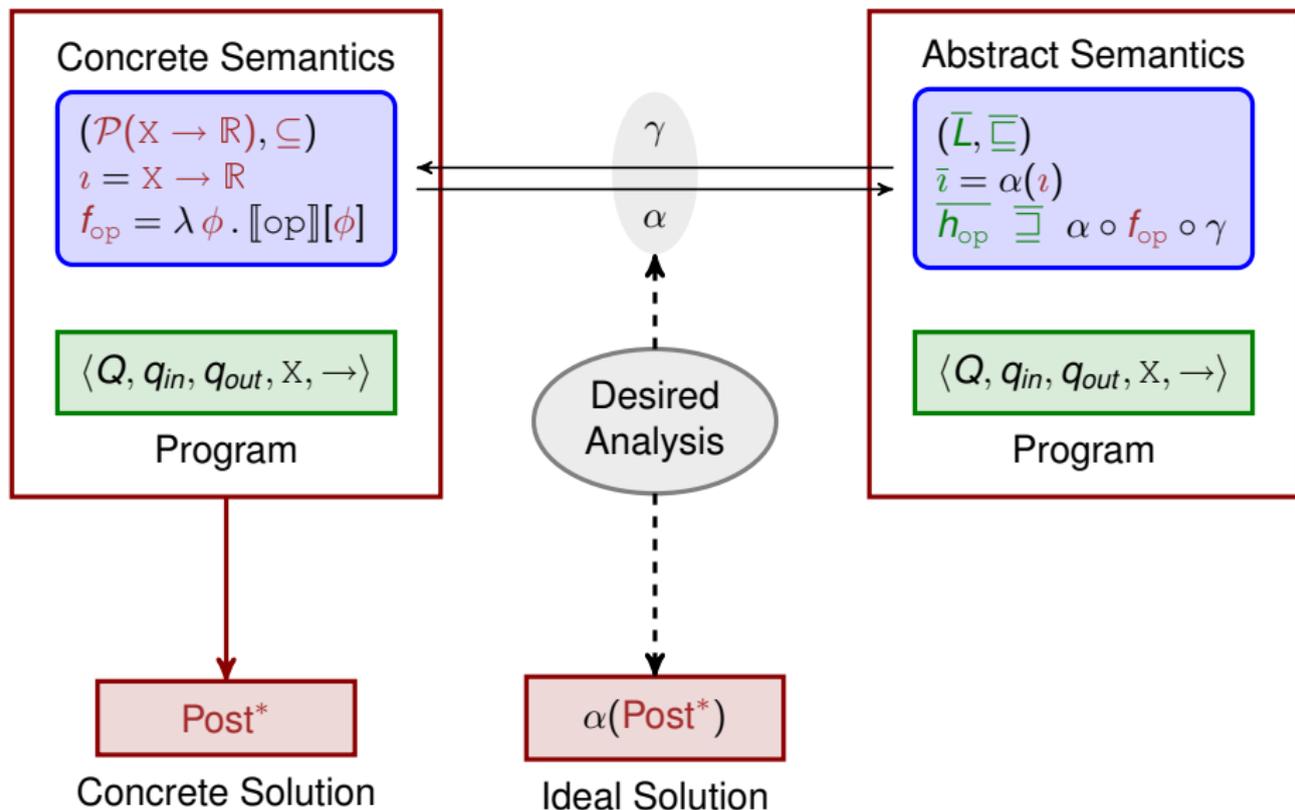
• Sign Analysis

• Range Analysis

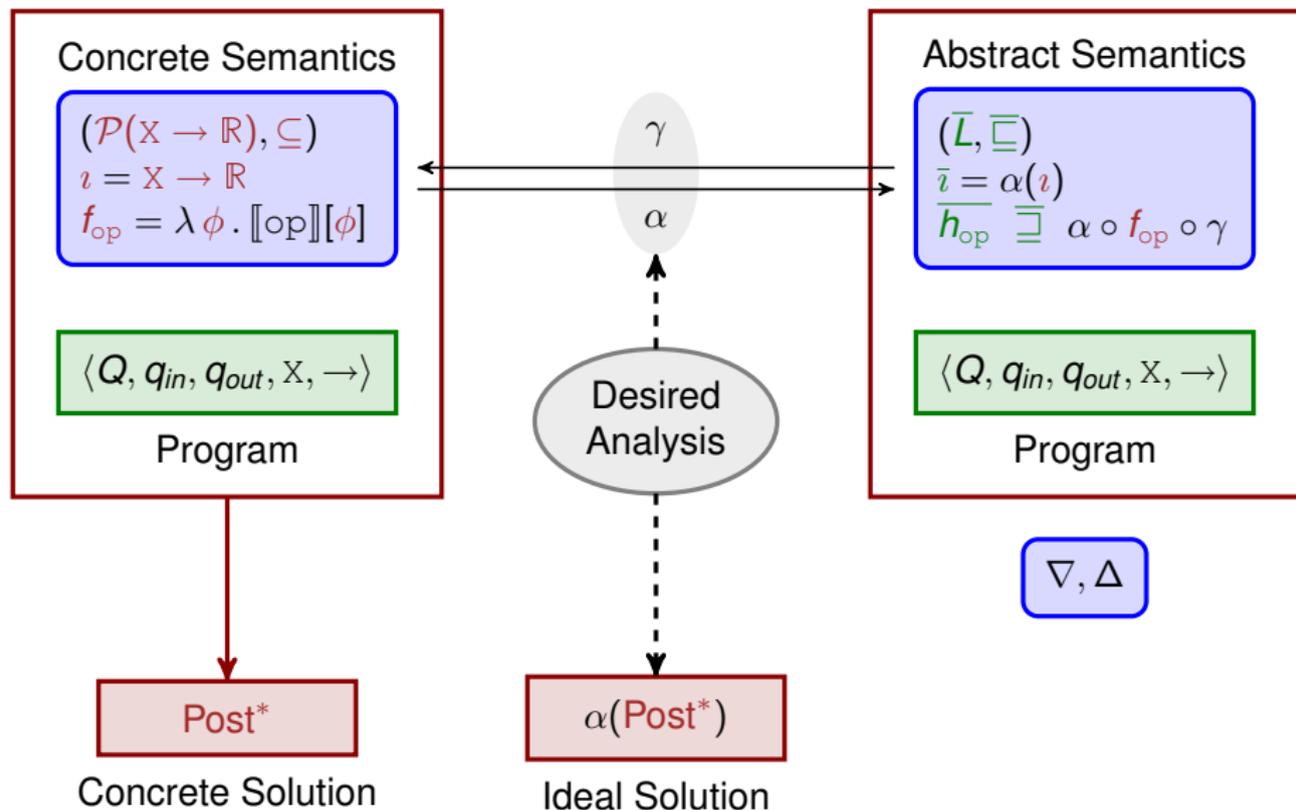
# Abstract Interpretation-Based Data Flow Analysis



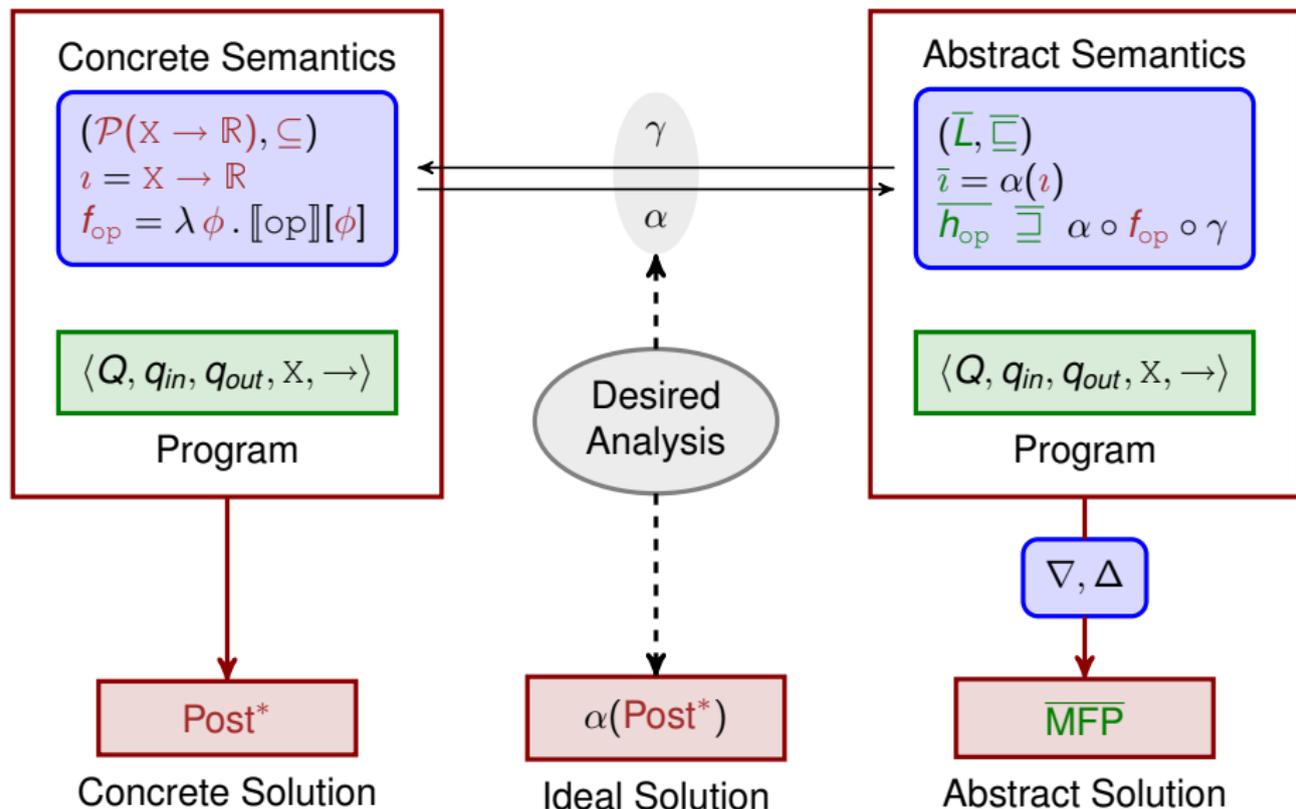
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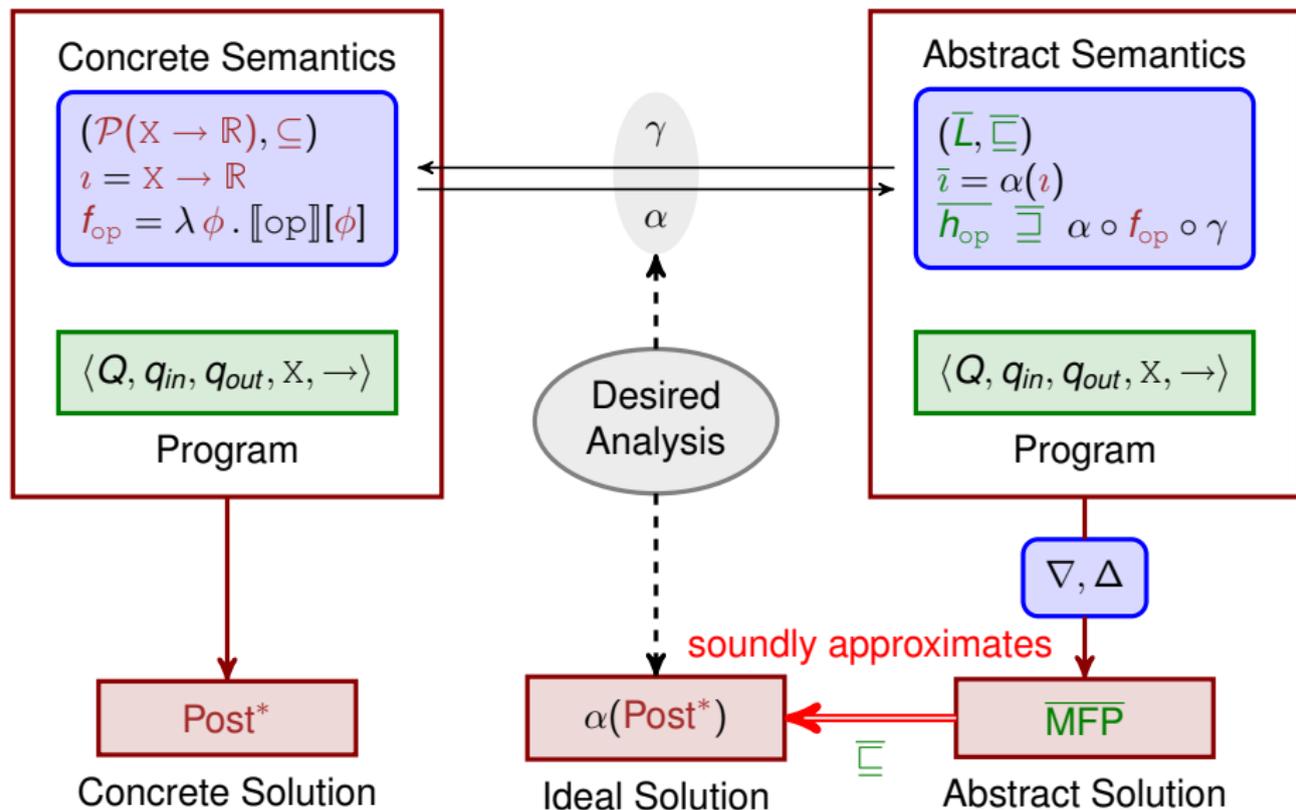
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11 Summary

**12 Applications of Static Analysis to Software Verification**

13 Limitations of Static Analysis for Software Verification

14 Some References

## Very Common Sources of Bugs

Uninitialized variables

Dead code

...

Can be detected by gen / kill data flow analyses

Data flow analysis in every compiler!

# Classical Data Flow Analysis in Compilers

```
1 class Fool {
2     static void fool(int x) {
3         int i, y;
4
5         for (i = 0 ; i < x ; i++) {
6             y = y + (i * i);
7         }
8     }
9 }
```

```
$ javac Fool.java
```

```
Fool.java:6: variable y might not have been initialized
```

```
    y = y + (i * i);
      ^
```

```
1 error
```

## Software Verification: Is $\text{Post}^*$ disjoint from $\text{Bad}$ ?

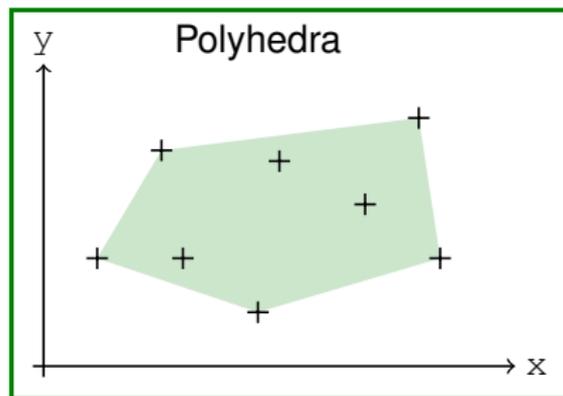
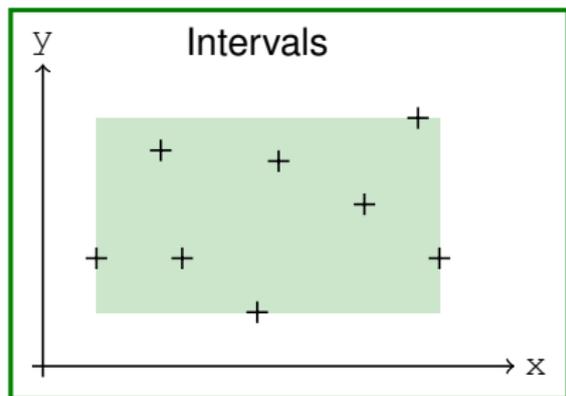
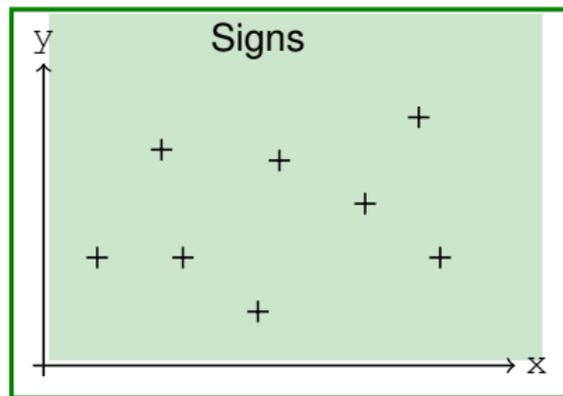
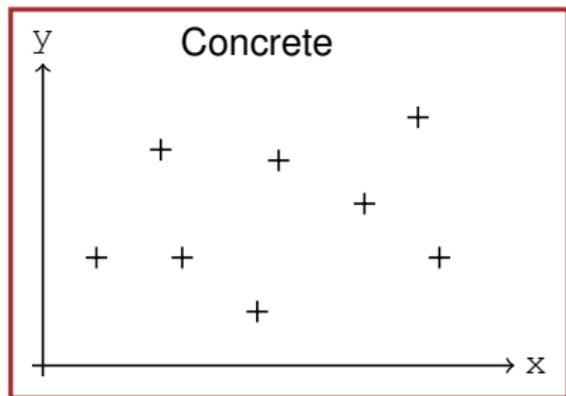
- 1 Compute an invariant  $\text{Inv} \supseteq \text{Post}^*$
- 2 If  $\text{Inv}$  is disjoint from  $\text{Bad}$  then return “program safe”

The MFP solution obtained by abstract interpretation is an invariant 😊

## Tradeoff between computational cost and precision

- Numerical abstract domains
- Approximate transfer mappings
- Widenings and narrowings

# Some Numerical Abstract Domains



# Some Commercial Static Analysis Tools

PolySpace™ Embedded Software Verification, The MathWorks™

*« PolySpace™ products verify C, C++, and Ada code for embedded applications by detecting run-time errors before code is compiled and executed. »*

Coverity Prevent™ Static Analysis for C/C++, for C#, and for Java

*« The foundation of Coverity's leading automated approach to identifying and resolving the most critical defects in C, C++, C# and Java source code. »*

Coverity periodically runs Coverity Prevent™ on open source projects

- Program Analyzer Generator (Saarland Univ. & AbsInt GmbH)
- Purify (IBM), Klocwork, ...

# Some Academic Static Analysis Tools

## ASTRÉE Static Analyzer — P. Cousot, R. Cousot, ...

- Abstract interpretation-based analysis of C
- Application to safety critical embedded software

Verification of the primary flight control software of the Airbus A340 and A380 fly-by-wire systems



## APRON Numerical Abstract Domain Library — B. Jeannet, ...

- Common interface to various abstract domains
  - includes intervals, polyhedra, octagons, linear congruences
- Online demonstration of the Interproc analyzer
- Open-source, released under the GNU LGPL

- 11 Summary
- 12 Applications of Static Analysis to Software Verification
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- 14 Some References

## False Positives and False Negatives

### Example of False Positive

Variable detected as **not initialized**, but in fact it is **initialized for all runs** of the program.

### Example of False Negative

No code detected as **dead**, but in fact **some program point cannot be reached** by any run.

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### Example of False Negative

No code detected as **dead**, but in fact **some program point cannot be reached** by any run.

# Classical Data Flow Analysis in Compilers

```
1 class Foo2 {  
2     static int foo2(int x) {  
3         int y;  
4  
5         if (x == 0) { y = 5; }  
6         else      { y = 2; }  
7         return y;  
8     }  
9 }
```

```
$ javac Foo2.java  
$
```

# Classical Data Flow Analysis in Compilers

```
1 class Foo3 {  
2     static int foo3(int x) {  
3         int y;  
4  
5         if (x == 0) { y = 5; }  
6         if (x != 0) { y = 2; }  
7         return y;  
8     }  
9 }
```

```
$ javac Foo3.java  
Foo3.java:7: variable y might not have been initialized  
    return y;  
           ^  
1 error
```

## False Positives

Software Verification: Is  $\text{Post}^*$  disjoint from  $\text{Bad}$ ?

- 1 Compute an invariant  $\text{Inv} \supseteq \text{Post}^*$
- 2 If  $\text{Inv}$  is disjoint from  $\text{Bad}$  then return “program safe”
- 3 If  $\text{Inv}$  intersects  $\text{Bad}$  then return “alarm”

In practice, there might be too many false alarms...

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In practice, there might be too **many false alarms**...

# What can we do about it?

## Software Verification by Static Analysis: Workflow

While the analysis returns alarms

- 1 **Inspect** alarms to determine whether they are spurious or not
- 2 If alarms are spurious then **refine** the analysis to gain precision

Why not automate this process?

Trade termination guarantee with fully automatic model-checking

- ☹ Not acceptable for compile-time static analyses
- ☺ Acceptable for verification

Topic of next part...

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# Some References

 F. Nielson, H. R. Nielson, and C. Hankin.

*Principles of Program Analysis.*

Springer, 1999.



P. Cousot and R. Cousot.

Systematic design of program analysis frameworks.

In *Proc. 6th ACM Symp. Principles of Programming Languages, San Antonio, TX, USA*, pages 269–282. ACM Press, 1979.

▶ The **ASTRÉE** Static Analyzer.

<http://www.astree.ens.fr/>

▶ The **APRON** Library for Numerical Abstract Domains.

<http://apron.cri.enscm.fr/library/>

▶ **Coverity's Scan.**

<http://scan.coverity.com/>

## Part VI

# Abstract Model Refinement

- 15 Introduction and Overview
- 16 Basic Theory on Property-Preserving Abstractions
- 17 Abstraction Schemes
- 18 Counterexample Guided Refinement

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Alarms must be inspected manually 😞

If an alarm is a real bug, then the analysis is useful 😊

Otherwise...

An improved analysis must be designed to eliminate alarms

# Software Verification by Static Analysis (Repetition)

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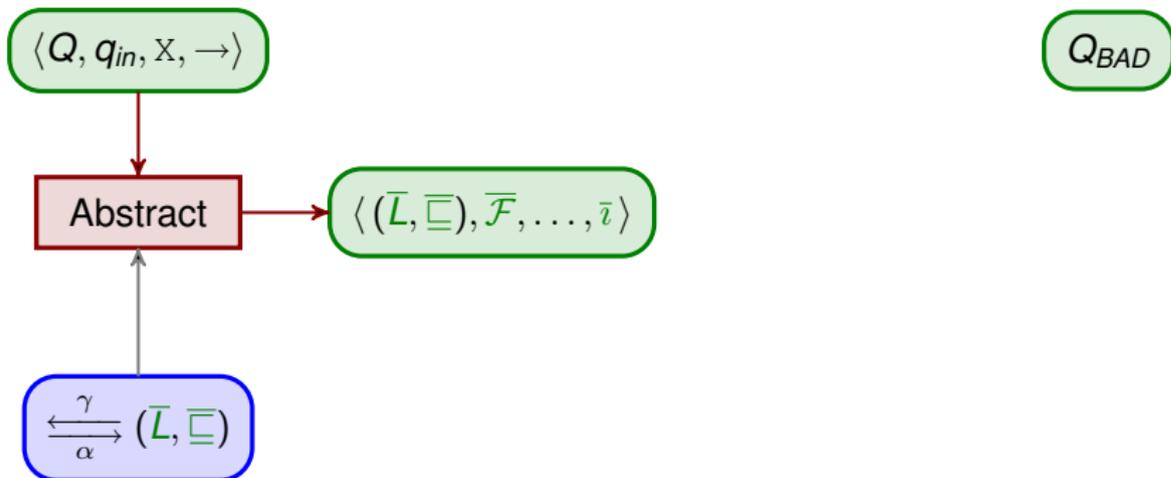
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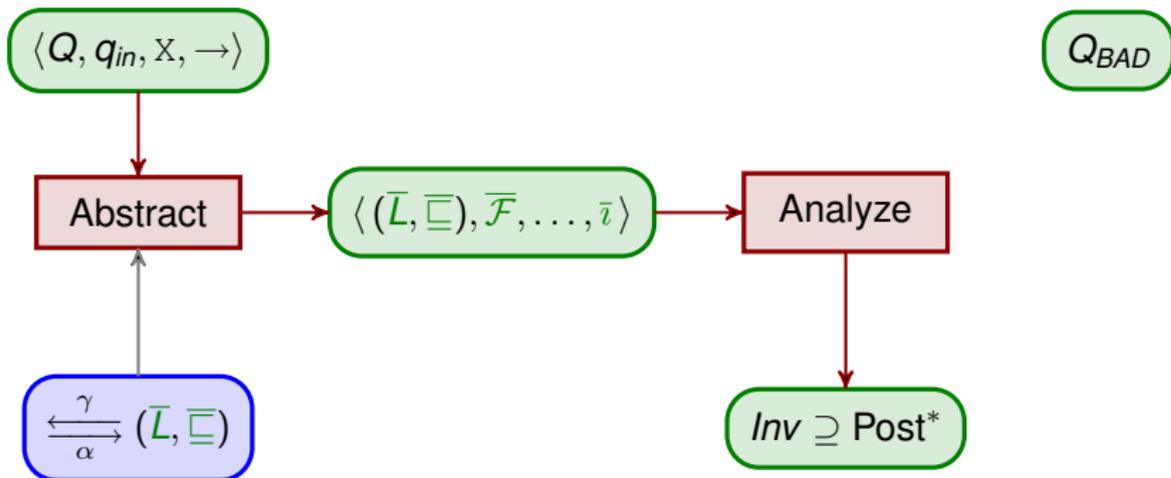
$\langle Q, q_{in}, X, \rightarrow \rangle$

$Q_{BAD}$

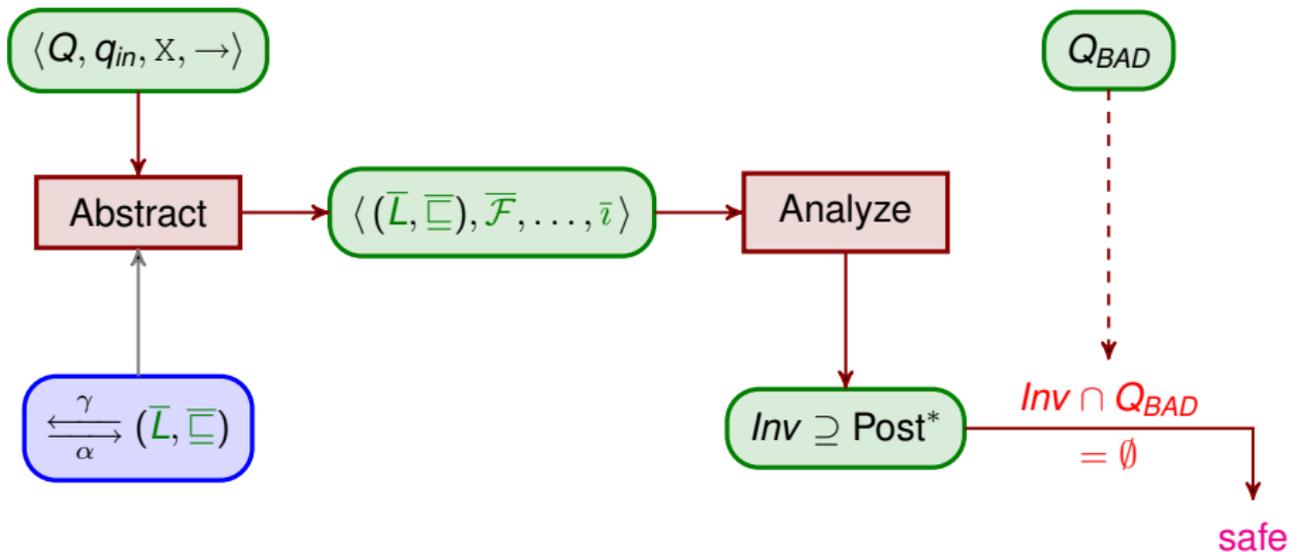
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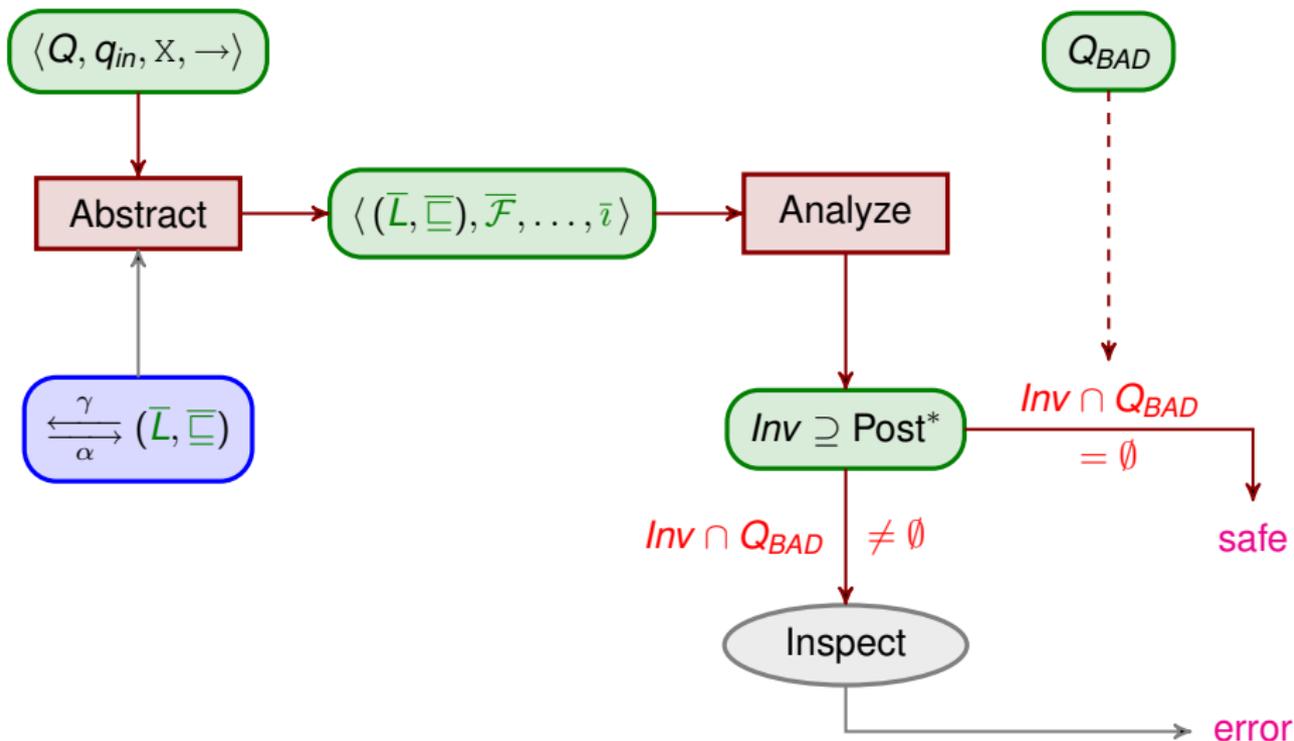
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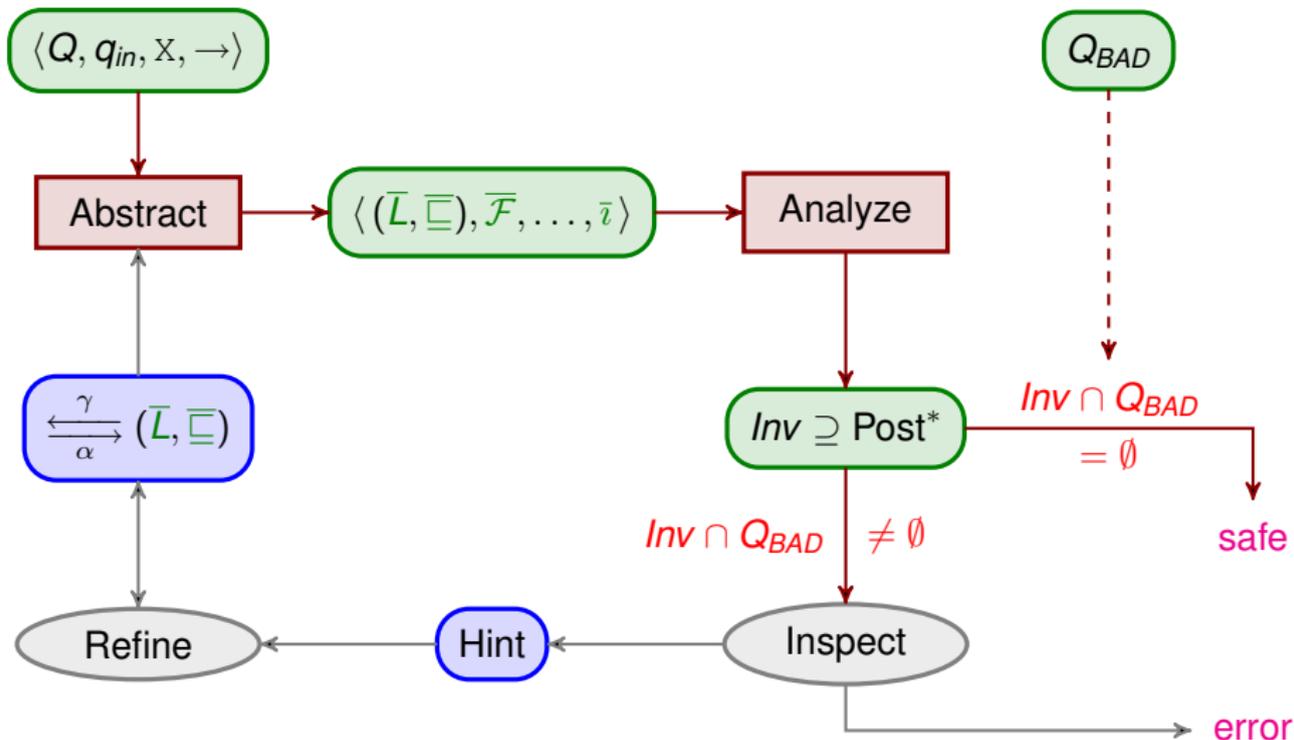
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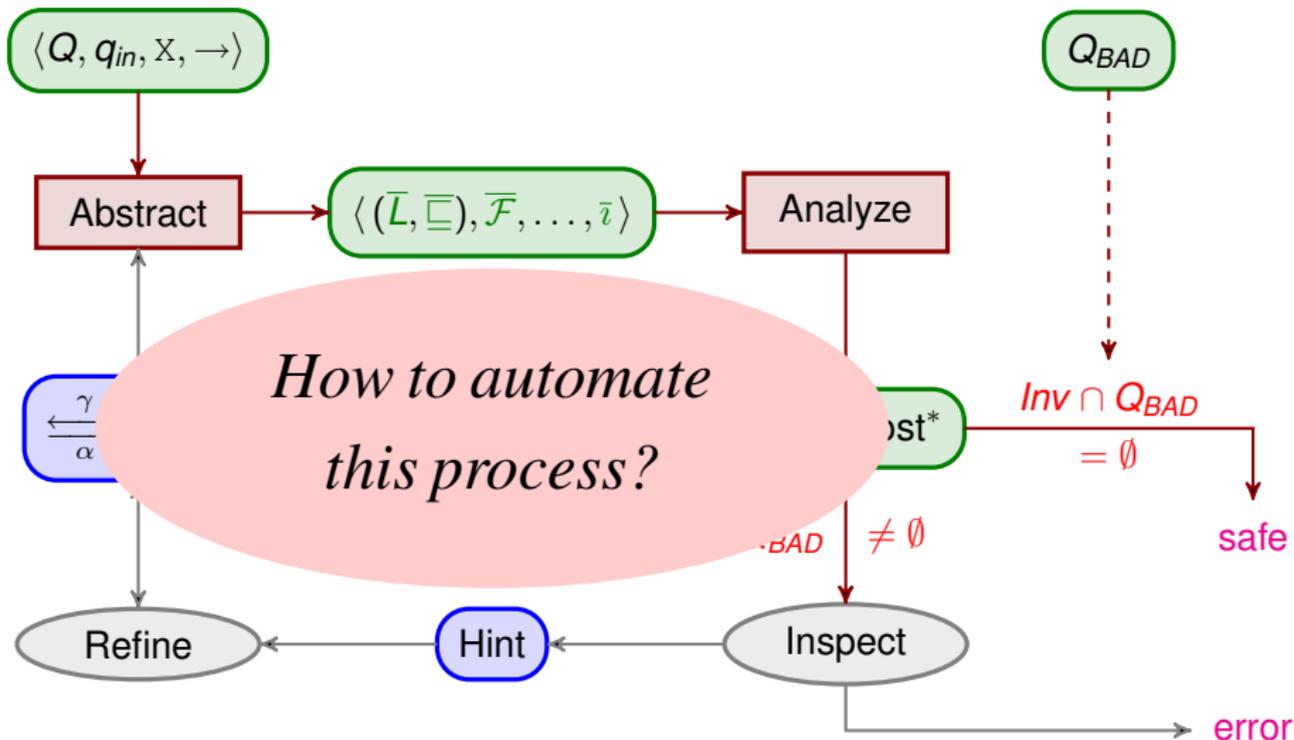
# Software Verification by Static Analysis: Workflow



# Software Verification by Static Analysis: Workflow



# Software Verification by Static Analysis: Workflow



# Inspection of Alarms: Not a Simple Task!

## Objective

Given an abstract invariant  $\overline{Inv}$  whose concretization  $\gamma(\overline{Inv})$  intersects  $Bad = Q_{BAD} \times (X \rightarrow \mathbb{R})$ , determine whether  $Post^*$  intersects  $Bad$ .

All configurations in  $\gamma(\overline{Inv}) \cap Bad$  are potentially reachable. . .

How are these configurations potentially reached?

It would be nice to have an “abstract run” of the form:

$$(q_{in}, \overline{\psi}_0) \xrightarrow{op_0} \dots \xrightarrow{op_{k-1}} (q_k, \overline{\psi}_k) \quad \text{with} \quad \begin{cases} q_k \in Q_{BAD} \\ \gamma(\overline{\psi}_k) \cap Bad \neq \emptyset \end{cases}$$

Checking whether this abstract run is spurious reduces to checking emptiness of the relation:  $([[op_k]] \circ \dots \circ [[op_0]])$ .

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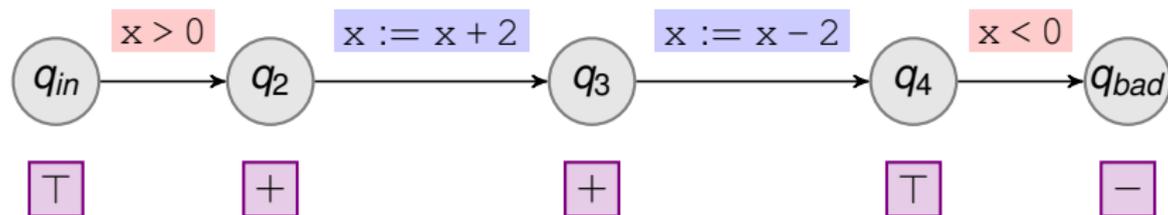
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# Inspection of Abstract Runs: Example



## Semantics of Operations (Repetition)

$(v, v') \in \llbracket g \rrbracket$  if  $v \models g$  and  $v' = v$

$(v, v') \in \llbracket x := e \rrbracket$  if  $\begin{cases} v'(x) = \llbracket e \rrbracket_v \\ v'(y) = v'(y) \end{cases}$  for all  $y \neq x$

$$\llbracket x < 0 \rrbracket \circ \llbracket x := x - 2 \rrbracket \circ \llbracket x := x + 2 \rrbracket \circ \llbracket x > 0 \rrbracket = \emptyset$$

$x > 0 \wedge x' = x + 2 \wedge x'' = x' - 2 \wedge x'' < 0$  *unsatisfiable*

# Refinement of Abstract Domains: Not a Simple Task!

## Objective

Given an abstract invariant  $\overline{Inv}$  and a subset  $U \subseteq \gamma(\overline{Inv}) \setminus \text{Post}^*$ , design a new abstract domain where the resulting  $\overline{Inv}$  is disjoint from  $U$ .

$U$  would be a set of configurations identified as false alarms.

Quite challenging!

## More Reasonable Objective

Given a spurious “abstract run”, design a new abstract domain that eliminates this “abstract run”.

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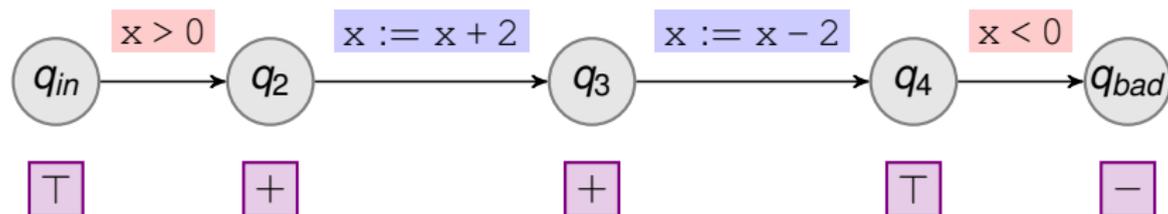
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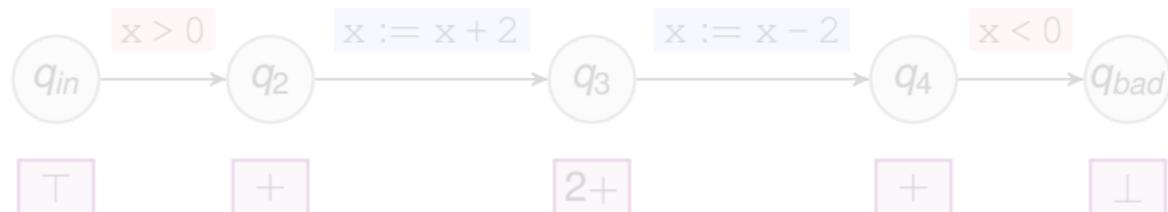
# Refinement Based on Abstract Runs: Example



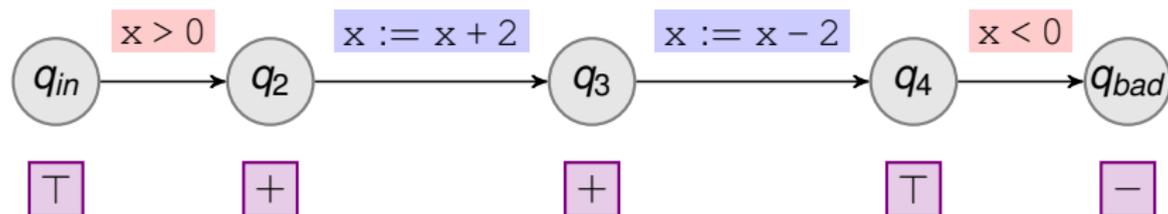
In  $q_3$ , the set of reachable valuations is:

$$\llbracket x := x + 2 \rrbracket \circ \llbracket x > 0 \rrbracket [(X \rightarrow \mathbb{R})] = \{v \in X \rightarrow \mathbb{R} \mid v(x) > 2\}$$

We lack the “property”  $x > 2$ . Let us add it (as 2+) to the *Sign* domain.



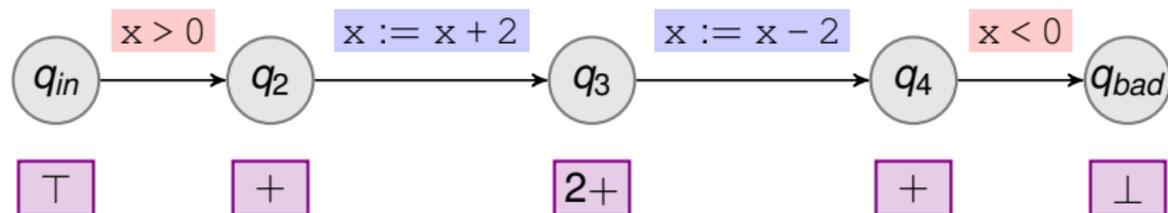
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# Hypothetical Workflow Based on Abstract Runs

Abstract “counterexample” runs are key to:

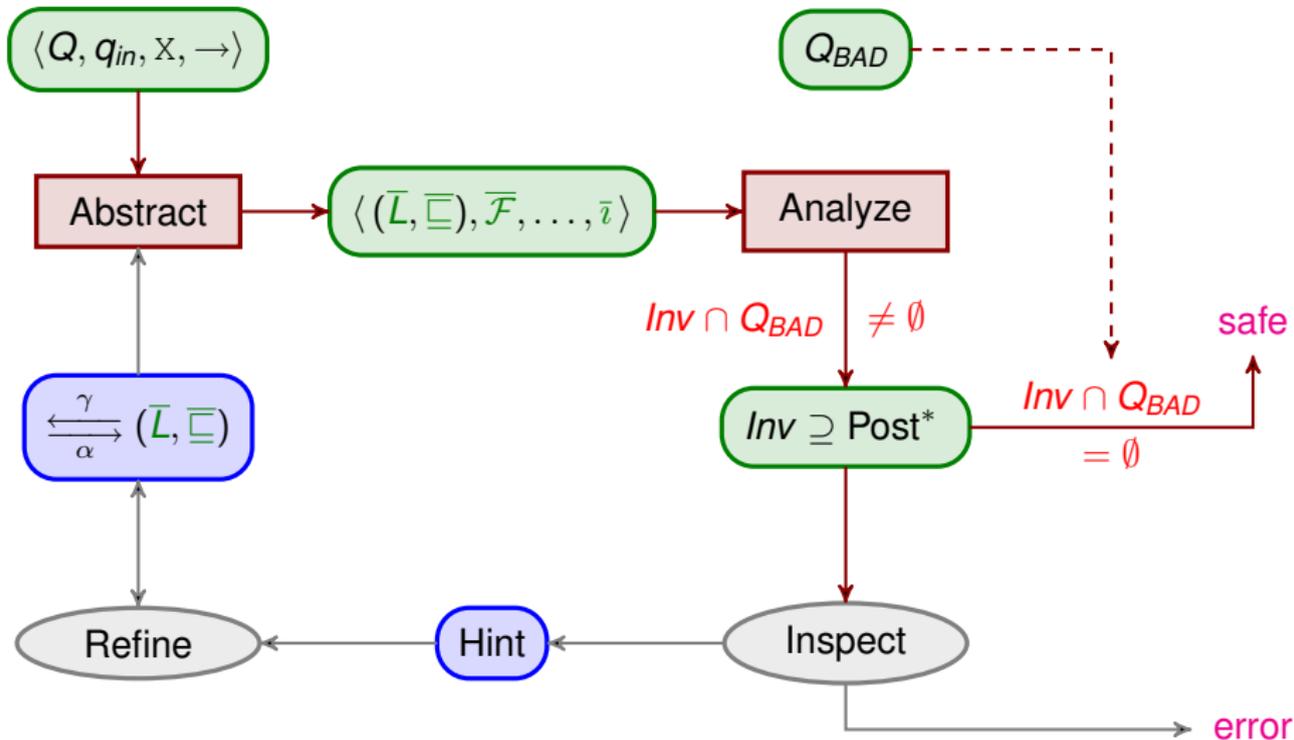
- inspection of alarms
- refinement of abstract domains

## Enhanced Workflow Based on Abstract Runs

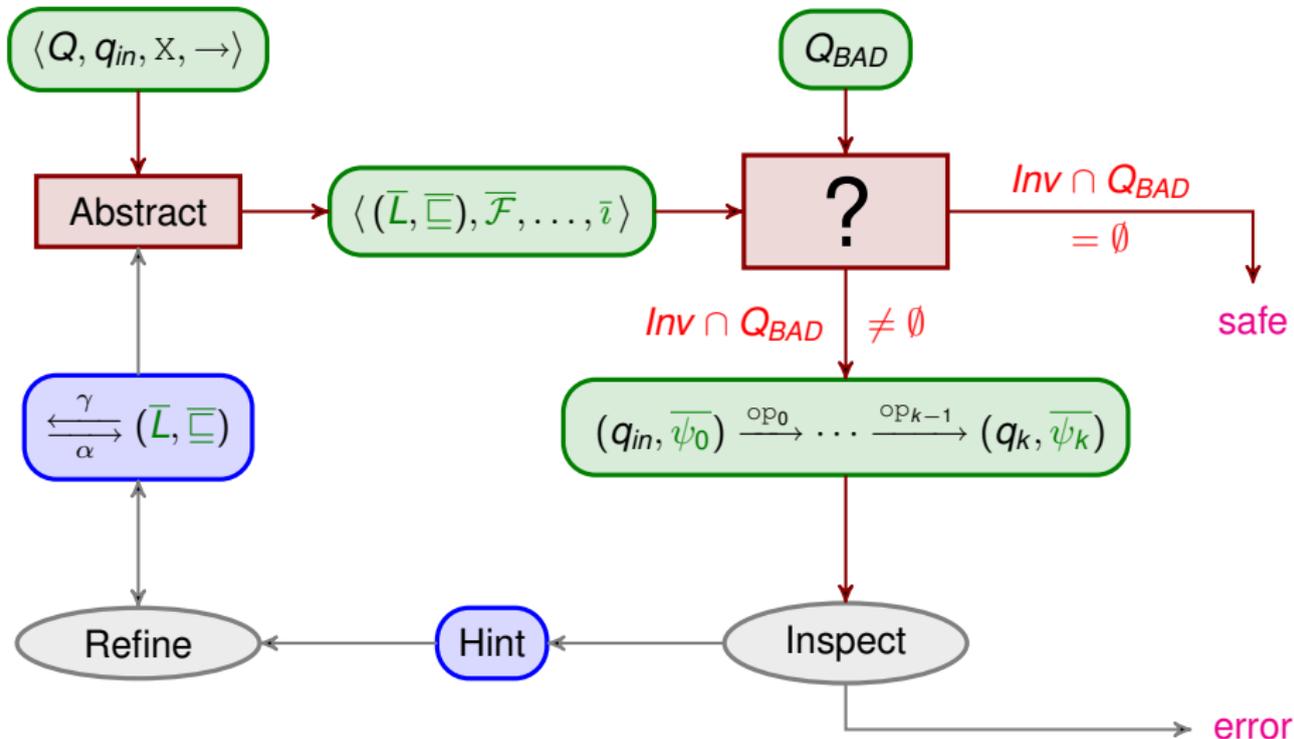
Imagine a hypothetical workflow where the analyzer returns:

- either “program safe” if it finds an invariant  $Inv$  disjoint from  $Bad$
- or “alarm” with an abstract run as a potential counterexample

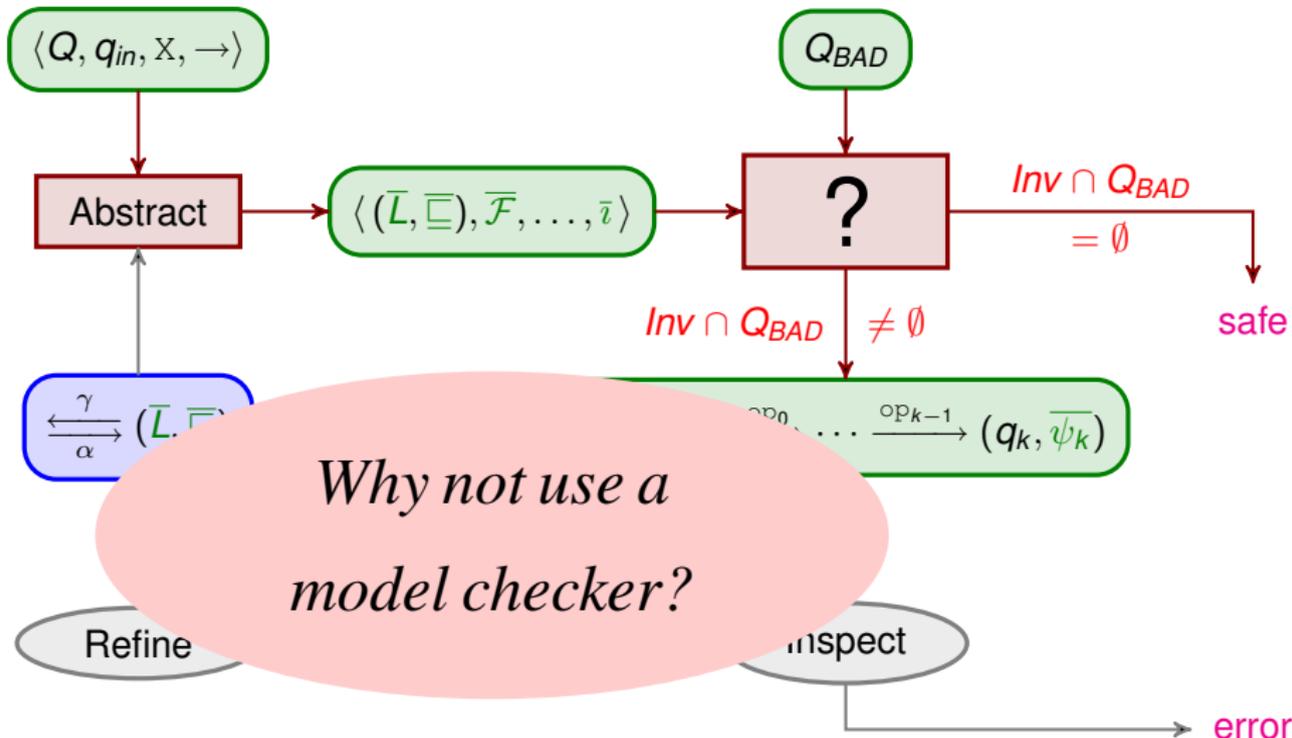
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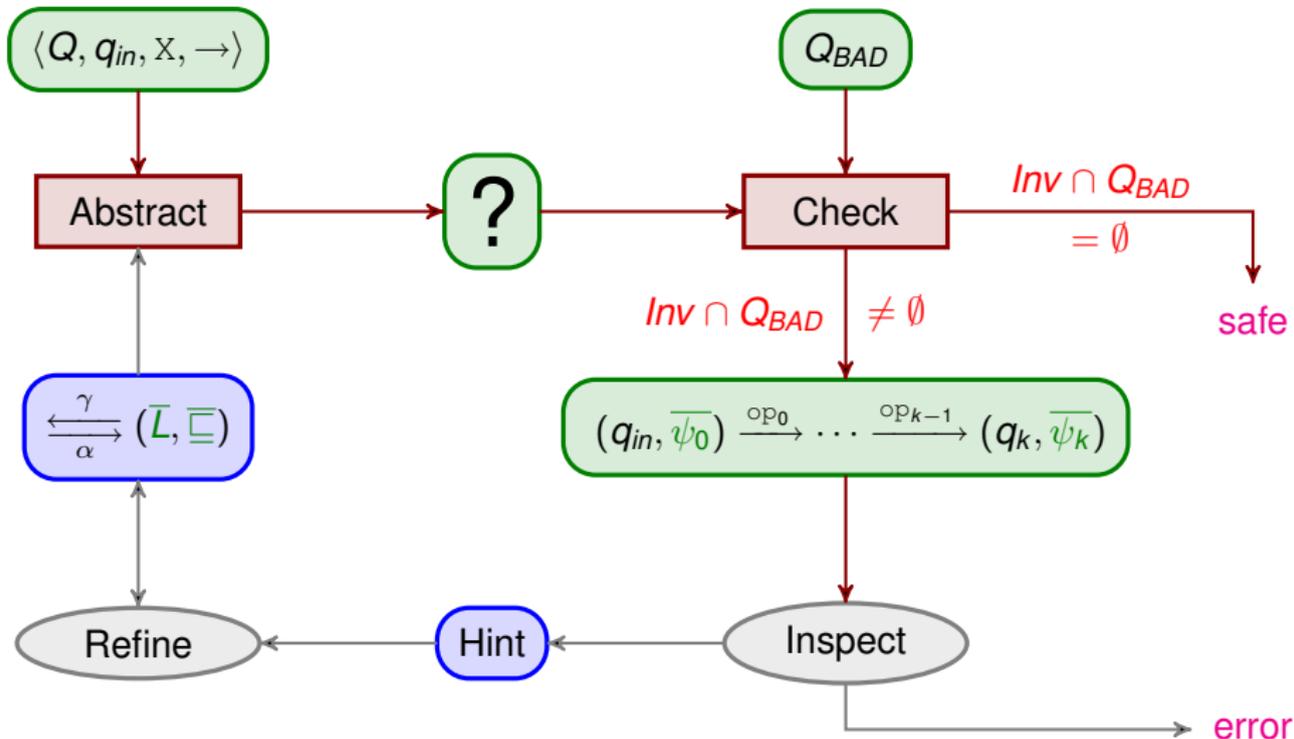
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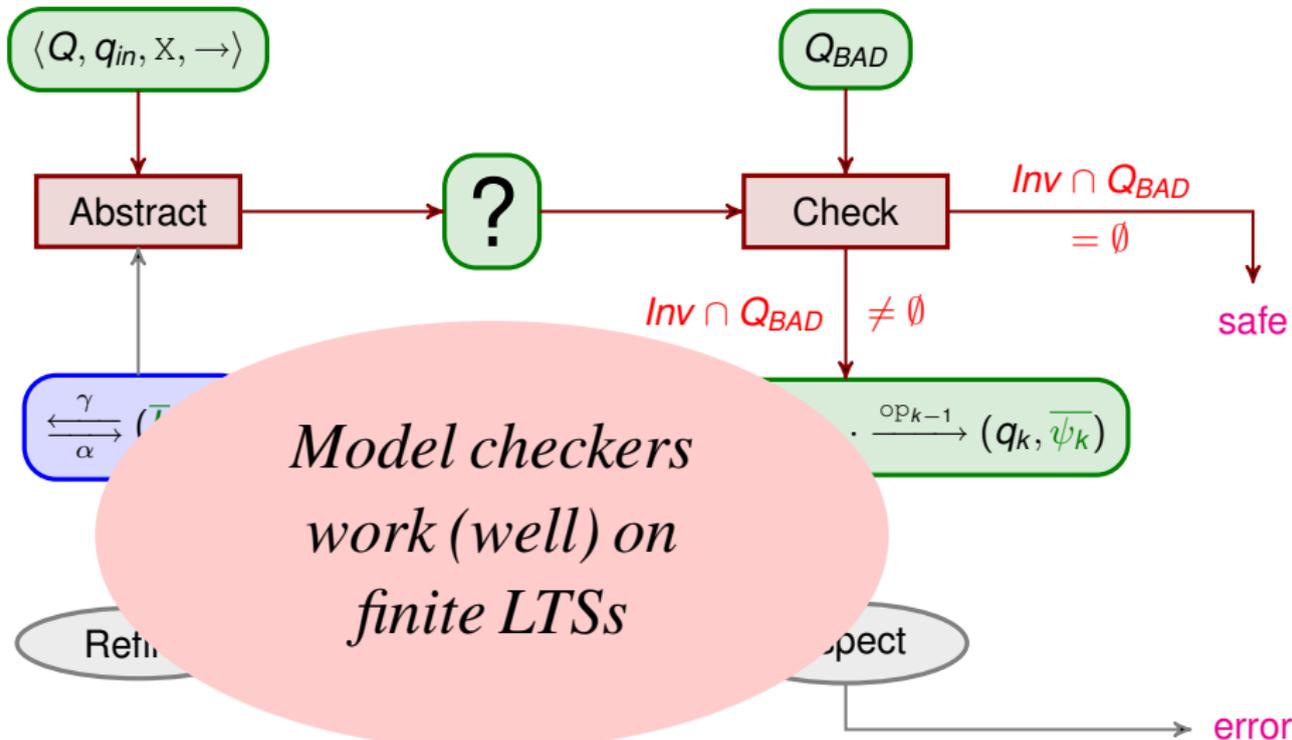
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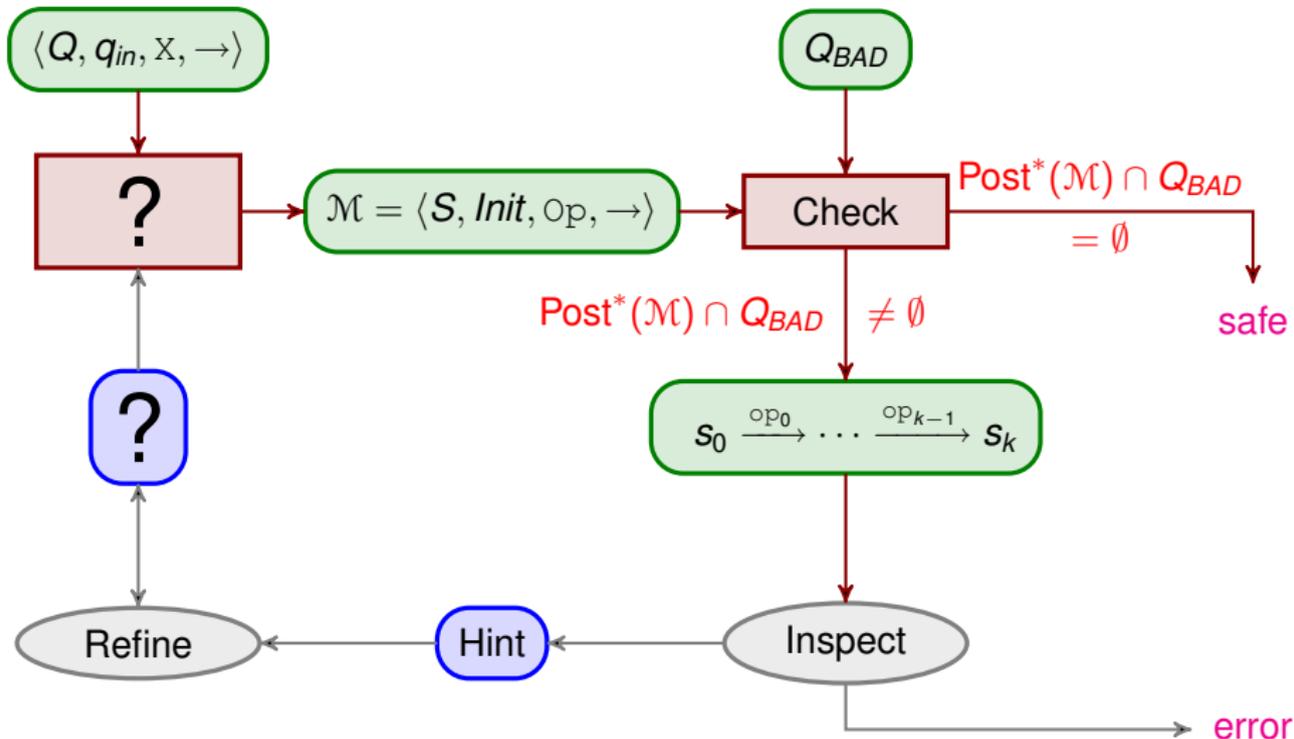
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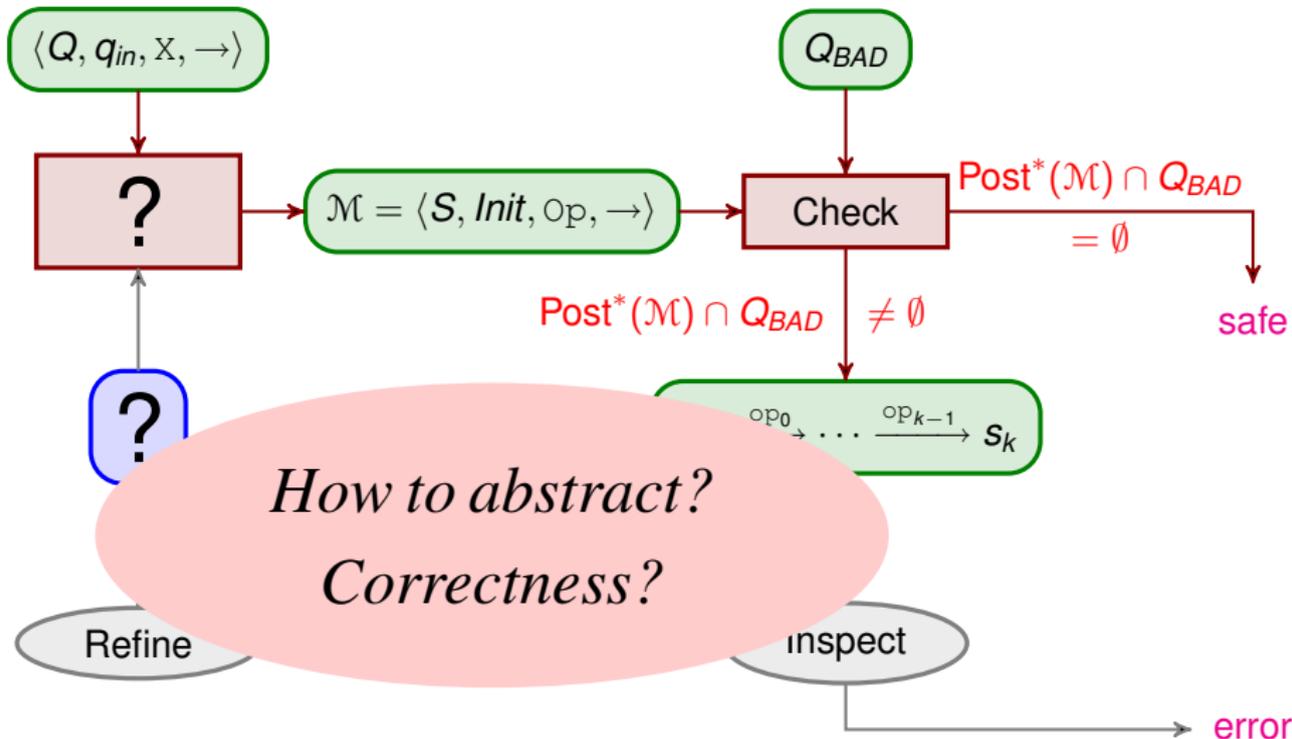
# Hypothetical Workflow Based on Abstract Runs



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# Hypothetical Workflow Based on Abstract Runs



# Verification by Model-Checking Abstract Models

This hypothetical workflow...

... is not hypothetical at all!

## Automatic Generation of Property-Preserving Abstractions

- First designed for large finite-state concurrent systems
- Inspired from abstract interpretation (use of Galois connections)
- Extended to (infinite-state) programs with theorem provers

## Credits: Pioneers (1990's)

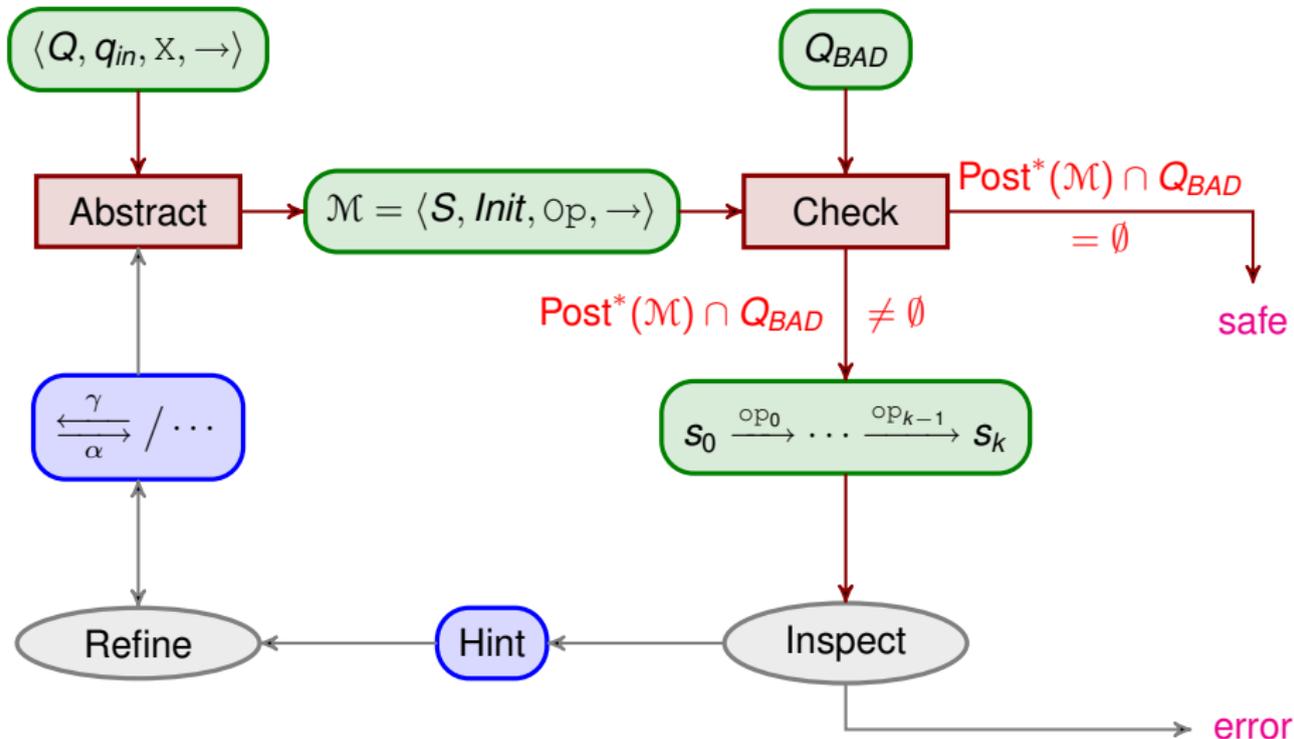
Joseph Sifakis & Claire Loiseaux

Dennis Dams & Rob Gerth & Orna Grumberg

Susanne Graf & Hassen Saïdi

...

# Verification by Model-Checking Abstract Models



# Automatic Inspection and Refinement: a Dream?

## Goal

Automate the tasks **Inspect** and **Refine**

## Counterexample Guided Refinement (2000)

- First designed for large finite-state systems (hardware)
- Extended to (infinite-state) programs with theorem provers
- Subject of active research

## Credits: Pioneers (2000)

Edmund Clarke     &     Orna Grumberg

Thomas Ball     &     Sriram Rajamani

...

# Summary and Outlook: Key Ingredients

## Property-Preserving Abstraction

Conservatively extract finite-state models from programs

## Model-Checking

Can use a readily available finite-state model checker 😊

## Inspection of Abstract Counterexamples

Reduces to satisfiability checking (use of theorem provers)

## Refinement Guided by Abstract Counterexamples

Driven by the safety property to check: precision where required

Monotonic: the model after refinement has less counterexamples

All these tasks can be automated 😊

15 Introduction and Overview

**16 Basic Theory on Property-Preserving Abstractions**

17 Abstraction Schemes

18 Counterexample Guided Refinement

# Objectives of the Basic Theory

## Property-Preserving Abstraction

Conservatively extract finite-state models from programs

We focus on safety properties

### Model

Labeled Kripke Structure

=

LTS + Bad

### Notations

Concrete LKS:  $\mathcal{M}^c$

Abstract LKS:  $\mathcal{M}^a$

## Theory Intentionally Limited (Only What We Need...)

- Notions of abstraction and refinement (simpler than  $\begin{matrix} \xleftarrow{\gamma} \\ \xrightarrow{\alpha} \end{matrix}$  😊)
- Theorem for preservation of safety

## Definition

A **labeled Kripke structure** is a quintuple  $\langle S, Init, Bad, \Sigma, \rightarrow \rangle$  where :

- $S$  is a set of *states*
- $Init \subseteq S$  is a set of *initial states*
- $Bad \subseteq S$  is a set of *bad states*
- $\Sigma$  is a finite set of *actions*
- $\rightarrow \subseteq S \times \Sigma \times S$  is a set of *transitions*

## Simplified Definition!

Kripke structures are classically defined with a mapping from  $S$  to  $\mathcal{P}(AP)$  where  $AP$  is a finite set of atomic propositions.

In our context  $AP = \{bad\}$ , hence it suffices to take  $Bad \subseteq S$ .

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## Labeled Transition System

$$\langle C, \text{Init}, \Sigma, \rightarrow \rangle$$

Elements of  $C$  are called **configurations**.

Use: **concrete** operational semantics of control flow automata.

## Labeled Kripke Structures

$$\begin{aligned} \mathcal{M} &= \langle S, \text{Init}, \text{Bad}, \Sigma, \rightarrow \rangle \\ &= \text{LTS} + \text{Bad} \end{aligned}$$

Elements of  $S$  are called **states**.

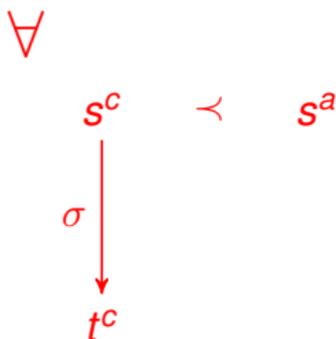
Use: models (in general **abstract ones**) for abstraction refinement.

# Simulation Relation: Definition

Consider two labeled Kripke structures:

$$\mathcal{M}^c = \langle S^c, Init^c, Bad^c, \Sigma, \rightarrow^c \rangle \quad \mathcal{M}^a = \langle S^a, Init^a, Bad^a, \Sigma, \rightarrow^a \rangle$$

A **simulation relation** from  $\mathcal{M}^c$  to  $\mathcal{M}^a$  is any binary relation  $\prec \subseteq S^c \times S^a$  satisfying:

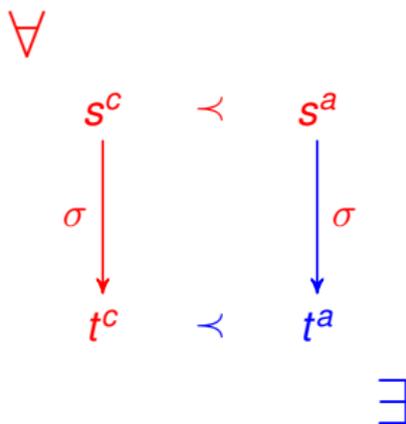


# Simulation Relation: Definition

Consider two labeled Kripke structures:

$$\mathcal{M}^c = \langle S^c, Init^c, Bad^c, \Sigma, \rightarrow^c \rangle \quad \mathcal{M}^a = \langle S^a, Init^a, Bad^a, \Sigma, \rightarrow^a \rangle$$

A **simulation relation** from  $\mathcal{M}^c$  to  $\mathcal{M}^a$  is any binary relation  $\prec \subseteq S^c \times S^a$  satisfying:



# Abstraction and Refinement

Consider two labeled Kripke structures:

$$\mathcal{M}^c = \langle S^c, Init^c, Bad^c, \Sigma, \rightarrow^c \rangle \quad \mathcal{M}^a = \langle S^a, Init^a, Bad^a, \Sigma, \rightarrow^a \rangle$$

If there exists a simulation relation  $\prec$  from  $\mathcal{M}^c$  to  $\mathcal{M}^a$  such that

$$\begin{cases} \forall s^c \in Init^c \cdot \exists s^a \in Init^a \cdot s^c \prec s^a \\ \forall (s^c, s^a) \in \prec \cdot s^c \in Bad^c \implies s^a \in Bad^a \end{cases}$$

then we say that:

$\mathcal{M}^a$  is an **abstraction** of  $\mathcal{M}^c$

$\mathcal{M}^c$  is a **refinement** of  $\mathcal{M}^a$

# Preservation of Safety Properties

A labeled Kripke structure  $\mathcal{M} = \langle S, Init, Bad, \Sigma, \rightarrow \rangle$  is **safe** if it contains **no** path

$$s_0 \xrightarrow{\sigma_0} s_1 \cdots s_{k-1} \xrightarrow{\sigma_k} s_k \quad \text{with} \quad \begin{cases} s_0 \in Init \\ s_k \in Bad \end{cases}$$

## Theorem (Safety Preservation)

*For any two labeled Kripke structures  $\mathcal{M}^c$  and  $\mathcal{M}^a$ ,*

*if  $\mathcal{M}^a$  is an abstraction of  $\mathcal{M}^c$  and  $\mathcal{M}^a$  is safe then  $\mathcal{M}^c$  is safe.*

The converse does not hold.

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The converse does not hold.

# Preservation of Safety Properties: Application

We want to show that a concrete labeled Kripke structure  $\mathcal{M}^c$  is safe.

If  $\mathcal{M}^c$  cannot be directly model-checked then:

- 1 design an abstract labeled Kripke structure  $\mathcal{M}^a$ , simpler than  $\mathcal{M}^c$ , and exhibit a simulation relation  $\prec$  that shows that  $\mathcal{M}^a$  is an abstraction of  $\mathcal{M}^c$ .
- 2 check that  $\mathcal{M}^a$  is safe

If  $\mathcal{M}^a$  is safe then  $\mathcal{M}^c$  is safe

However, If  $\mathcal{M}^a$  is not safe then we cannot conclude that  $\mathcal{M}^c$  is not safe.

# Preservation of Safety Properties: Completeness

Consider a labeled Kripke structure  $\mathcal{M}^c = \langle S^c, Init^c, Bad^c, \Sigma, \rightarrow^c \rangle$ .

$$(\mathcal{M}^a) \quad \begin{array}{ll} S^a = \{reach\} & Init^a = \{reach\} \\ \rightarrow^a = \{reach\} \times \Sigma \times \{reach\} & Bad^a = \emptyset \end{array}$$

The relation  $\prec = Post^*(\mathcal{M}^c) \times \{reach\}$  is obviously a simulation relation from  $\mathcal{M}^c$  to  $\mathcal{M}^a$ . Note that  $\mathcal{M}^a$  is safe. Moreover:

$$\text{if } \mathcal{M}^c \text{ is safe then } \begin{cases} \forall s^c \in Init^c \cdot \exists s^a \in Init^a \cdot s^c \prec s^a \\ \forall (s^c, s^a) \in \prec \cdot s^c \in Bad^c \implies s^a \in Bad^a \end{cases}$$

## Theorem (Relative Completeness)

*For any safe labeled Kripke structure  $\mathcal{M}^c$ , there exists a finite-state abstraction  $\mathcal{M}^a$  of  $\mathcal{M}^c$  such that  $\mathcal{M}^a$  is safe.*

Finite-state abstractions are sufficient to prove safety of any model.

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**17 Abstraction Schemes**

- Partition Abstraction
- Boolean Predicate Abstraction
- Cartesian Predicate Abstraction
- Application of Predicate Abstraction to Programs
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Presentation of abstraction schemes at the **Semantic Level**

**Forget about control flow automata and programs**

But keep them in mind for intuitions 😊

Implementation of predicate abstraction for control flow automata

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Implementation of predicate abstraction for control flow automata

# Partition Abstraction: Definition

Consider a labeled Kripke structure  $\mathcal{M}^c = \langle S^c, \text{Init}^c, \text{Bad}^c, \Sigma, \rightarrow^c \rangle$ .

Partition given by  $(S^a, \alpha)$  where  $S^a$  is a finite set and  $\alpha : S^c \rightarrow S^a$

## Partition Abstraction $\mathcal{M}^a$ Induced by $(S^a, \alpha)$

$$\text{Init}^a = \{\alpha(s^c) \mid s^c \in \text{Init}^c\}$$

$$\text{Bad}^a = \{\alpha(s^c) \mid s^c \in \text{Bad}^c\}$$

$$\rightarrow^a = \{(\alpha(s^c), \sigma, \alpha(t^c)) \mid (s^c, \sigma, t^c) \in \rightarrow^c\}$$

The simulation relation  $\prec = \{(s^c, \alpha(s^c)) \mid s^c \in S^c\}$  shows that

$\mathcal{M}^a$  is an abstraction of  $\mathcal{M}^c$

# Partition Abstraction: Explanation

## Partition $(S^a, \alpha)$

- $S^a$  finite set
- $\alpha : S^c \rightarrow S^a$

## Partition Abstraction Induced by $(S^a, \alpha)$

$$\begin{aligned} Init^a &= \{\alpha(s^c) \mid s^c \in Init^c\} && (Bad^a \dots) \\ \rightarrow^a &= \{(\alpha(s^c), \sigma, \alpha(t^c)) \mid (s^c, \sigma, t^c) \in \rightarrow^c\} \end{aligned}$$

Induced equivalence relation  $\sim$  defined by:  $s^c \sim t^c$  if  $\alpha(s^c) = \alpha(t^c)$ .

## Abstraction Function $\alpha : S^c \rightarrow S^a$

All concrete states in an equivalence class are merged together.

## Induced Concretization Function $\gamma : S^a \rightarrow \mathcal{P}(S^c)$

$$\gamma(s^a) = \{s^c \mid \alpha(s^c) = s^a\}$$

## Not a Galois Connection

$(\alpha, \gamma)$  becomes a Galois Connection when lifted to powersets.

# Partition Abstraction: Explanation

## Partition $(S^a, \alpha)$

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$Init^a$ ,  $Bad^a$  and  $\rightarrow^a$  are **existential** lifts of their concrete counterparts:

$$s^a \in Init^a \quad \text{iff} \quad \exists s^c . \begin{cases} \alpha(s^c) = s^a & \wedge \\ s^c \in Init^c \end{cases}$$

$$(s^a, \sigma, t^a) \in \rightarrow^a \quad \text{iff} \quad \exists s^c \exists t^c . \begin{cases} \alpha(s^c) = s^a & \wedge \\ \alpha(t^c) = t^a & \wedge \\ (s^c, \sigma, t^c) \in \rightarrow^c \end{cases}$$

# Partition Abstraction: Computation of $\mathcal{M}^a$

Consider a labeled Kripke structure  $\mathcal{M}^c = \langle S^c, Init^c, Bad^c, \Sigma, \rightarrow^c \rangle$ .

## Computation of $Init^a$

```
I ← ∅  
foreach  $s^a \in S^a$   
  if  $\exists s^c \cdot (s^c \in \gamma(s^a) \wedge s^c \in Init^c)$   
    I ← I ∪ { $s^a$ }  
return I
```

## Computation of $\rightarrow^a$

```
R ← ∅  
foreach  $(s^a, \sigma, t^a) \in S^a \times \Sigma \times S^a$   
  if  $\exists s^c \exists t^c \cdot (s^c \in \gamma(s^a) \wedge t^c \in \gamma(t^a) \wedge (s^c, \sigma, t^c) \in \rightarrow^c)$   
    R ← R ∪ {( $s^a, \sigma, t^a$ )}  
return R
```

# Partition Abstraction: Implementation Issues

- Machine **representation** of  $\alpha : S^c \rightarrow S^a$  or  $\gamma : S^a \rightarrow \mathcal{P}(S^c)$ 
  - Examples: BDDs (if  $S^c = \{0, 1\}^n$ ), NDDs (if  $S^c = \mathbb{Z}^n$ ), ...
- Algorithms to **decide** the conditions

$$\exists s^c \cdot (s^c \in \gamma(s^a) \wedge s^c \in \mathit{Init}^c)$$

$$\exists s^c \cdot (s^c \in \gamma(s^a) \wedge s^c \in \mathit{Bad}^c)$$

$$\exists s^c \exists t^c \cdot (s^c \in \gamma(s^a) \wedge t^c \in \gamma(t^a) \wedge (s^c, \sigma, t^c) \in \rightarrow^c)$$

## Partial Algorithms (yes / no / ?) Are Sufficient

Safety preservation from  $\mathcal{M}^a$  to  $\mathcal{M}^c$  still holds if  $\mathit{Init}^a$ ,  $\mathit{Bad}^a$  and  $\rightarrow^a$  are larger than the “optimal ones”. We may soundly consider “?” as “yes”.

# Partition Abstraction: Refinement

Given two equivalence relations  $\sim_1$  and  $\sim_2$  on some set  $S$ , we say that  $\sim_2$  is **finer** than  $\sim_1$  if  $\sim_2 \subseteq \sim_1$ , or equivalently if each equivalence class of  $\sim_1$  is a union of equivalence classes of  $\sim_2$ .

Consider two partitions  $(S_1^a, \alpha_1)$  and  $(S_2^a, \alpha_2)$ .

If  $\sim_2$  is finer than  $\sim_1$  then  $\mathcal{M}^a(S_2^a, \alpha_2)$  is a refinement of  $\mathcal{M}^a(S_1^a, \alpha_1)$ .

## Informally

To refine a partition abstraction, split some equivalence classes.

## Recomputation of $\mathcal{M}^a$ after refinement

- Refinement is **local** to equivalence classes that are split.
- If  $\mathcal{M}^a$  is stored explicitly then the refined  $\rightarrow^a$  can be efficiently computed from the previous  $\rightarrow^a$ .

## Predicates

Formulas in **first-order logic** over some vocabulary

## Example

For control flow automata, take the same vocabulary as in expressions:

$$\langle \dots, -1, 0, 1, \dots ; +, -, * ; <, \leq, =, \neq, \geq, > \rangle$$

At the **semantic level**, we view predicates as **sets of states**.

## Predicates

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At the **semantic level**, we view predicates as **sets of states**.

# Boolean Predicate Abstraction: Definition

Consider a labeled Kripke structure  $\mathcal{M}^c = \langle S^c, \text{Init}^c, \text{Bad}^c, \Sigma, \rightarrow^c \rangle$ .

Support predicates given by a finite set  $\Phi$  of subsets of  $S^c$

## Characteristic Function of $\phi \in \Phi$

$$\mathbf{1}_\phi : S^c \rightarrow \{0, 1\}$$
$$s^c \mapsto \begin{cases} 1 & \text{if } s^c \in \phi \\ 0 & \text{if } s^c \notin \phi \end{cases}$$

## Partition $(S_\Phi^a, \alpha_\Phi)$

$$S_\Phi^a = \Phi \rightarrow \{0, 1\}$$
$$\alpha_\Phi(s^c) = \lambda \phi. \mathbf{1}_\phi(s^c)$$

## Boolean Predicate Abstraction $\mathcal{M}^a$ Induced by $\Phi$

Partition abstraction induced by the partition  $(S_\Phi^a, \alpha_\Phi)$

# Boolean Predicate Abstraction: Explanation

## Partition ( $S_\phi^a, \alpha_\phi$ )

$$S_\phi^a = \phi \rightarrow \{0, 1\}$$

$$\alpha_\phi(s^c) = \lambda \phi. \mathbf{1}_\phi(s^c)$$

## Intuition

Abstract state: truth value for each predicate

$\alpha_\phi$  merges concrete states that satisfy the same predicates.

## Induced Concretization Function $\gamma : S^a \rightarrow \mathcal{P}(S^c)$

$$\gamma_\phi(s^a) = \bigcap_{s^a(\phi)=1} \phi \cap \bigcap_{s^a(\phi)=0} S^c \setminus \phi$$

## Not a Galois Connection

$(\alpha_\phi, \gamma_\phi)$  becomes a Galois Connection when lifted to powersets.

# Boolean Predicate Abstraction: Computation of $\mathcal{M}^a$

$Init^a$ ,  $Bad^a$  and  $\rightarrow^a$  can be computed as for partition abstractions, but:

## Exponential complexity

Number of abstract states:  $2^{|\Phi|}$

Exponential number of decisions  $\exists s^c \exists t^c \cdot (\dots)$  to compute  $\rightarrow^a$

Exploit the structure of the partition to get better algorithms (in practice)

Computation of  $\alpha(U) = \{\alpha(s^c) \mid s^c \in U\}$  where  $U \subseteq S^c$

If  $U \subseteq \phi$  then every  $s^a \in \alpha(U)$  necessarily satisfies  $s^a(\phi) = 1$ .

In that case, there is no need to examine candidates where  $s^a(\phi) = 0$ .

$$\Phi_1 = \{\phi \in \Phi \mid U \subseteq \phi\}$$

$$\Phi_0 = \{\phi \in \Phi \mid U \subseteq S^c \setminus \phi\}$$

New complexity linear in  $|\Phi_0| + |\Phi_1|$  and exponential in  $|\Phi \setminus (\Phi_0 \cup \Phi_1)|$

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New complexity linear in  $|\Phi_0| + |\Phi_1|$  and exponential in  $|\Phi \setminus (\Phi_0 \cup \Phi_1)|$

# Boolean Predicate Abstraction: Implementation Issues

Each abstract state is a truth valuation of the predicates.

Sets of abstract states (e.g.  $Init^a$ ,  $Bad^a$ ) are sets of truth valuations.

## Natural Encoding

### Propositional Formulas

Introduce propositional variables  $p_\phi, p'_\phi$  for each predicate  $\phi$ .

$$s^a \rightsquigarrow \bigwedge_{\phi \in \Phi} \overline{p_\phi} \quad (\text{conjunction of literals})$$

$$Init^a, Bad^a \rightsquigarrow \bigvee_{\phi \in \Phi} \bigwedge \overline{p_\phi} \quad (\text{formula on } p_\phi)$$

$$\rightarrow^a \rightsquigarrow \bigvee_{\phi \in \Phi} \bigwedge \overline{p_\phi} \bigwedge_{\phi \in \Phi} \overline{p'_\phi} \quad (\text{formula on } p_\phi, p'_\phi)$$

Use BDDs to represent these propositional formulas ☺

# Boolean Predicate Abstraction: Implementation Issues

Each abstract state is a truth valuation of the predicates.

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## Natural Encoding

### Propositional Formulas

Introduce propositional variables  $p_\phi, p'_\phi$  for each predicate  $\phi$ .

$$\begin{aligned} s^a &\iff \bigwedge_{\phi \in \Phi} \overline{p_\phi} && \text{(conjunction of literals)} \\ Init^a, Bad^a &\iff \bigvee_{\phi \in \Phi} \bigwedge_{\phi \in \Phi} \overline{p_\phi} && \text{(formula on } p_\phi) \\ \rightarrow^a &\iff \bigvee_{\phi \in \Phi} \bigwedge_{\phi \in \Phi} \overline{p_\phi} \bigwedge_{\phi \in \Phi} \overline{p'_\phi} && \text{(formula on } p_\phi, p'_\phi) \end{aligned}$$

Use BDDs to represent these propositional formulas 😊

If  $\Phi_2 \supseteq \Phi_1$  then  $\mathcal{M}^a(\Phi_2)$  is a refinement of  $\mathcal{M}^a(\Phi_1)$ .

## Informally

To refine a boolean predicate abstraction, add new predicates.

## Recomputation of $\mathcal{M}^a$ after refinement

- Refinement is **global**, since it can impact all abstract states.

# Cartesian Predicate Abstraction: Introduction

Support predicates given by a finite set  $\Phi$  of subsets of  $S^c$

## Objective

Avoid exponential cost in the abstraction of a set  $U$  of concrete states

A **monomial** is a conjunction of literals  $\bigwedge_{\phi \in \Phi'} \overline{p_\phi}$  for some  $\Phi' \subseteq \Phi$ .

## Solution

Replace disjunctions of abstract states by the **most precise monomial**.

$$\text{Boolean: } U \subseteq S^c \xrightarrow{\alpha} \bigvee \bigwedge_{\phi \in \Phi} \overline{p_\phi}$$

$$\text{Cartesian: } U \subseteq S^c \xrightarrow{\alpha} \bigwedge_{\phi \in \Phi'} \overline{p_\phi} \quad (\Phi' \subseteq \Phi)$$

# Cartesian Predicate Abstraction: Trivectors

## Encoding of Monomials

Encode  $\bigwedge_{\phi \in \Phi'} \overline{p_\phi}$  as the valuation

$$v(\phi) = \begin{cases} 1 & \text{if } \overline{p_\phi} = p_\phi \\ 0 & \text{if } \overline{p_\phi} = \neg p_\phi \\ * & \text{if } \phi \notin \Phi' \end{cases}$$

## 3-Valued Characteristic Function

$$\mathbf{1}_\phi : \mathcal{P}(S^c) \rightarrow \{0, 1, *\}$$

$$U \neq \emptyset \mapsto \begin{cases} 1 & \text{if } U \subseteq \phi \\ 0 & \text{if } U \subseteq S^c \setminus \phi \\ * & \text{otherwise} \end{cases}$$

## Cartesian Abstraction and Concretization Functions

$$S_\Phi^a = \Phi \rightarrow \{0, 1, *\}$$

$$\alpha_\Phi(U) = \lambda \phi. \mathbf{1}_\phi(U) \quad (U \neq \emptyset)$$

$$\gamma_\Phi(s^a) = \bigcap_{s^a(\phi)=1} \phi \cap \bigcap_{s^a(\phi)=0} S^c \setminus \phi$$

# Cartesian Predicate Abstraction: Definition

## Notation: Concrete Post Operator

$$\text{Post}^c(U, \sigma) = \{t^c \in \text{State}^c \mid \exists s^c \in U \cdot (s^c, \sigma, t^c) \in \rightarrow^c\}$$

## Cartesian Predicate Abstraction $\mathcal{M}^a$ Induced by $\Phi$

$$S_\Phi^a = \Phi \rightarrow \{0, 1, *\}$$

$$\text{Init}^a = \{\alpha_\Phi(s^c) \mid s^c \in \text{Init}^c\}$$

$$\text{Bad}^a = \{s^a \mid s^a \in S^a, \gamma_\Phi(s^a) \cap \text{Bad}^c \neq \emptyset\}$$

$$\rightarrow^a = \{(s^a, \sigma, \alpha_\Phi \circ \text{Post}^c(\gamma_\Phi(s^a), \sigma)) \mid s^a \in S^a, \sigma \in \Sigma\}$$

The simulation relation  $\prec = \{(s^c, s^a) \mid s^c \in \gamma_\Phi(s^a)\}$  shows that

$\mathcal{M}^a$  is an abstraction of  $\mathcal{M}^c$

# Cartesian Predicate Abstraction: Remarks

## Cartesian Predicate Abstraction $\mathcal{M}^a$ Induced by $\Phi$

$$S_{\Phi}^a = \Phi \rightarrow \{0, 1, *\}$$

$$\gamma_{\Phi}(s^a) = \bigcap_{s^a(\phi)=1} \phi \cap \bigcap_{s^a(\phi)=0} S^c \setminus \phi$$

$$\rightarrow^a = \{(s^a, \sigma, \alpha_{\Phi} \circ \text{Post}^c \circ \gamma_{\Phi}(s^a)) \mid s^a \in S^a, \sigma \in \Sigma\}$$

Abstract state: truth value in  $\{0, 1, *\}$  for each  $\phi \in \Phi$ . **Not a partition!**

The special value  $*$  is conservatively treated as “don’t know” in  $\gamma_{\Phi}$ .

The transition relation  $\rightarrow^a$  is **deterministic** (at most one successor).

## Galois Connection

$(\alpha_{\Phi}, \gamma_{\Phi})$  is a Galois Connection (with  $0, 1 \sqsubseteq *$ ).

# Cartesian Predicate Abstraction: Computation of $\mathcal{M}^a$

## Computation of $Init^a$

Same as boolean case

## Computation of $\alpha(U)$

```
foreach  $\phi \in \Phi$ 
  if  $U \subseteq S^c \setminus \phi$ 
     $s^a[\phi] \leftarrow 0$ 
  else if  $U \subseteq \phi$ 
     $s^a[\phi] \leftarrow 1$ 
  else
     $s^a[\phi] \leftarrow *$ 
return  $s^a$ 
```

## Computation of $\rightarrow^a$

```
R  $\leftarrow \emptyset$ 
foreach  $(s^a, \sigma) \in S^a \times \Sigma$  |  $Post^c(\gamma(s^a), \sigma) \neq \emptyset$ 
  foreach  $\phi \in \Phi$ 
    if  $Post^c(\gamma(s^a), \sigma) \subseteq S^c \setminus \phi$ 
       $t^a[\phi] \leftarrow 0$ 
    else if  $Post^c(\gamma(s^a), \sigma) \subseteq \phi$ 
       $t^a[\phi] \leftarrow 1$ 
    else
       $t^a[\phi] \leftarrow *$ 
  R  $\leftarrow R \cup \{(s^a, \sigma, t^a)\}$ 
return R
```

Linear number of decisions  $Post^c(\gamma(s^a), \sigma) \subseteq \dots$  to compute the successor  $\rightarrow^a(s^a, \sigma)$  of a given abstract state  $s^a$  and action  $\sigma \in \Sigma$ .

Similar to boolean predicate abstraction:

- Encoding with 3-valued propositional variables  $p_\phi, p'_\phi$
- Representation with TDDs (or BDDs via binary encoding)

For concrete labeled Kripke structures obtained from programs, the cartesian predicate abstraction can be presented as boolean program.

# Cartesian Predicate Abstraction: Refinement

If  $\Phi_2 \supseteq \Phi_1$  then  $\mathcal{M}^a(\Phi_2)$  is a refinement of  $\mathcal{M}^a(\Phi_1)$ .

## Informally

To refine a cartesian predicate abstraction, add new predicates.

## Recomputation of $\mathcal{M}^a$ after refinement

- Refinement is **global**, since it can impact all abstract states.

# How about Programs?

Control flow automaton:  $\langle Q, q_{in}, X, \rightarrow \rangle$ . Set  $Q_{BAD} \subseteq Q$  of bad locations.

## Concrete Labeled Kripke Structure $\mathcal{M}^c$

$$S^c = Q \times (X \rightarrow \mathbb{R})$$

$$Init^c = \{q_{in}\} \times (X \rightarrow \mathbb{R})$$

$$\Sigma = \text{Op}$$

$$Bad^c = Q_{BAD} \times (X \rightarrow \mathbb{R})$$

$$\rightarrow^c = \left\{ ((q, u^c), \sigma, (q', v^c)) \mid q \xrightarrow{\text{op}} q' \text{ and } (u^c, v^c) \in \llbracket \text{op} \rrbracket^c \right\}$$

The usual semantics  $\llbracket \text{op} \rrbracket$  of operations is now written  $\llbracket \text{op} \rrbracket^c$ .

## Nothing Surprising Here!

This is the usual labeled transition system (operational semantics of control flow automata) equipped with the usual bad configurations.

# Predicate Language

Control flow automaton:  $\langle Q, q_{in}, X, \rightarrow \rangle$ .

## Vocabulary

$\langle \dots, -1, 0, 1, \dots ; +, -, * ; <, \leq, =, \neq, \geq, > \rangle$

Additive and multiplicative theory of the reals is decidable.

## Finite Set $\phi$ of Support Predicates

(Quantifier-free) first-order formulas with **free variables in  $x$**

## Semantics of Support Predicates

The interpretation  $\llbracket \varphi \rrbracket$  of a predicate  $\varphi$  is a subset of  $X \rightarrow \mathbb{R}$ .

## Link With Semantic Level Abstraction Schemes

The interpretations  $\llbracket \varphi \rrbracket$  replace the “semantic support predicates”  $\phi$ .

## Boolean Predicate Abstraction $\mathcal{M}^a$ Induced by $\Phi$

$$S^a = Q \times (\Phi \rightarrow \{0, 1\}) \qquad \text{Init}^a = \{q_{in}\} \times (\Phi \rightarrow \{0, 1\})$$

$$\Sigma = \text{Op} \qquad \text{Bad}^a = Q_{BAD} \times (\Phi \rightarrow \{0, 1\})$$

$$\rightarrow^a = \left\{ ((q, u^a), \sigma, (q', v^a)) \mid q \xrightarrow{\text{op}} q' \text{ and } (u^a, v^a) \in \llbracket \text{op} \rrbracket^a \right\}$$

- Concrete valuations in  $X \rightarrow \mathbb{R}$  are replaced by abstract valuations in  $\Phi \rightarrow \{0, 1\}$ .
- The control flow automaton's graph is kept intact.
- All the work is done in the abstract semantics of operations.

## Syntactic Concretization

Concretization formula  $\gamma(v^a)$  of a valuation  $v^a \in \Phi \rightarrow \{0, 1\}$  defined by

$$\gamma(v^a) = \bigwedge_{\substack{\varphi \in \Phi \\ v^a(\varphi)=1}} \varphi \quad \wedge \quad \bigwedge_{\substack{\varphi \in \Phi \\ v^a(\varphi)=0}} \neg\varphi$$

Abstract semantics  $\llbracket \text{op} \rrbracket^a$  of operations defined as a binary relation

$$\llbracket \text{op} \rrbracket^a \subseteq (\Phi \rightarrow \{0, 1\}) \times (\Phi \rightarrow \{0, 1\})$$

Guards:  $(u^a, v^a) \in \llbracket g \rrbracket^a$  if  $v^a = u^a$  and  $\gamma(u^a) \wedge g$  sat.

Assignments:  $(u^a, v^a) \in \llbracket x := e \rrbracket^a$  if  $\gamma(u^a) \wedge \gamma(v^a)[e/x]$  sat.

# Boolean Predicate Abstraction: Computation of $\mathcal{M}^a$

Safety checking of  $\mathcal{M}^a$  usually performed by forward graph exploration.

Computation of  $\{v^a \in \Phi \rightarrow \{0, 1\} \mid (u^a, v^a) \in \llbracket g \rrbracket^a\}$

```
if  $\gamma(u^a) \wedge g$  is satisfiable
  return  $\{u^a\}$ 
else
  return  $\emptyset$ 
```

Computation of  $\{v^a \in \Phi \rightarrow \{0, 1\} \mid (u^a, v^a) \in \llbracket x := e \rrbracket^a\}$

```
S  $\leftarrow \emptyset$ 
foreach  $v^a \in \Phi \rightarrow \{0, 1\}$  (exponential ☹)
  if  $\gamma(u^a) \wedge \gamma(v^a)[e/x]$  is satisfiable
    S  $\leftarrow S \cup \{v^a\}$ 
return S
```

# Cartesian Predicate Abstraction: Definition

## Cartesian Predicate Abstraction $\mathcal{M}^a$ Induced by $\Phi$

$$S^a = Q \times (\Phi \rightarrow \{0, 1, *\}) \quad \text{Init}^a = \{q_{in}\} \times \{\lambda \varphi . *\}$$

$$\Sigma = \text{Op} \quad \text{Bad}^a = Q_{BAD} \times \{\lambda \varphi . *\}$$

$$\rightarrow^a = \left\{ ((q, u^a), \sigma, (q', v^a)) \mid q \xrightarrow{\text{op}} q' \text{ and } v^a = \llbracket \text{op} \rrbracket^a(u^a) \right\}$$

## Syntactic Concretization

Concretization formula  $\gamma(v^a)$  of a valuation  $v^a \in \Phi \rightarrow \{0, 1, *\}$  defined by

$$\gamma(v^a) = \bigwedge_{\substack{\varphi \in \Phi \\ v^a(\varphi)=1}} \varphi \quad \wedge \quad \bigwedge_{\substack{\varphi \in \Phi \\ v^a(\varphi)=0}} \neg \varphi$$

# Cartesian Predicate Abstraction: Definition

Abstract semantics  $\llbracket \text{op} \rrbracket^a$  of operations defined as a partial function

$$\llbracket \text{op} \rrbracket^a : (\Phi \rightarrow \{0, 1, *\}) \rightarrow (\Phi \rightarrow \{0, 1, *\})$$

## Guards

If  $\gamma(u^a) \wedge g$  is unsatisfiable then  $\llbracket \text{op} \rrbracket^a(u^a)$  is undefined

$$\text{Otherwise } \llbracket \text{op} \rrbracket^a(u^a) = \lambda \varphi. \begin{cases} 0 & \text{if } (\gamma(u^a) \wedge g) \Rightarrow \neg \varphi \text{ is valid} \\ 1 & \text{if } (\gamma(u^a) \wedge g) \Rightarrow \varphi \text{ is valid} \\ * & \text{otherwise} \end{cases}$$

## Assignments

If  $\gamma(u^a)$  is unsatisfiable then  $\llbracket x := e \rrbracket^a(u^a)$  is undefined

$$\text{Otherwise } \llbracket x := e \rrbracket^a(u^a) = \lambda \varphi. \begin{cases} 0 & \text{if } \gamma(u^a) \Rightarrow \neg \varphi[e/x] \text{ is valid} \\ 1 & \text{if } \gamma(u^a) \Rightarrow \varphi[e/x] \text{ is valid} \\ * & \text{otherwise} \end{cases}$$

# Cartesian Predicate Abstraction: Computation of $\mathcal{M}^a$

Safety checking of  $\mathcal{M}^a$  usually performed by forward graph exploration.

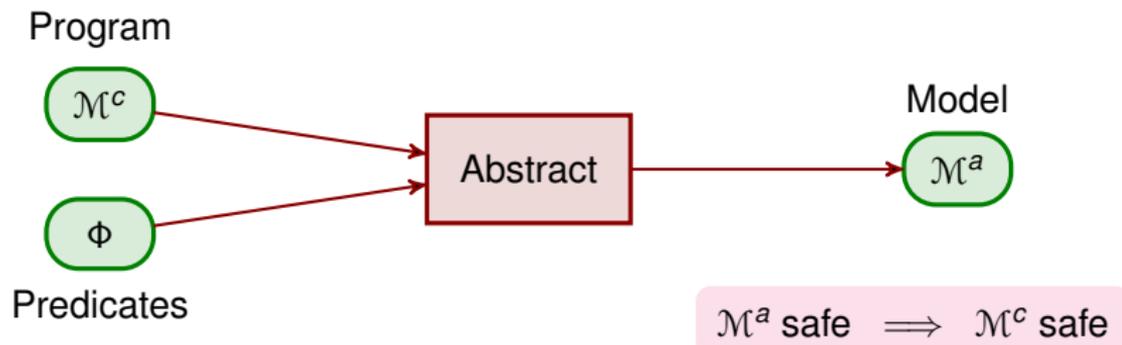
## Computation of $\llbracket g \rrbracket^a(u^a)$

```
if  $\gamma(u^a) \wedge g$  is unsatisfiable
  return undefined
foreach  $\varphi \in \Phi$  (linear ☺)
  if  $\models (\gamma(u^a) \wedge g) \Rightarrow \neg\varphi$ 
     $v^a[\varphi] \leftarrow 0$ 
  else if  $\models (\gamma(u^a) \wedge g) \Rightarrow \varphi$ 
     $v^a[\varphi] \leftarrow 1$ 
  else
     $v^a[\varphi] \leftarrow *$ 
return  $v^a$ 
```

## Computation of $\llbracket x := e \rrbracket^a(u^a)$

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if  $\gamma(u^a)$  is unsatisfiable
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  else if  $\models \gamma(u^a) \Rightarrow \varphi[e/x]$ 
     $v^a[\varphi] \leftarrow 1$ 
  else
     $v^a[\varphi] \leftarrow *$ 
return  $v^a$ 
```

# Summary: Automatic Predicate Abstraction



Refinement consists in adding new support predicates

## Boolean Abstraction

**Partition** induced by  $\Phi$

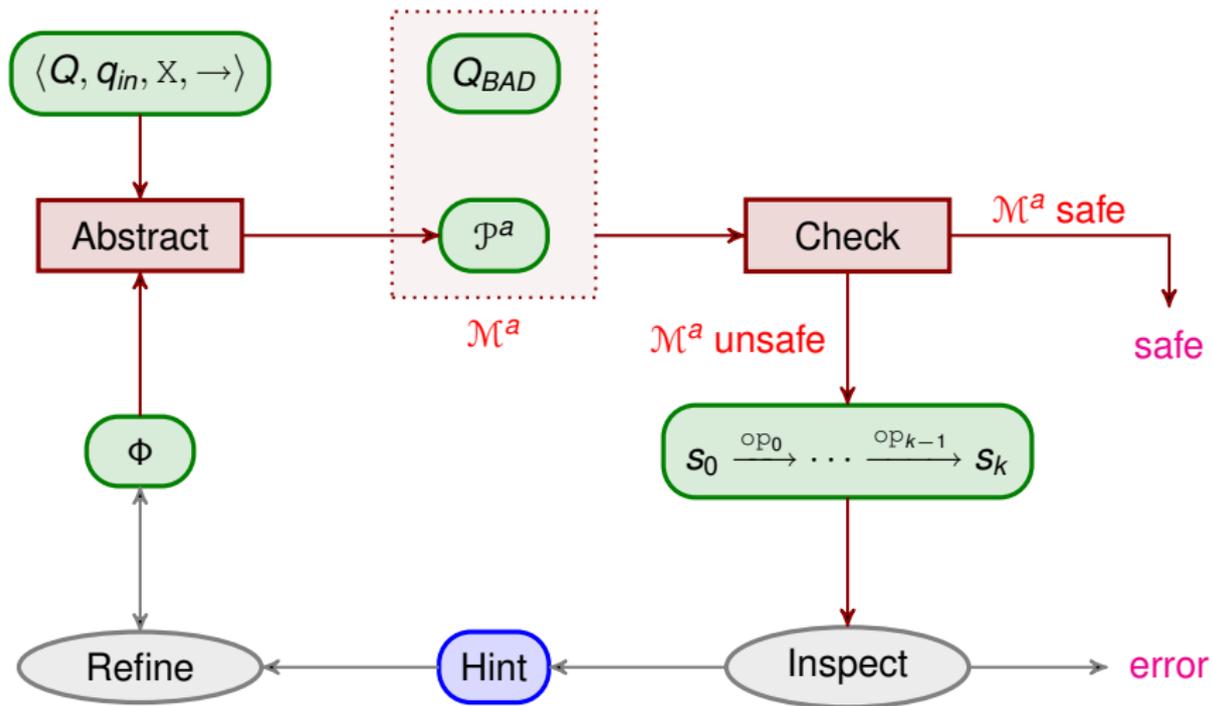
- ☺ Most precise abstraction based of  $\Phi$
- ☹ Exponential (for successors)

## Cartesian Abstraction

**Monomials** induced by  $\Phi$

- ☹ Less precise than boolean abstraction
- ☺ Linear (for successors)

# Verification by Model-Checking Abstract Models



- 15 Introduction and Overview
- 16 Basic Theory on Property-Preserving Abstractions
- 17 Abstraction Schemes
- 18 Counterexample Guided Refinement**
  - Inspection of Abstract Counterexamples
  - Counterexample Guided Refinement
  - Counterexample Guided Abstraction Refinement Algorithms

# Inspection of Abstract Counterexamples

Control flow automaton:  $\langle Q, q_{in}, X, \rightarrow \rangle$ . Set  $Q_{BAD} \subseteq Q$  of bad locations.

$\mathcal{M}^a = \langle S^a, Init^a, Bad^a, \Sigma, \rightarrow^a \rangle$  obtained by **predicate abstraction**

## Abstract counterexample

$$(q_{in}, v_0^a) \xrightarrow{op_0} (q_1, v_1^a) \cdots (q_k, v_k^a) \xrightarrow{op_k} (q_{bad}, v_{k+1}^a)$$

The abstract counterexample is **feasible** if there is a concrete run

$$(q_{in}, v_0^c) \xrightarrow{op_0} (q_1, v_1^c) \cdots (q_k, v_k^c) \xrightarrow{op_k} (q_{bad}, v_{k+1}^c) \quad \text{with } v_i^c \in \gamma(v_i^a)$$

Better to directly check for all possible abstract predicate valuations!

## Objective

Check whether a control path  $q_{in} \xrightarrow{op_0} q_1 \cdots q_k \xrightarrow{op_k} q_{bad}$  is feasible

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Better to directly check for all possible abstract predicate valuations!

## Objective

Check whether a control path  $q_{in} \xrightarrow{op_0} q_1 \cdots q_k \xrightarrow{op_k} q_{bad}$  is feasible

# Checking Feasibility of Control Paths

## Feasibility at the Semantic Level

$$q_{in} \xrightarrow{\text{op}_0} q_1 \cdots q_k \xrightarrow{\text{op}_k} q_{bad} \text{ feasible} \quad \text{iff} \quad [[\text{op}_k]] \circ \cdots \circ [[\text{op}_0]] \neq \emptyset$$

Recall that expressions  $e$  used in guards and assignments are over  $x$ .

## Syntactic Effect of Operations: Formula $\llbracket \text{op} \rrbracket$ over $x, x'$

$$\llbracket g \rrbracket = g \wedge \bigwedge_{x \in X} x' = x \quad \llbracket x := e \rrbracket = x' = e \wedge \bigwedge_{y \in X, y \neq x} y' = y$$

For each  $\text{op} \in \text{Op}$ :  $\llbracket \llbracket \text{op} \rrbracket \rrbracket = \llbracket \text{op} \rrbracket$

Multiply-primed copies of variables:  $x^{(i)}$  is the copy of  $x$  with  $i$  primes.

## Feasibility at the Syntactic Level

$$q_{in} \xrightarrow{\text{op}_0} q_1 \cdots q_k \xrightarrow{\text{op}_k} q_{bad} \text{ feasible} \quad \text{iff} \quad \llbracket \text{op}_0 \rrbracket^{(0)} \wedge \cdots \wedge \llbracket \text{op}_k \rrbracket^{(k)} \text{ sat.}$$

# Checking Feasibility of Control Paths

## Feasibility at the Semantic Level

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Number of variables grows linearly with the length of the control path.

To help the prover, we may replace  $\llbracket \text{op} \rrbracket$  with the **weakest precondition**

$$\text{wp}(\text{op}, \varphi) = \begin{cases} g \wedge \varphi & \text{if } \text{op} = g \\ \varphi[\mathbf{e}/\mathbf{x}] & \text{if } \text{op} = \mathbf{x} := \mathbf{e} \end{cases}$$

## Feasibility with Weakest Precondition

$q_{in} \xrightarrow{*} q_{bad}$  feasible iff  $\text{wp}(\text{op}_0, \text{wp}(\text{op}_1, \dots, \text{wp}(\text{op}_k, \text{true}) \cdots))$  sat.

But it might actually be better to rely on the prover's powerful engine!

# Checking Feasibility of Control Paths

## Feasibility at the Syntactic Level

$q_{in} \xrightarrow{\text{op}_0} q_1 \cdots q_k \xrightarrow{\text{op}_k} q_{bad}$  feasible iff  $\llbracket \text{op}_0 \rrbracket^{(0)} \wedge \cdots \wedge \llbracket \text{op}_k \rrbracket^{(k)}$  sat.

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## Feasibility with Weakest Precondition

$q_{in} \xrightarrow{*} q_{bad}$  feasible iff  $\text{wp}(\text{op}_0, \text{wp}(\text{op}_1, \dots, \text{wp}(\text{op}_k, \text{true}) \cdots))$  sat.

But it might actually be better to rely on the prover's powerful engine!

# Refinement Challenge: Finding Relevant Predicates

Assume that the counterexample  $q_{in} \xrightarrow{op_0} q_1 \cdots q_k \xrightarrow{op_k} q_{bad}$  is spurious

$$\langle\langle op_0 \rangle\rangle^{(0)} \wedge \cdots \wedge \langle\langle op_k \rangle\rangle^{(k)} \text{ unsatisfiable} \quad (1)$$

Refinement consists in adding new predicates, but as few as possible.

## Goal

Find predicates that remove the counterexample from the abstraction

## Practical Approach

Some conjuncts in (1) do not “participate” in unsatisfiability.

Natural idea: try to find a small unsatisfiable subset of **useful conjuncts**.

For instance pick the leaves in a **proof** of unsatisfiability.

Might or might not work...

... Let us **look back at the goal!**

# Refinement: Computation of Path Invariants

Consider an **unfeasible** control path  $q_{in} \xrightarrow{\text{op}_0} q_1 \cdots q_k \xrightarrow{\text{op}_k} q_{bad}$ .

## Path Safety Invariant

Sequence  $(\phi_i)_{0 \leq i \leq k+1}$  of subsets of  $X \rightarrow \mathbb{R}$  such that

$$\phi_0 = X \rightarrow \mathbb{R} \qquad \phi_{i+1} \supseteq \llbracket \text{op}_i \rrbracket [\phi_i] \qquad \phi_{k+1} = \emptyset$$

## Intuition

A path safety invariant gives an explanation of unfeasibility

## Example: Sequence of Reachable Valuations Along the Path

$$\phi_i = \llbracket \text{op}_{i-1} \rrbracket \circ \cdots \circ \llbracket \text{op}_0 \rrbracket [X \rightarrow \mathbb{R}]$$

## Objective

Compute **simple** path safety invariants

# Refinement: Path Safety Invariants from Proofs

Consider an **unfeasible** control path  $q_{in} \xrightarrow{\text{op}_0} q_1 \cdots q_k \xrightarrow{\text{op}_k} q_{bad}$ .

## Path Safety Invariant (Syntactic Definition)

Sequence  $(\varphi_i)_{0 \leq i \leq k+1}$  of formulas with free variables in  $X$  such that

$$\varphi_0 = \text{true} \quad \models \varphi_i \wedge \langle\langle \text{op}_i \rangle\rangle \Rightarrow \varphi_{i+1}^{(1)} \quad \varphi_{k+1} = \text{false}$$

Path safety **invariants** can be obtained **from proofs** of unsatisfiability

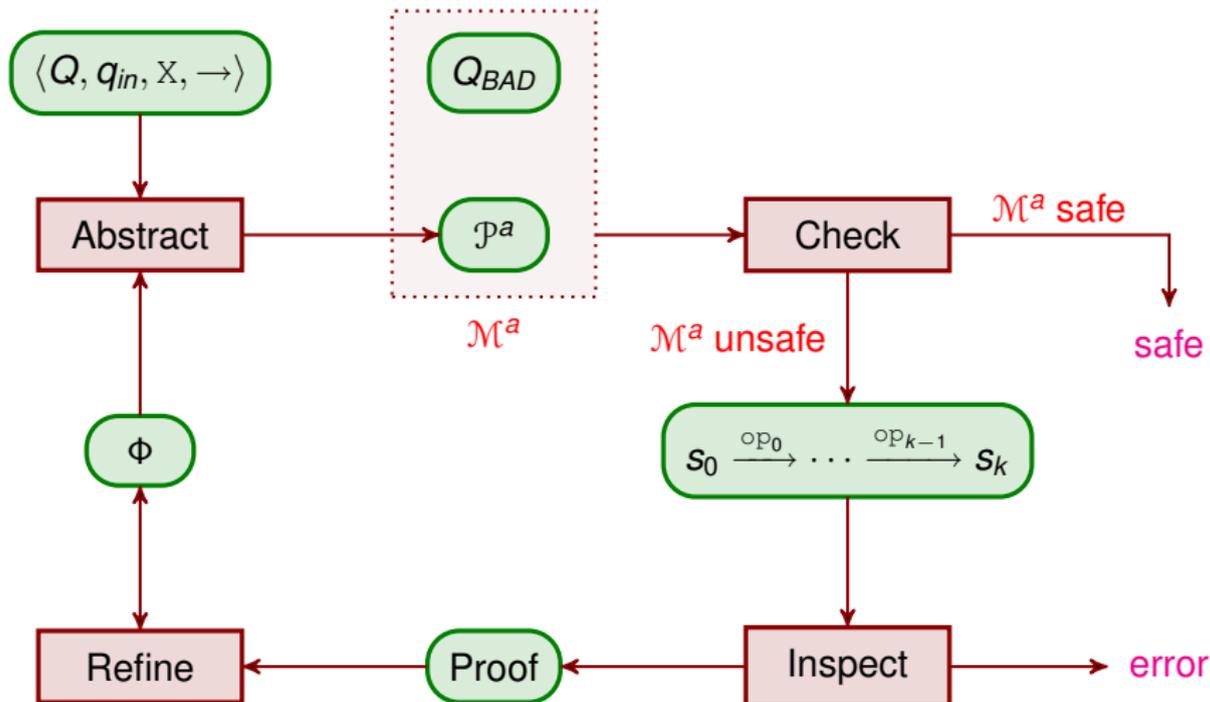
## Refinement

New predicates are atomic predicates from the path invariant.

This guarantees that the counterexample will be eliminated.

No quantifier ever introduced! 😊

# CounterExample-Guided Abstract Model Refinement



# Classical CEGAR Algorithm

```
CEGAR( $\mathcal{P} = \langle Q, q_{in}, X, \rightarrow \rangle, Q_{BAD}, \Phi_0$ )
   $\Phi \leftarrow \Phi_0$ 
  while (true)
     $\mathcal{M}^a \leftarrow \text{PredicateAbstraction}(\langle Q, q_{in}, X, \rightarrow \rangle, Q_{BAD}, \Phi)$ 
    check  $\leftarrow \text{ModelCheck}(\mathcal{M}^a)$ 
    if check is  $\mathcal{M}^a$  safe
      return  $\mathcal{P}$  safe
    // check is  $(q_{in}, v_0^a) \xrightarrow{op_0} (q_1, v_1^a) \cdots (q_k, v_k^a) \xrightarrow{op_k} (q_{bad}, v_{k+1}^a)$ 
    insp  $\leftarrow \text{Inspect}(q_{in} \xrightarrow{op_0} q_1 \cdots q_k \xrightarrow{op_k} q_{bad})$ 
    if insp is feasible
      return  $q_{in} \xrightarrow{op_0} q_1 \cdots q_k \xrightarrow{op_k} q_{bad}$  feasible
    // insp is unfeasible
    construct a path invariant and extract new predicates  $\Phi'$  from it
     $\Phi \leftarrow \Phi \cup \Phi'$ 
```

# Drawbacks of the Classical CEGAR Algorithm

## Batch-oriented integration

No sharing of data structures

No reuse of previous computations

Re-explores the same error-free parts of the configuration space

Abstraction fully computed before the model-checking phase

Useless expensive work

## Lazy CEGAR

Integrated CEGAR loop driven by the model-checker

Builds a reachability tree with abstract successors on demand

- Nodes labeled by support predicates
- Refinement only locally refines subparts of the tree

## Lazy Interpolation

Builds a reachability tree with no abstract successor computation

Uses interpolation to:

- rule out each spurious control path
- label counterexample paths in the tree with path invariants

# Part VII

## Conclusions

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- 21 Concluding Remarks
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# Summary: Abstract Model Refinement

Fully automatic software verification technique based on model-checking and refinement of finite-state abstractions

## Property-Preserving Abstraction

Conservatively extract finite-state models from programs

## Inspection of Abstract Counterexamples

Reduces to satisfiability checking

## Refinement Guided by Abstract Counterexamples

Based on the construction of path invariants

New predicates obtained from proofs of unsatisfiability

Each of these three phases relies on **theorem provers**

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21 Concluding Remarks

22 Some References

# Some CEGAR-based Software Verification Tools

## SLAM — Thomas Ball, Sriram Rajamani, ...

Analysis of programs written in C

- ☹ Classical batch-oriented CEGAR algorithm
- 😊 Interprocedural analysis (abstraction into boolean programs)

Now integrated in **Static Driver Verifier**, part of the Windows Driver Kit

## BLAST — Thomas Henzinger, ...

Analysis of programs written in C

- 😊 Lazy CEGAR algorithm
- ☹ Bounded-recursion interprocedural analysis

Open source, distributed under the BSD license

MAGIC, YASM, ...

## Why device drivers?

### High Impact

Bugs lead to system crash (e.g. BSOD)

Developed by third-party vendors

### Not So Complex

Simple safety properties (e.g. locking discipline)

Only a small part of the code is relevant to the properties

Medium-sized ( $\leq 25\,000$  lines)

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Verification of software: computation of strong enough invariants

## Abstraction Process

Interpret programs according to a simplified, “abstract” semantics.

## Property-Preserving Abstraction

Formally relate the “abstract” semantics with the “standard” semantics, so as to preserve relevant properties.

Main challenge: suitable refinement of abstractions

# Static Analysis versus Abstraction Refinement

## Static Analysis

- 😊 Always terminates
- 😞 False positives
- 😞 Manual refinement
- 😊 Infinite domains
- 😞 Same precision everywhere

Inspection & Refinement

Smart mind

## Abstraction Refinement

- 😞 May not terminate
- 😊 Definite answer (yes / no)
- 😊 Automatic refinement
- 😞 Finite abstract domains
- 😊 Adaptive precision
- 😊 Driven by the property

Inspection & Refinement

Smart prover

# Static Analysis versus Abstraction Refinement

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## Inspection & Refinement

Smart mind

## Inspection & Refinement

Smart prover

## Not Covered in the Lecture

### Computational Models

- Pointer analysis, arrays
- Recursion, threads
- Hybrid systems, ...

### Beyond Safety

- Termination
- Liveness properties
- $\mu$ -calculus (Modal LKS)

Software Verification remains a challenging problem!

### Room for Improvement

- Generation of smart predicates for refinement
- Path invariants for control paths with loops

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# Some References



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In *Proc. 9th Int. Conf. Computer Aided Verification, Haifa, Israel*, LNCS 1254, pages 72–83. Springer, 1997.



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▶ **The SLAM Project.**

<http://research.microsoft.com/slam/>

▶ **The Berkeley Lazy Abstraction Software Verification Tool.**

<http://mtc.epfl.ch/software-tools/blast/>

Thank you!