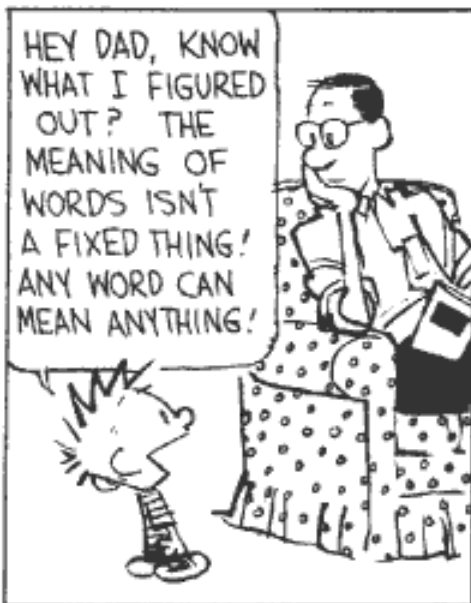


Abstract Interpretation

(Galois, Collections, Widening)



BY GIVING WORDS NEW MEANINGS, ORDINARY ENGLISH CAN BECOME AN EXCLUSIONARY CODE! TWO GENERATIONS CAN BE DIVIDED BY THE SAME LANGUAGE!



What Are We Doing?

- Sam Block Variable Renaming for Decompilation and Program Readability
- William Burns Surveying Malware's Relation to Programming Languages
- Tim Chaplin Generalizing the Burrows-Wheeler Transform
- Kirti Chawla Detection of Covert Channels in RFID Supply Chains
- Joel Coffman
- Derek Davis Classification of and Confidence in Repairs to Software Bugs
- Daniel Dougherty Branch Prediction in Dynamic Binary Translation Systems
- Zak Fry A Large-Scale Code Readability Metric
- Andrew Jurik Surveying The Use of Static Analyses to Verify Security Properties
- Dan Lepage A Bayesian Approach to SVBRDF Decomposition
- Ming Mao A Process Execution and Collaboration Policy Assertion Language
- Irwin Reyes The Accuracy of Face-Recognition Systems
- Arkaitz Ruiz Alvarez Verifying Programs that Use Intel's Threading Building Blocks Library
- Mona Sergi Improving Automatic Program Repair with Template-based Mutations
- Blake Sheridan Implementing Support for Hardware-based Profiling in Embedded Systems
- Michael Skalak Implementing and Interpreting Visual Languages
- Elizabeth Soechting Semantic Regression Testing for Tree-Structured Output
- Luther Tychonievich A Unified Compiler Representation to Support Debugging Optimized Code
- Kristen Walcott A Test Case Generation Technique for Multithreaded Programs
- Ren Xu Surveying Refinement Types

Tool Time

- How's Homework 5 going?
- Get started early
- Compilation problems?
 - See FAQ(trivia: what tool brand is this?)



More Power!

- You can handle it!

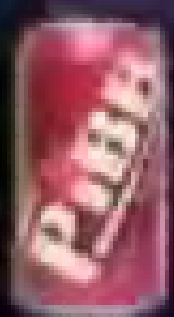


Abstract Interpretation

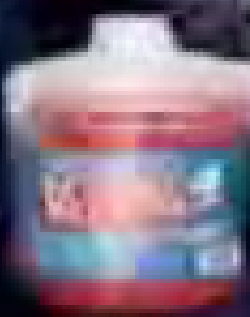
- We have an abstract domain A
 - e.g., $A = \{ \text{positive, negative, zero} \}$
 - An abstraction function $\beta : \mathbb{Z} \rightarrow A$
 - \mathbb{Z} is our concrete domain
 - A concretization function $\gamma : A \rightarrow \mathcal{P}(\mathbb{Z})$
- Positive + Positive = ???
- Positive + Negative = ???
- Positive / Zero = ???

We don't want security to get suspicious ...





+



=

**crazy
delicious**

Review

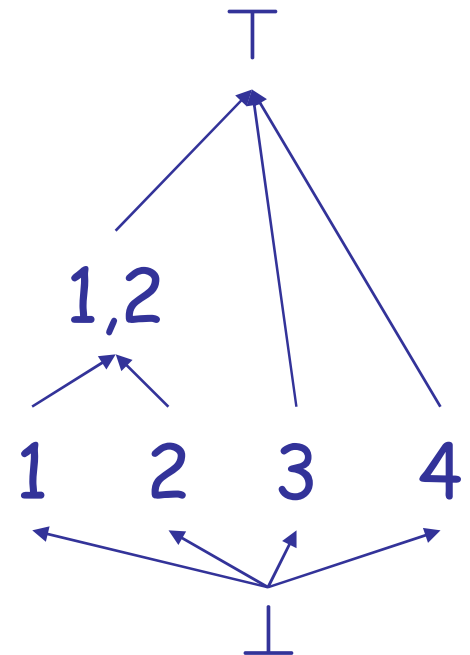
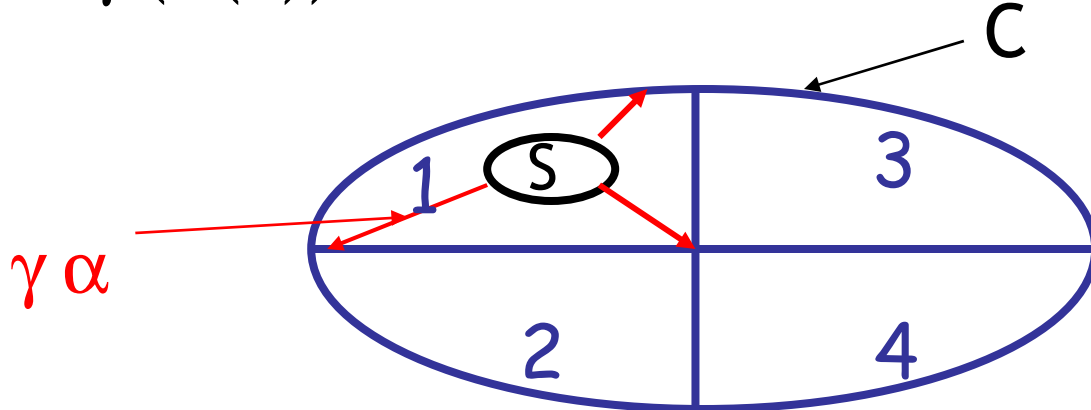
- We introduced **abstract interpretation**
- An abstraction mapping from concrete to abstract values
 - Has a concretization mapping which forms a Galois connection
- We'll look a bit more at Galois connections
- We'll lift AI from expressions to programs
- ... and we'll discuss the mythic “**widening**”

Why Galois Connections?

- We have an abstract domain A
 - An abstraction function $\beta : \mathbb{Z} \rightarrow A$
 - Induces $\alpha : \mathcal{P}(\mathbb{Z}) \rightarrow A$ and $\gamma : A \rightarrow \mathcal{P}(\mathbb{Z})$
- We argued that for correctness
$$\gamma(a_1 \text{ op } a_2) \supseteq \gamma(a_1) \text{ op } \gamma(a_2)$$
 - We wish for the set on the left to be as small as possible
 - To reduce the **loss of information through abstraction**
- For each set $S \subseteq \mathbb{C}$, define $\alpha(S)$ as follows:
 - Pick smallest S' that includes S and is in the image of γ
 - Define $\alpha(S) = \gamma^{-1}(S')$
 - Then we define: $a_1 \text{ op } a_2 = \alpha(\gamma(a_1) \text{ op } \gamma(a_2))$
- Then α and γ form a Galois connection

Galois Connections

- A Galois connection between complete lattices A and $\mathcal{P}(C)$ is a pair of functions α and γ such that:
 - γ and α are monotonic
 - (with the \subseteq ordering on $\mathcal{P}(C)$)
 - $\alpha(\gamma(a)) = a$ for all $a \in A$
 - $\gamma(\alpha(S)) \supseteq S$ for all $S \in \mathcal{P}(C)$



More on Galois Connections



- All Galois connections are **monotonic**
- In a Galois connection one function uniquely and absolutely **determines the other**

Abstract Interpretation for Imperative Programs

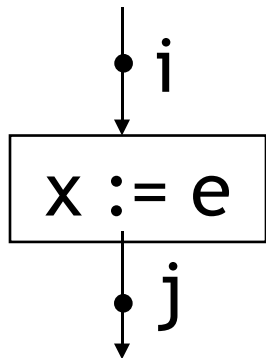
- So far we abstracted the value of **expressions**
- Now we want to abstract the **state** at each point in the program
- First we define the concrete semantics that we are abstracting
 - We'll use a **collecting semantics**

Collecting Semantics

- Recall
 - A state $\sigma \in \Sigma$. Any state σ has type $\text{Var} \rightarrow \mathbb{Z}$
 - States vary from program point to program point
- We introduce a set of program points: labels
- We want to answer questions like:
 - Is x always positive at label i ?
 - Is x always greater or equal to y at label j ?
- To answer these questions we'll construct
 - $C \in$ **Contexts**. C has type $\text{Labels} \rightarrow \mathcal{P}(\Sigma)$
 - For each label i , $C(i)$ = all possible states at label i
 - This is called the collecting semantics of the program
 - This is basically what SLAM (and BLAST, ESP, ...) approximate (using BDDs to store $\mathcal{P}(\Sigma)$ efficiently)

Defining the Collecting Semantics

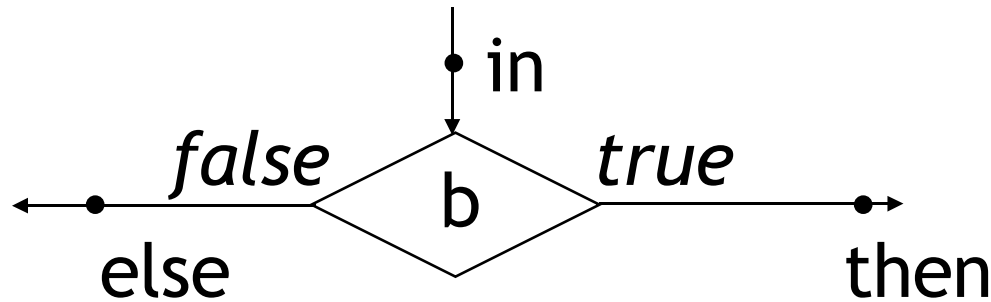
- We first define relations between the collecting semantics at different labels
 - We do it for **unstructured CFGs** (cf. HW5!)
 - Can do it for IMP with careful notion of program points
- Define a **label on each edge** in the CFG
- For assignment



$$C_j = \{\sigma[x := n] \mid \sigma \in C_i \wedge [[e]]\sigma = n\}$$

Defining the Collecting Semantics

- For conditionals



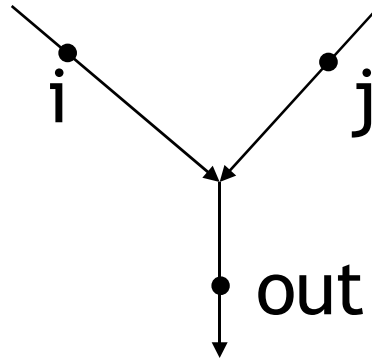
$$C_{\text{else}} = \{ \sigma \mid \sigma \in C_{\text{in}} \wedge \llbracket b \rrbracket \sigma = \text{false} \}$$

$$C_{\text{then}} = \{ \sigma \mid \sigma \in C_{\text{in}} \wedge \llbracket b \rrbracket \sigma = \text{true} \}$$

- Assumes b has **no side effects** (as in IMP or HW5)

Defining the Collecting Semantics

- For a join

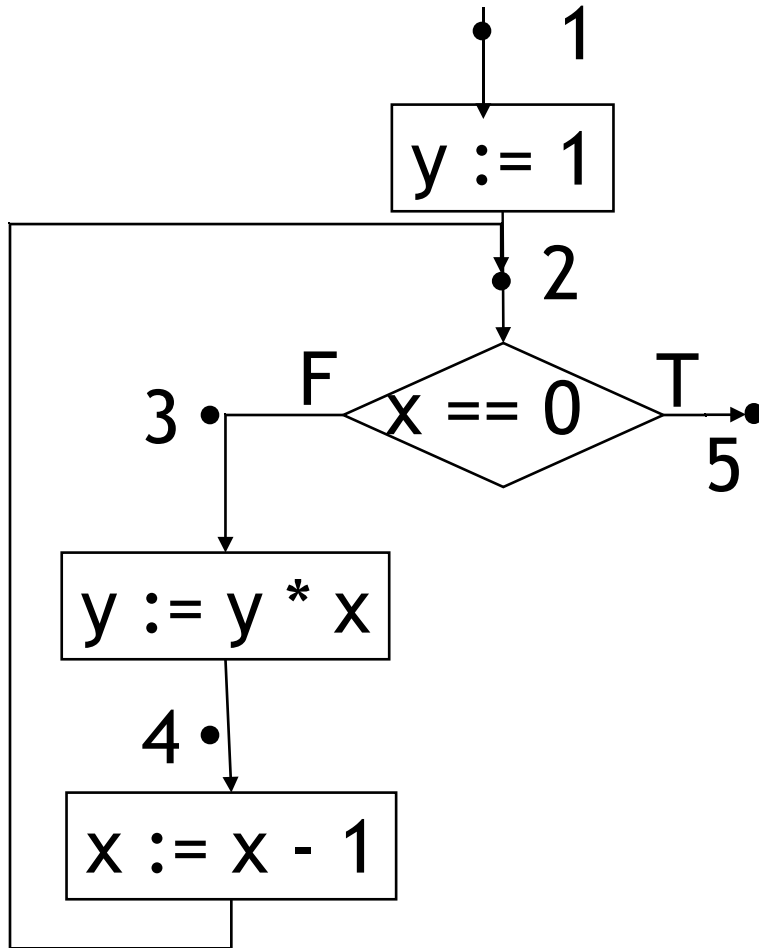


$$C_{\text{out}} = C_i \cup C_j$$

- Verify that these relations are **monotonic**
 - If we increase a C_x all other C_y can only increase

Collecting Semantics: Example

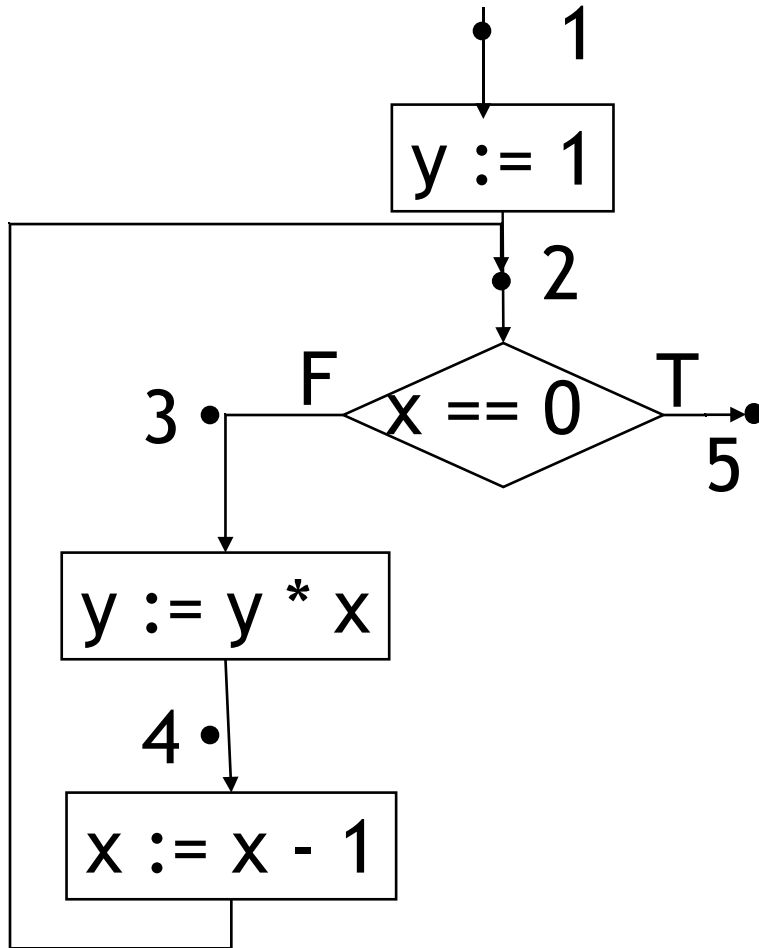
- Assume $x \geq 0$ initially (*explain this?*)



$$C_1 = \{\sigma \mid \sigma(x) \geq 0\}$$

Collecting Semantics: Example

- Assume $x \geq 0$ initially

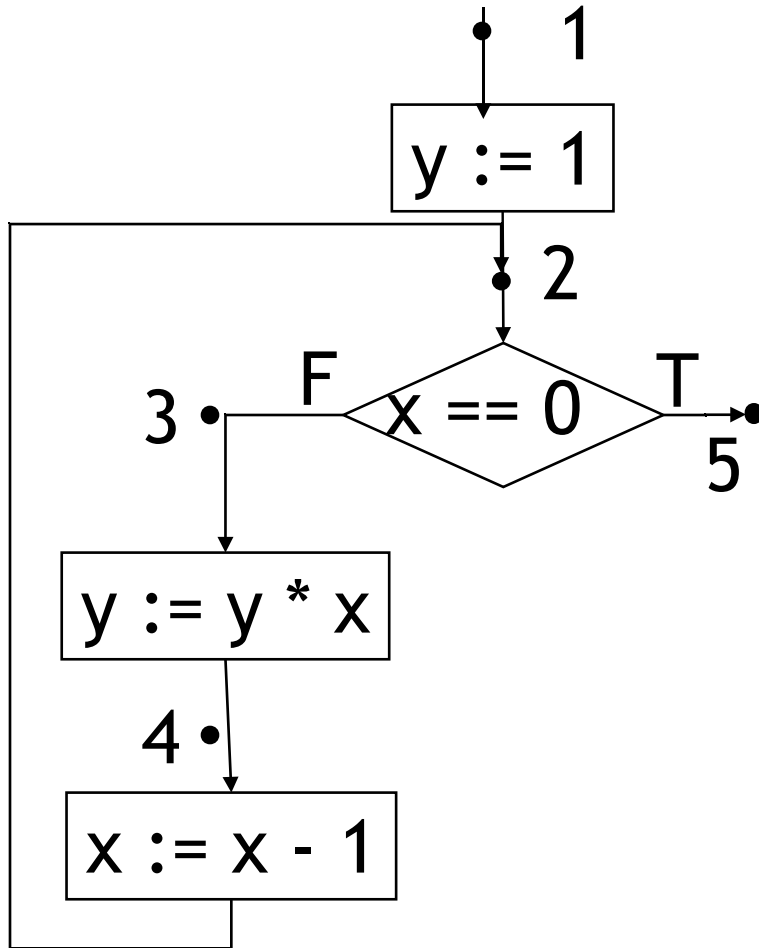


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Collecting Semantics: Example

- Assume $x \geq 0$ initially



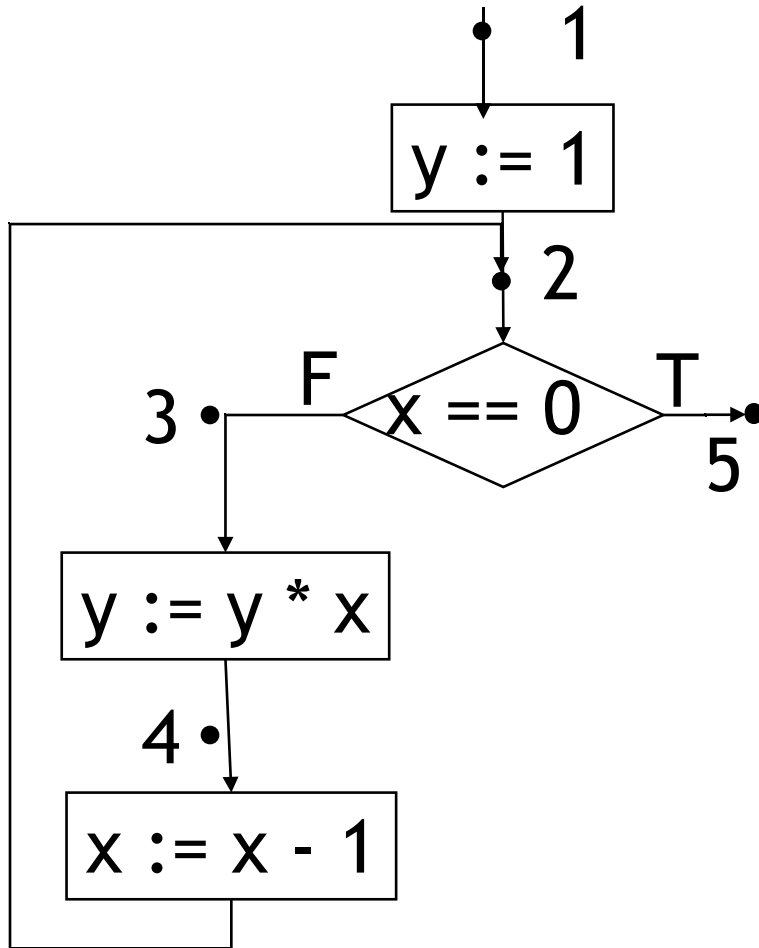
$$C_1 = \{\sigma \mid \sigma(x) \geq 0\}$$

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$$C_3 = C_2 \cap \{\sigma \mid \sigma(x) \neq 0\}$$

Collecting Semantics: Example

- Assume $x \geq 0$ initially



$$C_1 = \{\sigma \mid \sigma(x) \geq 0\}$$

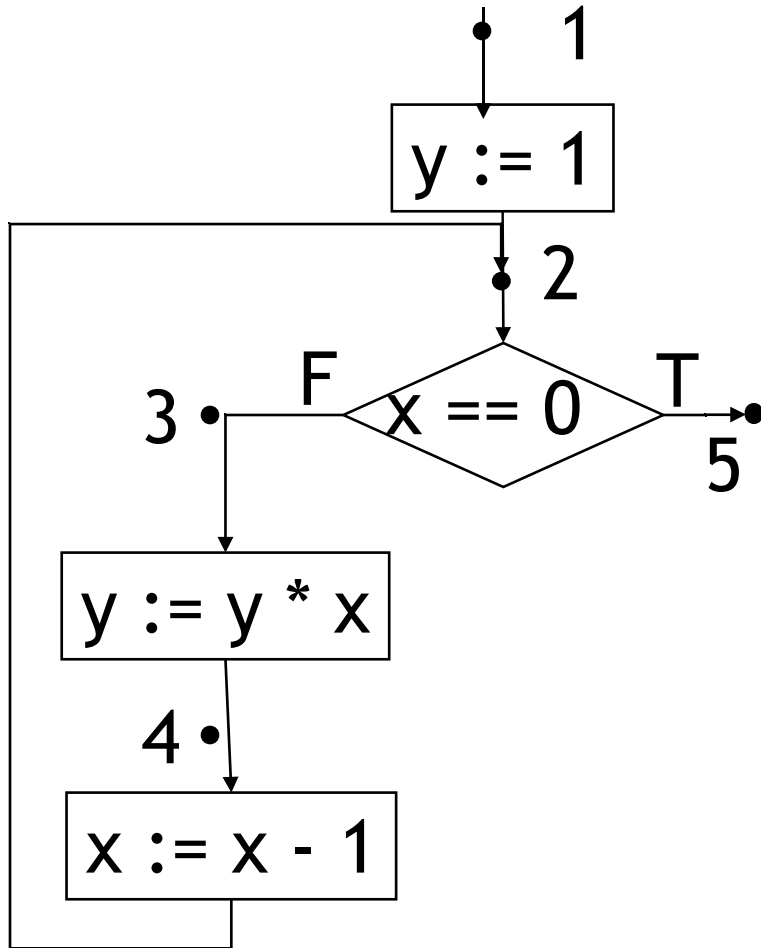
$$C_2 = \{\sigma[y:=1] \mid \sigma \in C_1\}$$

$$C_3 = C_2 \cap \{\sigma \mid \sigma(x) \neq 0\}$$

$$C_4 = \{\sigma[y:=\sigma(y)*\sigma(x)] \mid \sigma \in C_3\}$$

Collecting Semantics: Example

- Assume $x \geq 0$ initially



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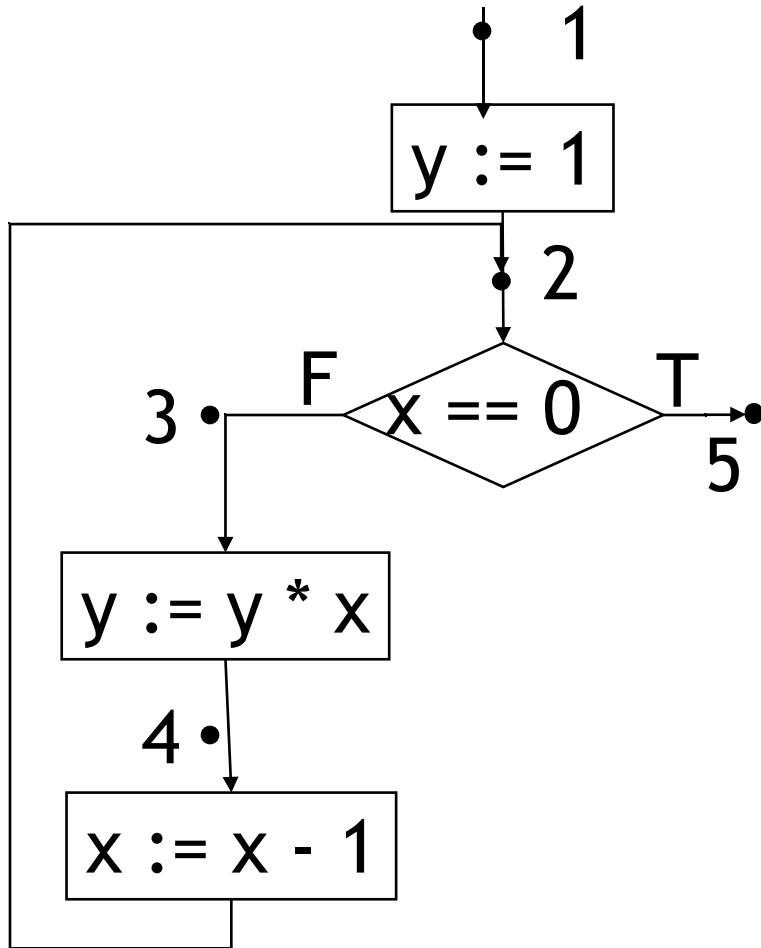
$$\cup \{\sigma[x:=\sigma(x)-1] \mid \sigma \in C_4\}$$

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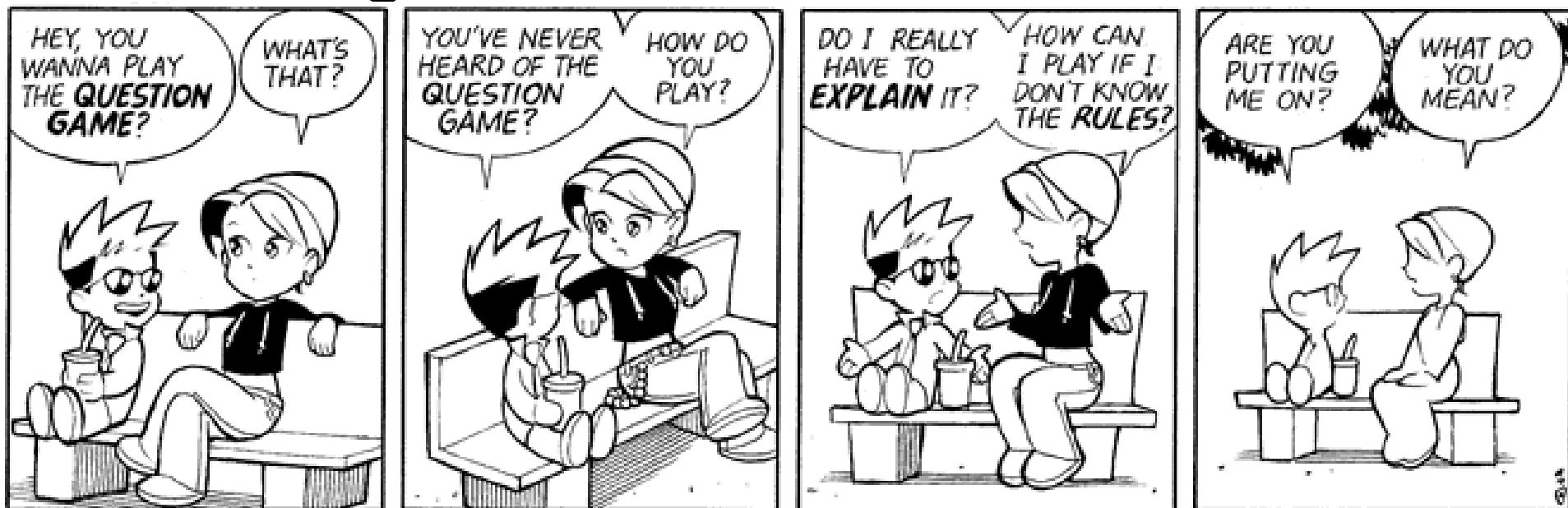
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$$C_5 = C_2 \cap \{\sigma \mid \sigma(x) = 0\}$$

Why Does This Work?

- We just made a system of **recursive equations** that are **defined largely in terms of themselves**
 - e.g., $C_2 = F(C_4)$, $C_4 = G(C_3)$, $C_3 = H(C_2)$
- Why do we have any reason to believe that this will get us what we want?

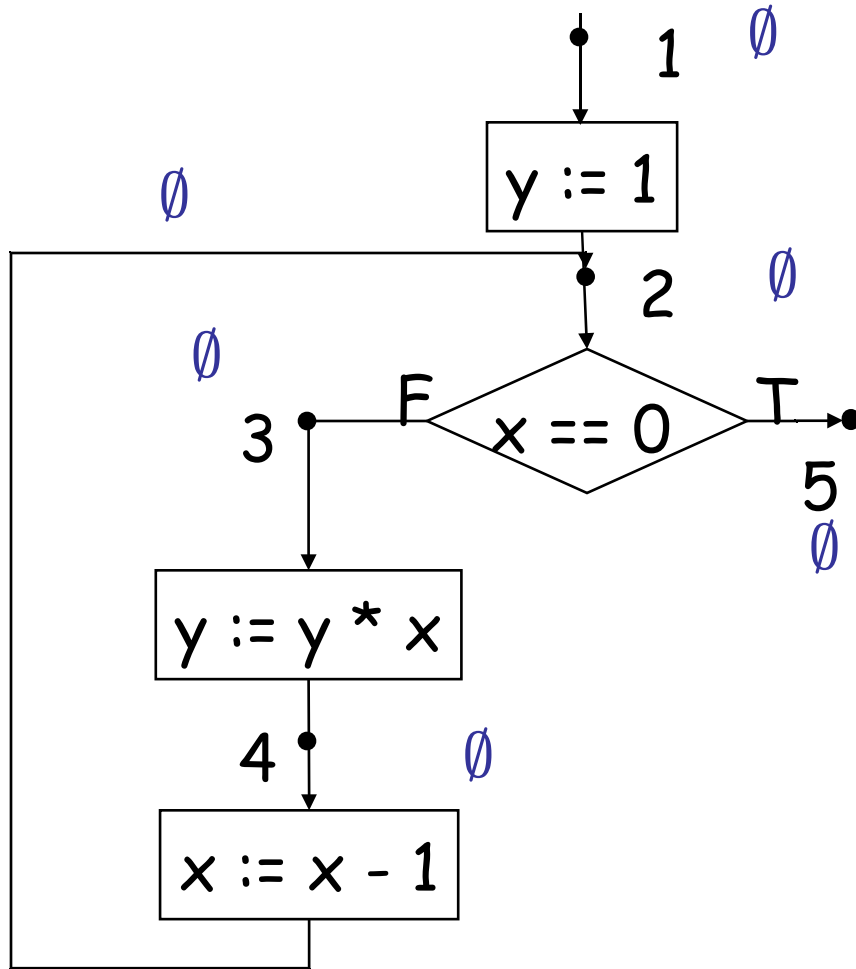


The Collecting Semantics

- We have an equation with the unknown C
 - The equation is defined by a **monotonic** and **continuous** function on domain $\text{Labels} \rightarrow \mathcal{P}(\Sigma)$
- We can use the **least fixed-point theorem**
 - Start with $C^0(L) = \emptyset$ (aka $C^0 = \lambda L. \emptyset$)
 - Apply the relations between C_i and C_j to get C^1_i from C^0_j
 - Stop when all $C^k = C^{k-1}$
 - Problem: **we'll go on forever for most programs**
 - But we know **the fixed point exists**

Collecting Semantics: Example

- (assume $x \geq 0$ initially)



$$C_1 = \{\sigma \mid \sigma(x) \geq 0\}$$

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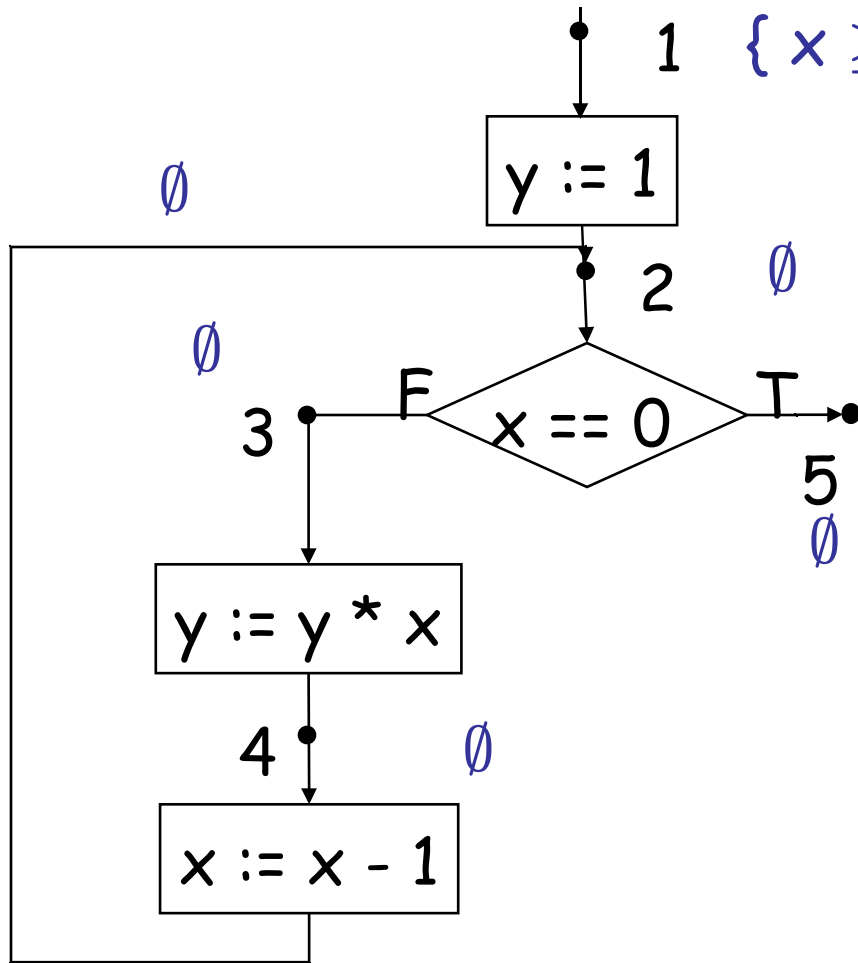
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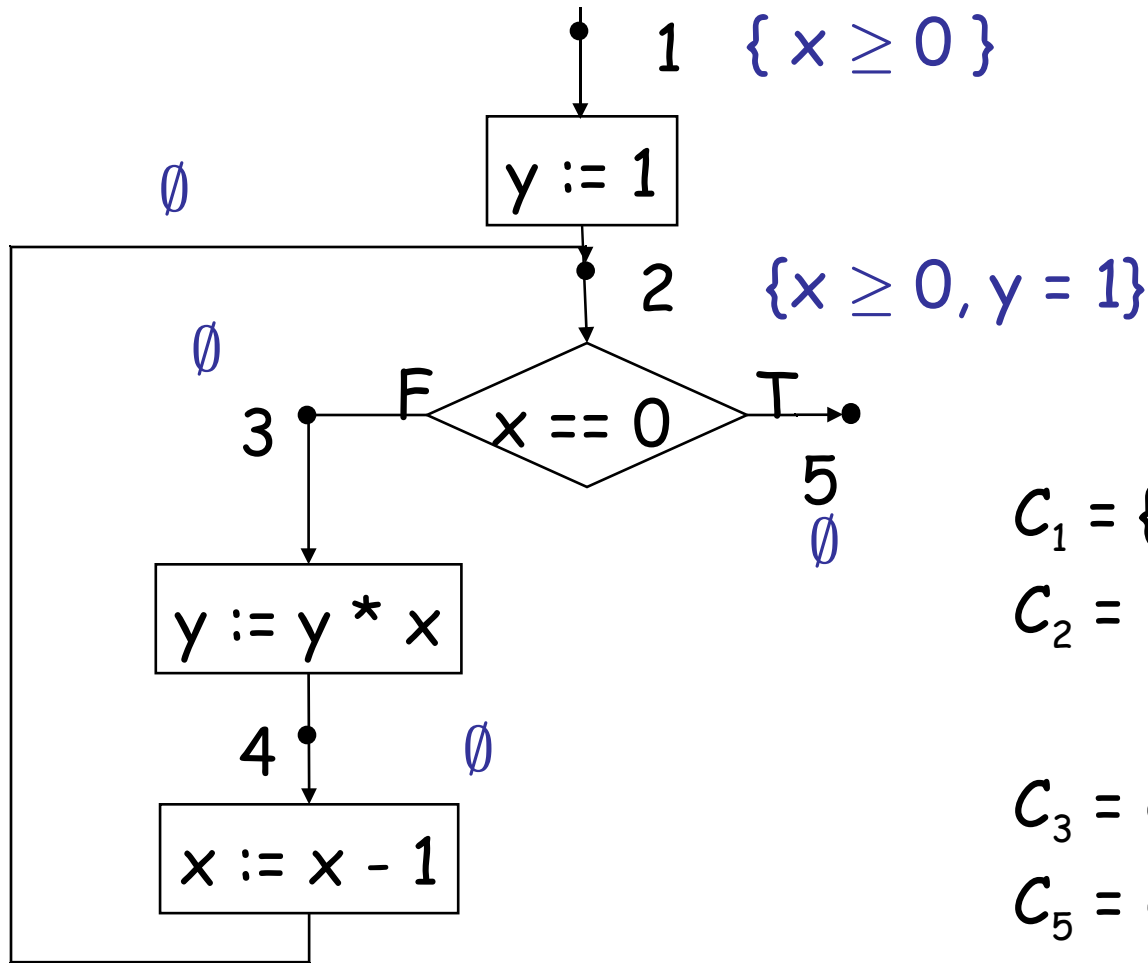
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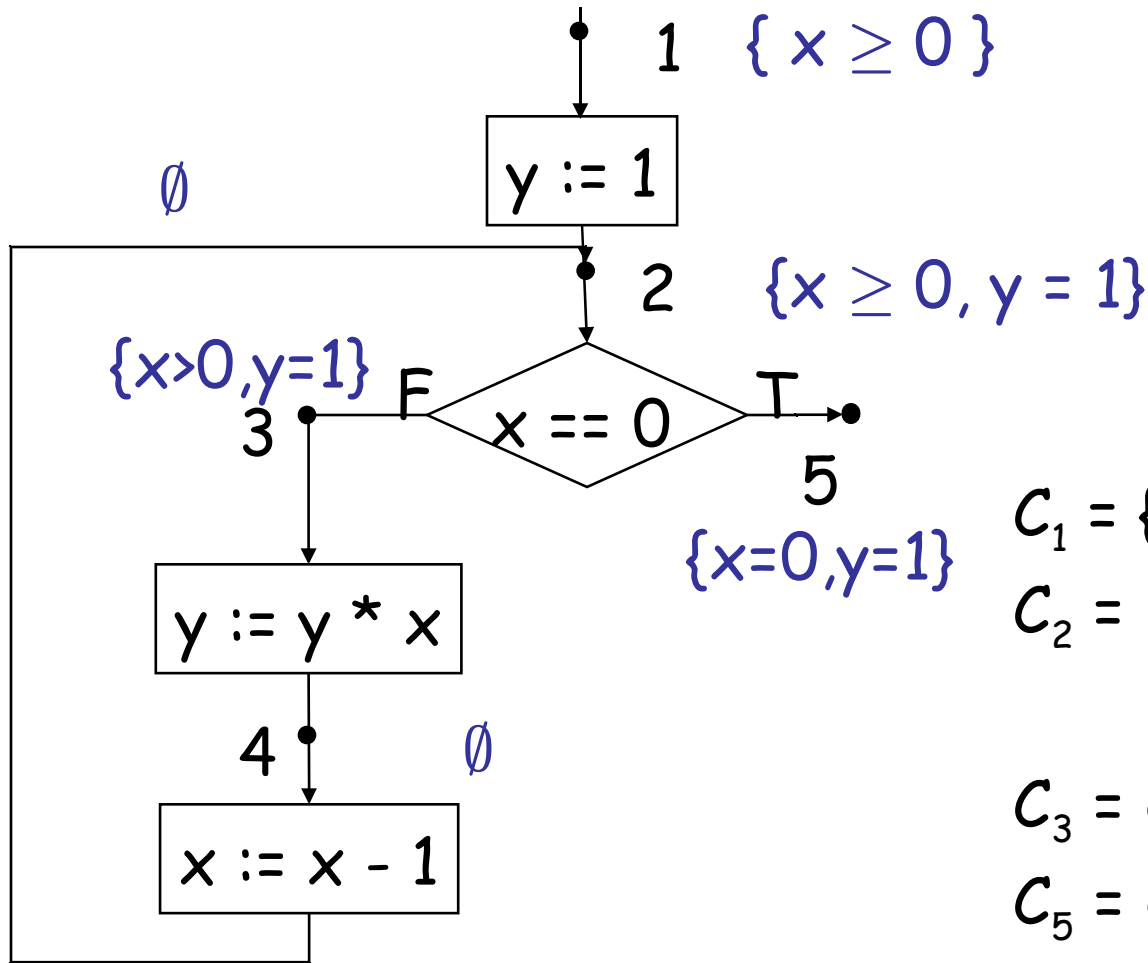
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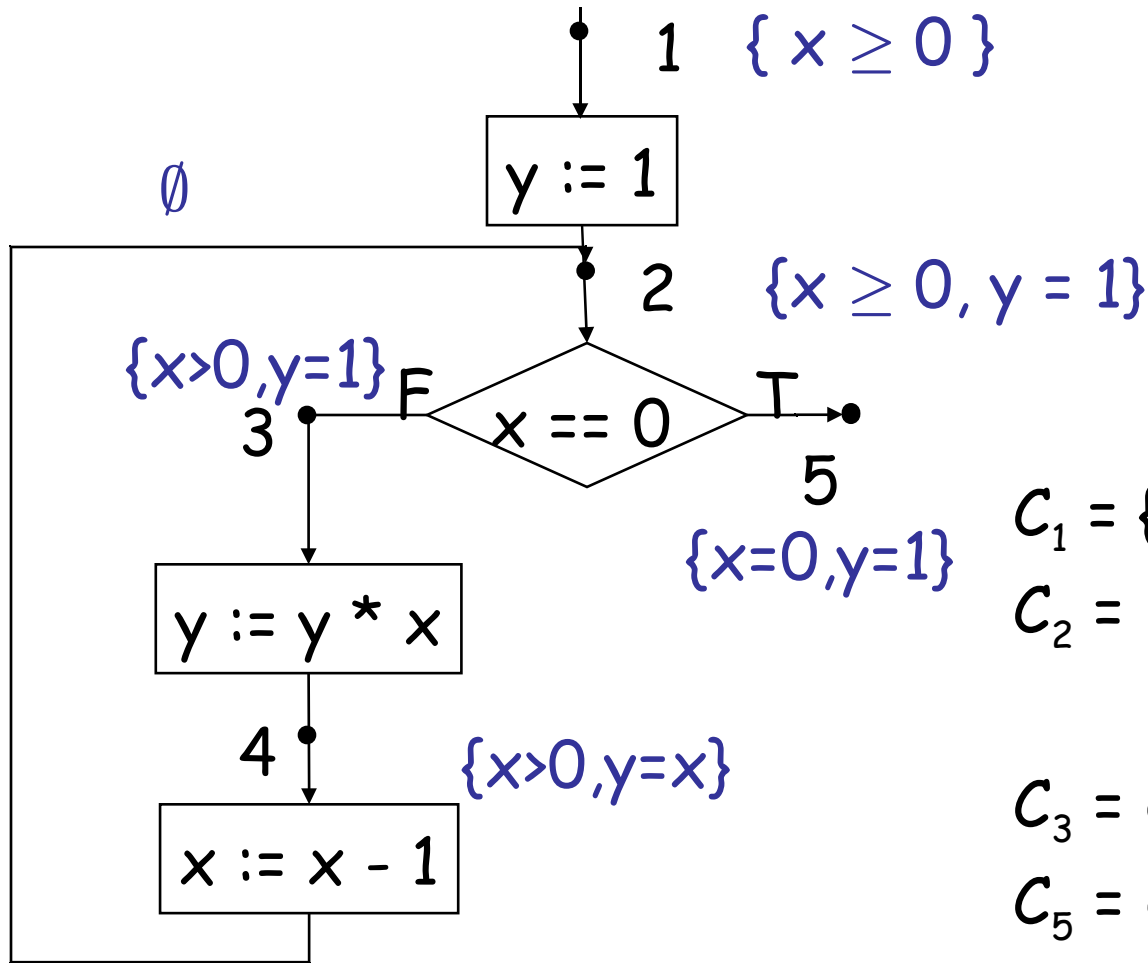
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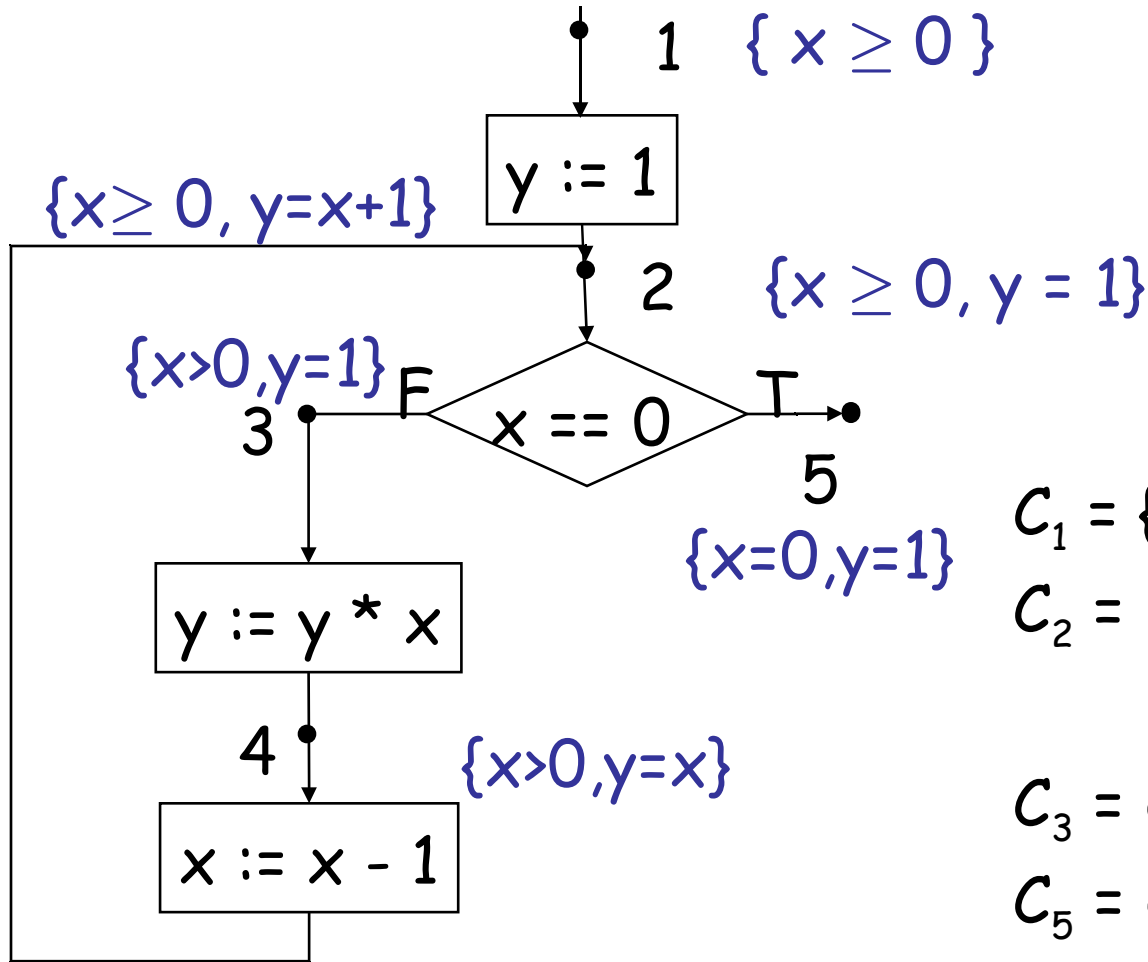
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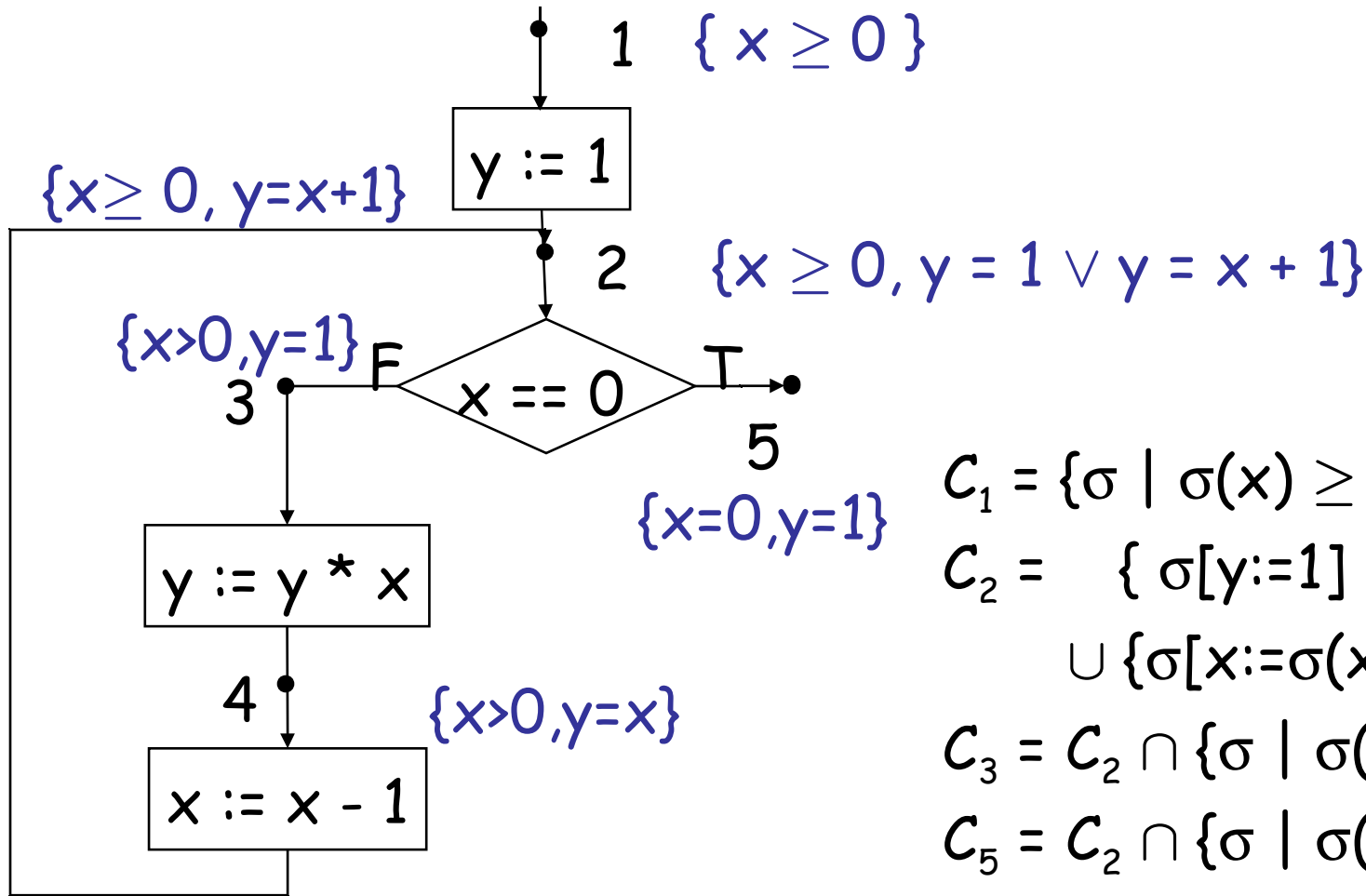
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Q: Theatre (006 / 842)

- Name the 1879 Gilbert & Sullivan operetta parodied by the following quote:
 - *I am the very model of a Newsgroup personality.*
 - *I intersperse obscenity with tedious banality.*
 - *Addresses I have plenty of, both genuine and ghosted too,*
 - *On all the countless newsgroups that my drivel is cross-posted to.*

Q: Movies (387 / 842)

- Name the movie quoted below and also name either character or either character's actor. In this 1987 Mel Brooks spoof, one character is revealed to be another character's *"father's brother's nephew's cousin's former roommate."*

Q: TV Music (040 / 842)

- Fill in the three blanks in this **Flintstones** theme song snippet:
 - *Let's ride with the family down the street*
 - *Through the courtesy of blank
blank blank*
 - *When you're with the Flintstones*
 - *Have a yabba dabba doo time*

Abstract Interpretation

- Pick a complete lattice A (abstractions for $\mathcal{P}(\Sigma)$)
 - Along with a monotonic abstraction $\alpha : \mathcal{P}(\Sigma) \rightarrow A$
 - Alternatively, pick $\beta : \Sigma \rightarrow A$
 - This uniquely defines its Galois connection γ
- Take the relations between C_i and move them to the abstract domain:

$$a : \text{Label} \rightarrow A$$

- Assignment

Concrete: $C_j = \{\sigma[x := n] \mid \sigma \in C_i \wedge \llbracket e \rrbracket \sigma = n\}$

Abstract: $a_j = \alpha \{\sigma[x := n] \mid \sigma \in \gamma(a_i) \wedge \llbracket e \rrbracket \sigma = n\}$

Abstract Interpretation

- Conditional

Concrete: $C_j = \{ \sigma \mid \sigma \in C_i \wedge \llbracket b \rrbracket \sigma = \text{false} \}$ and
 $C_k = \{ \sigma \mid \sigma \in C_i \wedge \llbracket b \rrbracket \sigma = \text{true} \}$

Abstract: $a_j = \alpha \{ \sigma \mid \sigma \in \gamma(a_i) \wedge \llbracket b \rrbracket \sigma = \text{false} \}$ and
 $a_k = \alpha \{ \sigma \mid \sigma \in \gamma(a_i) \wedge \llbracket b \rrbracket \sigma = \text{true} \}$

- Join

Concrete: $C_k = C_i \cup C_j$

Abstract: $a_k = \alpha (\gamma(a_i) \cup \gamma(a_j)) = \text{lub} \{a_i, a_j\}$

Least Fixed Points In The Abstract Domain

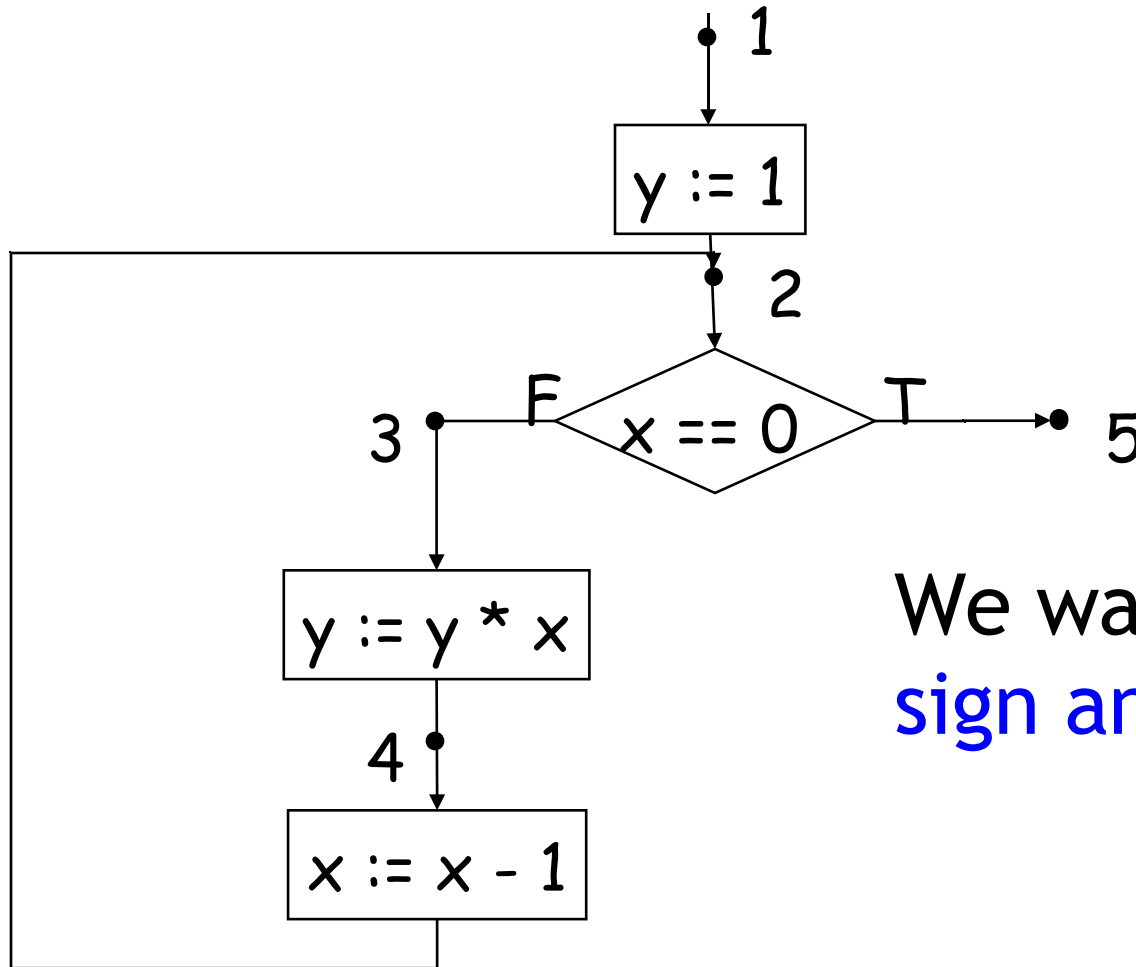
- We have a recursive equation with unknown “a”
 - Defined by a **monotonic** and **continuous** function on the domain $\text{Labels} \rightarrow A$
- We can use the **least fixed-point theorem**:
 - Start with $a^0 = \lambda L. \perp$ (aka: $a^0(L) = \perp$)
 - Apply the monotonic function to compute a^{k+1} from a^k
 - Stop when $a^{k+1} = a^k$
- Exactly the same computation as for the collecting semantics
 - **What is new?**
 - *“There is nothing new under the sun but there are lots of old things we don't know.” - Ambrose Bierce*

Least Fixed Points In The Abstract Domain

- We have a **hope of termination!**
- Classic setup: A has only uninteresting chains (finite number of elements in each chain)
 - A has finite height h (= “finite-height lattice”)
- The computation takes $O(h \times |\text{Labels}|^2)$ steps
 - At each step “a” makes progress on at least one label
 - We can only make progress h times
 - And each time we must compute $|\text{Labels}|$ elements
- This is a **quadratic analysis**: good news
 - This is exactly the same as Kildall’s 1973 analysis of dataflow’s polynomial termination given a finite-height lattice and monotonic transfer functions.

Abstract Interpretation: Example

- Consider the following program, $x > 0$



We want to do the **sign analysis** on it.

Abstract Domain for Sign Analysis

- **Invent** the complete sign lattice

$$S = \{ \perp, -, 0, +, \top \}$$

- Construct the complete lattice

$$A = \{x, y\} \rightarrow S$$

- With the usual point-wise ordering
- Abstract state gives the sign for x and y
- We start with $a^0 = \lambda L. \lambda v \in \{x, y\}. \perp$
 - aka: $a^0(L, v) = \perp$

Let's Do It!

Label		Iterations →									
1	x	+									+
	y	T									T
2	x	⊥	+			T					T
	y	⊥	+						T		T
3	x	⊥		+			T				T
	y	⊥		+						T	T
4	x	⊥			+			T			T
	y	⊥			+			T			T
5	x	⊥					0				0
	y	⊥					+			T	T

Notes, Weaknesses, Solutions

- We abstracted the state of each **variable independently**

$$A = \{x, y\} \rightarrow \{\perp, -, 0, +, \top\}$$

- We lost relationships between variables
 - e.g., at a point x and y may always have the same sign
 - In the previous abstraction we get $\{x := \top, y := \top\}$ at label 2 (when in fact **y is always positive!**)
- We can also abstract the state as a whole

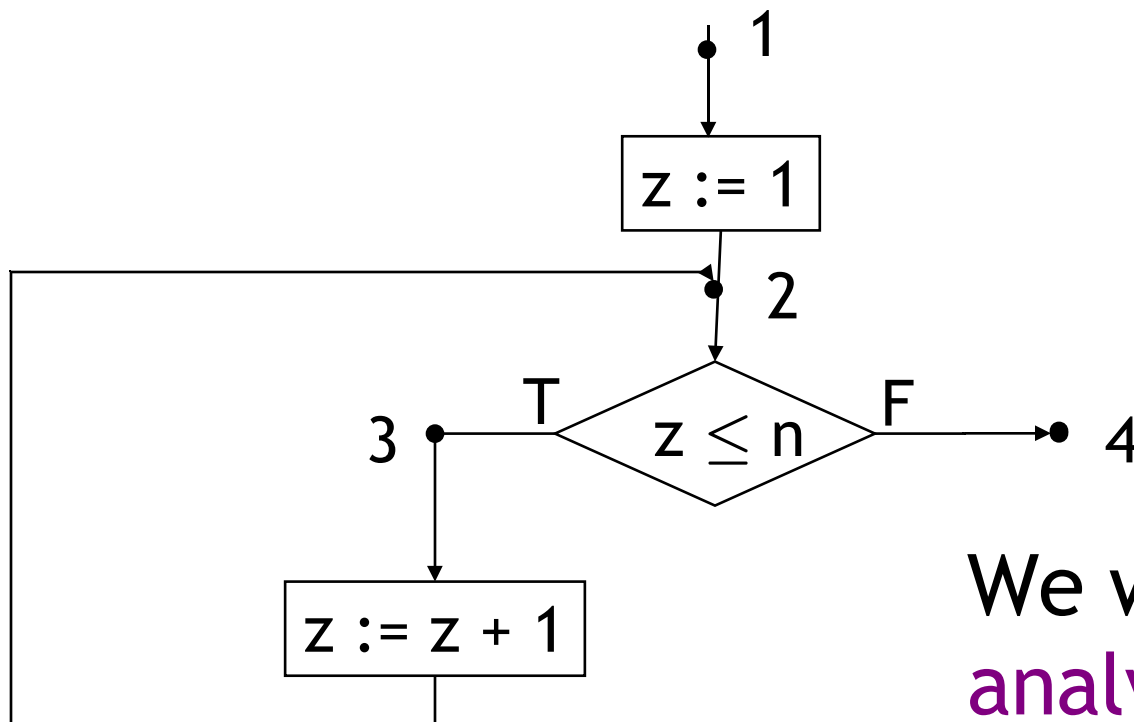
$$A = \mathcal{P}(\{\perp, -, 0, +, \top\} \times \{\perp, -, 0, +, \top\})$$

Other Abstract Domains

- Range analysis
 - Lattice of ranges: $R = \{ \perp, [n..m], (-\infty, m], [n, +\infty), \top \}$
 - It is a complete lattice
 - $[n..m] \sqcup [n'..m'] = [\min(n, n').. \max(m, m')]$
 - $[n..m] \sqcap [n'..m'] = [\max(n, n').. \min(m, m')]$
 - With appropriate care in dealing with ∞
 - $\beta : \mathbb{Z} \rightarrow R$ such that $\beta(n) = [n..n]$
 - $\alpha : \mathcal{P}(\mathbb{Z}) \rightarrow R$ such that $\alpha(S) = \text{lub} \{ \beta(n) \mid n \in S \} = [\min(S).. \max(S)]$
 - $\gamma : R \rightarrow \mathcal{P}(\mathbb{Z})$ such that $\gamma(r) = \{ n \mid n \in r \}$
- This lattice has **infinite-height chains**
 - So the abstract interpretation **might not terminate!**

Example of Non-Termination

- Consider this (common) program fragment



We want to do **range analysis** on it.

Example of Non-Termination

- Consider the sequence of abstract states at point 2
 - $[1..1], [1..2], [1..3], \dots$
 - The analysis **never terminates**
 - Or terminates very late if the loop bound is known statically
- It is time to approximate even more: widening
- We redefine the join (lub) operator of the lattice to ensure that from $[1..1]$ upon union with $[2..2]$ the result is $[1..+\infty)$ and not $[1..2]$
- Now the sequence of states is
 - $[1..1], [1, +\infty), [1, +\infty)$ Done (no more infinite chains)

Formal Definition of Widening

(Cousot 16.399 “Abstract Interpretation”, 2005)

- A widening $\nabla : (P \times P) \rightarrow P$ on a poset $\langle P, \sqsubseteq \rangle$ satisfies:
 - $\forall x, y \in P . x \sqsubseteq (x \nabla y) \wedge y \sqsubseteq (x \nabla y)$
 - For all **increasing chains** $x^0 \sqsubseteq x^1 \sqsubseteq \dots$ the increasing chain $y^0 =_{\text{def}} x^0, \dots, y^{n+1} =_{\text{def}} y^n \nabla x^{n+1}, \dots$ is not strictly increasing.
- Two different main uses:
 - Approximate missing lubs. (*Not for us.*)
 - **Convergence acceleration.** (*This is the real use.*)
 - A widening operator can be used to effectively compute an upper approximation of the least fixpoint of $F \in L \nabla L$ starting from below when L is computer-representable but does not satisfy the ascending chain condition.

Formal Widening Example

$$[1, 1] \nabla [1, 2] = [1, +\infty)$$

- Range Analysis on z:

L0: $z := 1$;

L1: while $z < 99$ do

L2: $z := z + 1$

L3: done /* $z \geq 99$ */

L4:

$x_j^{L_i}$ =^{def} the j th iterative attempt to compute an abstract value for z at label L_i

Recall $\text{lub } S = [\min(S).. \max(S)]$
 $\text{lub } \{[2, +\infty), [1, +\infty)\} = \{[1, +\infty)\}$

Original x^i	Widened y^i
$x_0^{L0} = \perp$	$y_0^{L0} = \perp$
$x_0^{L1} = [1, 1]$	$y_0^{L1} = [1, 1]$
$x_0^{L2} = [1, 1]$	$y_0^{L2} = [1, 1]$
$x_0^{L3} = [2, 2]$	$y_0^{L3} = [2, 2]$
$x_1^{L2} = [1, 2]$	$y_1^{L2} = [1, +\infty)$
$x_1^{L3} = [2, +\infty)$	$y_1^{L3} = [2, +\infty)$
$x_0^{L4} = [99, +\infty)$	$y_0^{L4} = [99, +\infty)$
stable (fewer than 99 iterations!)	

Other Abstract Domains

- Linear relationships between variables
 - A convex [polyhedron](#) is a subset of \mathbb{Z}^k whose elements satisfy a number of inequalities:
$$a_1x_1 + a_2x_2 + \dots + a_kx_k \geq c_i$$
 - This is a complete lattice; linear programming methods compute lubs
- Linear relationships with at most two variables
 - Convex polyhedra but with ≤ 2 variables per constraint
 - Octagons ($x \pm y \geq c$) have efficient algorithms
- Modulus constraints (e.g. even and odd)

Abstract Chatter

- **AI, Dataflow and Software Model Checking**
 - The big three (aside from flow-insensitive type systems) for program analyses
- Are in fact quite related:
 - David Schmidt. *Data flow analysis is model checking of abstract interpretation*. POPL '98.
- AI is usually flow-sensitive (per-label answer)
- AI can be path-sensitive (if your abstract domain includes \vee , for example), which is just where model checking uses BDD's
- Metal, SLAM, ESP, ... can all be viewed as AI

Abstract Interpretation

Conclusions

- AI is a very powerful technique that underlies a large number of program analyses
- AI can also be applied to functional and logic programming languages
- There are a few success stories
 - Strictness analysis for lazy functional languages
 - PolySpace for linear constraints
- In most other cases however AI is still slow
- When the lattices have infinite height and widening heuristics are used the result becomes unpredictable