

# Generation of Inductive Assertion Network using Interprocedural Data Flow Analysis

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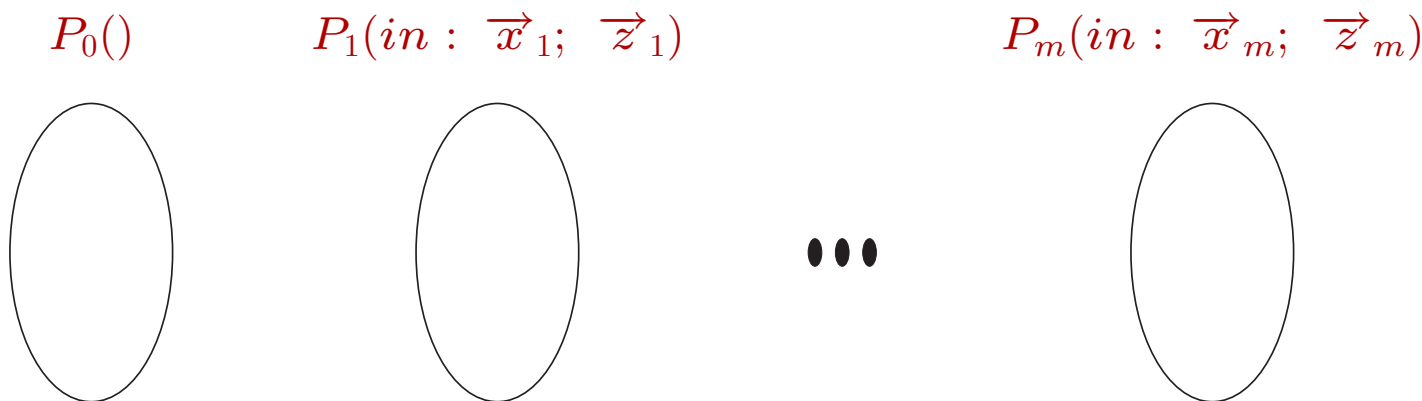
Joint work with: Amir Pnueli

# Introduction

- This is a preliminary result.
- Interprocedural Translation Validation as motivation.
- Constant propagation is one of the central optimizations.
- Use source invariants in order to strengthen our proof rules.
- Introduce an existing method for obtaining the precise solution for the constant propagation problem.
- Obtain the source invariants using the solution.

# Transition Graphs to Model Programs with Procedures

An application  $\mathcal{A}$  consists of  $m + 1$  modules, where  $P_0$  represents the main procedure:



The variables of each module  $P_i$  are partitioned into  $\vec{y} = (\vec{x}; \vec{z}; \vec{w})$ , where:

- $\vec{x}$  are the input parameters passed by **value**;
- $\vec{z}$  are the input parameters passed by **reference**;
- $\vec{w}$  denotes **local(working)** variables.

# Instructions

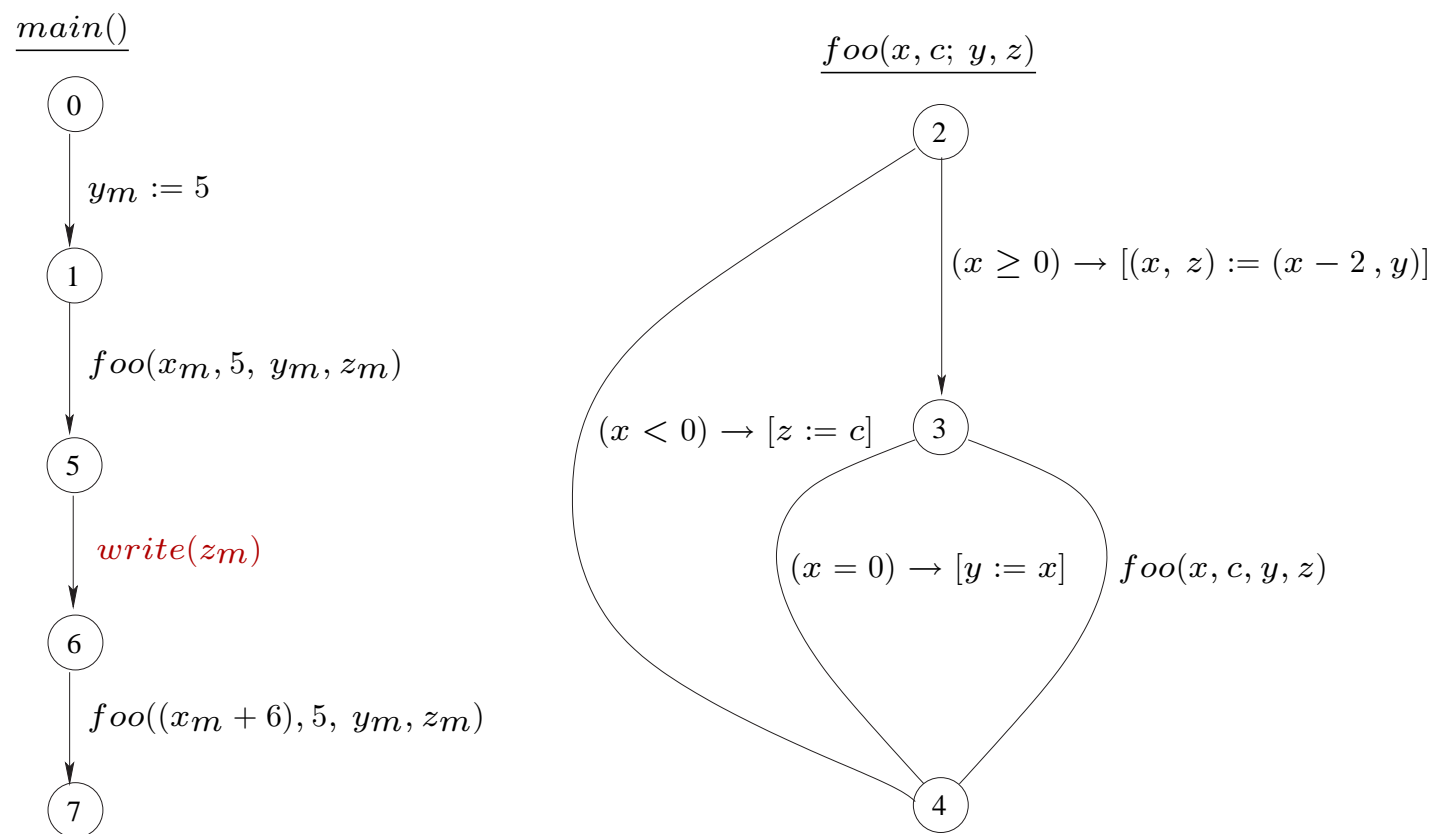
Nodes of the graph are connected by directed edges labeled by instructions:

- A **Guarded assignment** is an instruction of the form  $c \rightarrow [\vec{u} := \text{exp}(\vec{y})]$ , where  $c$  is a boolean expression over  $\vec{y}$ , where  $\vec{u} \subseteq \vec{y}$ .
- **Read** and **write** instructions are denoted by  $\text{read}(\vec{u})$  and  $\text{write}(\vec{u})$ .
- **Procedure call** instruction  $f(\text{exp}(\vec{y}), \vec{u})$  denotes a call to module  $f(\text{in} : \vec{x}_f; \vec{z}_f)$ , passing input parameters  $\text{exp}(\vec{y})$  and  $\vec{u}$ .

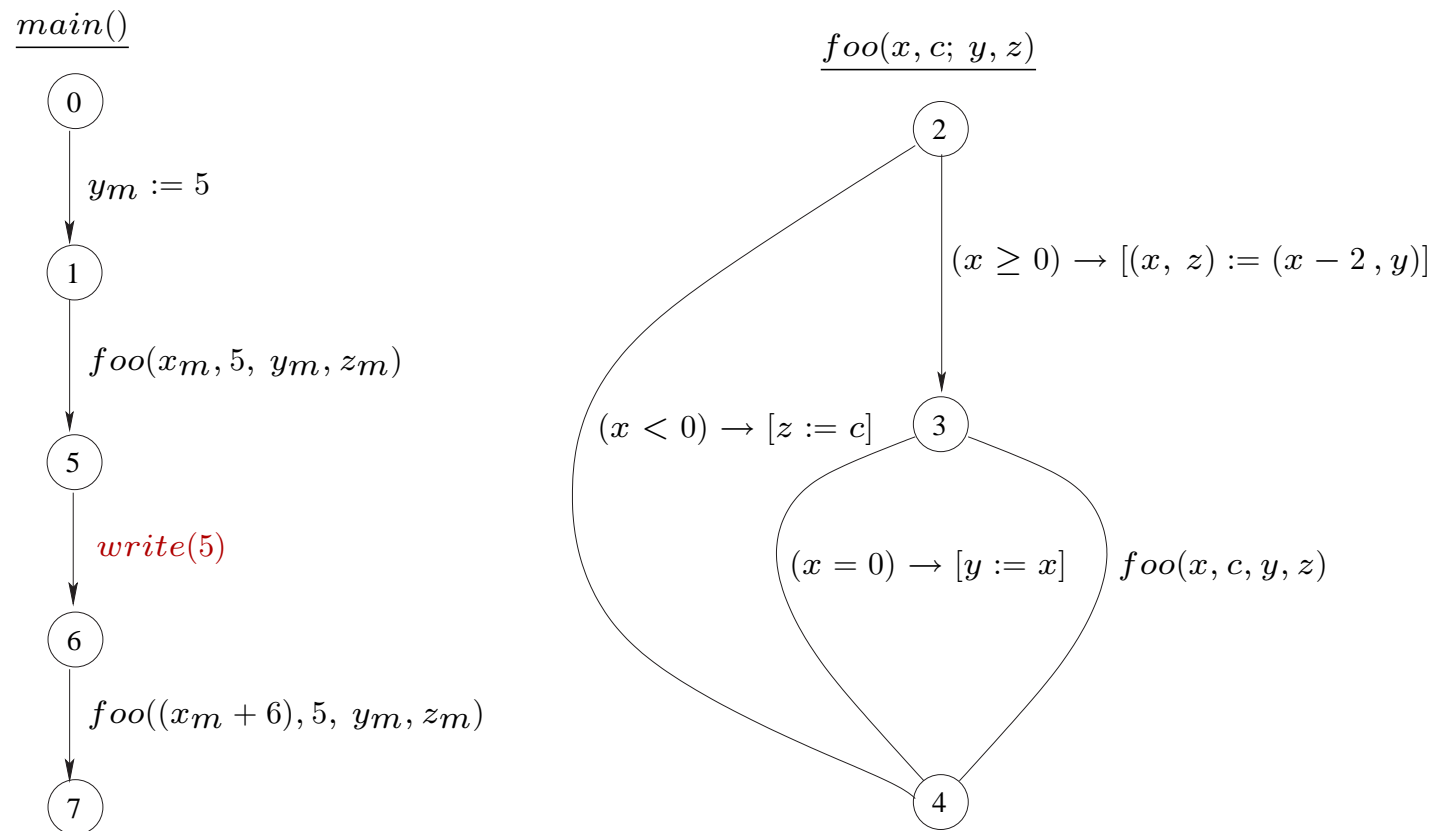
## From Programs to Transition Graphs

- We choose the set of **cut-points** of a procedure  $P_k$  to be the set of **nodes** for the corresponding transition graph.
- If there is a path  $\pi$  from cut-point  $i$  to cut-point  $j$ , which does not pass through any other cut-point, we add **edge**  $(i, j)$  to the graph and label it by the instruction that summarizes the effect of executing the path  $\pi$ .
- The **global variables** are efficiently modeled using input parameters.
- The representation of **functions** is straightforward.

# Example of Constant Propagation: Source



# Example of Constant Propagation: Target



# Assertion Network

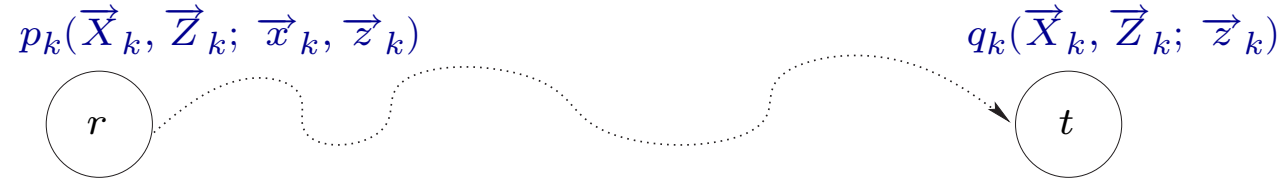
Let fictitious variables  $\vec{X}$  and  $\vec{Z}$  represent the values of the input variables  $\vec{x}$  and  $\vec{z}$  at the procedure entry.

An **assertion network**  $\mathcal{N} = \{\varphi_0, \dots, \varphi_n\}$  associates an assertion  $\varphi_l$  with each program location  $l$ :

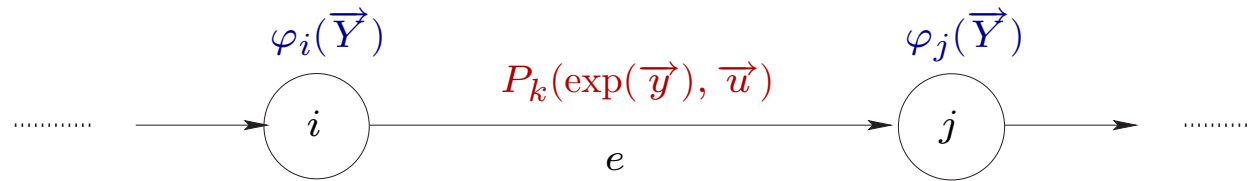
- For each module  $P_k$ , the **input predicate** associated with the procedure entry location is denoted by  $p_k(\vec{X}, \vec{Z}; \vec{x}, \vec{z})$ .
- The assertion associated with the exit location is denoted by  $q_k(\vec{X}, \vec{Z}; \vec{z})$ . The **output predicate**  $q_k$  is the procedure summary: it specifies the relation between the input and output values.
- The assertions at all other locations of the procedure  $\varphi_l(\vec{Y})$  may depend on any of the variables.

# Verification Conditions: Procedure Call

$$\underline{P_k(in : \vec{x}_k; \vec{z}_k)}$$



$$\underline{P(in : \vec{x}; \vec{z})}$$



We associate the following two conditions with a procedure call  $P_k(\text{exp}(\vec{y}), \vec{u})$ :

$$\mathcal{VC}_{call} : \quad \varphi_i(\vec{Y}) \rightarrow p_k(\text{exp}(\vec{y}), \vec{u}; \text{exp}(\vec{y}), \vec{u})$$

$$\mathcal{VC}_{return} : \quad \varphi_i(\vec{Y}) \wedge q_k(\text{exp}(\vec{y}), \vec{u}; \vec{z}_k) \rightarrow \varphi_j(\vec{Y})[\vec{u} \mapsto \vec{z}_k]$$

# Inductive Assertion Network

- An **assertion network**  $\mathcal{N}$  for a program  $\mathcal{A}$  is said to be **inductive** if all the verification conditions for all edges in  $\mathcal{A}$  are valid.
- Every inductive network is **invariant**.

Our goal is to construct an **inductive assertion network** for the source program that would be strong enough to prove **translation** in presence of **context-sensitive constant propagation**.

Since the interprocedural constant propagation is not a trivial problem, we compute the inductive network utilizing the **existing dataflow analysis**.

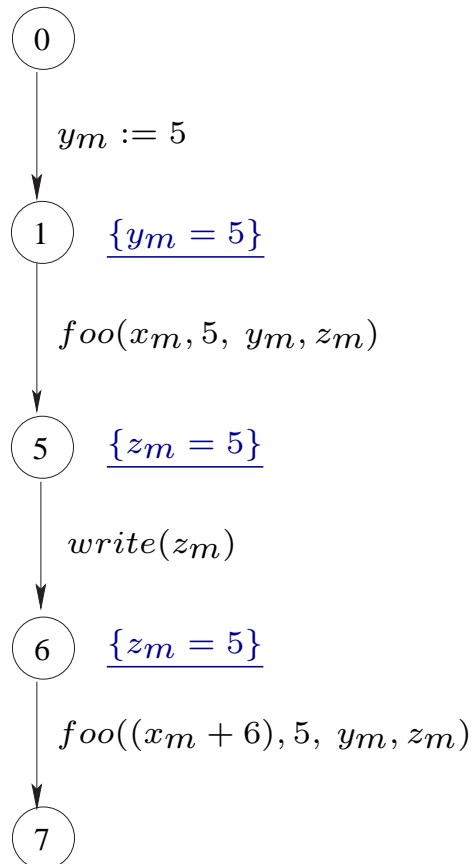
## First Attempt

On the first thought, any precise **solution** to the interprocedural constant propagation problem should suffice:  $\varphi_l$  should be extended with conjunct  $x = 5$  if  $x$  always evaluates to constant **5** at location  $l$ .

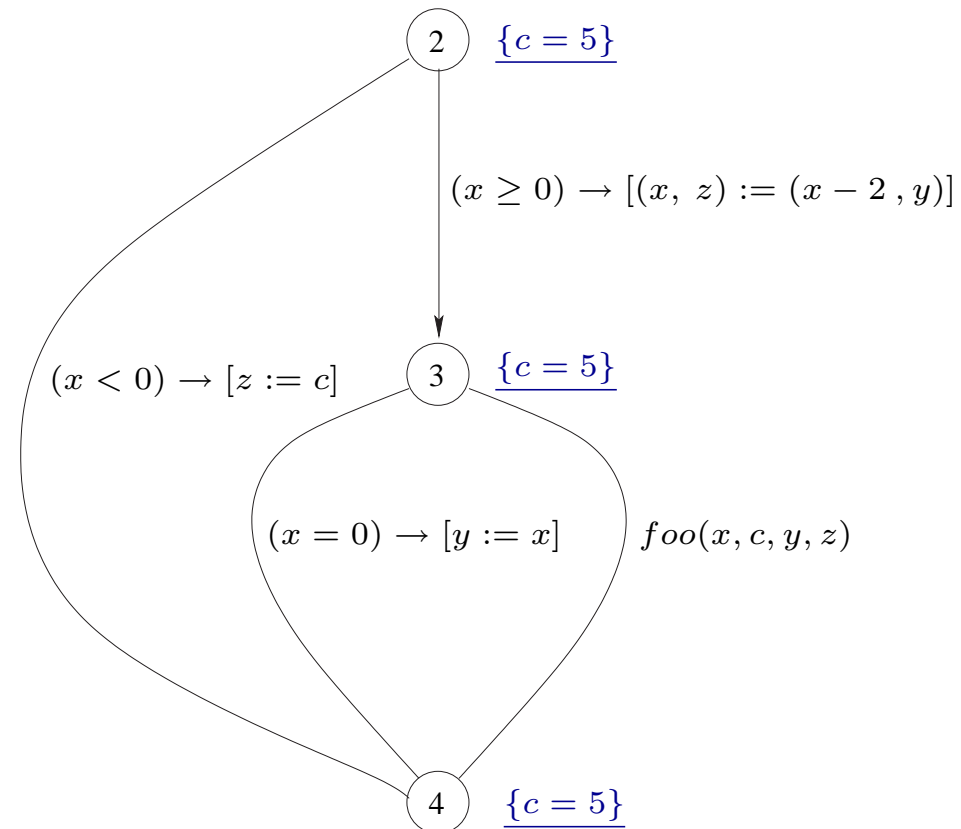
However, the resulting **assertion network**  $\mathcal{N}$  may **not** be **inductive**.

# Back to our Example

main()



foo(x, c; y, z)



## Solution

As our interprocedural dataflow analysis algorithm, we are going to use the one presented in:

- Precise Interprocedural Dataflow Analysis with Applications to Constant Propagation by M. Sagiv, T. Reps, and S. Horwitz.
- Precise Interprocedural Dataflow Analysis via Graph Reachability by M. Sagiv, T. Reps, and S. Horwitz.
- Two Approaches to Interprocedural Dataflow Analysis (Functional Approach) by M. Sharir and A. Pnueli.

The interprocedural dataflow analysis algorithm not only provides a solution to the problem, but also finds a fixpoint for the corresponding set of dataflow equations. Intuitively, we are going to use the information about the fixpoint itself to strengthen our network so it would be inductive.

# Precise Interprocedural Dataflow Analysis with Applications to Constant Propagation

- Solves **copy constant propagation** and **linear constant propagation** (both distributive problems);
- Produces **precise** ("meet-over-all-valid-paths" solution) results for both recursive and non-recursive programs;
- The **dynamic-programming algorithm** solves these constant propagation problems in polynomial time  $O(E * MaxVisible^3)$ ;
- Uses a **compact representation** of distributive dataflow functions ("environment transformers").

# Distributive Constant Propagation Problems

- In **copy constant propagation**, a variable  $x$  is discovered to be 10 only if
  - $x := 10$
  - $y := 10; x := y$
- In **linear constant propagation**, a variable  $x$  is discovered to be 10 only if
  - $x := 10$
  - $y := 3; x := 2 * y + 4$

## Environment Transformer

- Let  $V$  be a finite set of **program variables**. Let  $L = Z \perp \top$ .
- The set  $Env(V, L)$  of **environments** is the set of functions from  $V$  to  $L$ .
- A mapping  $T : Env(V, L) \mapsto Env(V, L)$  is called an **environment transformer**.

For an environment  $env \in Env(V, L)$ , if  $env(v) \in Z$  then the variable  $v$  has a known constant value in the environment  $env$ ; the value  $\perp$  denotes non constant, and  $\top$  denotes an unknown value.

A **dataflow problem** is specified by annotating each edge of the supergraph of  $\mathcal{A}$  with an environment transformer. We are interested in meet over all valid paths solution.

# Environment Transformers for Constant Propagation Problems

- Linear constant propagation:

$$\textit{operation} \quad T : Env(V, L) \mapsto Env(V, L)$$

$$x := c \quad \lambda env. env[x \rightarrow c]$$

$$x := c_1 * y + c_2 \quad \lambda env. env[x \rightarrow c_1 * env(y) + c_2]$$

$$x := y + z \quad \lambda env. env[x \rightarrow \perp]$$

- The last transformer is a safe approximation; the exact transformer  $\lambda env. env[x \rightarrow env(y) + env(z)]$  cannot be used in this framework because it is not distributive.
  - Consider two environments:  $env_1 : [y \rightarrow 3; z \rightarrow 2]$  and  $env_2 : [y \rightarrow 4; z \rightarrow 1]$
  - Distributivity:  $T(env_1 \sqcap env_2) = T(env_1) \sqcap T(env_2)$
  - $T_{x+y}(env_1 \sqcap env_2) = T_{x+y}([y \rightarrow \perp; z \rightarrow \perp]) = [x \rightarrow \perp + \perp] = [x \rightarrow \perp]$
  - $T_{x+y}(env_1) \sqcap T_{x+y}(env_2) = [x \rightarrow 3 + 2] \sqcap [x \rightarrow 4 + 1] = [x \rightarrow 5]$

## Pointwise Representation of Environment Transformers

Every **distributive** transformer  $T : Env(V, L) \mapsto Env(V, L)$  can be represented using a set of functions:

$$F^T = \{f_{v,v'} \mid v, v' \in V \cup \{\Lambda\}\}, \text{ each of type } L \mapsto L.$$

- Function  $f_{v,v'}$  captures the effect that the value of variable  $v$  in the argument environment has on the value of  $v'$  in the result environment; if  $v'$  does not depend on  $v$ , then  $f_{v,v'} = \lambda.l.\top$ .
- Function  $f_{\Lambda,v'}$  is used to represent the effects of on the variable  $v'$  that are independent of the argument environment.

For any symbol  $v'$ , the value  $T(env)(v')$  can be determined by taking the meet of the values of  $|V| + 1$  individual function applications:

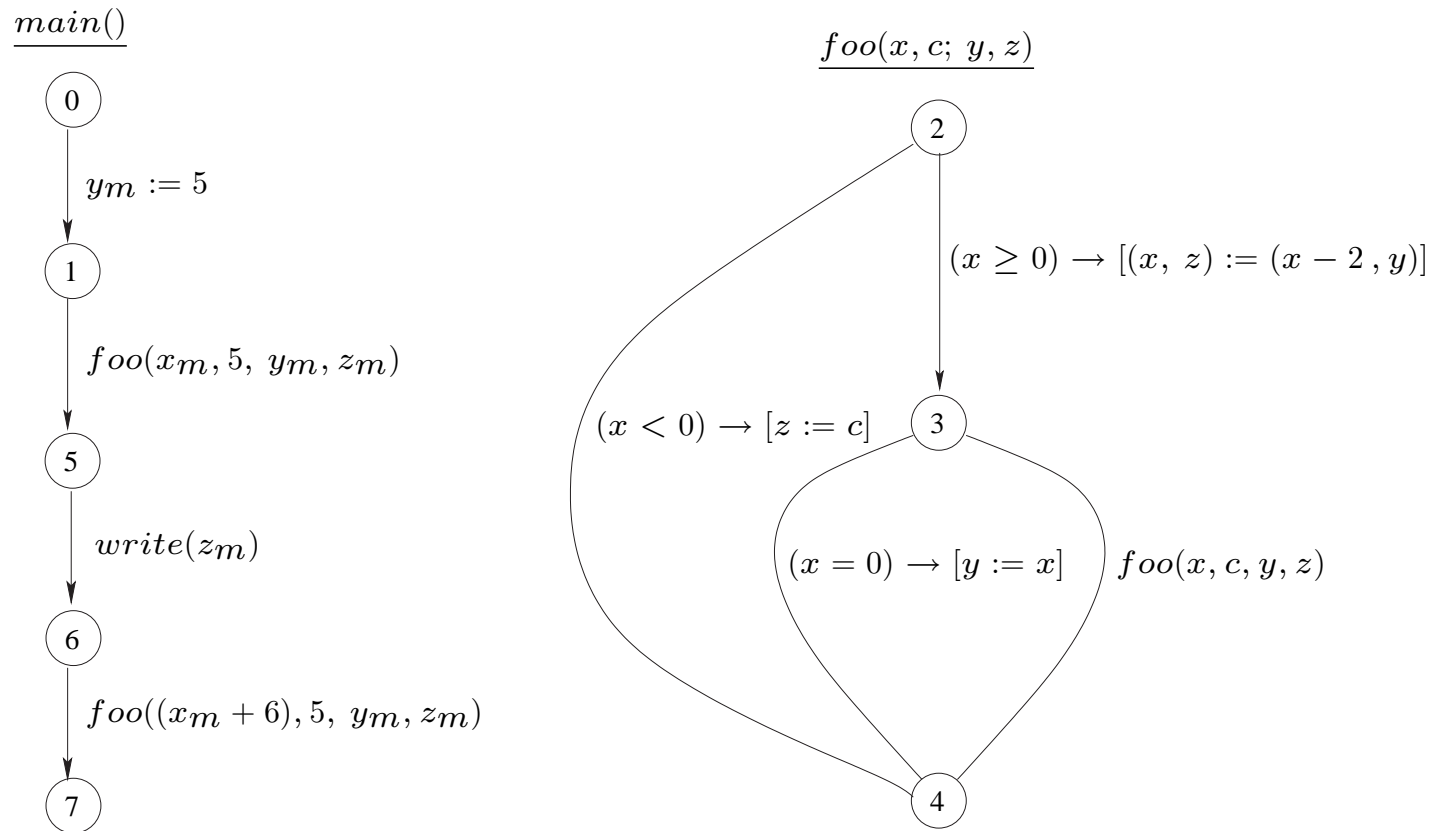
$$T(env)(v') = f_{\Lambda,v'} \sqcap (\sqcap_{v \in V} f_{v,v'}((env)(v))).$$

# A Dynamic Programming Algorithm

Two phase algorithm:

- Compute the dataflow solution at the nodes of a procedure as a function of the initial values at the procedure entry - **path functions**.
- Compute the dataflow **values** at every point using the path functions.

# Phase One: environment transformers



Below is the list of environment transformers computed for procedure *foo*. We omit all the functions that evaluate to top  $f_{(v,v')} = \lambda l. \top$ .

$$F_{(2,2)} = \{ f_{x,x} = \lambda l.l, f_{c,c} = \lambda l.l, f_{y,y} = \lambda l.l, f_{z,z} = \lambda l.l \}$$

$$F_{(2,3)} = \{ f_{c,c} = \lambda l.l, f_{y,y} = \lambda l.l, f_{y,z} = \lambda l.l \}$$

$$F_{(2,4)} = \{ f_{c,c} = \lambda l.l, f_{c,z} = \lambda l.l, f_{y,z} = \lambda l.l \}$$

## Computation of Invariants using the Env Transformers

Given all dataflow facts and the transformer represented by  $F_{(i,j)}$ , we follow the following rules to compute an invariant  $\varphi_l$  at location  $l$  of  $P_k$ :

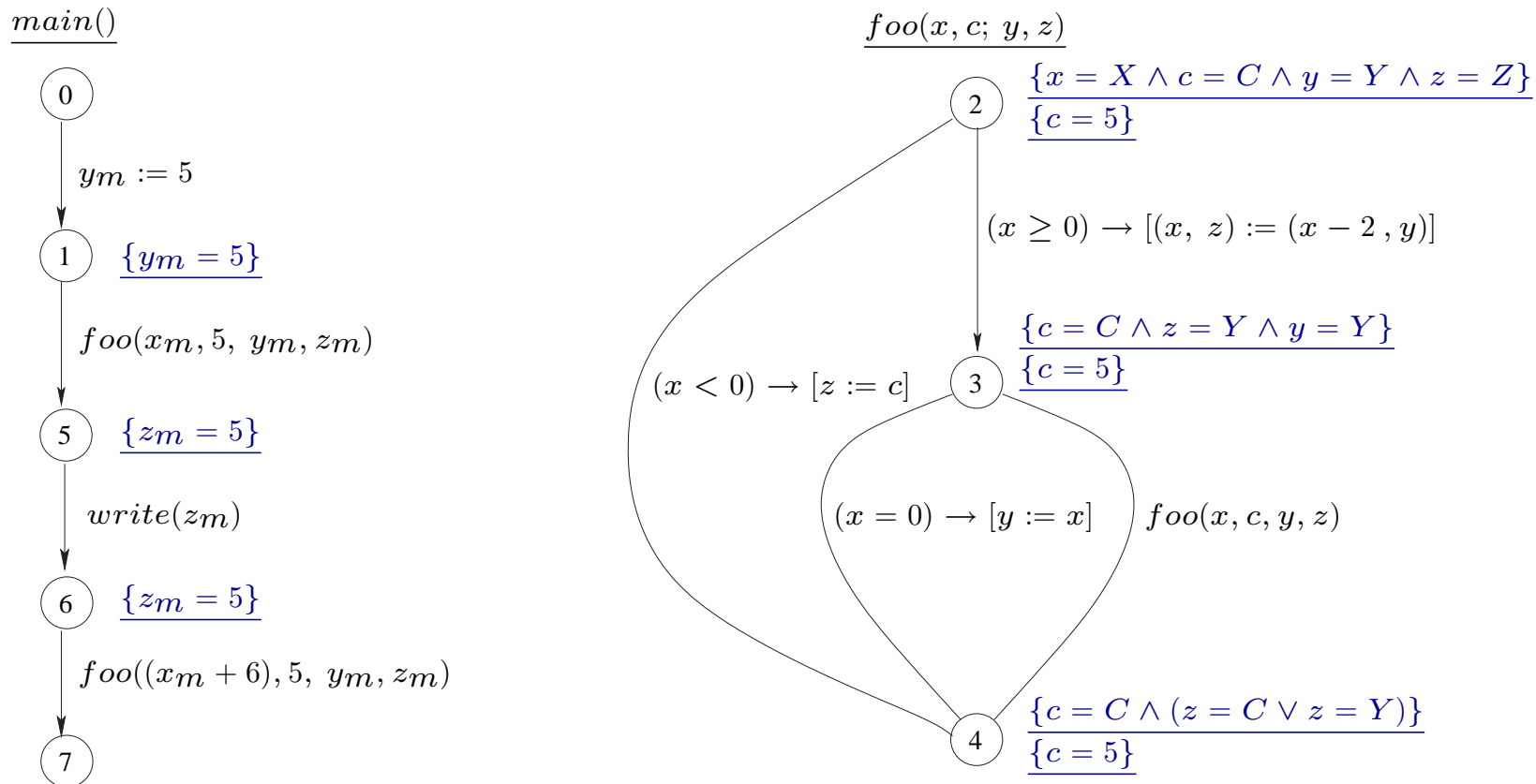
- We ignore all functions of the form  $f_{(v,v')} = \lambda l. \top$ .
- For each variable  $v'$  that is not set to  $\perp$  by  $f_{(\Lambda,v')} \in F_{(r_k,l)}$  we add the following conjunct to  $\varphi_l$ :

$$\bigvee_{f_{v,v'} \in F_{(r_k,l)}} v' = f_{v,v'}(V)$$

Note that we use **disjunction** to model the effect of the **meet operator**.

- We also add the conjunct  $x = \text{const}$  if  $x$  was determined to evaluate to constant  $\text{const}$  at location  $l$ . We need this addition since  $T_{(r_k,l)}$  does not propagate the the information from the callers.

# Resulting Inductive Assertion Network



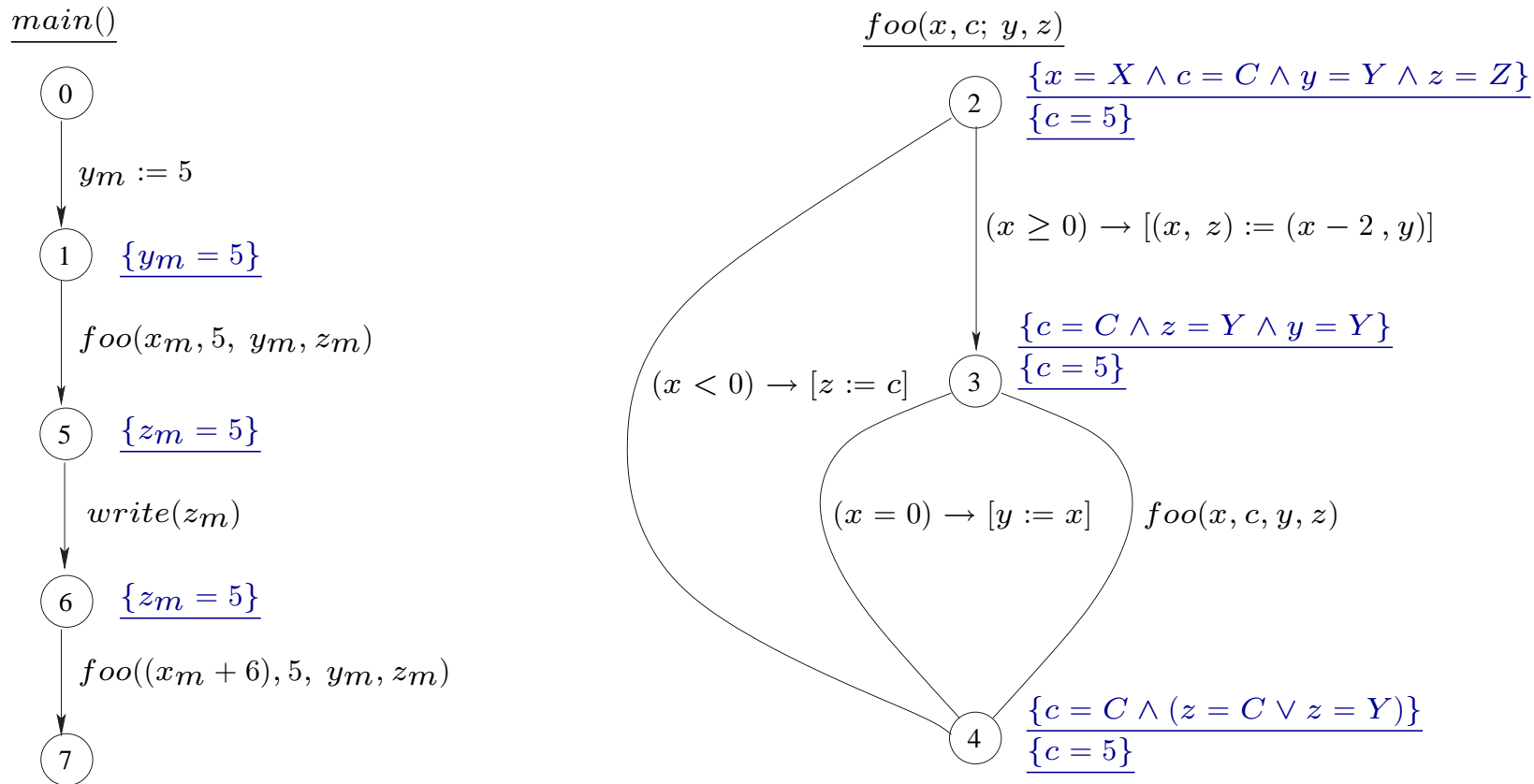
The list of environment transformers computed for procedure *foo*:

$$F_{(2,2)} = \{ f_{x,x} = \lambda l.l, f_{c,c} = \lambda l.l, f_{y,y} = \lambda l.l, f_{z,z} = \lambda l.l \}$$

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# Resulting Inductive Assertion Network



For example, let's show that the return verification condition for call edge (1, 5) of our example holds.

$$\begin{array}{lclclclclcl}
 \mathcal{VC}_{ret}: & \varphi_1 & \wedge & \varphi_4[(C, Y) \mapsto (5, y_m)] & \rightarrow & \varphi_5[z_m \mapsto z] & \Leftrightarrow \\
 & y_m = 5 & \wedge & c = 5 \wedge (z = 5 \vee z = y_m) \wedge c = 5 & \rightarrow & z = 5 & 
 \end{array}$$

Questions?