Using Widenings/Narrowings in Data Flow Analyses

An introduction

Vesal Maynard Vojdani

vesal@ut.ee

Tartu University

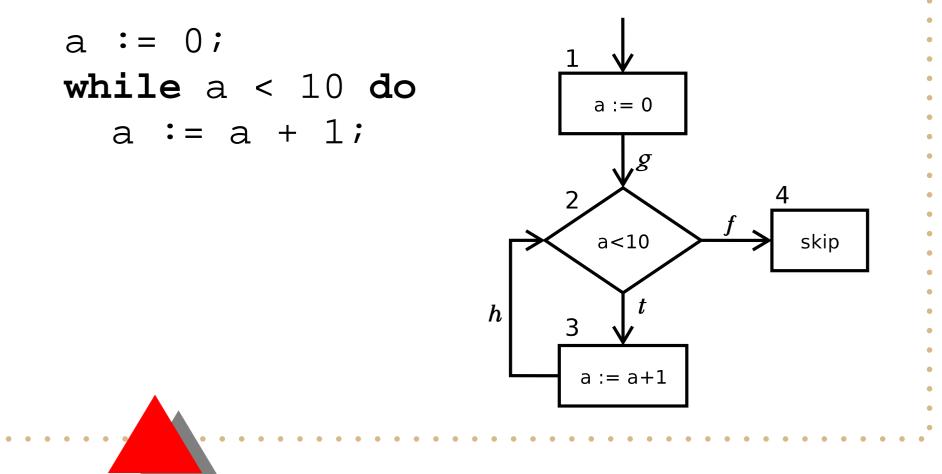
Data Flow Analysis – p. 1/31

Overview

- Introduction ← you are here!
- Galois Connections
- From MOP to MFP
- Insufficiency of Galois connections
- Widenings/Narrowings
- Examples

Interval Analysis

• The goal is to do an interval analysis on the following simple program.



Standard Semantics

- The standard semantics of a programming language defines how expression modify the state.
- The semantics is embodied in the transfer functions of the graph.
- The analysis must take into account
 - all possible input states
 - all possible executions

· · · · · · · · · · · · · · · · · · ·	
	•
	•
	•
	•
	•
	•
	•
	•
	•
	•
	•
A 11 • 1 1 • ,	•
All possible inputs	
	•
	•
	•
	•
	•
	•
	•
	•
	•
	•
	•
	•
	•
	•
	•
	•
	•
· · · · · · · · · · · · · · · · · · ·	• • • • • • •
	Data Flow Analysis – p. 5/3

Sets of States Semantics

- To analyse to program for all possible input states, we have to use the Set of States semantics.
- It is naturally derrived from the Standard semantics.
- Most precise domain, but impossible to use.
- Approximation is necessary.

The Lattice ordering

- Interesting properties are undecidable.
- Analyses must err on the safe side.
- The Lattice ordering: $x \sqsubseteq y$ means:
 - The set of programs satisfying x is included in the set of programs satisfying y.
 - The state y is a correct approximation of x.
 - The state x is more precise than y.

Galois connections

 Galois connection (more precisely adjunctions):

 $\alpha \colon X \to Y$ $\gamma \colon Y \to X$

total functions that satisfy

$$\alpha(x) \sqsubseteq y \iff x \sqsubseteq \gamma(y)$$

Meaning of Galois connections

$\alpha(x) \sqsubseteq y \iff x \sqsubseteq \gamma(y)$

- $\alpha(x)$ is the most precise approximation of x.
- γ(y) is the most general element, which can be soundly approximated by y.

Moving to the interval domain

- We can now abstract
 - Sets of State Semantics (visualize as points in space)
 - Collecting Semantics (abstract with projection, concretize as grid)
 - Interval domain (project to interval, concreetizes to rectangle)
- The transfer functions are induced by our abstraction:

$$f' = \alpha \circ f \circ \gamma$$

All possible executions

Notation

- The Control Flow Graph G = (N, E, s), where
 - N is the set of nodes.
 - $E = N \times N$
 - s is the initial node.
- The analysis is an annotation $N \rightarrow D$.
- The transfer functions $tf: E \to (D \to D)$
- ι is the initial state.

Merge Over all Paths

Definition 1. The path semantics $[\![\pi]\!]_{tf}$ is simply the composition of the transfer functions along that path

$$\llbracket \varepsilon \rrbracket_{tf} = id_{D \to D}$$
$$\llbracket e_1, \dots, e_n \rrbracket_{tf} = \llbracket e_2, \dots, e_n \rrbracket_{tf} \circ tf(e_1)$$

Definition 2. The MOP solution to the data flow problem is defined at each node by

$$\mathsf{MOP}(n) = \bigsqcup \left\{ \llbracket \pi \rrbracket_{tf} \left(\iota \right) \mid \pi \text{ is a path from } s \text{ to } n \right\}$$

Least Fixed-point

- MOP is not calculable.
- **Definition 3.** The least (minimal) fixed point MFP(n) is the least solution to the following system:

$$\mathsf{MFP}(n) = \begin{cases} \iota & \text{if } n = s \\ \bigsqcup \left\{ tf(e)(\mathsf{MFP}(n')) \mid e = (n', n) \in E \right\} & \text{otherwise} \end{cases}$$

This is correct with respect to MOP, i.e. $\forall n : MOP(n) \sqsubseteq MFP(n)$.

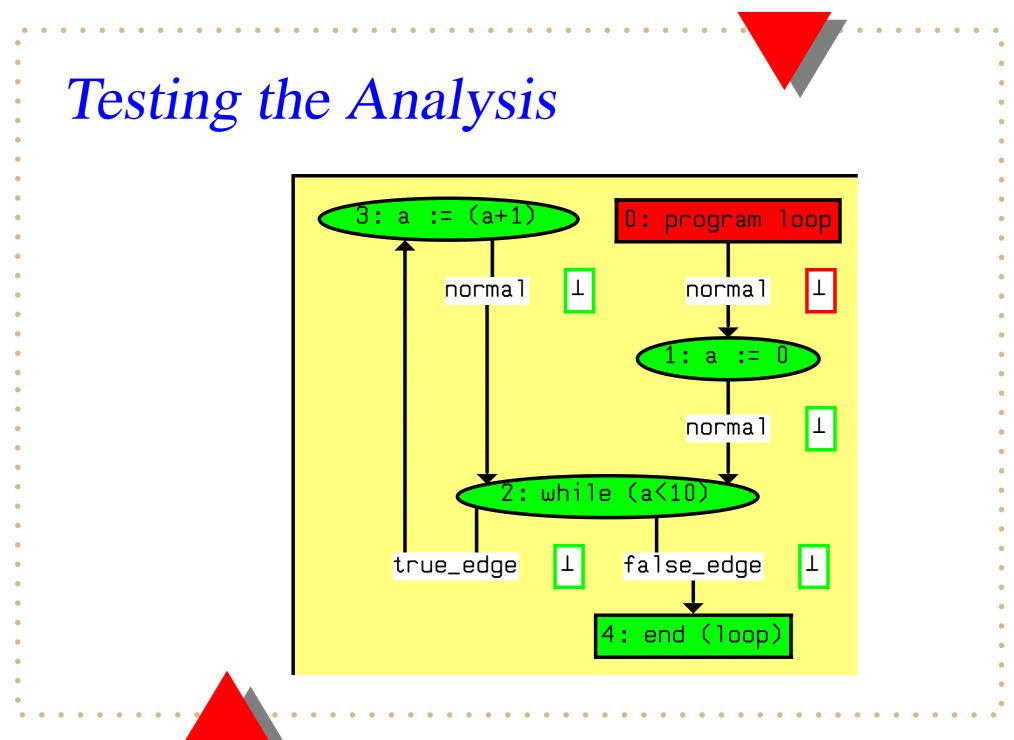
PAG

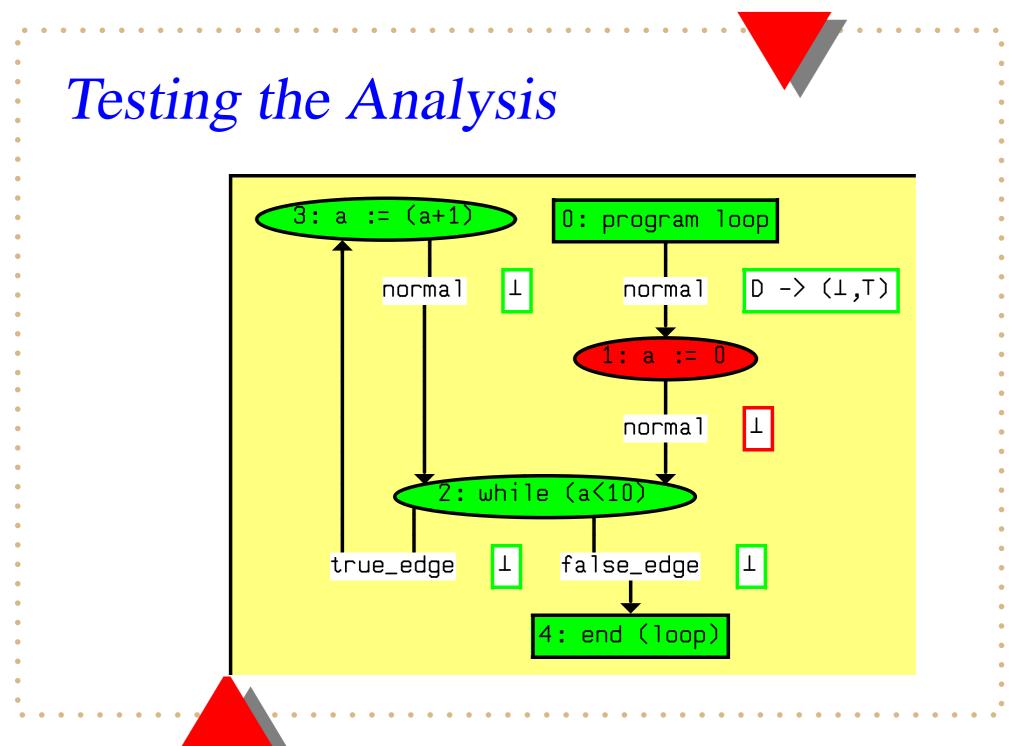
- The analysis is implemented in PAG a Program Analysis Generator. See www.absint.com for further information.
- The output represents functions.

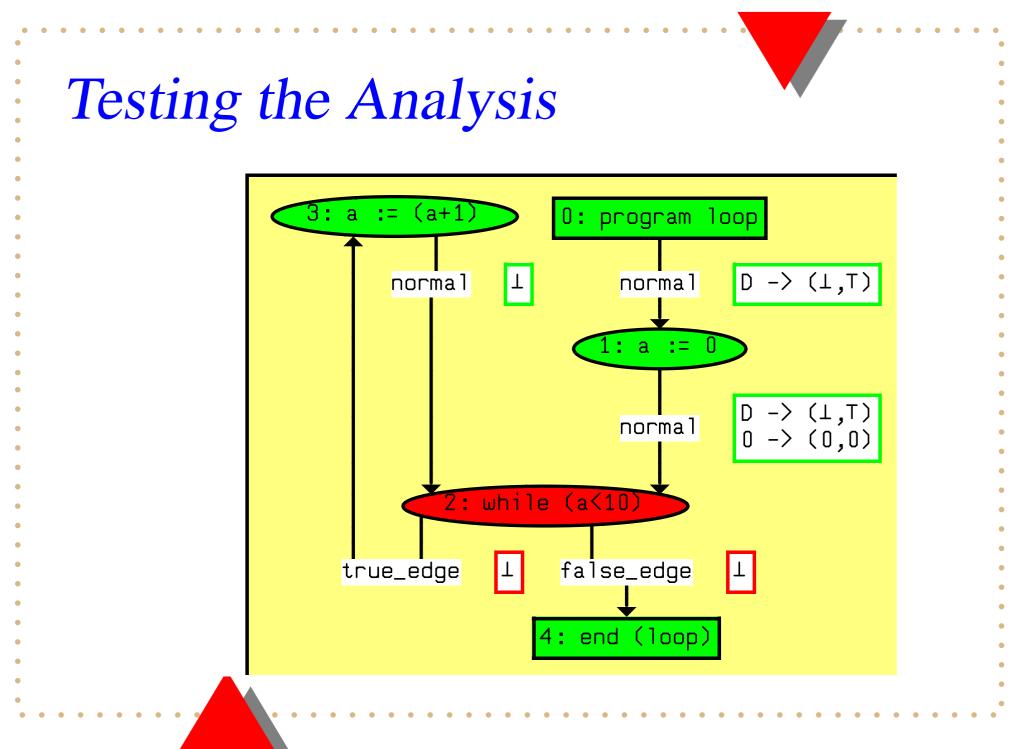
$$f(x) = \begin{cases} (0, 10) & \text{if } x = a\\ (-\infty, \infty) & \text{otherwise} \end{cases}$$

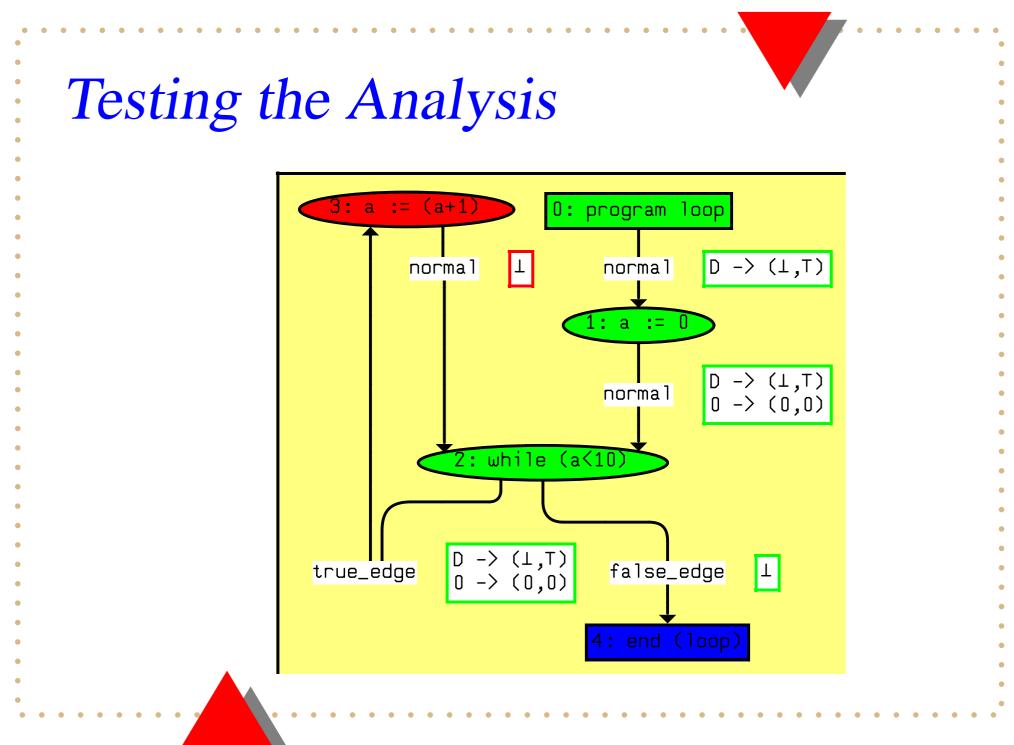
is written as

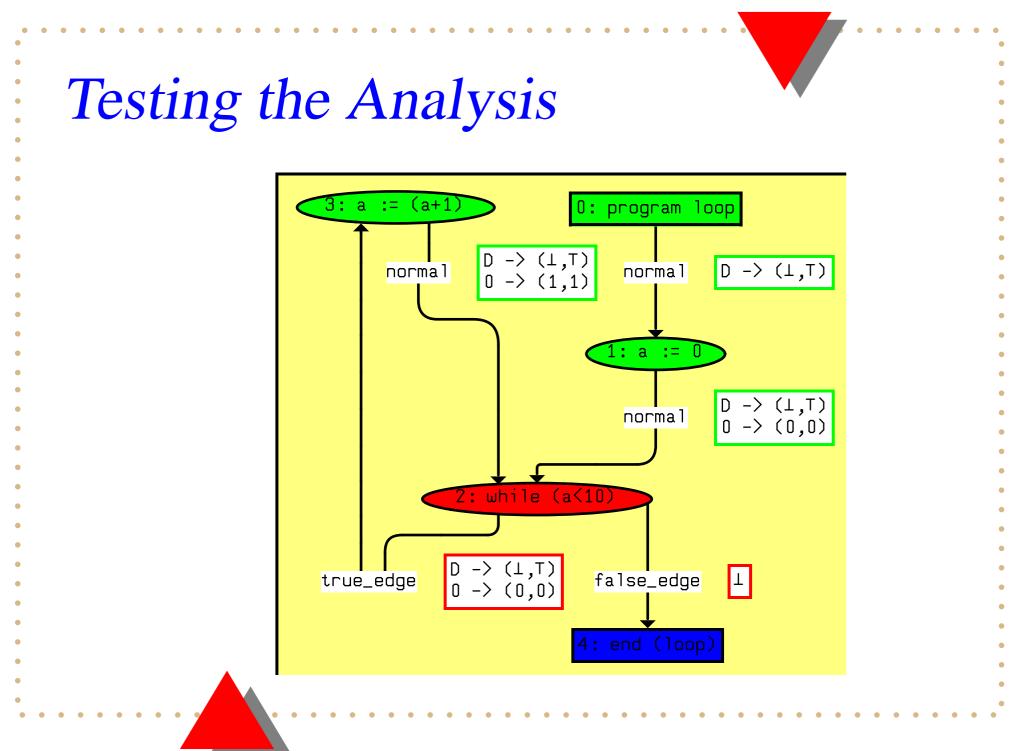
$$D \to (\bot, \top)$$
$$0 \to (0, 10)$$

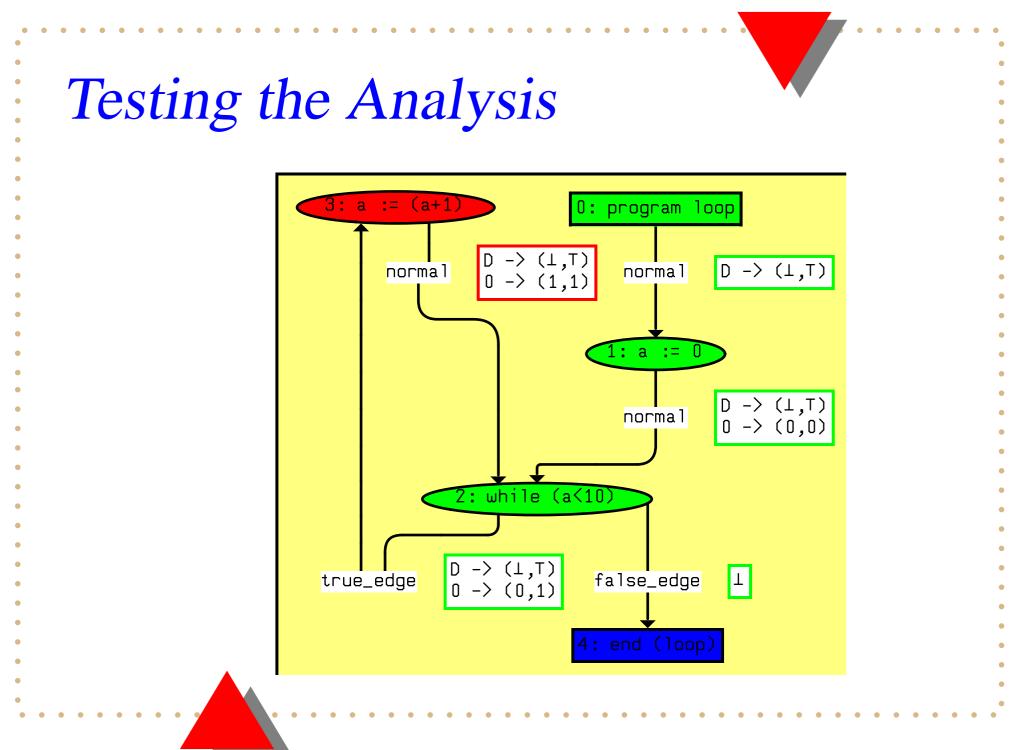


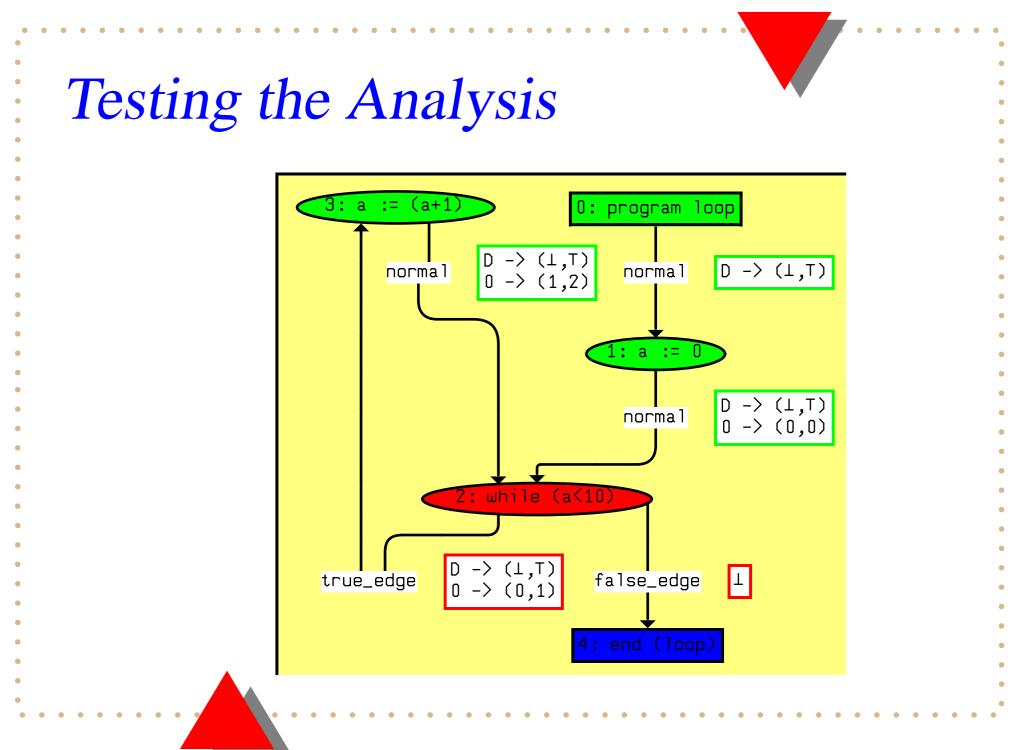


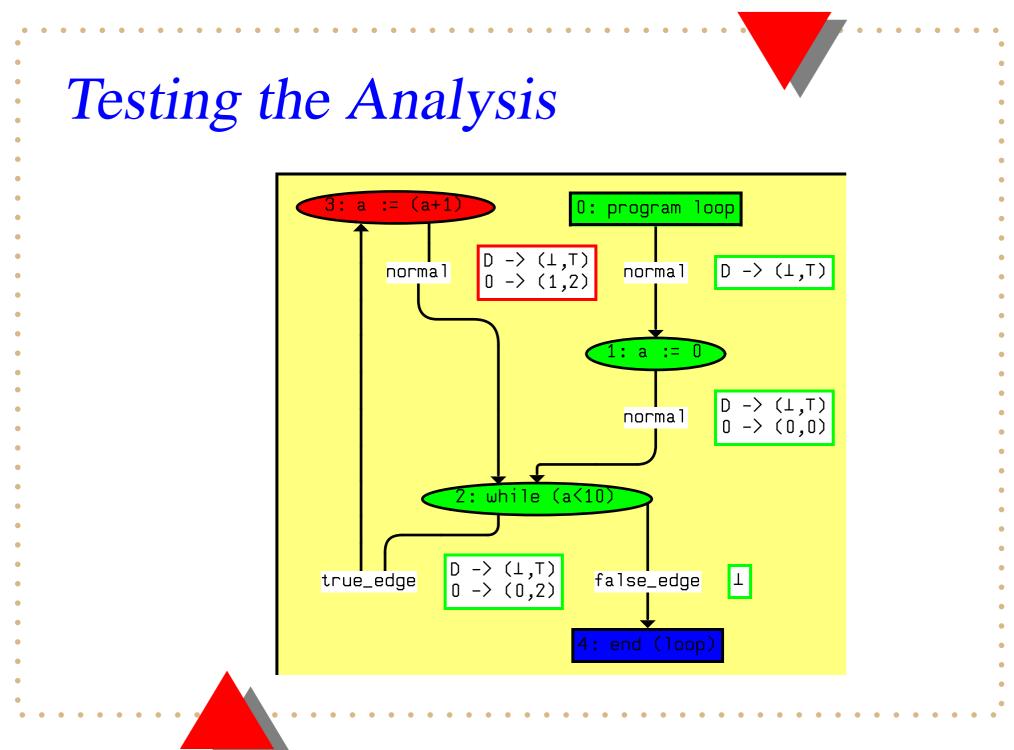


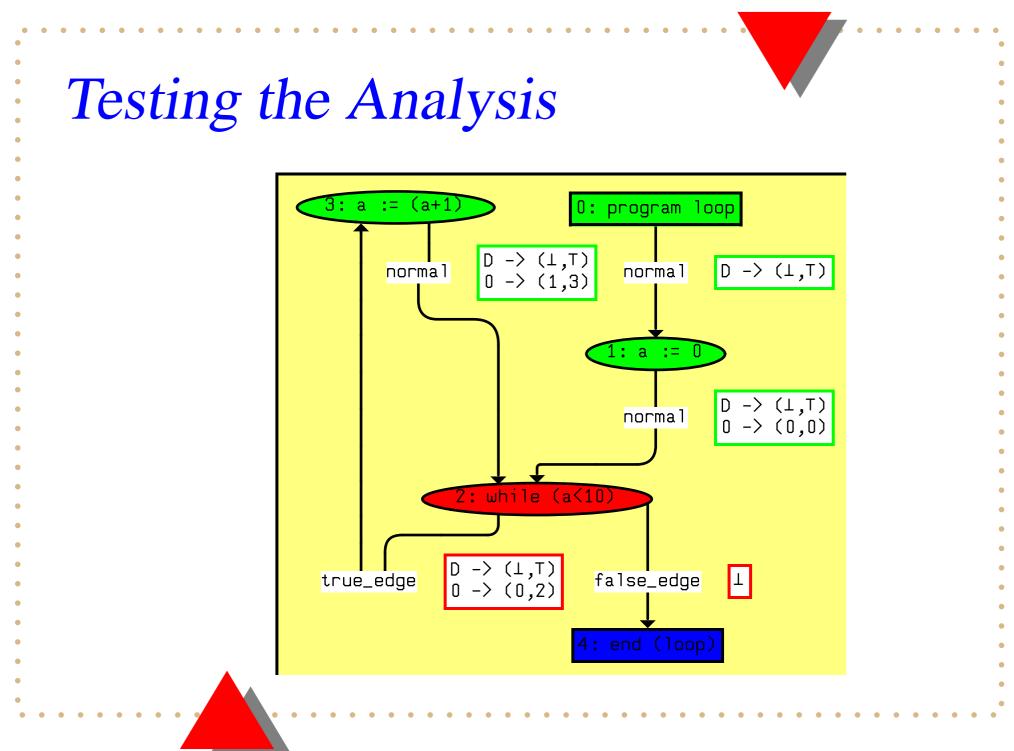


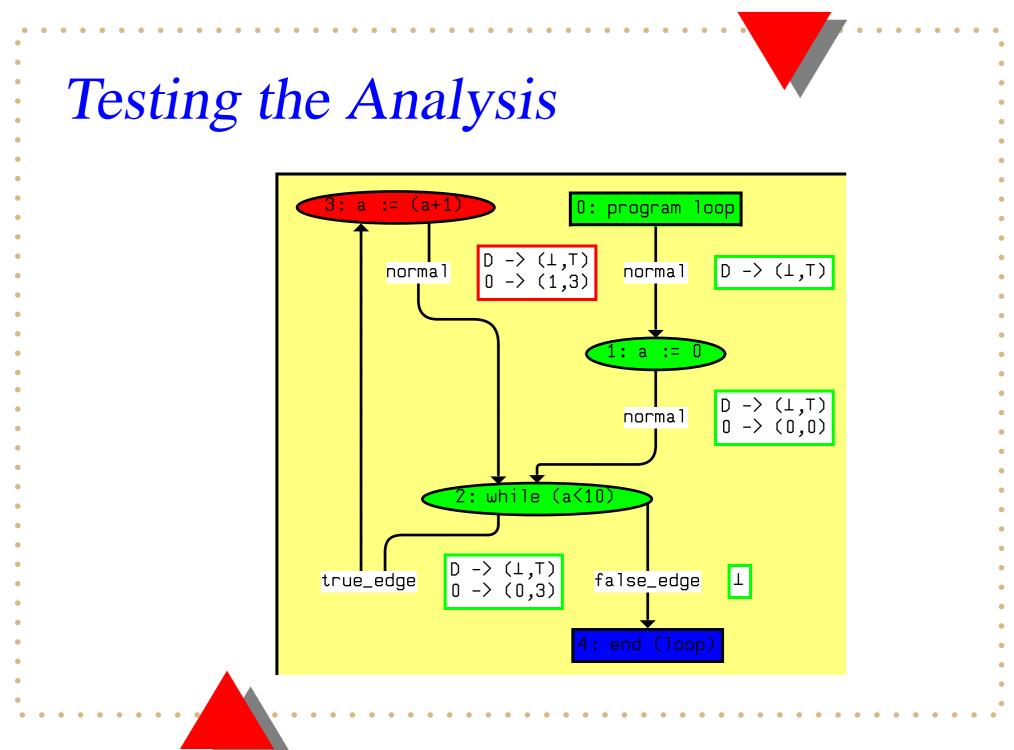


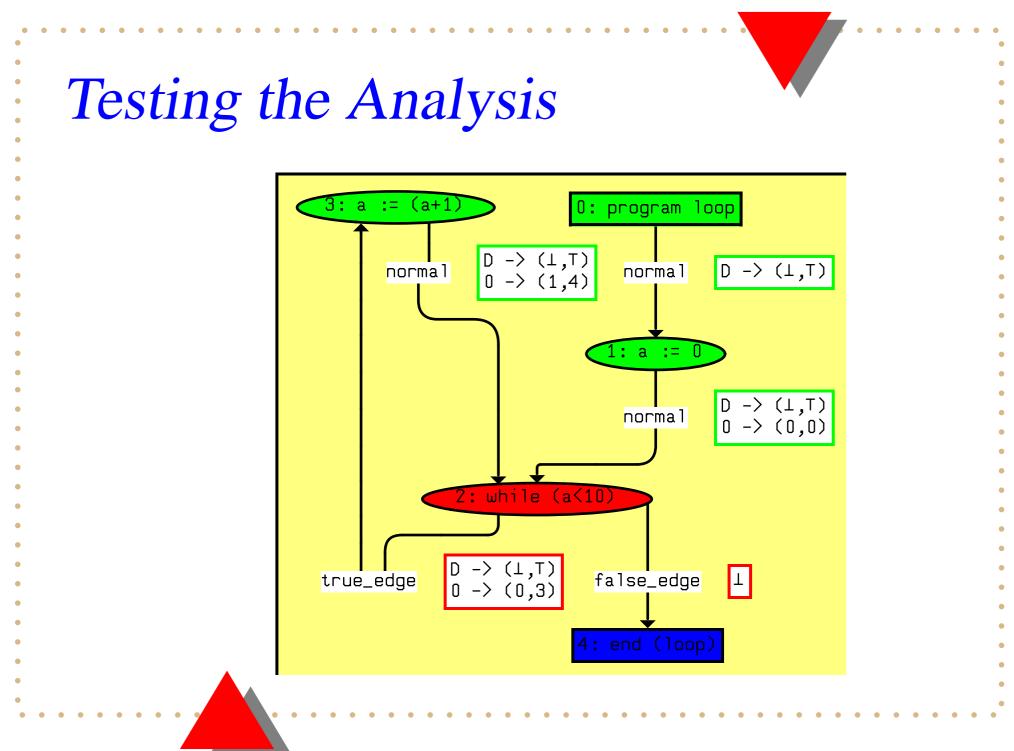


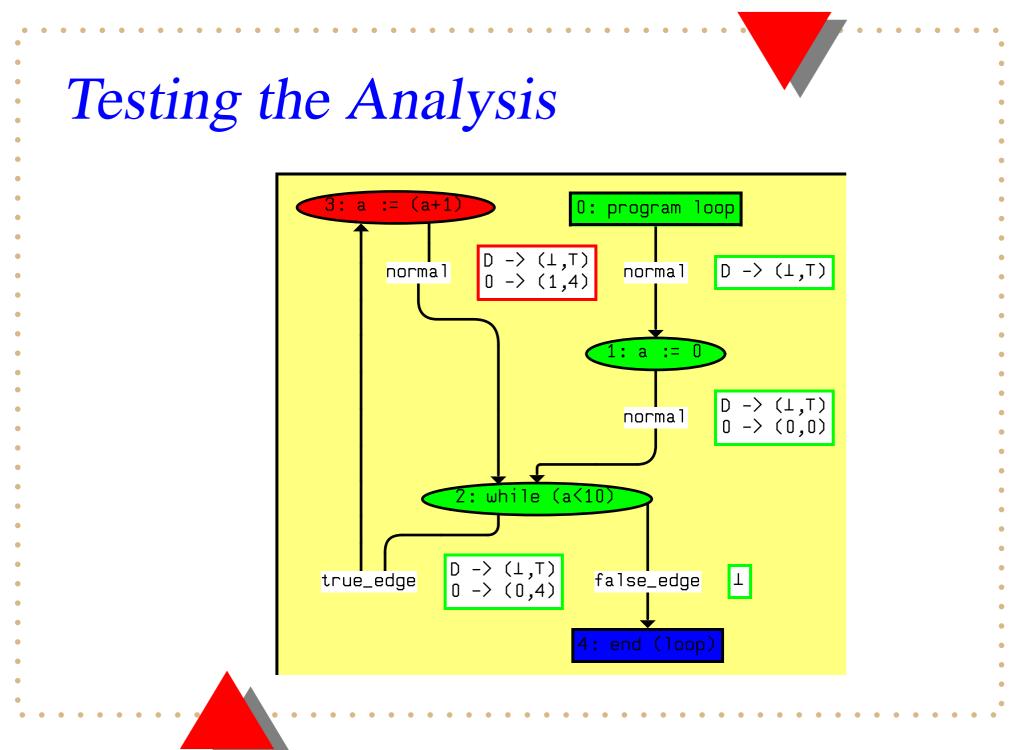


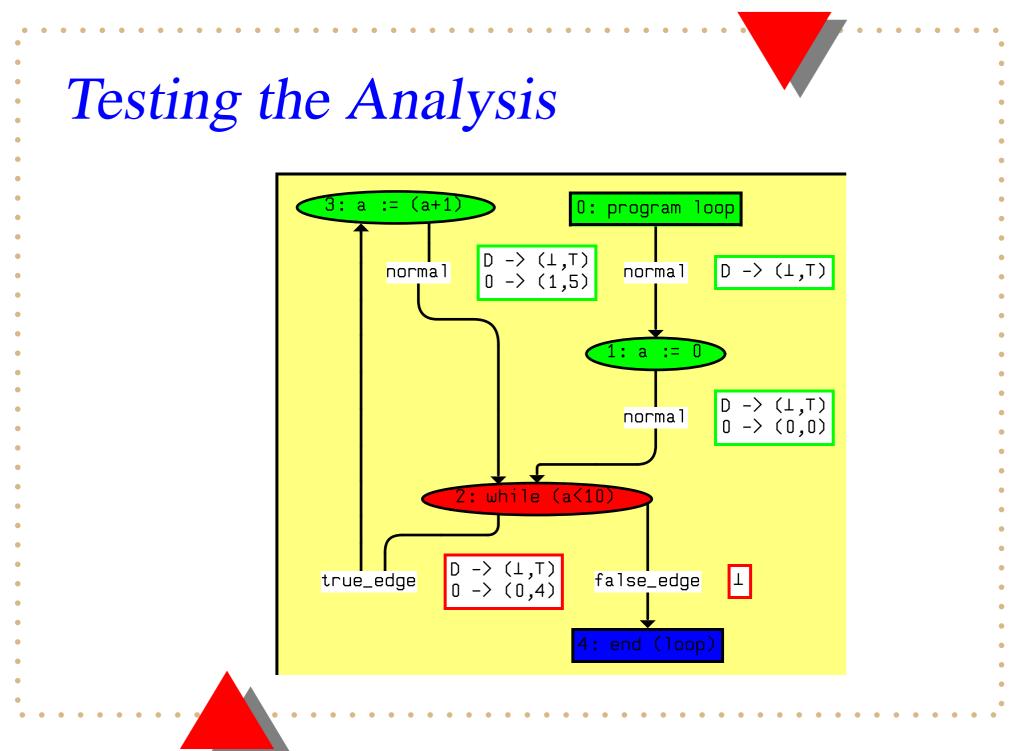


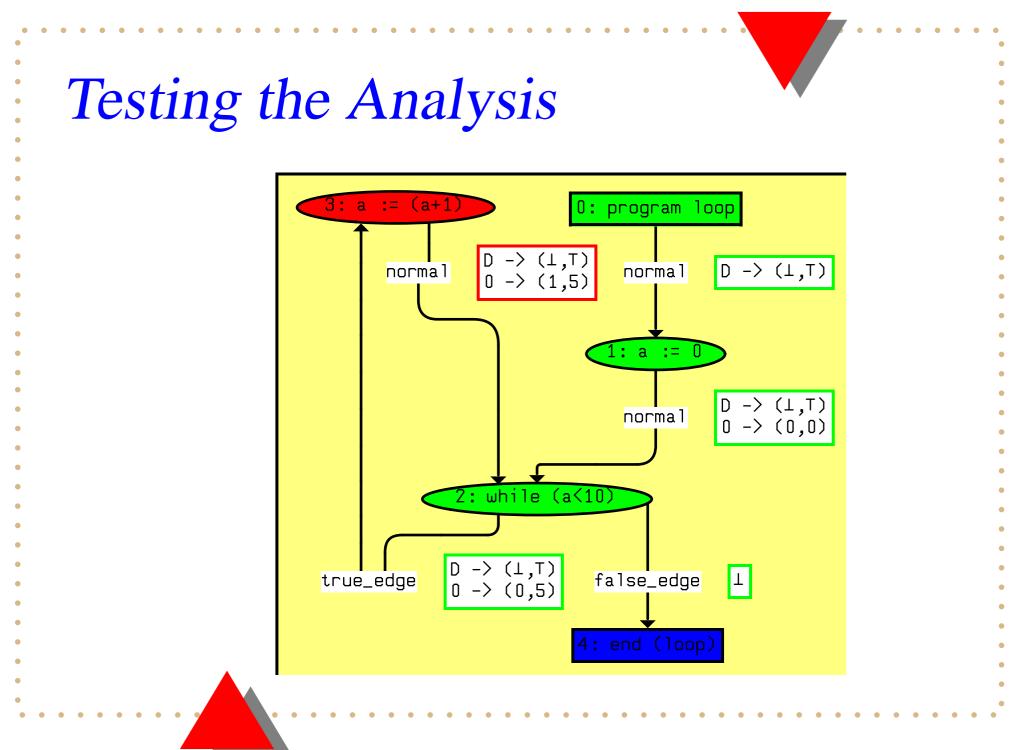


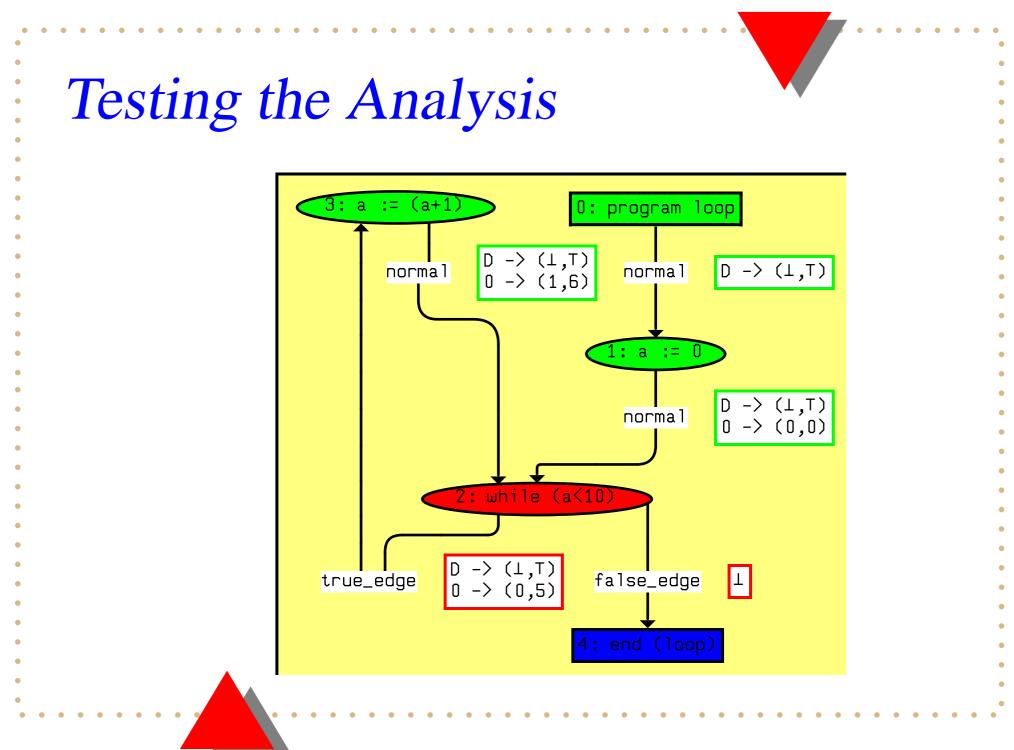


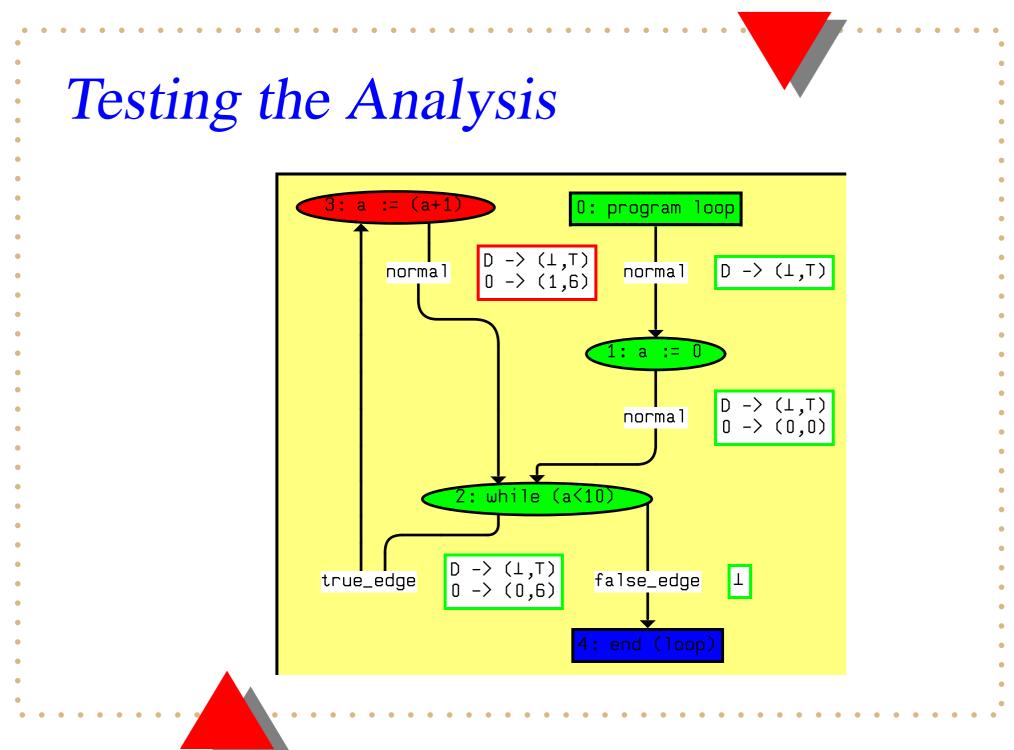


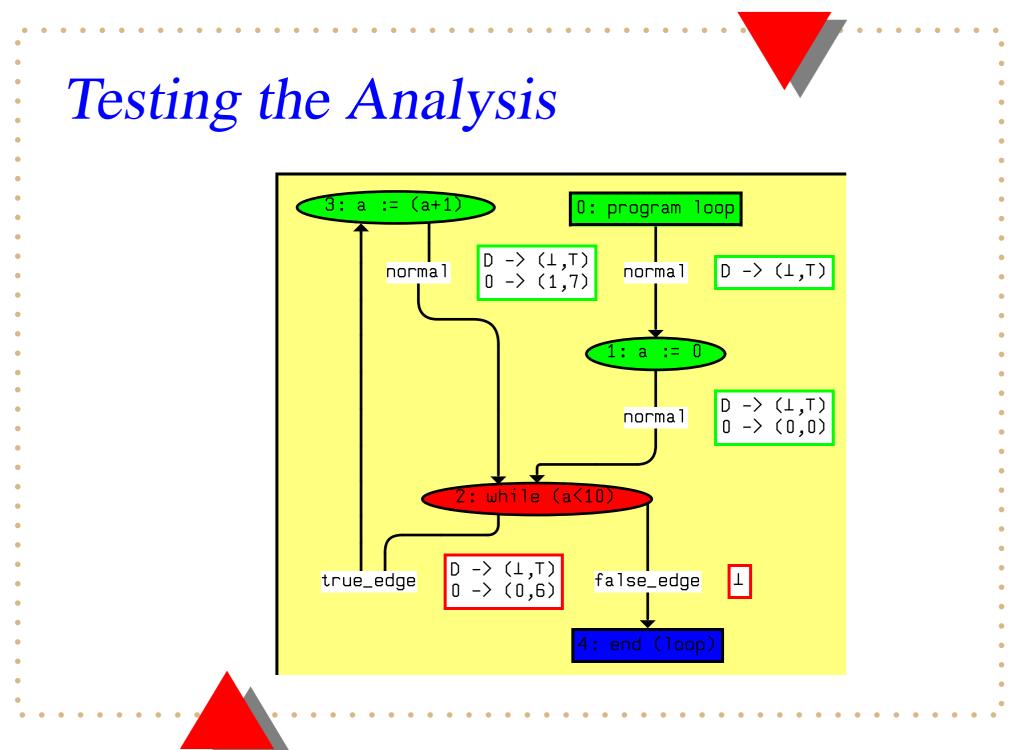


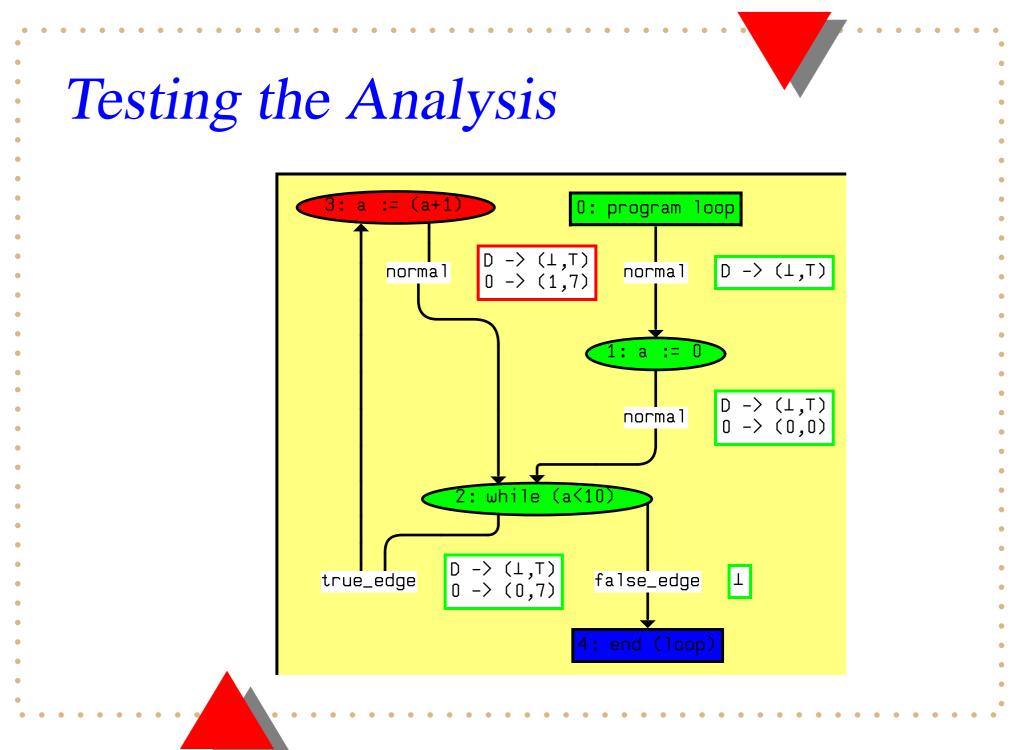


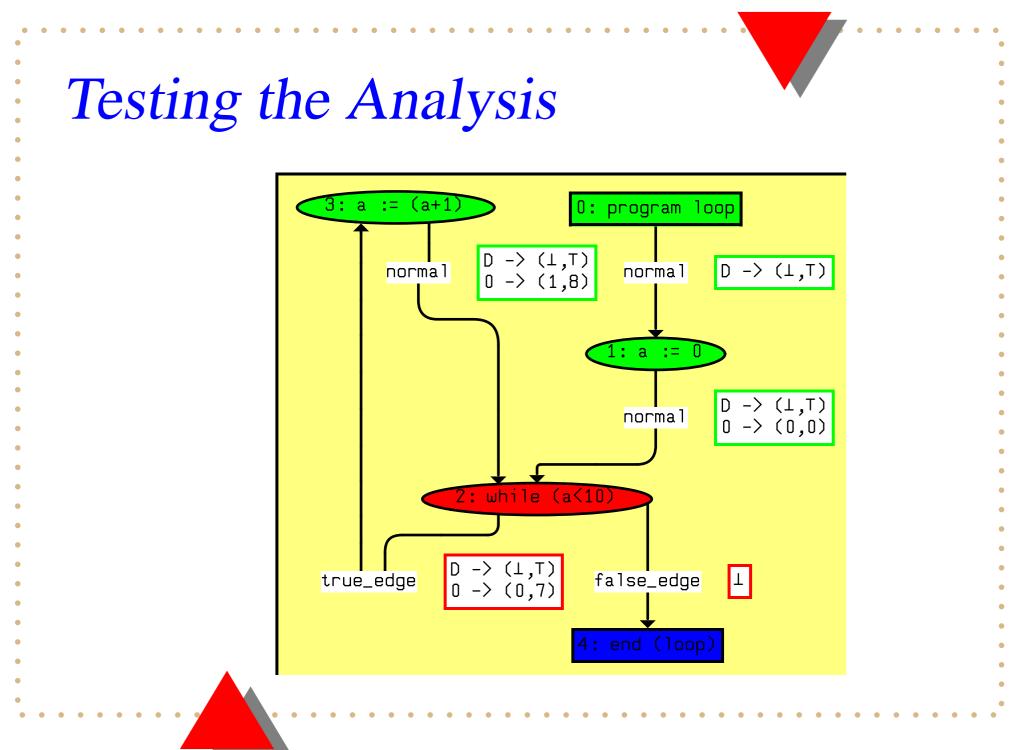


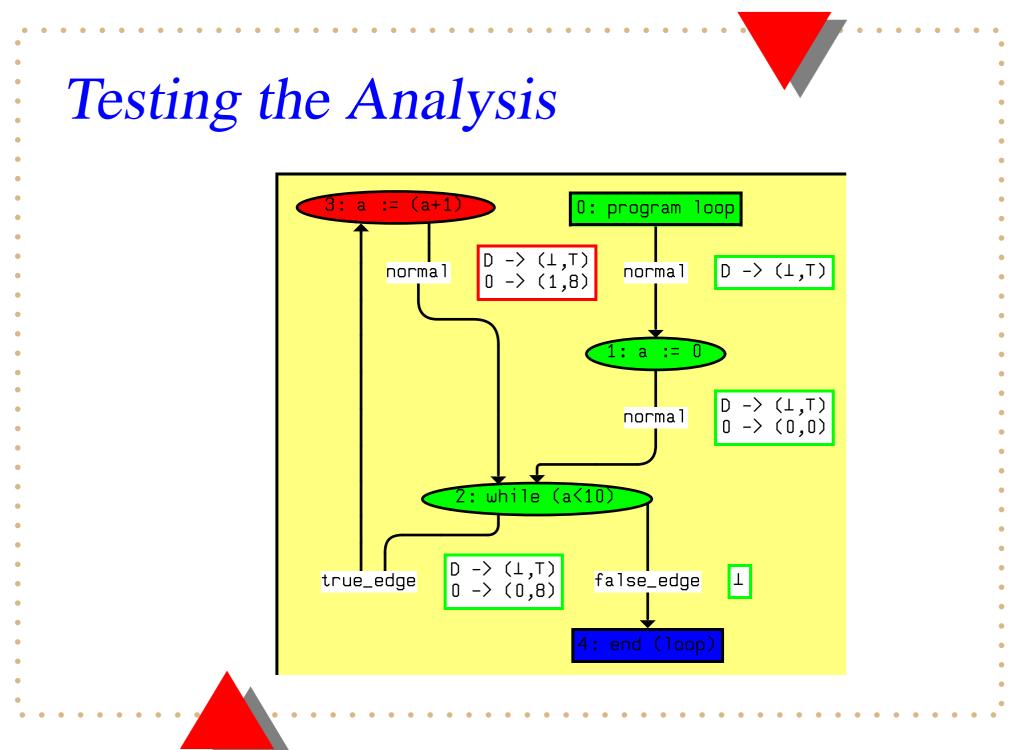


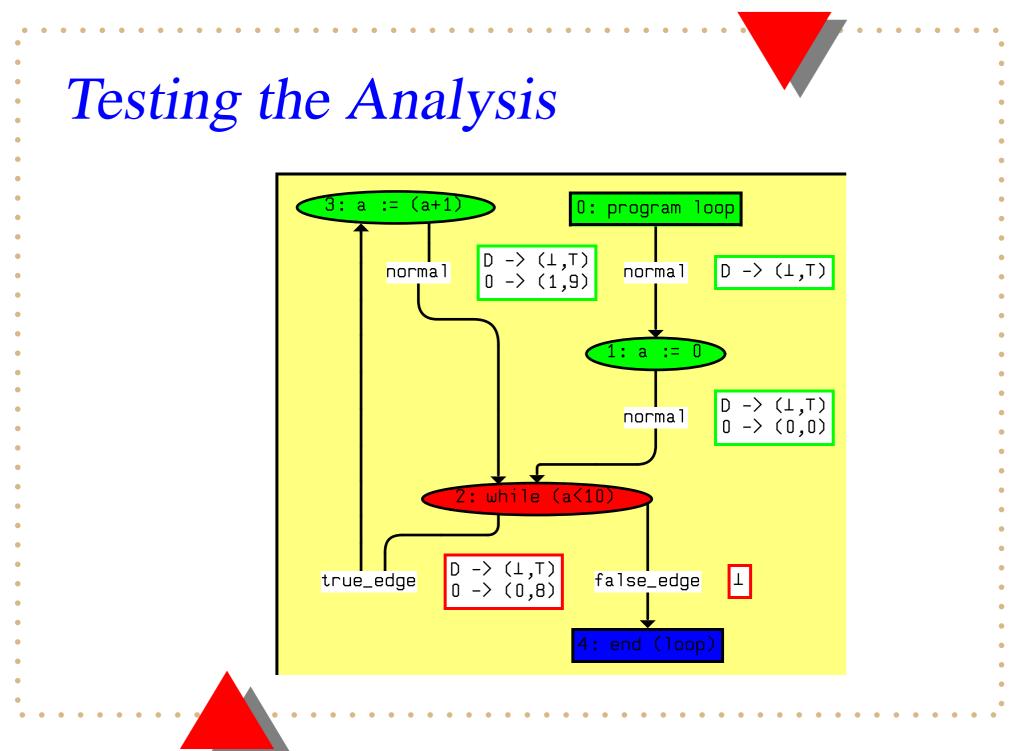


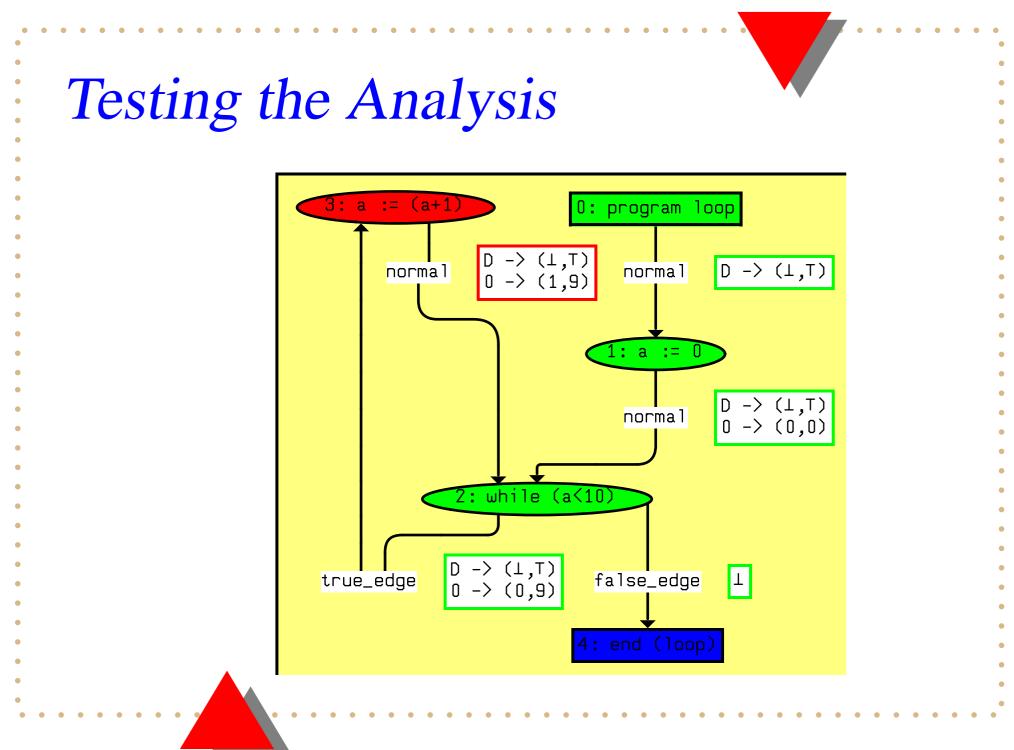


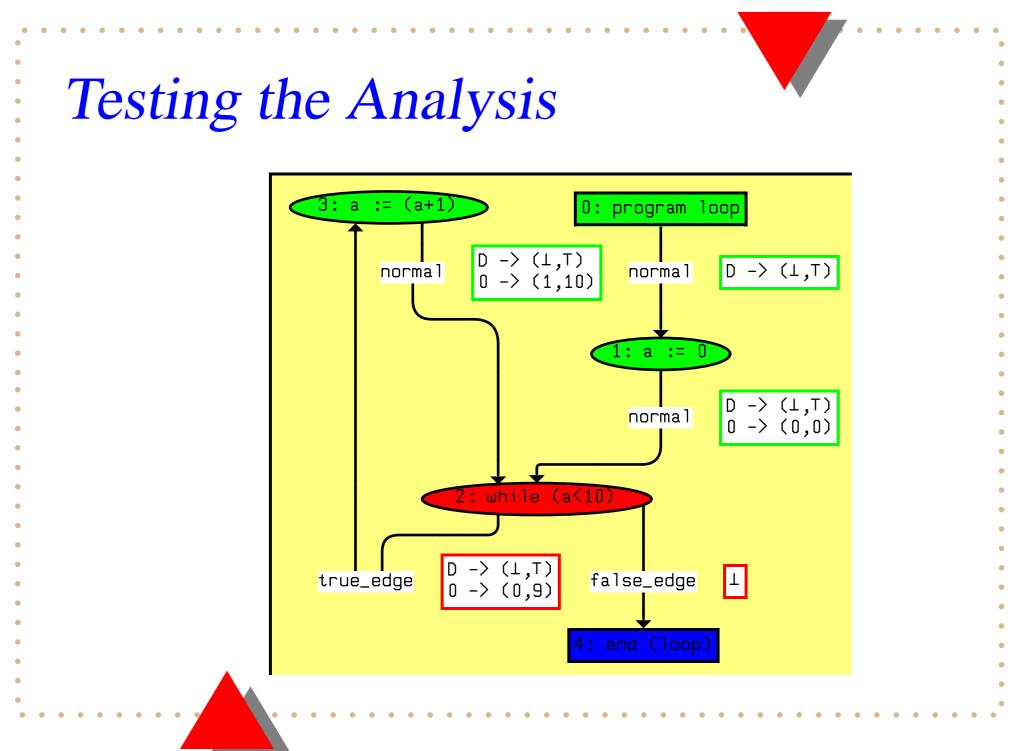


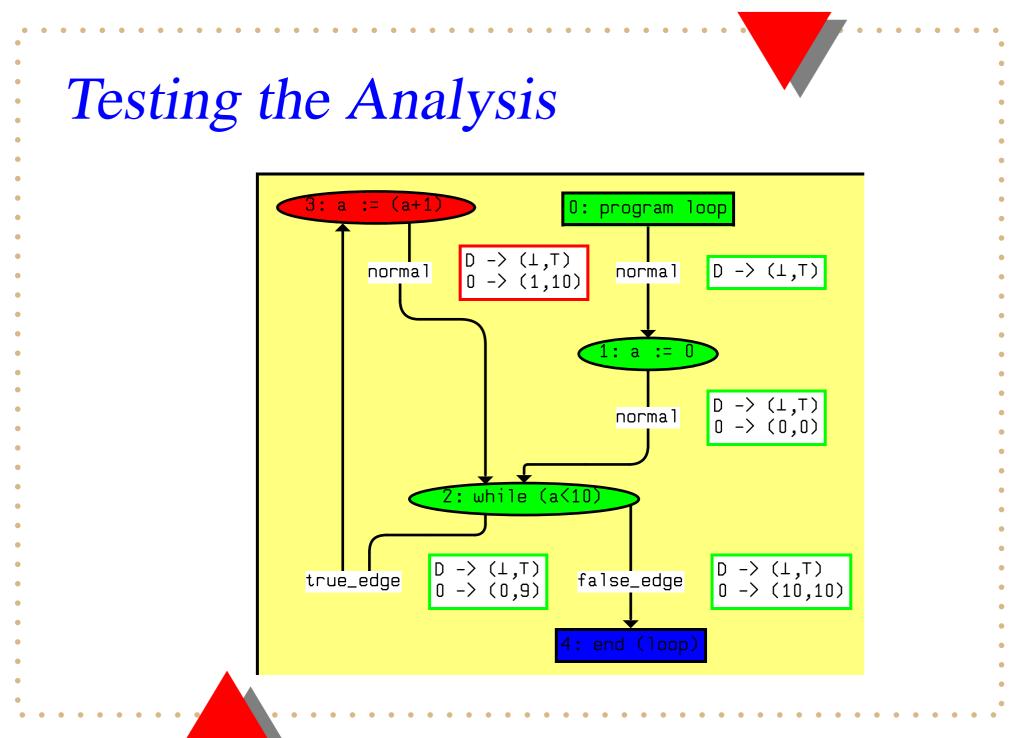


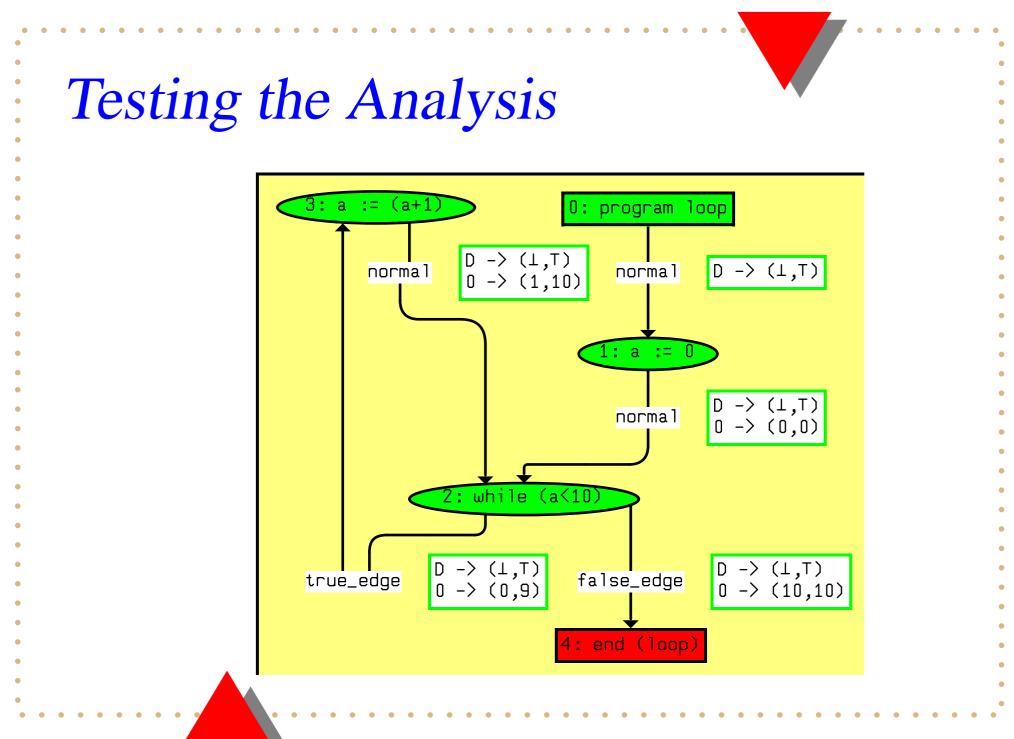


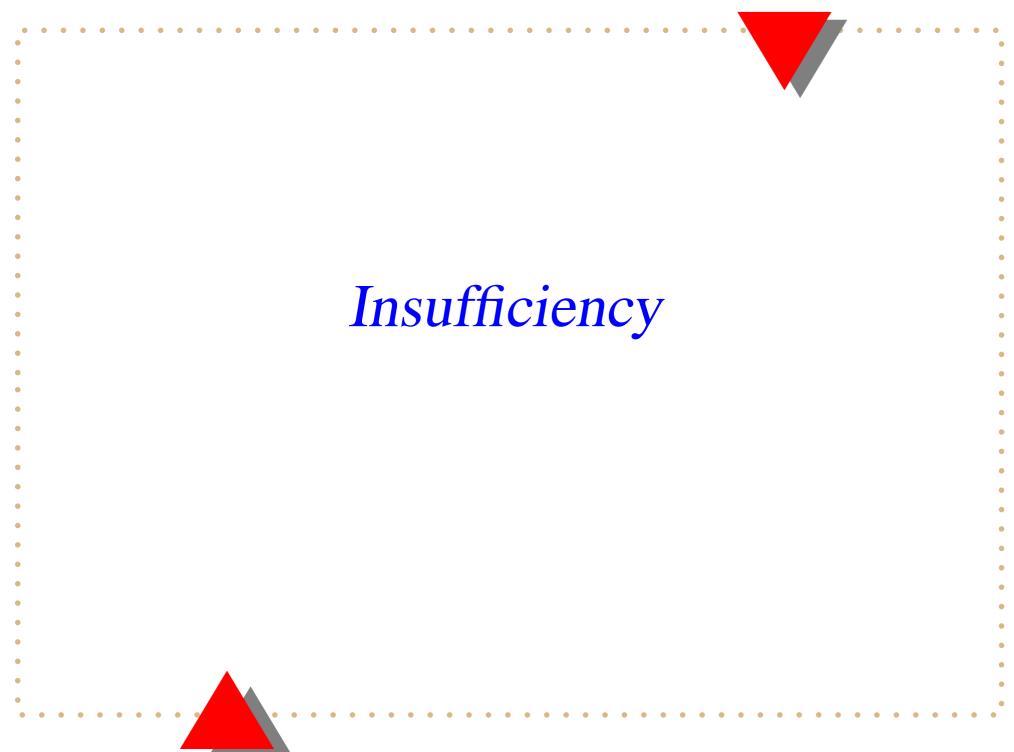












This will not do

- You might as well run the program. We were lucky the analysis terminated.
- We need further approximation...But that's not possible!
 - For each program there is a finite domain.
 - But no single finite domain will do for all programs.
 - Can we find the suitable domain by some form of textual analysis?

Textual analysis?

- Cousot brings the following examples
 - McCarty's 91-function (Bourdoncle)
 - Rational congruence analysis (Granger)
 - Linear inequality analysis (Cousot/Halbwachs)
- Here's the 91-function

$$f(n) = \begin{cases} n - 10 & \text{if } n > 100 \\ f(f(n + 11)) & \text{otherwise} \end{cases}$$

How did they do it?

Widenings

- A widening is a lattice operator ∇: L × L → L such that
 - It's an upper bound operator
 - For all increasing chains $x_0 \sqsubseteq x_1 \sqsubseteq \cdots$, the increasing chain defined by

$$y_0 = x_0$$
$$y_{i+1} = y_i \nabla x_{i+1}$$

is not strictly increasing.

Iterating with widenings

The upward iteration sequence with widening

$$X_{0} = \bot$$

$$X_{i+1} = \begin{cases} X_{i} & \text{if } F(X_{i}) \sqsubseteq X_{i} \\ X_{i} \bigtriangledown F(X_{i}) & \text{otherwise} \end{cases}$$

stabilizes to a safe approximation of the fix-point of F.

• F is reductive at that point.

Why is that?

• If we reach a point in Red(F), then we're ok.

$$X_{0} = \bot$$

$$X_{i+1} = \begin{cases} X_{i} & \text{if } F(X_{i}) \sqsubseteq X_{i} \\ X_{i} \bigtriangledown F(X_{i}) & \text{otherwise} \end{cases}$$

- The sequence is clearly ascending.
- If $\exists i : F(X_i) \sqsubseteq X_i$, then
 - The sequence will immediately stabilize.
 - Since $X_i \in \text{Red}(F)$, we have $lfp(F) \sqsubseteq X_i$.

Will we get there?
• Assume
$$\forall i : X_i \sqsubset F(X_i)$$
, then
 $X_0 = \bot$
 $X_{i+1} = X_i \bigtriangledown F(X_i)$

 This is just the widening of the following ascending chain

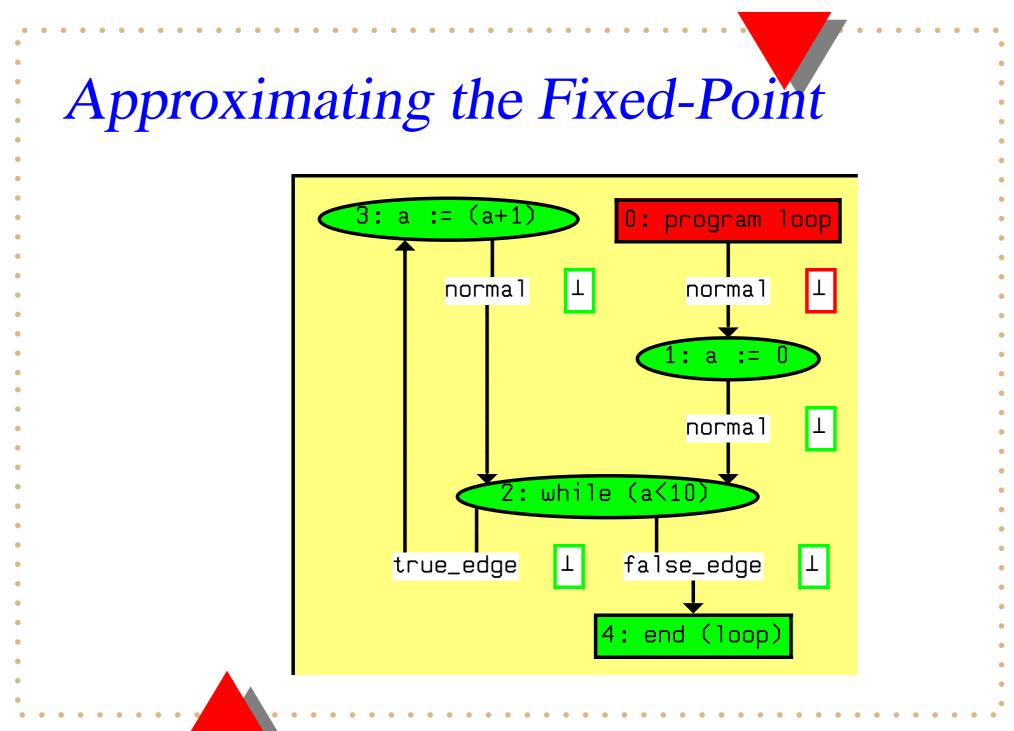
$$Y_0 = \bot$$
$$Y_{i+1} = F(X_i)$$

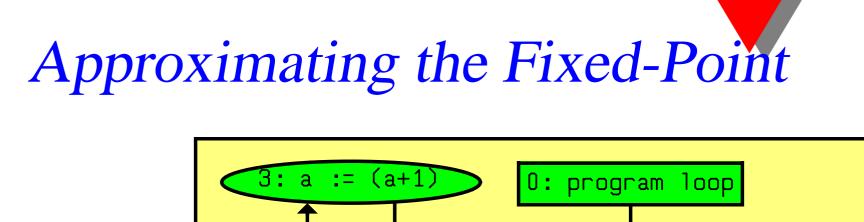
Widening for Upper bounds

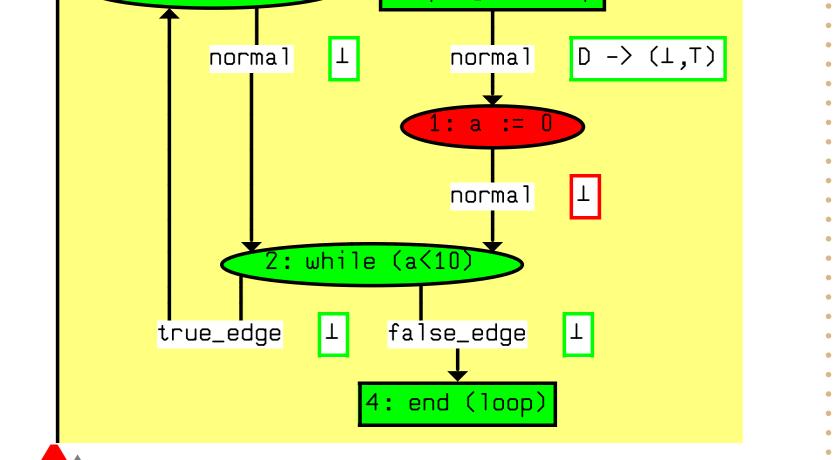
The upper bound of an interval, can be widened by

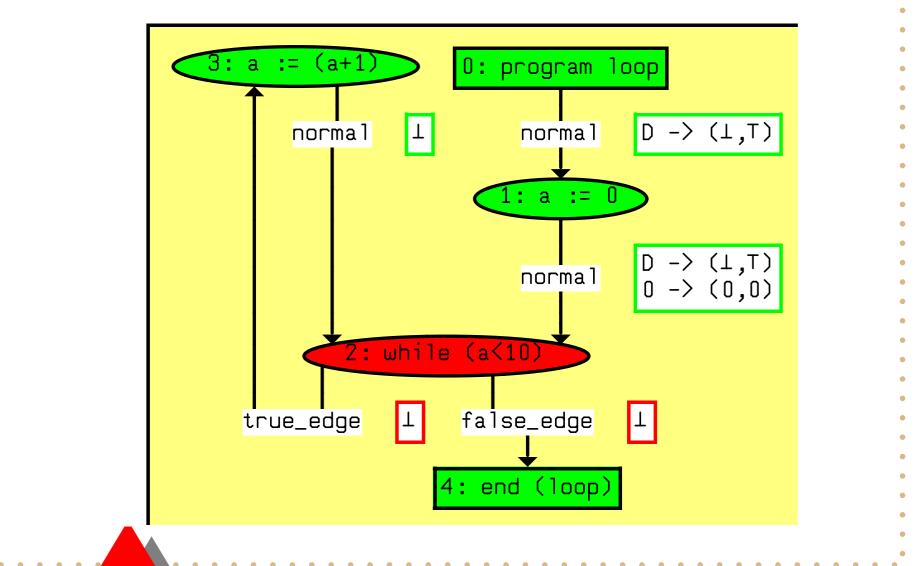
$$x \nabla y = \begin{cases} \infty & \text{if } x < y \\ x & \text{otherwise} \end{cases}$$

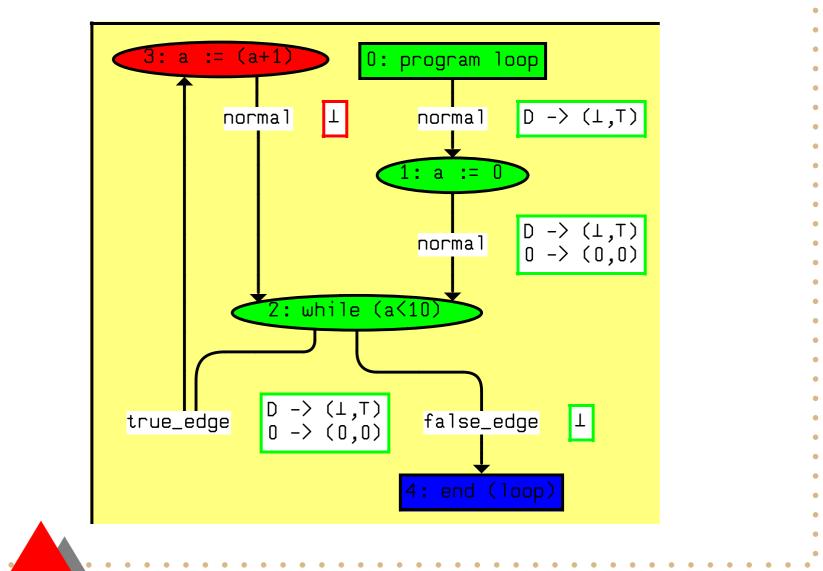
It's a widening! Any chain is stablized immediately.

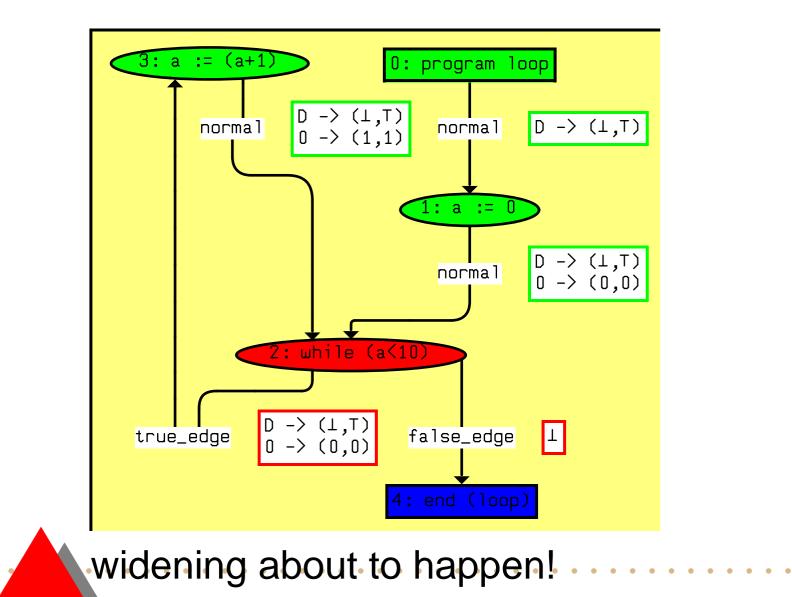


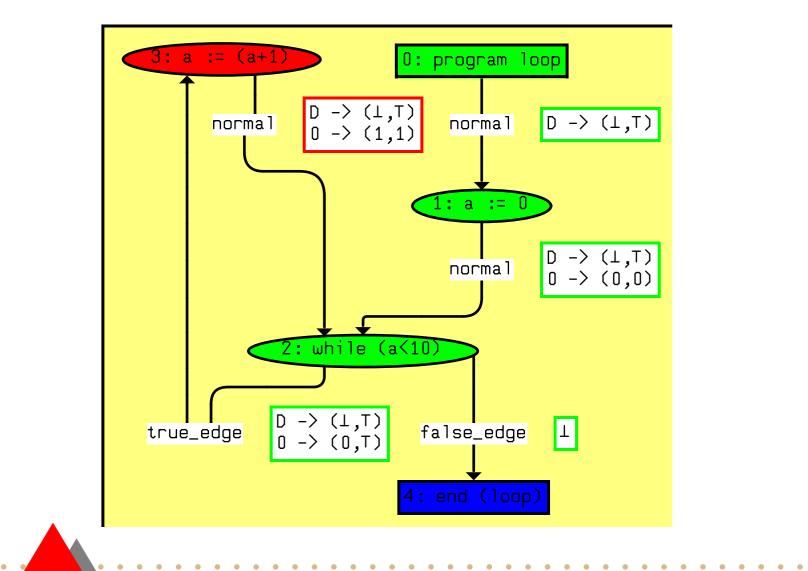


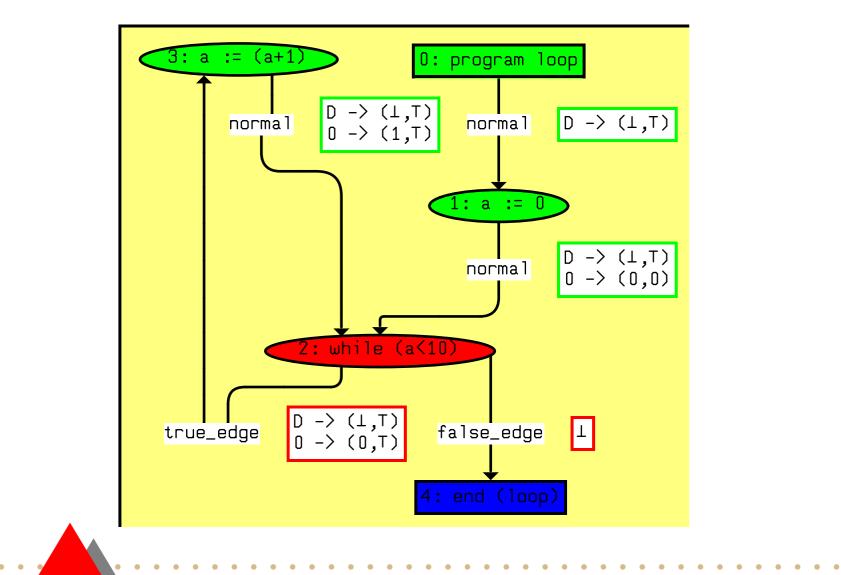


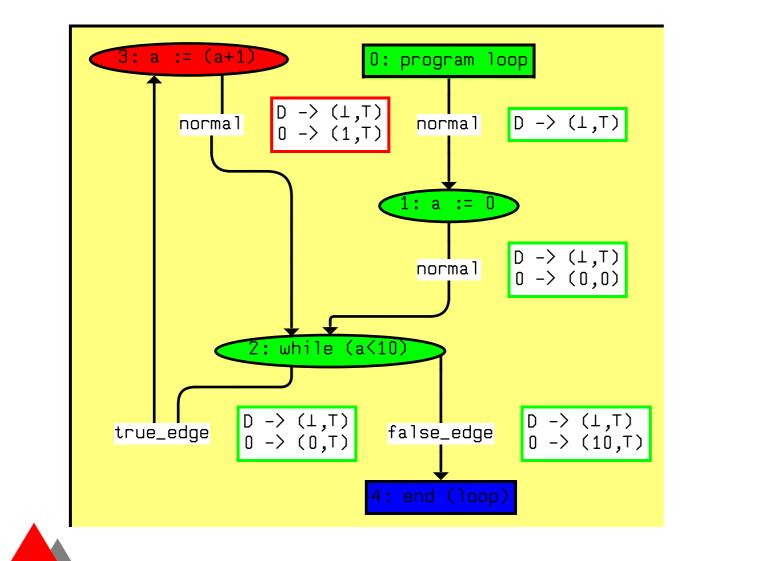


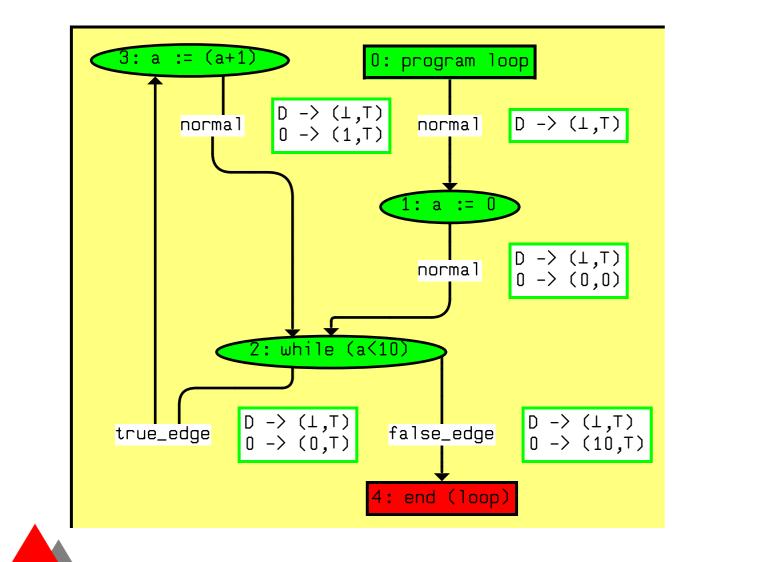












Narrowings

- Since *F* is reductive at the point reached through the upward iteratoin, we can fine-tune the result.
- Downward sequence may also not reach the fix-point, so we might need a similar procedure, called *narrowings*, to ensure termination.
- The only difference:

 $\forall x, y \in L : (y \sqsubseteq x) \implies (y \sqsubseteq (x \bigtriangleup y) \sqsubseteq x)$

Not the dual of widening

• The difference:

 $\forall x, y \in L : (y \sqsubseteq x) \implies (y \sqsubseteq (x \vartriangle y) \sqsubseteq x)$

means it is not a lower bound operator.

• While narrowings keep the sequence in Red, the dual of widenings would step out of Red and eventually into Ext.

Fine-tuning the result 0: program loop D -> (⊥,⊤) O -> (1,⊤) D -> (⊥,⊤) normal normal D -> (⊥,⊤) O -> (O,O)

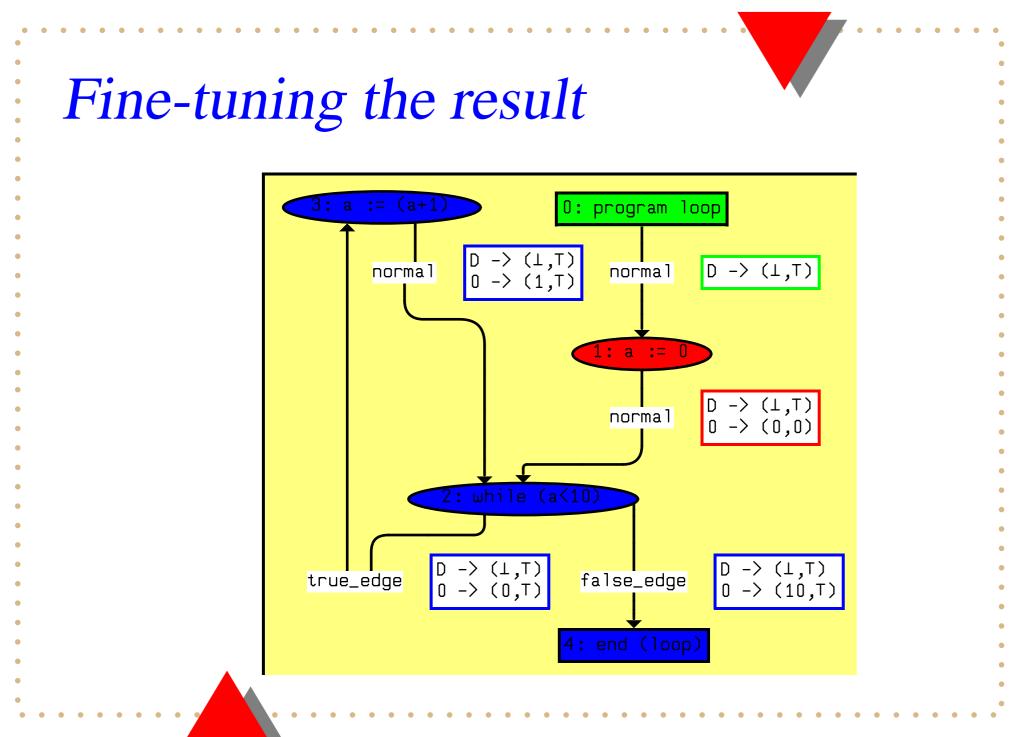
normal

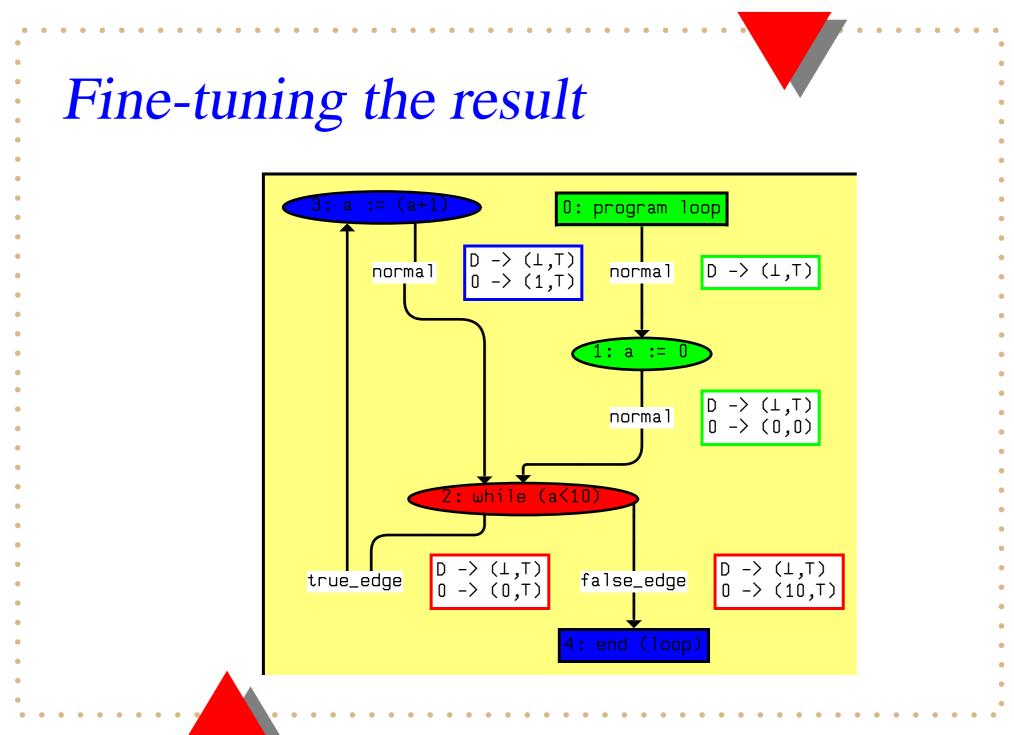
-> (⊥,T)

0 -> (10,T)

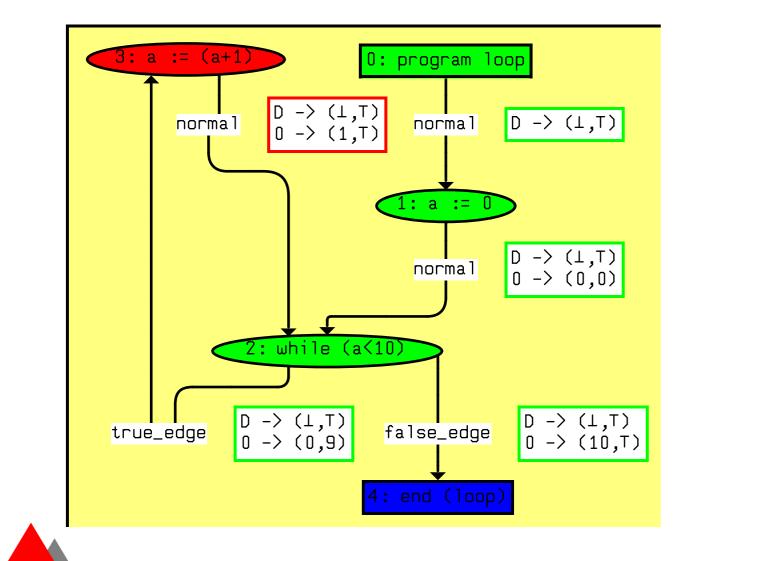
D

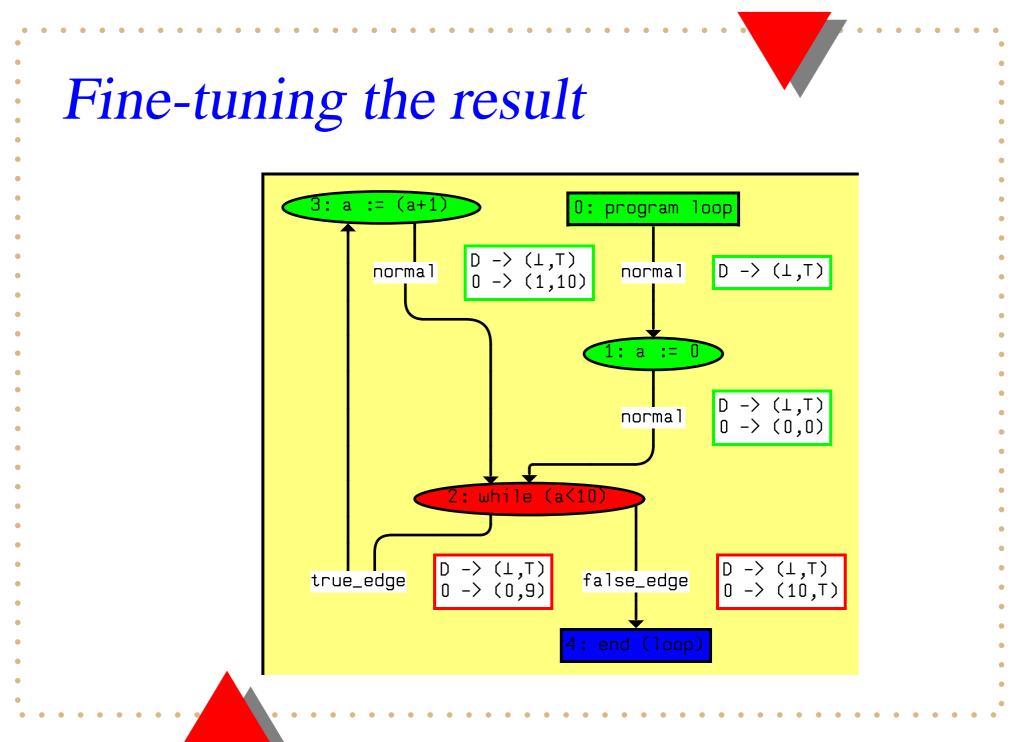
2: while (a<10) D -> (⊥,T) O -> (O,T) false_edge true_edge



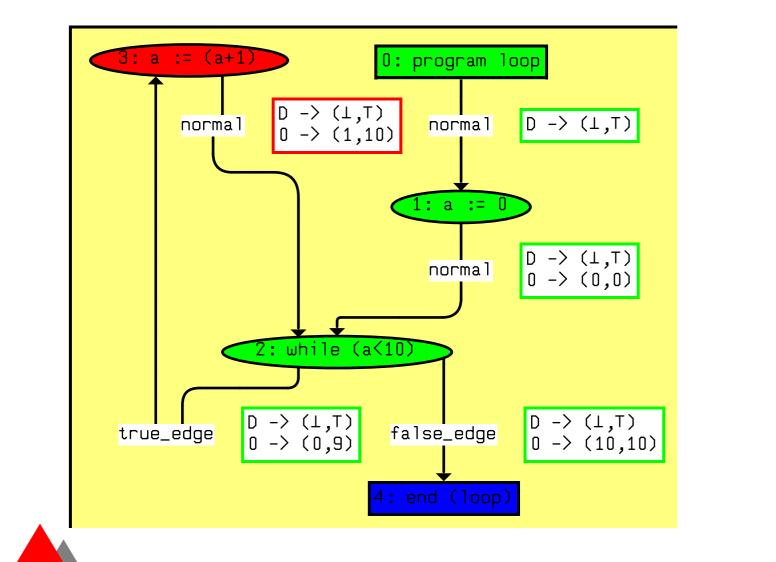


Fine-tuning the result

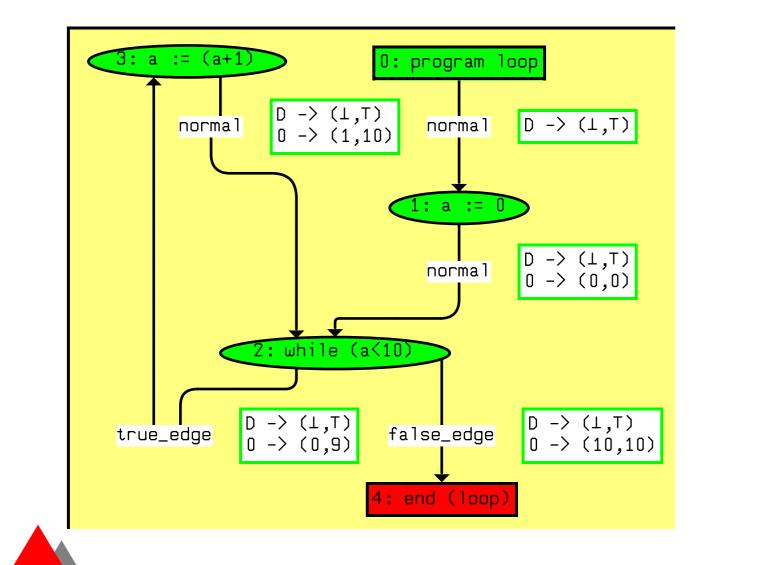




Fine-tuning the result



Fine-tuning the result



Data Flow Analysis - p. 29/31

