



LUND
UNIVERSITY

EDAP15: Program Analysis

DATAFLOW ANALYSIS 3
INTERPROCEDURAL ANALYSIS

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Inter- vs. Intra-Procedural Analysis

- ▶ **Intra**procedural: Within one procedure
 - ▶ Data flow analysis so far
- ▶ **Inter**procedural: Across multiple procedures
 - ▶ Type Analysis, especially. with polymorphic type inference

Limitations of Intra-Procedural Analysis

Teal-0

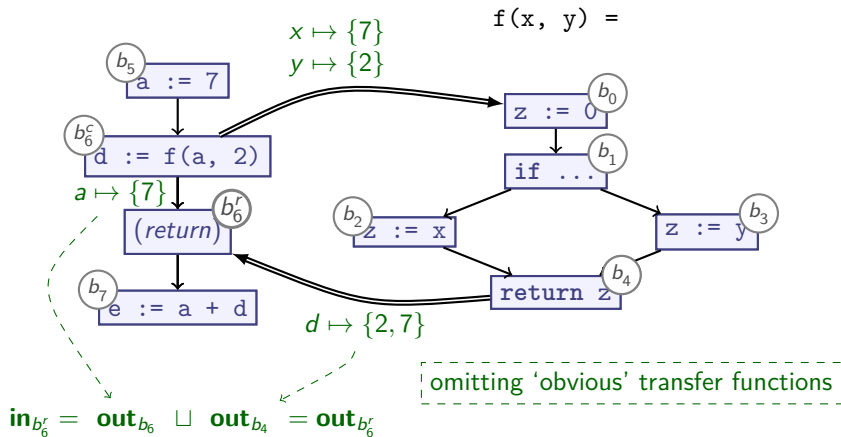
```
a := 7;  
d := f(a, 2);  
e := a + d;
```

Teal-0

```
fun f(x, y) = {  
  z := 0;  
  if x > y {  
    z := x;  
  } else {  
    z := y;  
  }  
  return z;  
}
```

How can we compute Reachable Definitions here?

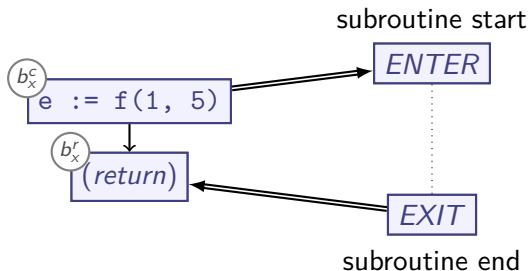
A Naïve Inter-Procedural Analysis



► **out_{b₇}:** $e \mapsto \{9, 14\}$

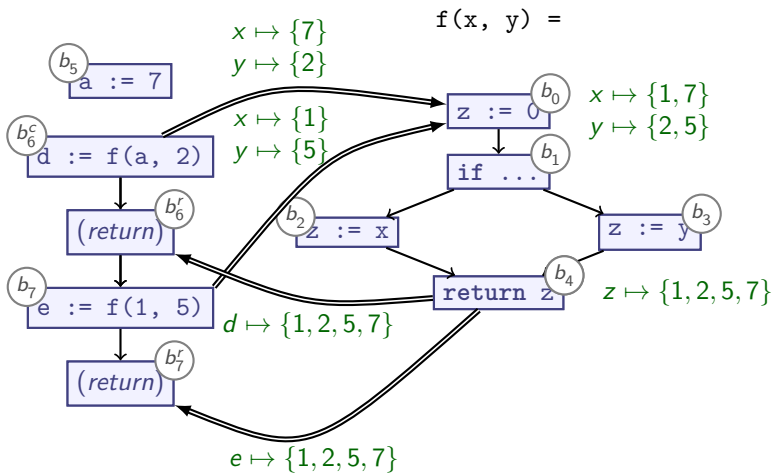
Works rather straightforwardly!

Inter-Procedural Data Flow Analysis



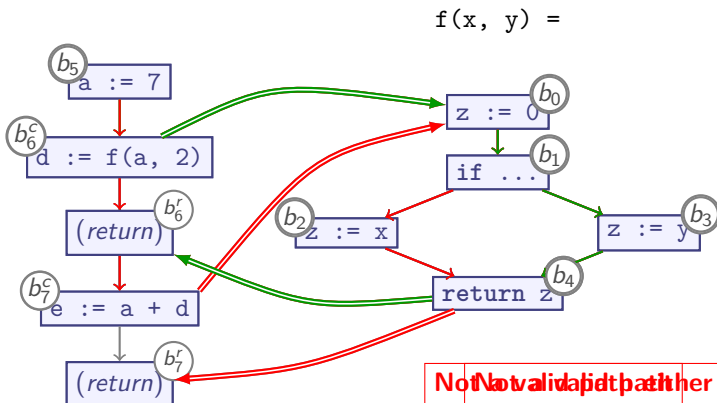
- ▶ Split call sites b_x into *call* (b_x^c) and *return* (b_x^r) nodes
- ▶ Intra-procedural edge $b_x^c \longrightarrow b_x^r$ carries environment/store
- ▶ Inter-procedural edge (\Rightarrow):
 - ▶ Caller \Rightarrow subroutine, substitutes parameters (for pass-by-value)
 - ▶ Caller \Leftarrow return, substitutes result (for pass-by-result)
 - ▶ Otherwise as intra-procedural data flow edge

A Naïve Inter-Procedural Analysis



Imprecision!

Context Sensitivity: Valid Paths



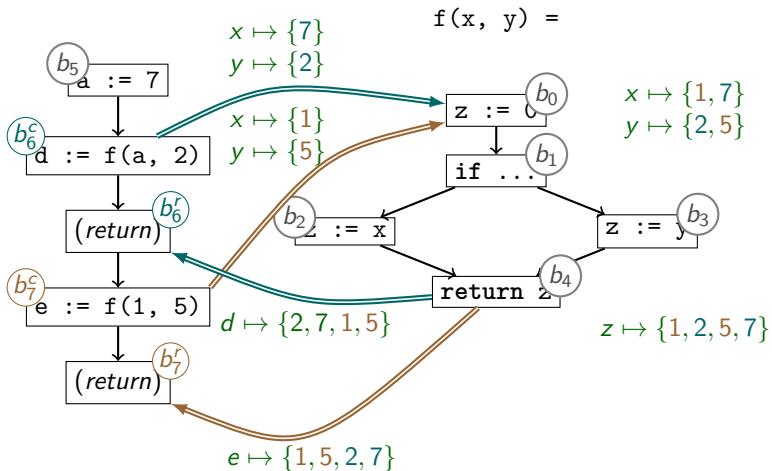
► $[b_5, b_6^c, b_0, b_1, b_3, b_4, b_6^r]$

Context-sensitive interprocedural analyses consider only valid paths

Summary

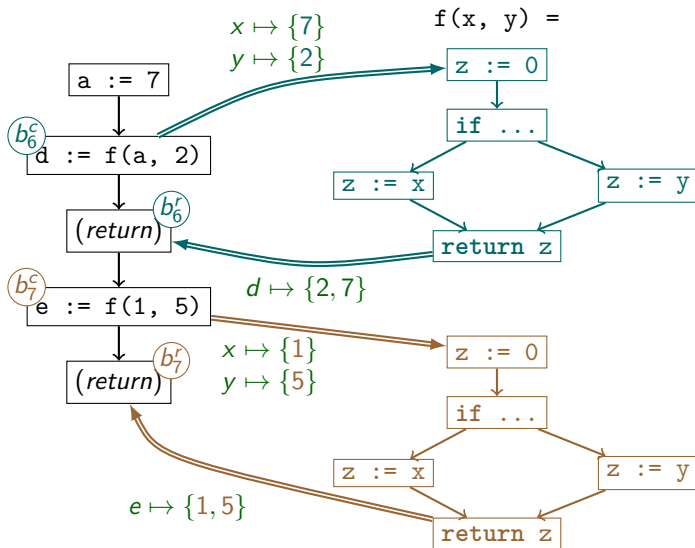
- ▶ **Intraprocedural** Data Flow Analysis is highly imprecise with subroutine calls
- ▶ **Interprocedural** Data Flow Analysis is more precise:
 - ▶ Split call site into call site + return site
 - ▶ Add flow edges between call sites, subroutine entry
 - ▶ Add flow edges between subroutine return, return site
 - ▶ Carry environment from call site to return site
- ▶ Interprocedural analysis must typically consider the entire program
 - ⇒ **whole-program analysis**
- ▶ Naïve interprocedural analysis is **context-insensitive**
 - ▶ Merge all callers into one

Interprocedural Data Flow Analysis



Context-insensitive: analysis merges all callers to $f()$

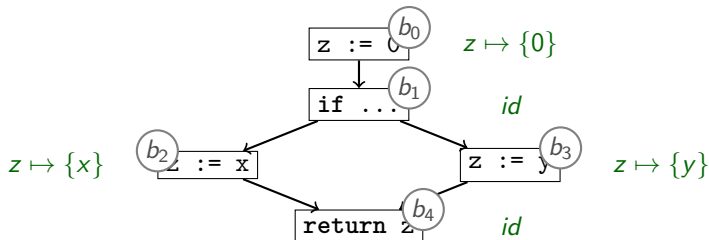
Inlining



Clone subroutine IRs for each *calling context*

Alternative to Inlining: Summarise Procedure (Here: Reaching Defs.)

$f(x, y) =$



► Compose transfer functions:

- $trans_{b_0} \circ trans_{b_1} = [z \mapsto 0]$
- $trans_{b_0} \circ trans_{b_1} \circ trans_{b_2} = [z \mapsto \{x\}]$
- $trans_{b_0} \circ trans_{b_1} \circ trans_{b_3} = [z \mapsto \{y\}]$
- $trans_{b_0} \circ trans_{b_1} \circ (trans_{b_2} \sqcup trans_{b_3}) = [z \mapsto \{x, y\}]$
- $trans_{b_0} \circ trans_{b_1} \circ (trans_{b_2} \sqcup trans_{b_3}) \circ trans_{b_4} = [z \mapsto \{x, y\}]$

Procedure Summaries vs Recursion

`f calls g calls h calls f`

- ▶ Requires additional analysis to identify who calls whom
- ▶ Compute summaries of mutually recursive functions together
- ▶ Recursive call edges analogous to loops

Procedure Summaries

- ▶ Composing transfer functions yields a combined transfer function for $f()$:

$$trans_f = [\mathbf{return} \mapsto \{x, y\}]$$

- ▶ Use $trans_f$ as transfer function for $f()$, discard f 's body

- ▶ **Advantages:**

- ▶ Can yield compact subroutine descriptions
- ▶ Can speed up call site analysis dramatically

- ▶ **Disadvantages:**

- ▶ More complex to implement
- ▶ Recursion is challenging

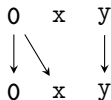
- ▶ **Limitations:**

- ▶ Requires suitable representation for summary
- ▶ Requires mechanism for abstracting and applying summary
- ▶ Worst cases:
 - ▶ $trans_f$ is symbolic expression as complex as f itself

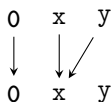
Representation Relations

Example procedure summary representation:

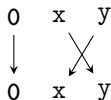
```
x := null;
y := y;
```



```
if x != y {
    x := y;
}
y := 1;
```



```
{ t := x
  x := y
  y := t }
```



'May be null' analysis

- $c \rightarrow d$:
 if $P(c) \in \mathbf{in}_b$ then $P(d) \in \mathbf{out}_b$
- Representation Relations relate \mathbf{in}_b and \mathbf{out}_b variables \mathcal{V}
- $R \subseteq (\mathcal{V} \cup \{\mathbf{0}\}) \times (\mathcal{V} \cup \{\mathbf{0}\})$
- if $\langle \mathbf{0}, X \rangle \in R$:
 X always 'may be null' in \mathbf{out}_b
- if $\langle Y, X \rangle \in R$:
 If Y 'may be null' in \mathbf{in}_b :
 $\Rightarrow X$ 'may be null' in \mathbf{out}_b

Summary

- ▶ **Context-sensitive** analysis distinguishes ‘calling context’ when analysing subroutine
 - ▶ ‘Who called me’?
 - ▶ Can go deeper: ‘And who called them?’
- ▶ **Inlining** is one strategy for **context-sensitive** analysis
- ▶ Copy subroutine bodies for each caller
- ▶ Alternative: **Procedure summaries** built from composed transfer functions
- ▶ Can speed up context-sensitive analysis of popular functions, compared to inlining
- ▶ Needs some suitably abstract analysis *for the given program*
 - ▶ Example: IFDS-style **Representation Relations**
- ▶ Recursion is nontrivial:
 - ▶ Analyse function calls (*call graph*)
 - ▶ Analyse strongly connected components together

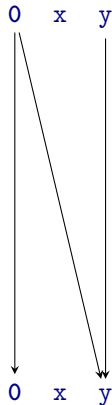
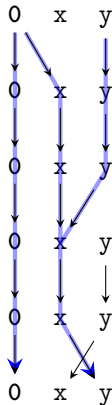
Composing Representation Relations

Recall Representation Relations (*may be null* analysis):

```
x := null;  
y := y;
```

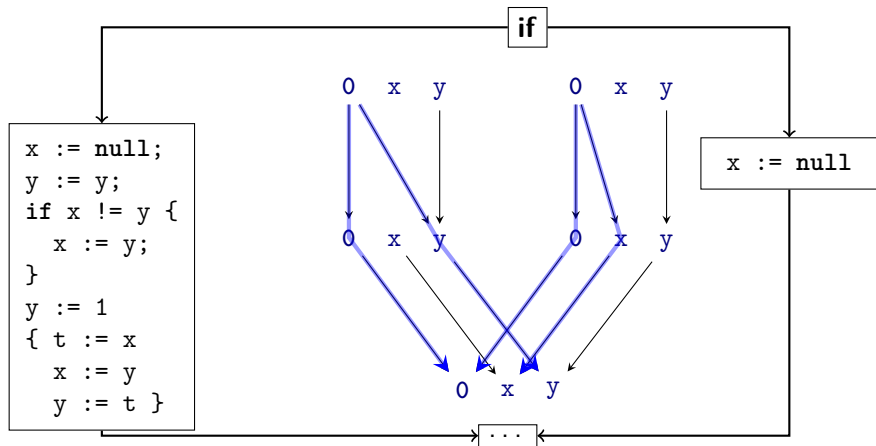
```
if x != y {  
  x := y;  
}  
y := 1;
```

```
{ t := x;  
  x := y;  
  y := t; }
```



Composed representation relations are again representation relations

Merging Control-Flow Paths



Logical "Or"

Dataflow via Graph Reachability

$$n = \langle b, v \rangle$$

- ▶ Assume binary lattice $(\{\top, \perp\}, \sqsubseteq, \sqcap, \sqcup)$
 - ▶ $a \sqcup b = \top$ iff $a = \perp$ and $b = \perp$, otherwise $a \sqcup b = \top$
 - ▶ Typical for 'May be X' analysis ('may be null')
- ▶ We can encode Dataflow problem as *Graph-Reachability*
- ▶ Graph nodes $n = \langle b, v \rangle$
 - ▶ b : CFG node
 - ▶ v : Variable or $\mathbf{0}$
 - ▶ Variable: Property of interest connected to variable
 - ▶ $\mathbf{0}$: Property of interest connected to executing this statement/block

Dataflow via Graph Reachability

$$n = \langle b, v \rangle$$

- ▶ Assume binary lattice $(\{\top, \perp\}, \sqsubseteq, \sqcap, \sqcup)$
 - ▶ $a \sqcup b = \top$ iff $a = \perp$ and $b = \perp$, otherwise $a \sqcup b = \top$
 - ▶ Typical for 'May be X' analysis ('may be null')
 - ▶ Equivalently for 'Must' analysis:
'must be null' = not ('may be non-null')
- ▶ We can encode Dataflow problem as *Graph-Reachability*
- ▶ Graph nodes $n = \langle b, v \rangle$
 - ▶ b : CFG node
 - ▶ v : Variable or **0**
 - ▶ Variable: Property of interest connected to variable
 - ▶ **0**: Property of interest connected to executing this statement/block

A Dataflow Worklist Algorithm: IFDS

- ▶ Context-sensitive interprocedural dataflow algorithm
- ▶ Historical name: IFDS
(Interprocedural **F**inite **D**istributive **S**ubset problems)
- ▶ 'Exploded Supergraph': $G^\# = (N^\#, E^\#)$
 - ▶ $N^\# = N_{CFG} \times \mathcal{V} \cup \{0\}$
 - ▶ Plus parameter/return call edges
- ▶ b_{main}^s is the CFG *ENTER* node of the main entry point
- ▶ Property-of-interest holds if reachable from $\langle b_{main}^s, \mathbf{0} \rangle$
- ▶ **Key ideas:**
 - ▶ Worklist-based
 - ▶ Construct Representation Relations on demand
 - ▶ Construct 'Exploded Supergraph'
 - ▶ CFG of all functions $\times \mathcal{V} \cup \{\mathbf{0}\}$

IFDS Datastructures

Instead of $\langle\langle b_0, v_0 \rangle, \langle b_3, v_0 \rangle\rangle$ we also write:

$$\langle b_0, v_0 \rangle \rightarrow \langle b_3, v_0 \rangle$$

WORKLIST edge

$$\langle b_0, v_0 \rangle \dashrightarrow \langle b_3, v_0 \rangle$$



PATHEDGE edge

All WORKLIST edges are also PATHEDGE edges

Result of our analysis

N^\sharp -edge



SUMMARYINST

Generated from summary nodes
Otherwise equivalent to N^\sharp -edges

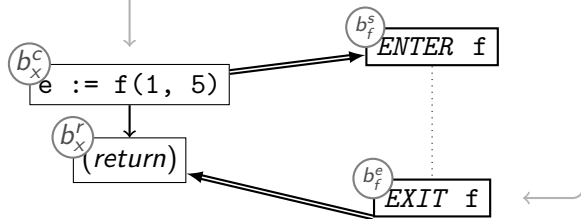
IFDS Strategy

- ▶ Algorithm distinguishes between three types of nodes:

- ▶ Exit nodes (b_f^e)

- ▶ Call nodes (b_x^c)

- ▶ Other nodes



On-demand processing

```
Procedure propagate( $n_1 \rightarrow n_2$ ):  
begin  
  if  $n_1 \rightarrow n_2 \in \text{PATHEDGE}$  then  
    return  
   $\text{PATHEDGE} := \text{PATHEDGE} \cup \{n_1 \rightarrow n_2\}$   
   $\text{WORKLIST} := \text{WORKLIST} \cup \{n_1 \rightarrow n_2\}$   
end
```

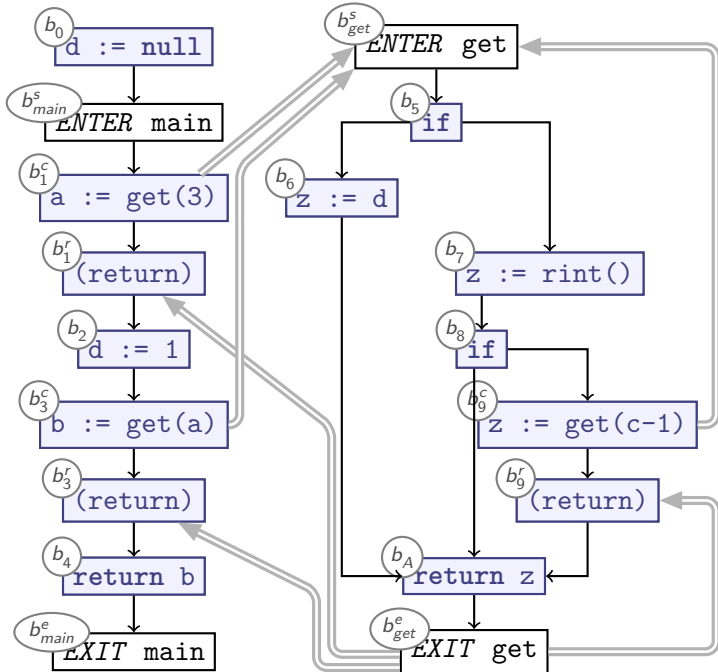
Running Example

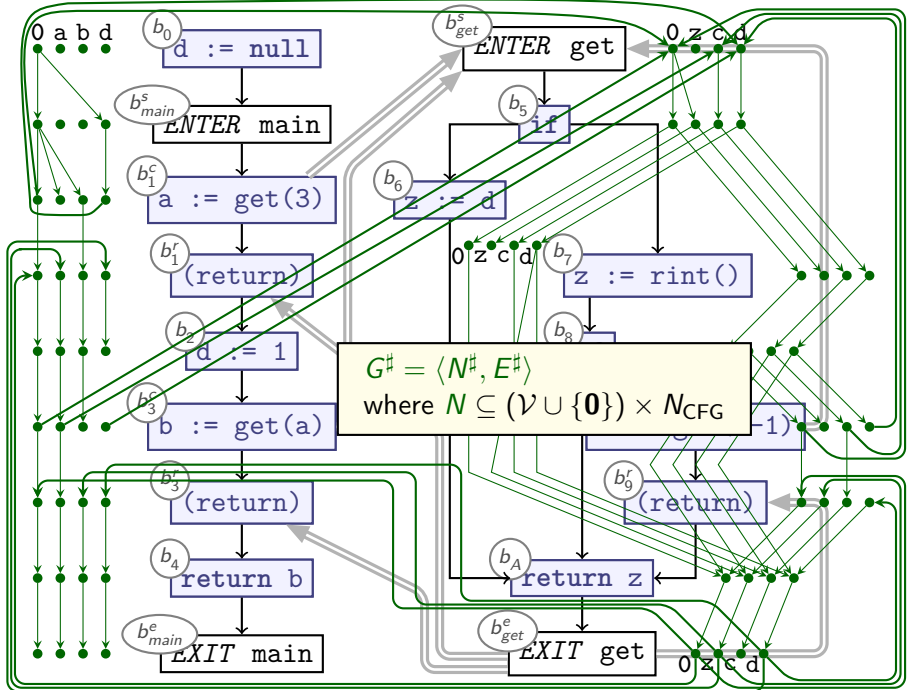
Teal-0: *main()*

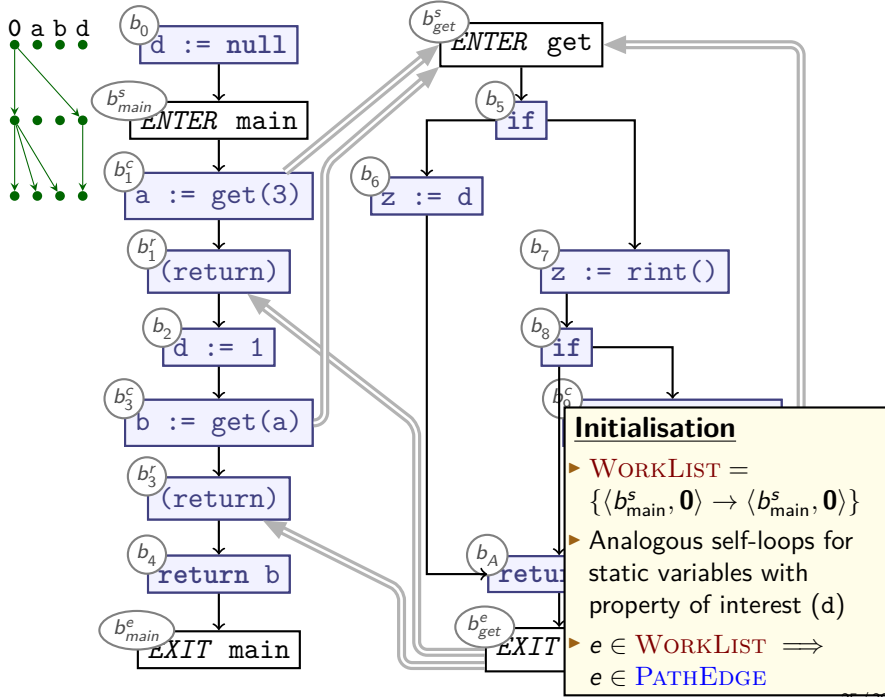
```
var default := null;
fun main() = {
  var a := get(3);
  default := 1;
  var b := get(3);
  return b;
}
```

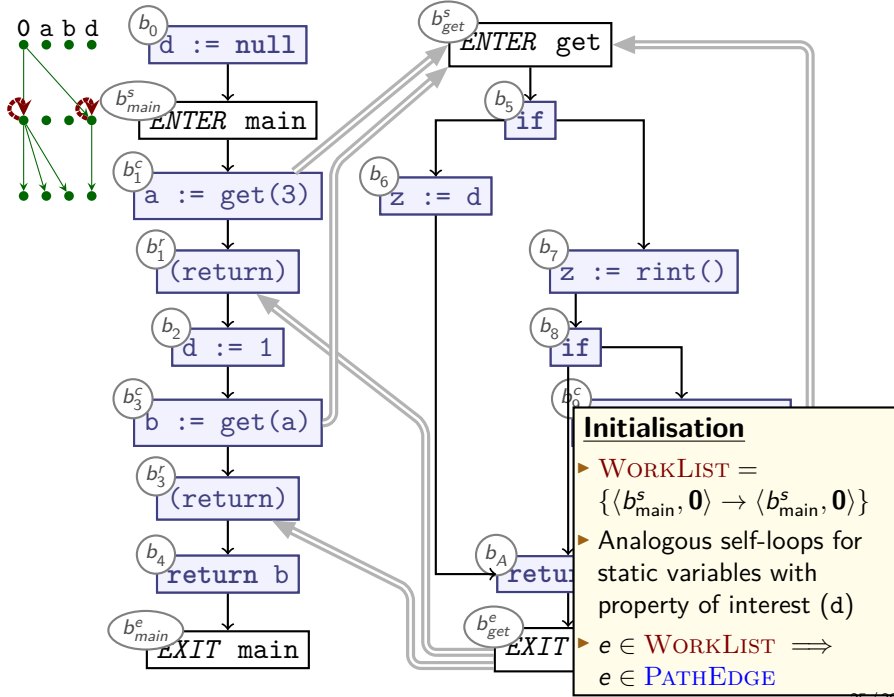
Teal-0: *get()*

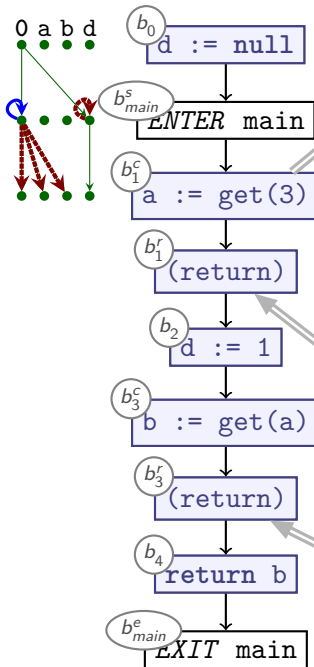
```
fun get(c) = {
  if c == 0 {
    z := default;
  } else {
    z := read_int();
    if z < 0 {
      z := get(c - 1);
    }
  }
  return z;
}
```









Procedure propagate($n_1 \rightarrow n_2$):

begin

if $n_1 \rightarrow n_2 \in \text{PATHEDGE}$ **then**

return

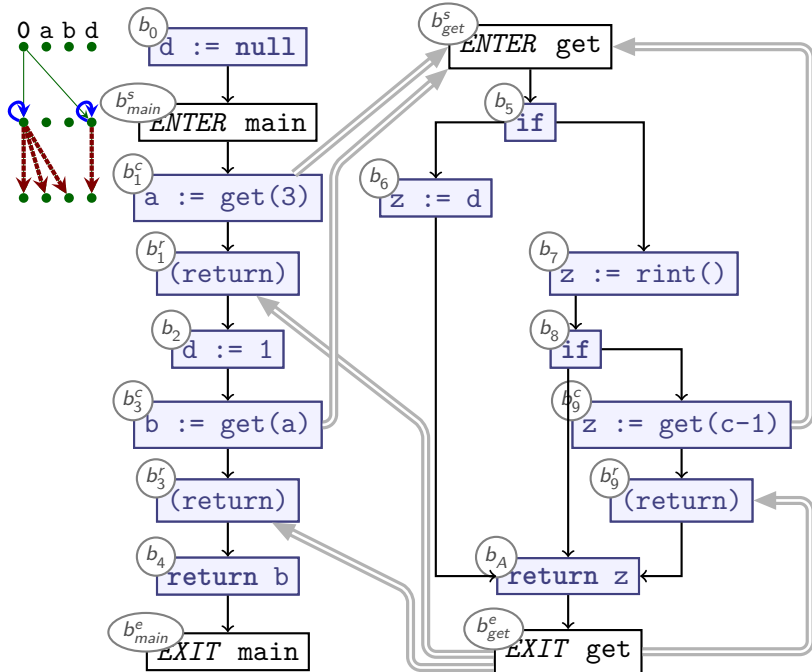
$\text{PATHEDGE} := \text{PATHEDGE} \cup \{n_1 \rightarrow n_2\}$

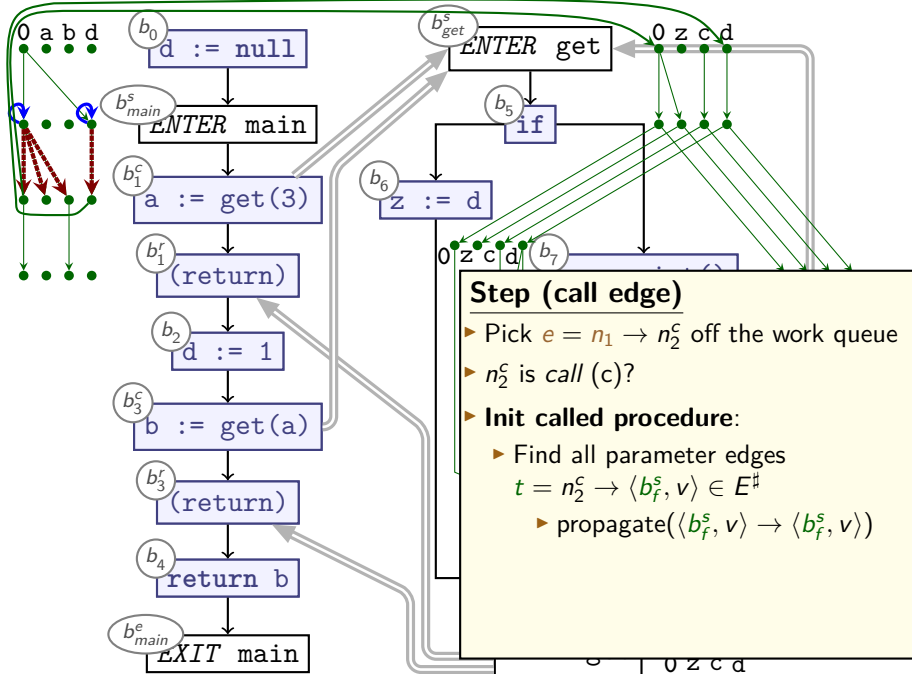
$\text{WORKLIST} := \text{WORKLIST} \cup \{n_1 \rightarrow n_2\}$

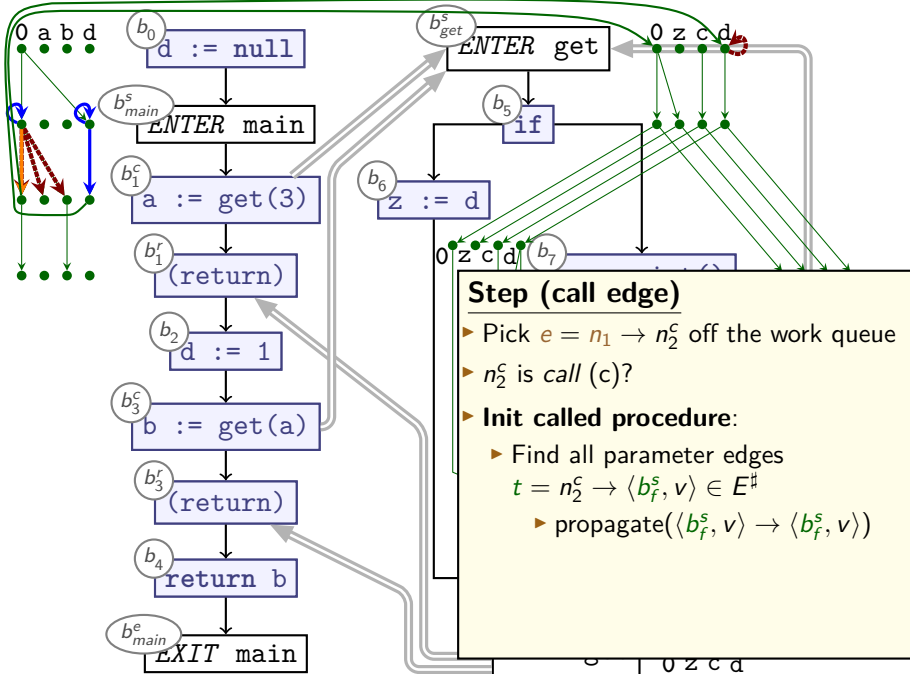
end

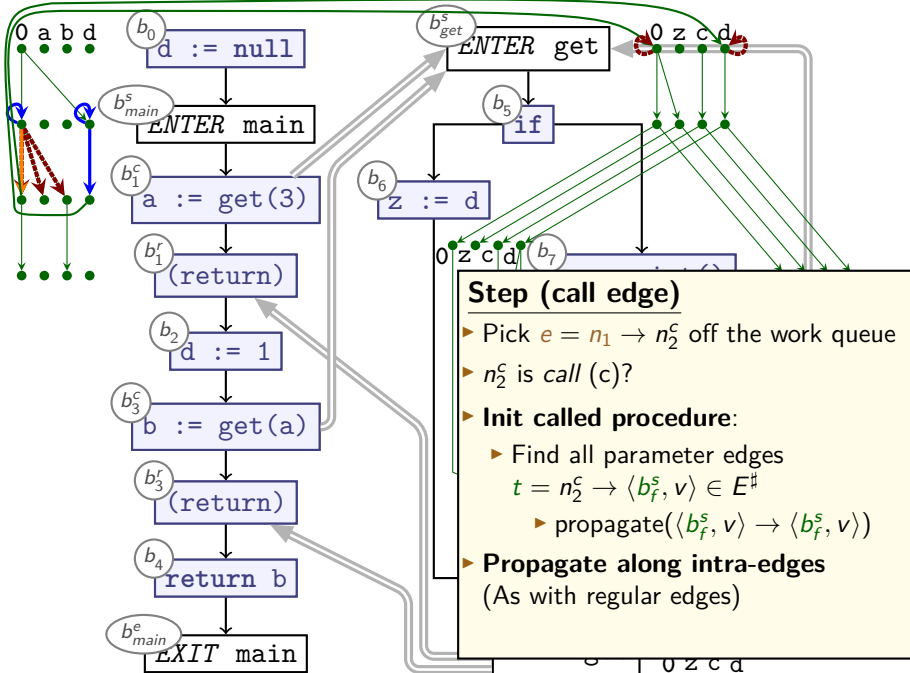
Step (regular edge)

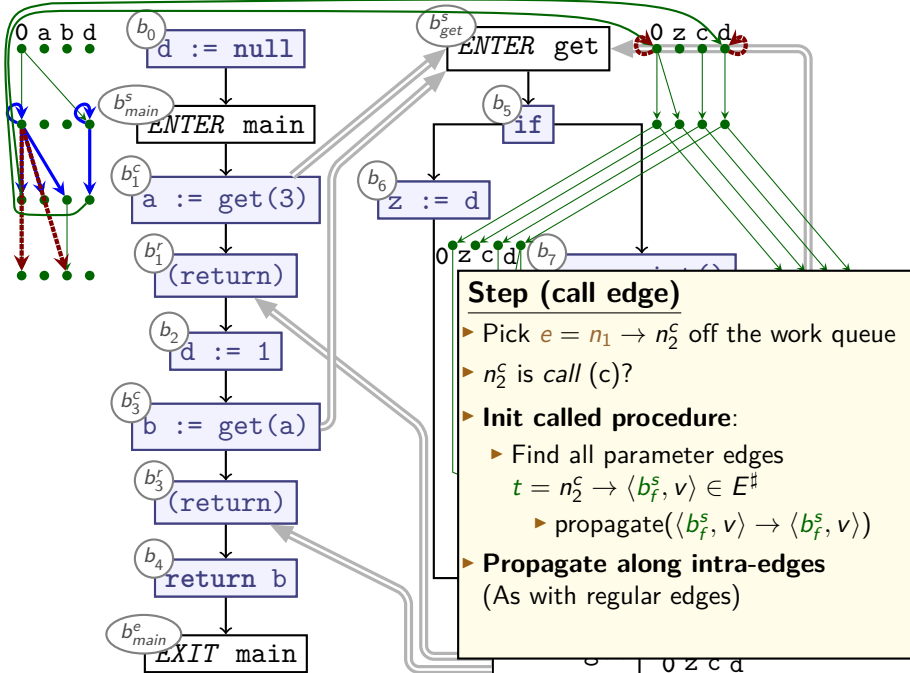
- ▶ Pick e off the work queue
 $e = n_1 \rightarrow n_2$
- ▶ n_2 neither call (c) nor exit (e)?
- ▶ Find all $n_2 \rightarrow n_3$
propagate($n_1 \rightarrow n_3$)
- ▶ Remove e from WORKLIST
- ▶ e remains in PATHEDGE

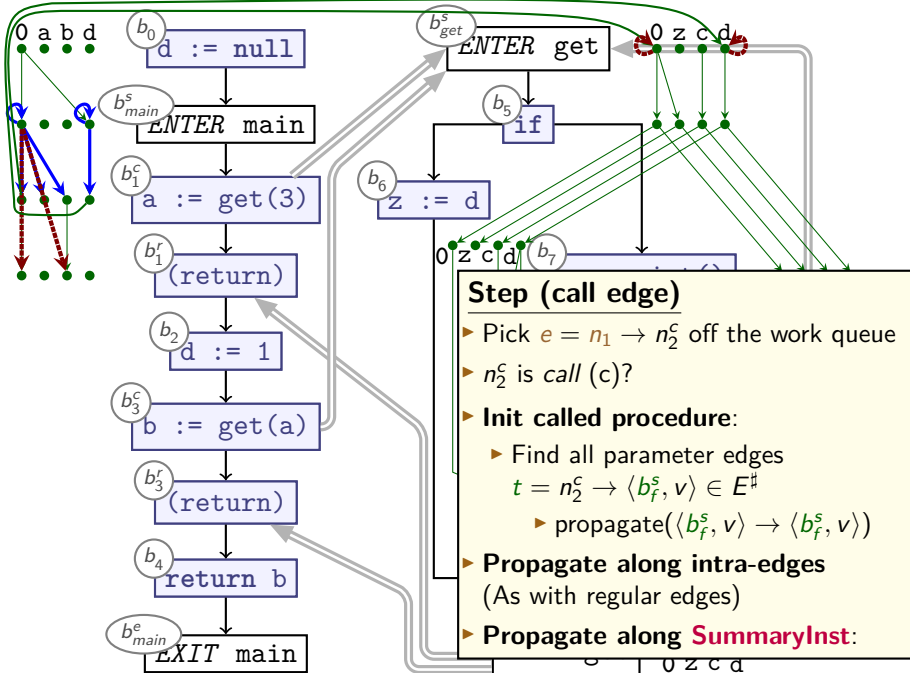


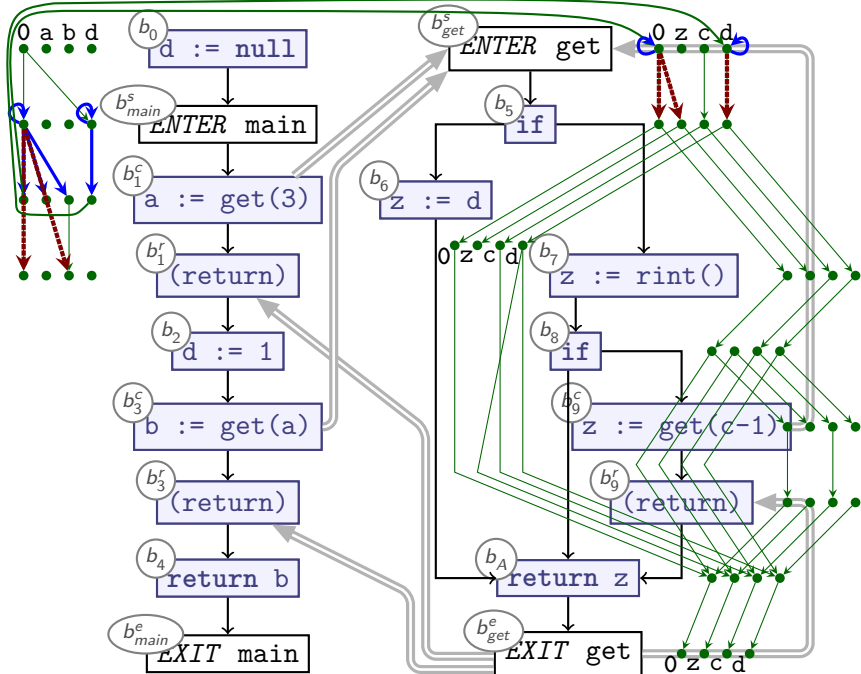


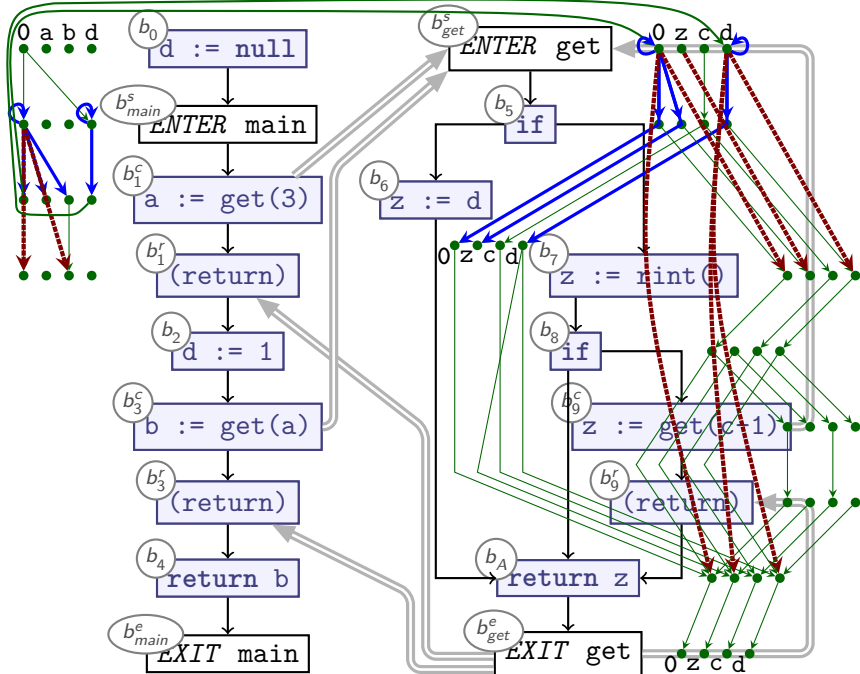


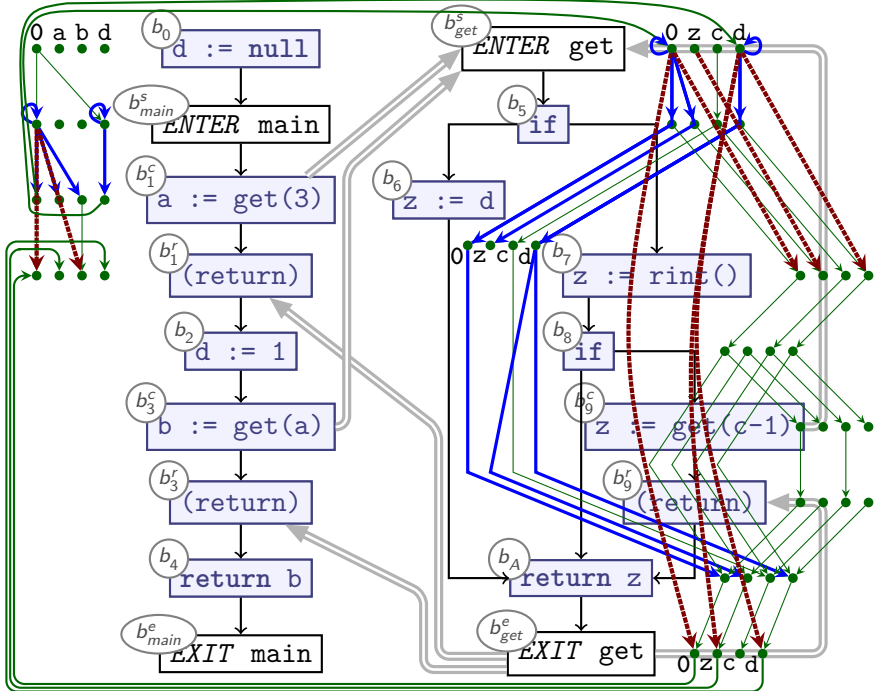


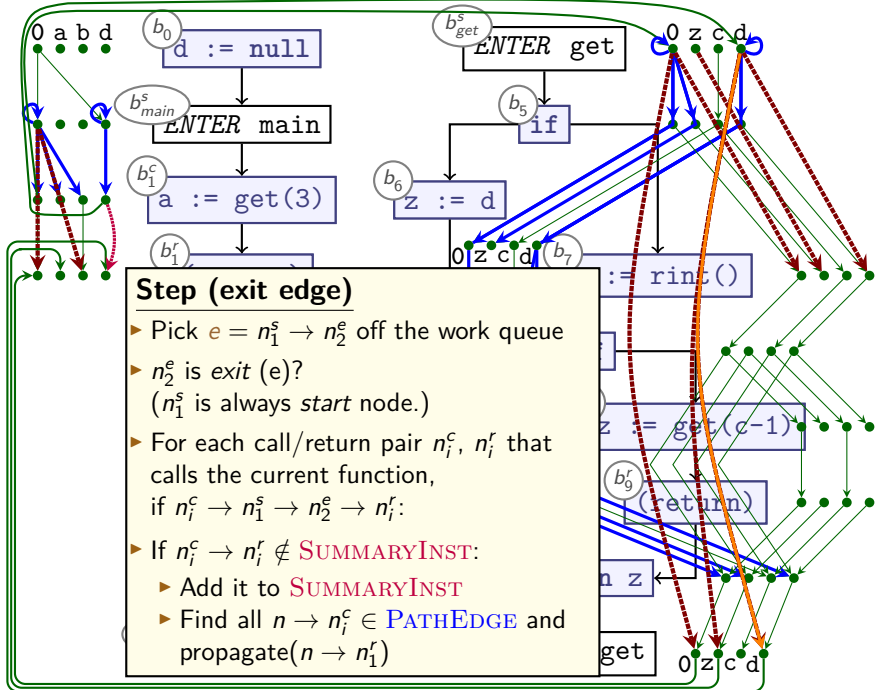


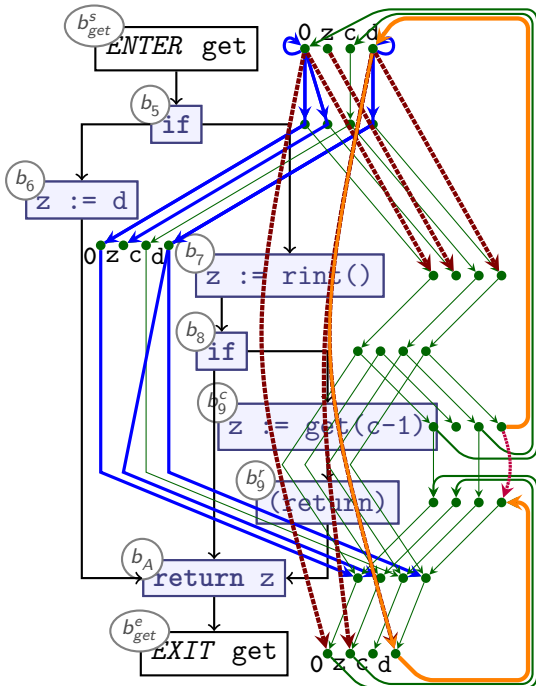
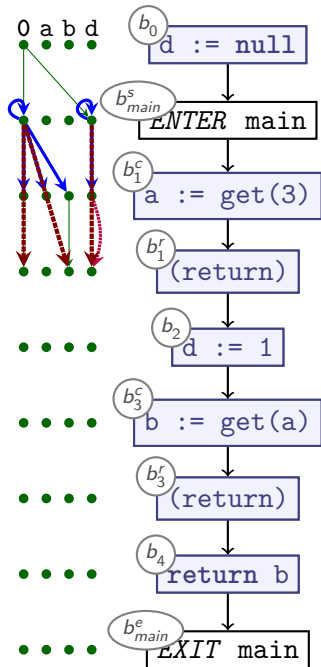


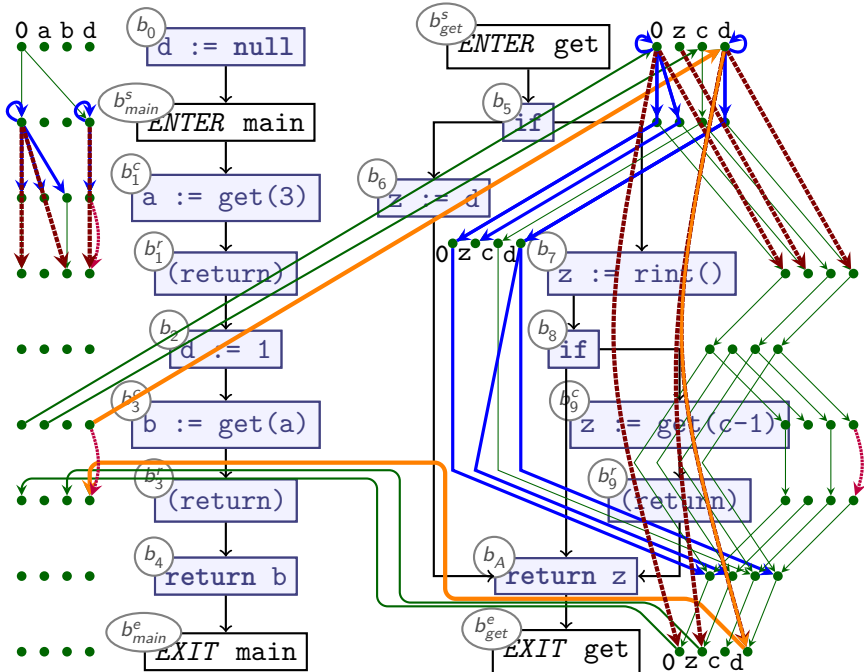


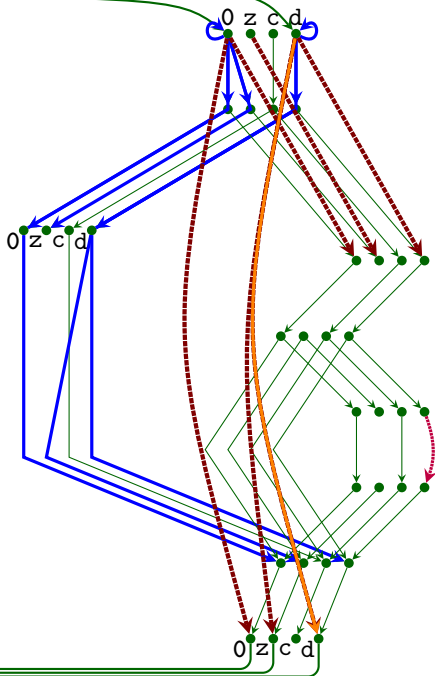
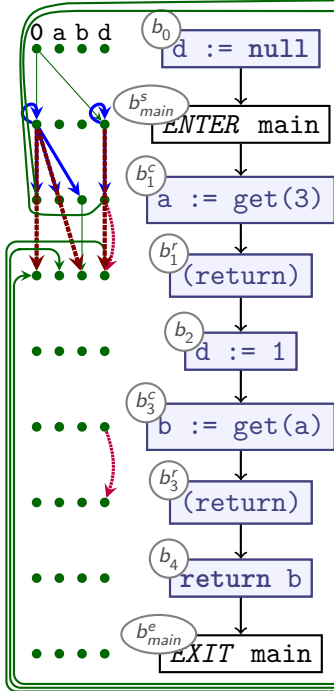


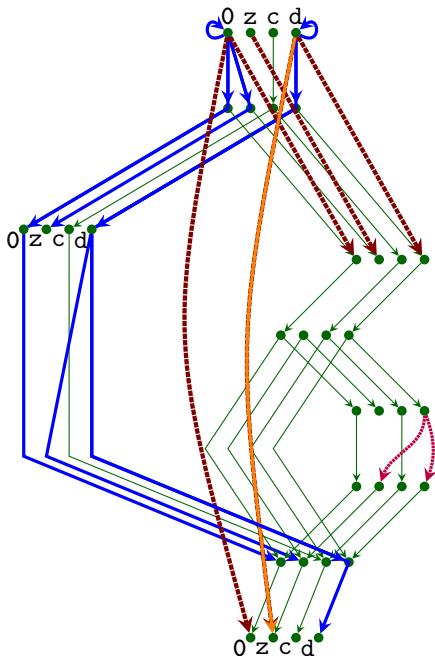
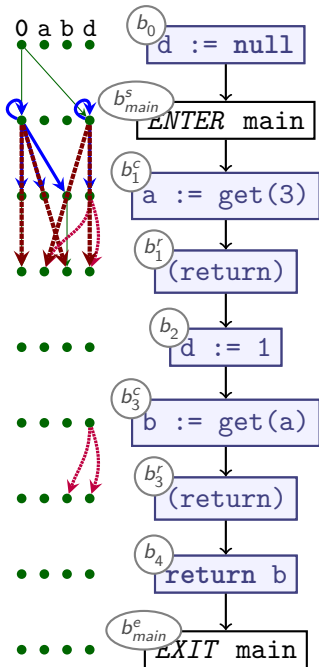


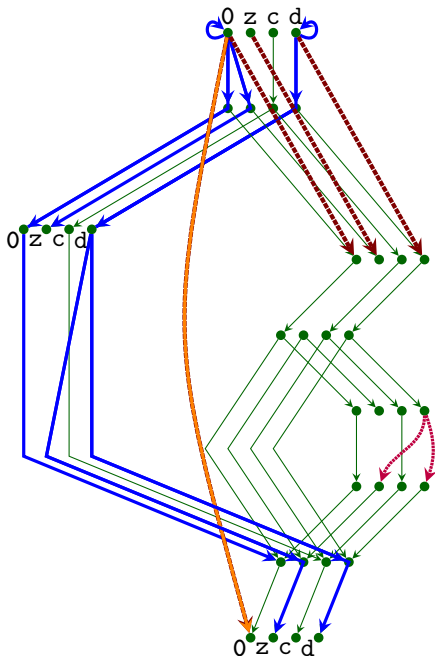
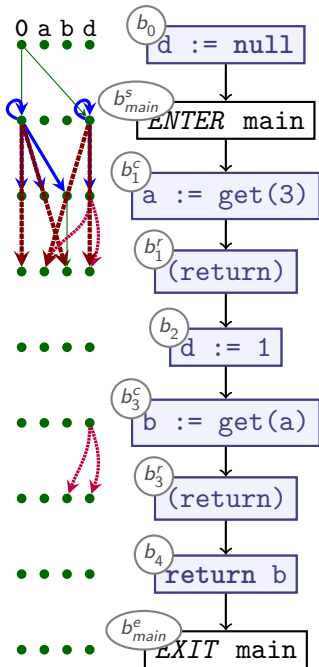


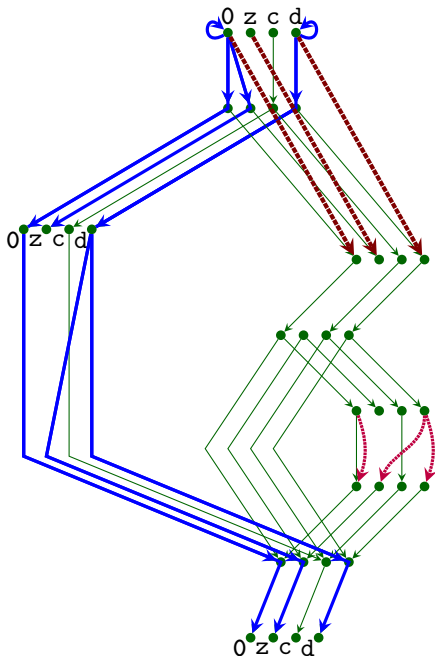
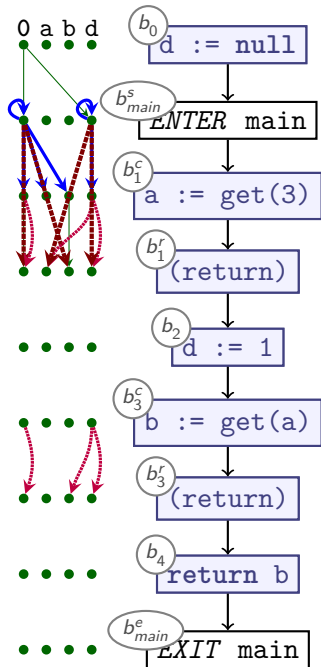


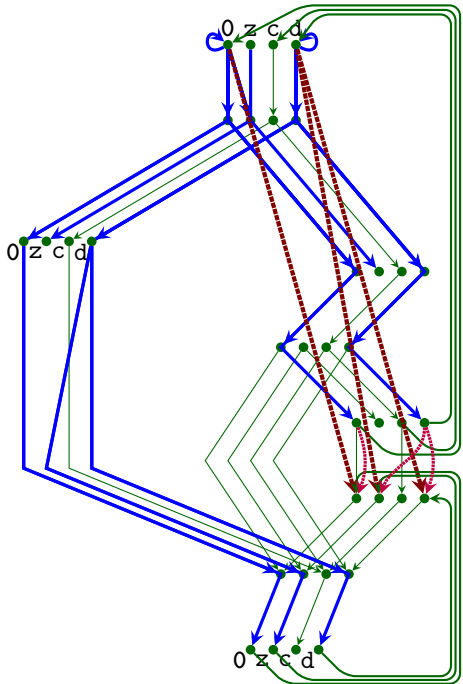
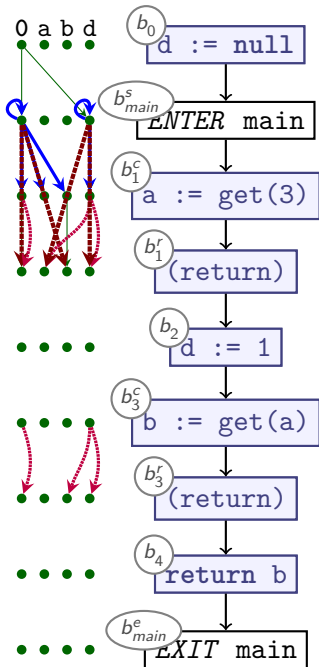


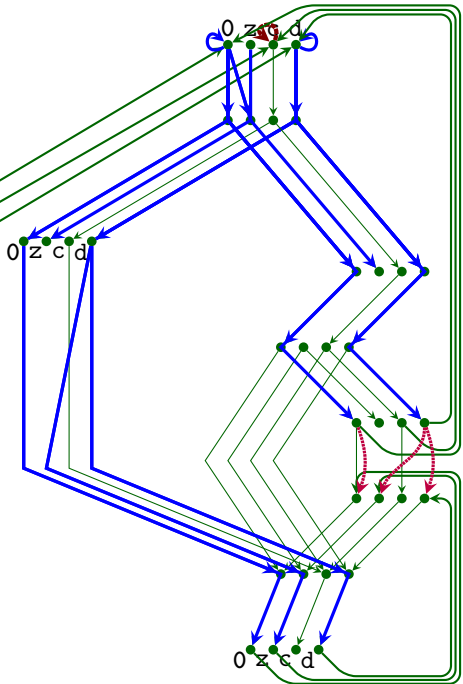
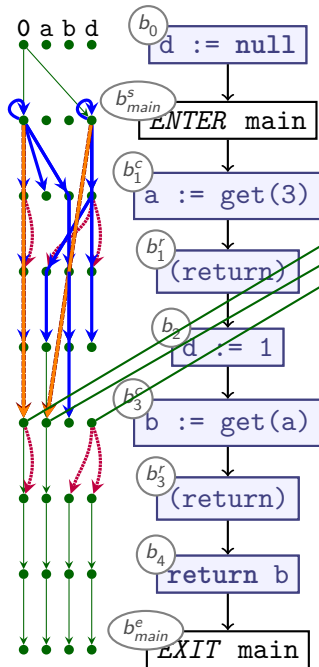


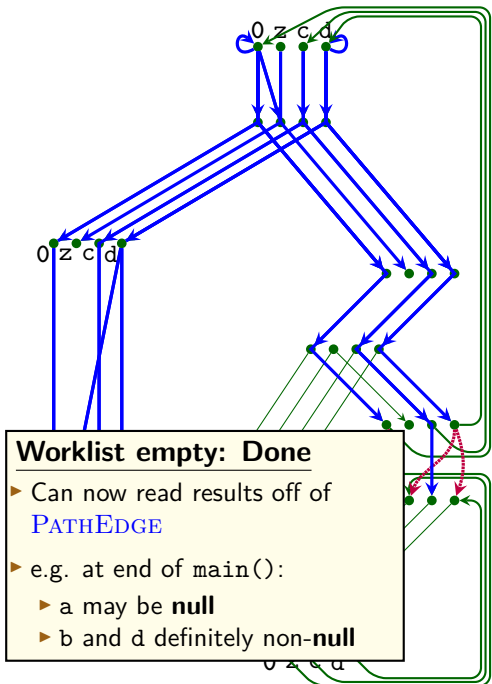
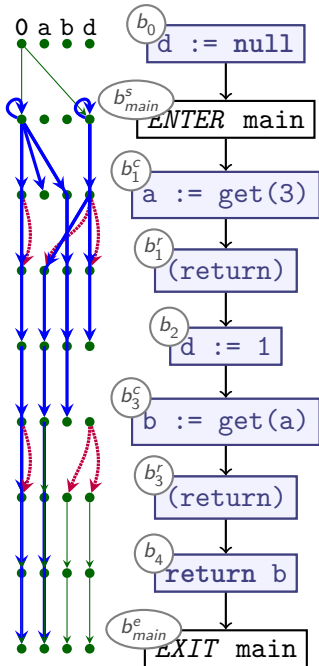












The IFDS Algorithm: Initialisation and Propagation)

Procedure Init():

begin

WORKLIST := **PATHEDGE** := \emptyset

propagate($\langle b_{\text{main}}^s, \mathbf{0} \rangle \rightarrow \langle b_{\text{main}}^s, \mathbf{0} \rangle$)

ForwardTabulate()

end

Procedure propagate($n_1 \rightarrow n_2$):

begin

if $n_1 \rightarrow n_2 \in \mathbf{PATHEDGE}$ **then**

return

PATHEDGE := **PATHEDGE** $\cup \{n_1 \rightarrow n_2\}$

WORKLIST := **WORKLIST** $\cup \{n_1 \rightarrow n_2\}$

end

IFDS: Forward Tabulation

Procedure ForwardTabulate():

begin

while $n_0 \rightarrow n_1 \in \text{WORKLIST}$ **do**

WorkList := **WorkList** $\setminus \{n_0 \rightarrow n_1\}$

$\langle b_0, v_0 \rangle = n_0$; $\langle b_1, v_1 \rangle = n_1$

if b_1 is neither *Call* nor *Exit* node **then**

foreach $n_1 \rightarrow n_2 \in E^\#$:

propagate($n_0 \rightarrow n_2$)

else if b_1 is *Call* node **then begin**

foreach call edge $n_1 \rightarrow n_2 \in E^\#$:

propagate($n_2 \rightarrow n_2$)

foreach non-call edge $n_1 \rightarrow n_2 \in E^\# \cup \text{SUMMARYINST}$:

propagate($n_0 \rightarrow n_2$)

end else if b_1 is *Exit* node **then begin**

foreach caller/return node pair b_i^c, b_i^r that calls b_0 **and** vars v_0, v_1 **do**

$n_s = \langle b_i^c, v_0 \rangle$; $n_r = \langle b_i^r, v_1 \rangle$

if $\{n_s \rightarrow n_0, n_0 \rightarrow n_1, n_1 \rightarrow n_r\} \subseteq E^\#$ **and not** $n_s \rightarrow n_r \in \text{SUMMARYINST}$ **then**

SUMMARYINST := **SUMMARYINST** $\cup \{n_s \rightarrow n_r\}$

foreach $n_z \rightarrow n_s \in \text{PATHEDGE}$:

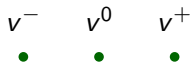
propagate(n_z, n_r)

end done end done end

Summary: IFDS Algorithm

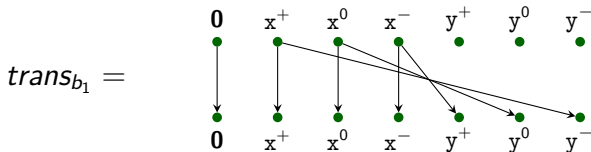
- ▶ Computes yes-or-no analysis on all variables
 - ▶ Original notion of 'variables' is slightly broader)
- ▶ Represents facts-of-interest as nodes $\langle b, v \rangle$:
 - ▶ b is node (basic block) in CFG
 - ▶ v is variable that we are interested in
- ▶ Uses
 - ▶ '*Exploded Supergraph*' $G^\#$
 - ▶ All CFGs in program in one graph
 - ▶ Plus interprocedural call edges
 - ▶ *Representation relations*
 - ▶ *Graph reachability*
 - ▶ *A worklist*
- ▶ Distinguishes between *Call* nodes, *Exit* nodes, others
- ▶ **Demand-driven**: only analyses what it needs
- ▶ **Whole-program analysis**
- ▶ **Computes Least Fixpoint on distributive frameworks**

Beyond True and False



- ▶ What if abstract domain is not boolean?
 - ▶ e.g., $\{\top, A^+, A^-, A^0, \perp\}$
- ▶ Multiple boolean properties per variable
 - ▶ easy for powerset lattice $\mathcal{P}(\{+, -, 0\})$
- ▶ *Limitation*: Transfer functions only depend on one variable
- ▶ Some problems not representable, others must adapt lattice

Consider $b_1 = \boxed{y := 0 - x}$:



This is how the algorithm was originally proposed

BONUS SLIDES

Extending IFDS?

- ▶ Not all analyses map well to IFDS
- ▶ Core ideas are appealing:
 - ▶ Automatically compute procedure summaries
 - ▶ Exploit graph reachability + worklist for *dependency tracking*

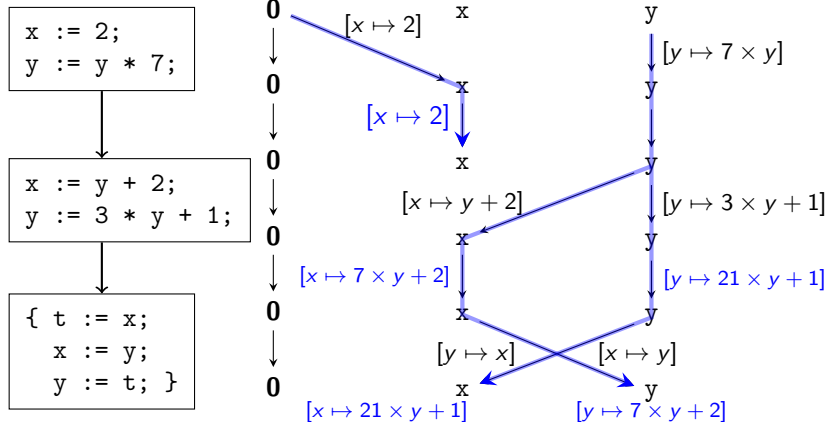
It is possible to extend this to other classes of problems

Linear Reaching Values

Statement	in_b	out_b
$x := 42$	M	$M \text{ with } [x \mapsto 42]$
$x := y + 1$	$M = \{[y \mapsto c], \dots\}$	$M \text{ with } [x \mapsto c + 1]$
$x := y * 7$	$M = \{[y \mapsto c], \dots\}$	$M \text{ with } [x \mapsto c \times 7]$
$x := y + z$	M	$M \text{ with } [x \mapsto \top]$

- ▶ “ $M \text{ with } [x \mapsto e]$ ” means “Remove from M any $[x \mapsto \dots]$ if it exists, and then add $[x \mapsto e]$ ”.
- ▶ The above sketches a *distributive* reaching values analysis
 - ▶ Each annotation of form $v_1 \mapsto c_1 \times v_2 + c_2$
 - ▶ Tradeoff: no support for adding / multiplying / ... (multiple variables)
- ▶ Encode in IFDS?

Labelling Graph Edges



- ▶ Extending IFDS to support information processing
- ▶ Carrying over key techniques:
 - ▶ *Track dependencies*
 - ▶ *Generate procedure summaries on the fly*

Representation

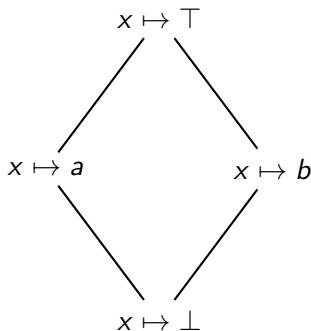
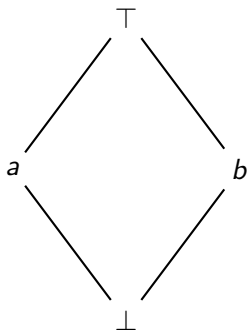
$$\left\{ \begin{array}{l} [x \mapsto c_{x,1} \times x + d_{x,1}] \\ [y \mapsto c_{y,1} \times y + d_{y,1}] \end{array} \right\} \circ \left\{ \begin{array}{l} [x \mapsto c_{x,2} \times v_1 + d_{x,2}] \\ [y \mapsto c_{y,2} \times v_2 + d_{y,2}] \end{array} \right\} \\ = \\ \left\{ \begin{array}{l} [x \mapsto (c_{x,2} \times c_{x,1}) \times v_1 + (d_{x,2} + c_{x,1} \times d_{x,1})] \\ [y \mapsto (c_{y,2} \times c_{y,1}) \times v_1 + (d_{y,2} + c_{y,1} \times d_{y,1})] \end{array} \right\}$$

- ▶ c_i, d_i : constants
- ▶ v_i : program variables
- ▶ (Maps of) linear functions are closed under composition
- ▶ Must support \sqcup to merge, map to \top on mismatch

$$\left\{ \begin{array}{l} [x \mapsto c_{x,1} \times v_1 + d_{x,1}] \\ [y \mapsto c_{y,1} \times v_3 + d_{y,1}] \end{array} \right\} \sqcup \left\{ \begin{array}{l} [x \mapsto c_{x,1} \times v_1 + d_{x,1}] \\ [y \mapsto c_{y,2} \times v_2 + d_{y,2}] \end{array} \right\} \\ = \\ \left\{ \begin{array}{l} [x \mapsto c_{x,1} \times x + d_{x,1}] \\ [y \mapsto \perp] \end{array} \right\}$$

Micro-Functions and Lattices

- Extend lattices to such 'Micro-Functions':



Micro-Functions, Efficient Representation

- ▶ Micro-Functions must support:

Encoding		$O(1)$ space
Computation	$f(x)$	$O(1)$ time
Equality testing	$f = f'$	$O(1)$ time
Composition	$f \circ f'$	$O(1)$ time
Meet	$f \sqcup f'$	$O(1)$ time

- ▶ Micro-functions are **efficiently representable** if they satisfy space / time constraints
 - ▶ Required for the algorithm's time bounds
- ▶ Other examples:
 - ▶ IFDS problems
 - ▶ Value bounds analysis

The IDE Algorithm (1/1)

- ▶ Interprocedural **D**istributive **E**nvironments algorithm
- ▶ Extends IFDS to 'labelled' edges as described above
- ▶ Assumes distributive framework over micro-functions
- ▶ Algorithmic changes:
 - ▶ First phase analogous to IFDS
 - ▶ Second phase applies computed functions to read out results
- ▶ Maintain/update mapping from path edges to micro-functions f :

$$\text{PATHEDGE} = \{ \langle b_0, v_0 \rangle \xrightarrow{f_0} \langle b_1, v_1 \rangle, \dots \}$$

- ▶ 'Missing edges' equivalent to $x \mapsto \perp$
- ▶ Initialise:

$$\text{PATHEDGE} = \{ \langle b_0, v_0 \rangle \xrightarrow{v_1 \mapsto \perp} \langle b_1, v_1 \rangle, \dots \}$$

- ▶ Always exactly one f per $\{ \langle b_0, v_0 \rangle \xrightarrow{f} \langle b_1, v_1 \rangle \} \in \text{PATHEDGE}$

The IDE Algorithm (2/2)

Procedure propagate($n_1 \rightarrow n_2$): -- IFDS version

begin

if $n_1 \rightarrow n_2 \in \text{PATHEDGE}$ **then**

return

$\text{PATHEDGE} := \text{PATHEDGE} \cup \{n_1 \rightarrow n_2\}$

$\text{WORKLIST} := \text{WORKLIST} \cup \{n_1 \rightarrow n_2\}$

end



Procedure propagate_{IDE}($n_1 \xrightarrow{f} n_2$): -- IDE version

begin

let $n_1 \xrightarrow{f'} n_2 \in \text{PATHEDGE}$

$f_{\text{upd}} := f \sqcup f'$

if $f_{\text{upd}} = f'$ **then**

return

$\text{PATHEDGE} := (\text{PATHEDGE} \setminus \{n_1 \xrightarrow{f'} n_2\}) \cup \{n_1 \xrightarrow{f_{\text{upd}}} n_2\}$

$\text{WORKLIST} := \text{WORKLIST} \cup \{n_1 \rightarrow n_2\}$

end

Summary

- ▶ IDE strictly generalises IFDS
- ▶ Utilises **Micro-Functions** to ensure efficient summaries:
 - ▶ Intra-procedural summaries via **PATHEDGE**
 - ▶ Inter-procedural procedure summaries via **SUMMARYINST**
- ▶ Runtime is $O(LED^3)$ if micro-functions are **efficiently representable**
 - ▶ L : Lattice height
 - ▶ IFDS: 1
 - ▶ IDE: length of longest descending chain
 - ▶ E : Number of control-flow edges
 - ▶ D : Number of variables
- ▶ IFDS supported by many popular dataflow frameworks