

Inter- vs. Intra-Procedural Analysis

- ▶ Intraprocedural: Within one procedure
- ▶ Interprocedural: Across multiple procedures

Limitations of Intra-Procedural Analysis

Teal-0

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

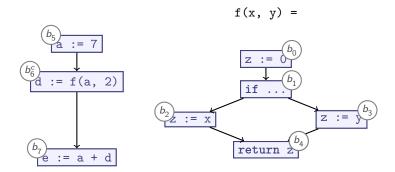
```
Teal-0

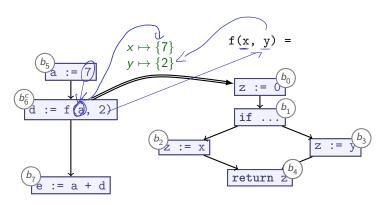
fun f(x, y) = {
  z := 0;
  if x > y {
```

```
if x > y {
   z := x;
} else {
   z := y;
}
return z;
}
```

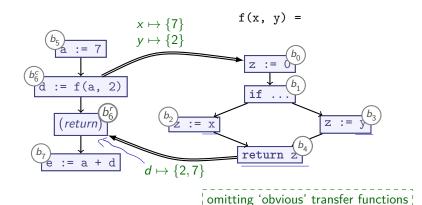
How can we compute Reachable Definitions here?

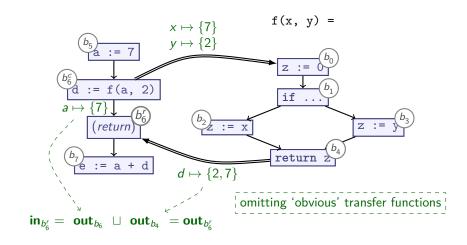
leaching

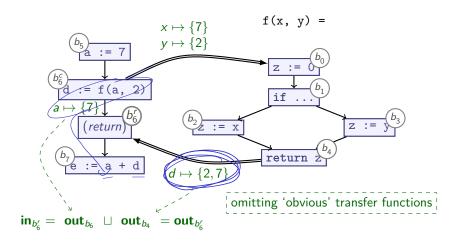




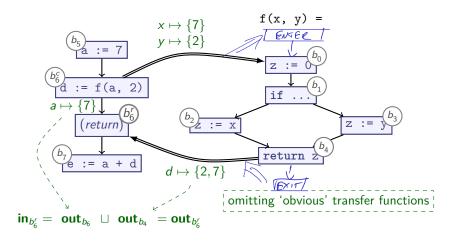
omitting 'obvious' transfer functions







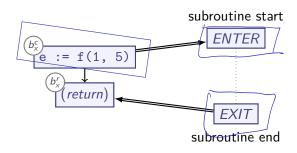
▶ **out**_{b_7}: $e \mapsto \{9, 14\}$



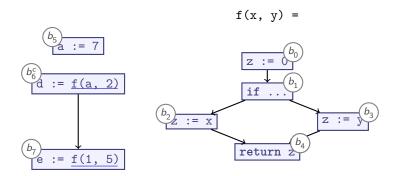
▶ **out**_{b_7}: $e \mapsto \{9, 14\}$

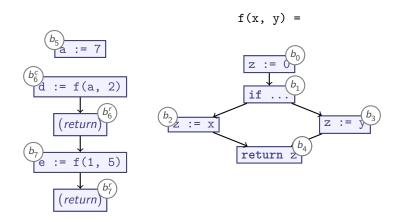
Works rather straightforwardly!

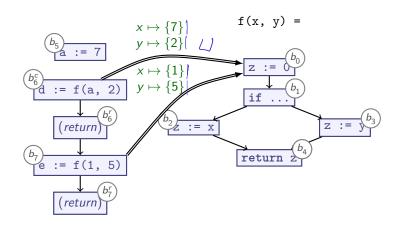
Inter-Procedural Data Flow Analysis

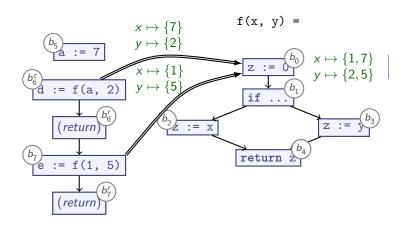


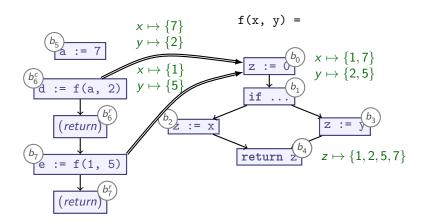
- ▶ Split call sites b_x into call (b_x^c) and return (b_x^r) nodes
- ▶ Intra-procedural edge $b_x^c \longrightarrow b_x^r$ carries environment/store
- ► Inter-procedural edge (→):
 - Caller → subroutine, substitutes parameters (for pass-by-value)
 - ► Caller ← return, substitutes result (for pass-by-result)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

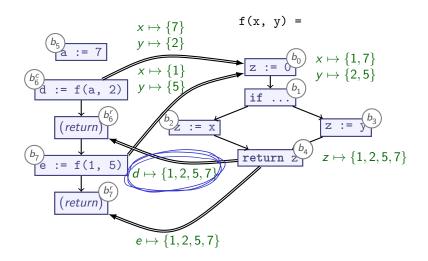


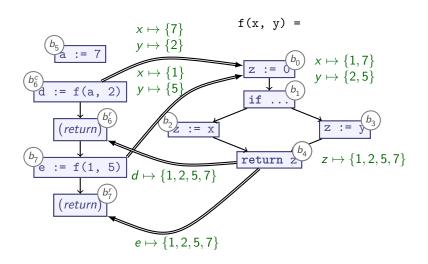




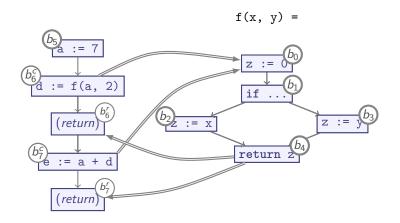


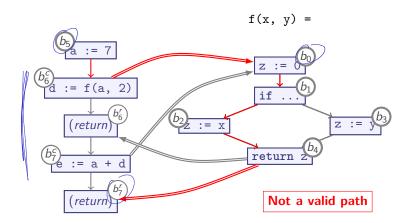


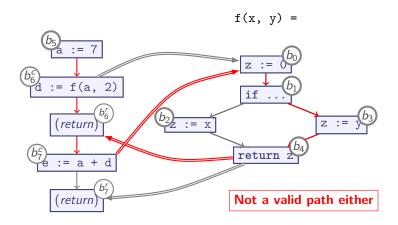


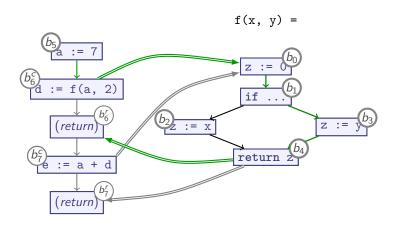


Imprecision!









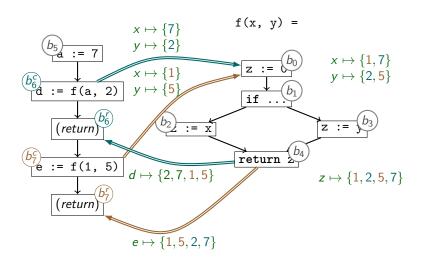
 $\triangleright [b_5, b_6^c, b_0, b_1, b_3, b_4, b_6^r]$

Context-sensitive interprocedural analyses consider only valid paths

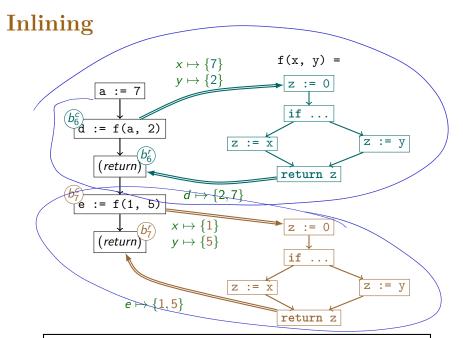
Summary

- 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

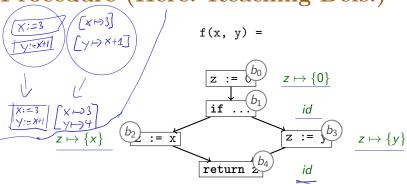
Interprocedural Data Flow Analysis



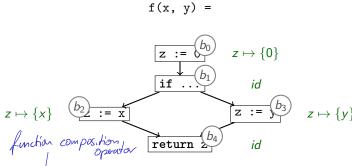
Context-insensitive: analysis merges all callers to f()



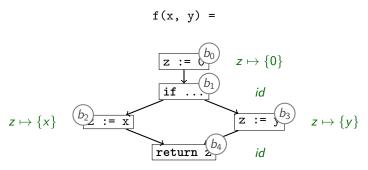
Clone subroutine IRs for each calling context



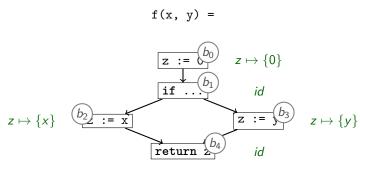
► Compose transfer functions:



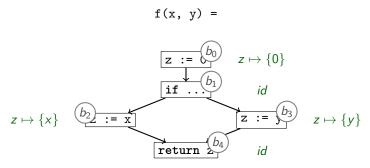
- ► Compose transfer functions:
 - ightharpoonup $trans_{b_0}$ $\circ trans_{b_1} = [z \mapsto 0]$



- ► Compose transfer functions:
 - ightharpoonup trans $b_0 \circ trans_{b_1} = [z \mapsto 0]$
 - ▶ $trans_{b_0} \circ trans_{b_1} \circ trans_{b_2} = [z \mapsto \{x\}]$

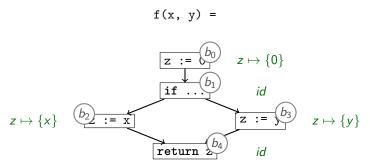


- ► Compose transfer functions:
 - ightharpoonup trans_{b1} \circ trans_{b1} = $[z \mapsto 0]$
 - ightharpoonup trans_{b1} \circ trans_{b2} = $[z \mapsto \{x\}]$
 - $\blacktriangleright trans_{b_0} \circ trans_{b_1} \circ trans_{b_3} = [z \mapsto \{y\}]$



- Compose transfer functions:
 - ightharpoonup trans_{b1} $= [z \mapsto 0]$
 - ▶ $trans_{b_1} \circ trans_{b_1} \circ trans_{b_2} = [z \mapsto \{x\}]$

 - ► $trans_{b_0} \circ trans_{b_1} \circ trans_{b_3} = [z \mapsto \{y\}]$ ► $trans_{b_0} \circ trans_{b_1} \circ (trans_{b_2} \not w trans_{b_3}) = [z \mapsto \{x, y\}]$



- ► Compose transfer functions:
 - ightharpoonup trans $b_0 \circ trans_{b_1} = [z \mapsto 0]$
 - $\blacktriangleright trans_{b_0} \circ trans_{b_1} \circ trans_{b_2} = [z \mapsto \{x\}]$
 - ightharpoonup trans_{b₁} \circ trans_{b₃} = $[z \mapsto \{y\}]$
 - $trans_{b_0} \circ trans_{b_1} \circ (trans_{b_2} \not \neg trans_{b_3}) = [z \mapsto \{x, y\}]$
 - ▶ $trans_{b_0} \circ trans_{b_1} \circ (trans_{b_2} \bowtie trans_{b_3}) \circ trans_{b_4} = [z \mapsto \{x, y\}]$

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

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

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

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

- ▶ Use transf as transfer function for f(), discard f's body
- Advantages:
 - Can yield compact subroutine descriptions
 - Can speed up call site analysis dramatically

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

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

- ▶ Use transf 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

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

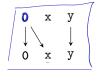
$$trans_f = [\mathbf{return} \mapsto \{x, y\}]$$

- ▶ Use transf 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:
 - transf is symbolic expression as complex as f itself

Representation Relations

Example procedure summary representation:

```
\underline{x} := \underline{\text{null}};
\underline{y} := \underline{y};
```



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

'May be null' analysis

- P(v): v may be null
- P(0) always holds

Representation Relations

Example procedure summary representation:

$$\begin{array}{cccc}
\bullet & x & \lambda & \bullet \\
& & & & & \\
\bullet & & & & \\
\bullet & & & \\
\bullet & & & \\
\bullet &$$

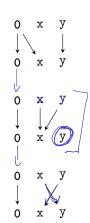


'May be null' analysis

- ▶ P(v): v may be null
- P(0) always holds
- if $P(c) \in [\underline{\mathbf{in}_b}]$ then $P(d) \in \underline{\mathbf{out}_b}$

Representation Relations

Example procedure summary representation:



'May be null' analysis

- P(v): v may be null
- $P(\mathbf{0})$ always holds
- ► $c \longrightarrow d$: if $P(c) \in \mathbf{in}_b$ then $P(d) \in \mathbf{out}_b$

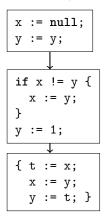
Summary

- 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





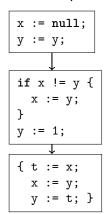


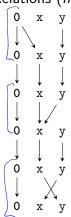


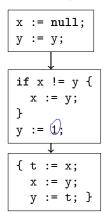


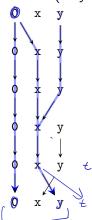




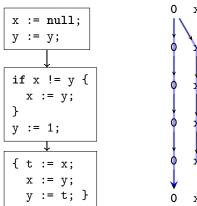


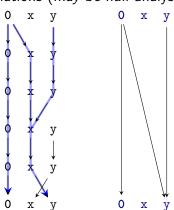




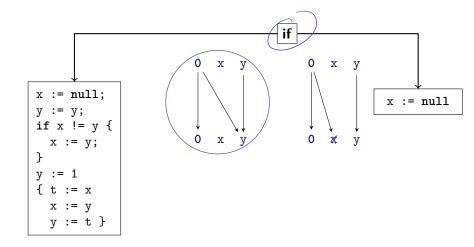


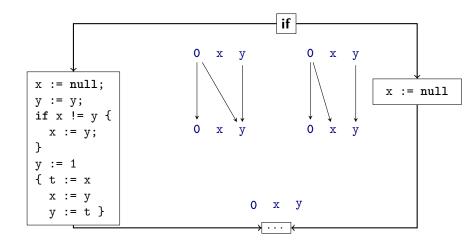
Recall Representation Relations (may be null analysis):

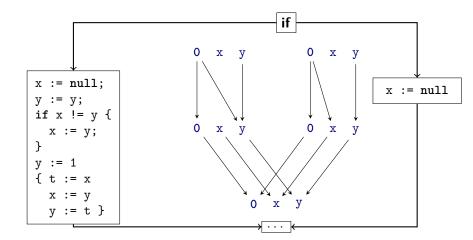


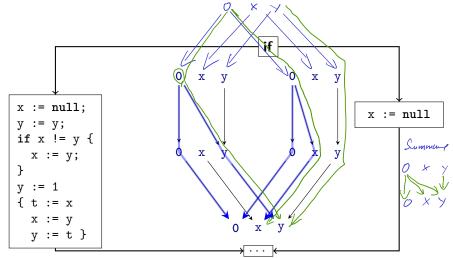


Composed representation relations are again representation relations









Dataflow via Graph Reachability

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

- ▶ Assume binary latice $(\{\top,\bot\},\sqsubseteq,\sqcap,\sqcup)$
 - ▶ $a \sqcup b = \top$ iff $a = \bot$ and $b = \bot$, otherwise $a \sqcup b = \top$
 - 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 Finite Distributive Subset problems)
- 'Exploded Supergraph': $G^{\sharp} = (N^{\sharp}, E^{\sharp})$
 - $\blacktriangleright N^{\sharp} = N_{\mathsf{CFG}} \times \mathcal{V} \cup \{0\}$
 - ► Plus parameter/return call edges



- $\blacktriangleright b_{\text{main}}^{s}$ is the CFG *ENTER* node of the main entry point
- Property-of-interest holds if reachable from $\langle b_{\text{main}}^s, \mathbf{0} \rangle$
- ► 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 (b_0, v_0) (b_3, v_0) PATHEDGE edge

WORKLIST edge All WORKLIST edges are also PATHEDGE edges

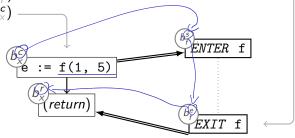
Result of our analysis



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

IFDS Strategy

- Algorithm distinguishes between three types of nodes:
 - ► Exit nodes (b_f)
 - ► Call nodes (b_x)
 - ▶ Other nodes



On-demand processing

```
Procedure propagate (n_1 \rightarrow n_2):

begin

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

return

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

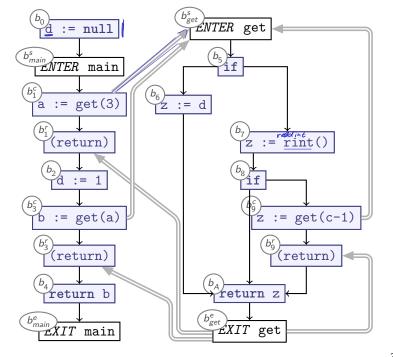
WORKLIST: = WORKLIST \cup \{n_1 \rightarrow n_2\}

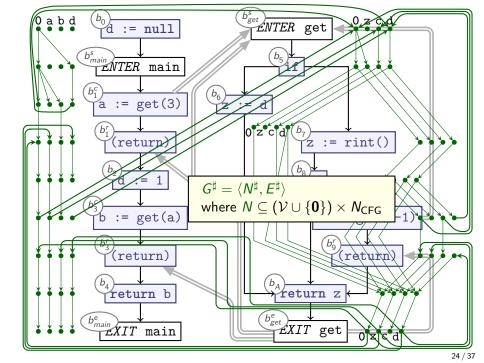
end
```

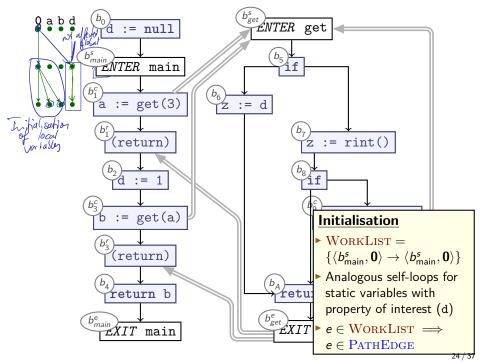
Running Example

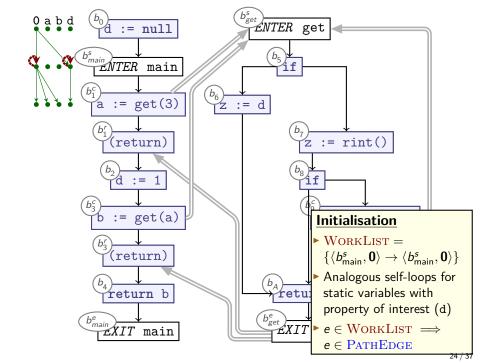
Teal-0: main() var default := null; fun main() = { var a := get(3); default := 1; var b := get(3); return b; }

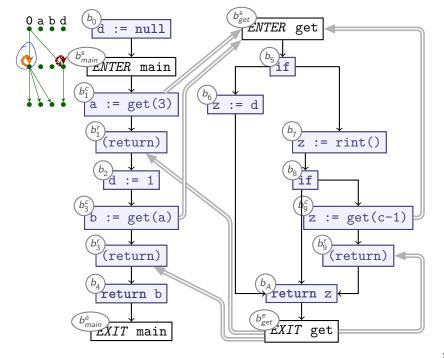
```
Teal-0: get()
fun get(c) = {
  if c == 0 {
     z := default;
  } else {
     z := read int();
     if z < 0 {
       z := get(c - 1);
   return z;
```

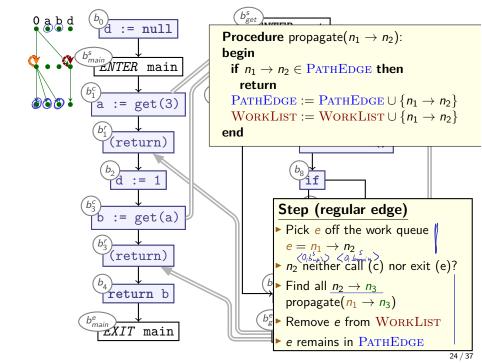


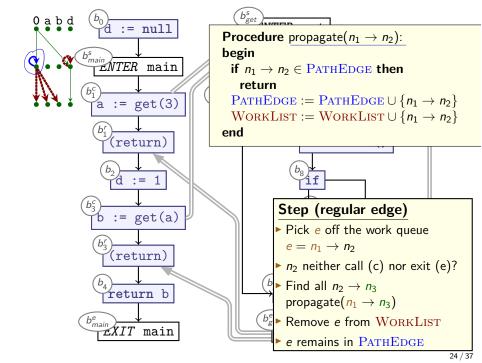


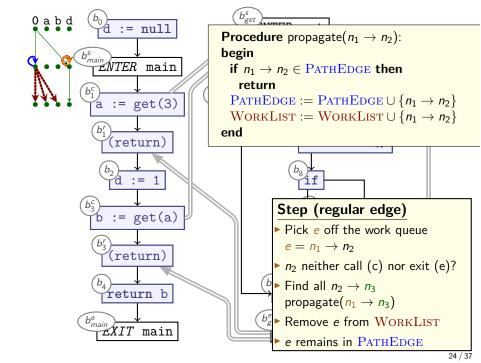


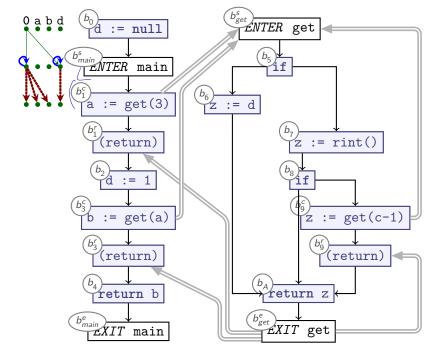


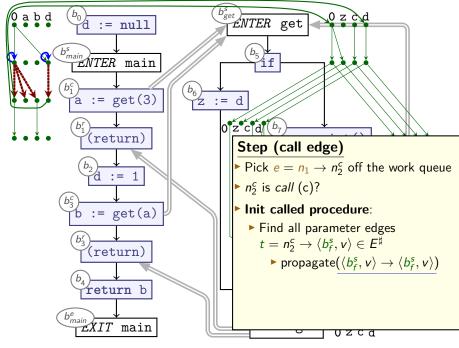


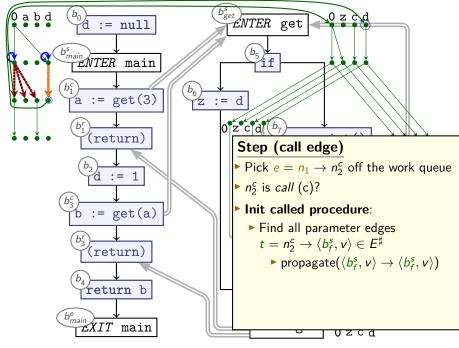


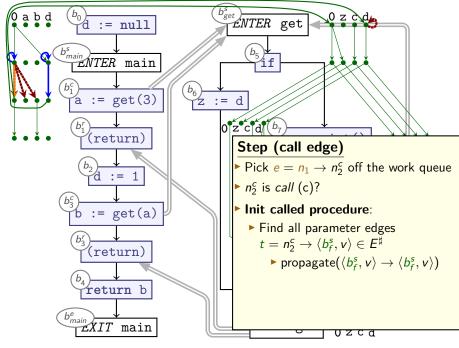


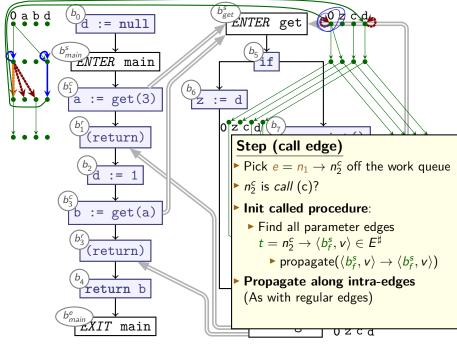


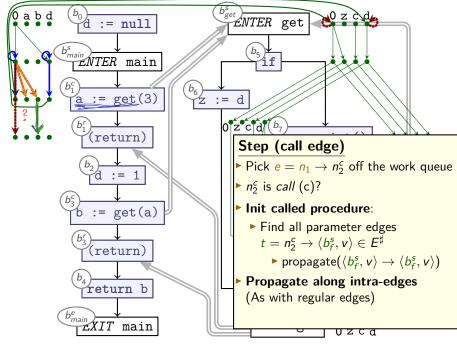


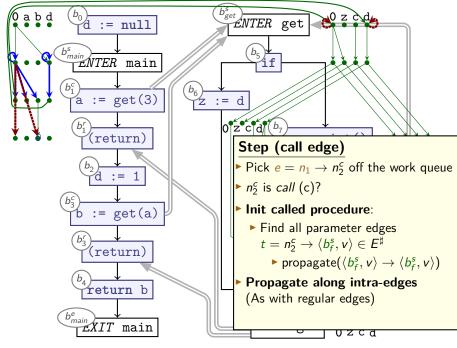


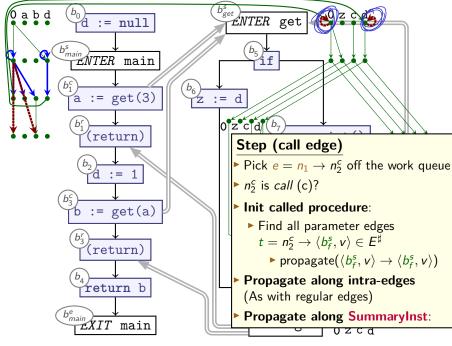


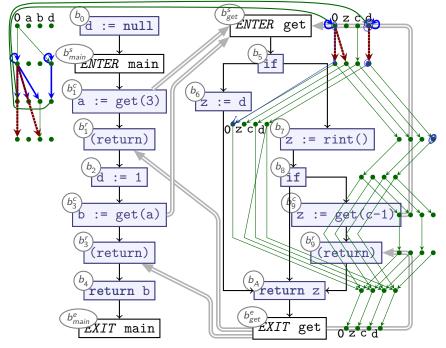


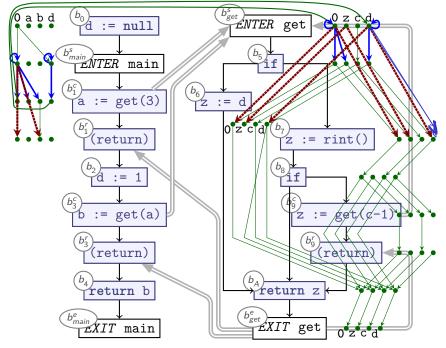


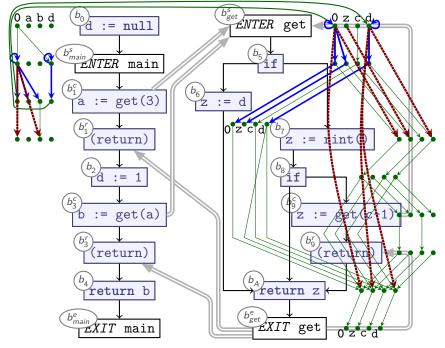


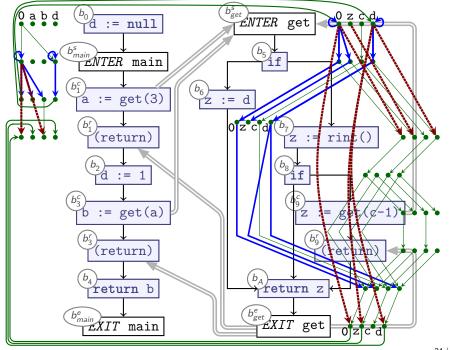


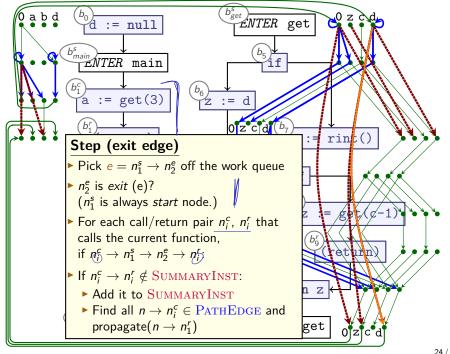


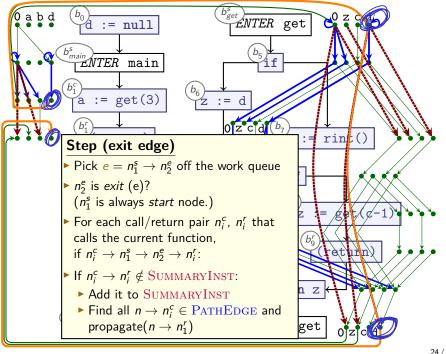


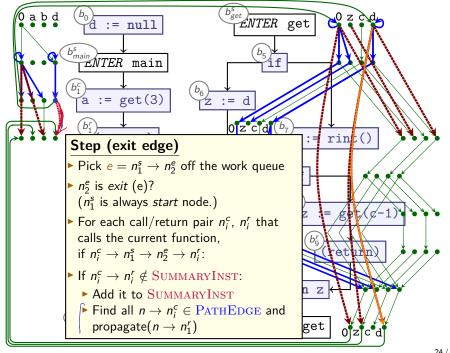


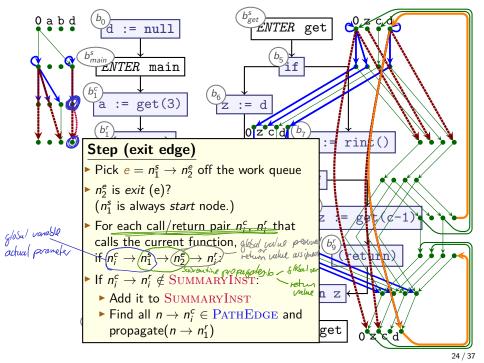


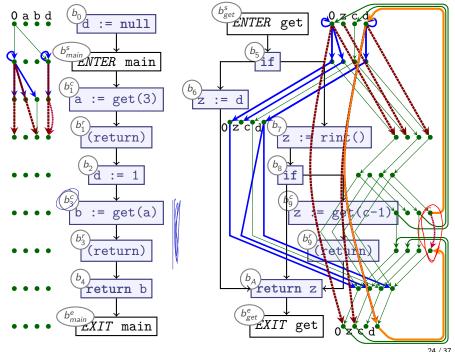


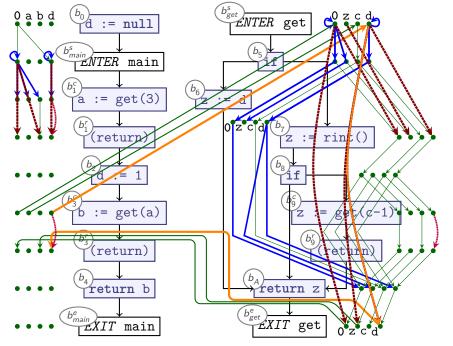


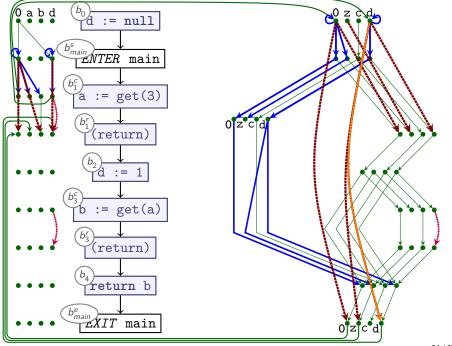


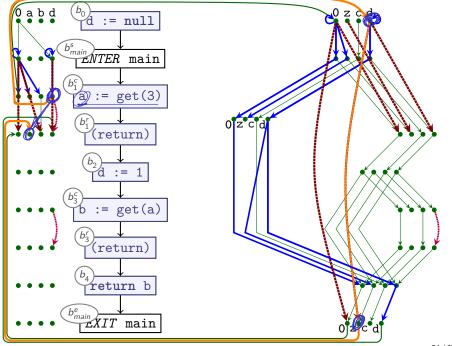


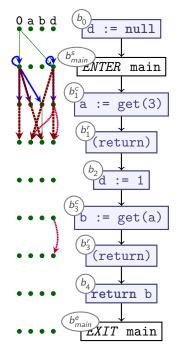


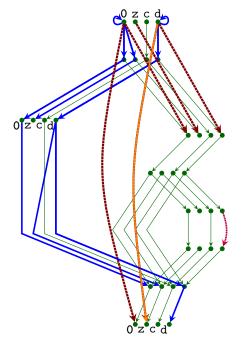


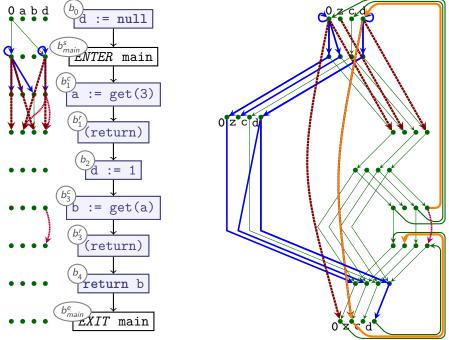


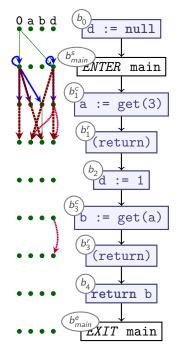


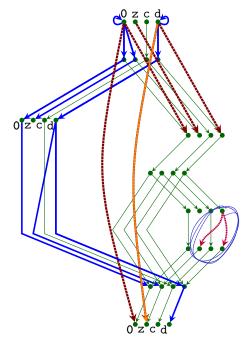


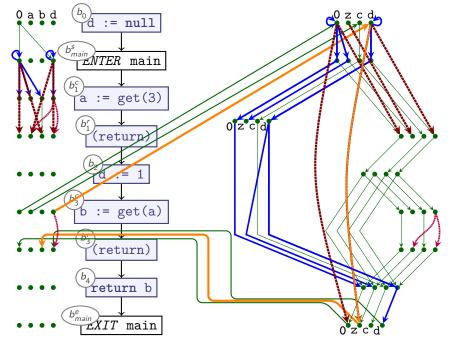


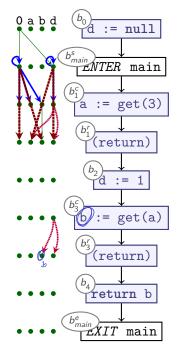


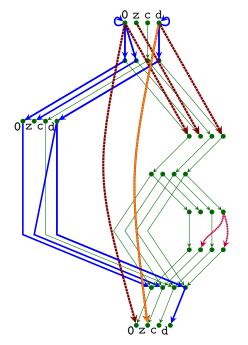


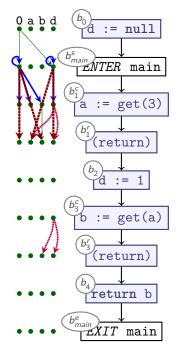


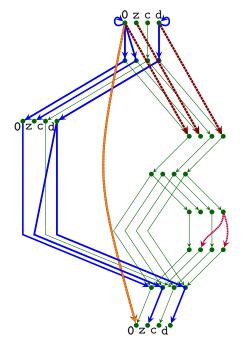


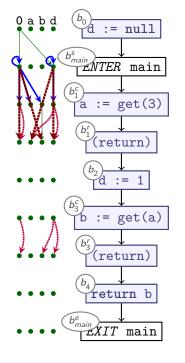


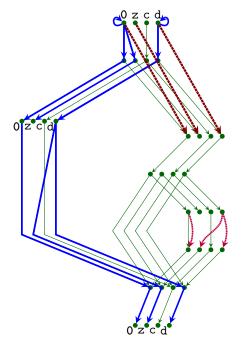


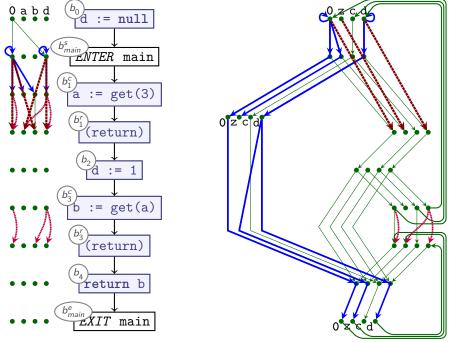


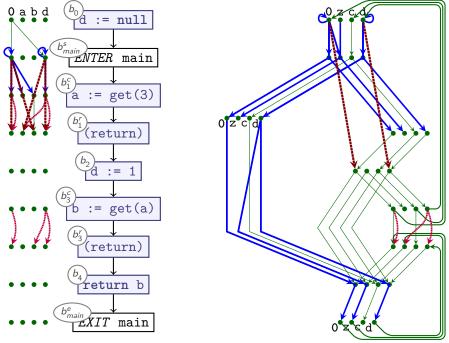


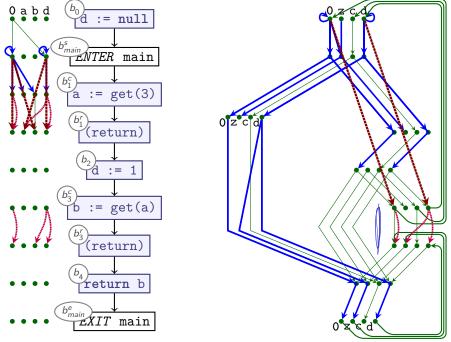


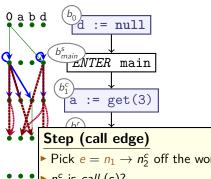




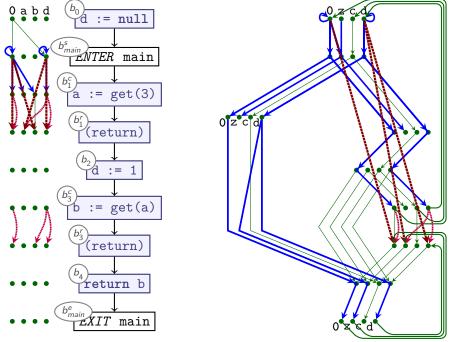


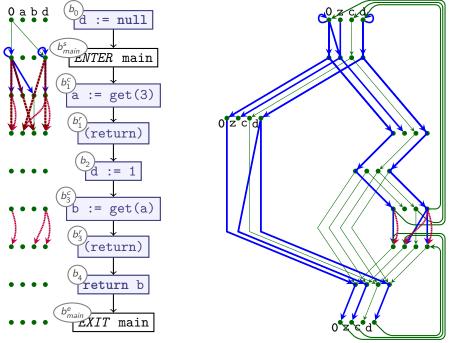


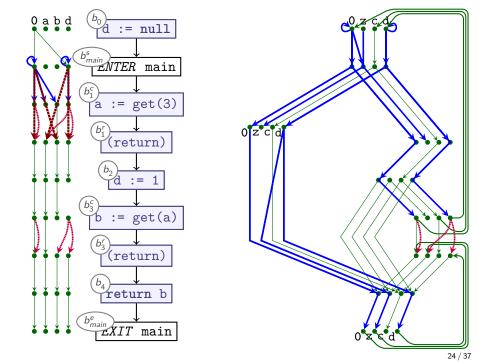


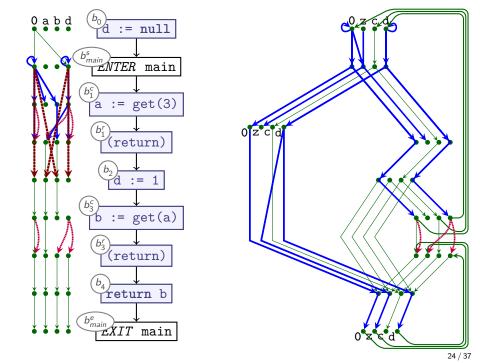


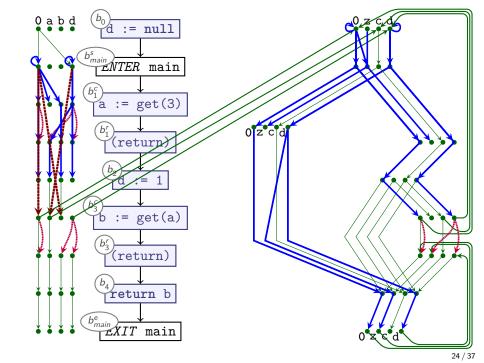
- ▶ Pick $e = n_1 \rightarrow n_2^c$ off the work queue
- \triangleright n_2^c is call (c)?
 - Init called procedure:
 - Find all parameter edges $t = n_2^c \rightarrow \langle b_f^s, v \rangle \in E^{\sharp}$
 - ▶ propagate($\langle b_f^s, v \rangle \rightarrow \langle b_f^s, v \rangle$)
 - Propagate along intra-edges (As with regular edges)
 - Propagate along SummaryInst:
 - (As with regular edges)

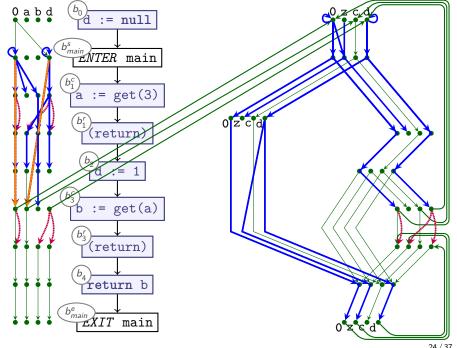


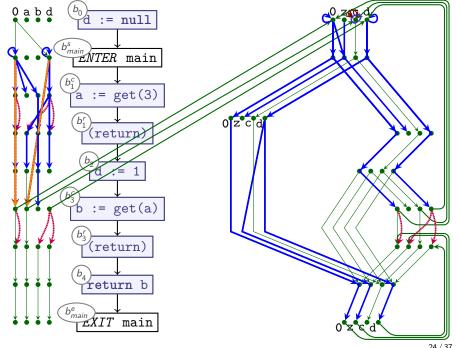


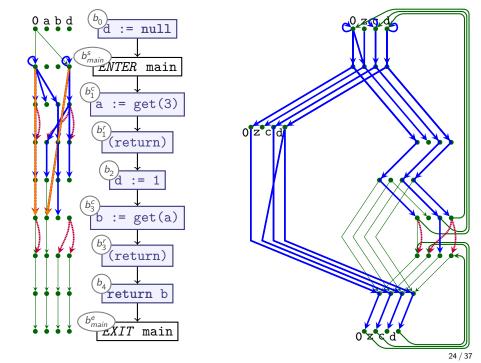


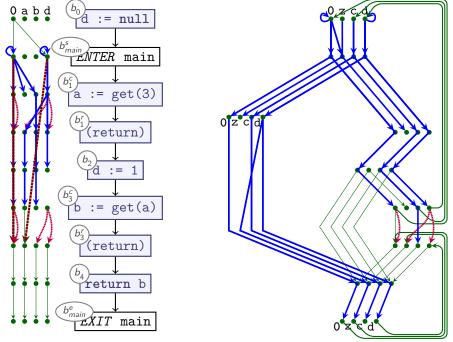


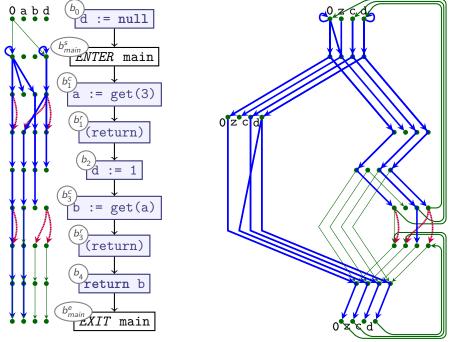


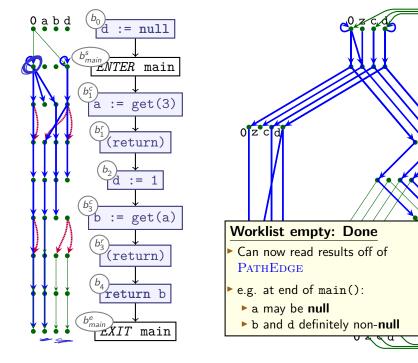












The IFDS Algorithm: Initialisation and Propagation)

```
Procedure Init():
begin
  WorkList := PathEdge := \emptyset
  propagate(\langle b_{main}^s, \mathbf{0} \rangle \rightarrow \langle b_{main}^s, \mathbf{0} \rangle)
  ForwardTabulate()
end
Procedure propagate(n_1 \rightarrow n_2):
begin
  if n_1 \rightarrow n_2 \in \text{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 \ \{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^{\sharp}:
        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^{\sharp}:
        propagate(n_2 \rightarrow n_2)
      foreach non-call edge n_1 \rightarrow n_2 \in E^{\sharp} \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^c, v_1 \rangle
        if \{n_s \to n_0, n_0 \to n_1, n_1 \to n_r\} \subset E^{\sharp} and not n_s \to n_r \in \text{SUMMARYINST} then
          SUMMARYINST := SUMMARYINST \cup \{n_s \rightarrow n_r\}
          foreach n_z \rightarrow n_s \in 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., $\{\top, A^+, A^-, A^0, \bot\}$

Beyond True and False

$$v^ v^0$$
 v^+

- What if abstract domain is not boolean?
 - e.g., $\{\top, A^+, A^-, A^0, \bot\}$
- ▶ 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 $b_1 = y := 0 x$:

Beyond True and False

$$v^ v^0$$
 v^+

- ▶ What if abstract domain is not boolean?
 - e.g., $\{\top, A^+, A^-, A^0, \bot\}$
- ► 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 $b_1 = y := 0 x$:

Beyond True and False

$$v^ v^0$$
 v^+

- ▶ What if abstract domain is not boolean?
 - e.g., $\{\top, A^+, A^-, A^0, \bot\}$
- ▶ 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 $b_1 = y := 0 x$:

Extending IFDS?

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

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

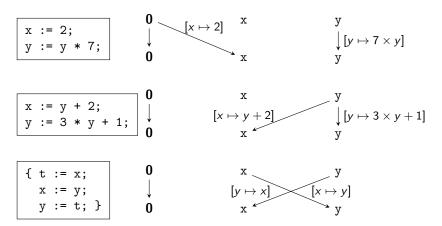
Statement	\mid in $_b$	out _b
	М	$\{[x \mapsto 42]\} \cup (M \setminus [x \mapsto _])$
x := y + 1	$M = \{[y \mapsto c], \ldots\}$	$\{[x \mapsto c+1]\} \cup (M \setminus [x \mapsto _])$
x := y * 7	$M = \{[y \mapsto c], \ldots\}$	$\{[x \mapsto c \times 7]\} \cup (M \setminus [x \mapsto _])$
x := y + z	M	$\{[x \mapsto \bot]\} \cup (M \setminus [x \mapsto _])$

Linear Reaching Values

Statement	in_b	out_b
		$\{[x \mapsto 42]\} \cup (M \setminus [x \mapsto _])$
x := y + 1	$M = \{[y \mapsto c], \ldots\}$	$\{[x \mapsto c+1]\} \cup (M \setminus [x \mapsto _])$
x := y * 7	$M = \{[y \mapsto c], \ldots\}$	$\{[x \mapsto c \times 7]\} \cup (M \setminus [x \mapsto _])$
x := y + z	M	$\{[x \mapsto \bot]\} \cup (M \setminus [x \mapsto \bot])$

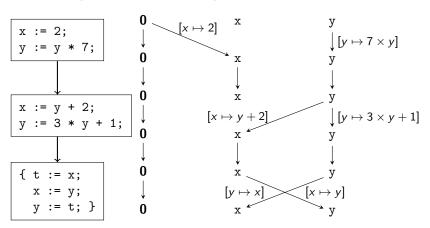
- ▶ The above sketches a *distributive* reaching values analysis
 - ▶ Each annotation of form $v_1 \mapsto c_1 \times v_2 + c_2$
 - ► Tradeoff: no support for adding / multiplying / ... (multiple variables)
- ► Encode in IFDS?

Labelling Graph Edges



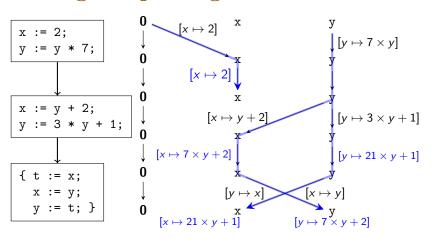
- ▶ Extending IFDS to support information processing
- Carrying over key techniques:
 - Track dependencies
 - ► Generate procedure summaries on the fly

Labelling Graph Edges



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

Labelling Graph Edges



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

Representation

$$\begin{cases}
[x \mapsto c_{x,1} \times x + d_{x,1}] \\
[y \mapsto c_{y,1} \times y + d_{y,1}]
\end{cases} \circ \begin{cases}
[x \mapsto c_{x,2} \times v_1 + d_{x,2}] \\
[y \mapsto c_{y,2} \times v_2 + d_{y,2}]
\end{cases}$$

$$= \begin{cases}
[x \mapsto (c_{x,2} \times c_{x,1}) \times v_1 + (d_{x,2} + c_{x_1} \times d_{x_1})] \\
[y \mapsto (c_{y,2} \times c_{y,1}) \times v_1 + (d_{y,2} + c_{y_1} \times d_{y_1})]
\end{cases}$$

- $ightharpoonup c_i, d_i$: constants
- ▶ v_i: program variables

Representation

$$\begin{cases}
[x \mapsto c_{x,1} \times x + d_{x,1}] \\
[y \mapsto c_{y,1} \times y + d_{y,1}]
\end{cases} \circ \begin{cases}
[x \mapsto c_{x,2} \times v_1 + d_{x,2}] \\
[y \mapsto c_{y,2} \times v_2 + d_{y,2}]
\end{cases}$$

$$= \begin{cases}
[x \mapsto (c_{x,2} \times c_{x,1}) \times v_1 + (d_{x,2} + c_{x_1} \times d_{x_1})] \\
[y \mapsto (c_{y,2} \times c_{y,1}) \times v_1 + (d_{y,2} + c_{y_1} \times d_{y_1})]
\end{cases}$$

- $ightharpoonup c_i, d_i$: constants
- ▶ v_i: program variables
- ► (Maps of) linear functions are closed under composition

Representation

$$\begin{cases}
[x \mapsto c_{x,1} \times x + d_{x,1}] \\
[y \mapsto c_{y,1} \times y + d_{y,1}]
\end{cases} \circ \begin{cases}
[x \mapsto c_{x,2} \times v_1 + d_{x,2}] \\
[y \mapsto c_{y,2} \times v_2 + d_{y,2}]
\end{cases}$$

$$= \begin{cases}
[x \mapsto (c_{x,2} \times c_{x,1}) \times v_1 + (d_{x,2} + c_{x_1} \times d_{x_1})] \\
[y \mapsto (c_{y,2} \times c_{y,1}) \times v_1 + (d_{y,2} + c_{y_1} \times d_{y_1})]
\end{cases}$$

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

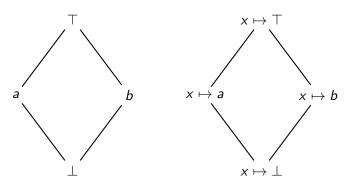
$$\left\{
\begin{bmatrix}
x \mapsto c_{x,1} \times v_1 + d_{x,1} \\
y \mapsto c_{y,1} \times v_3 + d_{y,1}
\end{bmatrix}
\right\} \sqcup
\left\{
\begin{bmatrix}
x \mapsto c_{x,1} \times v_1 + d_{x,1} \\
y \mapsto c_{y,2} \times v_2 + d_{y,2}
\end{bmatrix}
\right\}$$

$$=
\left\{
\begin{bmatrix}
x \mapsto c_{x,1} \times v_1 + d_{x,1} \\
y \mapsto c_{y,2} \times v_2 + d_{y,2}
\end{bmatrix}
\right\}$$

$$=
\left\{
\begin{bmatrix}
x \mapsto c_{x,1} \times x + d_{x,1} \\
y \mapsto \bot
\end{bmatrix}
\right\}$$

Micro-Functions and Lattices

Extend lattices to such 'Micro-Functions':



Micro-Functions, Efficient Representation

Micro-Functions must support:

```
Encoding
Computation f(x)
Equality testing f = f'
Composition f \circ f'
Meet f \sqcup f'
```

- ► Other examples:
 - ▶ IFDS problems
 - ► Value bounds analysis

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 Distributive Environments 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*:

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

- 'Missing edges' equivalent to $x \mapsto \bot$
- ► Initialise:

PATHEDGE =
$$\{\langle b_0, v_0 \rangle \stackrel{v_1 \mapsto \bot}{\longrightarrow} \langle b_1, v_1 \rangle, \ldots \}$$

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

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
  PATHEDGE := PATHEDGE \cup \{n_1 \rightarrow n_2\}
  WORKLIST := WORKLIST \cup \{n_1 \rightarrow n_2\}
end
Procedure propagate<sub>IDE</sub>(n_1 \stackrel{f}{\rightarrow} n_2): -- IDE version
begin
  let n_1 \stackrel{f'}{\rightarrow} n_2 \in \text{PATHEDGE}
  f_{\text{und}} := f \sqcup f'
  if f_{upd} = f' then
    return
  PATHEDGE := (PATHEDGE \setminus \{n_1 \stackrel{f'}{\rightarrow} n_2\}) \cup \{n_1 \stackrel{f_{upd}}{\rightarrow} n_2\}
  WORKLIST := 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