

Lecture 9: Interprocedural Analysis

17-355/17-655/17-819: Program Analysis

Rohan Padhye and Jonathan Aldrich

March 4, 2021

* Course materials developed with Claire Le Goues

Extend WHILE with functions

Extend WHILE3ADDR with functions

$$\begin{array}{lcl} F & ::= & \mathbf{fun}\ f(x)\ \{\ \overline{n : I}\ \} \\ I & ::= & \dots \mid \mathbf{return}\ x \mid y := f(x) \end{array}$$

Extend WHILE3ADDR with functions

$F ::= \text{fun } f(x) \{ \overline{n : I} \}$
 $I ::= \dots \mid \text{return } x \mid y := f(x)$

1 : $\text{fun } double(x) : int$
2 : $y := 2 * x$
3 : $\text{return } y$
4 : $\text{fun } main() : void$
5 : $z := 0$
6 : $w := double(z)$

Extend WHILE3ADDR with functions

```
1 : fun divByX(x) : int  
2 :   y := 10/x  
3 :   return y  
  
4 : fun main() : void  
5 :   z := 5  
6 :   w := divByX(z)
```

```
1 : fun double(x) : int  
2 :   y := 2 * x  
3 :   return y  
  
4 : fun main() : void  
5 :   z := 0  
6 :   w := double(z)
```

Data-Flow Analysis

HOW DO WE ANALYZE THESE PROGRAMS?

Approach #1: Analyze functions independently

- Pretend function $f()$ cannot see the source of function $g()$
- Simulates separate compilation and dynamic linking (e.g. C, Java)
- Create CFG for each function body and run **intraprocedural** analysis
- **Q:** What should be is σ_0 and $f_z[x := g(y)]$ and $f_z[\text{return } x]$ for zero analysis?

$$\sigma_0 =$$

$$f[x := g(y)](\sigma) =$$

$$f[\text{return } x](\sigma) =$$

Can we show that division on line 2 is safe?

```
1 : fun divByX(x) : int
2 :     y := 10/x
3 :     return y
4 : fun main() : void
5 :     z := 5
6 :     w := divByX(z)
```

Approach #2: User-defined Annotations

$\text{@NonZero} \rightarrow \text{@NonZero}$

```
1 : fun divByX(x) : int
2 :     y := 10/x
3 :     return y
4 : fun main() : void
5 :     z := 5
6 :     w := divByX(z)
```

$$f[x := g(y)](\sigma) = \sigma[x \mapsto \text{annot}[g].r] \quad (\text{error if } \sigma(y) \not\models \text{annot}[g].a)$$

$$f[\text{return } x](\sigma) = \sigma \quad (\text{error if } \sigma(x) \not\models \text{annot}[g].r)$$

Approach #2: User-defined Annotations

$\text{@NonZero} \rightarrow \text{@NonZero}$

```
1 : fun divByX(x) : int  
2 :   y := 10/x  
3 :   return y  
  
4 : fun main() : void  
5 :   z := 5  
6 :   w := divByX(z)
```

$\text{@NonZero} \rightarrow \text{@NonZero}$

```
1 : fun double(x) : int  
2 :   y := 2 * x  
3 :   return y  
  
4 : fun main() : void  
5 :   z := 0  
6 :   w := double(z) Error!
```

$$f[x := g(y)](\sigma) = \sigma[x \mapsto \text{annot}[g].r] \quad (\text{error if } \sigma(y) \not\models \text{annot}[g].a)$$

$$f[\text{return } x](\sigma) = \sigma \quad (\text{error if } \sigma(x) \not\models \text{annot}[g].r)$$

Approach #2: User-defined Annotations

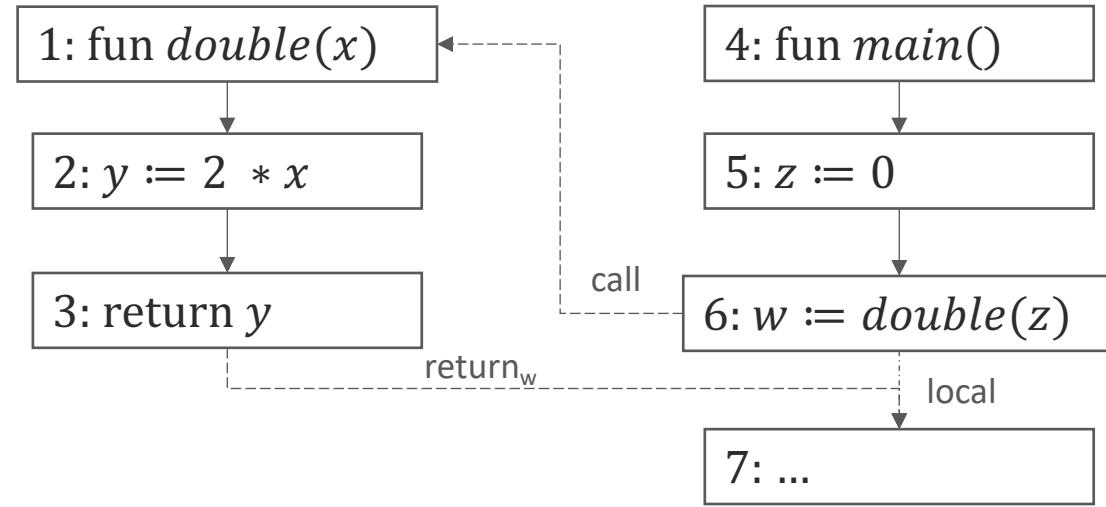
```
@NonZero -> @NonZero
1 : fun divByX(x) : int
2 :   y := 10/x
3 :   return y
4 : fun main() : void
5 :   z := 5
6 :   w := divByX(z)
```

```
@Any -> @NonZero
1 : fun double(x) : int
2 :   y := 2 * x
3 :   return y Error!
4 : fun main() : void
5 :   z := 0
6 :   w := double(z)
```

$$f[x := g(y)](\sigma) = \sigma[x \mapsto \text{annot}[g].r] \quad (\text{error if } \sigma(y) \not\models \text{annot}[g].a)$$

$$f[\text{return } x](\sigma) = \sigma \quad (\text{error if } \sigma(x) \not\models \text{annot}[g].r)$$

Approach #3: Interprocedural CFG



```
1 : fun double(x) : int
2 :     y := 2 * x
3 :     return y
4 : fun main() : void
5 :     z := 0
6 :     w := double(z)
7: ...
```

$$f_z[x := g(y)]_{local}(\sigma) = \sigma \setminus (\{x\} \cup Globals)$$

$$f_z[x := g(y)]_{call}(\sigma) = \{formal(g) \rightarrow \sigma(y)\}$$

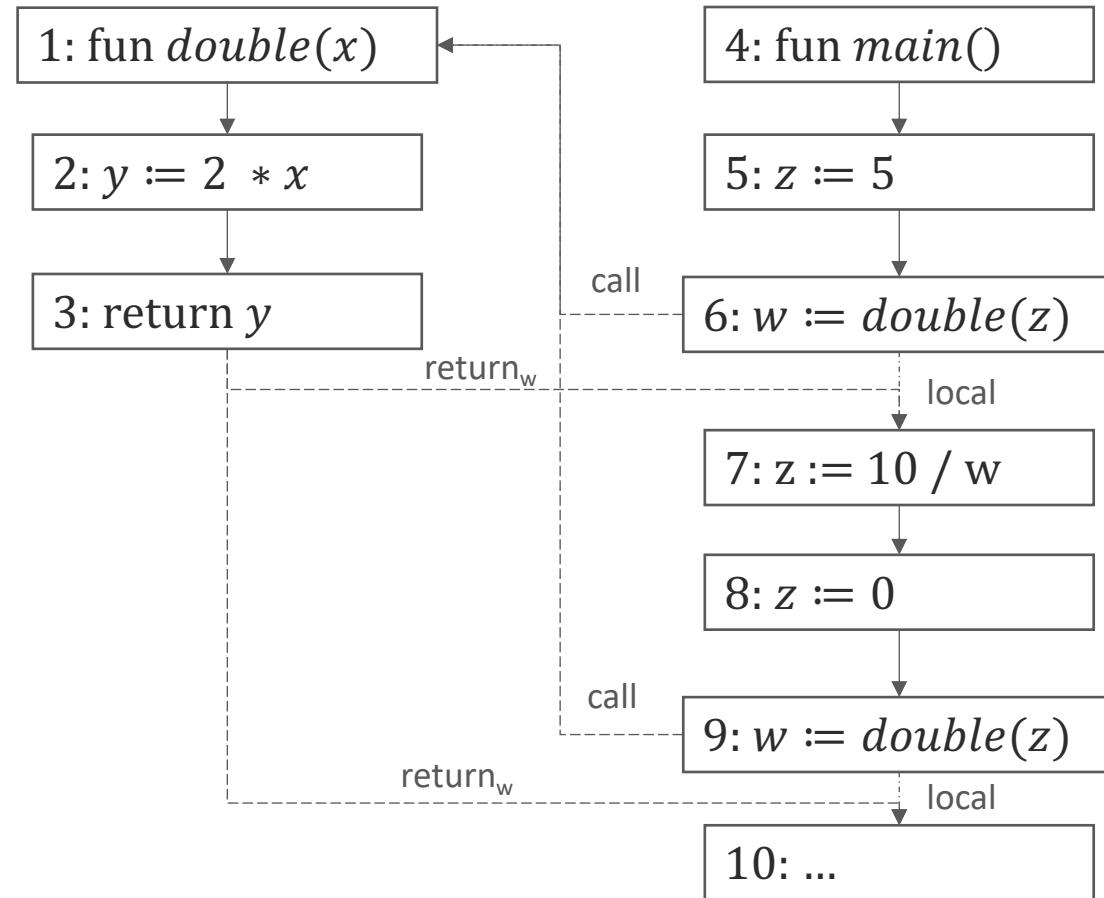
$$f_z[\text{return } x]_{ret}(\sigma) = \{z \rightarrow \sigma(z) \mid z \in Globals\} \cup \{\text{ret} \rightarrow \sigma(x)\}$$

Approach #3: Interprocedural CFG

Exercise: What would be the result of zero analysis for this program on line 7 and at the end?

```
1 : fun double(x) : int
2 :     y := 2 * x
3 :     return y
4 : fun main()
5 :     z := 5
6 :     w := double(z)
7 :     z := 10/w
8 :     z := 0
9 :     w := double(z)
```

Approach #3: Interprocedural CFG



```
1 : fun double(x) : int
2 :   y := 2 * x
3 :   return y
4 : fun main()
5 :   z := 5
6 :   w := double(z)
7 :   z := 10 / w
8 :   z := 0
9 :   w := double(z)
10: ...
```

Problems with Interprocedural CFG

- Merges (joins) information across call sites to same function
- Loses precision
- Models infeasible paths (call from one site and return to another)
- Can we “remember” where to return data-flow values?

Enter:

CONTEXT-SENSITIVE ANALYSIS

Context-Sensitive Analysis Example

```
1 : fun double(x) : int
2 :     y := 2 * x
3 :     return y
4 : fun main()
5 :     z := 5
6 :     w := double(z)
7 :     z := 10/w
8 :     z := 0
9 :     w := double(z)
```

Key idea: Separate analyses for functions called in different "contexts".

("context" = some statically definable condition)

Context-Sensitive Analysis Example

```
1 : fun double(x) : int
2 :     y := 2 * x
3 :     return y
4 : fun main()
5 :     z := 5
6 :     w := double(z)
7 :     z := 10/w
8 :     z := 0
9 :     w := double(z)
```

Context	σ_{in}	σ_{out}
Line 6	{x->N}	{x->N, y->N}
Line 9	{x->Z}	{x->Z, y->Z}

Context-Sensitive Analysis Example

```
1 : fun double(x) : int
2 :     y := 2 * x
3 :     return y
4 : fun main()
5 :     z := 5
6 :     w := double(z)
7 :     z := 10/w
8 :     z := 0
9 :     w := double(z)
```

Context	σ_{in}	σ_{out}
<main, T>	T	{w->Z, Z->Z}
<double, N>	{x->N}	{x->N, y->N}
<double, Z>	{x->Z}	{x->Z, y->Z}

```

type Context
  val fn : Function
  val input :  $\sigma$ 

type Summary
  val input :  $\sigma$ 
  val output :  $\sigma$ 

```

```
val results : Map[Context, Summary]
```

```

function ANALYZE( $ctx, \sigma_{in}$ )
   $\sigma'_{out} \leftarrow \text{INTRAPROCEDURAL}(ctx, \sigma_{in})$ 
  results[ $ctx$ ]  $\leftarrow \text{Summary}(\sigma_{in}, \sigma'_{out})$ 
  return  $\sigma'_{out}$ 
end function

```

```

function FLOW([n:  $x := f(y)$ ],  $ctx, \sigma_n$ )
   $\sigma_{in} \leftarrow [\text{formal}(f) \mapsto \sigma_n(y)]$ 
  calleeCtx  $\leftarrow \text{GETCTX}(f, ctx, n, \sigma_{in})$ 
   $\sigma_{out} \leftarrow \text{RESULTSFOR}(calleeCtx, \sigma_{in})$ 
  return  $\sigma_n[x \mapsto \sigma_{out}[result]]$ 
end function

```

Context	σ_{in}	σ_{out}
<main, T>	T	{w->Z, Z->Z}
<double, N>	{x->N}	{x->N, y->N}
<double, Z>	{x->Z}	{x->Z, y->Z}

Works for non-recursive contexts!

```

function GETCTX( $f, callingCtx, n, \sigma_{in}$ )
  return Context( $f, \sigma_{in}$ )
end function

```

```

function RESULTSFOR( $ctx, \sigma_{in}$ )
  if  $ctx \in \text{dom}(\text{results})$  then
    if  $\sigma_{in} \sqsubseteq \text{results}[ctx].input$  then
      return  $\text{results}[ctx].output$ 
    else
      return ANALYZE( $ctx, \text{results}[ctx].input \sqcup \sigma_{in}$ )
    end if
  else
    return ANALYZE( $ctx, \sigma_{in}$ )
  end if
end function

```

Recursion makes this a bit harder

```
int fact(int x) {  
    if (x == 1)  
        return 1;  
    else  
        return x * fact(x-1);  
}  
  
void main() {  
    int y = fact(2);  
    int z = fact(3);  
    int w = fact(getInputFromUser());  
}
```

Key Idea: Worklist of Contexts

val *worklist* : *Set[Context]*

val *analyzing* : *Stack[Context]*

val *results* : *Map[Context, Summary]*

val *callers* : *Map[Context, Set[Context]]*

Key Idea: Worklist of Contexts

```
val worklist : Set[Context]
val analyzing : Stack[Context]
val results : Map[Context, Summary]
val callers : Map[Context, Set[Context]]
```

```
function ANALYZEPROGRAM
    worklist  $\leftarrow \{Context(\text{main}, \top)\}$ 
    results[Context(main,  $\top$ )].input  $\leftarrow \top$ 
    while NOTEMPTY(worklist) do
        ctx  $\leftarrow$  REMOVE(worklist)
        ANALYZE(ctx, results[ctx].input)
    end while
end function
```

Key Idea: Worklist of Contexts

```
val worklist : Set[Context]
val analyzing : Stack[Context]
val results : Map[Context, Summary]
val callers : Map[Context, Set[Context]]
```

```
function ANALYZEPROGRAM
  worklist ← {Context(main, ⊤)}
  results[Context(main, ⊤)].input ← ⊤
  while NOTEMPTY(worklist) do
    ctx ← REMOVE(worklist)
    ANALYZE(ctx, results[ctx].input)
  end while
end function
```

```
function ANALYZE(ctx, σin)
  σout ← results[ctx].output
  PUSH(analyzing, ctx)
  σ'out ← INTRAPROCEDURAL(ctx, σin)
  POP(analyzing)
  if σ'out ≉ σout then
    results[ctx] ← Summary(σin, σout ∪ σ'out)
    for c ∈ callers[ctx] do
      ADD(worklist, c)
    end for
  end if
  return σ'out
end function
```

Key Idea: Worklist of Contexts

```
val worklist : Set[Context]
val analyzing : Stack[Context]
val results : Map[Context, Summary]
val callers : Map[Context, Set[Context]]
```

```
function FLOW(⟦n: x := f(y)⟧, ctx,  $\sigma_n$ )
   $\sigma_{in} \leftarrow [formal(f) \mapsto \sigma_n(y)]$ 
  calleeCtx  $\leftarrow \text{GETCTX}(f, ctx, n, \sigma_{in})$ 
   $\sigma_{out} \leftarrow \text{RESULTSFOR}(calleeCtx, \sigma_{in})$ 
  ADD(callers[calleeCtx], ctx)
  return  $\sigma_n[x \mapsto \sigma_{out}[result]]$ 
```

```
function ANALYZE(ctx,  $\sigma_{in}$ )
   $\sigma_{out} \leftarrow \text{results}[ctx].output$ 
  PUSH(analyzing, ctx)
   $\sigma'_{out} \leftarrow \text{INTRAPROCEDURAL}(ctx, \sigma_{in})$ 
  POP(analyzing)
  if  $\sigma'_{out} \not\sqsubseteq \sigma_{out}$  then
    results[ctx]  $\leftarrow \text{Summary}(\sigma_{in}, \sigma_{out} \sqcup \sigma'_{out})$ 
    for c  $\in$  callers[ctx] do
      ADD(worklist, c)
    end for
  end if
  return  $\sigma'_{out}$ 
end function
```

Key Idea: Worklist of Contexts

```
if  $ctx \in analyzing$  then  
    return  $\perp$ 
```

```
function FLOW([[n:  $x := f(y)$ ]], ctx,  $\sigma_n$ )  
     $\sigma_{in} \leftarrow [formal(f) \mapsto \sigma_n(y)]$   
    calleeCtx  $\leftarrow GETCTX(f, ctx, n, \sigma_{in})$   
     $\sigma_{out} \leftarrow RESULTSFOR(calleeCtx, \sigma_{in})$   
    ADD(callers[calleeCtx], ctx)  
    return  $\sigma_n[x \mapsto \sigma_{out}[result]]$ 
```

```
function ANALYZE(ctx,  $\sigma_{in}$ )  
     $\sigma_{out} \leftarrow results[ctx].output$   
    PUSH(analyzing, ctx)  
     $\sigma'_{out} \leftarrow INTRAPROCEDURAL(ctx, \sigma_{in})$   
    POP(analyzing)  
    if  $\sigma'_{out} \not\models \sigma_{out}$  then  
        results[ctx]  $\leftarrow Summary(\sigma_{in}, \sigma_{out} \sqcup \sigma'_{out})$   
        for c  $\in callers[ctx]$  do  
            ADD(worklist, c)  
        end for  
    end if  
    return  $\sigma'_{out}$   
end function
```

On Precision: Why return \perp when analyzing?

Exercise: Try running constant propagation on this program

```
int iterativeIdentity(x, y)
    if x <= 0
        return y
    else
        iterativeIdentity(x-1, y)

void main(z)
    w = iterativeIdentity(z, 5)
```

On Termination and Complexity

- Add to worklist $C \times H$ times ($C = \# \text{contexts}$, $H = \text{lattice height}$)
- After each analysis, propagate result to N callers
- $O(C \times N \times H)$ intraprocedural analyses
- $= O(E \times H)$ where E is $\#\text{edges}$ in context-sensitive call graph
- Is C finite???

Types of Context-Sensitivity

- No context sensitivity
- Call strings
- Value contexts
- Limited call strings
- Limited value contexts

Limited Context-Sensitivity

No context-sensitivity

```
type Context
  val fn : Function

function GETCTX(f, callingCtx, n, σin)
  return Context(f)
end function
```

Value-based context-sensitivity

```
function GETCTX(f, callingCtx, n, σin)
  return Context(f, σin)
end function
```

K-call-string context-sensitivity

```
type Context
  val fn : Function
  val string : List[Int]
```

```
function GETCTX(f, callingCtx, n, σin)
  newStr ← SUFFIX(callingCtx.string ++ n, CALL_STRING_CUTOFF)
  return Context(f, newStr)
end function
```