

# Interprocedural Analysis and Optimization



Mod/Ref Analysis  
Alias Analysis  
Constant Propagation  
Procedure Inlining and Cloning

# Introduction

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- Interprocedural Analysis
  - Gathering information about the whole program instead of a single procedure
    - Examples: side-effect analysis, alias analysis
- Interprocedural Optimization
  - Modifying more than one procedure, or
  - Using interprocedural analysis

# Interprocedural Side-effect Analysis

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## □ Modification and Reference Side-effect

- MOD(s): set of variables that may be modified as a side effect of call at s
- REF(s): set of variables that may be referenced as a side effect of call at s

```
COMMON X, Y
```

```
...
```

```
DO I = 1, N
```

```
S0: CALL P
```

```
S1: X(I) = X(I) + Y(I)
```

```
ENDDO
```

## □ Can vectorize if

- P neither modifies nor uses X
- P does not modify Y

# Interprocedural Alias Analysis

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```
SUBROUTINE S(A,X,N)
  COMMON Y
  DO I = 1, N
S0:   X = X + Y*A(I)
  ENDDO
END
```

- ❑ Could we keep X and Y in different registers?
  - What happens if S is called with parameters S(A,Y,N)?
    - ❑ Y is aliased to X on entry to S (Fortran uses call-by-ref)
    - ❑ Can't put X and Y in different registers
- ❑ For each parameter x, compute ALIAS(p,x)
  - The set of variables that may refer to the same location as formal parameter x on entry to p

# Call Graph construction

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- Interprocedural analysis must model how procedures call each other
  - Two approaches: call graph and interprocedural control flow graph
- Call Graph:  $G=(N,E)$  model call relations between procedures
  - N: one vertex for each procedure
  - E:  $p \rightarrow q$ : if procedure  $p$  calls  $q$ ; one edge for each possible call
- Construction must handle function pointers (procedure parameters)

```
SUBROUTINE S(X,P)
```

```
S0:   CALL P(X)
```

```
      RETURN
```

```
END
```

- P is a procedure parameter to S
  - What values can P have on entry to S?
  - CALL(s): set of all procedures that may be invoked at s (alias analysis)

# Flow Insensitive Side-effect Analysis

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- Goal: compute what variables may be modified by each procedure
  - Interprocedural analysis
- Assumptions
  - Procedure definitions are not nested inside one another
  - All parameters passed by reference
  - Each procedure has a constant number of parameters
  - Procedures may recursively invoke each other
- We will formulate and solve the MOD(s) problem

# Solving MOD

- MOD(s): variables modified by the call of procedure p at call site s

$$MOD(s) = DMOD(s) \cup \bigcup_{x \in DMOD(s)} ALIAS(p, x)$$

- DMOD(s): set of variables directly modified as side-effect of call at s

$$DMOD(s) = \{v \mid s \Rightarrow p, v \xrightarrow{s} w, w \in GMOD(p)\}$$

- GMOD(p): set of global variables and formal parameters of p that are modified, either directly or indirectly as a result of calling p

```
S0: CALL P(A,B,C)
SUBROUTINE P(X,Y,Z)
    INTEGER X,Y,Z
    X = X*Z
    Y = Y*Z
END
```

**GMOD(P) = {X, Y}**

**DMOD(S0) = {A, B}**

# Solving GMOD

---

- GMOD(p) contains two types of variables
  - IMOD(p): variables explicitly modified in body of P
  - Variables modified as a side-effect of some procedure invoked in p
    - Global variables are viewed as parameters to a called procedure

$$GMOD(p) = IMOD(p) \cup \bigcup_{s=(p,q)} \{z \mid z \xrightarrow{s} w, w \in GMOD(q)\}$$

- May take a long time to converge due to recursive procedure calls



# Solving GMOD

- Decompose  $GMOD(p)$  differently to get an efficient solution
- Key: Treat side-effects to global variables and reference formal parameters separately

$$GMOD(p) = \boxed{IMOD^+(p)} \cup \boxed{\bigcup_{s=(p,q)} GMOD(q) \cap \neg LOCAL}$$

- where

$$IMOD^+(p) = IMOD(p) \cup \bigcup_{s=(p,q)} \{z \mid z \xrightarrow{s} w, w \in RMOD(q)\}$$

- **$RMOD(p)$** : set of formal parameters that may be modified in  $p$ , either directly or by used as actual parameter to call another procedure  $q$

# Alias Analysis

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- Recall definition of MOD(s)

$$MOD(s) = DMOD(s) \cup \bigcup_{x \in DMOD(s)} ALIAS(p, x)$$

- Need to

- Compute ALIAS(p,x)
- Update DMOD to MOD using ALIAS(p,x)

- Key Observations

- Two global variables can never be aliased of each other.
- Global variables can only be aliased to formal parameters
- The number of aliases for each variable is bounded by the number of formal parameters and global variables
  - Not true in C/C++ code when data can be dynamically allocated

# Update DMOD to MOD

---

```
SUBROUTINE P
  INTEGER A
S0:   CALL S(A,A)
END
```

```
SUBROUTINE S(X,Y)
  INTEGER X,Y
S1:   CALL Q(X)
END
```

```
SUBROUTINE Q(Z)
  INTEGER Z
  Z = 0
END
```

$\text{GMOD}(Q)=\{Z\}$

$\text{DMOD}(S1)=\{X\}$

$\text{MOD}(S1)=\{X,Y\}$

# Interprocedural Optimizations

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- ❑ The goal of interprocedural analysis is to enable whole program optimizations
- ❑ Can we understand procedural calls just like regular statements?
  - MOD/REF -- set of variables modified/referenced in procedure
  - ALIAS -- set of aliased variables in a procedure.
- ❑ Eliminating the boundary of procedures
  - Procedure inlining and cloning(specialization)
- ❑ Can enhance the scope of many optimizations
  - Constant propagation
  - Redundancy elimination
  - Loop optimizations

# Procedure Inlining

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- Replace a procedure invocation with the body of the procedure being called
  - Advantages:
    - Eliminates procedure call overhead.
    - Allows more optimizations to take place
  - However, overuse can cause slowdowns
    - Breaks compiler procedure assumptions.
    - Function calls add needed register spills.
    - Changing function forces global recompilation.

# Procedure Cloning

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- Often specific values of function parameters result in better optimizations.

```
PROCEDURE UPDATE(A,N,IS)
  REAL A(N)
  INTEGER I = 1,N
    A(I*IS-IS+1)=A(I*IS-IS+1)+PI
  ENDDO
END
```

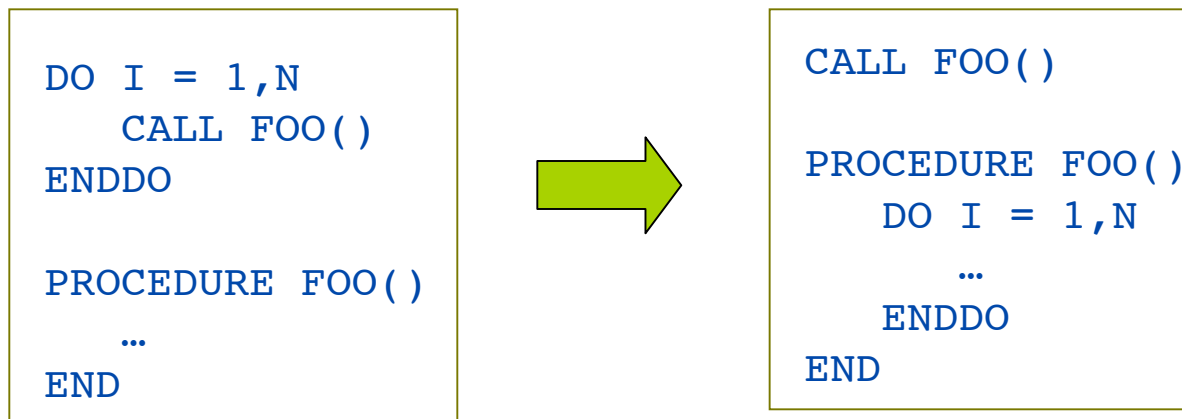
If we knew that  $IS \neq 0$  at a call, then loop can be vectorized.

**If we know that  $IS \neq 0$  at specific call sites, clone a vectorized version of the procedure and use it at those sites.**

# Hybrid optimizations

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- ❑ Combinations of procedures can have benefit.
- ❑ One example is loop embedding:



# Constant Propagation

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- Propagating constants between procedures can significantly improve performance
- Dependence testing can be made more precise

```
SUBROUTINE FOO(N)
  INTEGER N,M
  CALL INIT(M,N)
  DO I = 1,P
    B(M*I + 1) = 2*B(1)
  ENDDO
END
```

```
SUBROUTINE INIT(M,N)
  M = N
END
```

Enable more accurate dependence analysis if  
N is a constant

- Challenge: need to model data-flow across procedural boundaries



# Constant Propagation

- **Definition:** Let  $s = (p,q)$  be a call site, and let  $x$  be a parameter of  $q$ . The **jump function**  $J_s^x$ 
  - Gives the value of formal parameter  $x$  used to invoke  $q$  in terms of incoming parameter values of procedure  $p$
  - Models a transfer function for each call site
    - caller parameters  $\implies$  callee parameters
- We construct an interprocedural value graph:
  - Add a node to the graph for each jump function  $J_s^x$
  - If  $x$  is used to compute to  $J_{t'}^y$ , where  $t'$  is a call site in procedure  $q$ , then add an edge between  $J_s^x$  and  $J_{t'}^y$  for every call site  $s = (p,q)$  in some procedure  $p$
  - Model control flow (call relations) between jump functions
- Apply the constant propagation algorithm to this graph.
  - Might want to iterate with global propagation

# Jump Functions

```
PROGRAM MAIN
  INTEGER A
  α CALL PROCESS(15,A)
  PRINT A
END
SUBROUTINE PROCESS(N,B)
  INTEGER N,B,I
  β CALL INIT(I,N)
  γ CALL SOLVE(B,I)
END
SUBROUTINE INIT(X,Y)
  INTEGER X,Y
  X = 2*Y
END
SUBROUTINE SOLVE(C,T)
  INTEGER C,T
  C = T*10
END
```

- Need a way of building  $J_\gamma^I$
- For parameter  $x$  of procedure  $p$ , define  $R_p^x$  to be the output value of  $x$  in terms of input parameters of  $p$

$$R_{INIT}^X = \{2 * Y\} \quad R_{SOLVE}^C = \{T * 10\}$$

$$R_{PROCESS}^B = \begin{cases} R_{SOLVE}^C(J_\gamma^T(N)) & C \in MOD(\gamma) \\ \text{undefined} & \text{otherwise} \end{cases}$$

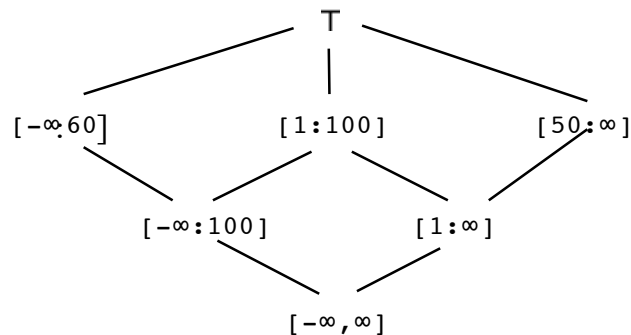
$$J_\gamma^T = \begin{cases} R_{INIT}^X(N) & I \in MOD(\beta) \\ \text{undefined} & \text{otherwise} \end{cases}$$

$$J_\alpha^N = 15 \quad J_\beta^Y = N$$

# Symbolic Analysis

- Prove facts about values of variables
  - Find a symbolic expression for a variable in terms of other variables.
  - Establish a relationship between pairs of variables at some point in program.
  - Establish a range of values for a variable at a given point.

## Range Analysis:



- Jump functions and return jump functions return ranges.
- Meet operation is now more complicated.
- If we can bound number of times upper bound increases and lower bound decreases, the finite-descending-chain property is satisfied.

# Array Section Analysis

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- Consider the following code:

```
DO I = 1,N  
  CALL SOURCE(A,I)  
  CALL SINK(A,I)  
ENDDO
```

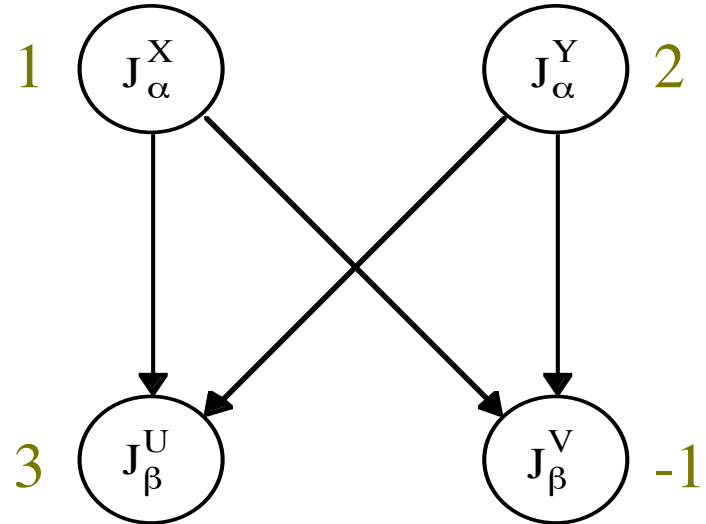
Does this loop carry dependence?

Let  $M_A(I)$  be the set of locations in array modified on iteration  $I$  and  $U_A(I)$  set of locations used on iteration  $I$ . Then has a carried true dependence iff

$$M_A(I_1) \cap U_A(I_2) \neq \emptyset \quad 1 \leq I_1 < I_2 \leq N$$

# Example

```
PROGRAM MAIN
  INTEGER A,B
  A = 1
  B = 2
 $\alpha$  CALL S(A,B)
END
SUBROUTINE S(X,Y)
  INTEGER X,Y,Z,W
  Z = X + Y
  W = X - Y
 $\beta$  CALL T(Z,W)
END
SUBROUTINE T(U,V)
  PRINT U,V
END
```



The constant-propagation algorithm will  
Eventually converge to above values.

# Whole Program Optimization

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- What we have covered
  - Call graph construction
  - Mod/ref analysis
  - Alias analysis
  - Constant propagation
  - Procedure inlining and cloning
- Practical concerns
  - Requires the source code of multiple procedures (whole program)
  - Requires recompilation of interdependent procedures when program is modified