

# CS 610: Data Dependence Analysis

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<https://www.iitk.ac.in/doaa/data/FAQ-2020-21-I.pdf>

# How to Write Efficient and Scalable Programs?

## Good choice of algorithms and data structures

- Determines number of operations executed

## Code that the compiler and architecture can effectively optimize

- Determines number of instructions executed

## Proportion of parallelizable and concurrent code

- Amdahl's law

## Sensitive to the architecture platform

- Efficiency and characteristics of the platform
- For e.g., memory hierarchy, cache sizes



# Role of a Good Compiler

Try and extract performance automatically

Optimize memory access latency

- Code restructuring optimizations
- Prefetching optimizations
- Data layout optimizations
- Code layout optimizations



# Parallelism Challenges for a Compiler

- On single-core machines
  - Focus is on register allocation, instruction scheduling, reduce the cost of array accesses
- On parallel machines
  - Find parallelism in sequential code, find portions of work that can be executed in parallel
  - Principle strategy is data decomposition – good idea since this can scale



# Can we parallelize the following loops?

```
do i = 1, 100  
  A(i) = A(i) + 1  
enddo
```

```
do i = 1, 100  
  A(i) = A(i-1) + 1  
enddo
```



# Dependences

S1  $a = b + c$

S2  $d = a * 2$

S3  $a = c + 2$

S4  $e = d + c + 2$



# Dependences

S1  $a = b + c$

S2  $d = a * 2$

S3  $a = c + 2$

S4  $e = d + c + 2$

## Execution constraints

- S2 must execute after S1
- S3 must execute after S2
- S3 must execute after S1
- S3 and S4 can execute in any order, and concurrently





# Data Dependence

- There is a data dependence from  $S1$  to  $S2$  if and only if
  - Both statements access the same memory location
  - At least one of the accesses is a write
  - There is a feasible execution path at run time from  $S1$  to  $S2$



# Types of Dependences

Flow (true)

Anti

Output

Input

S1  $X = \dots$   
S2  $\dots = X$

$S1 \delta S2$

S1  $\dots = X$   
S2  $X = \dots$

$S1 \delta^{-1} S2$

S1  $X = \dots$   
S2  $X = \dots$

$S1 \delta^0 S2$

S1  $\dots = a/b$   
S2  $\dots = b * c$



# Bernstein's Conditions

- Suppose there are two processes  $P_1$  and  $P_2$
- Let  $I_i$  be the set of all input variables for process  $P_i$
- Let  $O_i$  be the set of all output variables for process  $P_i$
- $P_1$  and  $P_2$  can execute in parallel (denoted as  $P_1 \parallel P_2$ ) if and only if
  - $I_1 \cap I_2 = \Phi$
  - $I_2 \cap O_1 = \Phi$
  - $O_2 \cap O_1 = \Phi$



# Bernstein's Conditions

- Suppose there are two processes  $P_1$  and  $P_2$  that can execute in parallel only
  - Let  $I_i$  be the set of all input variables for process  $P_i$
  - Let  $O_i$  be the set of all output variables for process  $P_i$
- Two processes can execute in parallel if they are flow-, anti-, and output-independent
- $O_2 \cap O_1 = \Phi$



# Bernstein's Conditions

- Suppose there are two processes  $P_1$  and  $P_2$  can execute in parallel only
- Let  $I_i$  be the set of all input variables for process  $P_i$
- Let  $O_i$  be the set of all output variables for process  $P_i$

Two processes can execute in parallel if they are flow-, anti-, and output-independent

$$O_2 \cap O_1 = \Phi$$

- If  $P_i \parallel P_j$ , does that imply  $P_j \parallel P_i$ ?
- If  $P_i \parallel P_j$  and  $P_j \parallel P_k$ , does that imply  $P_i \parallel P_k$ ?



# Find Parallelism in Loops – Is it Easy?

- Need to analyze array subscripts
- Need to check whether two array subscripts access the same memory location



# Dependence in Loops

```
    for i = 1 to 50
S1     A[i] = B[i-1] + C[i]
S2     B[i] = A[i+2] + C[i]
    endfor
```

- Unrolling loops can help figure out dependences

S1(1)      A[1] = B[0] + C[1]

S2(1)      B[1] = A[3] + C[1]

S1(2)      A[2] = B[1] + C[2]

S2(2)      B[2] = A[4] + C[2]

S1(3)      A[3] = B[2] + C[3]

S2(3)      B[3] = A[5] + C[3]

.....



# Dependence in Loops

```
    for i = 1 to 50
S1     A[i] = B[i-1] + C[i]
S2     B[i] = A[i+2] + C[i]
    endfor
```

- large loop bounds
- loop bounds may not be known at compile time

- Unrolling loops can help figure out dependences

```
S1(1)   A[1] = B[0] + C[1]
S2(1)   B[1] = A[3] + C[1]
S1(2)   A[2] = B[1] + C[2]
S2(2)   B[2] = A[4] + C[2]
S1(3)   A[3] = B[2] + C[3]
S2(3)   B[3] = A[5] + C[3]
```





# Dependence in Loops

- Parameterize the statement with the loop iteration number

```
DO I = 1, N
S1  A(I+1) = A(I) + B(I)
ENDDO
```

```
DO I = L, U, S
S1  ...
ENDDO
```



# Normalized Iteration Number

For an arbitrary loop in which the loop index  $I$  runs from  $L$  to  $U$  in steps of  $S$ , the *normalized iteration number*  $i$  of a specific iteration is equal to the value  $(I - L + 1) / S$ , where  $I$  is the value of the index on that iteration



# Iteration Vector

Given a nest of  $n$  loops, the *iteration vector*  $i$  of a particular iteration of the innermost loop is a vector of integers that contains the iteration numbers for each of the loops in order of nesting level.

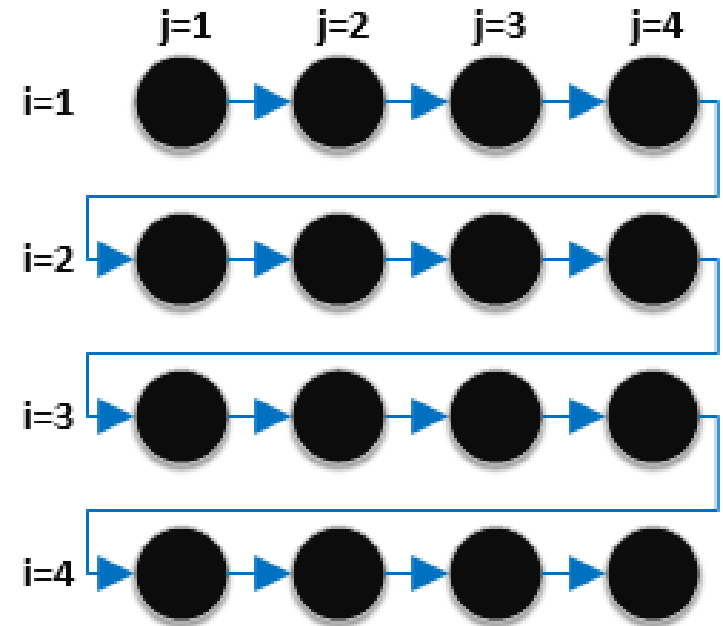
The iteration vector  $i$  is  $\{i_1, i_2, \dots, i_n\}$  where  $i_k$ ,  $1 \leq k \leq n$ , represents the iteration number for the loop at nesting level  $k$ .



# Iteration Space Graphs

- Represent each dynamic instance of a loop as a point in the graph
- Draw arrows from one point to another to represent dependences

```
S1:   for (i = 1; i <= 4; i++)  
      for (j = 1; j <= 4; j++)  
        a[i][j] = a[i][j-1] * x;
```



# Iteration Space Graph

- Dimension of iteration space is the loop nest level
- Not restricted to be rectangular

```
for i = 1 to 5 do
  for j = i to 5 do
    A(i, j) = B(i, j) + C(j)
  endfor
endfor
```



# Lexicographic Ordering of Iteration Vectors

- Assume  $i$  is a vector,  $i_k$  is the  $k^{\text{th}}$  element of the vector  $i$ , and  $i[1:k]$  is a  $k$ -vector consisting of the leftmost  $k$  elements of  $i$
- Iteration  $i$  precedes iteration  $j$ , denoted by  $i < j$ , if and only if
  - i.  $i[1:n-1] < j[1:n-1]$ , or
  - ii.  $i[1:n-1] = j[1:n-1]$  and  $i_n < j_n$



# Formal Definition of Loop Dependence

There exists a dependence from statement  $S1$  to statement  $S2$  in a common nest of loops if and only if there exist two iteration vectors  $i$  and  $j$  for the nest, such that

- i.  $i < j$  or  $i = j$  and there is a path from  $S1$  to  $S2$  in the body of the loop,
- ii. statement  $S1$  accesses memory location  $M$  on iteration  $i$  and statement  $S2$  accesses location  $M$  on iteration  $j$ , and
- iii. one of these accesses is a write.



# Distance Vectors

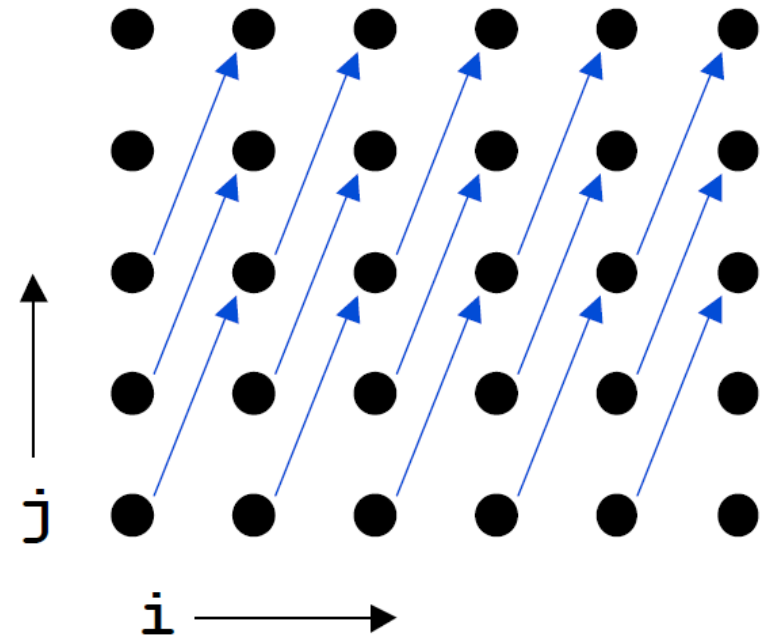
- For each dimension of an iteration space, the distance is the number of iterations between accesses to the same memory location

```
do i = 1, 6
  do j = 1, 5
    A(i, j) = A(i-1, j-2) + 1
  enddo
enddo
```

outer loop

- Distance vector: (1, 2)

inner loop





# Distance Vectors

- Suppose that there is a dependence from statement S1 on iteration  $i$  of a loop nest and statement S2 on iteration  $j$ , then the *dependence distance vector*  $d(i,j)$  is defined as a vector of length  $n$  such that  $\mathbf{d}(i,j)_k = j_k - i_k$ .
- A vector  $(d_1, d_2)$  is positive if  $(0,0) < (d_1, d_2)$ , i.e., its first (leading) non-zero component is positive



# Direction Vectors

- Suppose that there is a dependence from statement S1 on iteration  $i$  of a loop nest of  $n$  loops and statement S2 on iteration  $j$ , then the *dependence direction vector* is  $D(i,j)$  is defined as a vector of length  $n$  such that

$$D(i,j)_k = \begin{cases} - & \text{if } D(i,j)_k < 0 \\ 0 & \text{if } D(i,j)_k = 0 \\ + & \text{if } D(i,j)_k > 0 \end{cases}$$



# Distance and Direction Vectors

- Suppose that there is a dependence from statement S1 on iteration  $i$  of a loop nest of  $n$  loops and statement S2 on iteration  $j$  then the dependence vector is  $D(i, j)$ . In any valid dependence, the leftmost non-“0” component of the direction vector must be “+”

$$D(i, j)_k = \begin{cases} - & \text{if } D(i, j)_k < 0 \\ 0 & \text{if } D(i, j)_k = 0 \\ + & \text{if } D(i, j)_k > 0 \end{cases}$$



# Distance and Direction Vector Example

```
DO I = 1, N
  DO J = 1, M
    DO K = 1, L
S1      A(I+1,J,K-1) = A(I,J,K) + 10
    ENDDO
  ENDDO
ENDDO
```



# Distance and Direction Vector Example

```
FOR I = 1, 5
  DO J = 1, 5
S1    A(I,J) = A(I,J-3) + A(I-2,J) + A(I-1,J+2) + A(I+1,J-1)
  ENDFOR
ENDFOR
```



# Program Transformations and Validity



# Reordering Transformations

- A reordering transformation does not add or remove statements from a loop nest
  - Only reorders the execution of the statements that are already in the loop

Does not add or remove  
statements



Does not add or remove  
any new dependences



# Validity of Dependence-Based Transformations

- A reordering transformation is said to be valid for the program to which it applies if it preserves all dependences in the program





# Direction Vector Transformation

- Let  $T$  be a transformation applied to a loop nest
  - Does not rearrange the statements in the body of the loop
- $T$  is valid if, after it is applied, none of the direction vectors for dependences with source and sink in the nest has a leftmost non-“0” component that is “-”



# Dependence Types

- If in a loop, statement  $S_2$  depends on  $S_1$ , then there are two possible ways of this dependence occurring
  - $S_1$  and  $S_2$  execute on different iterations - this is called a **loop-carried** dependence
  - $S_1$  and  $S_2$  execute on the same iteration - this is called a **loop-independent** dependence
- These types partition all possible data dependences



# Loop-Carried Dependences

- S1 can reference the common location on one iteration of a loop; on a subsequent iteration S2 can reference the same location

```
DO I = 1, N
  S1  A(I+1) = F(I)
  S2  F(I+1) = A(I)
ENDDO
```

- i. S1 references location  $M$  on iteration  $i$
- ii. S2 references  $M$  on iteration  $j$
- iii.  $d(i,j) > 0$  (that is, contains a “+” as leftmost non-“0” component)



# Level of Loop-Carried Dependence

- The *level* of a loop-carried dependence is the index of the leftmost non-“0” of  $D(i,j)$  for the dependence.

```
DO I = 1, 10
  DO J = 1, 10
    DO K = 1, 10
      S1      A(I,J,K+1) = A(I,J,K)
    ENDDO
  ENDDO
ENDDO
```



# Utility of Dependence Levels

- A reordering transformation preserves all level- $k$  dependences if it
  - i. preserves the iteration order of the level- $k$  loop
  - ii. does not interchange any loop at level  $< k$  to a position inside the level- $k$  loop and
  - iii. does not interchange any loop at level  $> k$  to a position outside the level- $k$  loop.



# Utility of Dependence Levels

- A reordering transformation preserves all level- $k$  dependences if it
  - i. preserves the iteration order of the level- $k$  loop
  - ii. does not interchange any loop at level  $< k$  to a position inside the level- $k$  loop and
  - iii. does not interchange any loop at level  $> k$  to a position outside the level- $k$  loop.

```
DO I = 1, 10
S1   A(I+1) = F(I)
S2   F(I+1) = A(I)
ENDDO
```

```
DO I = 1, 10
S2   F(I+1) = A(I)
S1   A(I+1) = F(I)
ENDDO
```



# Is this transformation valid?

```
DO I = 1, 10
  DO J = 1, 10
    DO K = 1, 10
S      A(I+1,J+2,K+3) = A(I,J,K) + B
    ENDDO
  ENDDO
ENDDO
```

```
DO I = 1, 10
  DO K = 10, 1, -1
    DO J = 1, 10
S      A(I+1,J+2,K+3) = A(I,J,K) + B
    ENDDO
  ENDDO
ENDDO
```



# Loop-Independent Dependences

- S1 and S2 can both reference the common location on the same loop iteration, but with S1 preceding S2 during execution of the loop iteration.

- i. S1 refers to memory location  $M$  on iteration  $i$
- ii. S2 refers to  $M$  on iteration  $j$  and  $i = j$
- iii. There is a control flow path from S1 to S2 within the iteration.

```
DO I = 1, N
S1   A(I+1) = F(I)
S2   G(I+1) = A(I+1)
ENDDO
```

```
DO I = 1, 9
S1   A(I) =
S2   ... = A(10-I)
ENDDO
```





# Validity of Transformations for Loop-Independent Dependences

- If there is a loop-independent dependence from  $S1$  to  $S2$ , any reordering transformation that does not move statement instances between iterations and preserves the relative order of  $S1$  and  $S2$  in the loop body preserves that dependence.



# Is this transformation valid?

```
DO I = 1, N
S1:  A(I) = B(I) + C
S2:  D(I) = A(I) + E
ENDDO
```

```
D(1) = A(1) + E
DO I = 2, N
S1:  A(I-1) = B(I-1) + C
S2:  D(I) = A(I) + E
ENDDO
A(N) = B(N) + C
```



# Dependency Testing



# Dependence Testing

- Dependence question
  - Can  $4*I$  be equal to  $2*I+1$  for  $I$  in  $[1, N]$ ?

```
DO I=1, N
  A(4*I) = ...
  ... = A(2*I+1)
ENDDO
```

Given (i) two subscript functions  $f$  and  $g$ , and (ii) lower and upper loop bounds  $L$  and  $U$  respectively, does  $f(i_1) = g(i_2)$  have a solution such that  $L \leq i_1, i_2 \leq U$ ?



# Multiple Loop Nests

```
DO i=1,n
  DO j=1,m
    X(a1*i + b1*j + c1) = ...
    ... = X(a2*i + b2*j + c2)
  ENDDO
ENDDO
```

- Dependence test

$$\begin{aligned} a_1 * i_1 + b_1 * j_1 + c_1 &= a_2 * i_2 + b_2 * j_2 + c_2 \\ 1 \leq i_1, i_2 &\leq n \\ 1 \leq j_1, j_2 &\leq m \end{aligned}$$



# Multiple Loop Indices, Multi-Dimensional Array

```
DO i=1,n
  DO j=1,m
    X(a1*i1 + b1*j1 + c1, d1*i1 + e1*j1 + f1) = ...
    ... = X(a2*i2 + b2*j2 + c2, d2*i2 + e2*j2 + f2)
  ENDDO
ENDDO
```

- Dependence test

$$a_1 i_1 + b_1 j_1 + c_1 = a_2 i_2 + b_2 j_2 + c_2$$

$$d_1 i_1 + e_1 j_1 + f_1 = d_2 i_2 + e_2 j_2 + f_2$$

$$1 \leq i_1, i_2 \leq n$$

$$1 \leq j_1, j_2 \leq m$$



# Multiple Loop Indices, Multi-Dimensional Array

```
DO i=1,n
  DO j=1,m
    X(a1*i1 + b1*j1 + c1, d1*i1 + e1*j1 + f1) = ...
    ... = X(a2*i2 + b2*j2 + c2, d2*i2 + e2*j2 + f2)
  ENDDO
ENDDO
```

- Dependence test

$$\begin{aligned}a_1 i_1 + b_1 j_1 + c_1 &= a_2 i_2 + b_2 j_2 + c_2 \\d_1 i_1 + e_1 j_1 + f_1 &= d_2 i_2 + e_2 j_2 + f_2 \\1 \leq i_1, i_2 &\leq n \\1 \leq j_1, j_2 &\leq m\end{aligned}$$

complex



$$ax + by = c$$

# Data Dependence Testing

- Variables in loop indices are integers  $\rightarrow$  Diophantine equations
- The Diophantine equation  $a_1i_1 + a_2i_2 + \dots + a_ni_n = c$  has an integer solution if and only if  $\text{gcd}(a_1, a_2, \dots, a_n)$  evenly divides  $c$
- If there is a solution, we can test if it lies within the loop bounds
  - If not, then there is no dependence





# Complexity in Dependence Testing

- Subscript: A pair of subscript positions in a pair of array references
  - $A(i, j) = A(i, k) + C$
  - $\langle i, i \rangle$  is the first subscript,  $\langle j, k \rangle$  is the second subscript
- A subscript is said to be
  - Zero index variable (ZIV) if it contains no index
  - Single index variable (SIV) if it contains only one index
  - Multi index variable (MIV) if it contains more than one index
    - $A(5, I+1, j) = A(1, I, k) + C$
    - First subscript is ZIV, second subscript is SIV, third subscript is MIV



# Separability and Couple Subscript Groups

- A subscript is separable if its indices do not occur in other subscripts
- If two different subscripts contain the same index they are coupled
  - $A(I+1, j) = A(k, j) + C$  : Both subscripts are separable
  - $A(I, j, j) = A(I, j, k) + C$  : Second and third subscripts are coupled
- Coupling can cause imprecision in dependence testing

```
DO I = 1, 100
S1   A(I+1, I) = B(I) + C
S2   D(I) = A(I, I) * E
ENDDO
```



# Overview of Dependency Testing

1. Partition subscripts of a pair of array references into separable and coupled groups
2. Classify each subscript as ZIV, SIV or MIV
3. For each separable subscript apply single subscript test
  - If not done, goto next step
4. For each coupled group apply multiple subscript test like Delta Test
5. If still not done, merge all direction vectors computed in the previous steps into a single set of direction vectors



# Simple Dependence Testing: Delta Notation

- Notation represents index values at the source and sink

```
      DO I = 1, N
S      A(I + 1) = A(I) + B
      ENDDO
```

- Let source iteration be denoted by  $I_0$ , and sink iteration be denoted by  $I_0 + \Delta I$
- Valid dependence implies  $I_0 + 1 = I_0 + \Delta I$
- We get  $\Delta I = 1 \Rightarrow$  Loop-carried dependence with distance vector (1) and direction vector (+)



# Simple Dependence Testing: Delta Notation

```
DO I = 1, 100
  DO J = 1, 100
    DO K = 1, 100
      A(I+1,J,K) = A(I,J,K+1) + B
    ENDDO
  ENDDO
ENDDO
```

- $I_0 + 1 = I_0 + \Delta I$ ;      $J_0 = J_0 + \Delta J$ ;      $K_0 = K_0 + \Delta K + 1$
- Solutions:  $\Delta I = 1$ ;    $\Delta J = 0$ ;    $\Delta K = -1$
- Corresponding direction vector:  $(+,0,-)$



# Simple Dependence Testing: Delta Notation

- If a loop index does not appear, its distance is unconstrained and its direction is "\*"

```
DO I = 1, 100
  DO J = 1, 100
    A(I+1) = A(I) + B(J)
  ENDDO
ENDDO
```

flow W → R  
 $I_0 + 1 = I_0 + \Delta I$   
 $\Rightarrow \Delta I = 1$   
 $(+, *) \rightarrow i$   
 $(1, 0)$   
 $(1, 1)$   
 $(1, 2)$

- The direction vector for the dependence is (+, \*)



# Simple Dependence Testing: Delta Notation

- \* denotes union of all 3 directions

```
DO I = 1, 100
    DO J = 1, 100
        A(I+1) = A(I) + B(J)
    ENDDO
ENDDO
```

- (\*, +) denotes { (+, +), (0, +), (-, +) }
  - (-, +) denotes a level 1 anti-dependence with direction vector (+, -)



# Other Dependence Tests

- GCD test is simple but not accurate

- It can tell us that there is no solution

- Other tests

- Banerjee-Wolfe test: widely used test
- Power Test: improvement over Banerjee test
- Omega test: “precise” test, most accurate for linear subscripts
- Range test: handles non-linear and symbolic subscripts
- many variants of these tests

$$2x + 2y = 4$$

$$\Rightarrow x + y = 2 \quad \text{for } i = 1 \text{ to } 10$$

$$S1 \quad a[i] = b[i] + c[i]$$

$$S2 \quad d[i] = a[i-100];$$

$$x[a * i + b]$$

$$x[c * i + d]$$

$$\text{GCD}(c, a) \quad (d - b)$$

$$x[2 * i + 3] = x[2 * i] + 5;$$

$$-3 \mid \text{gcd}(2, 2)$$





# Dependence Testing is Hard

Unknown loop bounds can lead to false dependences

Need to be conservative about aliasing

Triangular loops add new constraints



# Why is Dependence Analysis Important?

- Dependence information can be used to drive other important loop transformations
  - For example, loop parallelization, loop interchange, loop fusion
- We will see many examples soon



# References

- R. Allen and K. Kennedy – Optimizing Compilers for Modern Architectures, Chapter 2.
- Michelle Strout – CS 553: Compiler Construction, Fall 2007.