

Dependence Analysis

(CS 3006)

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The Big Picture

1. What are our goals?

Simple Goal: Make execution time as small as possible

2. Which leads to:

- Achieve execution of many (all, in the best case)
 instructions in parallel
- But, you have to find <u>INDEPENDENT</u> instructions



Data Dependence

 Data must be produced and consumed in the correct order

Simple example of data dependence:

```
S_1 PI = 3.14

S_2 R = 5.0

S_3 AREA = PI * R ** 2
```

 Statement 53 cannot be moved before either 51 or 52 without compromising correct results



Motivation

DOALL loops: loops whose iterations can execute in parallel

```
for i = 11, 20
a[i] = a[i] + 3
```



Examples

for
$$i = 11, 20$$

 $a[i] = a[i] + 3$

Parallel

```
for i = 11, 20
a[i] = a[i-1] + 3
```

NOT Parallel?

```
for i = 11, 20
 a[i] = a[i-10] + 3
```

Parallel?



Dependence Analysis

- A dependence is a relationship between 2 computations that places constraints on their execution order
- Dependence analysis identifies these constraints
- Constraints are used to determine whether a particular transformation can be applied without changing the computation's semantics
- 2 types of dependences: control and data dependences
- **Both of them must be considered** when parallelizing programs.



Control Dependence

- There is a control dependence between \$1 and \$2, when \$1 determines whether \$2 will be executed or not
- Example:

```
S_1 IF (T .NE. 0.0)

S_2 A = A / T

S_3 CONTINUE
```

- Executing S2 before S1 could cause a divide by zero exception (in this example).
- S2 is conditional upon the execution of the branch in S1.



Data Dependence

- Two statements have a data dependence if they cannot be executed simultaneously due to conflicting uses of the same data.
- Ensure that data is produced and consumed in the right order:
 - 1. do not interchange loads and stores to the same location
 - 2. two stores take place in the correct order
- Formally:
 - There is a data dependence from statement S1 to statement S2 (S2 depends on S1) if:
 - Both statements access the <u>same memory location</u> and at least <u>one of them stores</u> onto it, and
 - There is a feasible run-time execution path from S1 to S2

Load/Store Classification

- Dependences classified in terms of load-store order:
 - 1. True dependences
 - S₂ depends on S₁ is denoted by S₁ δ S₂

$$S_1 X=...$$

 $S_2 ... = X$

This is a crucial dependence!

- 2. Antidependence
 - S₂ depends on S₁ is denoted by S₁ δ⁻¹ S₂

$$S_1 \dots = X$$

 $S_2 \quad X = \dots$

- 3. Output dependence
 - S₂ depends on S₁ is denoted by S₁ δ⁰ S₂

$$S_1 X = ...$$

 $S_2 X = ...$

Data Dependence of Scalar Variables

True/Flow dependence

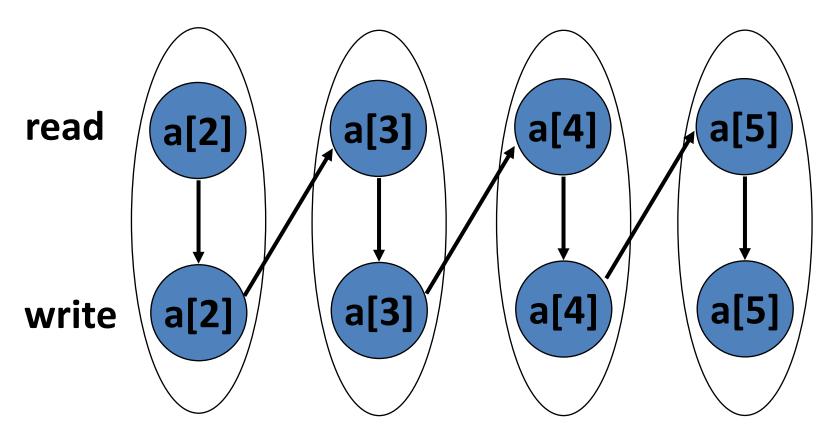
Anti-dependence

Input dependence

- Only data flow dependences are true dependences.
- Anti and output can be removed by renaming

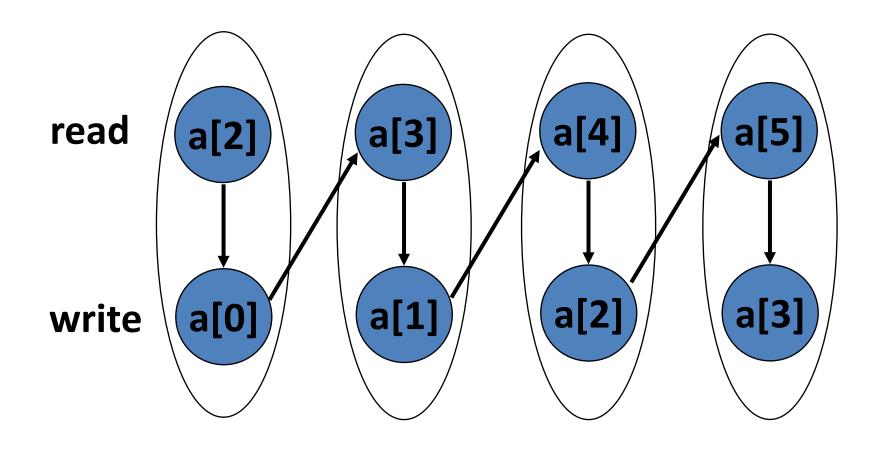


Array Accesses in a Loop



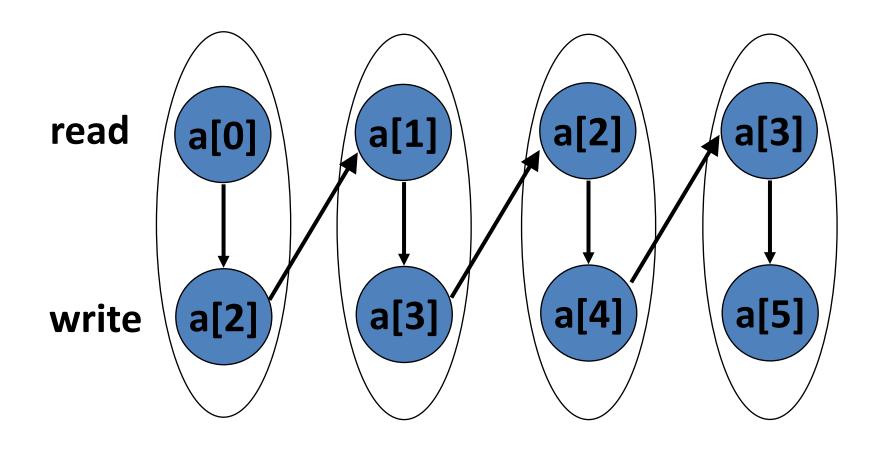


Array Anti-dependence





Array True-dependence





A Parallel DOALL Loop

 A loop is fully parallel if <u>no dependencies flow across</u> iterations:

DO
$$I = 2$$
, N
 $A(I) = A(I) + 1$
ENDDO

DO I = 2, N

$$A(I) = A(I-1) + 1$$
ENDDO

Parallel loops are found through dependence analysis and dependence tests

Usually done at the source-code level, focus is on arrays



Recognizing DOALL Loops

Find data dependences in loop

Definition: a dependence is loop-carried if it crosses an iteration boundary

 If there are no loop-carried dependences only then loop is parallelizable



Example: Loop Parallelization

Which of the following loops are parallelizable?



Dependence in Loops

```
1: A(2) = A(1) + B(1)

2: A(3) = A(2) + B(2)

2: A(4) = A(2) + B(2)

3: A(4) = A(3) + B(3)

3: A(5) = A(4) + B(3)

4: A(6) = A(4) + B(4)
```

Let us look at two different loops:

```
DO I = 1, N

S_1 = A(I+2) = A(I) + B(I)

ENDDO
```

- In both cases, statement S₁ depends on itself
- However, there is a significant difference.
- We need a formalism to describe and distinguish such dependences



Iteration Numbers

 The <u>iteration number</u> of a <u>loop</u> is <u>equal</u> to the <u>value</u> of the <u>loop index</u>

Definition:

For an arbitrary loop in which the loop index I runs from L to U in steps of S, the iteration number i of a specific iteration is equal to the index value I on that iteration:

Example:

DO I = 0, 10, 2

$$S_1$$

ENDDO

end normalizeLoop;

Algorithm: Normalizing Loops

Procedure $normalizeLoop(L_0)$; i = a unique compiler-generated S1: replace the loop header for L_0 DO I = L, U, Swith the adjusted loop header DO i = 1, (U - L + S) / SS2: replace each reference to I within the loop by i * S - S + L: S3: insert a finalization assignment I = i * S - S + L;immediately after the end of the loop;

Normalizing Loops: Examples

Example: before loop normalization

DO I = 3, 11, 2

$$S_1$$
 $A(I) = A(I+1)+10$
ENDDO

Example: after loop normalization

DO In = 1, 5

$$S_1$$
 A(2*In+1) = A(2*In+2) + 10
ENDDO
I=In*2-2+3

Normalizing Loops: Examples

normalize so that the loop

- starts at 1 and
- have stride 1

```
L: do I = 1000, 1, -1

A(I) = ...

end do
```

Normalizing Loops: Examples

normalize so that the loop

- starts at 1 and
- have stride 1

```
L: do I = 1000, 1, -1

A(I) = ...

end do
```

$\overline{\mathbf{Q}}$

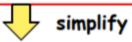
normalize

```
Lnorm: do $I = 1, ((1-1000+(-1)/(-1)

A(1000 + (\$I-1)*(-1)=...

end do

I = 1000 + MAX((1-1000+(-1))/(-1)), 0)*(-1)
```





Normalized Iteration Space

```
DO I = 100, 20, -10
A(I) = B(100-I) + C(I/5)
ENDDO
```



```
DO i = 1, 9

A(110-10*i) = B(10*i-10) + C(22-2*i)

ENDDO
```



Iteration Vectors

What do we do for nested loops?

- Need to consider the nesting level of a loop
- Nesting level of a loop is equal to one more than the number of loops that enclose it.
- 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 Nesting level 1 level.

enddo

• Thus, the iteration vector is: $\{i_1, i_2, ..., i_n\}$ where i_k , $1 \le k \le n$ represents the iteration number for the loop at nesting level k

```
I_1 = L_1, U_1

do I_2 = L_2, U_2

\vdots
do I_n = L_n, U_n
a(f_1(\vec{I}), f_2(\vec{I}), \dots, f_m(\vec{I})) = \dots
\dots = a(g_1(\vec{I}), g_2(\vec{I}), \dots, g_m(\vec{I}))
enddo
\vdots
enddo
```



Iteration Vectors - Example

Example:

```
DO I = 1, 2

DO J = 1, 2

S_1 <some statement>

ENDDO

ENDDO
```

 The iteration vector S₁[(2, 1)] denotes the instance of S₁ executed during the 2nd iteration of the I loop and the 1st iteration of the J loop



Iteration Space

Iteration Space: The **set** of all **possible iteration vectors** for a statement.

```
Example:
```

```
DO I = 1, 2

DO J = 1, 2

S_1 <some statement>

ENDDO

ENDDO
```

• The iteration space for S_1 is $\{ (1,1), (1,2), (2,1), (2,2) \}$

Ordering of iteration vectors

- Useful to define an ordering for iteration vectors
- i is a vector and i_k is the k-th element of i
 and i[1:k] is a k-vector consisting of the leftmost k
 elements of i
- Define an order
 - Iteration i precedes iteration j, denoted i < j, iff:

$$i[1:q] = j[1:q]$$
 and $i[q+1] < j[q+1]$ with $0 \le q \le n$

Formal Definition: Loop Dependence

Theorem: Loop Dependence

- There exists a **dependence** from statements **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 < j or i = j and there is a path from S1 to S2 in the body of the loop,
 - statement S1 accesses memory location M on iteration
 i and statement S2 accesses location M on iteration j,
 and
 - 3. one of these accesses is a write



Transformations

 We call a transformation safe if the transformed program has the same "meaning" as the original program

But, what is the "meaning" of a program?

"Meaning" of a program:

- Two computations are equivalent if, on the same inputs they produce the same outputs in the same order.
- Of course the performance of the program may change!



Re-Ordering Transformations

• A reordering transformation is any program transformation that merely changes the order of execution of the code, without adding or deleting any executions of any statements.

- A <u>reordering transformation</u> does <u>not eliminate</u> dependences.
- However, it can change the ordering of the dependence:
 - e.g. change from true anti-dependence or vice versa

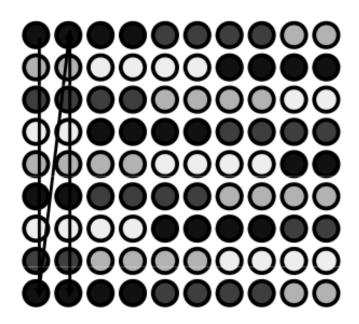


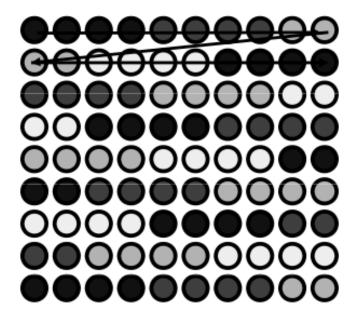
Re-Ordering Transformations

• A reordering transformation preserves a dependence if it preserves the <u>relative execution order</u> of the <u>source</u> and <u>sink</u> of that <u>dependence</u>.

 Reordering transformations changing the <u>control flow</u> can be applied to improve the code <u>unless any</u> <u>dependence is violated.</u>

Example: Improving Data Locality





```
Do i=1,n

do j=1,m

b(i,j)=5.0

enddo

enddo
```

Fundamental Theorem of Dependence

- Fundamental Theorem of Dependence:
 - Any reordering transformation that preserves every dependence in a program preserves the meaning of that program
- A transformation is said to be valid for the program to which it applies if it preserves all dependences in the program.
- A <u>valid transformation preserves</u> the <u>order of loads</u> and stores to every <u>memory location</u> in the program – <u>only input accesses can be reordered</u>.



Distance Vectors

- Distance vectors describe dependences among iterations.
- A dependence is loop carried or independent is crucial to determine --> a loop can be executed in parallel or not?
- They are important to reason about dependences inside of loops.
- Consider a dependence in a loop nest of n loops

$$S_1 \delta S_2$$

- S_2 depends on S_1 :
 - Statement \$1 on iteration *i* is the source of the dependence
 - Statement S2 on iteration j is the sink of the dependence

Distance Vectors

 The distance vector d(i, j) is a vector of length n such that:

$$d(i,j)_k = j_k - i_k$$

- We shall normalize distance vectors for loops in which the index step size is not equal to 1.
- Note that distance vectors describe dependences among iterations not among array elements.



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 D**(*i*, *j*) is defined as a **vector** of **length** *n* such that:

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

$$d(i,j)_k = j_k - i_k$$



DO I = 1, 10

$$S_1 = A(2*I) = B(I) + 1$$

 $S_2 = C(I) = A(I)$
ENDDO

Iteration Vector	I	\mathbf{S}_1	S_2
1	1	A(2)=	=A(1)
2	2	A(4)=	=A(2)
3	3	A(6)=	=A(3)
4	4	A(8)=	=A(4)
5	5	A(10)=	=A(5)
6	6	A(12)=	=A(6)
7	7	A(14)=	=A(7)
8	8	A(16)=	=A(8)
9	9	A(18)=	=A(9)
10	10	A(20)=	=A(10)



DO I = 1, 10

$$A(2*I) = B(I) + 1$$

 $C(I) = A(I)$
ENDDO

Iteration Vector	I	\mathbf{S}_1	S_2
1	1	A(2)=	=A(1)
2	2	A(4)=	=A(2)
3	3	A(6)=	=A(3)
4	4	A(8)=	=A(4)
5	5	A(10)=	=A(5)
6	6	A(12)=	=A(6)
7	7	A(14)=	=A(7)
8	8	A(16)=	=A(8)
9	9	A(18)=	=A(9)
10	10	A(20)=	=A(10)

dependence relation	array element
$S_1[1] \delta S_2[2]$	A(2)
$S_1[2] \delta S_2[4]$	A(4)
$S_1[3] \delta S_2[6]$	A(6)
$S_1[4] \delta S_2[8]$	A(8)
$S_1[5] \delta S_2[10]$	A(10)

- S_2 is true dependent on S_1 .
- •Distance Vector: (*) as distance varies from 1 to 5
- •Direction Vector: (<)

Example:

```
DO I = 1, N

DO J = 1, M

DO K = 1, L

S<sub>1</sub> A(I+1, J, K-1) = A(I, J, K) + 10

ENDDO

ENDDO

ENDDO
```

Example:

```
DO I = 1, N
DO J = 1, M
```

TWO Sample Iterations:

```
Source Iteration I=2, J=2, K=2 accesses the data item A(3,2,1)
Sink iteration I=3, J=2, K=1 accesses the data item A(3,2,1)
Sink-Source = (3,2,1) - (2,2,2) = (1,0,-1)
```

- S₁ has a true dependence on itself.
- Distance Vector: (1, 0, -1)
- Direction Vector: (<, =, >)



Source and Sink of Dependence

- Dependence source is the earlier statement (the statement at the tail of the dependence arrow)
- Dependence sink is the later statement (the statement at the head of the dependence arrow)

```
i = 3
 i = 1
            i = 2
                                      i = 4
                                                  i = 5
W(a[1])
            W(a[2])
                                     W(a[4])
                                                 W(a[5])
                        W(a[3])
R(b[1])
                        R(b[3])
                                                 R(b[5])
            R(b[2])
                                     R(b[4])
W(c[1])
                        W(c[3])
            W(c[2])
                                     W(c[4])
                                                 W(c[5])
R(a[0])
                        R(a[2])
```

Dependences can only go forward in time: always from an earlier iteration to a later iteration.



Direction Vectors

- A dependence cannot exist if it has a direction vector whose leftmost non "=" component is not "<"
 - As this would imply that the sink of the dependence occurs before the source (which is not true)
- How many dependences are there between a pair of statements in a loop nest?
 - ♦ One dependence for every statement instance that is the source of a dependence to another statement instance in the same loop nest.
 - ♦ Compiler collect only the different direction vectors for every different pair of statements.



```
do i = 2, 4

S<sub>1</sub>: a(i) = b(i) + c(i)

S<sub>2</sub>: d(i) = a(i)

end do
```



```
do i = 2, 4 

S<sub>1</sub>[2] S<sub>2</sub>[2] S<sub>1</sub>[3] S<sub>2</sub>[3] S<sub>1</sub>[4] S<sub>2</sub>[4] 

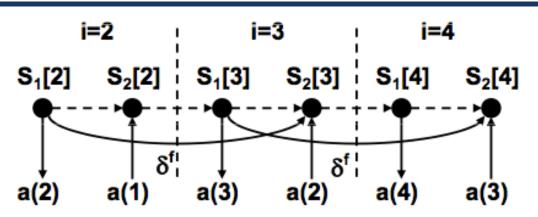
S<sub>1</sub>: a(i) = b(i) + c(i) 

S<sub>2</sub>: d(i) = a(i) 

end do
```

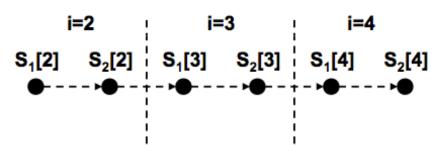
- There is an instance of S1 that precedes an instance of S2 in execution, and S1 produces data that S2 consumes.
- S1 is the source of the dependence; S2 is the sink of the dependence.
- The dependence flows between instances of statements in the same iteration (loop-independent dependence).
- The number of iterations between source and sink (dependence distance) is 0. Dependence direction is (=).





- There is an instance of S1 that precedes an instance of S2 in execution, and S1 produces data that S2 consumes.
- S1 is the source of the dependence; S2 is the sink of the dependence.
- The dependence flows between iterations (*loop-carried dependence*).
- The number of *iterations* between source and sink (dependence distance) is 1. Dependence direction is (<).



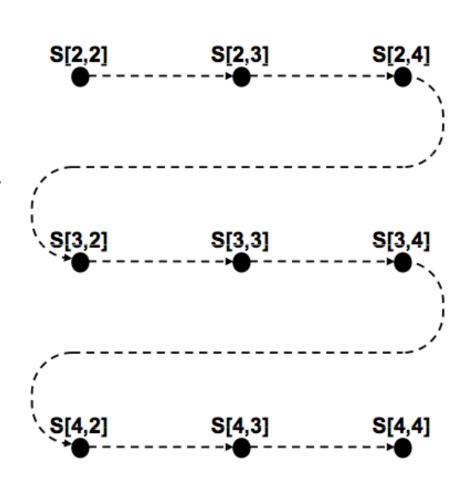


- There is an instance of S₂ that precedes an instance of S₁ in execution and S₂ consumes data that S₁ reassigns.
- S₂ is the source of the dependence with iteration vector I; S₁ is the sink with iteration vector J
- The dependence is loop-carried.
- The distance is 1. The direction is positive with a direction vector (<).
- S_1 is before S_2 in the loop body, so why < direction? $S_2 \delta^{-1} S_1$

Iteration Vector	I	\mathbf{S}_1	S_2
2	2	A(2)=	=A(3)
3	3	A(3)=	=A(4)
4	4	A(4)=	=A(5)



- An instance of S precedes another instance of S and S produces data that S consumes.
- S is both source and sink.
- The dependence is loop-carried.





Example-4 continued...

The dependence distance is (1,-1)
 Iteration vector of sink – iteration vector of source.

$$(3,2)$$
- $(2,3)$ = $(1,-1)$
 $(3,3)$ - $(2,4)$ = $(1,-1)$
 $(4,2)$ - $(3,3)$ = $(1,-1)$
 $(4,3)$ - $(3,4)$ = $(1,-1)$

		source	sink
I	J	S: a(i,j)	S: a(i-1,j+1)
2	2	a(2,2)	a(1,3)
2	3	a(2,3)	a(1,4)
2	4	a(2,4)	a(1,5)
3	2	a(3,2)	a(2,3)
3	3	a(3,3)	a(2,4)
3	4	a(3,4)	a(2,5)
4	2	a(4,2)	a(3,3)
4	3	a(4,3)	a(3,4)
4	4	a(4,4)	a(3,5)

• Direction vector (<,>)



Loop Carried Dependence

Statement S2 has a loop-carried dependence on statement S1, if and only if S1 references location M on iteration i, S2 references M on iteration j and d(i,j) > 0 (that is, D(i,j) contains a "<" as leftmost non "=" component).

Example:

- Array A: $S_1 \delta S_2$ dep. distance (1), direction vector (<)
- Array F: $S_1 \delta^{-1} S_2$ dep. distance (1), direction vector (<)



Loop Carried Dependence

 Level of a loop-carried dependence is the index of the leftmost non-"=" of D(i,j) for the dependence.

For instance:

```
DO I = 1, 10

DO J = 1, 10

DO K = 1, 10

S<sub>1</sub> A(I, J, K+1) = A(I, J, K)

ENDDO

ENDDO

ENDDO
```

- Direction vector for S1 is (=, =, <)
- Level of the dependence is 3
- A level-k dependence between S₁ and S₂ is denoted by S₁ δ_k S₂

Loop Carried Transformations

Example:

DO I = 1, 10

$$S_1$$
 A(I+1) = F(I)
 S_2 F(I+1) = A(I)
ENDDO

Can be transformed or Not?

Will the dependencies be retained or inverted?

can be transformed to:



Loop-Independent Dependence

Definition: Statement **S2** has a *loop-independent dependence* on statement **S1**, **if and only if** there exist **two iteration vectors** *i* and *j* such that:

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

Example:

```
DO I = 1, 10

S_1 A(I) = ...

S_2 ... = A(I)

ENDDO
```

Loop-Independent Dependence

More complicated example:

```
DO I = 1, 9

S_1 A(I) = ...

S_2 = A(10-I)

ENDDO
```

- S_1 stores and S_2 reads A(5) during iteration I=5
- all other dependences are loop carried

Parallelization and Vectorization

Theorem: it is valid to convert a sequential loop to a parallel loop if the loop carries no dependence.

Want to convert loops like:

DO I=1,N

$$X(I) = X(I) + C$$

ENDDO

• to X(1:N) = X(1:N) + C (Fortran 77 to Fortran 90)

However:

DO I=1, N

$$X(I+1) = X(I) + C$$

ENDDO

is not equivalent to X(2:N+1) = X(1:N) + C

Loop Distribution

Can statements in loops which carry dependences be vectorized?

D0 I = 1, N

$$S_1$$
 A(I+1) = B(I) + C
 S_2 D(I) = A(I) + E
ENDDO

• Dependence: $S_1 \delta_1 S_2$ can be converted to:

```
S_1 A(2:N+1) = B(1:N) + C

S_2 D(1:N) = A(1:N) + E
```

Loop Distribution – Example 2

```
DO I = 1, N

S_1   A(I+1) = B(I) + C

S_2   D(I) = A(I) + E

ENDDO
```

transformed to:

```
    leads to:
    S<sub>1</sub> A(2:N+1) = B(1:N) + C
    S<sub>2</sub> D(1:N) = A(1:N) + E
```



Loop Distribution

Loop distribution fails if there is a cyclic dependence

DO I = 1, N

$$S_1$$
 A(I+1) = B(I) + C
 S_2 B(I+1) = A(I) + E
ENDDO

 $S_1 \delta_1 S_2$ and $S_2 \delta_1 S_1$

Loop can be distributed

→ only 1 loop carried

dependence

What about:

Dependence Testing Complications

- Un-known Loop bounds:

What is the relationship between N and 10?

If N<=10 → No loop carried dependence

If $N>10 \rightarrow$ loop carried dependences

Dependence Testing Complications

- Triangular loops:

```
for (j = 0; j < height; j++) {
    for (i = 0; i \le j; i++) {
         printf("*");
    printf("\n");
                              doi = 1, N
Example run:
                               do j = 1, i-1
height of triangle: 6
*
                                a(i,j) = a(j,i)
                          S:
**
                               end do
***
                             end do
***
****
```

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Dependence Testing Complications

```
do i = 1, N
do j = 1, i-1
S: a(i,j) = a(j,i)
end do
end do
```

_	_		
i	j	A (I,j)	A(j,1)
1	1	A(1,1)←	→A(1,1)
1	2	A(1,2)	A(2,1)
2	1	A(2,1)	A(1,2)
2	2	A(2,2)	→A(2,2)
3	1	A(3,1)	A(1,3)
3	2	A(3,2)	A(2,1)

Dependence Testing Complications

- User Variables:

- Serious Problem: *Aliases/Pointers*

Eliminating Anti and Output Dependence

 Anti- and output dependences can always be eliminated through variable renaming.

DO I = 1, N

$$S_1$$
 A(I) = A(I+1)
ENDDO



Anti dependence across loops does not prevent loop parallelization



Loop Parallelization

- A dependence is said to be carried by a loop, if the loop is the outermost loop whose removal eliminates the dependence.
- If a dependence is not carried by the loop, it is loop-independent.

Outermost loop with non "=" dependence carries it.



Loop Parallelization

The iterations of a loop may be executed in parallel with one another *if and only if NO dependences are carried by the loop!*



Loop Parallelization Example-1

```
do i = 2, n-1
do j = 2, m-1
b(i, j) =
= b(i, j-1)
end do
end do
```

- Iterations of loop j must be executed sequentially, but the iterations of loop i may be executed in parallel.
- Outer loop parallelism.

Loop Parallelization Example-2

```
do i = 2, n-1
do j = 2, m-1
b(i, j) =
= b(i-1, j)
end do
end do
```

- Iterations of loop i must be executed sequentially, but the iterations of loop j may be executed in parallel.
- Inner-loop parallelism.



Techniques for Breaking Dependencies and for Dealing with Scalars

Privatization

 remove dependencies created by use of temporary workspaces

Induction Variable Substitution

find closed solutions for basic induction variables

Reduction

 Use reductions as compared to ownimplemented code etc.



Privatization or Scalar Expansion

```
INTEGER J INTEGER J, Jx(N)

DO I = 1, N

J = 

A(I) = J

ENDDO

INTEGER J, Jx(N)

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```

- All scalar assignments will cause loop-carried dependencies
- Can create local per-iteration or (more practically) per thread copies
 - scalar expansion or privatization



Induction Variable Substitution

DO I = 1, N
$$J = J + K$$

$$A(I) = I * K$$

$$A(I) = J$$
ENDDO

- Basic induction variables cause flow dependencies
- Can be replaced by a closed-form solution
 - an induction variable derived from the loop control variable



Reduction

Reduction(ADD, A, N, sum)



Any Questions