

# Automatic Parallelization - 2

Y.N. Srikant

Department of Computer Science  
Indian Institute of Science  
Bangalore 560 012

NPTEL Course on Principles of Compiler Design

# Data Dependence Relations

Flow or true  
dependence

S1:  $X = \dots$



S2:  $\dots = X$



$\delta$

Anti-  
dependence

S1:  $\dots = X$



S2:  $X = \dots$



$\delta^|$

Output  
dependence

S1:  $X = \dots$



S2:  $X = \dots$



$\delta^o$

# Data Dependence Direction Vector

- Data dependence relations are augmented with a direction of data dependence (direction vector)
- There is one direction vector component for each loop in a nest of loops
- The *data dependence direction vector* (or direction vector) is  $\Psi = (\Psi_1, \Psi_2, \dots, \Psi_d)$ , where  $\Psi_k \in \{<, =, >, \leq, \geq, \neq, *\}$
- Forward or “<” direction means dependence from iteration  $i$  to  $i + k$  (*i.e.*, computed in iteration  $i$  and used in iteration  $i + k$ )
- Backward or “>” direction means dependence from iteration  $i$  to  $i - k$  (*i.e.*, computed in iteration  $i$  and used in iteration  $i - k$ ). This is not possible in single loops and possible in two or higher levels of nesting
- Equal or “=” direction means that dependence is in the same iteration (*i.e.*, computed in iteration  $i$  and used in iteration  $i$ )

# Direction Vector Example 1

```
for J = 1 to 100 do {  
S: X(J) = X(J) + c  
}
```

S  $\bar{\delta}_=$  S

```
X(1) = X(1) + c  
X(2) = X(2) + c
```

```
for J = 1 to 99 do {  
S: X(J+1) = X(J) + c  
}
```

S  $\delta_<$  S

```
X(2) = X(1) + c  
X(3) = X(2) + c
```

```
for J = 1 to 99 do {  
S: X(J) = X(J+1) + c  
}
```

S  $\bar{\delta}_<$  S

```
X(1) = X(2) + c  
X(2) = X(3) + c
```

```
for J = 99 downto 1 do {  
S: X(J) = X(J+1) + c  
}
```

S  $\delta_<$  S

```
X(99) = X(100) + c  
X(98) = X(99) + c  
note '-ve' increment
```

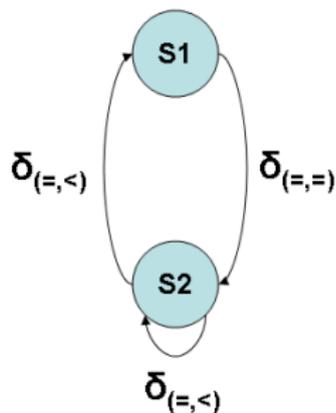
```
for J = 2 to 101 do {  
S: X(J) = X(J-1) + c  
}
```

S  $\delta_<$  S

```
X(2) = X(1) + c  
X(3) = X(2) + c
```

# Direction Vector Example 2

```
for I = 1 to 5 do {  
  for J = 1 to 4 do {  
S1:    A(I, J) = B(I, J) + C(I, J)  
S2:    B(I, J+1) = A(I, J) + B(I, J)  
  }  
}
```



## Demonstration of direction vector

I=1, J=1:  $A(1,1)=B(1,1)+C(1,1)$   $\leftarrow$  S1  $\delta_{(=,=)}$  S2  
 $B(1,2)=A(1,1)+B(1,1)$   $\leftarrow$  S2  $\delta_{(=,<)}$  S1  
J=2:  $A(1,2)=B(1,2)+C(1,2)$   
 $B(1,3)=A(1,2)+B(1,2)$   $\leftarrow$  S2  $\delta_{(=,<)}$  S2  
J=3:  $A(1,3)=B(1,3)+C(1,3)$   
 $B(1,4)=A(1,3)+B(1,3)$

# Direction Vector Example 3

S1  $\delta_{(<,>)}$  S2

```
for I = 1 to N do {  
  for J = 1 to N do {  
S1:   A(I+1, J) = ...  
S2:   ... = A(I, J+1)  
  }  
}
```

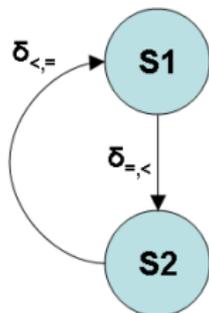
```
I = 1, J = 2  
S1:   A(2,2) = ...  
  
I = 2, J = 1  
S2:   ... = A(2,2)
```

S2  $\delta_{(<,>)}$  S1

```
for I = 1 to N do {  
  for J = 1 to N do {  
S1:   ... = A(I, J+1)  
S2:   A(I+1, J) = ...  
  }  
}
```

```
I = 1, J = 2  
S2:   A(2,2) = ...  
  
I = 2, J = 1  
S1:   ... = A(2,2)
```

# Direction Vector Example 4



```
for I = 1 to 100 do {  
  for J = 1 to 100 do {  
    for K = 1 to 100 do {  
      S1: X(I, J+1, K) = A(I, J, K) + 10  
    }  
    for L = 1 to 50 do {  
      S2: A(I+1, J, L) = X(I, J, L) + 5  
    }  
  }  
}
```

	I = 1	I = 2
J = 1	X(1,2,K) = A(1,1,K) A(2,1,L) = X(1,1,L)	X(2,2,K) = A(2,1,K) A(3,1,L) = X(2,1,L)
J = 2	X(1,3,K) = A(1,2,K) A(2,2,L) = X(1,2,L)	X(2,3,K) = A(2,2,K) A(3,2,L) = X(2,2,L)
J = 3	X(1,4,K) = A(1,3,K) A(2,3,L) = X(1,3,L)	X(2,4,K) = A(2,3,K) A(3,3,L) = X(2,3,L)

Annotations: A blue dashed arrow labeled  $\delta_{=,<}$  points from the J=1 row to the J=2 row. A red dashed arrow labeled  $\delta_{<=,}$  points from the I=1 column to the I=2 column.

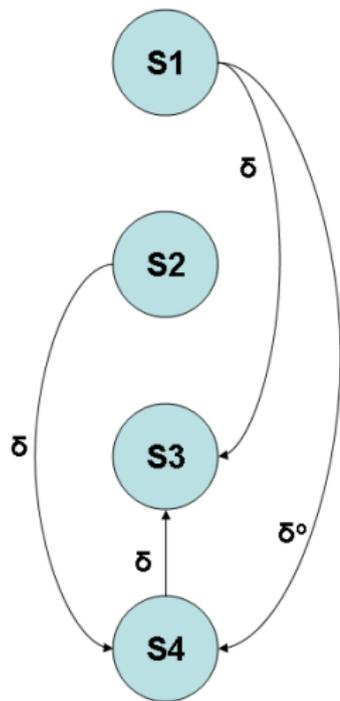
# Data Dependence Graph and Vectorization

- Individual nodes are statements of the program and edges depict data dependence among the statements
- If the DDG is acyclic, then vectorization of the program is possible and is straightforward
  - Vector code generation can be done using a topological sort order on the DDG
- Otherwise, find all the strongly connected components of the DDG, and reduce the DDG to an acyclic graph by treating each SCC as a single node
  - SCCs cannot be fully vectorized; the final code will contain some sequential loops and possibly some vector code

# Data Dependence Graph and Vectorization

- If all the dependence relations in a loop nest have a direction vector value of “=” for a loop, then the iterations of that loop can be executed in parallel with no synchronization between iterations
- Any dependence with a forward (<) direction in an outer loop will be satisfied by the serial execution of the outer loop
- If an outer loop L is run in sequential mode, then all the *dependences* with a forward (<) direction at the outer level (of L) will be automatically satisfied (even those of the loops inner to L)
- However, this is not true for those dependences with with (=) direction at the outer level; the dependences of the inner loops will have to be satisfied by appropriate statement ordering and loop execution order

# Vectorization Example 1



```
for l = 1 to 99 {  
  S1: X(l) = l  
  S2: B(l) = 100 - l  
}
```

Loop A

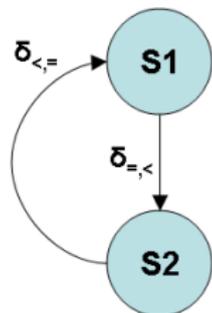
```
for l = 1 to 99 {  
  S3: A(l) = F(X(l))  
  S4: X(l+1) = G(B(l))  
}
```

Loop B

```
X(1:99) = (/1:99/)  
B(1:99) = (/99:1:-1/)  
X(2:100) = G(B(1:99))  
A(1:99) = F(X(1:99))
```

Loop A is parallelizable, but loop B is not, due to forward dependence of S3 on S4

# Vectorization Example 2.1

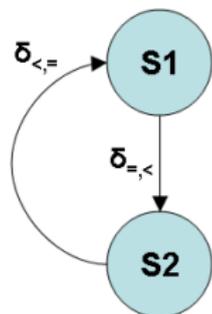


```
for I = 1 to 100 do {  
  for J = 1 to 100 do {  
    for K = 1 to 100 do {  
      S1: X(I, J+1, K) = A(I, J, K) + 10  
    }  
    for L = 1 to 50 do {  
      S2: A(I+1, J, L) = X(I, J, L) + 5  
    }  
  }  
}
```

	I = 1	I = 2
J = 1	X(1,2,K) = A(1,1,K) A(2,1,L) = X(1,1,L)	X(2,2,K) = A(2,1,K) A(3,1,L) = X(2,1,L)
J = 2	X(1,3,K) = A(1,2,K) A(2,2,L) = X(1,2,L)	X(2,3,K) = A(2,2,K) A(3,2,L) = X(2,2,L)
J = 3	X(1,4,K) = A(1,3,K) A(2,3,L) = X(1,3,L)	X(2,4,K) = A(2,3,K) A(3,3,L) = X(2,3,L)

Annotations: A blue dashed arrow labeled  $\delta_{=,<}$  points from the J=1 row to the J=2 row. A red dashed arrow labeled  $\delta_{<=,=}$  points from the J=2 row to the J=3 row. Red dashed arrows also point from the right side of the table to the right side of the J=1 and J=2 rows.

# Vectorization Example 2.2



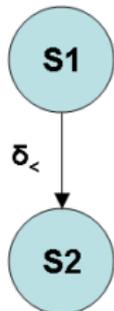
I loop cannot be vectorized due to the cycle.

I and J loops cannot be parallelized, due to '<' direction vector. K and L loops can be parallelized

```
for I = 1 to 100 do {  
  for J = 1 to 100 do {  
    for K = 1 to 100 do {  
      S1:      X(I, J+1, K) = A(I, J, K) + 10  
    }  
    for L = 1 to 50 do {  
      S2:      A(I+1, J, L) = X(I, J, L) + 5  
    }  
  }  
}
```

```
for I = 1 to 100 do {  
  X(I, 2:101, 1:100) = A(I, 1:100, 1:100) + 10  
  A(I+1, 1:100, 1:50) = X(I, 1:100, 1:50) + 5  
}
```

# Vectorization Example 2.3

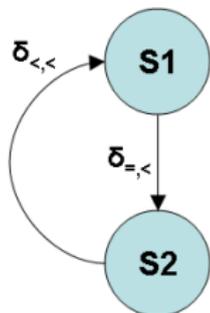


If the I loop is run sequentially, the I-loop dependences are satisfied; J-loop dependences change as shown and there are no more cycles. The loops can be vectorized. However, J-loop cannot be (still) parallelized.

```
for I = 1 to 100 do {  
  for J = 1 to 100 do {  
    for K = 1 to 100 do {  
S1:      X(I, J+1, K) = A(I, J, K) + 10  
    }  
    for L = 1 to 50 do {  
S2:      A(I+1, J, L) = X(I, J, L) + 5  
    }  
  }  
}
```

```
for I = 1 to 100 do {  
  X(I, 2:101, 1:100) = A(I, 1:100, 1:100) + 10  
  A(I+1, 1:100, 1:50) = X(I, 1:100, 1:50) + 5  
}
```

# Vectorization Example 2.4

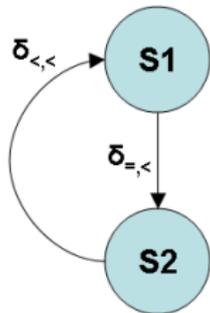


```
for I = 1 to 100 do {  
  for J = 1 to 100 do {  
    for K = 1 to 100 do {  
      S1: X(I, J+1, K) = A(I, J, K) + 10  
    }  
    for L = 1 to 50 do {  
      S2: A(I+1, J+1, L) = X(I, J, L) + 5  
    }  
  }  
}
```

	I = 1	I = 2
J = 1	X(1,2,K) = A(1,1,K) A(2,2,L) = X(1,1,L)	X(2,2,K) = A(2,1,K) A(3,2,L) = X(2,1,L)
J = 2	X(1,3,K) = A(1,2,K) A(2,3,L) = X(1,2,L)	X(2,3,K) = A(2,2,K) A(3,3,L) = X(2,2,L)
J = 3	X(1,4,K) = A(1,3,K) A(2,4,L) = X(1,3,L)	X(2,4,K) = A(2,3,K) A(3,4,L) = X(2,3,L)

The table illustrates the execution of the nested loops for I=1 and I=2. The rows represent J=1, 2, 3. The columns represent I=1 and I=2. The operations are shown as X(I, J+1, K) = A(I, J, K) + 10 and A(I+1, J+1, L) = X(I, J, L) + 5. A blue dashed arrow labeled  $\delta_{=,<}$  indicates the vectorization of the innermost loop (K). A red dashed arrow labeled  $\delta_{<,<}$  indicates the vectorization of the middle loop (L).

# Vectorization Example 2.5



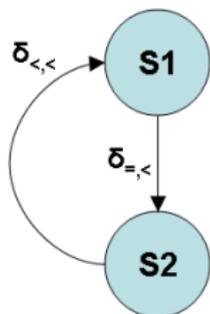
If the program is changed slightly, then dependences change as shown. I and J loops are not parallelizable. If I and J loops are interchanged and J-loop is run sequentially, I-loop can be parallelized. K and L loops are always parallelizable.

```
for I = 1 to 100 do {  
  for J = 1 to 100 do {  
    for K = 1 to 100 do {  
      S1:      X(I, J+1, K) = A(I, J, K) + 10  
    }  
    for L = 1 to 50 do {  
      S2:      A(I+1, J+1, L) = X(I, J, L) + 5  
    }  
  }  
}
```

```
for I = 1 to 100 do {  
  X(I, 2:101, 1:100) = A(I, 1:100, 1:100) + 10  
  A(I+1, 2:101, 1:50) = X(I, 1:100, 1:50) + 5  
}
```

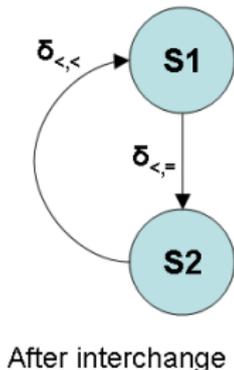
# Vectorization Example 2.6

Before interchange



```

for J = 1 to 100 do {
  for I = 1 to 100 do {
    for K = 1 to 100 do {
      S1:   X(I, J+1, K) = A(I, J, K) + 10
    }
    for L = 1 to 50 do {
      S2:   A(I+1, J+1, L) = X(I, J, L) + 5
    }
  }
}
  
```



	I = 1	I = 2
J = 1	X(1,2,K) = A(1,1,K) A(2,2,L) = X(1,1,L)	X(2,2,K) = A(2,1,K) A(3,2,L) = X(2,1,L)
J = 2	X(1,3,K) = A(1,2,K) A(2,3,L) = X(1,2,L)	X(2,2,K) = A(2,2,K) A(3,3,L) = X(2,2,L)
J = 3	X(1,4,K) = A(1,3,K) A(2,4,L) = X(1,3,L)	X(2,4,K) = A(2,3,K) A(3,4,L) = X(2,3,L)

Annotations: A blue dashed arrow labeled  $\delta_{<,-}$  points from the  $X(1,2,K)$  cell to the  $X(1,3,K)$  cell. Red dashed arrows labeled  $\delta_{<,<}$  point from the  $A(2,2,L)$  cell to the  $A(3,2,L)$  cell, and from the  $A(3,2,L)$  cell to the  $A(3,3,L)$  cell.

# Concurrentization Examples

```
for I = 2 to N do {  
  for J = 2 to N do {  
S1:   A(I,J) = B(I,J) + 2  
S2:   B(I,J) = A(I-1, J-1) - B(I,J)  
  }  
}
```

S1  $\delta_{(<,<)}$  S2, S1  $\bar{\delta}_{(=,=)}$  S2, S2  $\bar{\delta}_{(=,=)}$  S2

```
for I = 2 to N do {  
  for J = 2 to N do {  
S1:   A(I,J) = B(I,J) + 2  
S2:   B(I,J) = A(I, J-1) - B(I,J)  
  }  
}
```

S1  $\delta_{(=,<)}$  S2, S1  $\bar{\delta}_{(=,=)}$  S2, S2  $\bar{\delta}_{(=,=)}$  S2

	I = 1	I = 2
J = 1	A(2,2)= = A(1,1)	A(3,2)= = A(2,1)
J = 2	A(2,3)= = A(1,2)	A(3,3)= = A(2,2)
J = 3	A(2,4)= = A(1,3)	A(3,4)= = A(2,3)

If the I loop is run in serial mode then, the J loop can be run in parallel mode

	I = 1	I = 2
J = 1	A(2,2)= = A(2,1)	A(3,2)= = A(3,1)
J = 2	A(2,3)= = A(2,2)	A(3,3)= = A(3,2)
J = 3	A(2,4)= = A(2,3)	A(3,4)= = A(3,3)

The J loop cannot be run in parallel mode. However, the I loop can be run in parallel mode

# Loop Transformations for increasing Parallelism

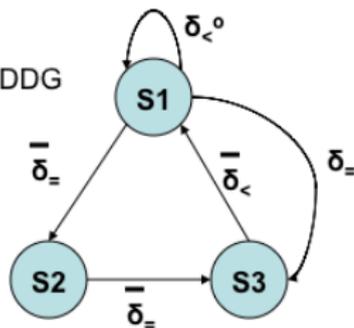
- Recurrence breaking
  - Ignorable cycles
  - Scalar expansion
  - Scalar renaming
  - Node splitting
  - Threshold detection and index set splitting
  - If-conversion
- Loop interchanging
- Loop fission
- Loop fusion

# Scalar Expansion

Not vectorizable or parallelizable

```
for I = 1 to N do {  
  S1: T = A(I)  
  S2: A(I) = B(I)  
  S3: B(I) = T  
}
```

Cyclic DDG



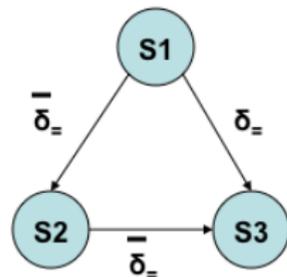
Vectorizable due to scalar expansion

```
for I = 1 to N do {  
  S1: Tx(I) = A(I)  
  S2: A(I) = B(I)  
  S3: B(I) = Tx(I)  
}
```

Parallelizable due to privatization

```
forall I = 1 to N do {  
  private temp  
  S1: temp = A(I)  
  S2: A(I) = B(I)  
  S3: B(I) = temp  
}
```

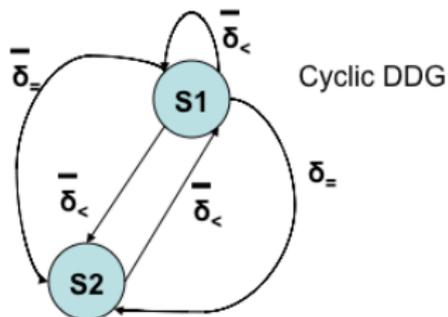
Acyclic DDG



# Scalar Expansion is not always profitable

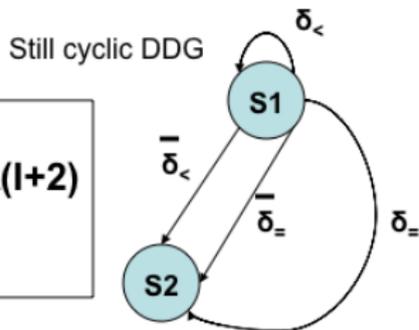
Not vectorizable or parallelizable

```
for l = 1 to N do {  
  S1:  T = T + A(l) + A(l+2)  
  S2:  A(l) = T  
}
```



Not vectorizable even  
after scalar expansion

```
for l = 1 to N do {  
  S1:  Tx(l) = Tx(l-1)+A(l)+A(l+2)  
  S2:  A(l) = Tx(l)  
}
```



# Scalar Renaming

The output dependence  
S1  $\delta^o$  S3 cannot be broken  
by scalar expansion

1.

```
for I = 1 to N do {  
  S1: T = A(I) + B(I)  
  S2: C(I) = T*2  
  S3: T = D(I) * B(I)  
  S4: A(I+2) = T + 5  
}
```

The output dependence  
S1  $\delta^o$  S3 CAN be broken  
by scalar renaming

2.

```
for I = 1 to N do {  
  S1: T1 = A(I) + B(I)  
  S2: C(I) = T1*2  
  S3: T2 = D(I) * B(I)  
  S4: A(I+2) = T2 + 5  
}
```

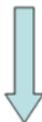
3.

```
S3: T2(1:100) = D(1:100) * B(1:100)  
S4: A(3:102) = T2(1:100) + 5(1:100)  
S1: T1(1:100) = A(1:100) + B(1:100)  
S2: C(1:100) = T1(1:100)*2(1:100)  
    T = T2(100)
```

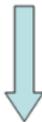
5(1:100) and 2(1:100)  
are vectors of constants

# If-Conversion

```
for l = 1 to 100 do {  
  if (A(l) <= 0) then continue  
  A(l) = B(l) + 3  
}
```



```
for l = 1 to 100 do {  
  BR(l) = (A(l) <= 0)  
  if (~ BR(l)) then  
    A(l) = B(l) + 3  
}
```

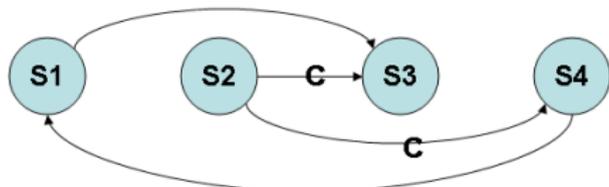


```
BR(1:N) = (A(1:N) <= 0)  
where (~ BR(1:N))  
A(1:N) = B(1:N) + 3
```

```
for l = 1 to N do {  
S1:   A(l) = D(l) + 1  
S2:   if (B(l) > 0) then  
S3:     C(l) = C(l) + A(l)  
S4:     D(l+1) = D(l+1) + 1  
      end if  
}
```



```
for l = 1 to N do {  
S2:   temp(1:N) = B(1:N) > 0  
S4:   where (temp(1:N))  
      D(2:N+1) = D(2:N+1) + 1  
S1:   A(1:N) = D(1:N) + 1  
S3:   where (temp(1:N))  
      C(1:N) = C(1:N) + A(1:N)  
}
```



# Loop Interchange

- For machines with vector instructions, inner loops are preferable for vectorization, and loops can be interchanged to enable this
- For multi-core and multi-processor machines, parallel outer loops are preferred and loop interchange may help to make this happen
- Requirements for simple loop interchange
  - 1 The loops L1 and L2 must be tightly nested (no statements between loops)
  - 2 The loop limits of L2 must be invariant in L1
  - 3 There are no statements  $S_v$  and  $S_w$  (not necessarily distinct) in L1 with a dependence  $S_v \delta_{(<, >)}^* S_w$

# Loop Interchange for Vectorizability

```
for I = 1 to N do {  
  for J = 1 to N do {  
S:   A(I,J+1) = A(I,J) * B(I,J) + C(I,J)  
  }  
}
```

Inner loop is not  
vectorizable

$S \delta_{(=, <)} S$

```
for J = 1 to N do {  
  for I = 1 to N do {  
S:   A(I,J+1) = A(I,J) * B(I,J) + C(I,J)  
  }  
}
```

Inner loop is  
vectorizable

$S \delta_{(<, =)} S$

```
for J = 1 to N do {  
S:   A(1:N, J+1) = A(1:N, J) * B(1:N, J) + C(1:N, J)  
}
```

# Loop Interchange for parallelizability

```
for I = 1 to N do {  
  for J = 1 to N do {  
S:   A(I+1,J) = A(I,J) * B(I,J) + C(I,J)  
  }  
}
```

Outer loop is not parallelizable, but inner loop is

$S \delta_{(<,=)} S$   
Less work per thread

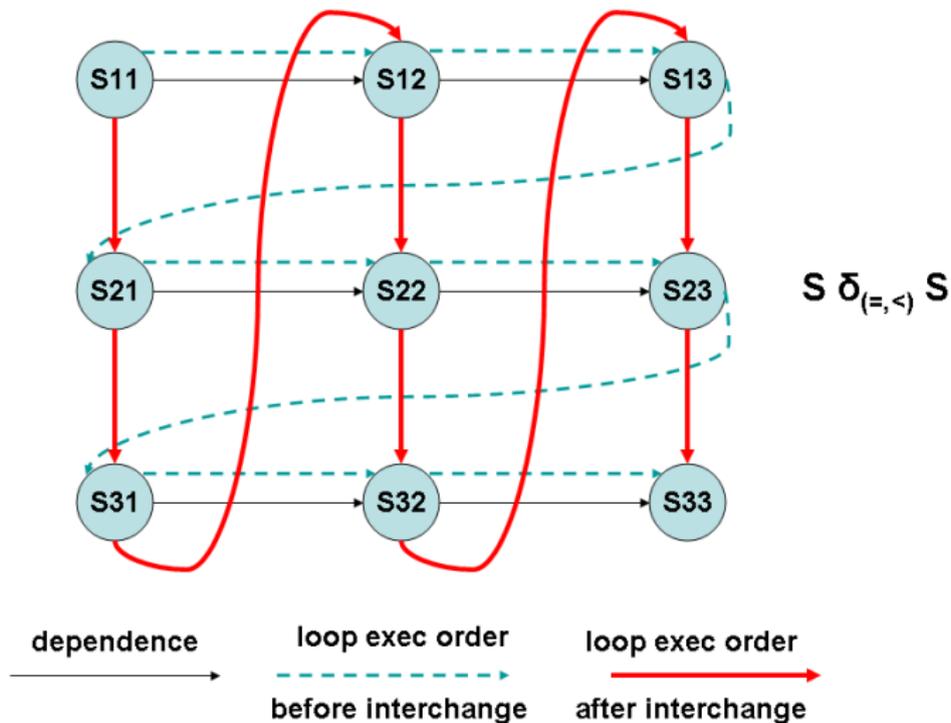
```
for J = 1 to N do {  
  for I = 1 to N do {  
S:   A(I+1,J) = A(I,J) * B(I,J) + C(I,J)  
  }  
}
```

Outer loop is parallelizable but inner loop is not

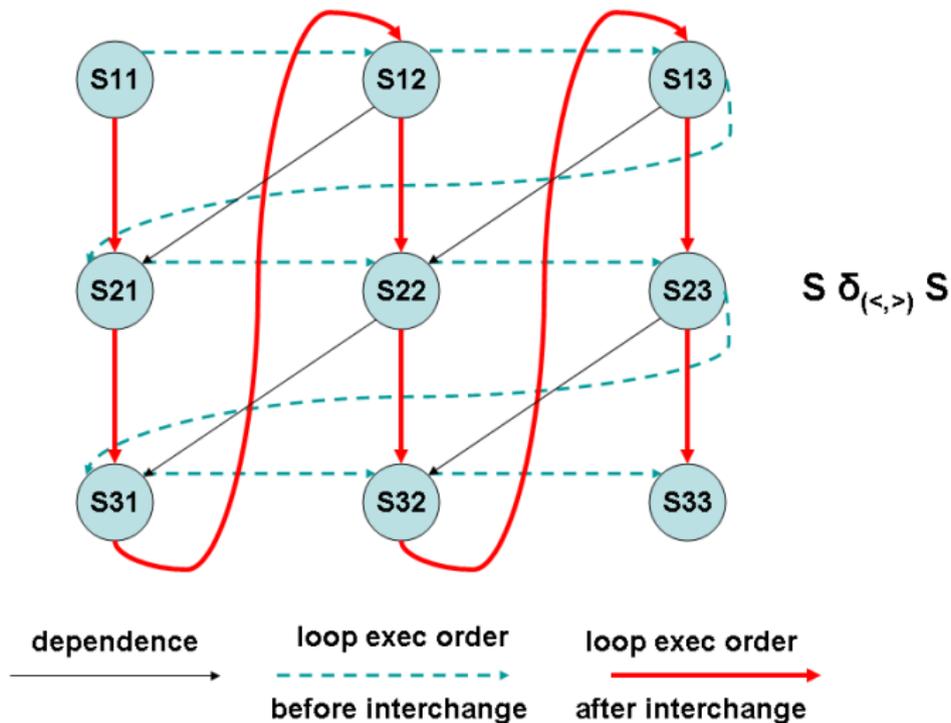
$S \delta_{(=,<)} S$   
More work per thread

```
forall J = 1 to N do {  
  for I = 1 to N do {  
S:   A(I+1,J) = A(I,J) * B(I,J) + C(I,J)  
  }  
}
```

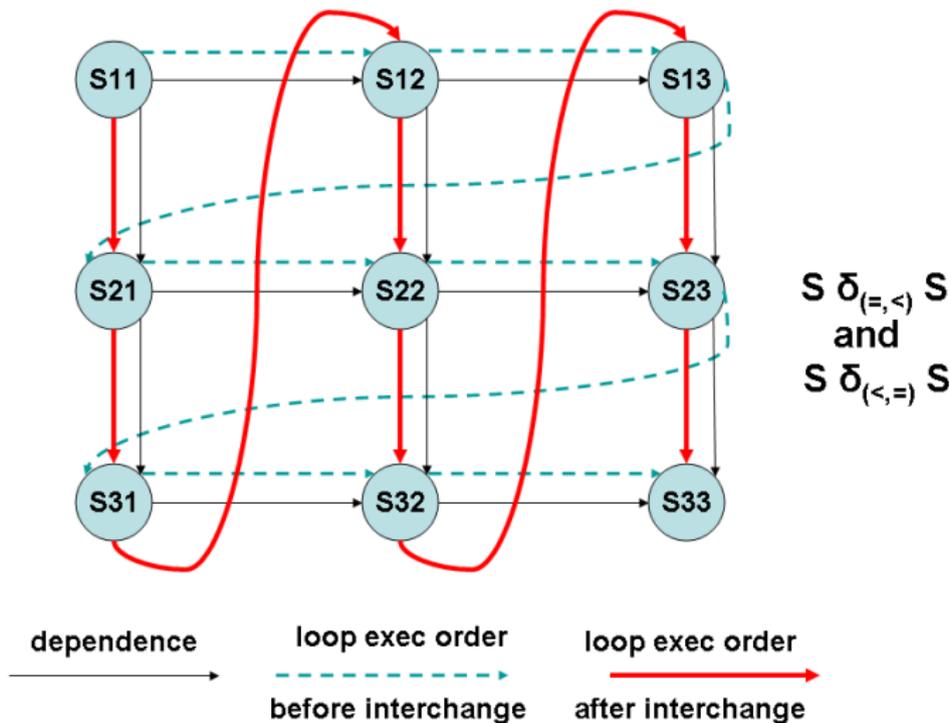
# Legal Loop Interchange



# Illegal Loop Interchange



# Legal but not beneficial Loop Interchange



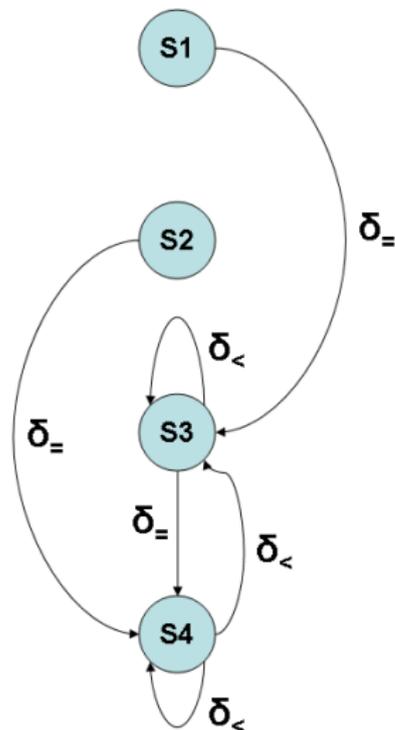
# Loop Fission - Motivation

```
for l = 1 to N do {  
S1:   A(l) = E(l) + 1  
S2:   B(l) = F(l) * 2  
S3:   C(l+1) = C(l) * A(l) + D(l)  
S4:   D(l+1) = C(l+1) * B(l) + D(l)  
}
```

The above loop cannot be vectorized

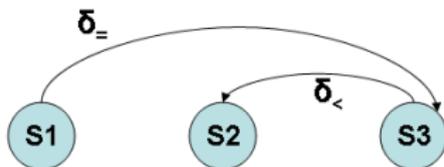
```
L1: for l = 1 to N do {  
S1:   A(l) = E(l) + 1  
S2:   B(l) = F(l) * 2  
}  
  
L2: for l = 1 to N do {  
S3:   C(l+1) = C(l) * A(l) + D(l)  
S4:   D(l+1) = C(l+1) * B(l) + D(l)  
}
```

L1 can be vectorized, but L2 cannot be



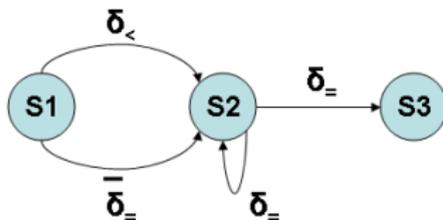
# Loop Fission: Legal and Illegal

```
for I = 1 to N do {  
S1:   A(I) = D(I) * T  
S2:   B(I) = (C(I) + E(I))/2  
S3:   C(I+1) = A(I) + 1  
}
```



In the above loop, S3  $\delta_{<}$  S2, and S3 follows S2. Therefore, cutting the loop between S2 and S3 is illegal. However, cutting the loop between S1 and S2 is legal.

```
for I = 1 to N do {  
S1:   A(I+1) = B(I) + D(I)  
S2:   B(I) = (A(I) + B(I))/2  
S3:   C(I) = B(I) + 1  
}
```



The above loop can be cut between S1 and S2, and also between S2 and S3