



# Multicore Computing

## Lecture06 – Loop Dependence



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## Rule of Thumb

- A loop that matches the following can be parallelized
  - Without restructuring
- 1. All assignments are to arrays
- 2. Each element is assigned by at most one iteration
- 3. No iteration reads elements assigned by any other iteration

```
// simplest case
for (i=0; i<N; i++){
    C[i] = A[i] + B[i-1];
}
```



## Loop-carried Dependences

- Q: Can we parallelize the following loop using OpenMP?

```
a[0] = 1;  
for (i=1; i<N; i++){  
    a[i] = a[i] + a[i-1];  
}
```

```
i=1: a[1] = a[1] + a[0];  
i=2: a[2] = a[2] + a[1];
```

```
// how about this?  
for (i=0; i<N; i+=2){  
    A[i] = A[i] + A[i-1];  
}
```

```
// and this?  
for (i=0; i<N/2; i+=2){  
    A[idx[i]] = A[idx[i]]+B[idx[i]];  
}
```





## Detecting Dependence

- Analyze how each variable is used within a loop
- Is the variable only read and never written?
  - → No dependence
- For each variable written: can there be any accesses in other iterations than the current?
  - → There exist dependences



## Data Dependence

- A *dependence* arises when
  - one operation depends on an earlier operation
- Executing Two Independent Statements
  - On one processor:
    - Statement 1;  
Statement 2;
  - On two processors in parallel:
    - Processor 1:            Processor 2:  
Statement 1;            Statement 2;
    - **Sequential consistency** is guaranteed if
      - Computation results are the same (independent of order)



# Data Dependences

## ■ Example 1

S1: a=1;

S2: b=1;

□ Statements are independent

## ■ Example 2

S1: a=1;

S2: b=2\*a;

$S_1$	$\mathbf{x} = \dots$
$S_2$	$\dots = \mathbf{x}$

□ Dependent (*true dependence*)

○ Second is dependent on first

○ Cannot remove dependency

## ■ Example 3

S1: a=b;

S2: b=1;

$S_1$	$\dots = \mathbf{x}$
$S_2$	$\mathbf{x} = \dots$

□ Dependent (*anti dependence*)

○ First is dependent on second

○ Can you remove dependency?

## ■ Example 4

S1: a=f(x);

S2: a=b;

$S_1$	$\mathbf{x} = \dots$
$S_2$	$\mathbf{x} = \dots$

□ Dependent (*output dependence*)

○ Second is dependent on first

○ Can you remove dependency?



# Data Dependences

- True dependence
  - Read-After-Write (RAW)
- False dependence
  - Anti-dependence
    - Write-After-Read (WAR)
      - S1.  $A = B + 1$     // S1 needs an old value before S2 overwrites it.
      - S2.  $B = 7$
  - Output dependence
    - Write-After-Write (WAW)
      - Ordering affects the final output value of a variable.
- Some dependences can be removed by
  - Rearranging statements
  - Eliminating statements
  - Rewriting with variable renaming



# Data Dependences

## ▪ Anti-dependence

- Write-After-Read (WAR)

- Rewrite with variable renaming

S1.  $A = B + 1$

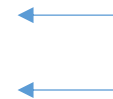
S2.  $B = 7$



N.  $B2 = B$

S1.  $A = B2 + 1$

S2.  $B = 7$



There's no anti-dependence.

S2 & S3 can be executed in parallel.

## ▪ Output dependence

- Write-After-Write (WAW)

- Rewrite with variable renaming

S1.  $B = 3$

S2.  $A = B + 1$

S3.  $B = 7$



S1.  $B2 = 3$

S2.  $A = B2 + 1$

S3.  $B = 7$



There's no output-dependence.

S1 & S3 can be executed in parallel.





## Loop Level Parallelism

- Determining whether two statements are dependent is not easy.
- Example
  - 1:  $a[i] = b[i] + c[i];$
  - 2:  $d[i] = a[i];$
  - There is a true dependence,
  - If we put this code in a loop body, the dependence flows within the same iteration.

```
for(int i=0;i<N;i++){  
     $a[i] = b[i] + c[i];$     /* 1 */  
     $d[i] = a[i];$           /* 2 */  
}
```

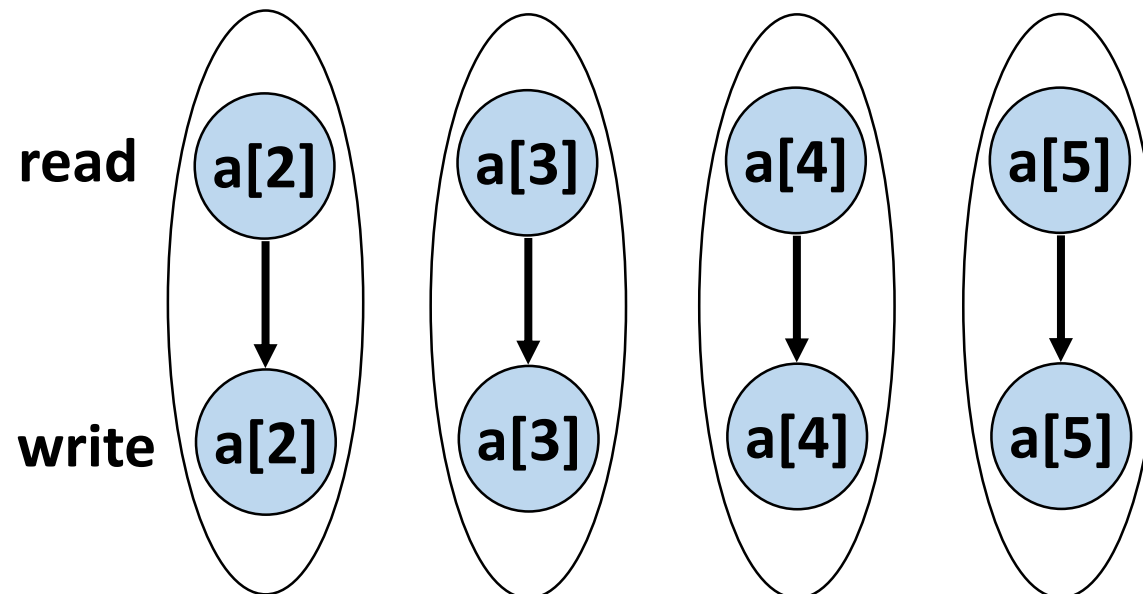
- This dependence is **loop-independent**.
  - aka.: the dependence distance is 0



## Loop Level Parallelism

Example

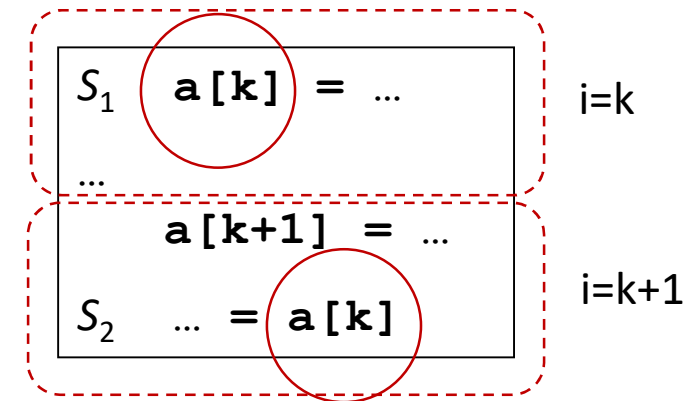
```
for (int i = 2; i <= 5; i++) {  
    a[i] = a[i] + 3  
}
```



## Loop-Carried Dependence: True Dependence

- Example

```
for(int i=0;i<N;i++){  
    a[i] = b[i] + c[i];    /* 1 */  
    d[i] = a[i-1];        /* 2 */  
}
```



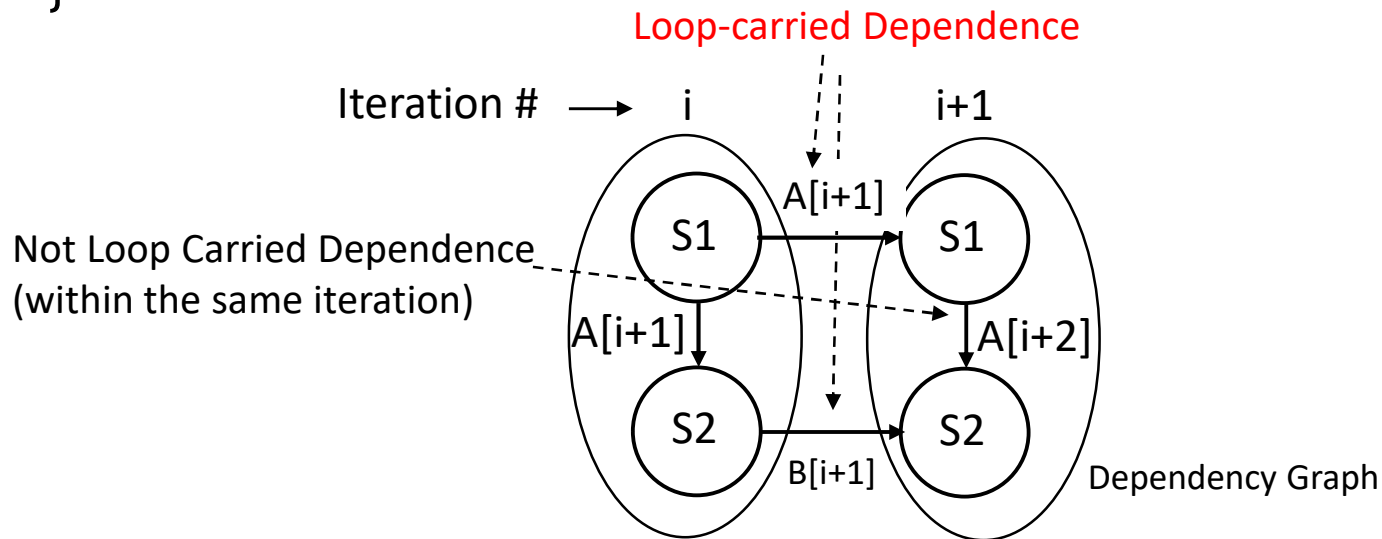
- There is a true dependence
- However, the dependence flows between instances of the statements in different iterations.
  - This is called **loop-carried dependence**.
  - Aka.: The dependence distance is 1.
  - Note: Loop-carried dependence may prevent parallelism.



## Loop-Carried Dependence: True Dependence

Example:

```
for (i=0; i<100; i++) {  
    A[i+1] = A[i] + C[i];    /* S1 */  
    B[i+1] = B[i] + A[i+1]; /* S2 */  
}
```



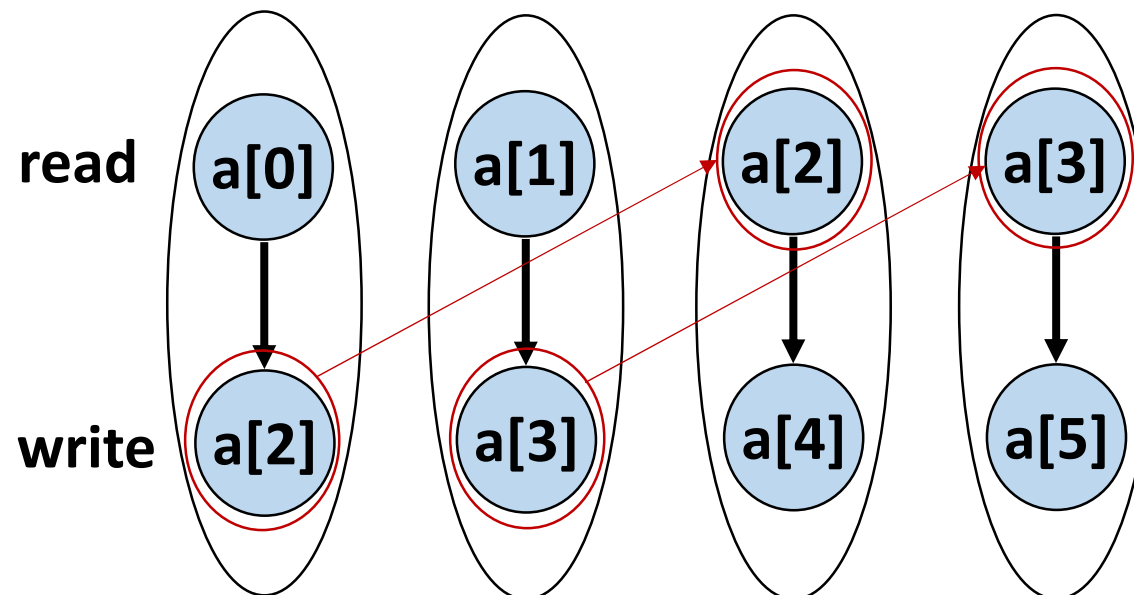
- S1 & S2 use values computed in the earlier iteration
  - These loop-carried dependences prevent loop parallelism.



## Loop-Carried Dependence: True Dependence

Example:

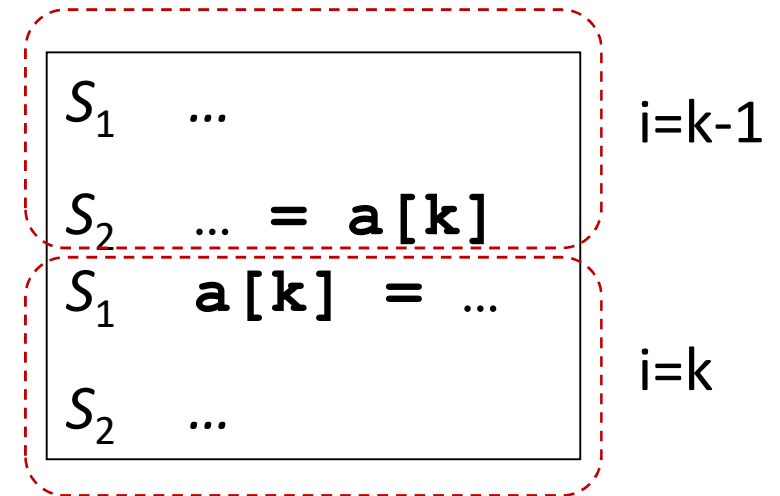
```
for(int i=2; i<=5; i++){  
    a[i] = a[i-2] + 3  
}
```



## Loop-Carried Dependence: Anti Dependence

### ■ Example

```
for(int i=0;i<N;i++){  
    a[i] = b[i] + c[i];    /* 1 */  
    d[i] = a[i+1];        /* 2 */  
}
```



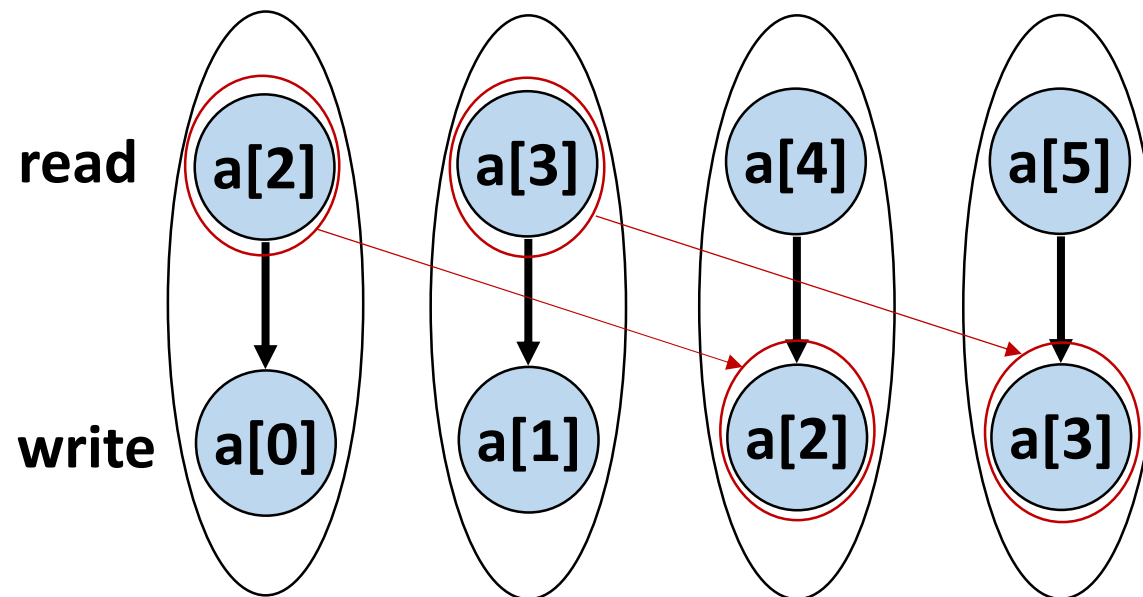
- There is an anti dependence
- This is also a **loop-carried dependence**.
- The dependence distance is -1.
- Note: Loop-carried dependence may prevent parallelism.



## Loop-Carried Dependence: Anti Dependence

Example:

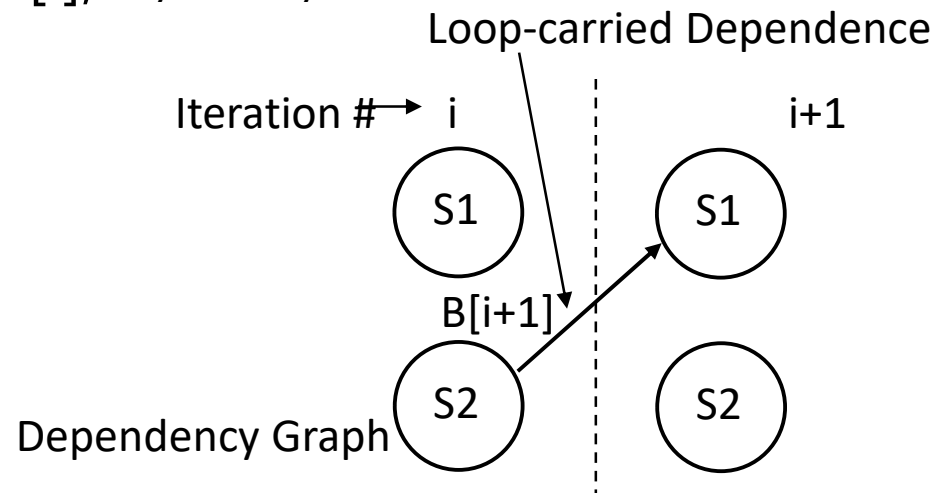
```
for( int i= 2; i<=5; i++){  
    a[i-2] = a[i] + 3  
}
```



## Removing Dependences

- In the loop:

```
for (i=0; i<100; i++) {  
    A[i] = A[i] + B[i];    /* S1 */  
    B[i+1] = C[i] + D[i]; /* S2 */  
}
```



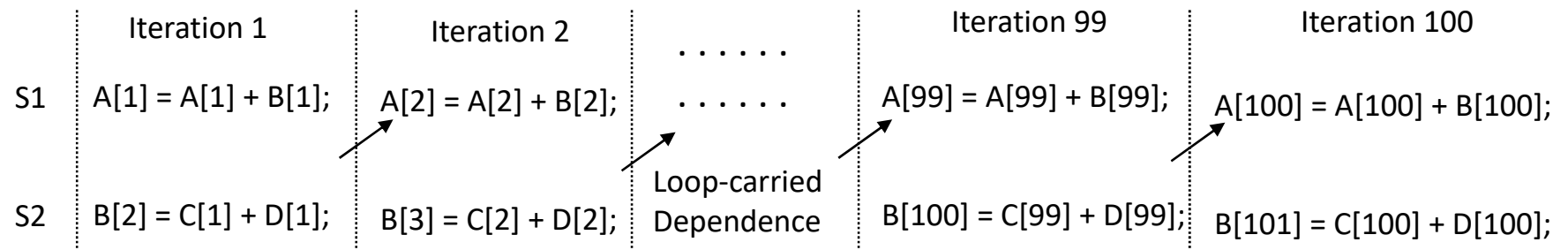
- S1 uses a value B[i] computed by S2 in the earlier iteration
- This dependence is not circular and does not form a chain.
  - Can be made parallel by replacing the code with ...



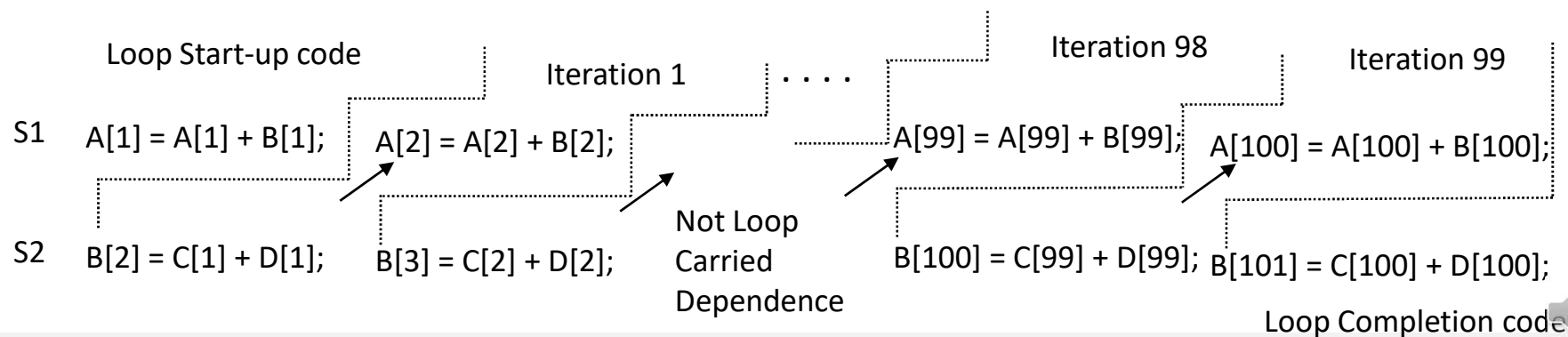


# Removing Dependences

## Original Loop:



## How about this?



## Removing Dependences

Original Loop:

```
for (i=1; i<100; i++) {  
    A[i] = A[i] + B[i];    /* S1 */  
    B[i+1] = C[i] + D[i]; /* S2 */  
}
```

Modified Parallel Loop:

```
A[1] = A[1] + B[1];  
for (i=1; i<99; i++) {  
    B[i+1] = C[i] + D[i];  
    A[i+1] = A[i+1] + B[i+1];  
}  
B[100] = C[99] + D[99];
```



## Removing Dependences

- What kind of dependences are there?

```
for (i=0; i<N; i++) {  
    x = (B[i] + C[i])/2;  
    A[i] = A[i+1] + x;  
}
```

- A: Output dependence for x &  
Anti dependence for  $A[i] = A[i+1]$



## Removing Dependences

```
#pragma omp parallel for  
for (i=0; i<N; i++){  
    A2[i] = A[i+1];  
}
```

```
#pragma omp parallel for private(x)  
for (i=0; i<N; i++) {  
    x = (B[i] + C[i])/2;  
    A[i] = A2[i+1] + x;  
}
```

## Loop Level Parallelism Analysis

```
for (i=0; i<n; i++) {  
    a[i] = b[i+1] * a[i-1];    /* S1 */  
    b[i] = b[i] * coef;        /* S2 */  
    c[i] = 0.5 * (c[i] + a[i]); /* S3 */  
    d[i] = d[i-1] * d[i];      /* S4 */  
}
```

Note that S4 has no dependences with other statements

```
for 'i</ : i; n: i** ( E  
    a@B< b[i+1] * a[i-1];    /* S1 */  
    b@B< b@B) coef:          /* S2 */  
    c@B< / -4 ) 'c@B* a@B:   /* S3 */  
G  
for 'i</ : i; n: i** ( E  
    d@B< d[i-1] * d@B        /* S4 */  
G
```

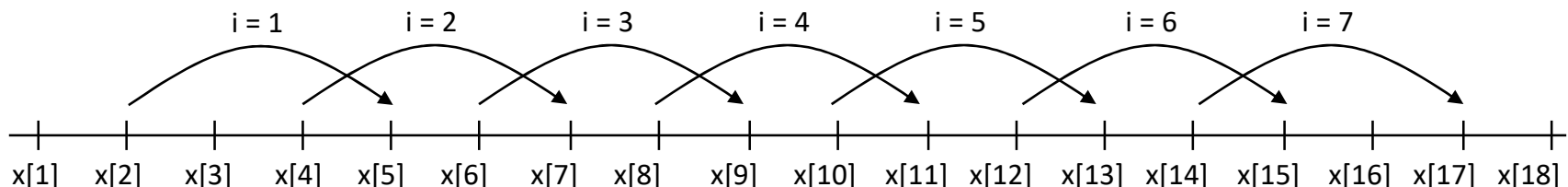
This is called “function parallelism”

## Loop Level Parallelism Analysis

- Is there any dependence?

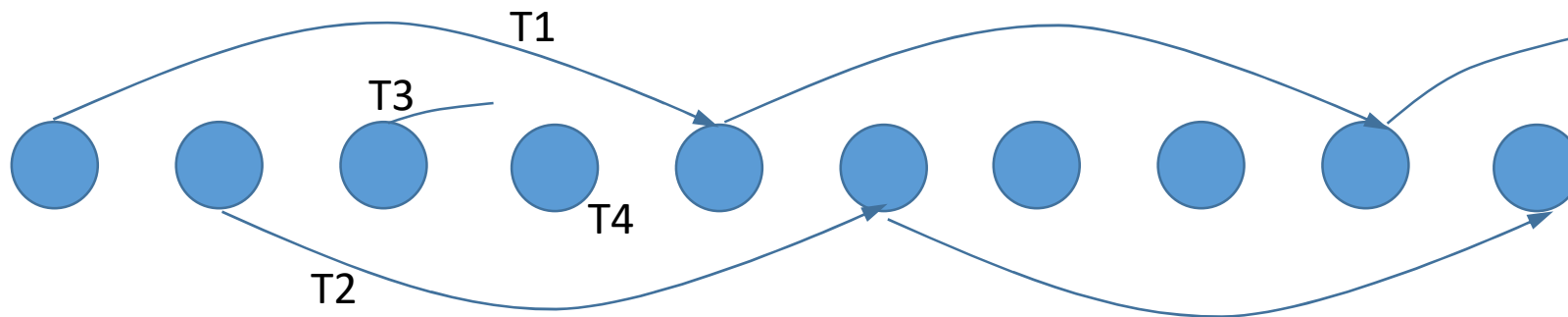
```
for (i=1; i<=100; i=i+1) {  
    A[2*i+3] = A[2*i] + 5;  
}
```

- GCD (Greatest Common Divisor) test to detect loop-carried dependence
  - If an array element with index:  $a*i+b$  is stored and  $c*i+d$  of the same array is loaded later,  $\text{GCD}(c, a)$  must divide  $(d-b)$ . I.e.,  $a*i+b = c*i+d$ 
    - E.g.)  $\text{GCD}(a,c) = 2$ ,  $d-b = -3$ . 2 does not divide -3
    - No loop carried dependence possible
  - GCD test is sufficient to guarantee no loop carried dependence.
  - GCD cannot tell if there 'is' a loop carried dependence.



## Loop Level Parallelism Analysis

- Is there any dependence?  
for (i=4; i<104; i++)  
    A[i] = 2 \* A[i-4];
  - Between a[10], a[6], ...
  - Between a[11], a[7], ...
- Some parallel execution is possible
  - How much?



We can divide this loop into four parallel tasks

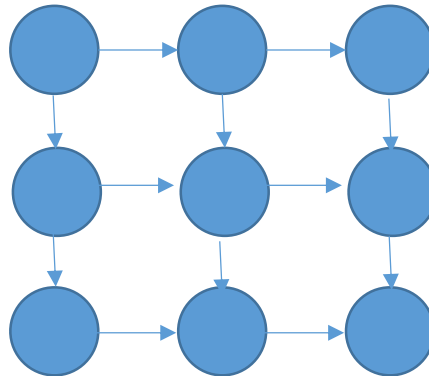
## Loop Level Parallelism Analysis

- Is there any dependence?

```
for (i=1; i<=n; i++)
```

```
  for (j=1; j<=n; j++)
```

```
    A[i][j] = A[i][j-1] + A[i-1][j];
```



Some parallel execution is possible

How?



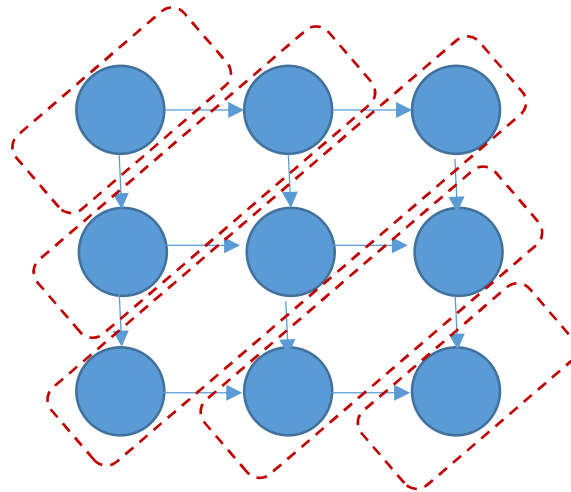
## Loop Level Parallelism Analysis: Recurrence Pattern

- Is there any dependence?

```
for (i=1; i<=n; i++)
```

```
  for (j=1; j<=n; j++)
```

```
    A[i][j] = A[i][j-1] + A[i-1][j];
```



In each diagonal, the nodes are independent of each other  
Let's rewrite the code to iterate over each diagonal

## Loop Level Parallelism Analysis: Recurrence Pattern

- Strategy

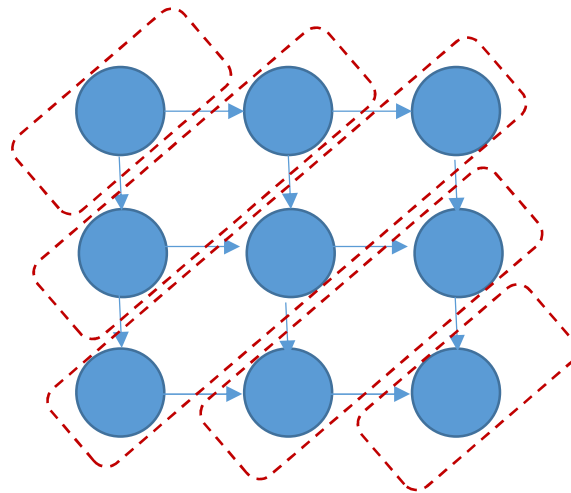
- Calculate number of diagonals

- **for each** diagonal do

Calculate the number of points in the current diagonal

**for each** point in the current diagonal do

Compute the value of the current point in the matrix





## Loop Level Parallelism Analysis: Recurrence Pattern

```
for (i=1; i <= 2*n-1; i++) { // 2n-1 anti-diagonals
    if (i <= n) {
        points = i;    // number of points in diag
        row = i;       // first pt (row,col) in diag
        col = 1;
    }
    else {
        points = 2*n - i;
        row = n;
        col = i-n+1;    // note that row+col = i+1 always
    }
    for_all (k=1; k <= points; k++) {
        a[row][col] = ... // update a[row][col]
        row--; col++;
    }
}
```