CS 314 Principles of Programming Languages

Lecture 18: Parallelism and Dependence Analysis

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Class Information

- Project 2 is released.
- Homework 7 will be released this weekend.

Programming with Concurrency

- A PROCESS or THREAD is a potentially-active execution context
- Classic *von Neumann* model of computing has single thread of control, however parallel programs have more than one
- A process or thread can be thought of as

 An abstraction of a physical PROCESSOR
- Processes/Threads can come from
 - Multiple CPUs
 - Kernel-level multiplexing of single physical machine
 - Language or library level multiplexing of kernel-level abstraction
- They can run
 - In true **parallel**
 - Unpredictably interleaved
 - ▶ Run-until-block

Dependence and Parallelization

Dependence analysis is fundamental to parallelization analysis

Dependence relation: all *task–to–task* execution orderings that must be preserved if the meaning of the program is to remain the same.

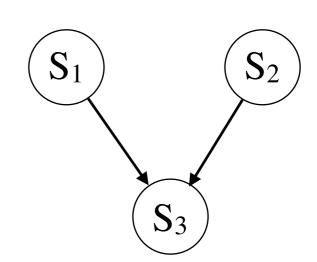
The dependence relation can be modeled as a directed graph such that if $A \rightarrow B$, the result of task A is required for the processing of task B

Example:

$$S_1$$
: $pi = 3.14$

$$S_2$$
: $R = 5$

$$S_3$$
: Area = pi * R^2

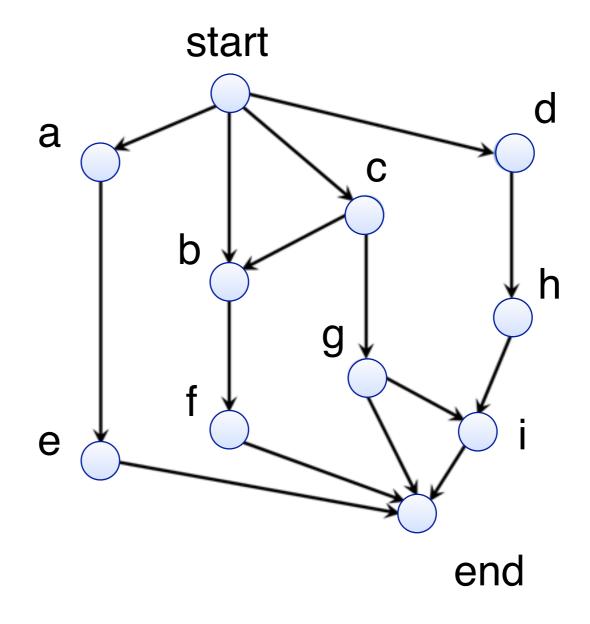


Statement-level dependence graph

Dependence Graph

- Directed acyclic graph (DAG)
- A node represents a task
- A directed edge represents precedence constraint

DAG example 1:



Dependence Graph

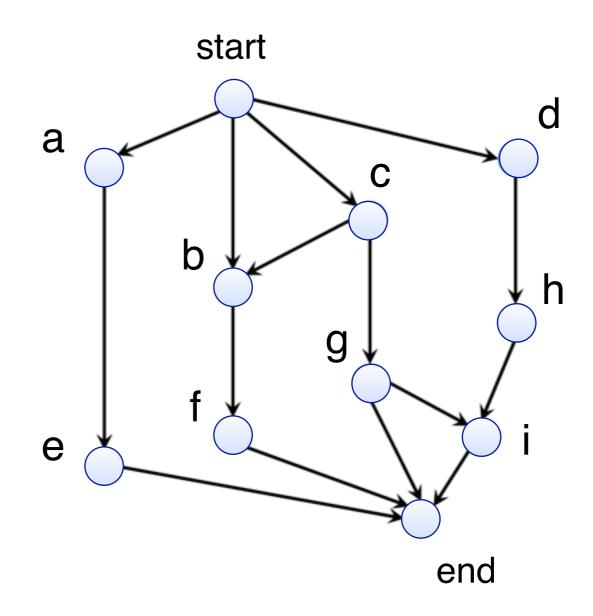
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DAG example 2: $A[1] A[2] A[3] A[4] \cdots A[N-1]$ A[N]S = sum(A[1], A[2], ..., A[N])

 T_p : time to perform computation with p processors

- T₁: work (total # operations)
- T_{∞} : critical path or span

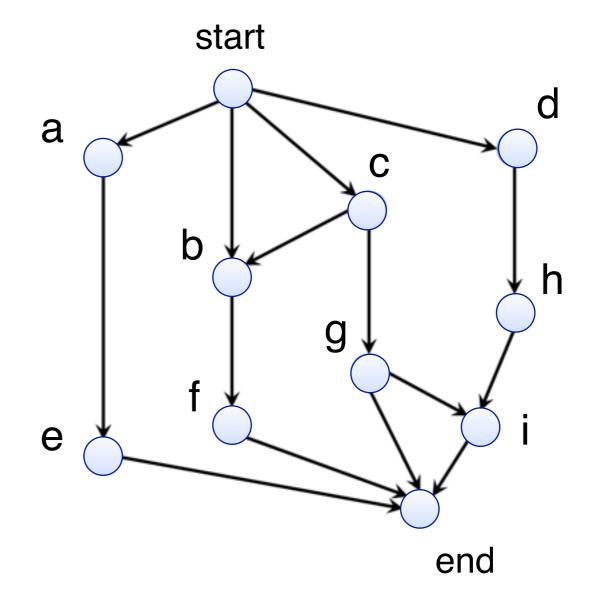
$$T_p \ge \frac{T_1}{p}, \qquad T_p \ge T_\infty$$



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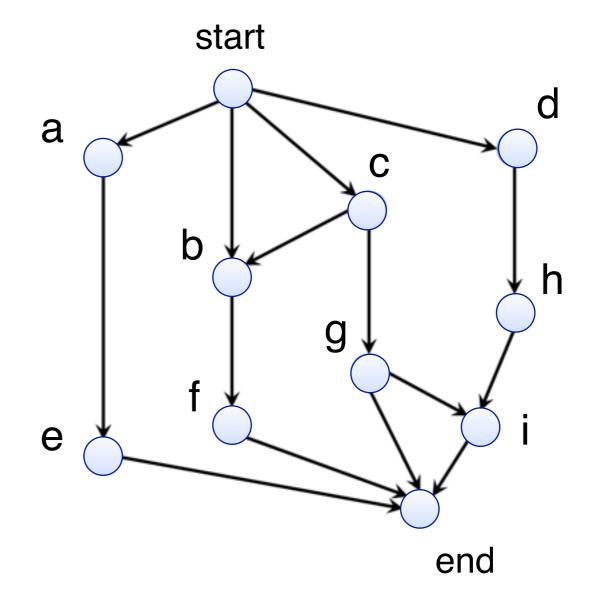


$$T_1 = ?$$

 T_p : time to perform computation with p processors

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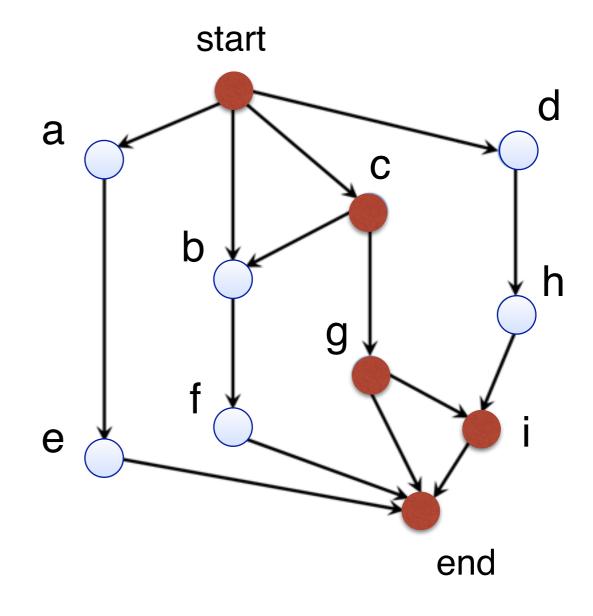


$$T_{\infty} = ?$$

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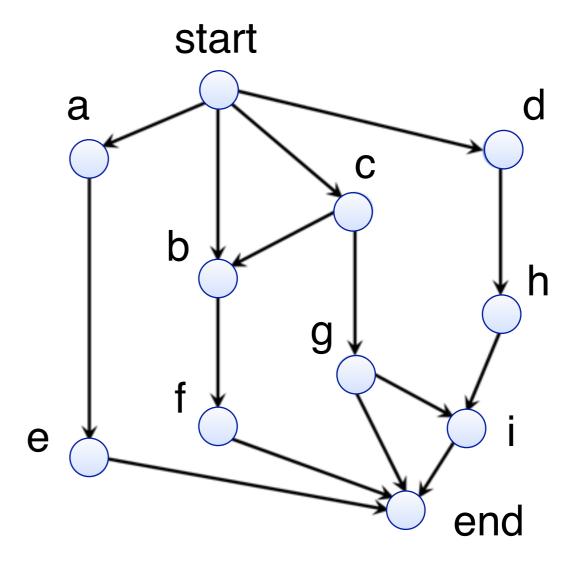


$$T_{\infty} = ?$$

Compute the earliest start time of each node

- Keep a value called S(n) associated with each node n
- For each node *n*

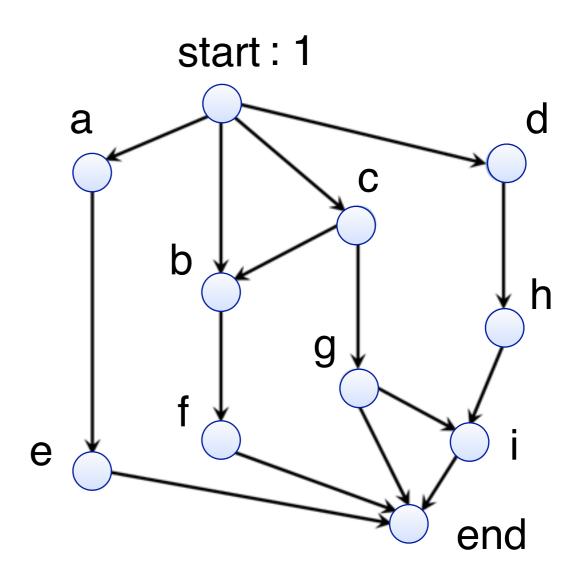
S(n) is the maximum of $\{S(p) + 1\}$, for all $p \in pred(n)$



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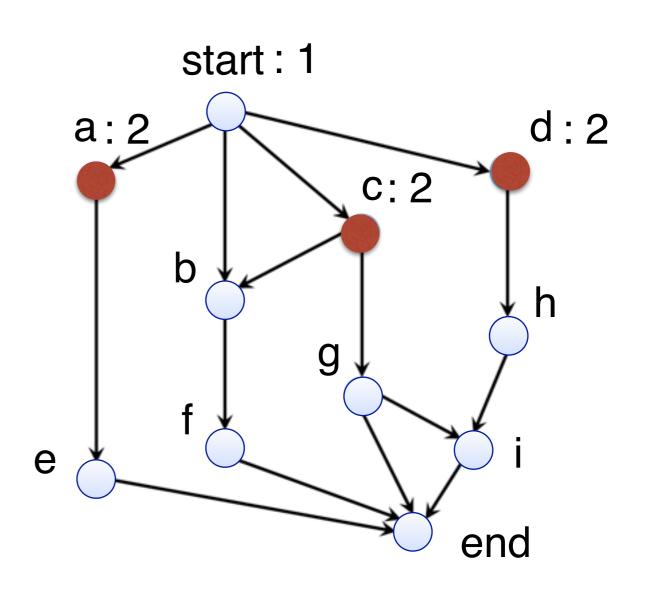
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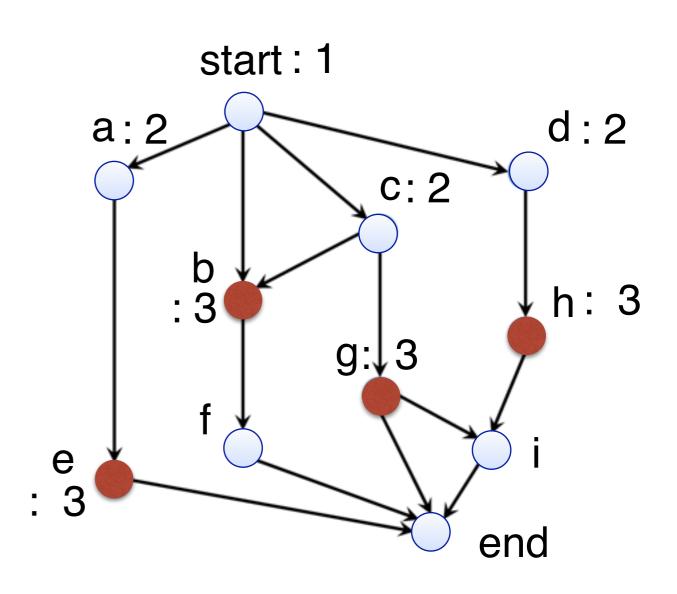
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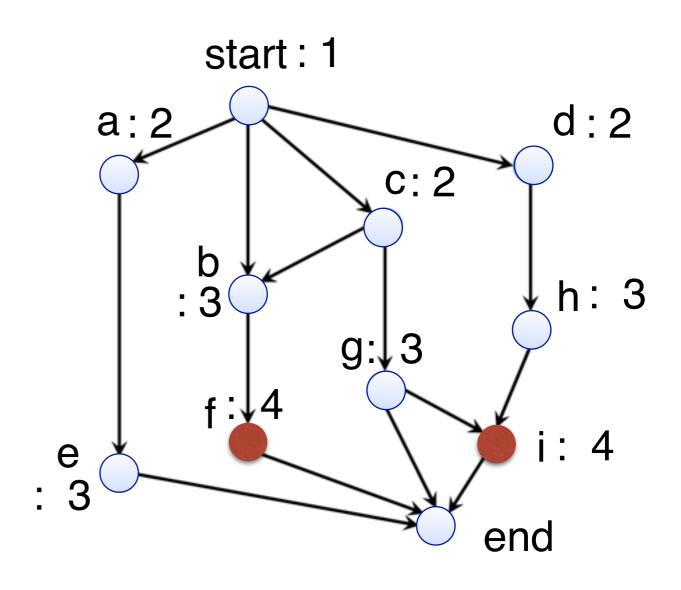
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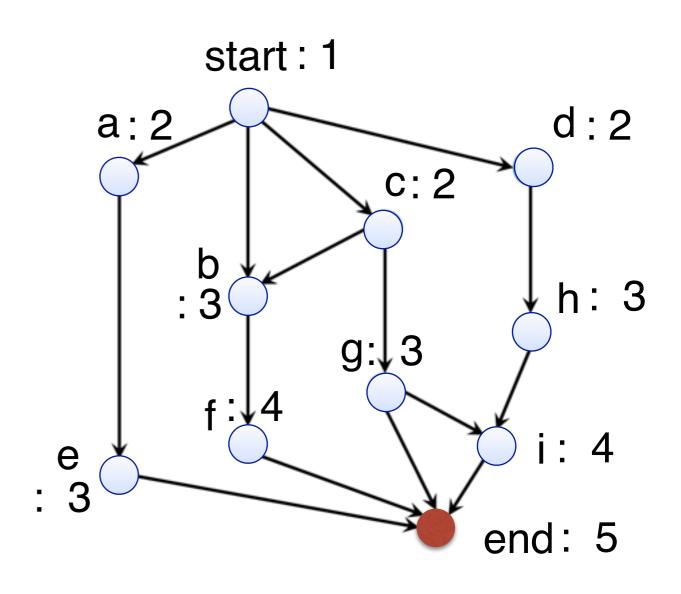
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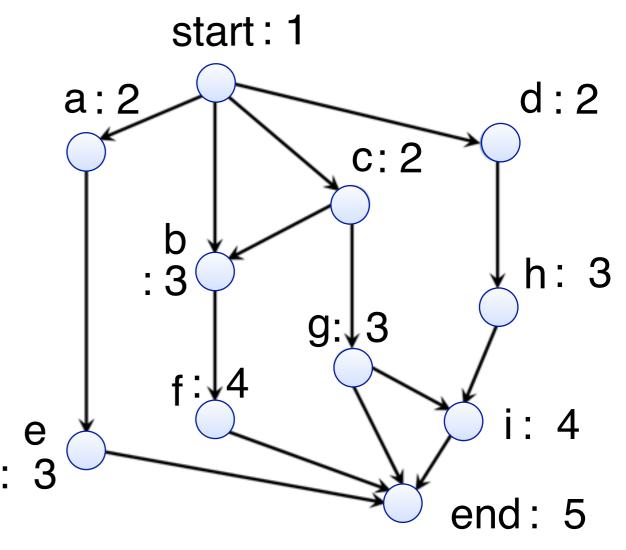
List Scheduling

Based on if the dependence constraints have been resolved

- Schedule the nodes that are ready at every time tick
- A completed operation at the end of one time step can lead to more ready operations at next time tick

Four threads T1, T2, T3, T4

	1	2	3	4	5
T1	start	a	b	f	end
T2		c	e	i	
Т3		d	g		
T4			h		



Automatic Parallelization

We will use **loop analysis** as an example to describe automatic dependence analysis and parallelization.

Assumptions:

- 1. We only have scalar and subscripted variables (no pointers and no control dependence) for loop dependence analysis.
- 2. We focus on *affine loops*: both loop bounds and memory references are affine functions of loop induction variables.

A function $f(x_1, x_2, ..., x_n)$ is **affine** if it is in such a form:

$$\mathbf{f} = c_0 + c_1 * x_1 + c_2 * x_2 + ... + c_n * x_n$$
, where c_i are all constants

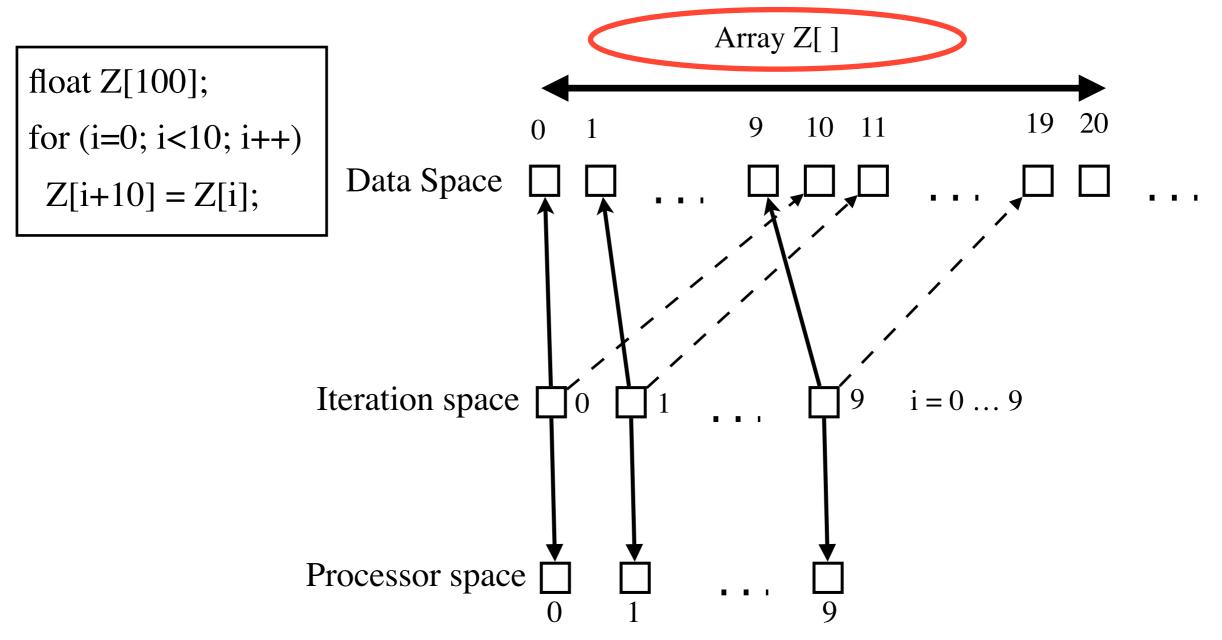
Affine Loops

Three spaces

- Iteration space
 - ▶ The set of dynamic execution instances
 - i.e. the set of value vectors taken by loop indices
 - ▶ A *k*-dimensional space for a *k*-level loop nest
- Data space
 - ▶ The set of array elements accessed
 - ▶ An *n*-dimensional space for an *n*-dimensional array
- Processor space
 - ▶ The set of processors in the system
 - ▶ In analysis, we may pretend there are unbounded # of virtual processors

Three Spaces

• Iteration space, data space, and processor space



Assuming one task is one loop iteration, what is the maximum parallelism?

Maximum parallelism: T_1/T_{∞}

Dependence Definition

Bernstein's Condition: — There is a data dependence from statement (instance) S_1 to statement S_2 (instance) if

- Both statements (instances) access the same memory locations
- One of them is a write
- There is a run-time execution path from S_1 to S_2

float Z[100]; for (i=0; i<10; i++) Z[i+10] = Z[i];

No dependence across any two loop iterations!

Data Dependence Classifications

"S₂ depends on S₁" — (S₁ δ S₂)

True (flow) dependence

occurs when S1 writes a memory location that S2 later reads (RAW).

Anti dependence

occurs when S1 reads a memory location that S2 later writes (WAR).

Output dependence

occurs when S1 writes a memory location that S2 later writes (WAW).

Input dependence

occurs when S1 reads a memory location that S2 later reads (RAR).

• Examples:

```
for (i = 1; i <= 100; i++) {

S1: A[i] = ...

S2: ...= A[i - 1]

}

float Z[100];

for (i =0; i < 12; i++) {

S: Z[i+10] = Z[i];
}
```

- 1. Is there dependence?
- 2. If so, what type of dependence?
- 3. From which statement (instance) to which statement (instance)?

Dependence Testing

Single Induction Variable (SIV) Test

• Single loop nest with constant lower (LB) and upper (UB) bound, and step 1.

• Two array references as affine function of loop induction variable

```
for i = LB, UB, 1
R1: X(a*i + c1) = ... \\ write
R2: ... = X(a*i + c2) ... \\ read
endfor
```

Question: Is there a true dependence between R1 and R2?

Dependence Testing

for
$$i = LB$$
, UB, 1

R1:
$$X(a*i + c1) = ...$$
 \\ write

R2: ... =
$$X(a*i + c2)$$
 ... \\ read

endfor

There is a dependence between R1 and R2 iff

$$\exists i, i': LB \le i \le i' \le UB \text{ and } (a*i+c_1) = (a*i'+c_2)$$

where i and i' represent two iterations in the iteration space. This means that in both iterations, the same element of array X is accessed.

So let's just solve the equation:

$$(a * i + c_1) = (a * i' + c_2)$$
 $(c_1 - c_2)/a = i' - i = \Delta d$

There is a dependence iff

- Δd is an integer value
- UB LB $\geq \Delta d \geq 0$

• Examples:

```
for (i = 1; i <= 100; i++) {

S1: A[i] = ...

S2: ...= A[i - 1]

}

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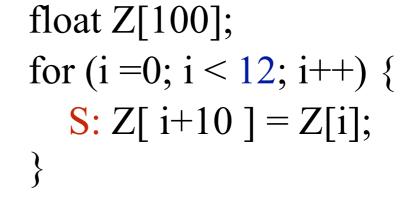
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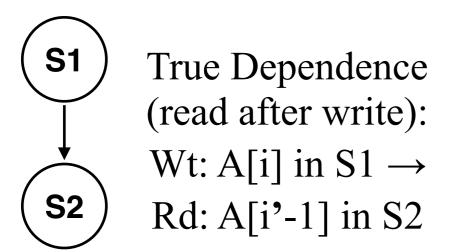
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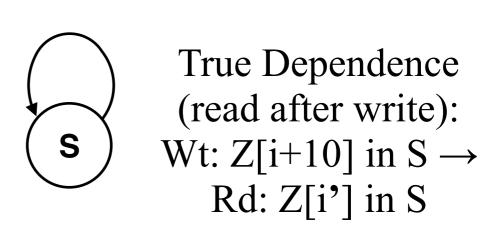
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• Examples:

```
for (i = 1; i <= 100; i++) {
    S1: A[i] = ...
    S2: ...= A[i - 1]
}
```







$$i' = i + 1$$

$$\Delta d = 1$$

$$i' = i + 10$$
$$\Delta d = 10$$

• More Examples:

```
for (i = 1; i \le 100; i++) {

R1: X(i) = ...

R2: ... = X(i + 2)

for (i = 3; i \le 15, i++) {

S1: X(2 * i) = ...

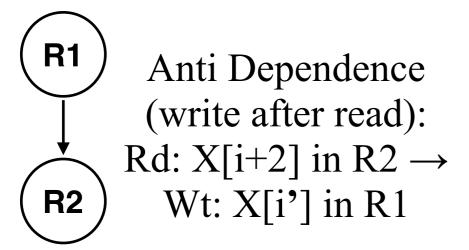
S2: ... = X(2 * i - 1)
}
```

- 1. Is there dependence?
- 2. If so, what type of dependence?
- 3. From which statement (instance) to which statement (instance)?

• More Examples:

```
for (i = 1; i <= 100; i++) {
   R1: X[i] = ...
   R2: ... = X[i + 2]
}
```

```
for (i = 3; i <= 15, i++) {
    S1: X[2 * i] = ...
    S2: ... = X[2 * i - 1]
}
```



S1

No dependence!

Next Class

Reading:

• ALSU, Chapter 11.1 - 11.3