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# COMP 322: Fundamentals of Parallel Programming

## Lecture 2: Computation Graphs, Ideal Parallelism

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# Async and Finish Statements for Task Creation and Termination (Recap)

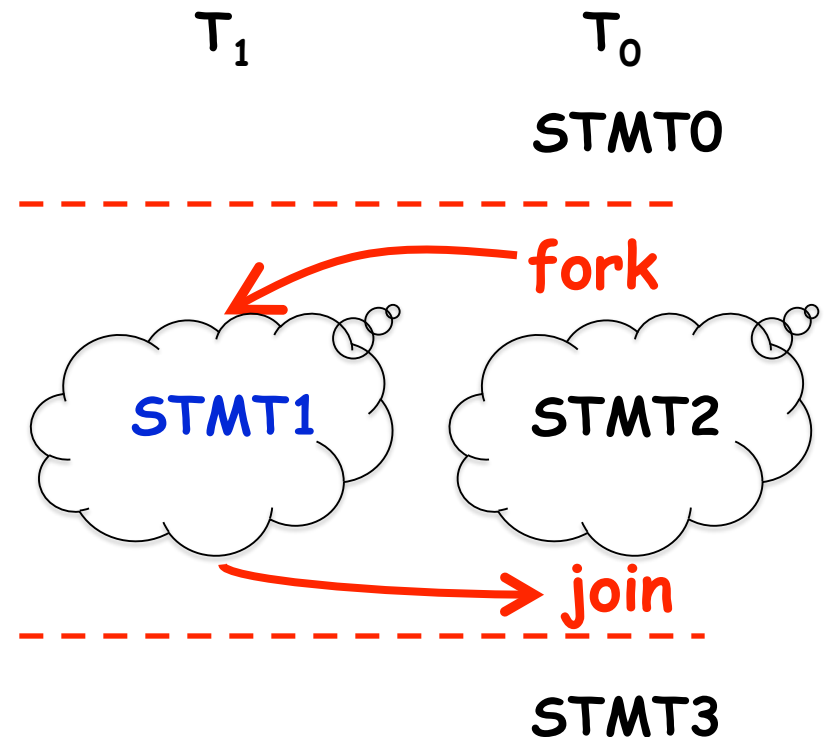
## async S

- Creates a new child task that executes statement S

```
// T0 (Parent task)
STMT0;
finish { //Begin finish
  async {
    STMT1; //T1 (Child task)
  }
  STMT2; //Continue in T0
          //Wait for T1
} //End finish
STMT3; //Continue in T0
```

## finish S

- Execute S, but wait until *all* asyncs in S's scope have terminated.



# One Possible Solution to Worksheet 1 (Parallel Matrix Multiplication)

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```
1. finish {
2.   for (int i = 0 ; i < N ; i++)
3.     for (int j = 0 ; j < N ; j++)
4.       async {
5.         for (int k = 0 ; k < N ; k++)
6.           C[i][j] += A[i][k] * B[k][j];
7.       } // async
8.} // finish
```

*This program generates  $N^2$  parallel async tasks, one to compute each  $C[i][j]$  element of the output array. Additional parallelism can be exploited within the inner  $k$  loop, but that would require more changes than inserting `async` & `finish`.*



# Is this a correct solution for Worksheet 1?

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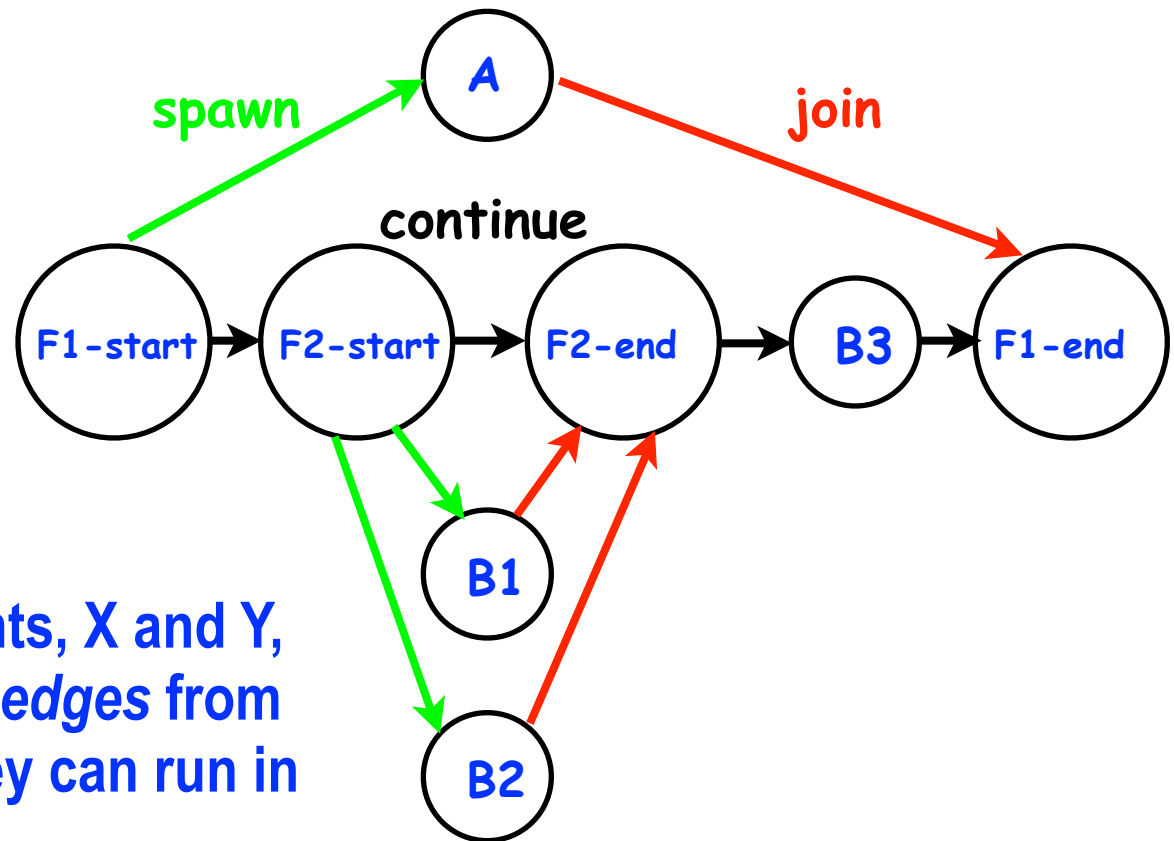
```
1. finish {
2.   for (int i = 0 ; i < N ; i++)
3.     for (int j = 0 ; j < N ; j++)
4.       for (int k = 0 ; k < N ; k++)
5.         async {
6.           C[i][j] += A[i][k] * B[k][j];
7.         } // async
8.} // finish
```



# Which statements can potentially be executed in parallel with each other?

```
1.  finish { // F1
2.    async A;
3.  finish { // F2
4.    async B1;
5.    async B2;
6.  } // F2
7.  B3;
8. } // F1
```

## Computation Graph



**Key idea:** If two statements, X and Y, have *no path of directed edges* from one to the other, then they can run in parallel with each other.



# Computation Graphs

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- A Computation Graph (CG) captures the dynamic execution of a parallel program, for a specific input
- CG nodes are “steps” in the program’s execution
  - A step is a sequential subcomputation without any async, begin-finish and end-finish operations
- CG edges represent ordering constraints
  - “Continue” edges define sequencing of steps within a task
  - “Spawn” edges connect parent tasks to child async tasks
  - “Join” edges connect the end of each async task to its IEF’s end-finish operations
- All computation graphs must be acyclic
  - It is not possible for a node to depend on itself
- Computation graphs are examples of “directed acyclic graphs” (dags)



# Complexity Measures for Computation Graphs

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## Define

- $\text{TIME}(N)$  = execution time of node  $N$
- $\text{WORK}(G)$  = sum of  $\text{TIME}(N)$ , for all nodes  $N$  in CG  $G$ 
  - $\text{WORK}(G)$  is the total work to be performed in  $G$
- $\text{CPL}(G)$  = length of a longest path in CG  $G$ , when adding up execution times of all nodes in the path
  - Such paths are called *critical paths*
  - $\text{CPL}(G)$  is the length of these paths (critical path length)
  - $\text{CPL}(G)$  is also the smallest possible execution time for the computation graph



# What is the critical path length of this parallel computation?

```
1. finish { // F1
2.   async A; // Boil pasta
3.   finish { // F2
4.     async B1; // Chop veggies
5.     async B2; // Brown meat
6.   } // F2
7.   B3; // Make pasta sauce
8. } // F1
```

Step B1



Step B2



Step A



Step B3





# Ideal Parallelism

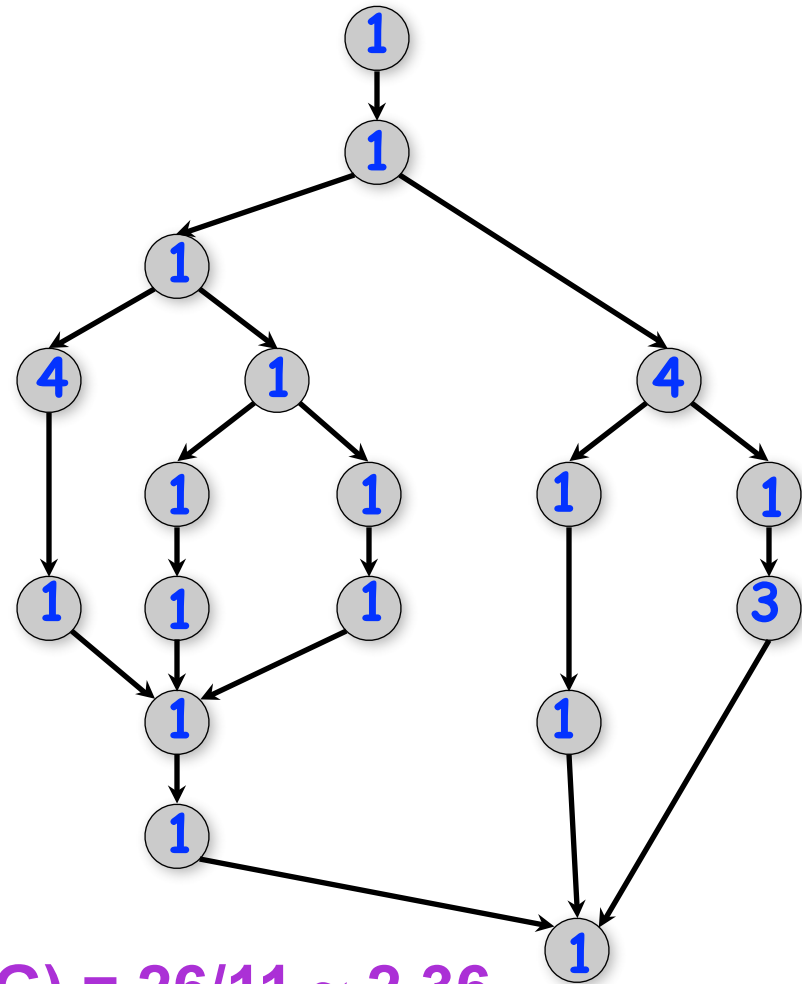
- Define **ideal parallelism** of Computation G Graph as the ratio,  $WORK(G)/CPL(G)$
- Ideal Parallelism is independent of the number of processors that the program executes on, and only depends on the computation graph

## Example:

$$WORK(G) = 26$$

$$CPL(G) = 11$$

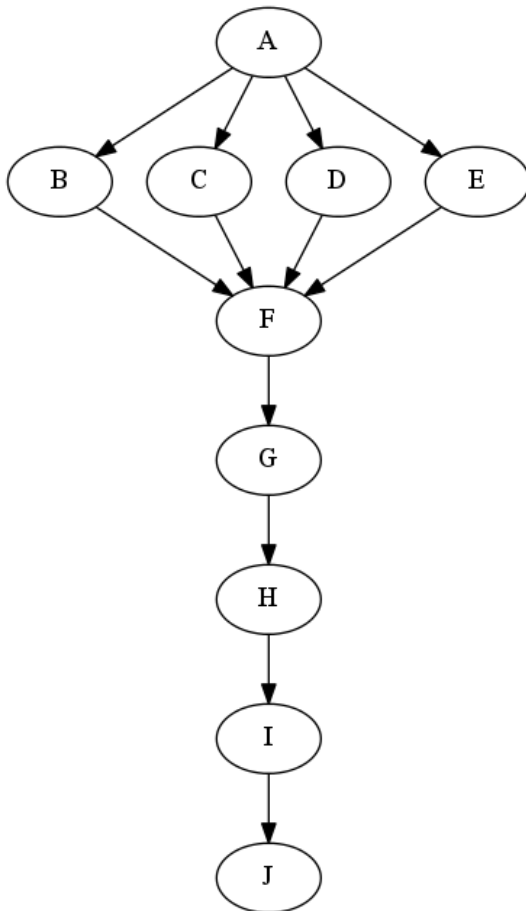
$$\text{Ideal Parallelism} = WORK(G)/CPL(G) = 26/11 \sim 2.36$$



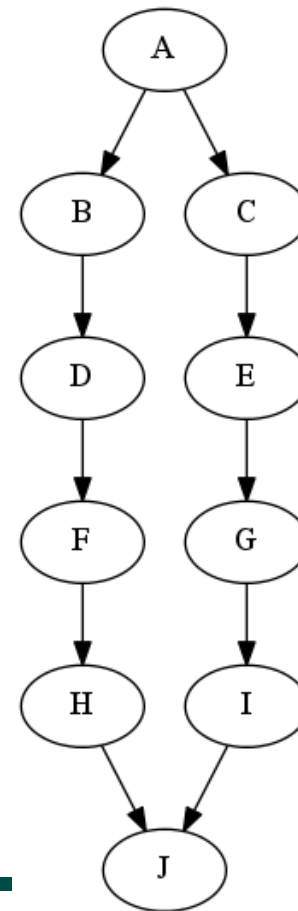
# Which Computation Graph has more ideal parallelism?

Assume that all nodes have  $\text{TIME} = 1$ , so  $\text{WORK} = 10$  for both graphs.

Computation Graph 1



Computation Graph 2



# Data Races

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A data race occurs on location  $L$  in a program execution with computation graph  $CG$  if there exist steps (nodes)  $S1$  and  $S2$  in  $CG$  such that:

1.  $S1$  does not depend on  $S2$  and  $S2$  does not depend on  $S1$ , i.e.,  $S1$  and  $S2$  can potentially execute in parallel, and
  2. Both  $S1$  and  $S2$  read or write  $L$ , and at least one of the accesses is a write.
- A data-race is an error. The result of a read operation in a data race is undefined. The result of a write operation is undefined if there are two or more writes to the same location.
  - Above definition includes all “potential” data races i.e., we consider it to be a data race even if  $S1$  and  $S2$  execute on the same processor.



# Reminders

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- **IMPORTANT:**
  - Send email to [comp322-staff@mailman.rice.edu](mailto:comp322-staff@mailman.rice.edu) if you did NOT receive a welcome email from us
  - Bring your laptop to this week's lab at 7pm TODAY (Section A01: DH 1064, Section A02: DH 1070)
  - Watch videos for topics 1.2 & 1.3 for next lecture on Wednesday
- Complete each week's assigned quizzes on edX by 11:59pm that Friday. This week, you should submit quizzes for lecture & demonstration videos for topics 1.1, 1.2, 1.3, 1.4
- HW1 will be assigned on Jan 16th and be due on Jan 28th
- See course web site for work assignments and due dates
  - <https://wiki.rice.edu/confluence/display/PARPROG/COMP322>

