



南方科技大学  
SOUTHERN UNIVERSITY OF SCIENCE AND TECHNOLOGY

# Algorithm Design and Analysis (H)

## CS216

Prof. Shiqi Yu (于仕琪)

yusq@sustech.edu.cn

<http://faculty.sustech.edu.cn/yusq/>



南方科技大学  
SOUTHERN UNIVERSITY OF SCIENCE AND TECHNOLOGY

# Greedy Algorithms



# Greedy algorithms

- Build up a solution in small steps,
- Choose a decision at each step myopically to optimize some underlying criterion.
- May not produce an optimal solution,
- But can yield locally optimal solutions that approximate a globally optimal solution in a reasonable amount of time.



南方科技大学  
SOUTHERN UNIVERSITY OF SCIENCE AND TECHNOLOGY

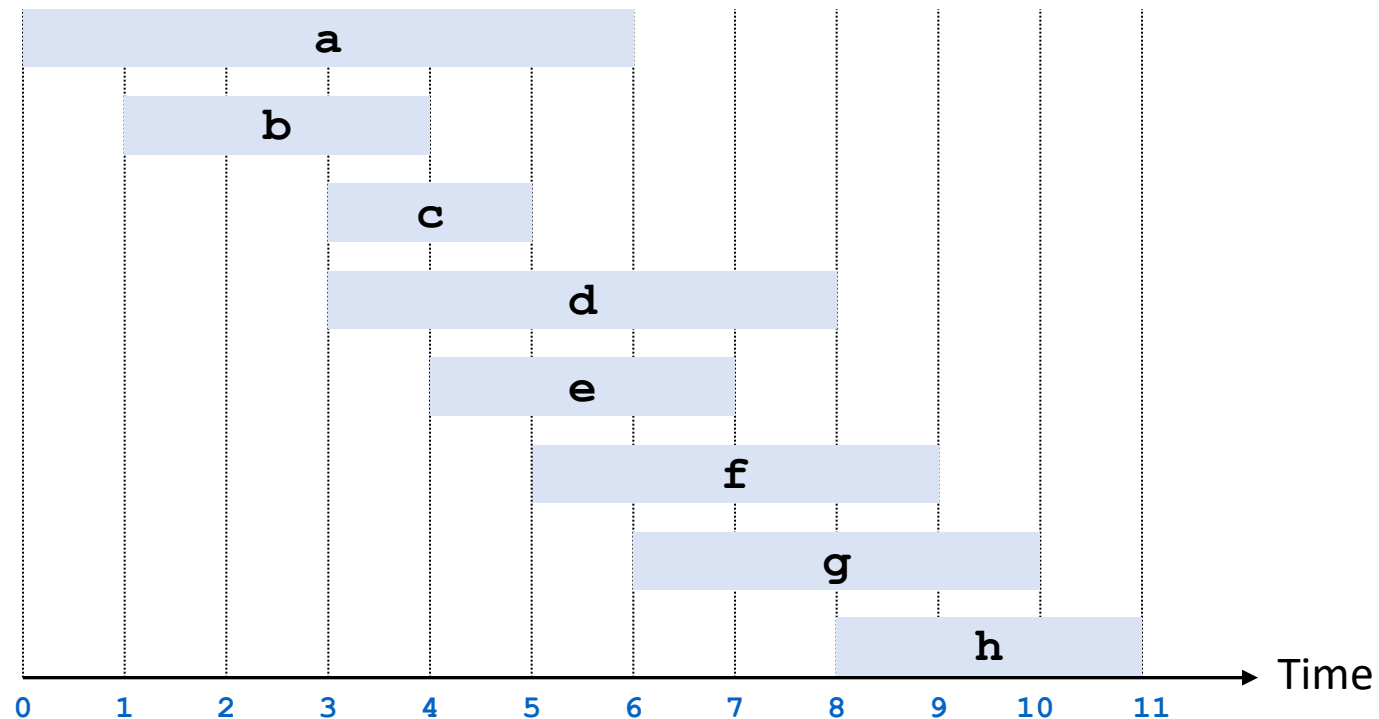
# 1. Interval Scheduling



# Interval Scheduling

- **Interval scheduling.**

- Job  $j$  starts at  $s_j$  and finishes at  $f_j$ .
- Two jobs **compatible** if they don't overlap.
- Goal: find maximum subset of mutually compatible jobs.





# Interval Scheduling: Greedy Algorithms

- **Greedy template.** Consider jobs in some natural order.  
Take each job provided it's compatible with the ones already taken.
  - **[Earliest start time]** Consider jobs in ascending order of  $s_j$ .
  - **[Earliest finish time]** Consider jobs in ascending order of  $f_j$ .
  - **[Shortest interval]** Consider jobs in ascending order of  $f_j - s_j$ .
  - **[Fewest conflicts]** For each job  $j$ , count the number of conflicting jobs  $c_j$ . Schedule in ascending order of  $c_j$ .



# Interval Scheduling: Greedy Algorithms

- **Greedy template.** Consider jobs in some natural order.  
Take each job provided it's compatible with the ones already taken.



counterexample for earliest start time



counterexample for shortest interval



counterexample for fewest conflicts



# Interval Scheduling: Greedy Algorithm

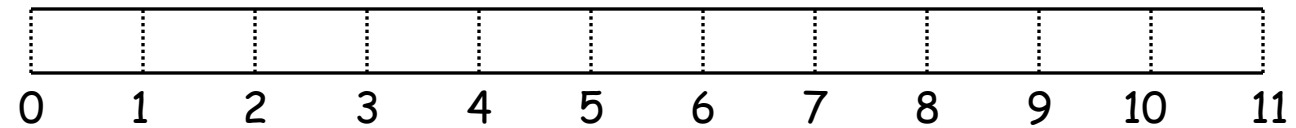
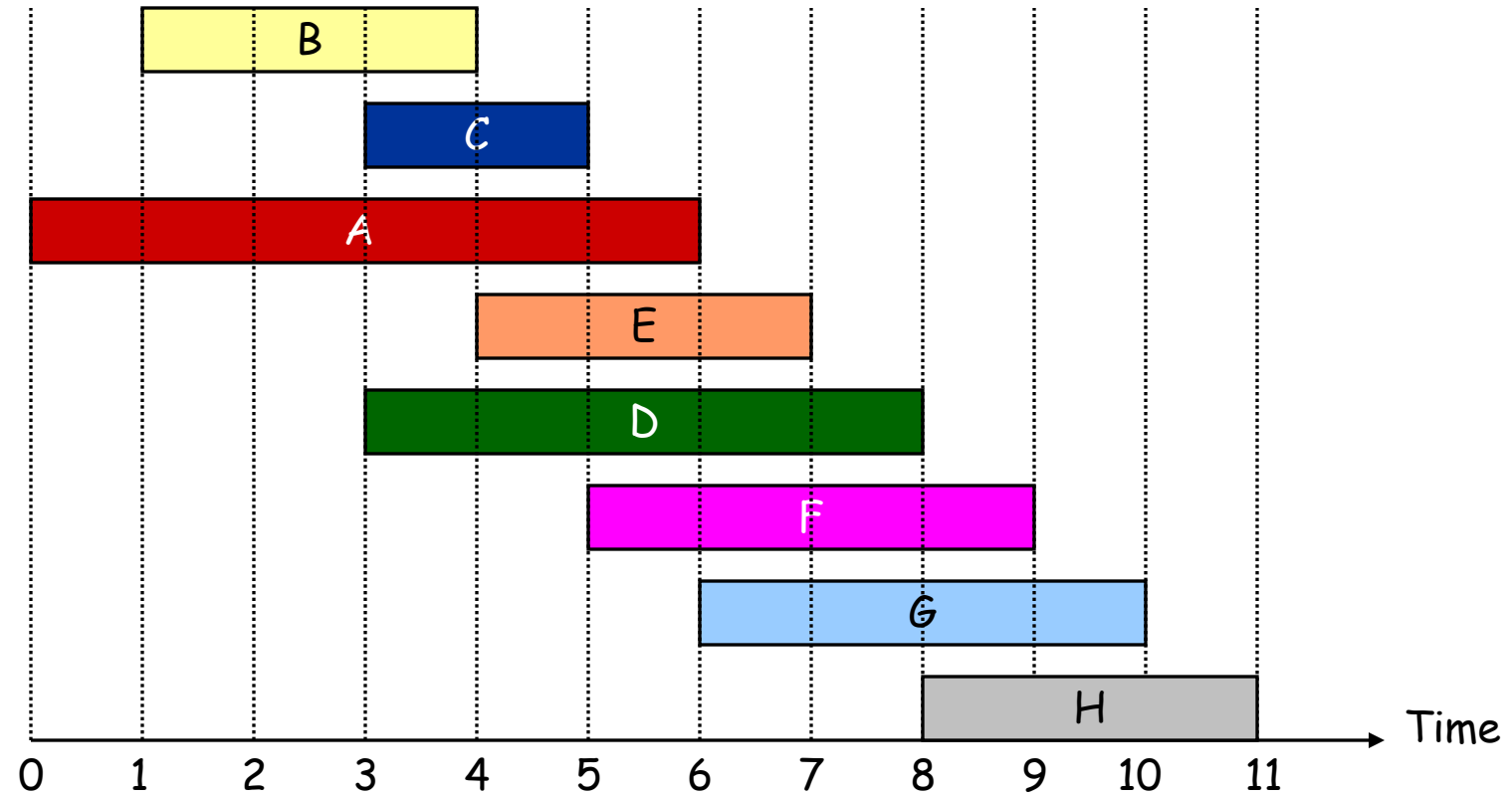
- **Greedy algorithm.** Consider jobs in increasing order of finish time. Take each job provided it's compatible with the ones already taken.

```
Sort jobs by finish times so that  $f_1 \leq f_2 \leq \dots \leq f_n$ .  
set of jobs selected  
A  $\leftarrow \phi$   
for j = 1 to n {  
    if (job j compatible with A)  
        A  $\leftarrow A \cup \{j\}$   
}  
return A
```

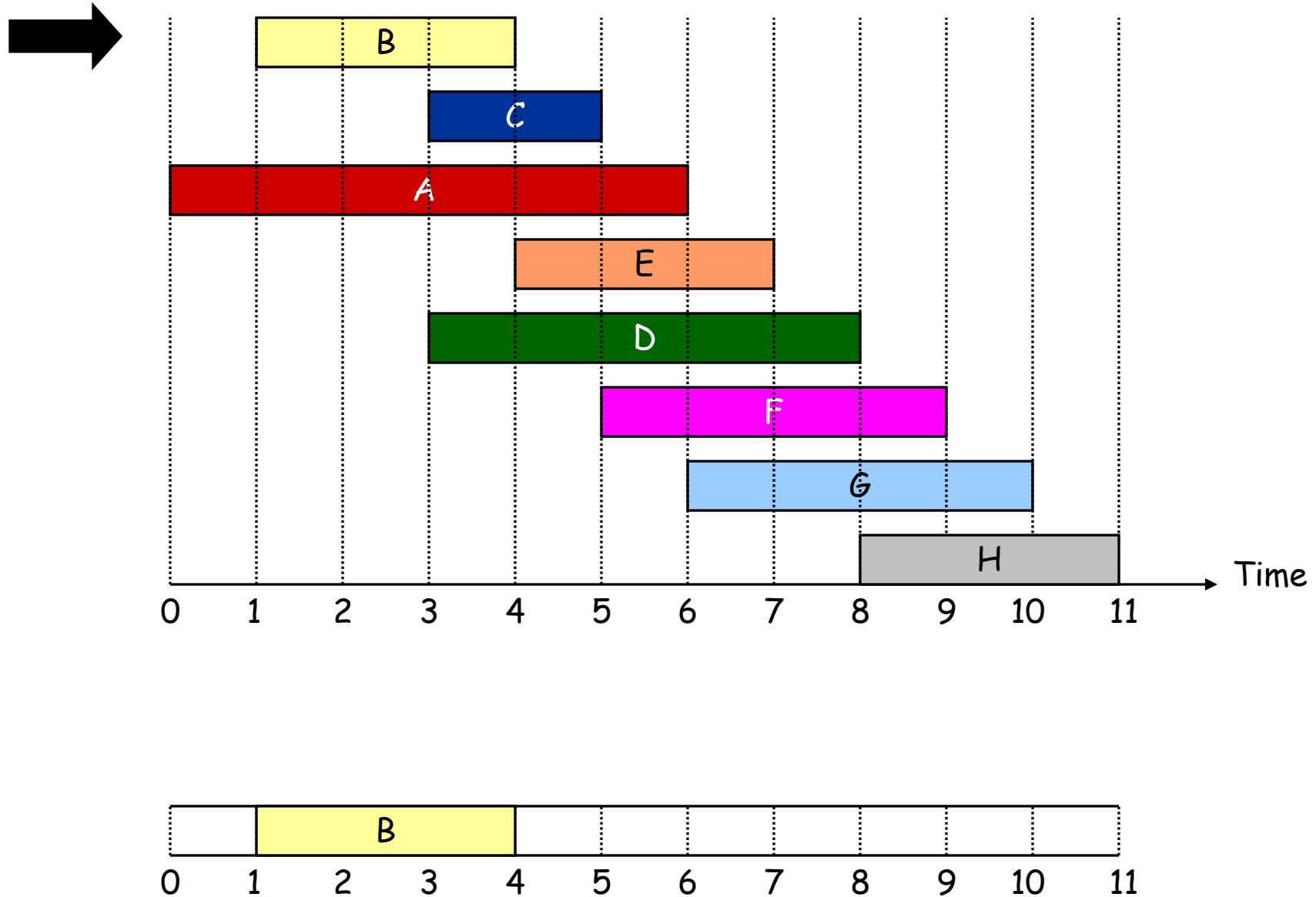
- Implementation.  $O(n \log n)$ .
  - Remember job  $j^*$  that was added last to A.
  - Job j is compatible with A if  $s_j \geq f_{j^*}$ .



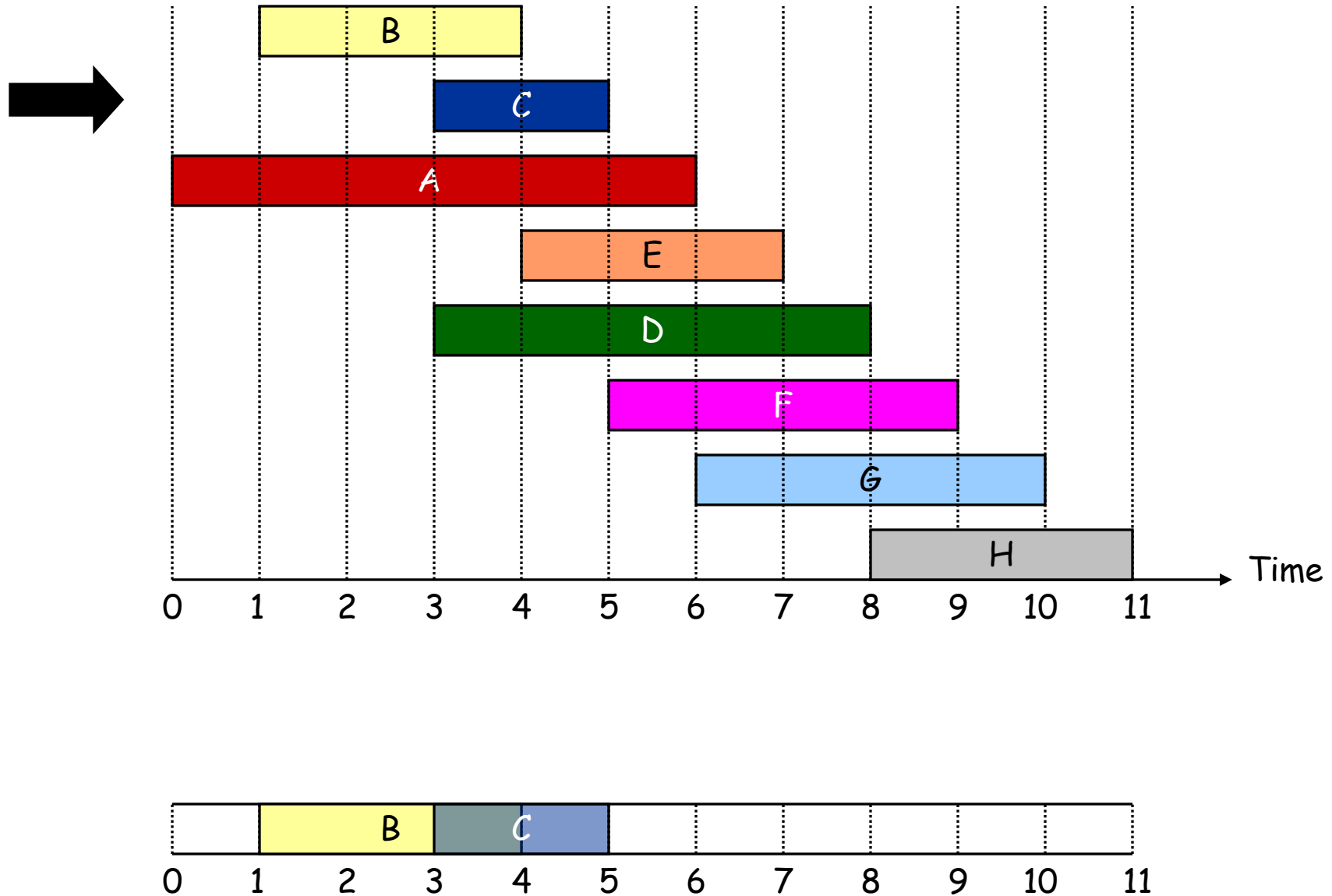
# Interval Scheduling



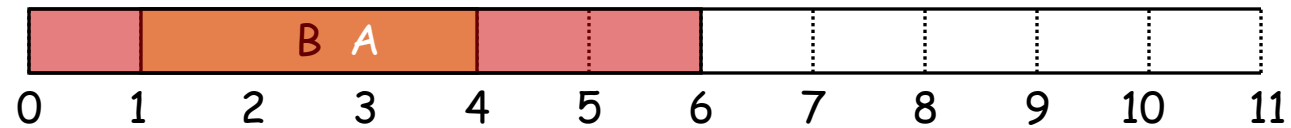
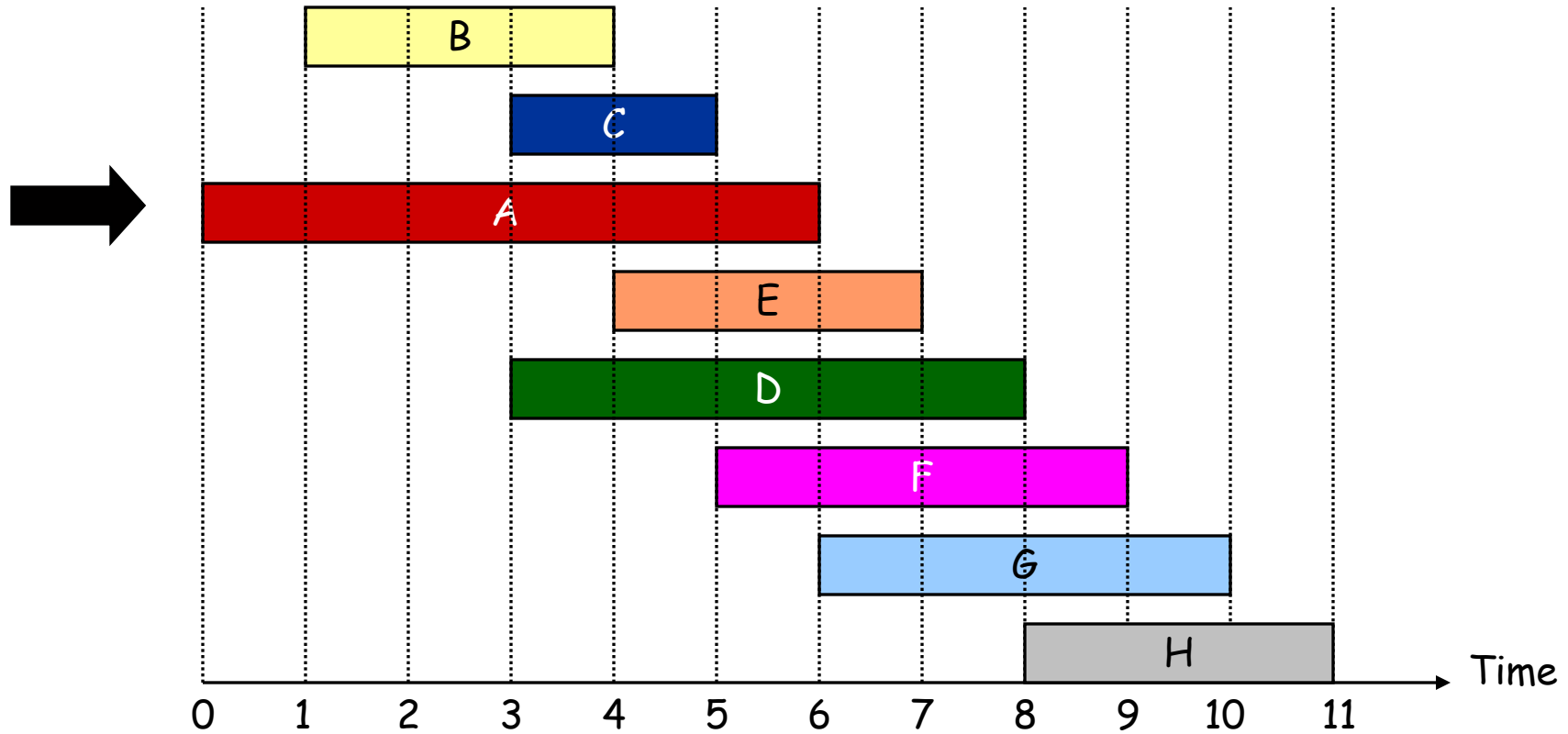
# Interval Scheduling



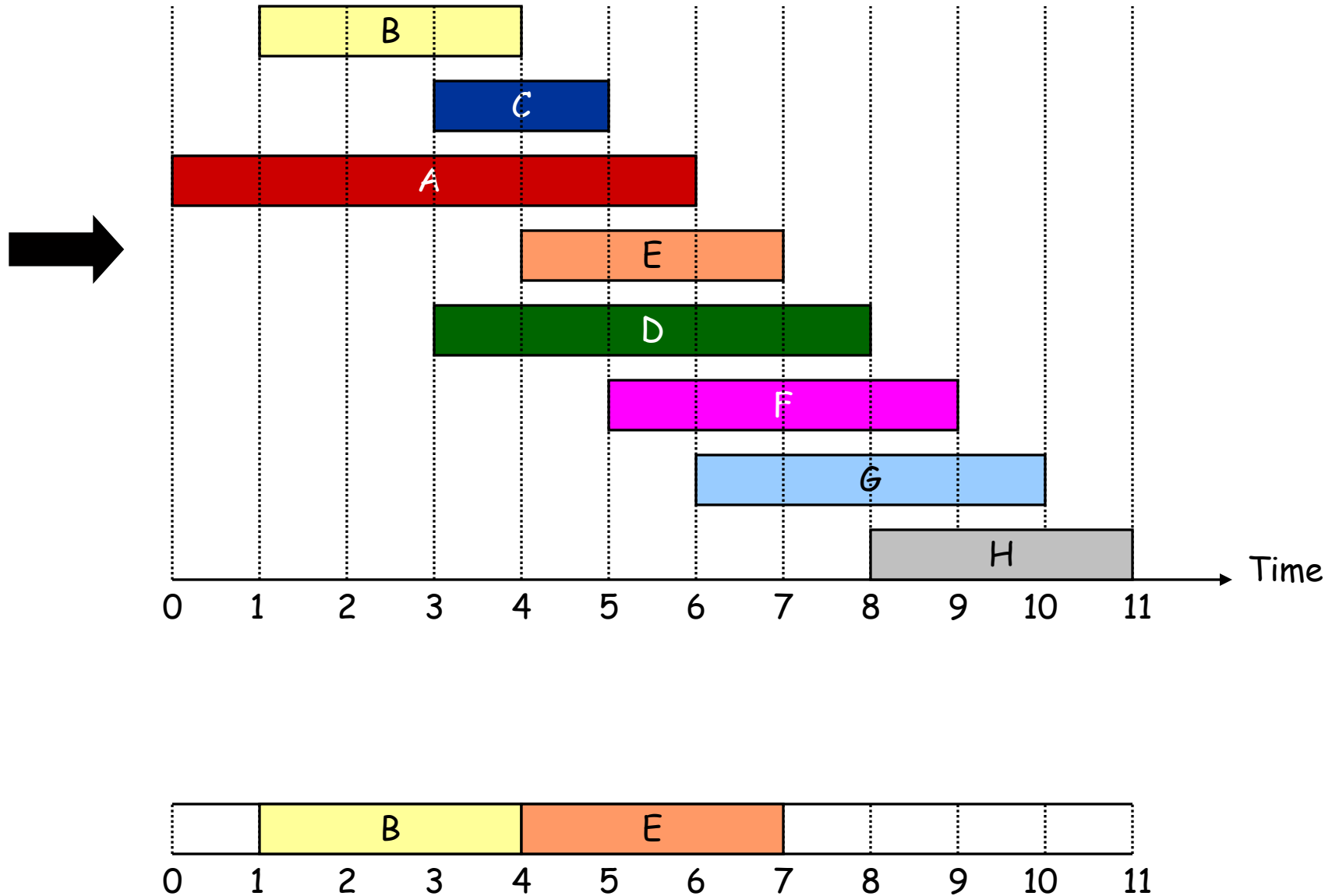
# Interval Scheduling



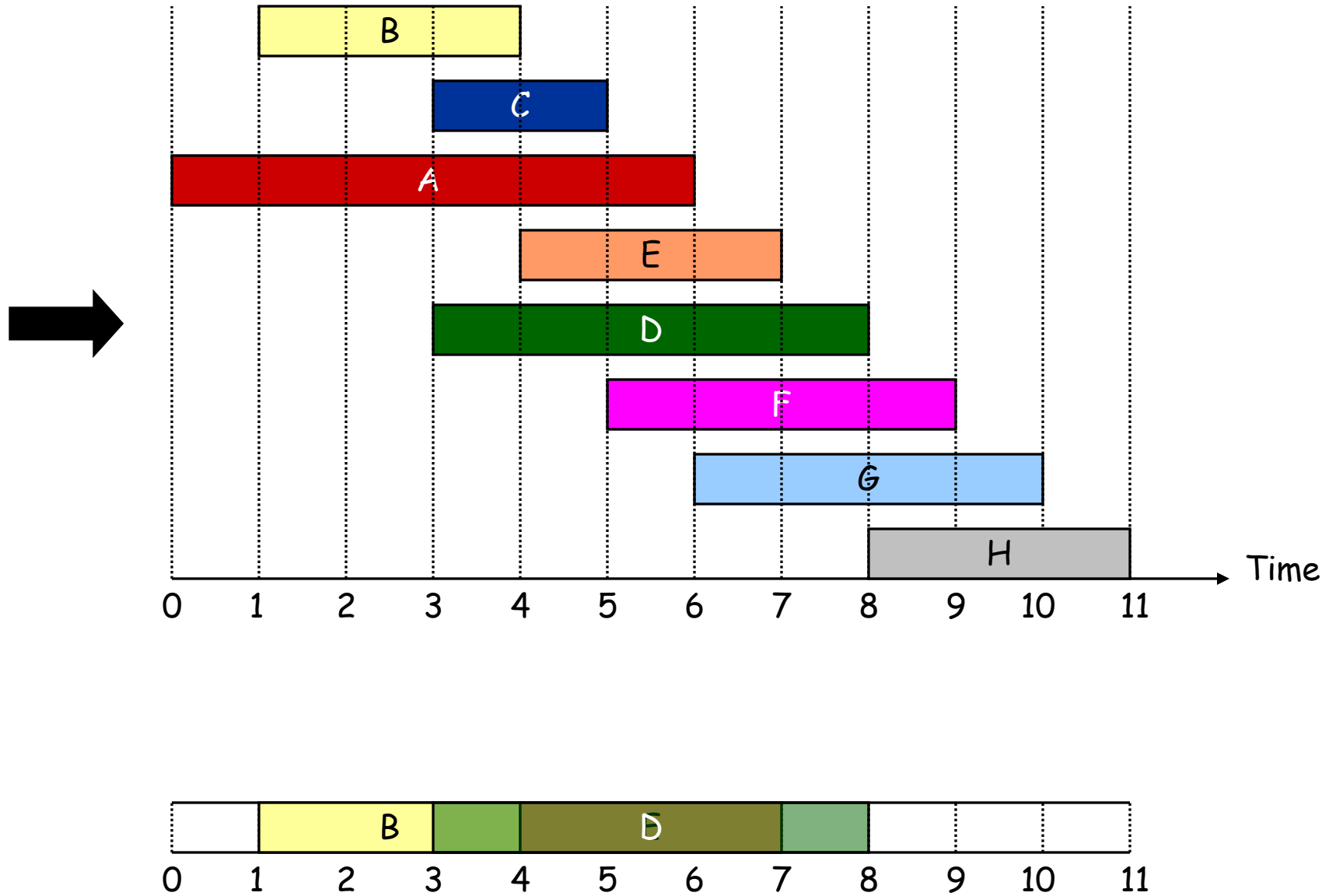
# Interval Scheduling



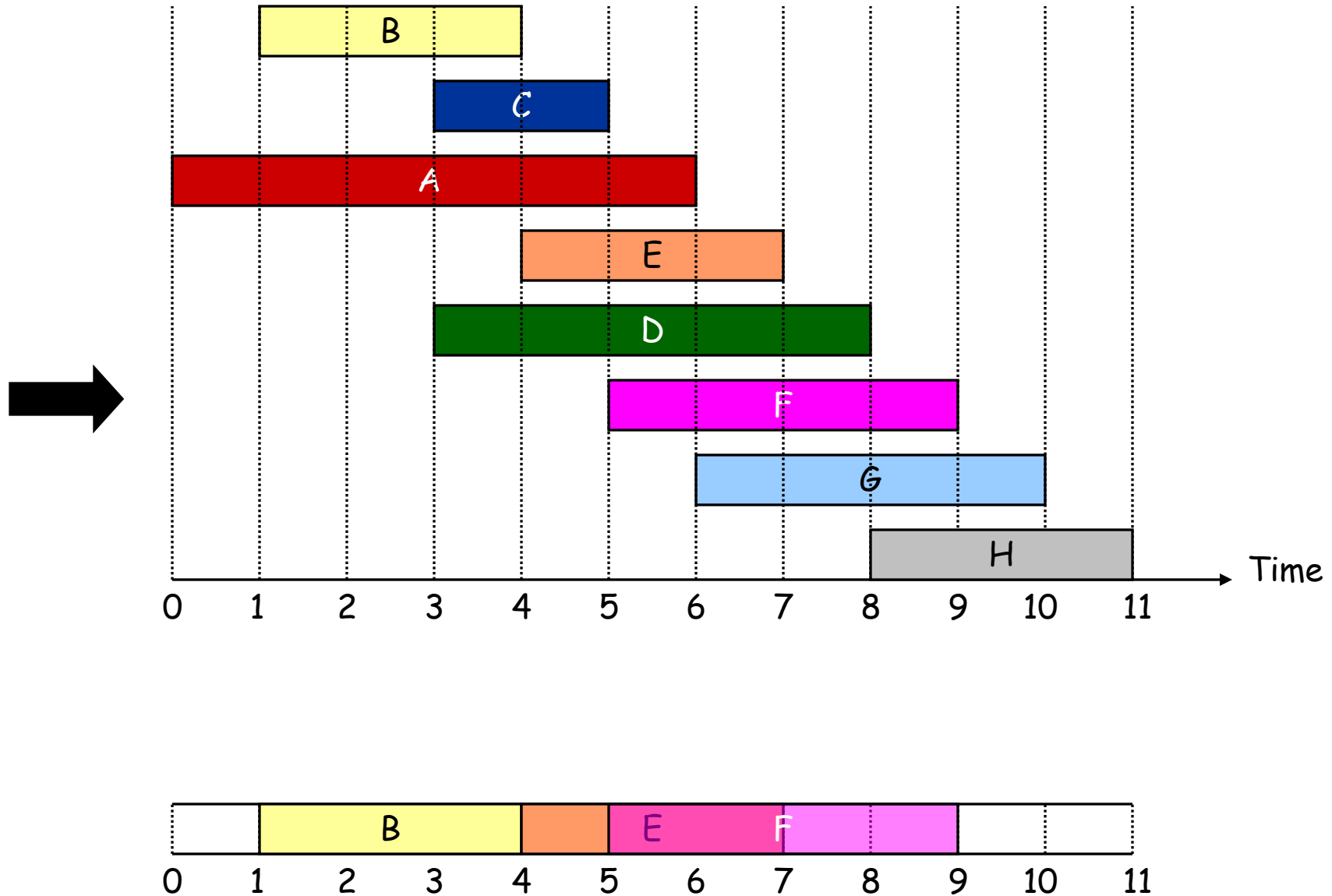
# Interval Scheduling



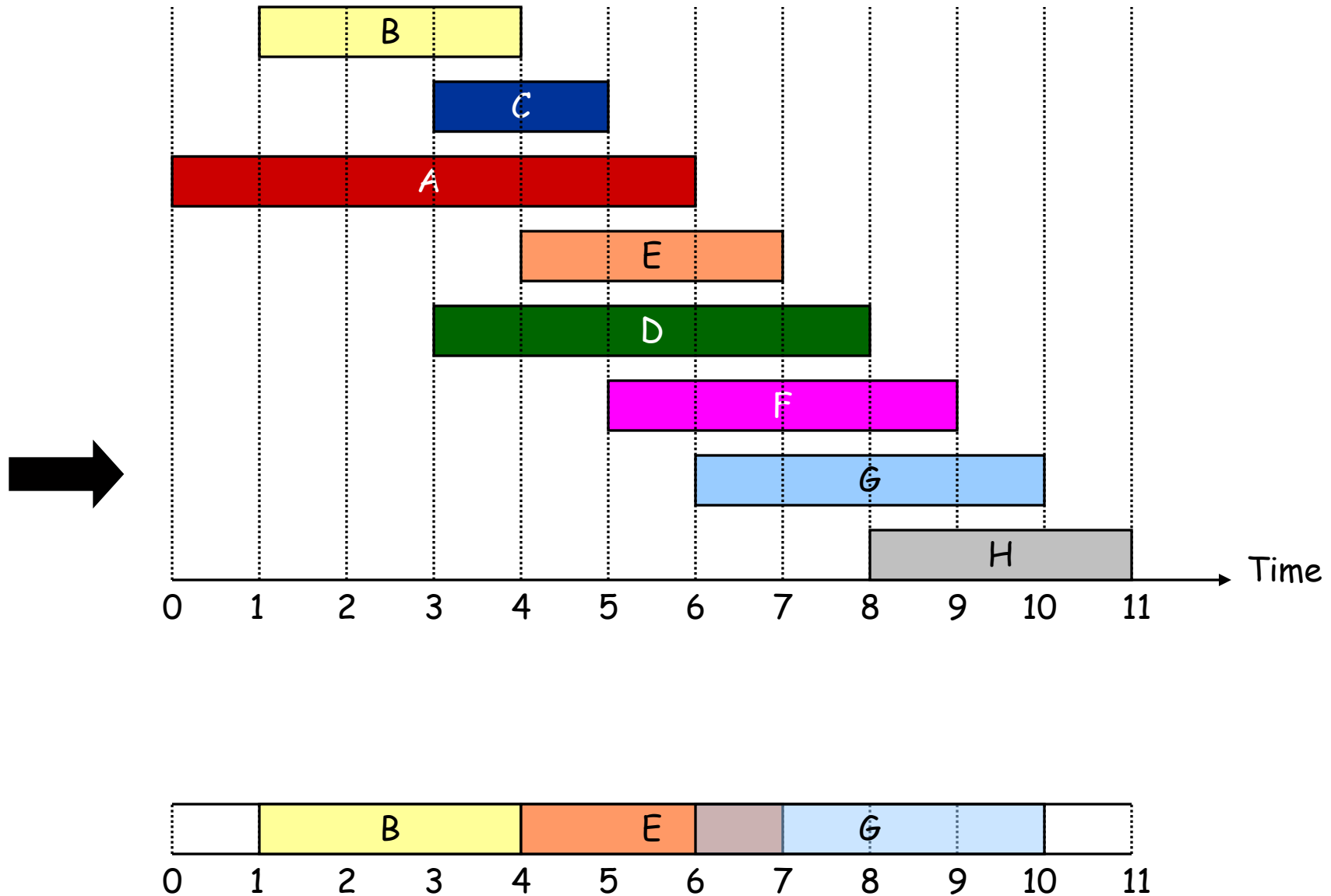
# Interval Scheduling



# Interval Scheduling

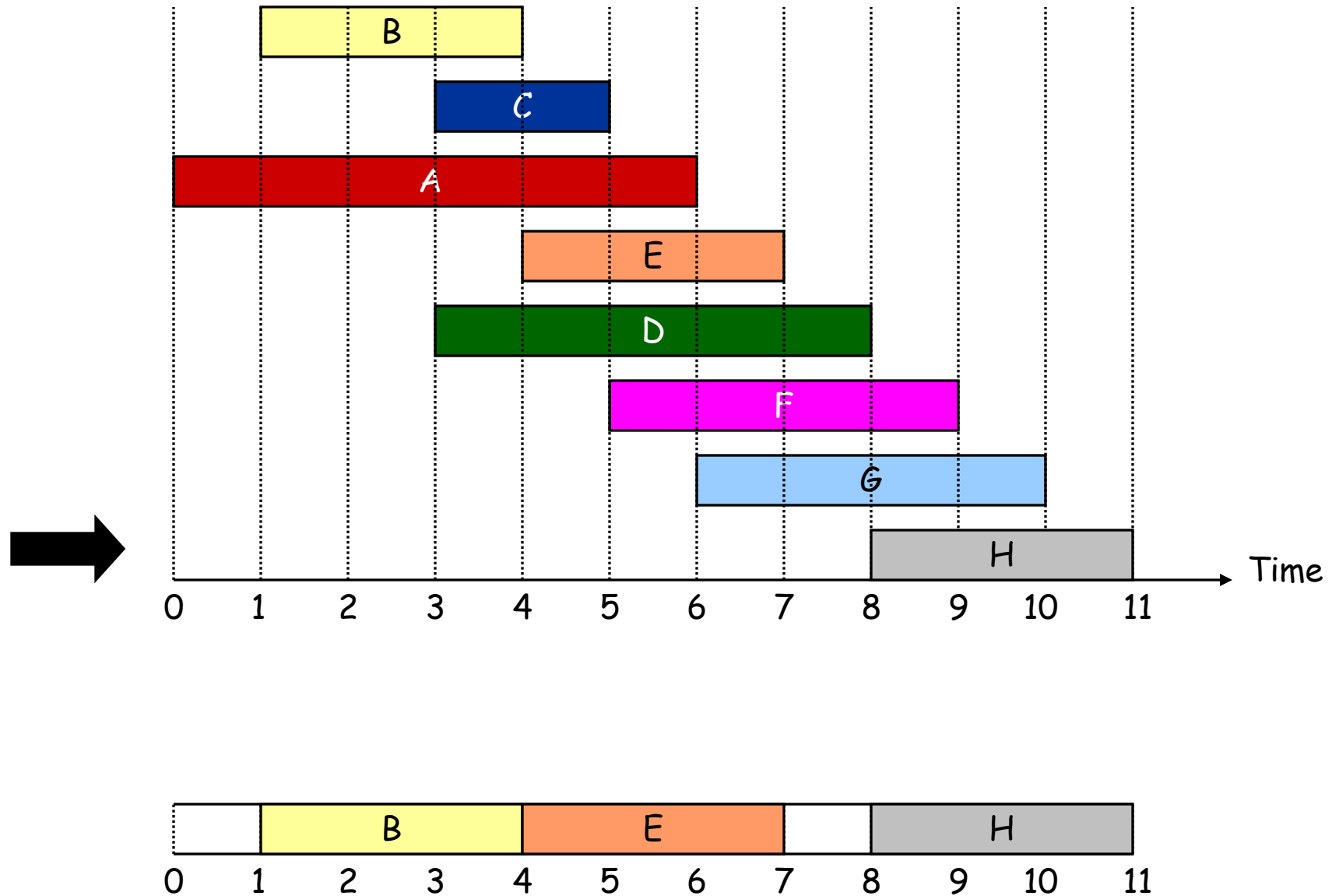


# Interval Scheduling





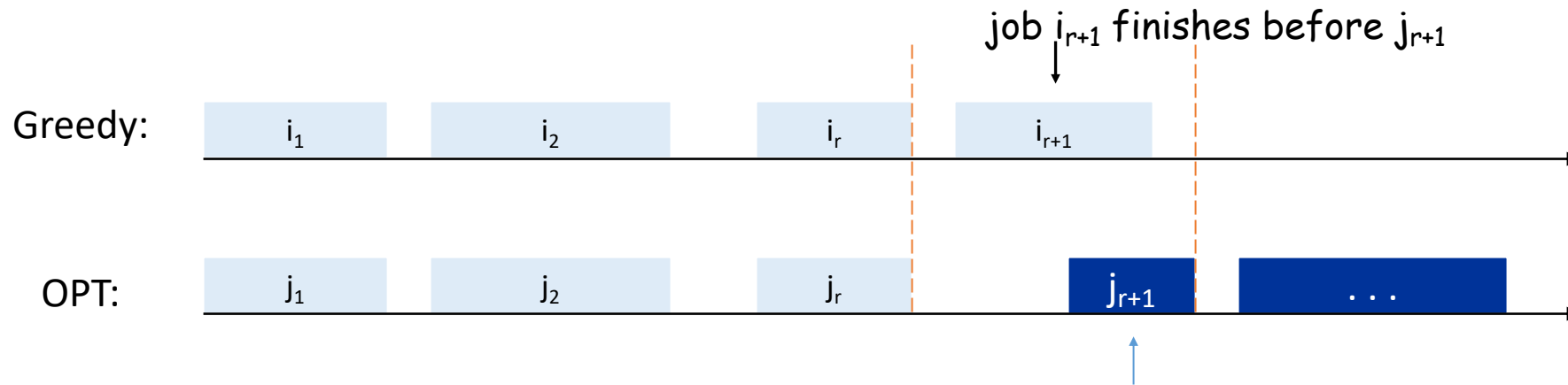
# Interval Scheduling





# Interval Scheduling: Analysis

- **Theorem.** Greedy algorithm is optimal.
- **Pf.** (by contradiction)
  - Assume greedy is not optimal, and let's see what happens.
  - Let  $i_1, i_2, \dots, i_k$  denote set of jobs selected by greedy.
  - Let  $j_1, j_2, \dots, j_m$  denote set of jobs in the optimal solution with  $i_1 = j_1, i_2 = j_2, \dots, i_r = j_r$  for the largest possible value of  $r$ .

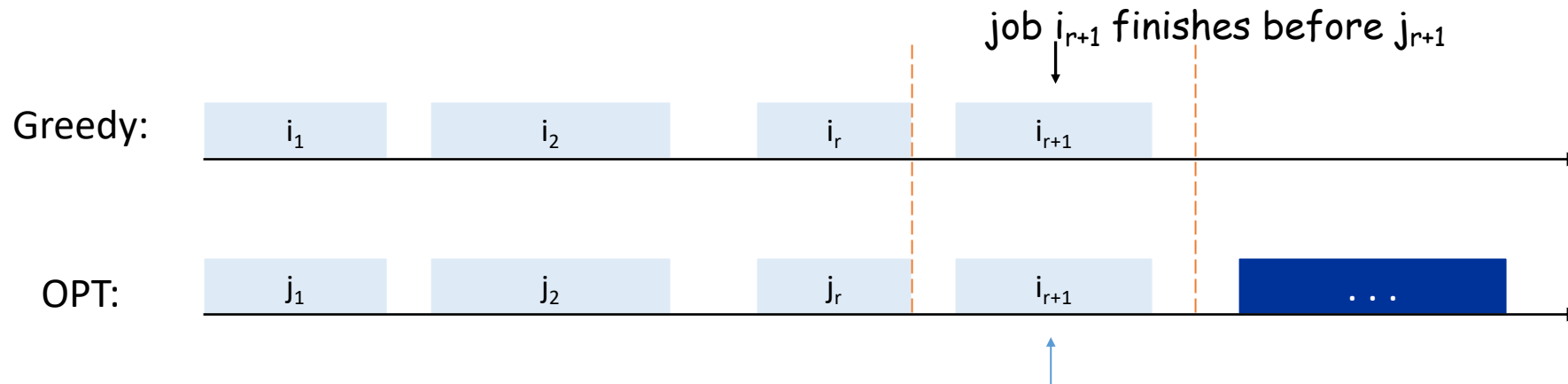


why not replace job  $j_{r+1}$   
with job  $i_{r+1}$ ?



# Interval Scheduling: Analysis

- **Theorem.** Greedy algorithm is optimal.
- **Pf.** (by contradiction)
  - Assume greedy is not optimal, and let's see what happens.
  - Let  $i_1, i_2, \dots, i_k$  denote set of jobs selected by greedy.
  - Let  $j_1, j_2, \dots, j_m$  denote set of jobs in the optimal solution with  $i_1 = j_1, i_2 = j_2, \dots, i_r = j_r$  for the largest possible value of  $r$ .



solution still feasible and optimal, but contradicts maximality of  $r$ .



南方科技大学  
SOUTHERN UNIVERSITY OF SCIENCE AND TECHNOLOGY

## 2. Interval Partitioning

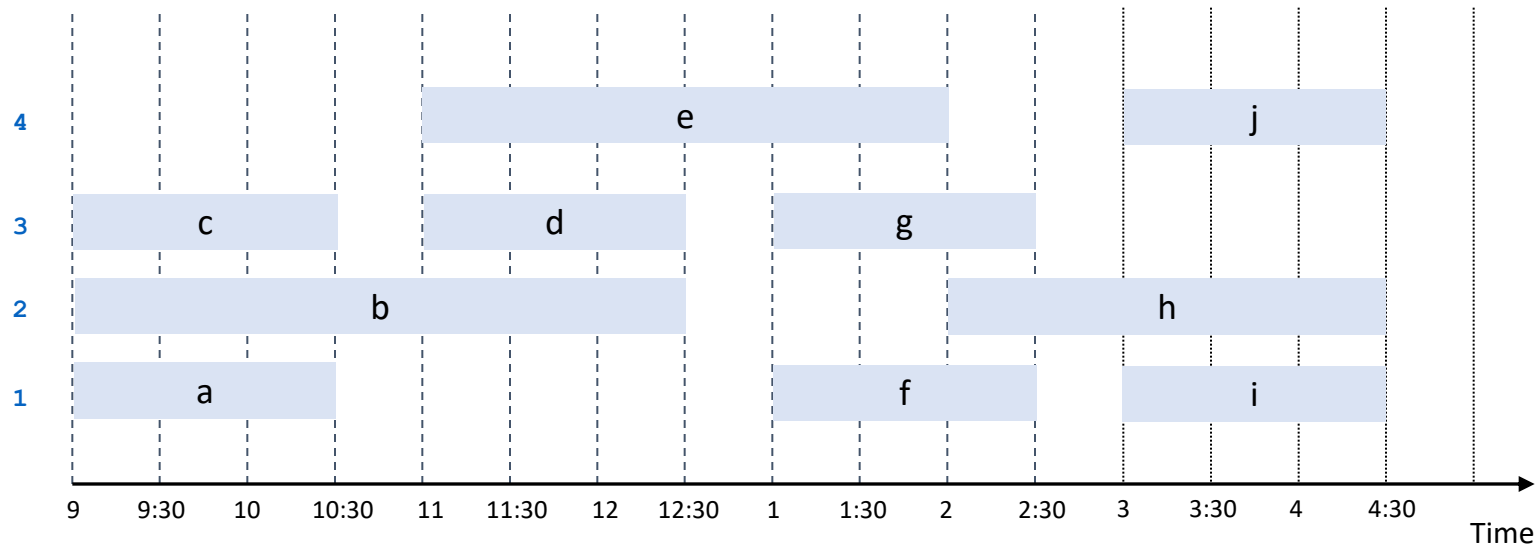


# Interval Partitioning

- Interval partitioning.

- Lecture  $j$  starts at  $s_j$  and finishes at  $f_j$ .
- Goal: find minimum number of classrooms to schedule all lectures so that no two occur at the same time in the same room.

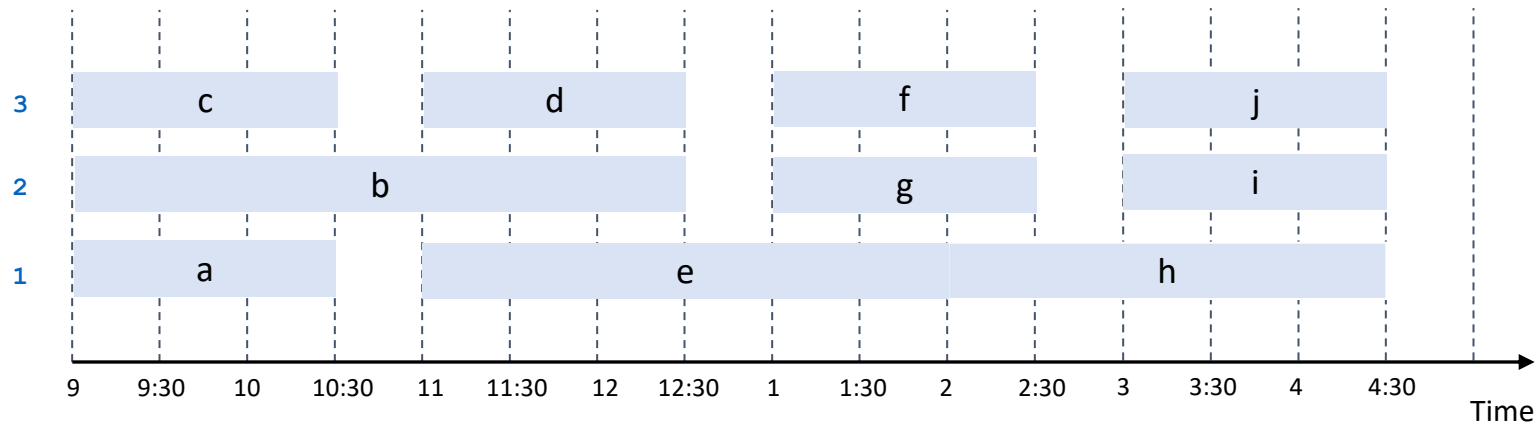
- Ex: This schedule uses 4 classrooms to schedule 10 lectures.





# Interval Partitioning

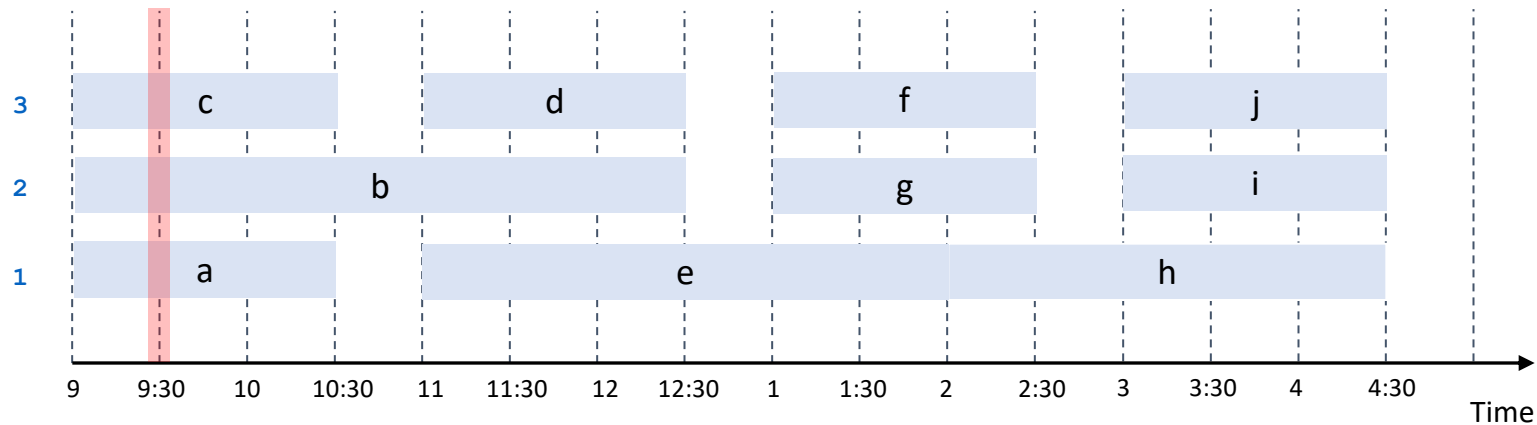
- Interval partitioning.
  - Lecture  $j$  starts at  $s_j$  and finishes at  $f_j$ .
  - Goal: find minimum number of classrooms to schedule all lectures so that no two occur at the same time in the same room.
- Ex: This schedule uses only 3.





# Interval Partitioning: Lower Bound on Optimal Solution

- **Def.** The **depth** of a set of open intervals is the maximum number that contain any given time.
- **Key observation.** Number of classrooms needed  $\geq$  depth.
- **Ex:** Depth of schedule below = 3  $\Rightarrow$  schedule below is optimal.  
a, b, c all contain 9:30
- **Q.** Does there always exist a schedule equal to depth of intervals?





# Interval Partitioning: Greedy Algorithm

- **Greedy algorithm.** Consider lectures in increasing order of start time: assign lecture to any compatible classroom.

```
Sort intervals by starting time so that  $s_1 \leq s_2 \leq \dots \leq s_n$ .  
d  $\leftarrow$  0 ← number of allocated classrooms  
  
for j = 1 to n {  
    if (lecture j is compatible with some classroom k)  
        schedule lecture j in classroom k  
    else  
        allocate a new classroom d + 1  
        schedule lecture j in classroom d + 1  
        d  $\leftarrow$  d + 1  
}
```

- **Implementation.**  $O(n \log n)$ .
  - For each classroom k, maintain the finish time of the last job added.
  - Keep the classrooms in a priority queue.





# Interval Partitioning: Greedy Analysis

- **Observation.** Greedy algorithm never schedules two incompatible lectures in the same classroom.
- **Theorem.** Greedy algorithm is optimal.
- **Pf.**
  - Let  $d$  = number of classrooms that the greedy algorithm allocates.
  - Classroom  $d$  is opened because we needed to schedule a job, say  $j$ , that is incompatible with all  $d-1$  other classrooms.
  - These  $d$  jobs each end after  $s_j$ .
  - Since we sorted by start time, all these incompatibilities are caused by lectures that start no later than  $s_j$ .
  - Thus, we have  $d$  lectures overlapping at time  $s_j + \epsilon$ .
  - Key observation  $\Rightarrow$  all schedules use  $\geq d$  classrooms. ■



# 3. Scheduling to Minimize Lateness



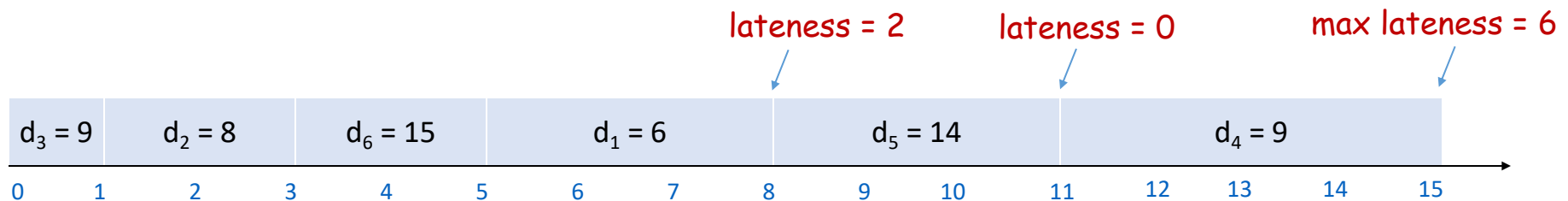
# Scheduling to Minimizing Lateness

- Minimizing lateness problem.

- Single resource processes one job at a time.
- Job  $j$  requires  $t_j$  units of processing time and is due at time  $d_j$ .
- If  $j$  starts at time  $s_j$ , it finishes at time  $f_j = s_j + t_j$ .
- Lateness:  $\ell_j = \max \{ 0, f_j - d_j \}$ .
- Goal: schedule all jobs to minimize **maximum** lateness  $L = \max \ell_j$ .

- Ex:

	1	2	3	4	5	6
$t_j$	3	2	1	4	3	2
$d_j$	6	8	9	9	14	15





# Minimizing Lateness: Greedy Algorithms

- Greedy template. Consider jobs in some order.
  - [Shortest processing time first] Consider jobs in ascending order of processing time  $t_j$ .
  - [Earliest deadline first] Consider jobs in ascending order of deadline  $d_j$ .
  - [Smallest slack] Consider jobs in ascending order of slack  $d_j - t_j$ .



# Minimizing Lateness: Greedy Algorithms

- Greedy template. Consider jobs in some order.

- [Shortest processing time first] Consider jobs in ascending order of processing time  $t_j$ .

	1	2
$t_j$	1	10
$d_j$	100	10

counterexample

- [Smallest slack] Consider jobs in ascending order of slack  $d_j - t_j$ .

	1	2
$t_j$	1	10
$d_j$	2	10

counterexample



# Minimizing Lateness: Greedy Algorithm

- Greedy algorithm. Earliest deadline first.

```
Sort n jobs by deadline so that  $d_1 \leq d_2 \leq \dots \leq d_n$ 
```

```
t  $\leftarrow$  0
```

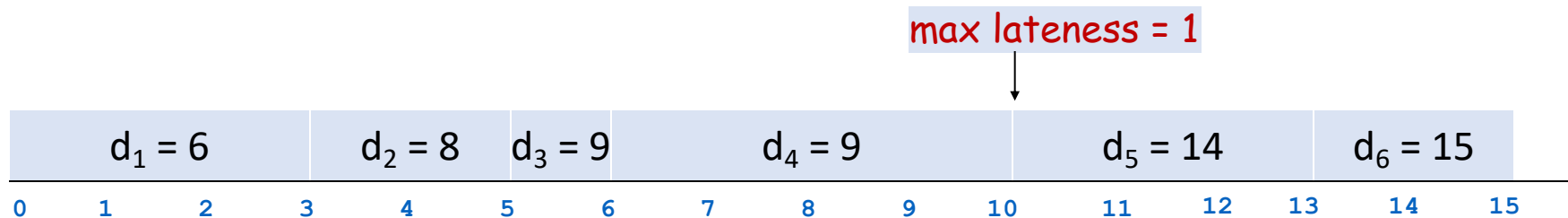
```
for j = 1 to n
```

```
    Assign job j to interval [t, t + tj]
```

```
    sj  $\leftarrow$  t, fj  $\leftarrow$  t + tj
```

```
    t  $\leftarrow$  t + tj
```

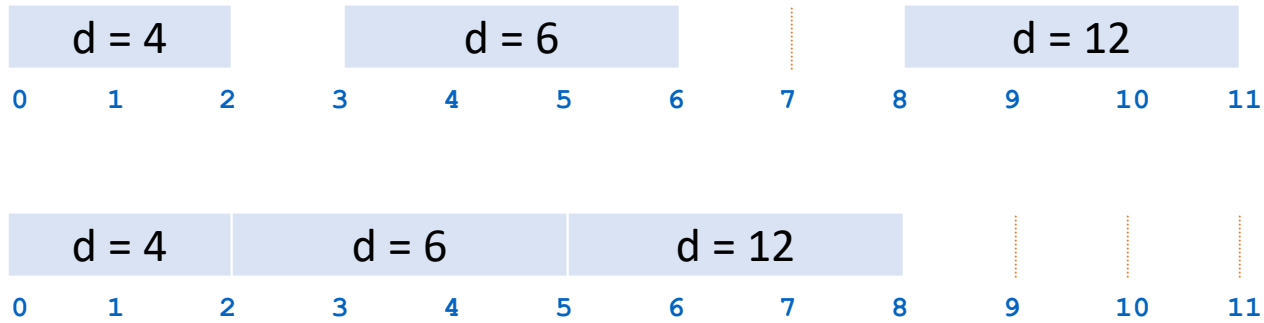
```
output intervals [sj, fj]
```





# Minimizing Lateness: No Idle Time

- **Observation.** There exists an optimal schedule with no **idle time**.

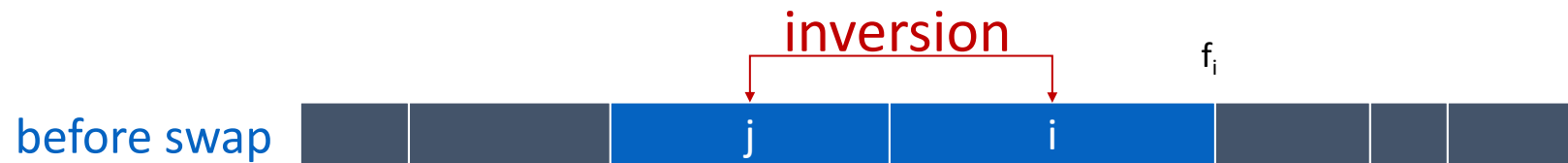


- **Observation.** The greedy schedule has no idle time.



# Minimizing Lateness: Inversions

- **Def.** Given a schedule  $S$ , an **inversion** is a pair of jobs  $i$  and  $j$  such that:  $d_i < d_j$  but  $j$  scheduled before  $i$ .



[ as before, we assume jobs are numbered so that  $d_1 \leq d_2 \leq \dots \leq d_n$  ]

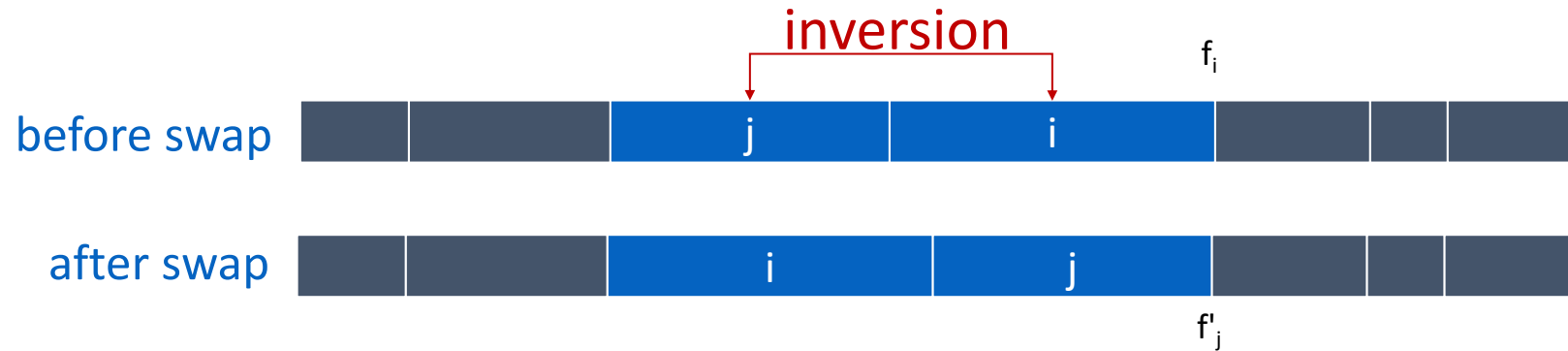
- **Observation.** Greedy schedule has no inversions.
- **Observation.** If a schedule (with no idle time) has an inversion, it has one with a pair of inverted jobs scheduled consecutively.





# Minimizing Lateness: Inversions

- **Def.** Given a schedule  $S$ , an **inversion** is a pair of jobs  $i$  and  $j$  such that:  $d_i < d_j$  but  $j$  scheduled before  $i$ .



- **Claim.** Swapping two consecutive, inverted jobs reduces the number of inversions by one and does not increase the max lateness.
- **Pf.** Let  $\ell$  be the lateness before the swap, and let  $\ell'$  be it afterwards.
  - $\ell'_k = \ell_k$  for all  $k \neq i, j$
  - $\ell'_i \leq \ell_i$
  - $\ell'_j = f'_j - d_j = f_i - d_j < f_i - d_i = \ell_i$





# Minimizing Lateness: Analysis of Greedy Algorithm

- **Theorem.** Greedy schedule  $S$  is optimal.
- **Pf.** Define  $S^*$  to be an optimal schedule that has the fewest number of inversions, and let's see what happens.
  - Can assume  $S^*$  has no idle time.
  - If  $S^*$  has no inversions, then  $S = S^*$ .
  - If  $S^*$  has an inversion, let  $i$ - $j$  be an adjacent inversion.
    - ✓ swapping  $i$  and  $j$  does not increase the maximum lateness and strictly decreases the number of inversions
    - ✓ this contradicts definition of  $S^*$  ■



# Greedy Analysis Strategies

- **Greedy algorithm stays ahead.** Show that after each step of the greedy algorithm, its solution is at least as good as any other algorithm's.
- **Structural.** Discover a simple "structural" bound asserting that every possible solution must have a certain value. Then show that your algorithm always achieves this bound.
- **Exchange argument.** Gradually transform any solution to the one found by the greedy algorithm without hurting its quality.
- **Other greedy algorithms.** Kruskal, Prim, Dijkstra, Huffman, ...