

Summary of Lecture 1

- Euler Theorem
- Algorithm for Euler Circuit.
- Basic techniques for designing algorithms:
greedy, divide-and-conquer, dynamic programming

Week 2: Greedy Algorithms



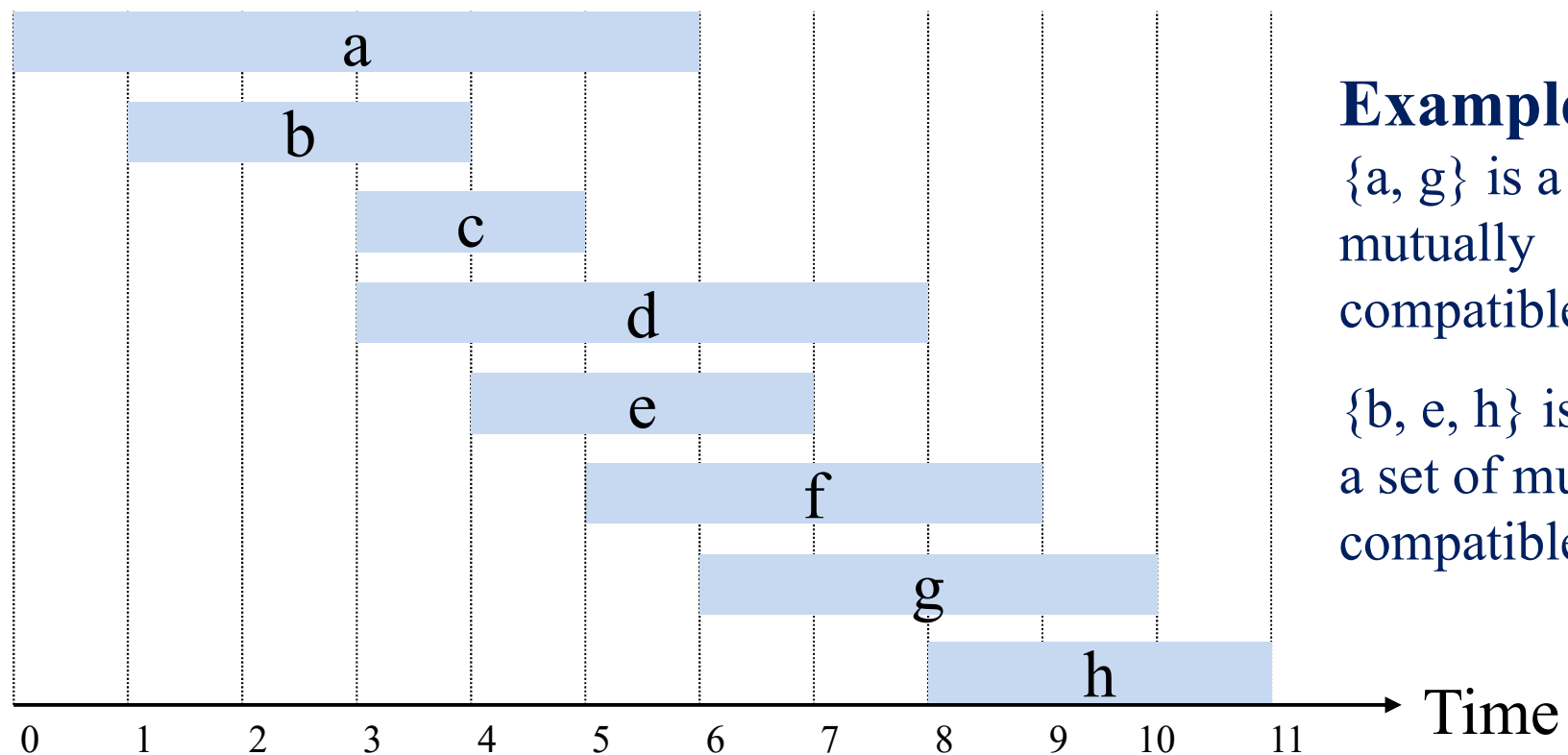
Greedy Algorithm

- A technique to solve problems:
 - always makes the **locally best** choice at the moment (local optimal).
 - Hopefully, a series of locally best choices will lead to a globally best solution.
- Greedy algorithms yield optimal solutions for many (but not all) problems.

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.



Examples:

$\{a, g\}$ is a set of mutually compatible jobs.

$\{b, e, h\}$ is also a set of mutually compatible jobs.

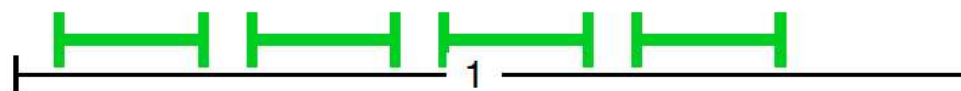
Ideas for Interval Scheduling

Greedy Algorithm: sort all the jobs in a list using a 'greedy' principle, and then choose it one by one

- What are possible rules for greedy sorting?
 - Choose the interval that starts earliest.
 - Rationale: start using the resource as soon as possible.
 - Choose the smallest interval.
 - Rationale: try to have lots of small jobs.
 - Choose the interval that overlaps (conflicts) with the fewest remaining intervals.
 - Rationale: keep our options open and eliminate as few intervals as possible.

Rules That Don't Work

① Earliest start time

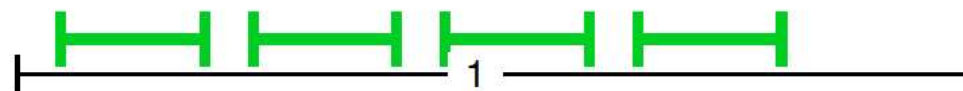


② Shortest job

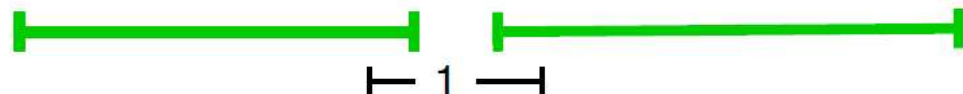
③ Fewest conflicts

Rules That Don't Work

① Earliest start time



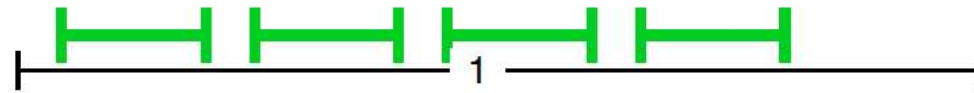
② Shortest job



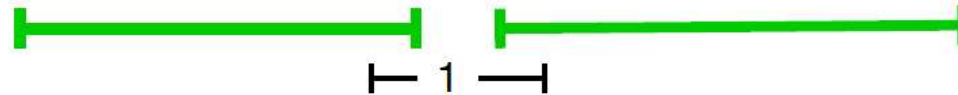
③ Fewest conflicts

Rules That Don't Work

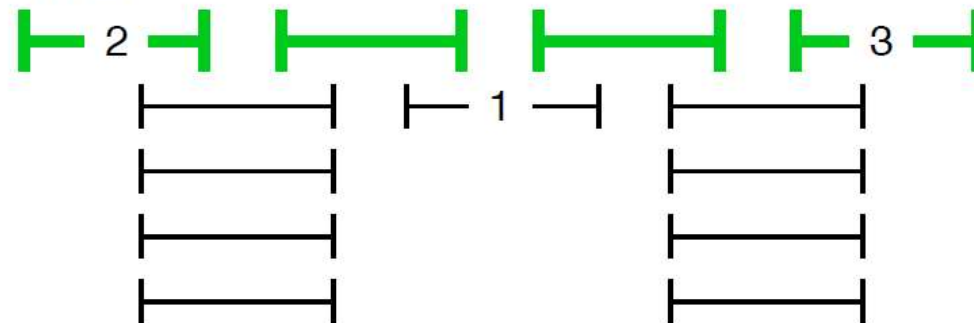
① Earliest start time



② Shortest job



③ Fewest conflicts



Example: sort based on finish time

- Jobs (s, f) : $(0, 10), (3, 4), (2, 8), (1, 5), (4, 5), (4, 8), (5, 6), (7, 9)$.
- Sorting based on f_i :
 - $(3, 4), (1, 5), (4, 5), (5, 6), (4, 8), (2, 8), (7, 9), (0, 10)$.
- Selecting jobs:
 - $(3, 4),$
 - $(4, 5),$
 - $(5, 6),$
 - $(7, 9),$

Interval Scheduling: Greedy Algorithm

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

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

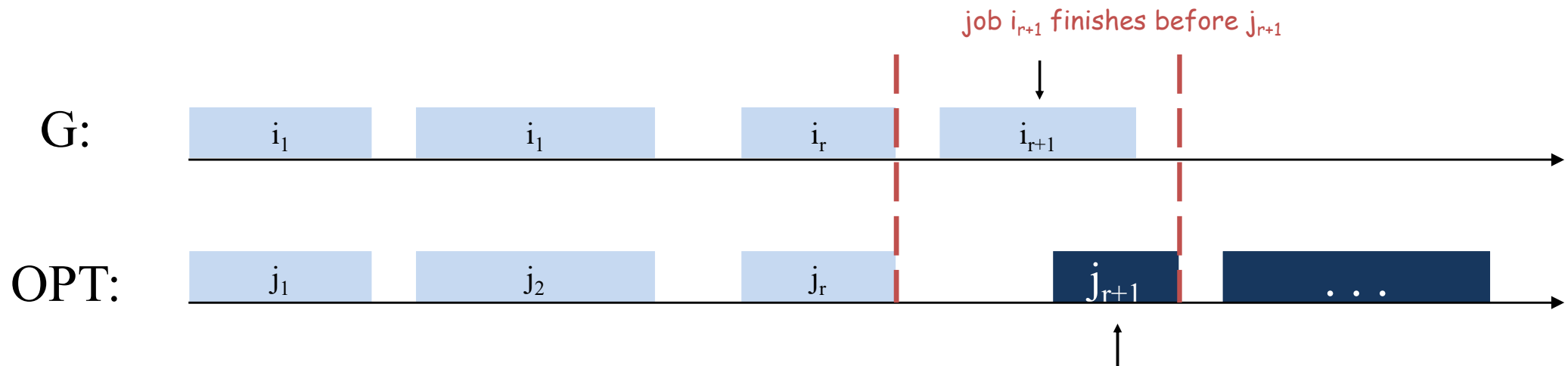
- Implementation. $O(n \log n)$.
 - Remember job j^* that was added last to A.
 - Job j is compatible with A if $s_j \geq f_{j^*}$.
- } how to decide if job j is compatible with A

How to Prove Optimality

- How can we prove the schedule returned is optimal?
 - Let A be the schedule returned by this algorithm.
 - Let OPT be some optimal solution (there may be many optimal solutions!).
- Might be hard to show that $A = OPT$, instead we need only to show that $|A| = |OPT|$ or **equivalently A is one of the optimal solutions.**
- **Note the distinction:** instead of proving directly that a choice of intervals A is the same as an optimal choice, we prove that it has the same number of intervals as an optimal. Therefore, it is optimal.

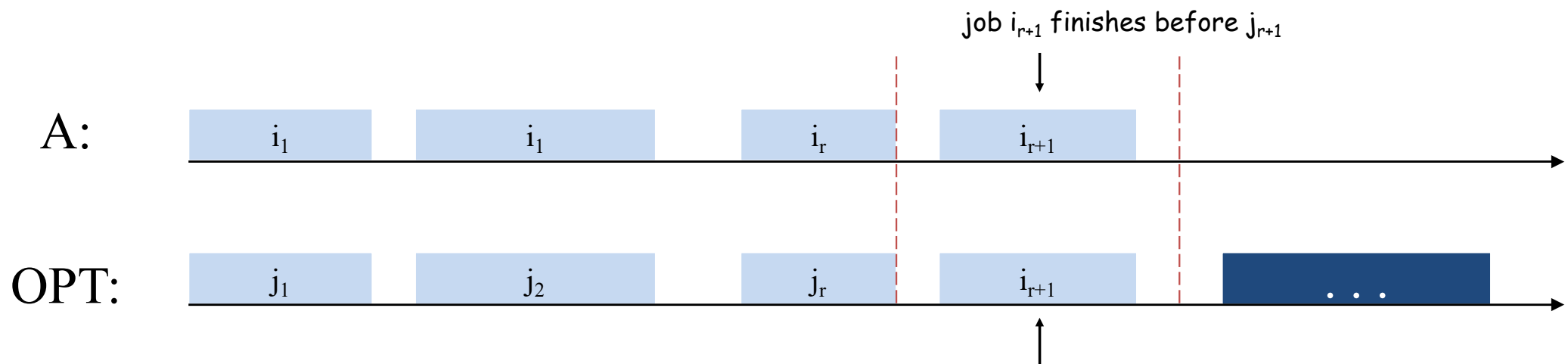
Interval Scheduling: Analysis

- Theorem. Greedy algorithm is optimal.
- Proof:
 - We compare the solution obtained from greedy algorithm with an optimal solution.
 - Let $G = i_1, i_2, \dots, i_k$ denote the set of jobs selected by greedy.
 - Let $\text{Opt} = j_1, j_2, \dots, j_n$ denote the set of jobs in the optimal solution.
 - The set of jobs are mutually compatible and the number of jobs is the largest.
 - Without loss of generality, we assume that
 - $i_1 = j_1, i_2 = j_2, \dots, i_r = j_r$ and $i_{r+1} \neq j_{r+1}$, where r could be 0, 1, 2,
 - Job i_{r+1} finishes before (or at the same time of) j_{r+1} due to our greedy algorithm.



Interval Scheduling: Analysis

- Theorem. Greedy algorithm is optimal.
- Pf.
 - Let $A = i_1, i_2, \dots, i_k$ denote set of jobs selected by greedy.
 - Let $\text{Opt} = j_1, j_2, \dots, j_m$ denote set of jobs in the optimal solution with $i_1 = j_1, i_2 = j_2, \dots, i_r = j_r$ and $i_{r+1} \neq j_{r+1}$ (Note that r can be 0, 1, 2, ...)



Another Proof for Interval Scheduling:

Let: $G = i_1 i_2 \dots i_m$ be the solution return by greedy

Let : $Opt = j_1 j_2 \dots j_k$ be an optimal solution

Both are in the order of finish time.

1. If $i_1 \neq j_1$, since i_1 finishes the earliest, and thus not later than j_1 , we can replace j_1 by i_1 in Opt. After this replacement, Opt is still optimal. Now

$$Opt = i_1 j_2 \dots j_k$$

2. If $i_2 \neq j_2$, since i_2 finishes earlier than j_2 , we can replace j_2 by i_2 in Opt. After this replacement , Opt is still optimal. Now

$$Opt = i_1 i_2 j_3 \dots j_k$$

.....

After a number of replacements:

$$Opt = i_1 i_2 \dots i_m$$

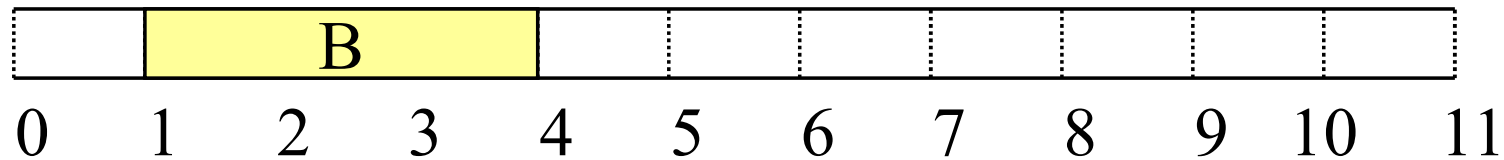
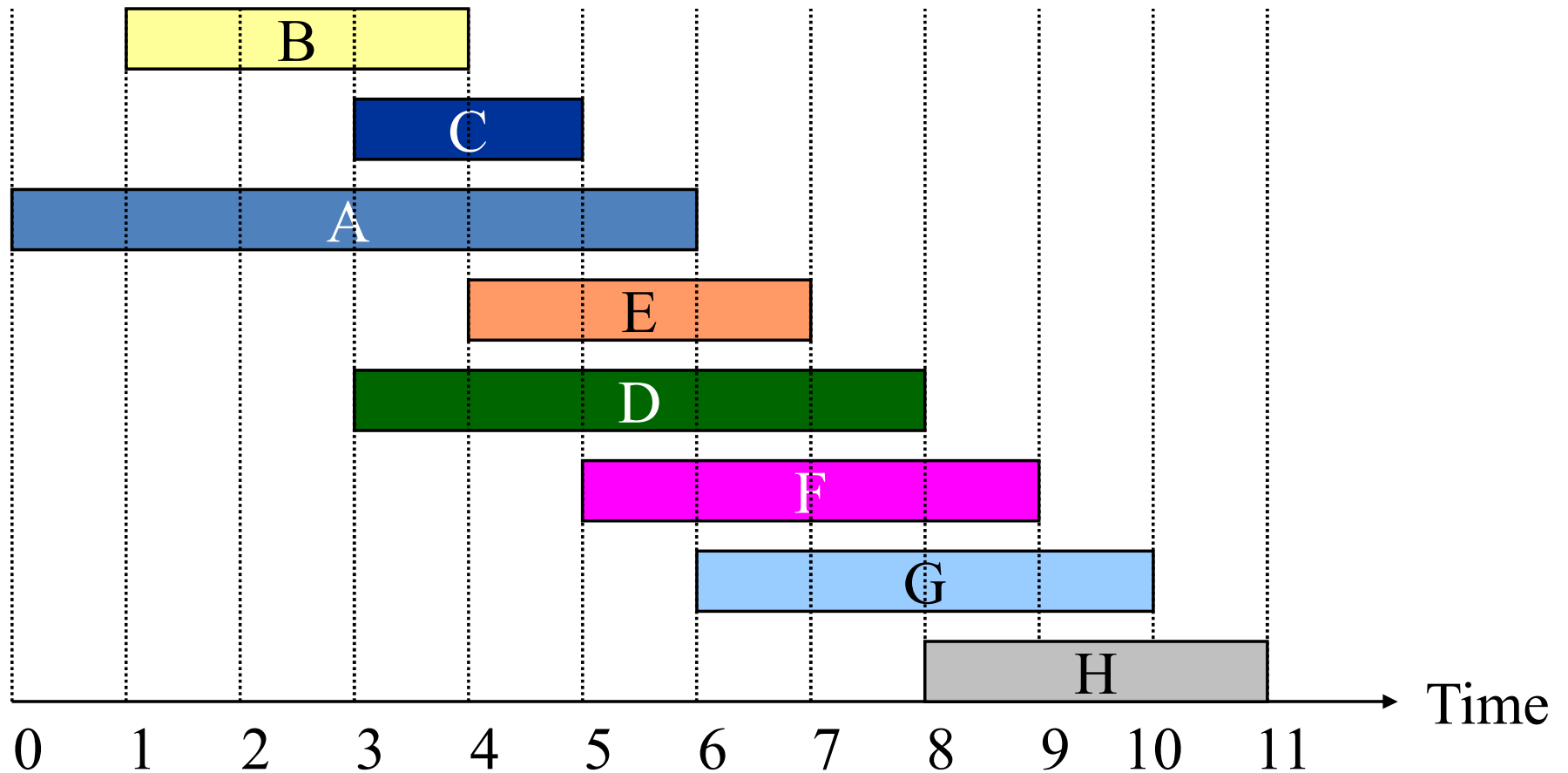
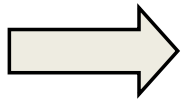
Or

$$Opt = i_1 i_2 \dots i_m j_{m+1} \dots j_k$$

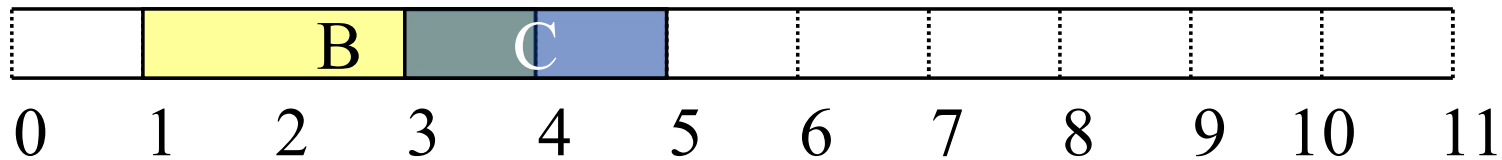
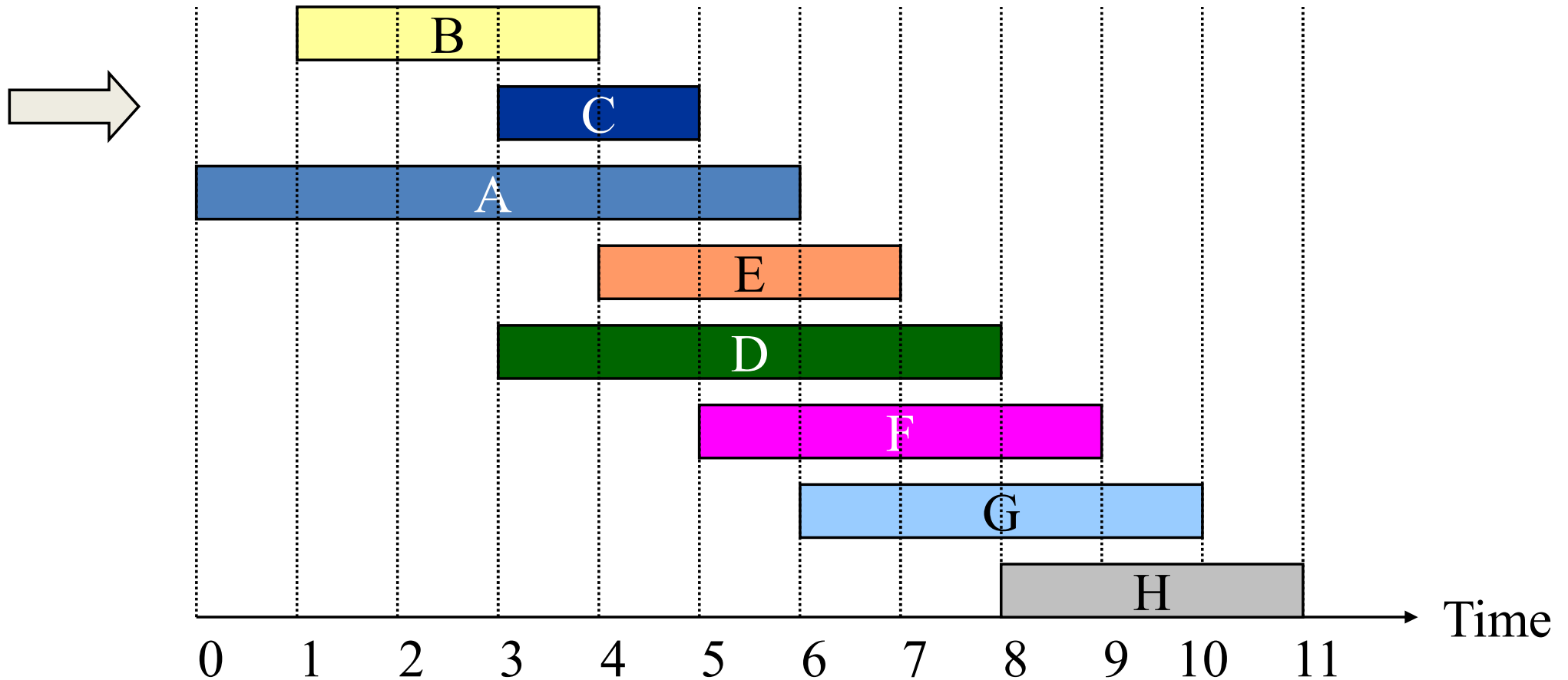
The last case is impossible, since no job starts later than finish time of i_m .

Therefore, G is optimal (it have the same number of jobs as Opt)

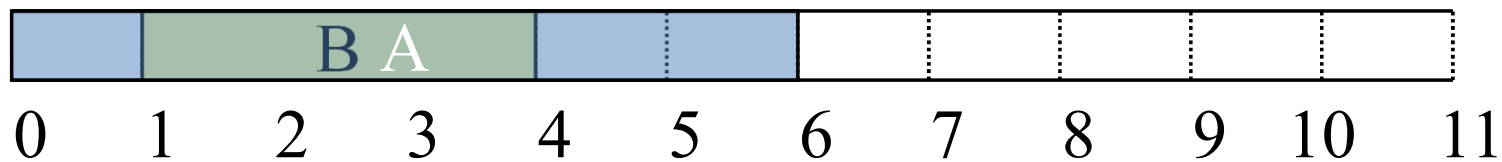
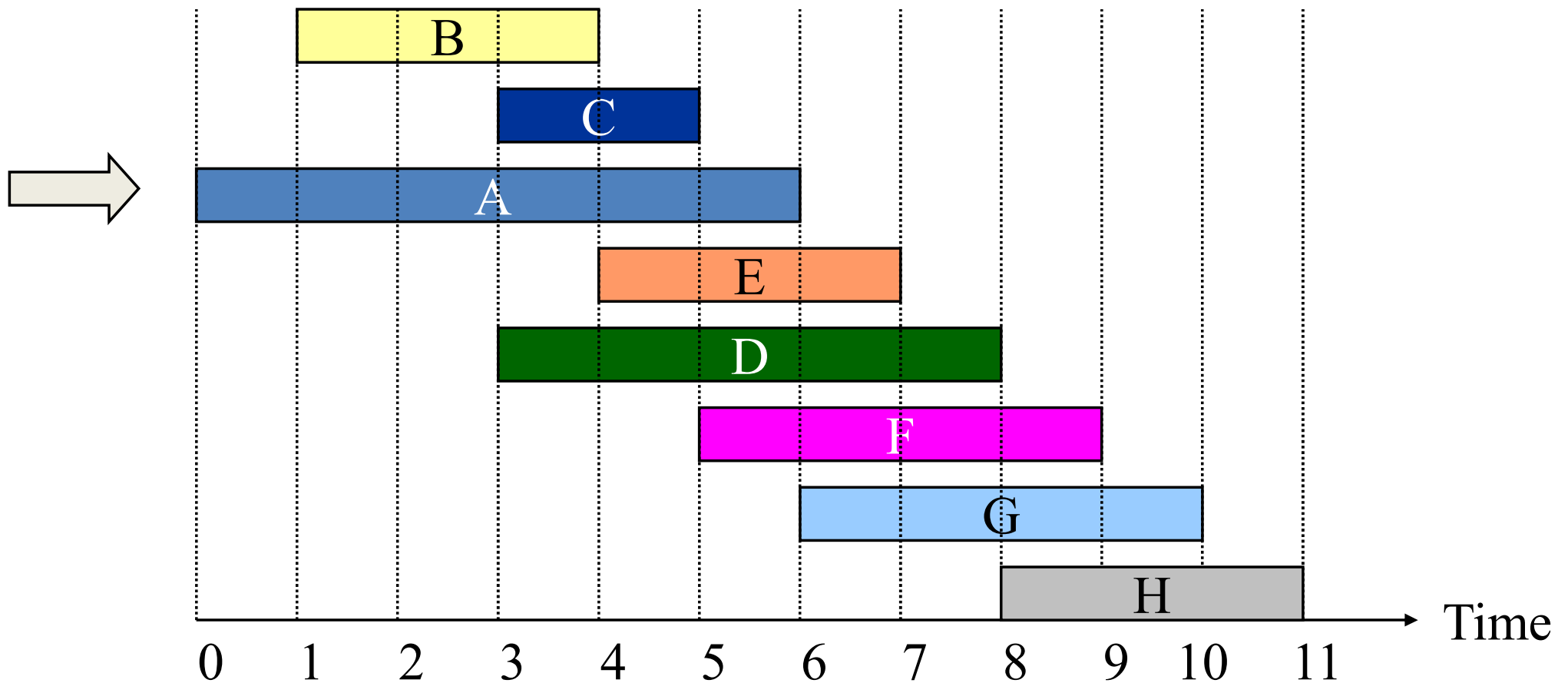
Interval Scheduling



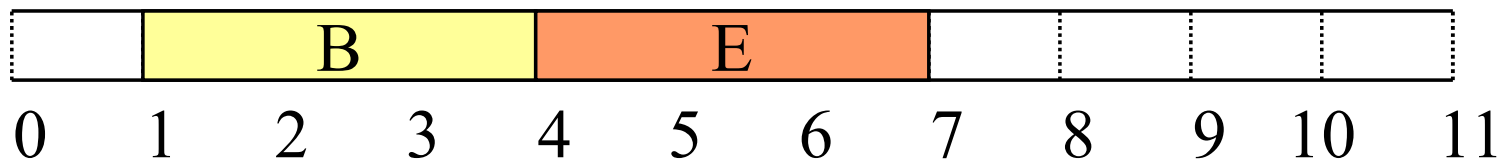
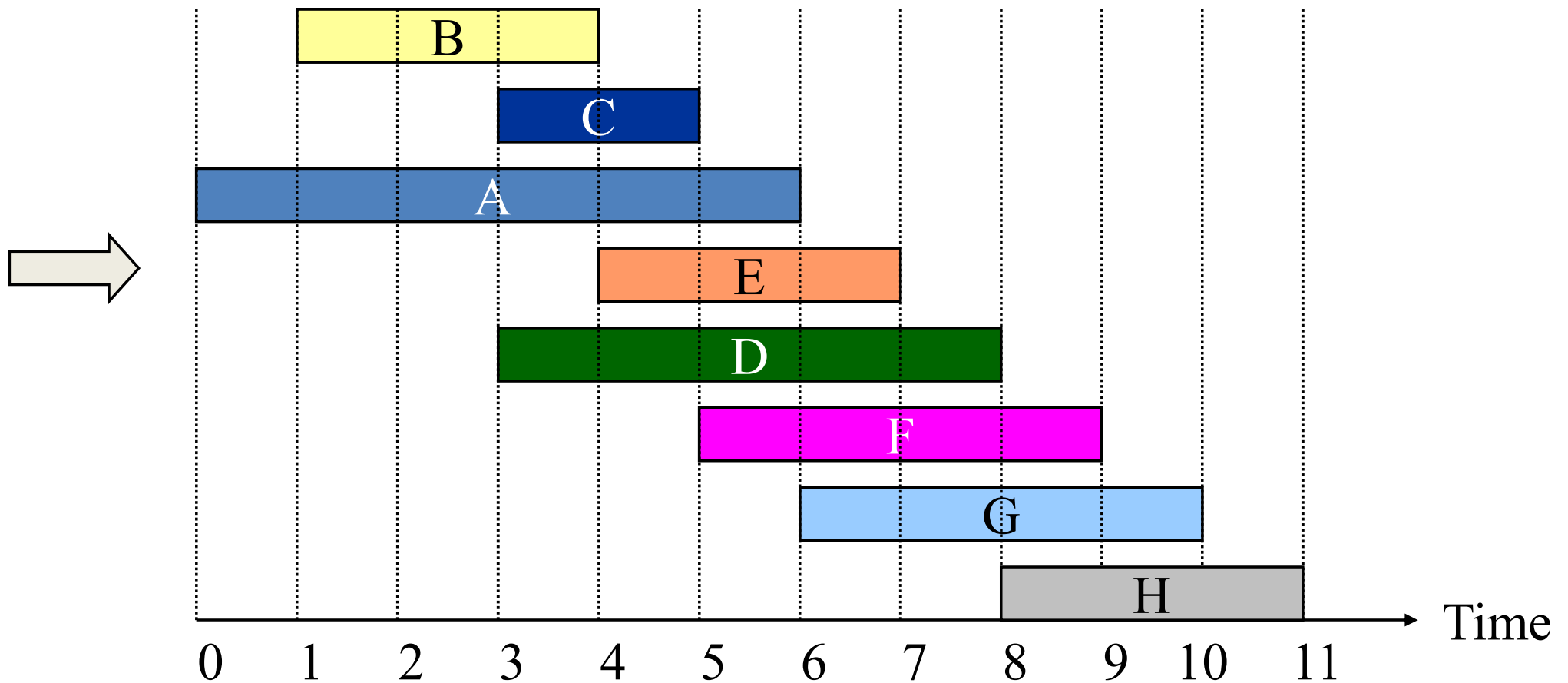
Interval Scheduling



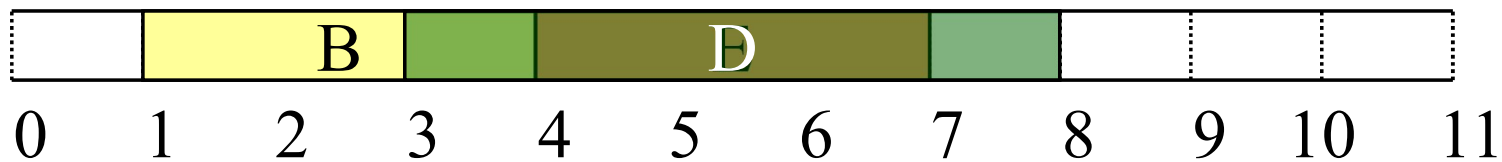
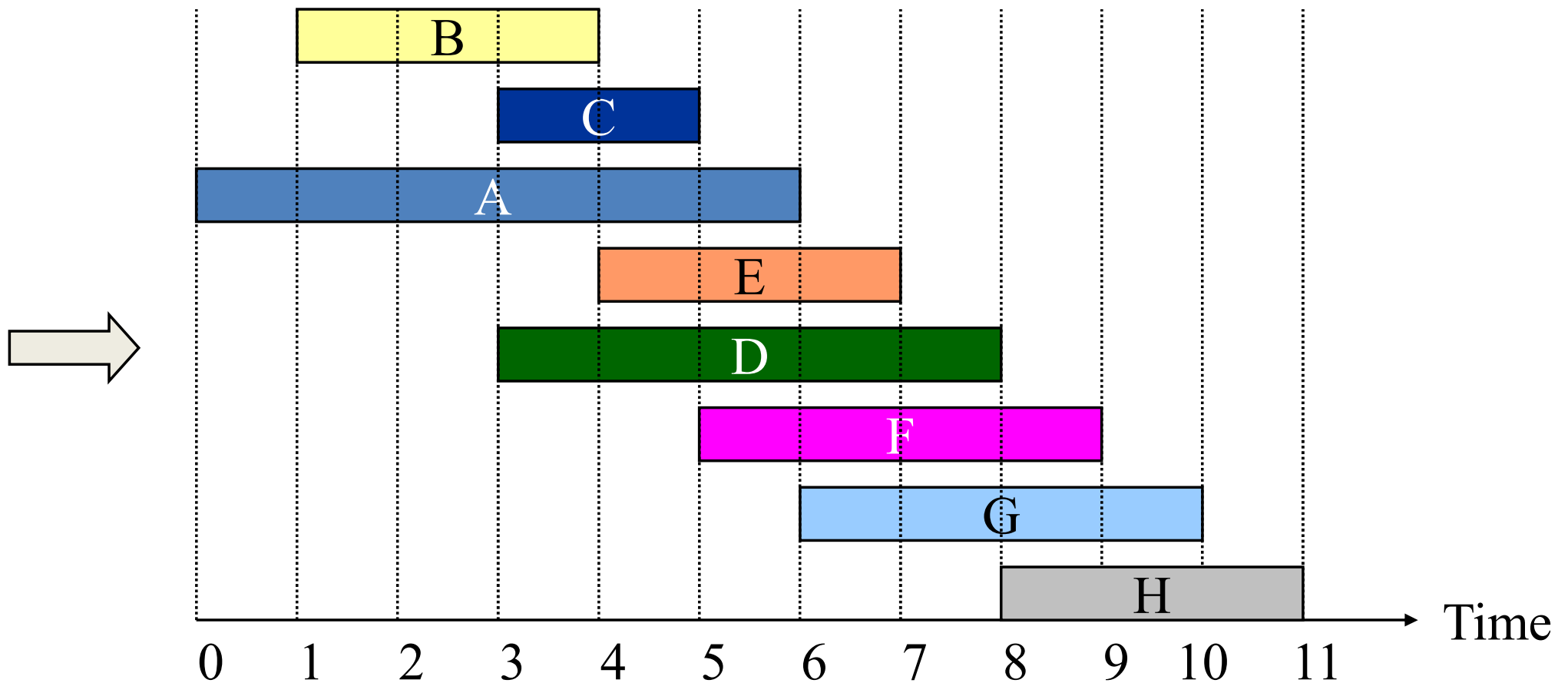
Interval Scheduling



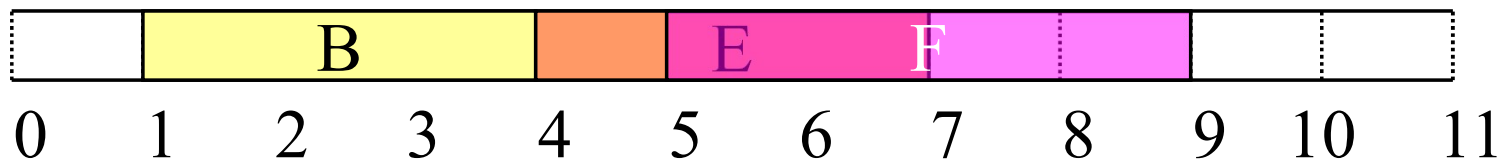
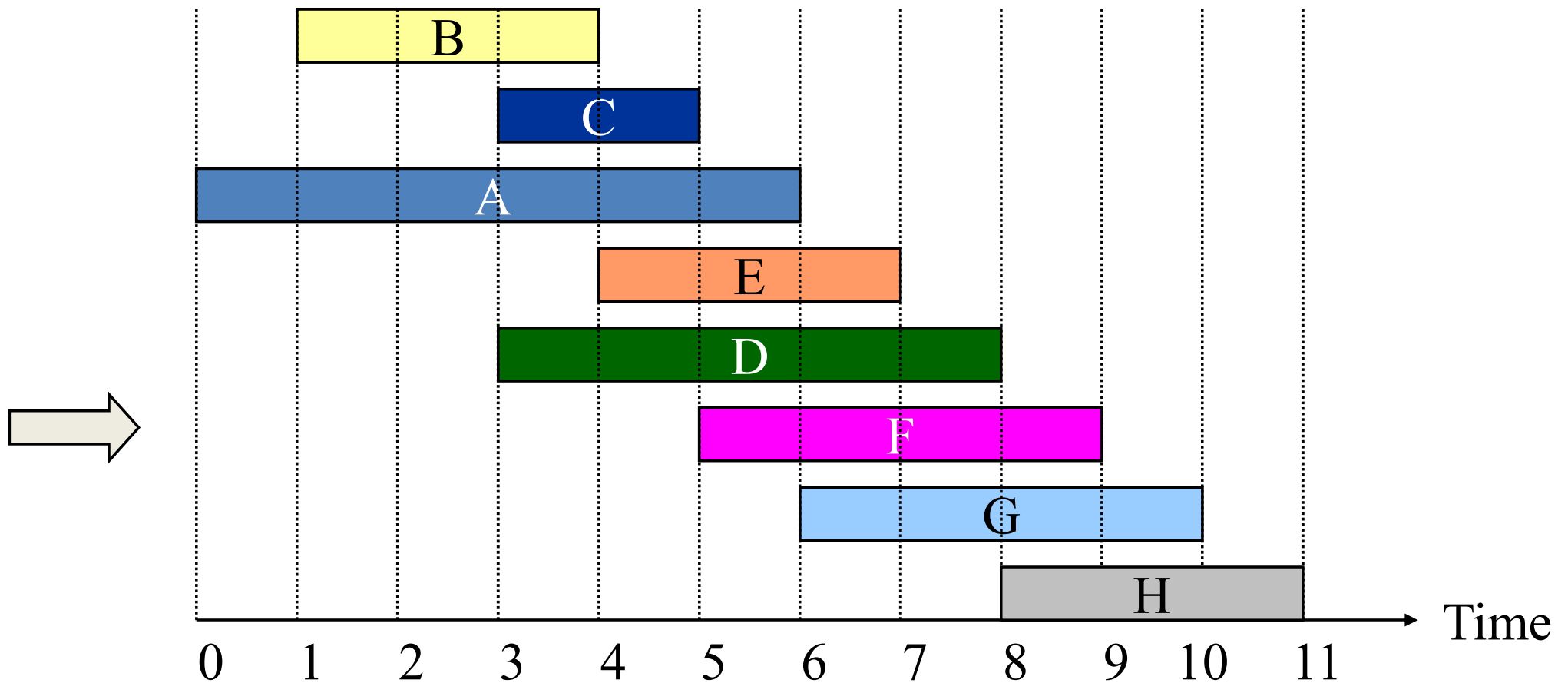
Interval Scheduling



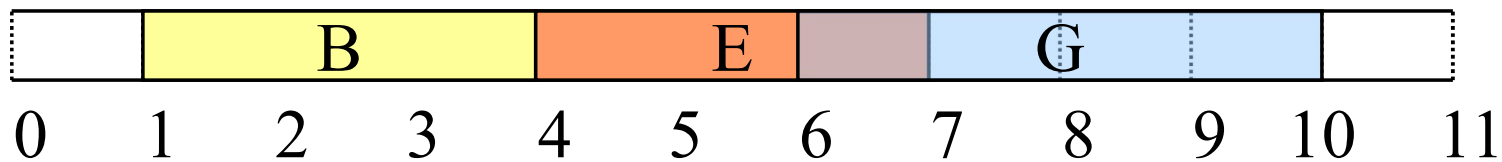
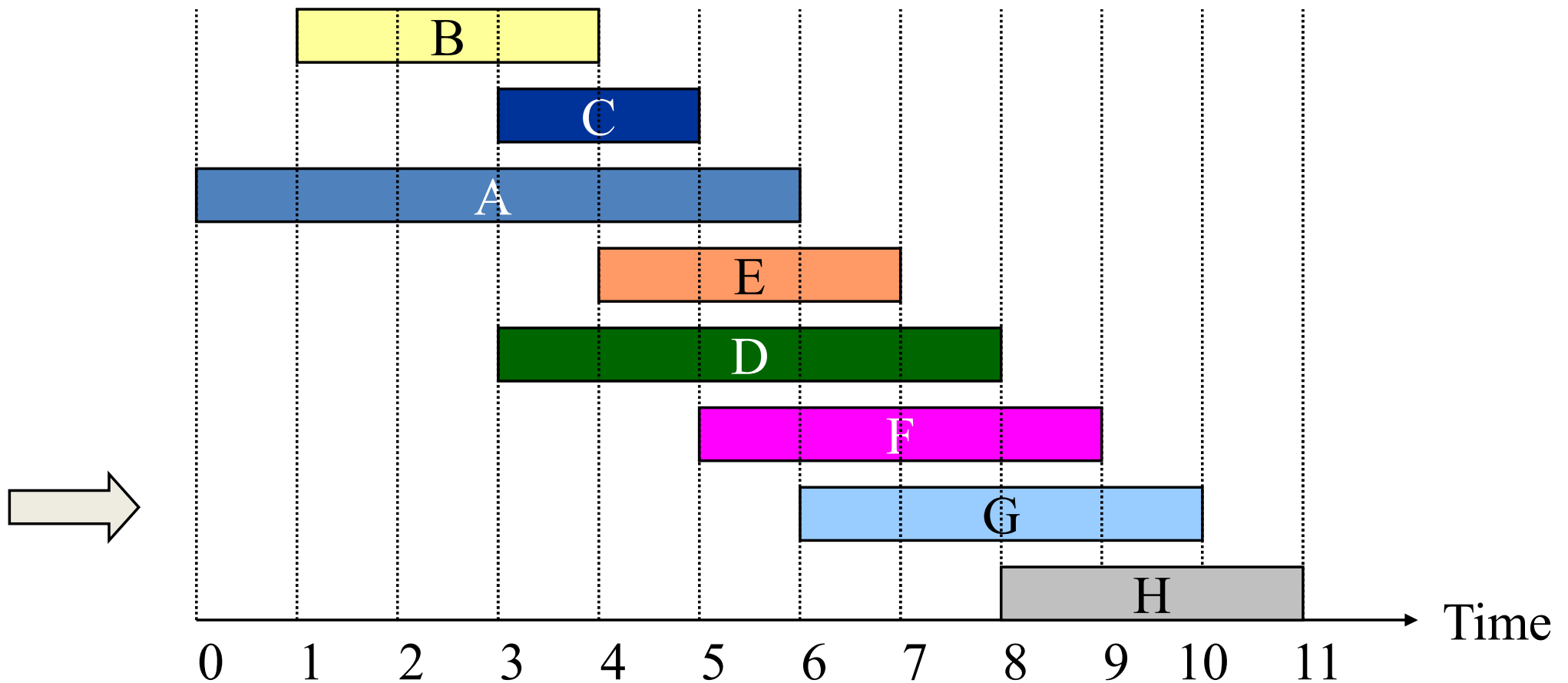
Interval Scheduling



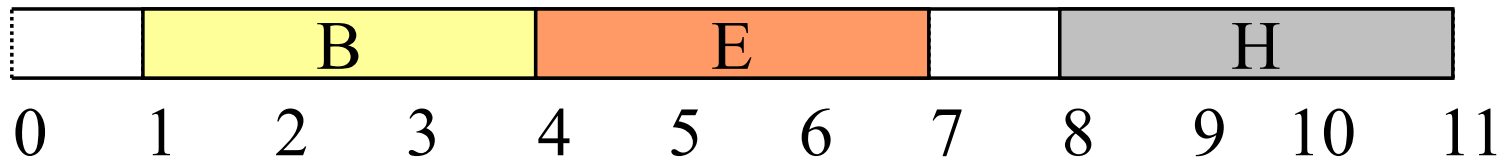
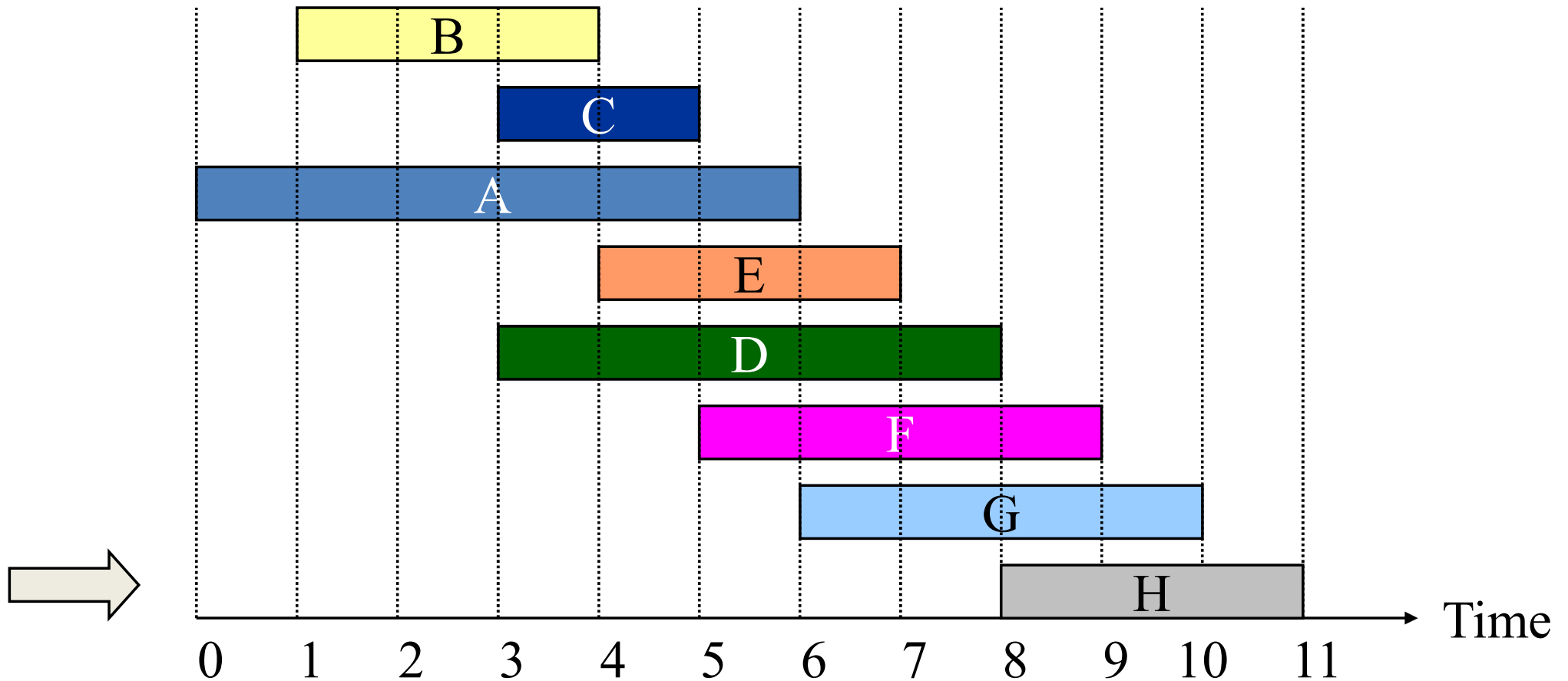
Interval Scheduling



Interval Scheduling



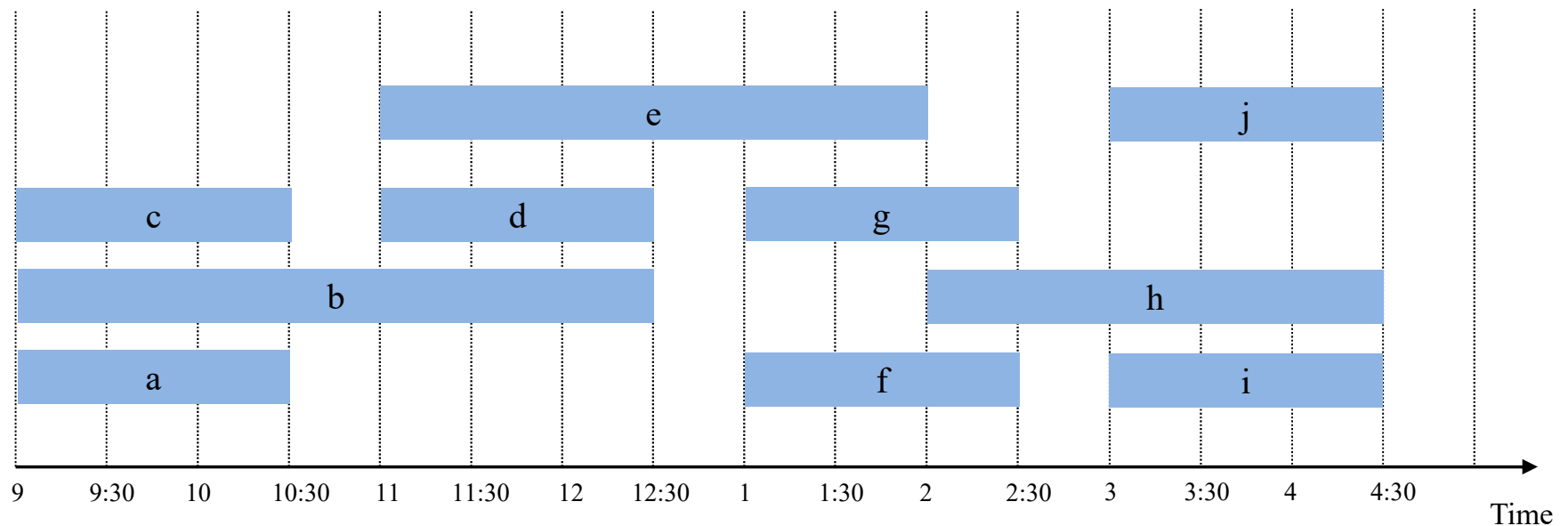
Interval Scheduling



Interval Partitioning

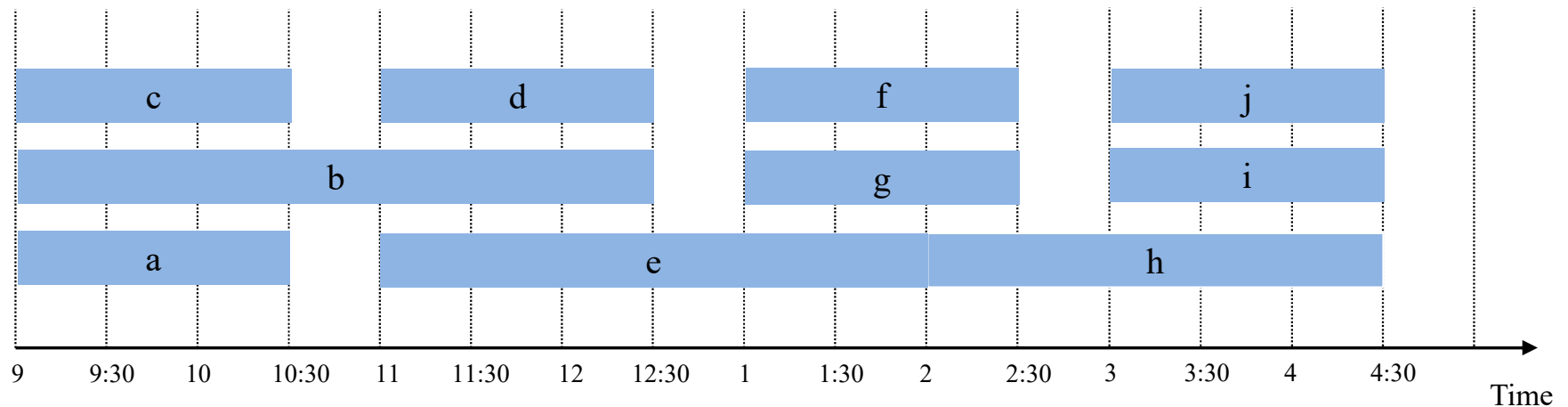
Interval Partitioning

- 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

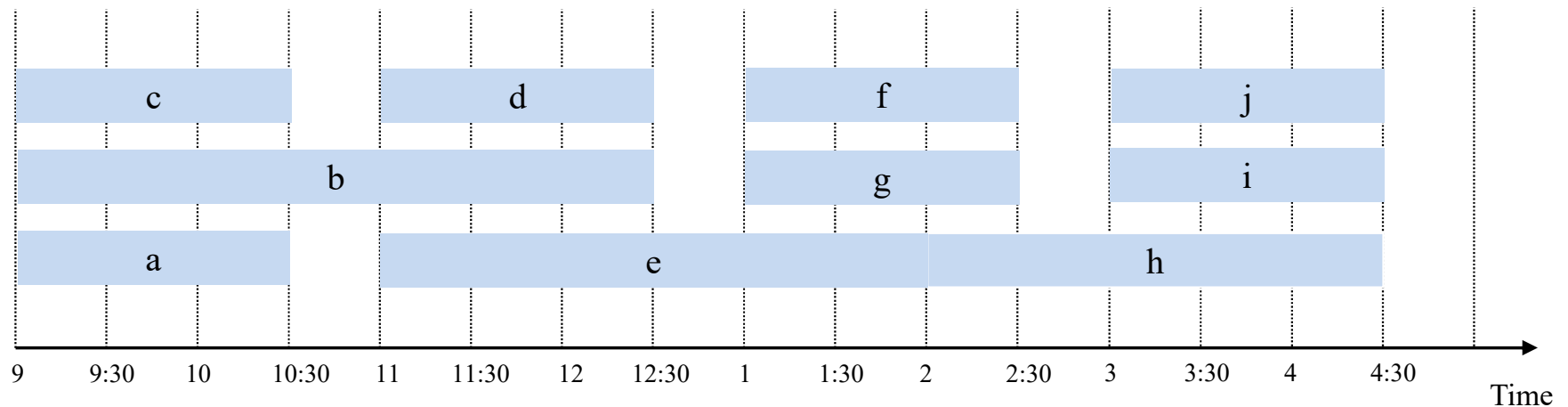
- Interval partitioning.
 - Lecture j starts at s_j and finishes at f_j .
 - Goal: find minimum number of classrooms to schedule all lectures so that no two occur at the same time in the same room.
- Ex: This schedule uses only 3.



Interval Partitioning: Lower Bound on Optimal Solution

- Def. The **depth** of a set of **open intervals** is the maximum number of overlapped lectures during the whole period.
- **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?



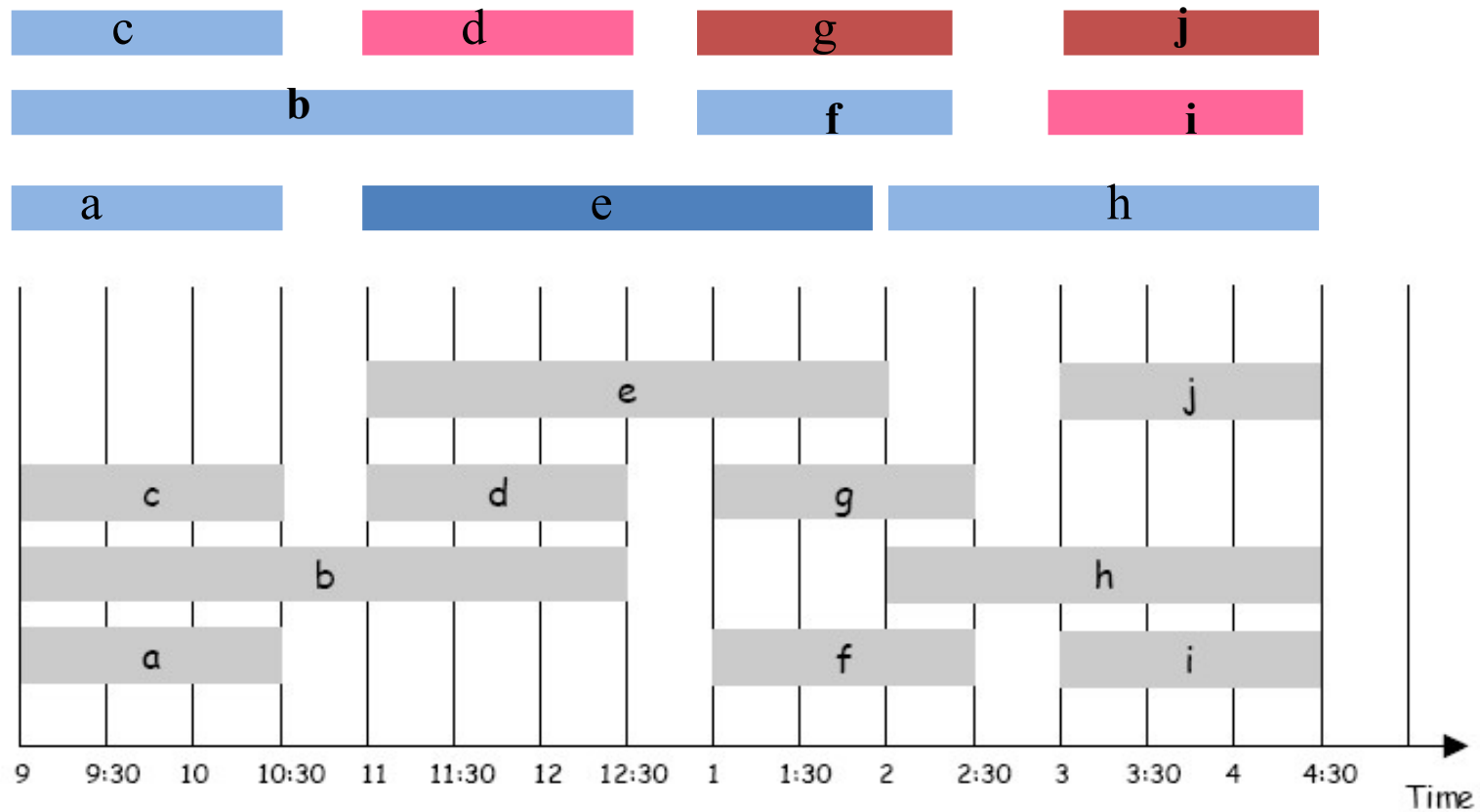
Interval Partitioning: Greedy Algorithm

- Greedy algorithm. Consider lectures in increasing order of start time: assign lecture to any compatible classroom (**Don't open any new classroom unless necessary**).

```
Sort intervals by starting time so that  $s_1 \leq s_2 \leq \dots \leq s_n$ .  
d  $\leftarrow$  0  $\leftarrow$  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.

Greedy Algorithm:



Interval Partitioning: Greedy Analysis

- **Observation.** Greedy algorithm never schedules two incompatible lectures in the same classroom (**its solution is always feasible**).
- **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.
 - 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 .
 - $d \leq \text{depth}$
 - Key observation \Rightarrow all schedules need to use $\geq \text{depth}$ classrooms.

Greedy Analysis Strategies

- **Exchange argument.** Gradually transform any solution to the one found by the greedy algorithm without hurting its quality.
- **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.