

# COMP 355

## Advanced Algorithms

### More on Greedy Algorithms



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## Greedy Algorithms

- **Optimization problem**
  - Given an input, compute a solution, subject to various constraints, that either minimizes cost or maximizes profit.
- **Efficiency**
  - Can we produce the optimal solution without using brute-force?
- Work for a number of optimization problems including MSTs (optimal solution)
- Provide fast heuristics (non-optimal solution strategies) = good approximations



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## Interval Scheduling: 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  $f_1 \leq f_2 \leq \dots \leq f_n$ .
  ↙ 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  $s_j \geq f_{j^*}$ .

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## Interval Scheduling: Correctness

Two issues

- Valid schedule (output is correct)?
- Is the schedule optimal (includes maximum number of activities)?

**Theorem.** Greedy algorithm produces a valid schedule.

**Pf.** A job j is only added to A if it is compatible with the other jobs already in A.

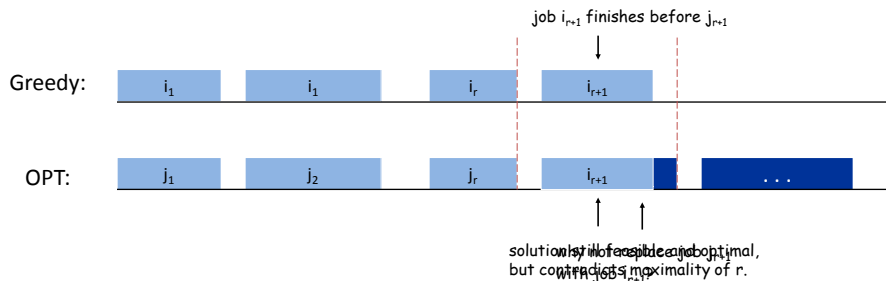
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## Interval Scheduling: Analysis

**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  $r$ .



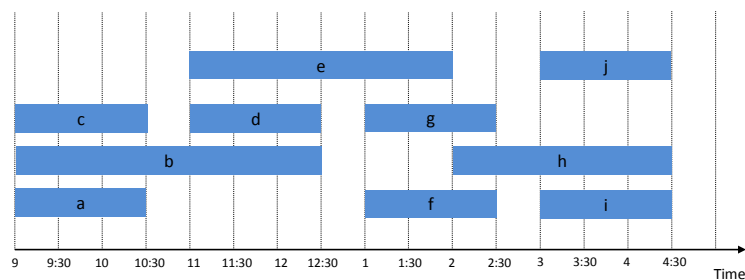
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## Interval Partitioning

**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.



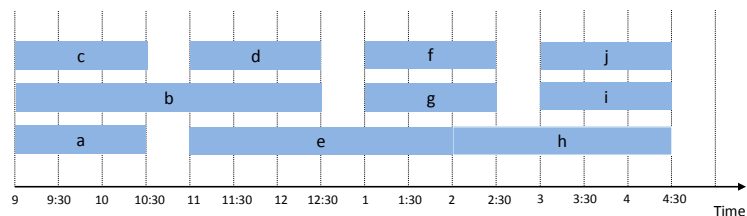
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# Interval Partitioning

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.



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## Interval Partitioning: Lower Bound on Optimal Solution

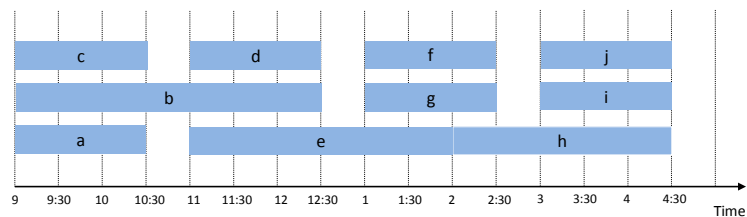
Def. The **depth** of a set of open intervals is the maximum number that contain any given time.

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?



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## 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.

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## Interval Partitioning: Greedy Algorithm

**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.
- 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.

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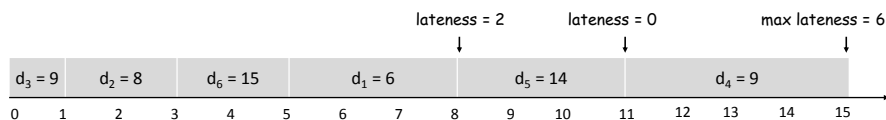
## Scheduling to Minimizing Lateness

### 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



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## 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  $d_j - t_j$ .

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## 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

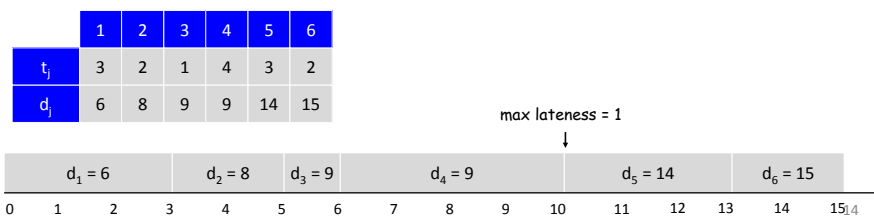
counterexample

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## Minimizing Lateness: Greedy Algorithms

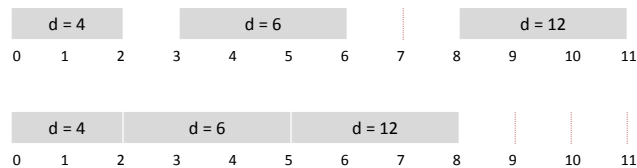
*Greedy algorithm.* Earliest deadline first.

```
Sort n jobs by deadline so that  $d_1 \leq d_2 \leq \dots \leq d_n$ 
t ← 0
for j = 1 to n
    Assign job j to interval [t, t + tj]
    sj ← t, fj ← t + tj
    t ← t + tj
output intervals [sj, fj]
```



## Minimizing Lateness: No Idle Time

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

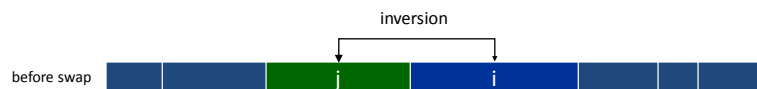


**Observation.** The greedy schedule has no idle time.

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## Minimizing Lateness: Inversions

**Def.** An **inversion** in schedule  $S$  is a pair of jobs  $i$  and  $j$  such that:  $i < j$  but  $j$  scheduled before  $i$ .



**Observation.** Greedy schedule has no inversions.

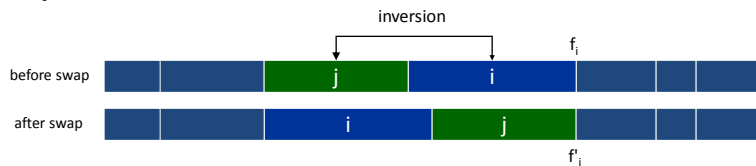
**Observation.** If a schedule (with no idle time) has an inversion, it has one with a pair of inverted jobs scheduled consecutively.

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## Minimizing Lateness: Inversions

**Def.** An **inversion** in schedule  $S$  is a pair of jobs  $i$  and  $j$  such that:  $i < j$  but  $j$  scheduled before  $i$ .



**Claim.** Swapping two adjacent, 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  $k \neq i, j$
- $\ell'_i \leq \ell_i$
- If job  $j$  is late:

$$\begin{aligned}
 \ell'_j &= f'_j - d_j && \text{(definition)} \\
 &= f_i - d_j && \text{(\textit{j} finishes at time } f_i\text{)} \\
 &\leq f_i - d_i && (i < j) \\
 &\leq \ell_i && \text{(definition)}
 \end{aligned}$$

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## 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  $S^*$

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## 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.

**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.