

## Objectives

- Greedy Algorithms
  - Interval partitioning
  - Minimizing Lateness
- Greedy stays ahead
- Exchange argument

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## Review: Greedy Algorithm Template

- Consider jobs (or whatever) in some order
  - Decision: What order is best?
- Take each job provided it's compatible with the ones already taken

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## Greedy Algorithms

- At each step, take as much as you can get
  - Feasible – satisfy problem's constraints
  - Locally optimal – best local choice among available feasible choices
  - Irrevocable – after decided, no going back

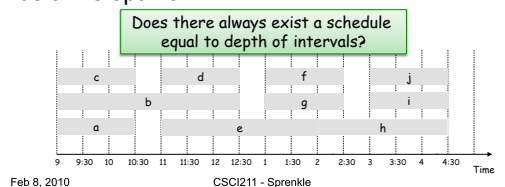
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## Interval Partitioning: Lower Bound on Optimal Solution

- Def. The depth of a set of open intervals is the maximum number that contain any given time.
- Key observation. # of classrooms needed  $\geq$  depth.  
 a, b, c all contain 9:30
- Ex: Depth of schedule below = 3  $\Rightarrow$  schedule below is optimal.



## Interval Partitioning Discussion

- Does there always exist a schedule equal to depth of intervals?
- Can we make decisions locally to get a global optimum?
  - Or are there long-range obstacles that require more resources?

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## Interval Partitioning: 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 = 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 = d + 1
  
```

Analyze algorithm

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## Interval Partitioning: 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 = 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 = 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 Analysis

- Observation.** Greedy algorithm never schedules two incompatible lectures in the same classroom
- Theorem.** Greedy algorithm is optimal
- Pf Intuition**
  - When do we add more classrooms?
  - When would we add the  $d+1$  classroom?

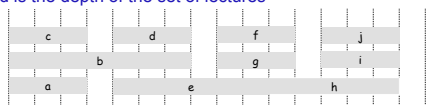
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## Interval Partitioning: Greedy Analysis

- Observation.** Greedy algorithm never schedules two incompatible lectures in the same classroom
- Theorem.** Greedy algorithm is optimal
- Pf.**
  - Let  $d$  = number of classrooms that 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  $s_j + \epsilon$
  - $d$  is the depth of the set of lectures



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## Proving Greedy Algorithms Work

- Specifically, produce an **optimal** solution
- Two approaches:
  - Greedy algorithm stays ahead** ←
    - Does better than any other algorithm at each step
  - Exchange argument**
    - Transform any solution into a greedy solution

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## Greedy Stays Ahead Proofs

- Define your solutions
  - Describe the form your greedy solution takes and what form some other solution takes (possibly the optimal solution)
  - Example: Let  $A$  be the solution constructed by the greedy algorithm and  $O$  be an optimal solution.
- Find a measure
  - Find a measure by which greedy stays ahead of the optimal solution
  - Ex: Let  $a_1, \dots, a_k$  be the first  $k$  measures of greedy algorithm and  $o_1, \dots, o_m$  be the first  $m$  measures of other solution (sometimes  $m = k$ )
- Prove greedy stays ahead
  - Show that greedy's partial solutions constructed are always just as good as the initial segments of the optimal solution, based on the measure
  - Ex: for all indices  $r \leq \min(k, m)$ , prove by induction that  $a_r \geq o_r$  or  $a_r \leq o_r$
  - Use the greedy algorithm to help you argue the inductive step
- Prove optimality
  - Prove that since greedy stays ahead of the other solution with respect to the measure, then the greedy solution is optimal.

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Exchange argument

## SCHEDULING TO MINIMIZE LATENESS

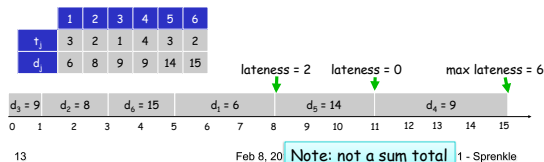
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## Scheduling to Minimizing Lateness

- Single resource processes one job at a time
- Job  $j$  requires  $t_j$  units of processing time and is due at time  $d_j$  (its deadline)
- 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**  $L = \max \ell_j$



## Greedy Algorithms

- **Greedy template.** Consider jobs in some order.
- What do we want to optimize?
- What order?
  - Intuition of order?
  - Counter examples for order being optimal?

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

Counter example

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

Counter example

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

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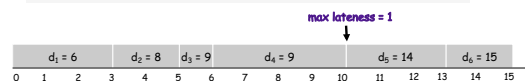
## Minimizing Lateness: Greedy Algorithm

- **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  $[t, t + t_j]$ 
   $s_j = t$ 
   $f_j = t + t_j$ 
   $t = t + t_j$ 
output intervals  $[s_j, f_j]$ 

```



What can we say about this algorithm/its results?

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## 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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## Proving Optimality

- **Goal:** Prove greedy algorithm produces optimal solution
- **Approach: Exchange argument**
  - Start with an optimal schedule Opt
  - Gradually modify Opt
    - Preserving its optimality
  - Transform into a schedule identical to greedy's schedule

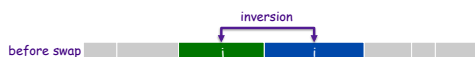
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## Minimizing Lateness: Inversions

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



Can Greedy's solution have any inversions?

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## Minimizing Lateness: Inversions

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



Greedy's schedule has no inversions!

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## Minimizing Lateness: Inversions

- Claim. Swapping two adjacent jobs with the same deadline does not increase the max lateness
- Pf Sketch. Let  $\ell$  be the lateness before the swap, and let  $\ell'$  be it afterwards
  - > Lateness of other jobs?
  - > Lateness of  $i$ ?  $j$ ?



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## Minimizing Lateness: Inversions

- Claim. Swapping two adjacent jobs with the same deadline does not increase the max lateness
- Pf. Let  $\ell$  be the lateness before the swap, and let  $\ell'$  be it afterwards
  - > Lateness remains the same for all other jobs:
    - $\ell'_k = \ell_k$  for all  $k \neq i, j$
  - > Lateness of  $i$  before is  $f_i - d_i = t_i + t_j - d_i$
  - > Lateness of  $j$  after is  $f'_j - d_j = t_i + t_j - d_j$ 
    - But  $d_i = d_j$



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## Minimizing Lateness: Inversions

- Claim. Swapping two adjacent, inverted jobs reduces the number of inversions by one and does **not increase the max lateness**
  - > How do we know inversions are adjacent?
- Pf Setup. Let  $\ell$  be the lateness before the swap, and let  $\ell'$  be it afterwards
  - > What can we say about how  $i$ 's,  $j$ 's, and other jobs' lateness changes?



By def of inversion,  $d_i < d_j$

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## Minimizing Lateness: Inversions

- 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:
 

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

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## Assignments

- Read Chapter 4
  - Wiki due next Wednesday
- Exam 1
  - Open book, open notes, open lecture notes
  - **NO OTHER RESOURCES**
  - I mention explicitly to analyze your algorithms' running times. I will not do that in the future.
  - Wed: half lecture, half questions

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