

Objectives

- Wrap up minimizing max lateness

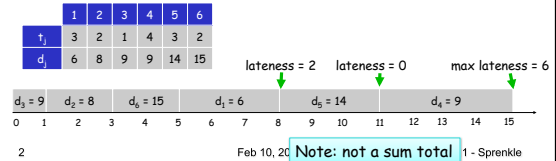
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1

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** $L = \max \ell_j$



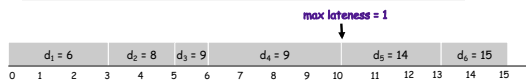
Minimizing Lateness: Greedy Algorithm

- **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  $[s_j, f_j]$ 

```



3

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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!

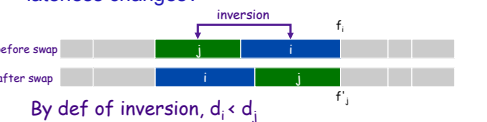
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4

Minimizing Lateness: Inversions

- **Claim.** Swapping two adjacent, inverted jobs reduces the number of inversions by one and does **not increase the max lateness**
- **Pf Setup.** Let ℓ be the lateness before the swap, and let ℓ' be it afterwards



5

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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**.

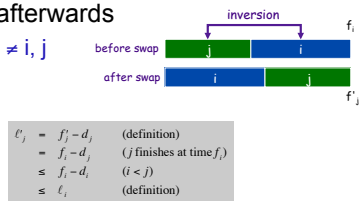
- **Pf.** Let ℓ be the lateness before the swap, and let ℓ' be it afterwards

➤ $\ell'_k = \ell_k$ for all $k \neq i, j$

➤ Know: $d_i < d_j$

➤ $\ell'_i \leq \ell_i$

➤ If job j is late:



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

6

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Minimizing Lateness: Analysis of Greedy Algorithm

- **Theorem.** Greedy schedule S is optimal
- **Pf idea.** Convert Opt to Greedy
 - Does opt schedule have idle time?
 - What if opt schedule has no inversions?
 - What if opt schedule has inversions?

7

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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^*

8

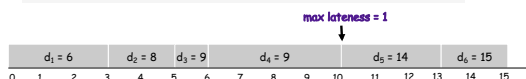
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Analyzing Running Time

- **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  $[s_j, f_j]$ 
```

 $O(n \log n)$ 

What is the runtime of this algorithm?

9

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Greedy Exchange Proofs

1. Label your algorithm's solution and a general solution.
 - For example, let $A = \{a_1, a_2, \dots, a_n\}$ be the solution generated by your algorithm, and let $O = \{o_1, o_2, \dots, o_m\}$ be an arbitrary (or optimal) feasible solution.
2. Compare greedy with other solution.
 - Assume that your arbitrary/optimal solution is not the same as your greedy solution (since otherwise, you are done).
 - Typically, you can isolate a simple example of this difference, such as one of the following:
 - There is an element of O that is not in A and an element of A that is not in O
 - There are 2 consecutive elements in O in a different order than they are in A (i.e., there is an *inversion*).
3. Exchange.
 - **Swap** the elements in question in O (either swap one element out and another in for the first case, or swap the order of the elements in the second case), and argue that you have a solution that is no worse than before.
 - Then argue that if you continue swapping, you eliminate all differences between O and A in a *finite* # of steps without worsening the solution's quality.
 - Thus, the greedy solution produced is just as good as any optimal solution, and hence is optimal itself.

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10

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.

11

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Assignments

- Read Chapter 4
 - Wiki due next Wednesday
- Friday: Exam 1 Due

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12