

## Objectives

- Wrap Up: Minimizing Lateness
  - Greedy exchange
- Problem: Shortest Path

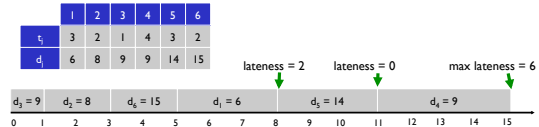
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## Review: Scheduling to Minimizing Lateness

- 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**  $L = \max \ell_j$



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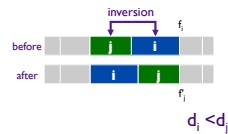
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Note: not a sum total

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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 to prove?



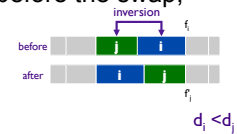
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## Minimizing Lateness: Inversions

- 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



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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  $k \neq i, j$
  - $\ell_i \leq \ell'_i$  because  $d_i < d_j$
  - Lateness of  $i$  before is  $\ell_i = f_i - d_i = T_{i-1} + t_i + t_j - d_i$
  - Lateness of  $j$  after is  $\ell'_j = f'_j - d_j = T_{i-1} + t_i + t_j - d_j$ 
    - But  $d_i < d_j$



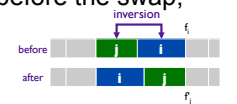
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## Minimizing Lateness: Inversions

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



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

➢  $\ell_j \leq \ell_i$ ,  $\ell'_i \leq \ell_i$

➢ If job  $j$  is late:

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

Shows that the maximum lateness of jobs does not increase after swap

## Greedy Exchange Proofs

1. Label your algorithm's solution and a general solution.
  - Example: let  $A = \{a_1, a_2, \dots, a_n\}$  be the solution generated by your algorithm, and let  $O = \{o_1, o_2, \dots, o_n\}$  be an optimal feasible solution.
2. Compare greedy with other solution.
  - Assume that the arbitrary/optimal solution is not the same as your greedy solution (since otherwise, you are done).
  - Typically, can isolate a simple example of this difference, such as:
    - ① There is an element  $e \in O$  that  $\notin A$  and an element  $f \in A$  that  $\notin O$
    - ② 2 consecutive elements in  $O$  are in a different order than in  $A$ 
      - i.e., there is an *inversion*
3. Exchange.
  - Swap the elements in question in  $O$  (either ① swap one element out and another in or ② swap the order of the elements) and argue that solution is no worse than before.
  - 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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## 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?

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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 (and no idle time), 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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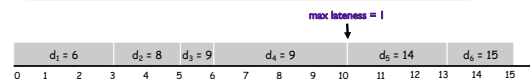
## Analyzing Running Time

- **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?

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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.

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## SHORTEST PATH

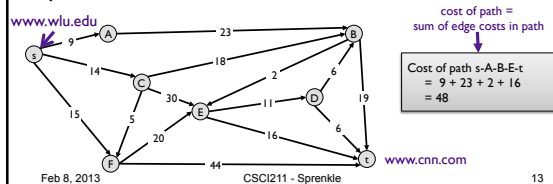
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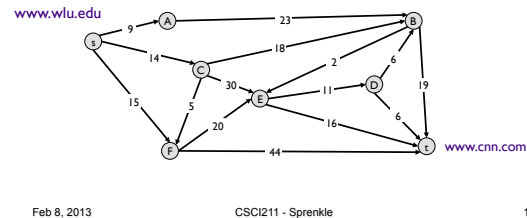
## Shortest Path Problem

- Given
  - Directed graph  $G = (V, E)$
  - Source  $s$ , destination  $t$
  - Length  $\ell_e$  = length of edge  $e$  (non-negative)
- Shortest path problem:** find shortest directed path from  $s$  to  $t$



## Shortest Path Problem

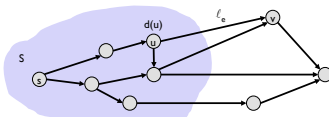
- Shortest path problem:** find shortest directed path from  $s$  to  $t$
- Brainstorming on solution ...



## Dijkstra's Algorithm

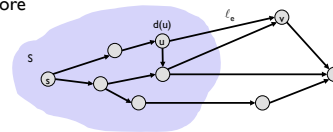
- Maintain a set of **explored nodes**  $S$ 
  - Keep the **shortest path distance**  $d(u)$  from  $s$  to  $u$
- Initialize  $S = \{s\}$ ,  $d(s) = 0$ ,  $\forall u \neq s, d(u) = \infty$
- Repeatedly choose unexplored node  $v$  which minimizes  $\pi(v) = \min_{e = (u,v) : u \in S} d(u) + \ell_e$ 
  - Add  $v$  to  $S$  and set  $d(v) = \pi(v)$ 

shortest path to some  $u$  in explored part followed by a single edge  $(u, v)$

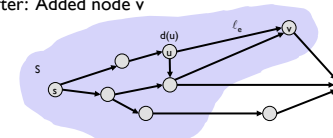


## Dijkstra's Algorithm

Before



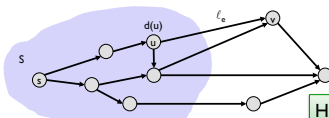
After: Added node  $v$



## Dijkstra's Algorithm

- Maintain a set of **explored nodes**  $S$ 
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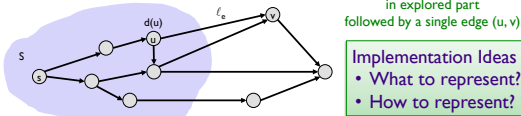


## How is Algorithm Greedy?

- We always form **shortest new s-v path** from a path in  $S$  followed by a *single edge*
- Proof of optimality:** *Stays ahead* of all other solutions
  - Each time selects a path to a node  $v$ , that path is shorter than every other possible path to  $v$

## Dijkstra's Algorithm

1. Maintain a set of **explored nodes**  $S$ 
  - Keep the shortest path distance  $d(u)$  from  $s$  to  $u$
2. Initialize  $S = \{s\}$ ,  $d(s) = 0$ ,  $\forall u \neq s, d(u) = \infty$
3. Repeatedly choose unexplored node  $v$  which minimizes  $\pi(v) = \min_{e = (u,v) : u \in S} d(u) + \ell_e$ ,
  - Add  $v$  to  $S$  and set  $d(v) = \pi(v)$



Implementation Ideas

- What to represent?
- How to represent?

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## Looking Ahead

- Exam due today at 5 p.m.
- Wiki due Tuesday for sections 3.4-3.6; chapter 4 (front matter), 4.1
  - Directed graphs, topological order
  - Greedy algorithms
- PS4 due Friday

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