# CS 341: Algorithms

### Lecture 6: Greedy algorithms, continued

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based on lecture notes by many other CS341 instructors

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# Minimizing lateness

#### Input:

- jobs  $J_1, \ldots, J_n$  with processing times  $t(1), \ldots, t(n)$  and deadlines  $d(1), \ldots, d(n)$
- can only do one thing at a time

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### **Output:**

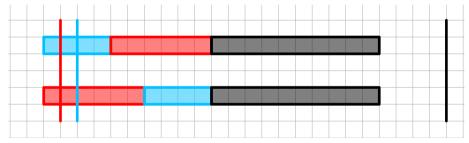
- $\bullet\,$  a scheduling of the jobs which minimizes maximal lateness
  - job  $J_i$  starts at time s(i) (TBD) and finishes at f(i) = s(i) + t(i)
  - if  $f(i) \ge d(i)$ , lateness  $\ell(i) = f(i) d(i)$ , otherwise 0
- maximal lateness =  $\max_i \ell(i)$

## Example: 3 jobs

- prepare my slides: need t(1) = 4 hours, deadline d(1) = 2 hours
- write solutions to assignments: need t(2) = 6 hours, deadline d(2) = 1 hour
- bake a panettone: need t(3) = 10 hours, deadline d(3) = 24 hours

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- 1, then 2, then 3: latenesses [2,9,0]
- 2, then 1, then 3: latenesses [8, 5, 0] (optimal)

### No breaks

#### **Observation:**

• if a scheduling has **idle time**, we can improve it by removing the breaks



• so the optimal has no idle time, and is given by a **permutation** of  $[1, \ldots, n]$ 

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,  $d(1) = 10$  so  $s(1) = 2$  and  $t(1) = 2$ ,  $d(2) = 5$  so  $s(2) = 3$ 

#### Attempt 3:

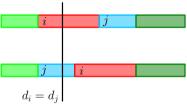
• do jobs in non-decreasing deadline order

slack = d(i) - t(i)

## **Non-uniqueness**

**Observation:** 

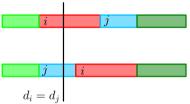
• if d(i) = d(j), the orderings  $[\ldots, i, j, \ldots]$  and  $[\ldots, j, i, \ldots]$  have the same max-lateness (because the second job is the latest)



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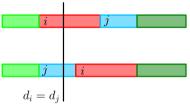


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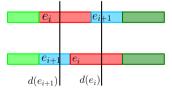
**Definition:** 

- take a permutation  $L = [e_1, \ldots, e_n]$  of  $[1, \ldots, n]$
- inversion: a pair (i, j) with i < j and  $d(e_i) > d(e_j)$ 
  - (= an inversion in  $[d(e_1), \ldots, d(e_n)]$  in the sense of lecture 3)
- no inversion  $\iff L$  in non-decreasing deadline order

- let  $L = [e_1, \ldots, e_n]$  be any permutation of  $[1, \ldots, n]$
- suppose that L is **not** in non-decreasing order of deadlines
- want:  $\max\_lateness(L) \ge \max\_lateness(L_{greedy})$

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- now, swap  $e_i$  and  $e_{i+1}$  to get a permutation L'. What about max\_lateness(L')?

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- we removed an inversion
- keep going: after at most n(n-1)/2 iterations, we have  $L_{\text{ord}}$  with **no inversion** and such that  $\max\_lateness(L_{\text{ord}}) \leq \max\_lateness(L)$
- we saw that  $\max\_lateness(L_{ord}) = \max\_lateness(L_{greedy})$

# Fractional knapsack

Input:

- items  $I_1, \ldots, I_n$  with weights  $w_1, \ldots, w_n$  and positive values  $v_1, \ldots, v_n$
- a capacity W

### **Output:**

- fractions  $E = e_1, \ldots, e_n$  such that
  - $0 \le e_j \le 1$  for all j
  - $e_1w_1 + \dots + e_nw_n \le W$
  - $e_1v_1 + \dots + e_nv_n$  maximal

#### Example:

- $w_1 = 10, v_1 = 60, w_2 = 30, v_2 = 90, w_3 = 20, v_3 = 100$
- W = 50

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#### Example:

- $w_1 = 10, v_1 = 60, w_2 = 30, v_2 = 90, w_3 = 20, v_3 = 100$
- W = 50
- optimal is  $e_1 = 1, e_2 = 2/3, e_3 = 1$ , total value 220

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### Remark:

- 0/1-version:  $e_j \in \{0, 1\}$  for all j
- dynamic programming

## The knapsack should be full

**Remark:** 

- if  $\sum_i w_i < W$ , just take all  $e_i = 1$
- so assume  $\sum_i w_i \ge W$

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- suppose we have an assignment with  $\sum_i e_i w_i < W$
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#### **Consequence:**

• it is enough to consider assignments with  $\sum_i e_i w_i = W$ 

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• pack with items in increasing weight  $w_i$ 

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- pack with items in increasing weight  $w_i$
- no:  $W = 10, w_1 = 10, v_1 = 100, w_2 = 5, v_2 = 1$

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### Attempt 3:

- pack with items in non-increasing "value per kilo"  $v_i/w_i$
- first example [6, 3, 5], second example [10, 1/5]

### Pseudo-code

GreedyKnapsack(v, w, W)1.  $E \leftarrow [0, \ldots, 0]$ sort items by non-increasing order of  $v_i/w_i$ 2.for  $k = 1, \ldots, n$  do 3. if  $w_k < W$  then 4.  $E[k] \leftarrow 1$ 5. $W \leftarrow W - w_k$ 6.  $\mathbf{else}$ 7.  $E[k] \leftarrow W/w_k$ 8. 9. return

**Remark:** output is  $S = [1, \ldots, 1, e_k, 0, \ldots, 0]$ **Runtime:**  $O(n \log(n))$ 

- let  $E = [e_1, \ldots, e_n]$  be the **output**, with  $\sum e_i w_i = W$
- let  $S = [s_1, \ldots, s_n]$  be any assignment, with  $\sum s_i w_i = W$
- assume that  $S \neq E$ , want  $value(E) \geq value(S)$

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- set  $s'_i = s_i + \alpha/w_i$  and  $s'_j = s_j \alpha/w_j$ , for  $\alpha$  TBD > 0, all other  $s'_k = s_k$
- in any case,  $\sum s'_i w_i = W$  and  $\text{value}(S') \ge \text{value}(S)$

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- in any case,  $\sum s'_i w_i = W$  and  $\text{value}(S') \ge \text{value}(S)$
- choose the first lpha such that either  $s_i'=e_i$  or  $s_j'=e_j$

$$\alpha = \min(w_i(e_i - s_i), w_j(s_j - e_j))$$

we found S' with one more common entry with E, and value(S') ≥ value(S)
if S' ≠ E, repeat, ..., until S'''... = E