#### **Fall'19 CSCE 629**

# Analysis of Algorithms

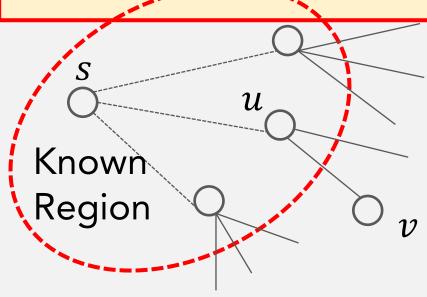
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#### Lecture 15

- Dijkstra's algorithm cont'd
- Interval scheduling

# Reflection on Dijkstra: greedy stays ahead

- Known region R: in which the shortest distance to s is known
- Growing R: adding v that has the shortest distance to s
- How to Identify v? The one that minimizes d(u) + l(u, v) for  $u \in R$ Shortest path to some u in known region, followed by a single edge (u, v)



```
Dijk(G,s) // initialize d(s) = 0, d(u) = \infty, R=Ø
While R ≠ V
Pick v \notin R w. smallest d(v) // by Priority Q
Add v to R
For all edges (v,w) \in E
If d(v) > d(u) + l(u,v)
d(v) \leftarrow d(u) + l(u,v)
```

#### **Contrast with Bellman-Ford**

■ Dijkstra (Greedy)  $O((m+n)\log n)$ 

$$\frac{d(v)}{d(v)} = \min_{u \in R} d(u) + l(u, v)$$

- Positive weight: no need to wait; more edges in a path do not help
- Bellman-Ford (Dynamic programming) O(mn)

$$\frac{\mathsf{OPT}(i,v)}{\mathsf{opt}(i-1,v)} = \min\left\{\mathsf{OPT}(i-1,v), \min_{v \to w \in E} \{\mathsf{OPT}(i-1,w) + l_{v \to w}\}\right\}$$

#### ❖Global vs. Local

- Dijkstra's requires global information: known region & which to add
- Bellman-Ford uses only local knowledge of neighbors, suits distributed setting

## Network routing: distance-vector protocol

#### Communication network

- Nodes: routers
- Edges: direct communication links
- Cost of edge: delay on link.

naturally nonnegative, but Bellman-Ford used anyway!

#### Distance-vector protocol ["routing by rumor"]

- Each router maintains a vector of shortest-path lengths to every other node (distances) and the first hop on each path (directions).
- Algorithm: each router performs separate computations for each potential destination node.
- Path-vector protocol: coping with dynamic costs

## Correctness of Dijkstra's algorithm

Known Region R

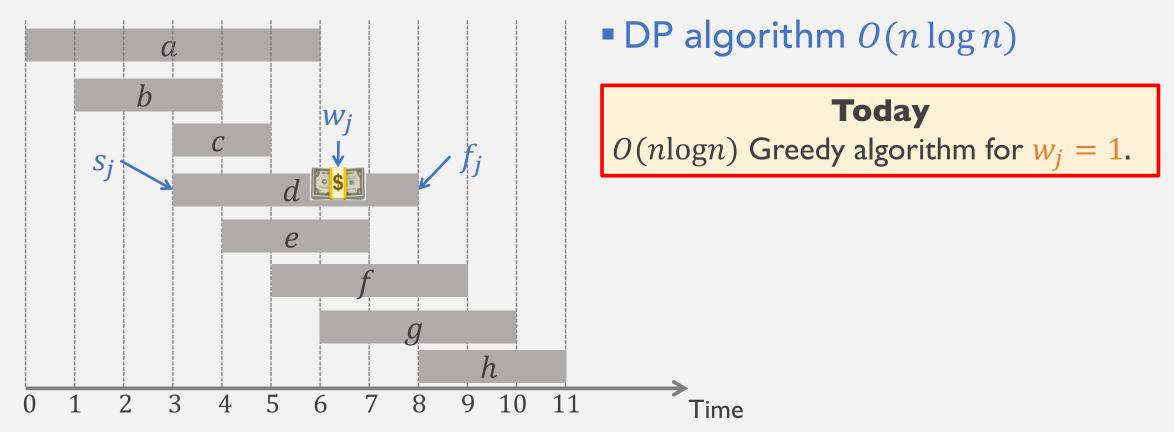
Invariant. For each node  $u \in R$ , d(u) is the length of a shortest s - u path

#### Proof. (By induction on size of R)

- Base case: |R| = 1 trivial
- Induction hypothesis: true for  $|R| = k \ge 1$ 
  - Let v be the next node added to R and (u, v) be the chosen edge. Call this s u v path P.
  - Consider any s v path Q. [Next show it's no shorter than P]
  - Let (x, y) be the first edge in Q leaving R; let Q' be the S x segment
  - $l(Q) \ge l(Q') + l(x,y) \ge d(x) + l(x,y) \ge l(P)$ ; because Dijkstra's picked v in this iteration (node outside R with shortest distance to s)

## Recall: weighted interval scheduling

- Input. n jobs; job j starts at  $s_j$ , finishes at  $f_j$ , weight  $w_j$
- Output. Subset of mutually compatible jobs of maximum weight



## **Greedy strategies**

Recall. DP recurrence.

OPT(j) = value of optimal solution to jobs 1,2, ..., j

$$\frac{OPT(j)}{OPT(j)} = \begin{cases}
0 & \text{if } j = 0 \\
max{OPT(j-1), } w_j + OPT(pre(j)) \end{cases} \text{ otherwise}$$

- Greedy: be lazy & pick the next compatible job that "looks nice"
  - Earliest start time: ascending order of  $s_i$ .
  - Earliest finish time: ascending order of  $f_i$ .
  - Shortest interval: ascending order of  $f_i s_i$ .
  - Fewest conflicts: the one that conflicts the least number of jobs go first.
- Exercise. Find counterexamples for each strategy (if possible)

#### **Greedy:** counterexamples

© Earliest start time: ⊗ Shortest interval: **©** Fewest conflicts:

© Earliest finishing time

## Greedy Algorithm: earliest finishing time

```
IntScheduling (\{s_j, f_j\})

1. Sort by finishing time so that f_1 \leq f_2 \leq \cdots \leq f_n \longrightarrow O(n \log n)

2. A \leftarrow \emptyset // set of selected jobs

3. For j = 1, ..., n

If j compatible with A

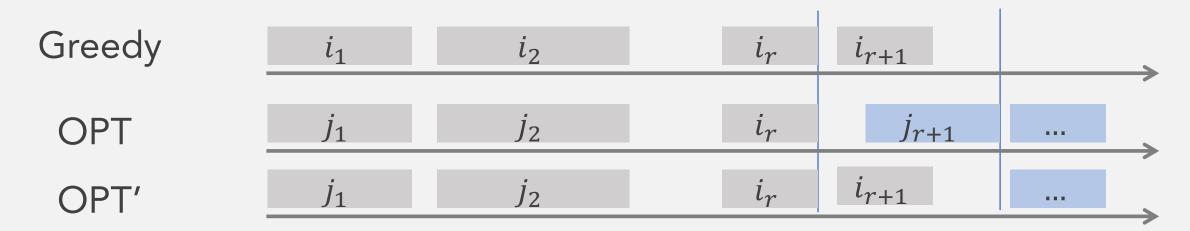
A \leftarrow A \cup \{j\}
```

- Running time:  $O(n \log n)$
- Correctness: proof by contradiction
  - Suppose greedy is not optimal
  - Consider an optimal strategy: one that agrees with Greedy for as many initial jobs as possible
  - Look at the first place that they differ: show a new optimal that agrees with greedy more

### **Greedy Algorithm: correctness**

#### Proof (by contradiction): Suppose greedy is not optimal

- Let  $i_1, i_2, ..., i_k$  denote set of jobs selected by greedy
- Let  $j_1, j_2, ..., j_m$  be set of jobs in the optimal solution OPT where  $i_1 = j_1, i_2 = j_2, ..., i_r = j_r$  for the largest possible value of r
- Sub  $i_{r+1}$  for  $j_{r+1}$  in OPT: still feasible and optimal (OPT'); but agrees with Greedy at r+1 positions; contradicts the maximality of r

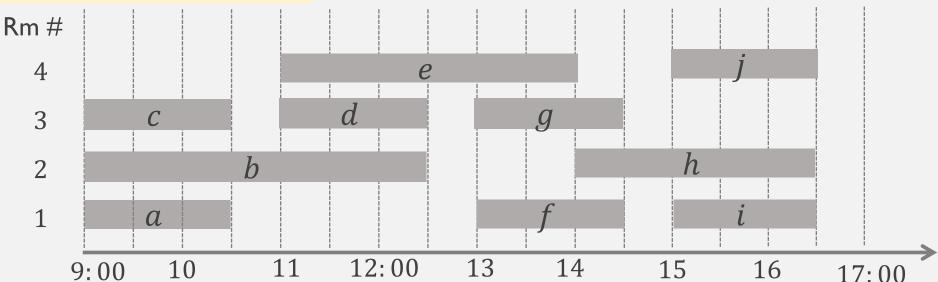


## **Interval Partitioning Problem**

#### Scheduling classes

- Input. Lectures  $\{s_j, f_j\}$
- Output. Minimum number of classrooms to schedule all lectures so that no two occur at the same time in the same room.

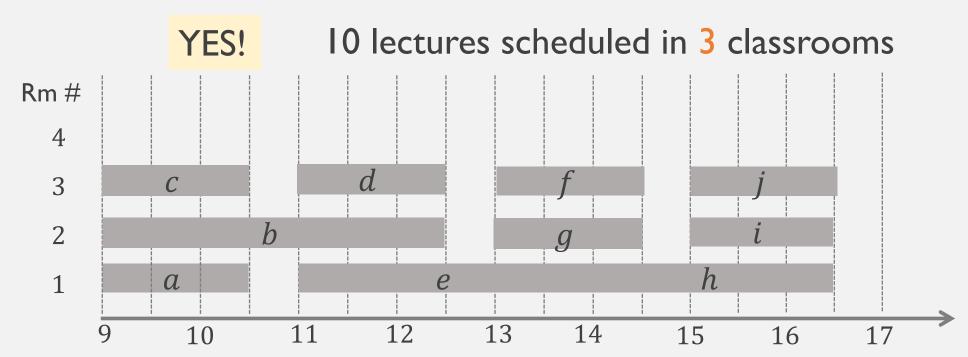
Can you do better? 10 lectures scheduled in 4 classrooms



## **Interval Partitioning Problem**

#### Scheduling classes

- Input. Lectures  $\{s_j, f_j\}$
- Output. Minimum number of classrooms to schedule all lectures so that no two occur at the same time in the same room.



## Greedy algorithm

 Idea. Sort lectures in increasing order of start time: assign lecture to any compatible classroom.

```
IntPartition({s<sub>j</sub>, f<sub>j</sub>}) // r ← 0 # of allocated rooms
1. Sort by starting time so that s_1 \le s_2 \le \cdots \le s_n
2. For j = 1, ..., n

If j compatible with some classroom k
Schedule j in room k

Else allocate new classroom r + 1
Schedule j in room r + 1
r \leftarrow r + 1

OBS.# rm needed ≥ depth of input intervals (i.e., Max. number of
```

- Running time.  $O(n \log n)$
- Optimality. #Rm allocated = depth of input intervals

lectures that overlap)