



Portland State University

W'21 CS 584/684

**Algorithm Design &
Analysis**

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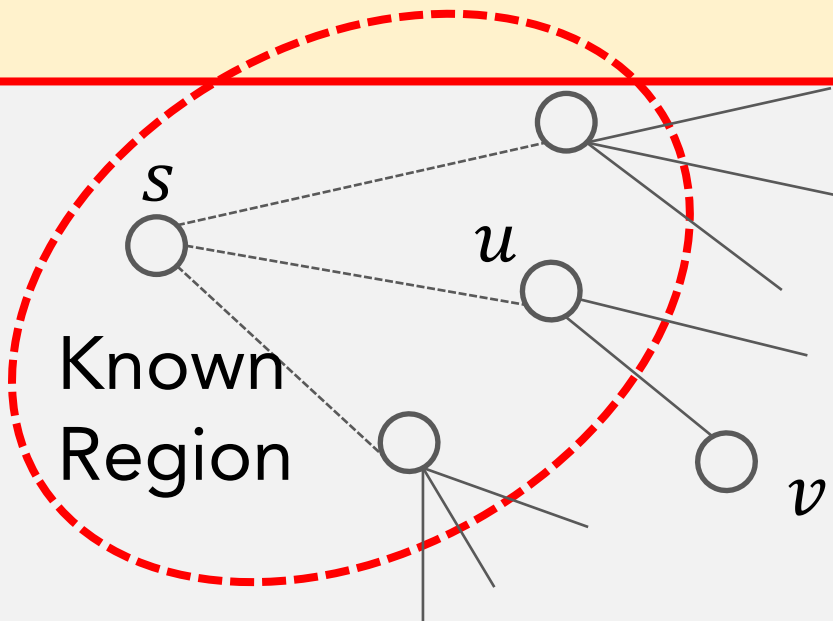
Lecture 11

- Dijkstra's algorithm cont'd
- Interval scheduling
- Minimum spanning tree

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  $d(s) = 0, d(u) = \infty, R = \emptyset$   
While  $R \neq V$   
    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) $O((m + n) \log n)$

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

- Positive weight: no need to wait; additional edges in a path do not help.

- Bellman-Ford (Dynamic programming) $O(mn)$

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

❖ 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.

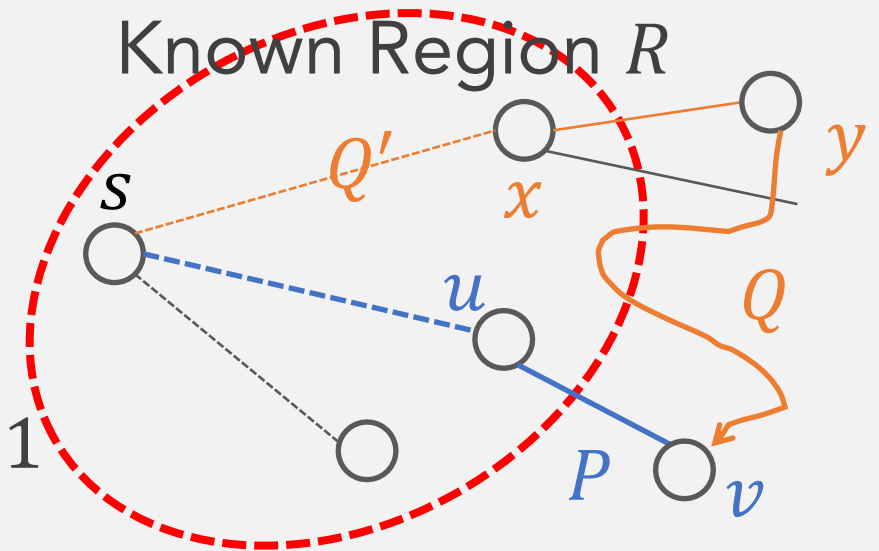
- Border Gateway Protocol ([BGP](#)).

Correctness of Dijkstra's algorithm

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 $|R| = k \geq 1$
- Show $|R| = k + 1$.
 - 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.
 - $l(Q) \geq l(Q') + l(x, y) \geq d(x) + l(x, y) \geq l(P)$; because Dijkstra's picked v in this iteration (node outside R with shortest distance to s)

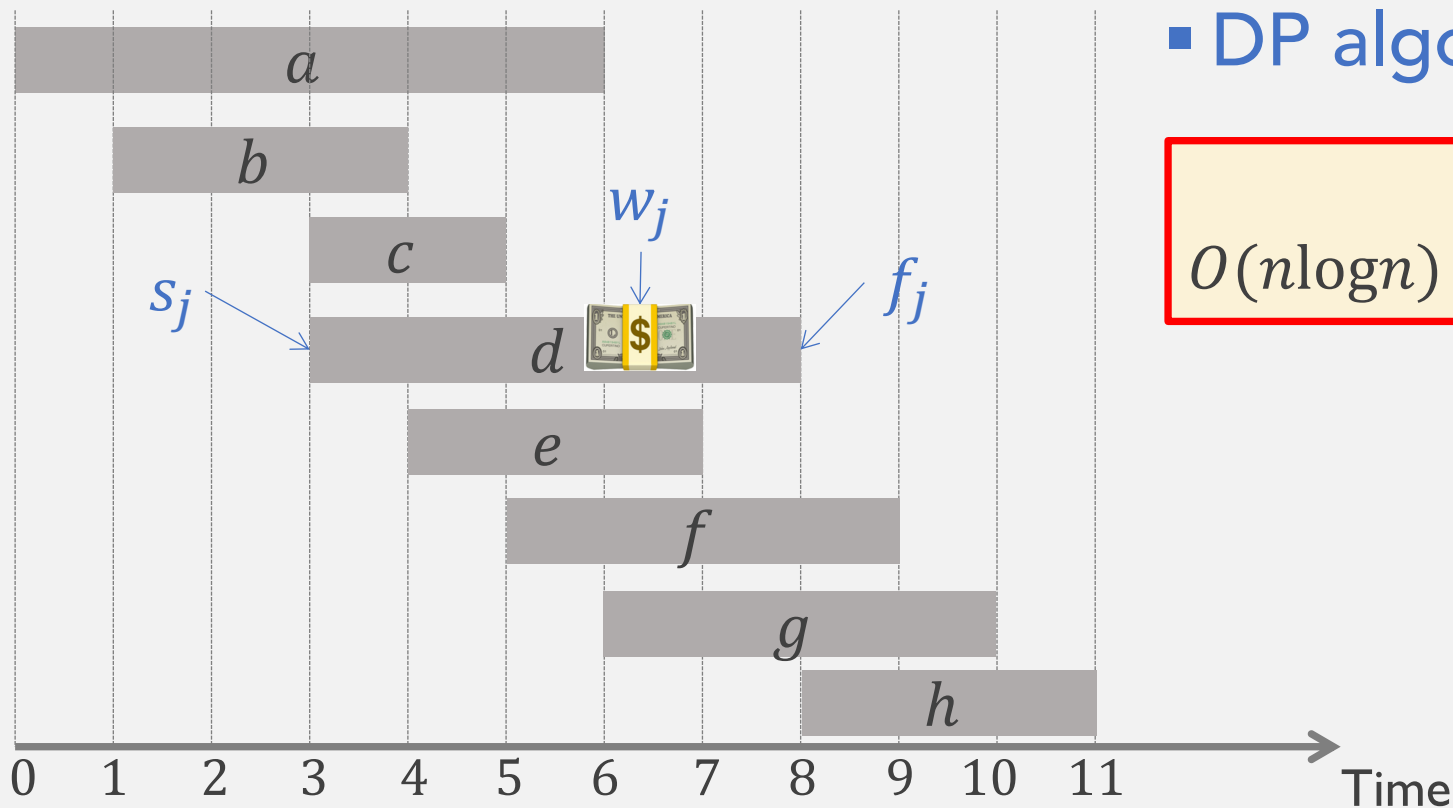


Recall: weighted interval scheduling

- **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

- DP algorithm $O(n \log n)$

Today
 $O(n \log n)$ Greedy algorithm for $w_j = 1$.



Greedy strategies

Recall. DP recurrence.

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

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

- **Greedy:** be lazy & pick the next compatible job that “looks nice”
 - **Earliest start time:** ascending order of s_j .
 - **Earliest finish time:** ascending order of f_j .
 - **Shortest interval:** ascending order of $f_j - s_j$.
 - **Fewest conflicts:** the one that conflicts the **least** number of jobs runs 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 ($\{s_j, f_j\}$)

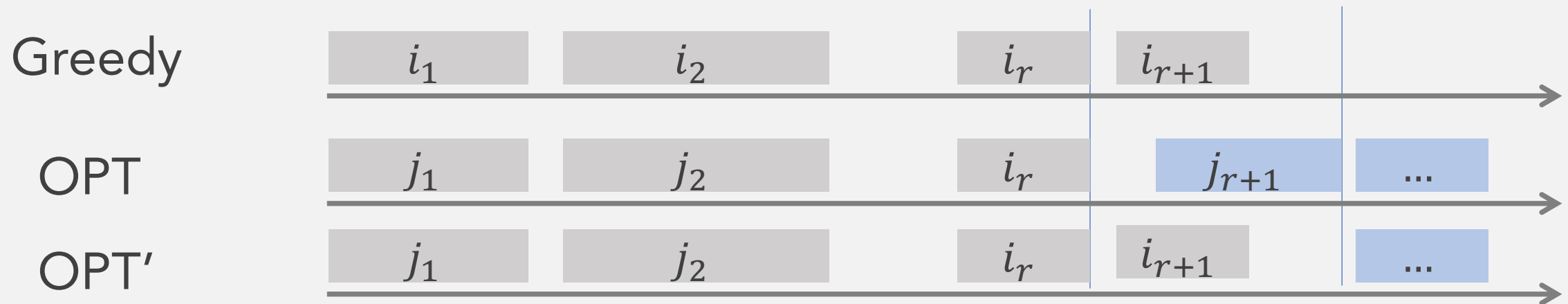
1. Sort by finishing time so that $f_1 \leq f_2 \leq \dots \leq f_n$ $\rightarrow O(n \log n)$
 2. $A \leftarrow \emptyset$ // set of selected jobs
 3. For $j = 1, \dots, n$
 - If j compatible with A
 - $A \leftarrow A \cup \{j\}$
- } $n \times O(1)$

- 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, \dots, i_k denote set of jobs selected by greedy.
- Let j_1, j_2, \dots, j_m be set of jobs in the optimal solution OPT where $i_1 = j_1, i_2 = j_2, \dots, i_r = j_r$ 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 .

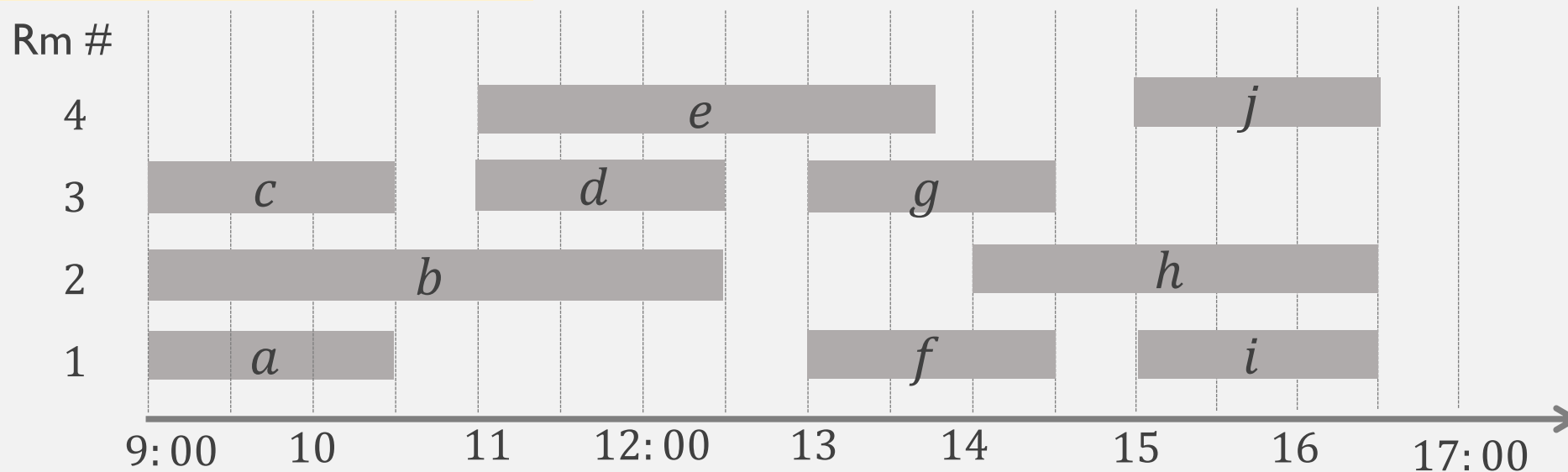


Interval Partitioning Problem

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



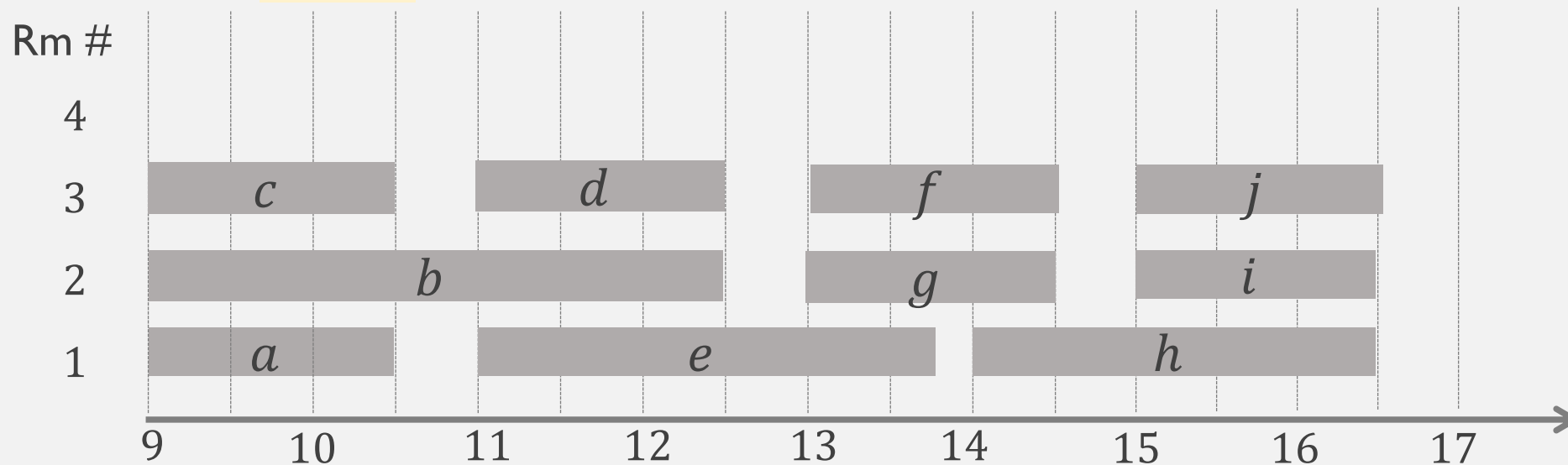
Interval Partitioning Problem

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.

YES!

10 lectures scheduled in **3** classrooms



Greedy algorithm

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

IntPartition($\{s_j, f_j\}$) // $r \leftarrow 0$ # of allocated rooms

1. Sort by **starting** time so that $s_1 \leq s_2 \leq \dots \leq s_n$

2. For $j = 1, \dots, n$

 If j compatible with some classroom k

 Schedule j in room k

 Else allocate new classroom $r + 1$

 Schedule j in room $r + 1$

$r \leftarrow r + 1$

} How to do it in $O(\log r)$

OBS. # rm needed \geq
depth of input intervals
(i.e., Max. number of
lectures that overlap)

- **Running time.** $O(n \log n)$

- **Optimality.** #Rm allocated = depth of input intervals

Minimum spanning tree (MST)

- **Input.** A connected undirected graph $G = (V, E)$.
 - **Weight function** $w: E \rightarrow \mathbb{R}$.
 - For now, assume all edge weights are **distinct**.

↳ A tree that connects all vertices

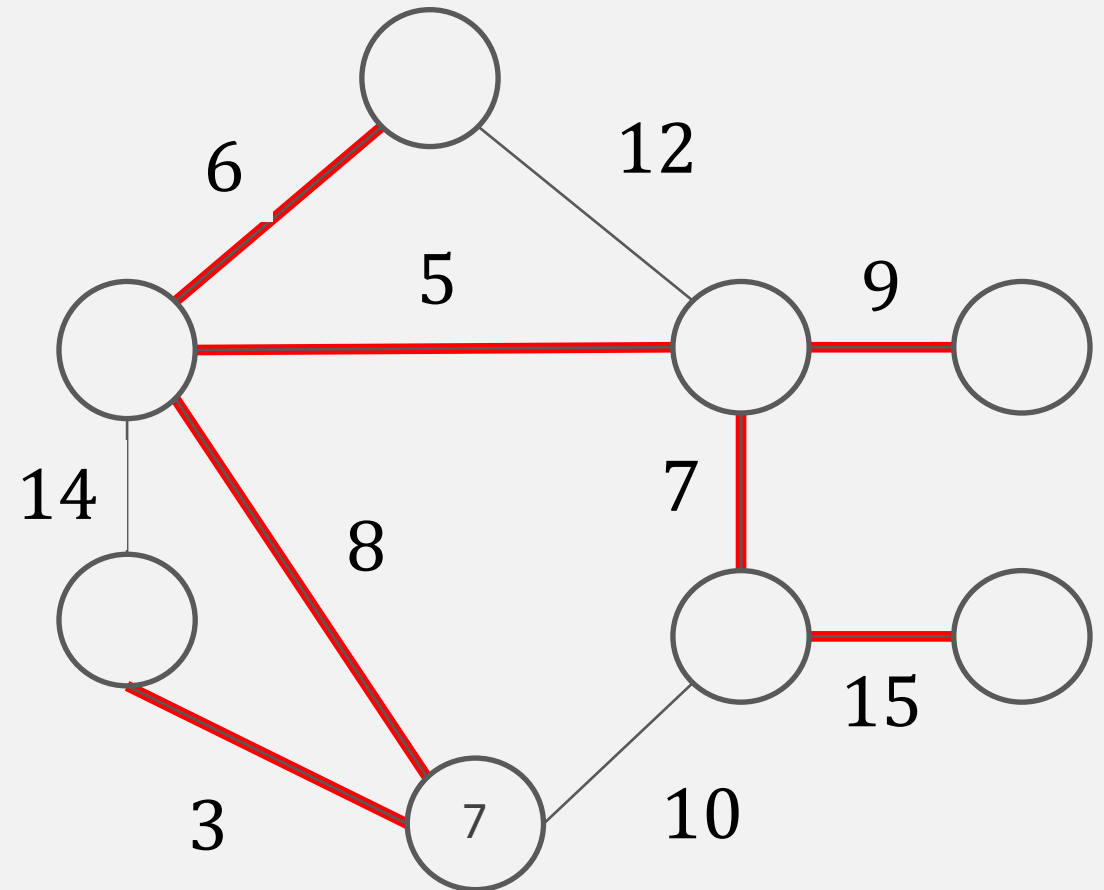
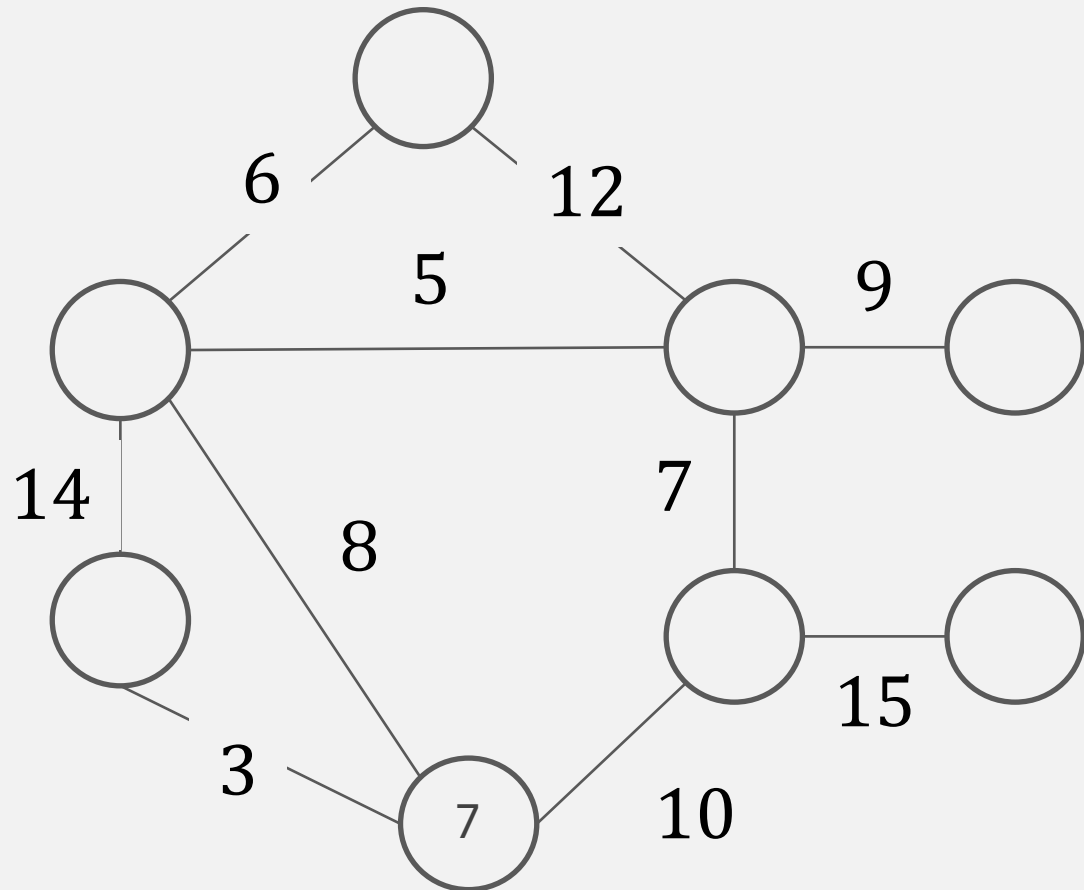
- **Output.** A **spanning tree** T of minimum weight.

$$w(T) := \sum_{(u,v) \in T} w(u, v)$$

Applications

- Cluster, Real-time face verification.
- Network design (communication, electrical, computer, road).
-

Example of MST



Pop quiz 1

Which of the following are true for all spanning trees?

- A. Contains exactly $|V| - 1$ edges
- B. The removal of any edge disconnects it
- C. The addition of any edge creates a cycle
- D. All of the above

Cayley's theorem.

The complete graph on n nodes has n^{n-2} spanning trees.
[Brute-force forbidden]

Brainstorming

Greedy strategies for computing an MST?

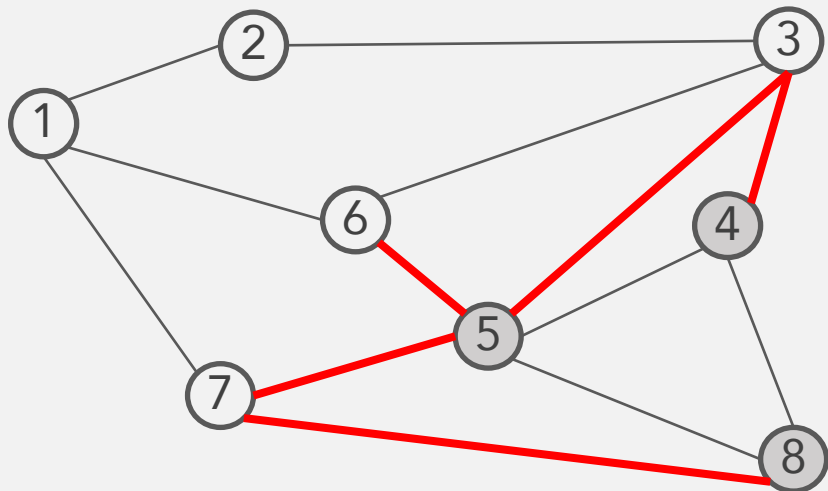
Greedy algorithms for MST

- **Kruskal's**. Start with $T = \emptyset$. Insert edges in **ascending** order of weights, unless it creates a cycle.
 - **Reverse-Delete**. Start with $T = E$. Remove edges in **descending** order of weights, unless it **disconnects** T .
- Edge-driven
- **Prim's**. Start with some **node** s . Grow a tree T from s outward. Add v to T such that $w(u, v)$ cheapest and $u \in T$.
- Node-driven
- Sounds familiar? Dijkstra's?

☺ In this extremely lucky case, all of them work! But correctness proofs are non-trivial. We need the following tools to prove them.

Cycles and cuts

- **Cycle:** set of **edges** of form $(a, b), (b, c), \dots, (z, a)$.
- **Cut:** a subset of **nodes** $S \subseteq V$.
- **Cutset $D(S)$:** subset of **edges** with **exactly** one endpoint in S .

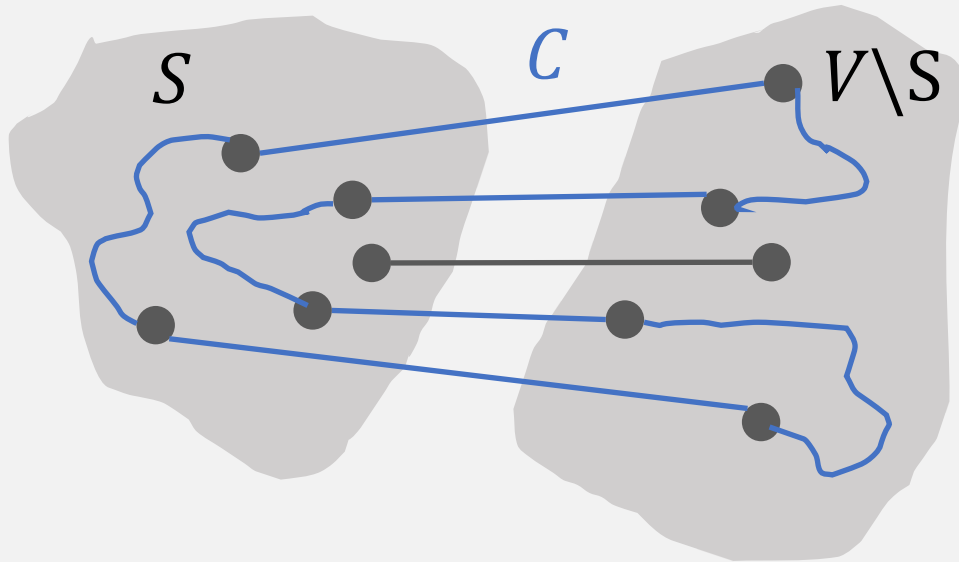


Ex. Cut $S = \{4, 5, 8\}$

Cutset $D(S) = \{(4, 3), (5, 7), (5, 6), (7, 8)\}$

Observation: cycle-cut intersection

Claim*. A cycle & a cutset intersect in an **even** number of edges.



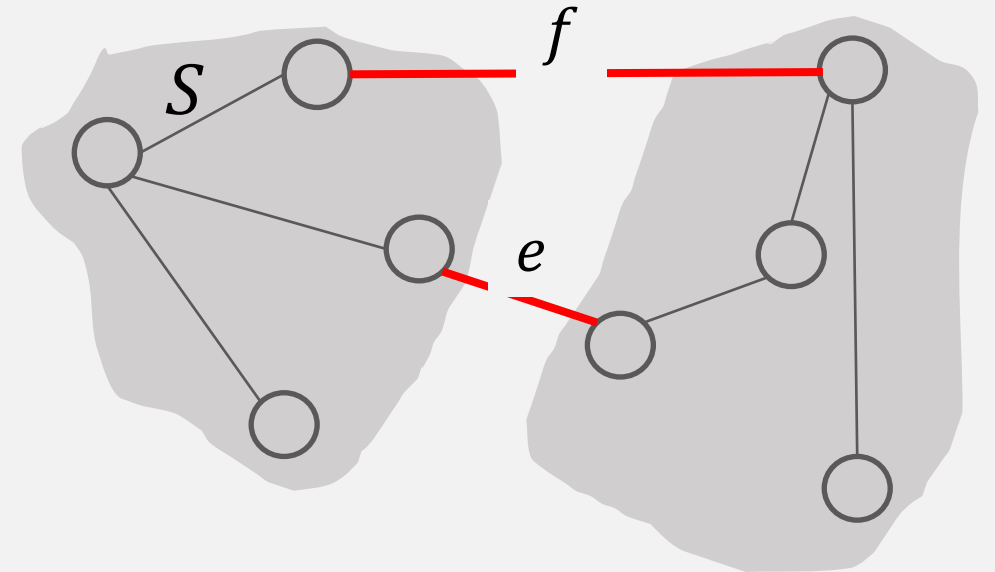
- **Proof.** A cycle has to leave & enter the cut the same number of times.

Cut Property

Cut property. Let S be a subset of nodes. Let e be the **min weight** edge with **exactly** one endpoint in S . Then any MST T contains e .

■ **Proof.** (exchange argument)

- Suppose e does not belong to T
- Adding e to T creates a cycle C
- Edge e is both in C and in the cutset $D(S)$
- there exists another edge, say f , that is in both C and D . [Claim*]
- $T' := T \cup \{e\} - \{f\}$ is also a spanning tree
- $w_e < w_f \rightarrow w(T') < w(T)$. Contradiction!



Cycle property

Cycle property. Let C be a cycle, and let f be the **max weight** edge in C . Then any MST T does not contain f .

■ **Proof.** (exchange argument)

- Suppose f belongs to T
- Deleting f creates a cut S
- Edge f is both in C and in the cutset $D(S)$
- ➔ there exists another edge, say e , that is in both C and D .
- $T' := T \cup \{e\} - \{f\}$ is also a spanning tree
- $w_e < w_f \rightarrow w(T') < w(T)$. Contradiction!

