


# LEC013 Greedy: Scheduling, MST

VG441 SS2020




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# Scheduling Problem

- We are given  $n$  jobs to schedule
- For each job  $j = 1, \dots, n$ , let
  - $w_j$  be the weight (or the importance)
  - $l_j$  be the length (or the time required)
- Define the completion time of job  $j$   
 $c_j = \text{sum of the lengths of jobs up to and including } l_j$  

Objective: to minimize the weighted sum of completion times


$$\sum_{j=1}^n w_j \cdot c_j$$

# Intuition

- If all jobs have the same length, we prefer **larger weighted** jobs to appear earlier in the order
- If all jobs have equal weights, we prefer **shorter length** jobs to appear earlier in the order

$$w_i = 1 \quad \forall i$$

1 1	2 3	3 6
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$$w_1c_1 + w_2c_2 + w_3c_3 = 1 + 3 + 6 = 10$$

$$w_i = 1 \quad \forall i$$

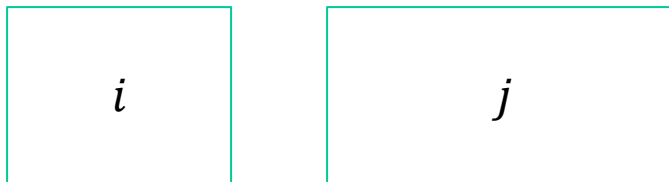
3	2	1
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$$w_1c_1 + w_2c_2 + w_3c_3 = 3 + 5 + 6 = 14$$

# Quandrum (Tricky Cases)

- What do we do in the cases where

$$l_i < l_j \text{ and } w_i < w_j$$



- Idea: give a priority score (prefers smaller length and larger weight at the same time)

$$\text{score-diff} = l_j - w_j$$

→ Guess #1

$$\text{score-ratio} = l_j / w_j$$

→ Guess #2



# Quandrum (Tricky Cases)

- Let a look at a simple example:  $l_i < l_j$  and  $w_i < w_j$



$$l_1 = 5 \text{ and } w_1 = 3$$



$$l_2 = 2 \text{ and } w_2 = 1$$

- Score-diff does not work so well...

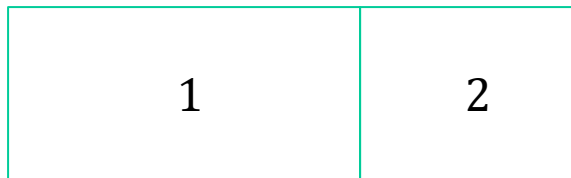
$$\text{score-diff} = l_j - w_j$$



$$\text{WC} = 2 + 21 = 23$$

- Score-ratio seems to work ✓

$$\text{score-ratio} = l_j / w_j$$



$$\text{WC} = 15 + 7 = 22$$

# Correctness Argument


- Claim: Ranking by score\_ratio is correct.
- Proof (by exchange argument):

Consider some input of  $n$  jobs, now rename the jobs according the score\_ratio, then our greedy algo picks the schedule

$$\sigma = 1, 2, 3, \dots, n \text{ with } \frac{l_1}{w_1} \leq \frac{l_2}{w_2} \leq \frac{l_3}{w_3} \leq \dots \leq \frac{l_n}{w_n}$$

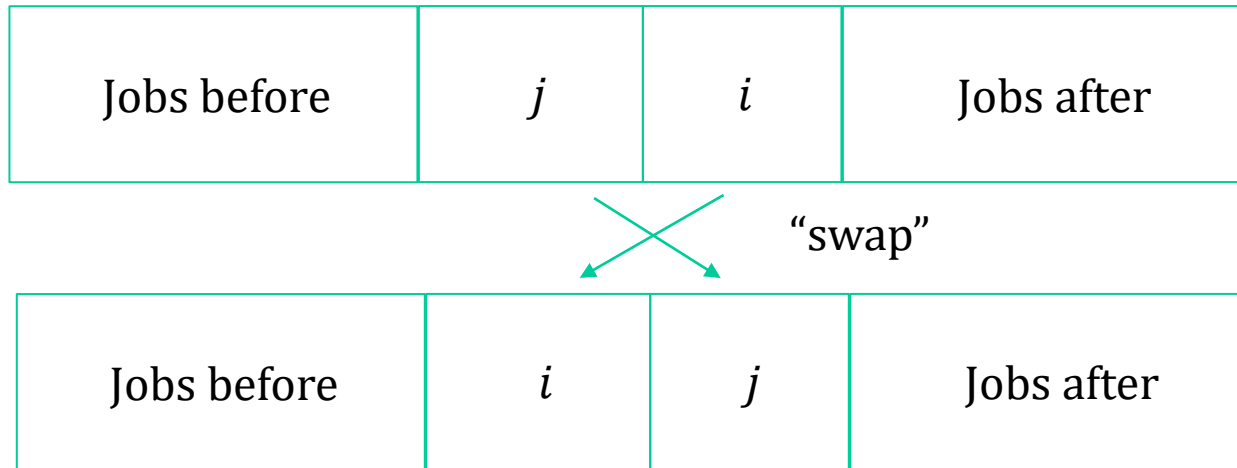
# Correctness Argument

- Consider any other schedule  $\sigma'$ , we want to show that  $\sigma$  is as good as  $\sigma'$

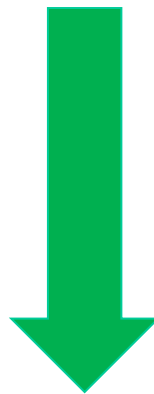
Now if  $\sigma' \neq \sigma$  then at some point in  $\sigma'$ , there is a job  $j$  right after a job  $i$  with  $j < i$  (why?) 

# Correctness Argument

$\sigma'$



$$WC^{\text{after swap}} - WC^{\text{before swap}} = w_j l_i - w_i l_j \leq 0$$



After at most  $(n \text{ choose } 2)$  steps,  
we recover our greedy  $\sigma$  with at least  
good as  $\sigma'$

$\sigma$



# MST

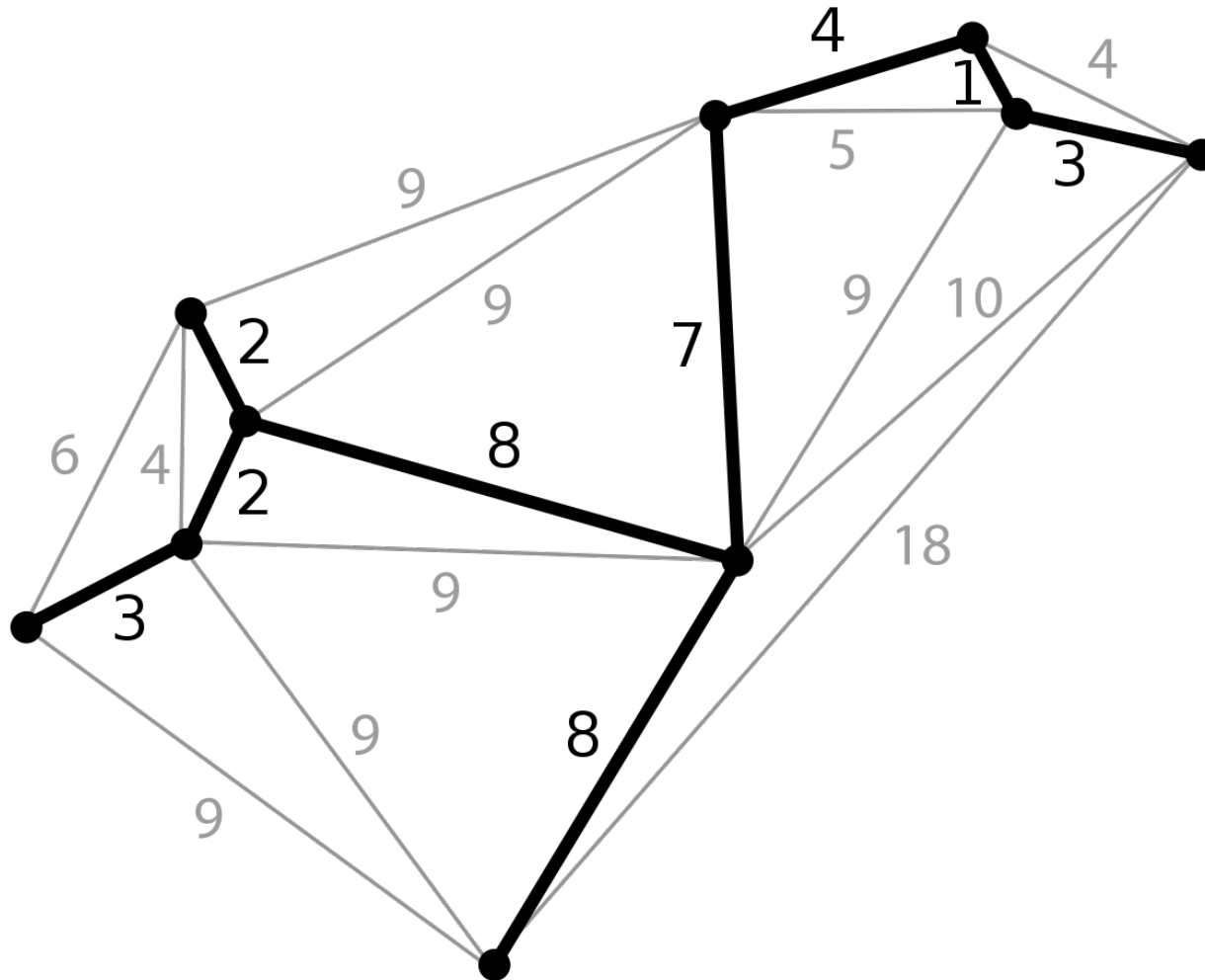
Given undirected connected graph,  $G = (V, E)$ . There is a function  $c : E \rightarrow \mathbb{R}$ , showing the cost of each edge. Assume that there are  $m$  edges in  $G$

Definition (tree):  $T = (V_T, E_T)$ , is an undirected graph in which any two vertices are connected by exactly one path. In other words, any acyclic connected graph is a tree.

Definition (a spanning tree):  $T = (V_T, E_T)$  in graph  $G = (V, E)$  is a tree with  $E_T \subseteq E$  that has all vertices covered, i.e.  $V_T = V$



# An Example of MST



# Greedy Algorithm for MST (Kruskal Algo)

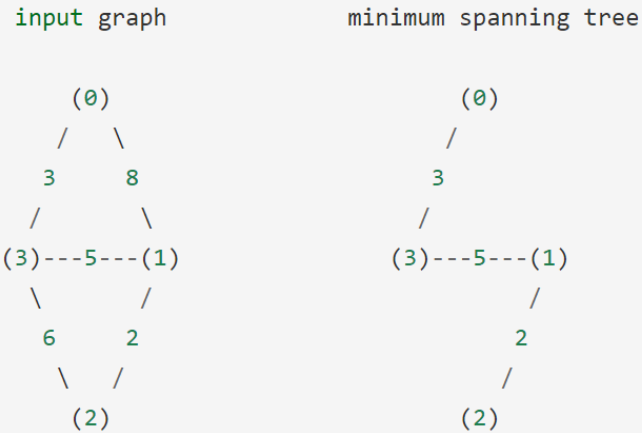
1. sort edges in non-decreasing order of cost,  $c(e_1) \leq c(e_2) \leq \dots \leq c(e_m)$
2. Let  $T_1 = \emptyset$
3. For  $i = 1, 2, \dots, m$ , if  $(T_i \cup \{e_i\})$  contains no cycle, let  $T_{i+1} = T_i \cup \{e_i\}$

We apply quicksort to assign indices to edges in the first step. We can apply a simple check of whether there is a cycle in graph  $T_i \cup e_i$  in step 3

1. Get both ends of edge  $e_i$ , vertex  $a$  and  $b$
2. Make a list  $L$  that contains all vertices connected to  $a$  in  $T_i$  🗃️
3. If vertex  $b$  is in  $L$ , there will be a cycle in  $T_i \cup \{e_i\}$ . If  $b$  is not in  $L$ , then  $T_i \cup \{e_i\}$  is acyclic.

Note that the algorithm above takes  $O(m)$  time. So the overall greedy algorithm runs in polynomial time.

# Python Code

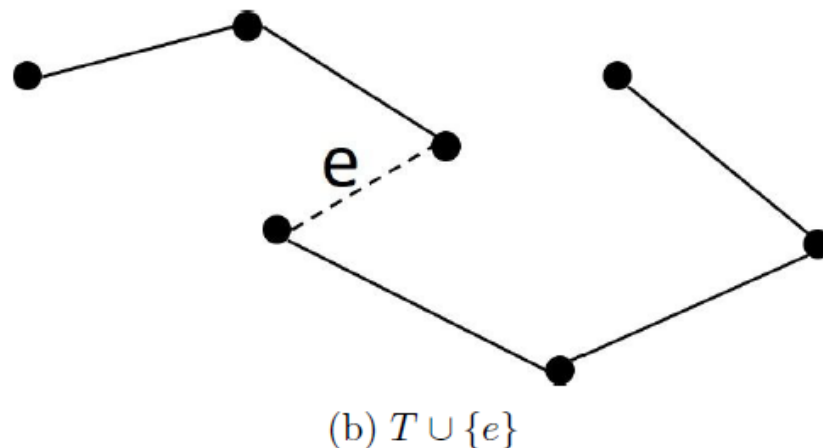
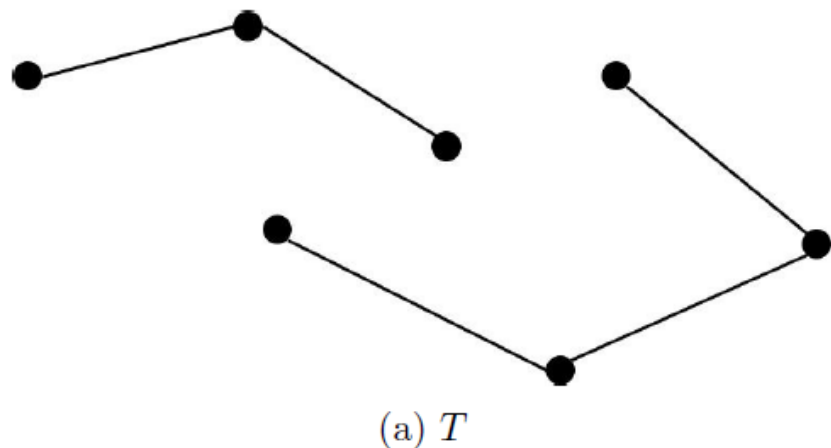


```
>>> from scipy.sparse import csr_matrix
>>> from scipy.sparse.csgraph import minimum_spanning_tree
>>> X = csr_matrix([[0, 8, 0, 3],
...                 [0, 0, 2, 5],
...                 [0, 0, 0, 6],
...                 [0, 0, 0, 0]])
>>> Tcsr = minimum_spanning_tree(X)
>>> Tcsr.toarray().astype(int)
array([[0, 0, 0, 3],
       [0, 0, 2, 5],
       [0, 0, 0, 0],
       [0, 0, 0, 0]])
```

# Analysis

**Lemma:** The algorithm gives a tree  $T$  that covers all vertices in  $G$ .

**Proof:** Given the graph is connected, we can prove this property by contradiction. If in the tree  $T$  generated by algorithm, not all the vertices are connected, there exist an edge  $e \in E$  between two disconnected components of  $T$  (see Figure 1 a). Because  $T$  is acyclic, and two components are not connected, we can declare that  $T \cup \{e\}$  is also acyclic. Then according to the algorithm,  $e$  must have been included.

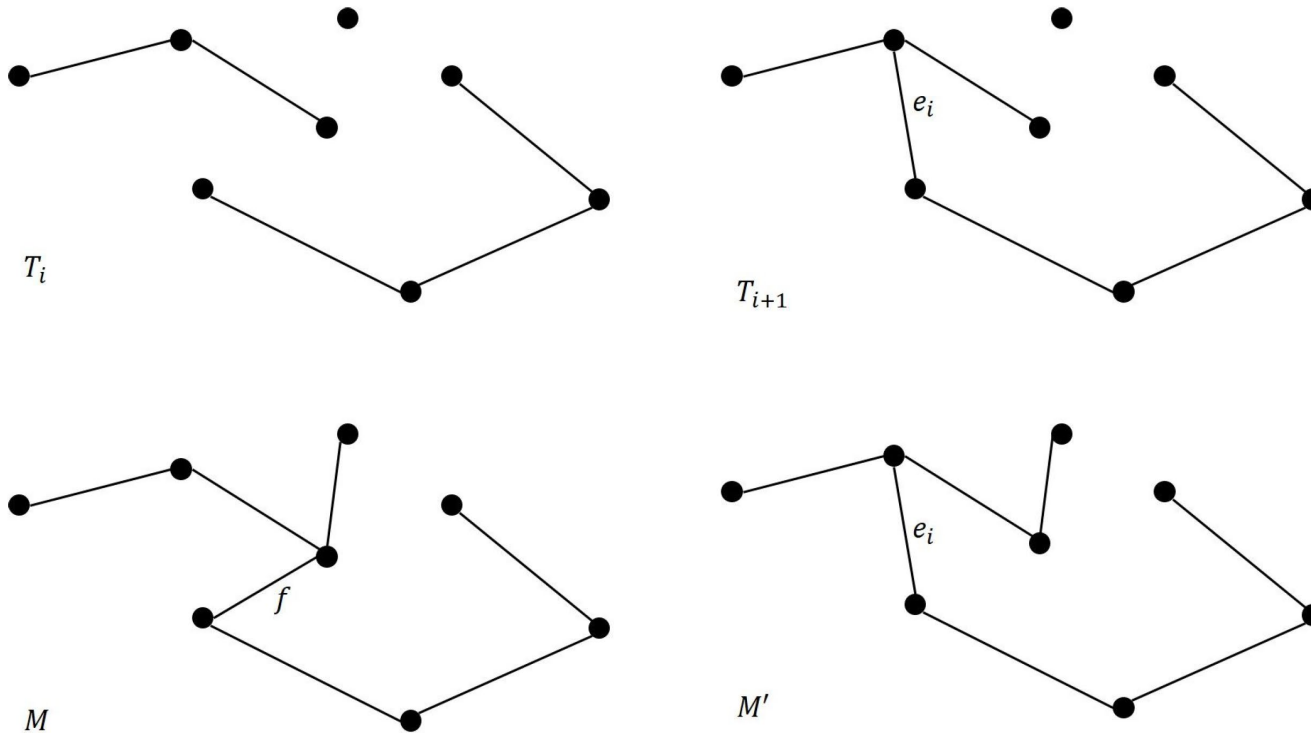


# Analysis

Let  $i = 1, 2, \dots, m, m + 1$  denote the iteration in the algorithm. Let  $T_i$  denote the solution at the beginning of  $i^{th}$  iteration.

**Lemma:** For each iteration  $i$ , there is a minimum spanning tree  $M$  with  $T_i \subset M$

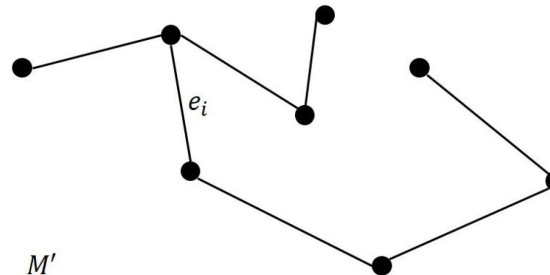
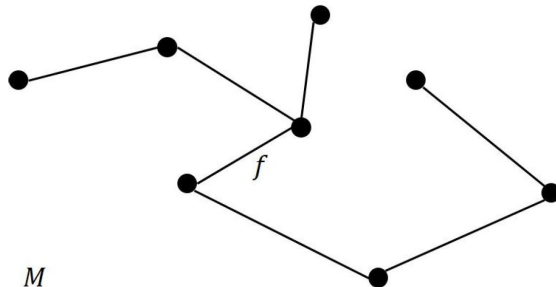
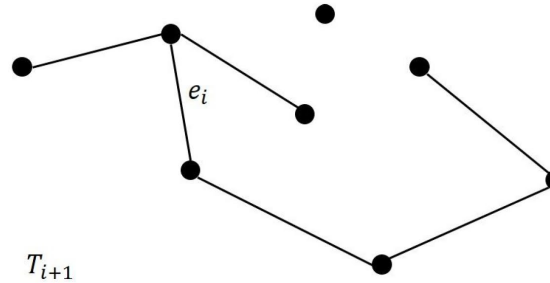
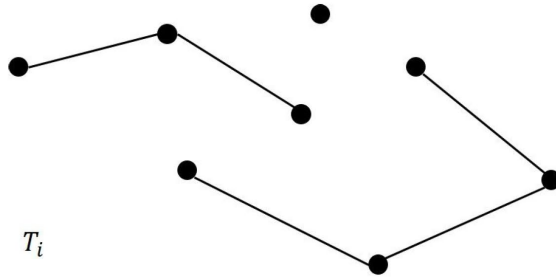
Quandrum (tricky case) Suppose  $T_i \subseteq M$  but  $T_{i+1} = T_i \cup \{e_i\} \not\subseteq M$



$$\text{cost}(M') = \text{cost}(M \setminus \{f\} \cup \{e_i\}) = \text{cost}(M) - \text{cost}(f) + \text{cost}(e_i) \leq \text{cost}(M)$$

# Analysis

Suppose  $T_i \subseteq M$  but  $T_{i+1} = T_i \cup \{e_i\} \not\subseteq M$



$$\text{cost}(M') = \text{cost}(M \setminus \{f\} \cup \{e_i\}) = \text{cost}(M) - \text{cost}(f) + \text{cost}(e_i) \leq \text{cost}(M)$$

- $M \cup e_i$  must contain a cycle (if not, then  $M$  is disconnected.)
- There must be an edge  $f$  in this cycle that is costlier than  $e_i$  (if not, then the greedy algorithm would not pick  $e_i$ )