

# Greedy Algorithms

# The Traveling Salesperson Problem

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NEARESTNEIGHBOR( $G = (V, E)$ )

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**Input:** A complete, weighted graph  $G$

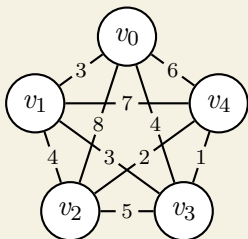
**Output:** The weight of a Hamiltonian cycle in  $G$

- 1: **for** all  $v \in V$  **do**
  - 2:     **let**  $v$  be “unexplored”
  - 3: **let**  $s = v$  be “explored”, any vertex in  $V$ , and  $W$  be 0
  - 4: **while** there exist “unexplored” vertices in  $V$  **do**
  - 5:     **let**  $e = (v, u)$  be the lightest edge such that  $u$  is “unexplored”
  - 6:     **let**  $v$  be  $u$  and  $W$  be  $W + w_e$
  - 7: **return**  $W + w_e$ , where  $e = (v, s)$
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# The Traveling Salesperson Problem

## Example

Given:



...the Hamiltonian cycle of minimum weight is  $(v_0, v_1, v_2, v_4, v_3, v_0)$ , weight 14. NEARESTNEIGHBOR returns  $(v_0, v_1, v_3, v_4, v_2, v_0)$ , weight 17, and does *not* solve the Traveling Salesperson Problem.

# Greedy Algorithms

## Definition

A **greedy** algorithm is one that assumes a locally optimal choice will always lead to the globally optimal solution.

- A divide and conquer approach may not improve complexity.
- A greedy approach may be *incorrect*.

## Example

SHORTESTPATH (correctly) assumes that at least one of the lightest paths from a vertex will always traverse the lightest available path.

## Example

NEARESTNEIGHBOR (incorrectly) assumes that the lightest Hamiltonian cycle will always traverse the lightest incident edge.

# Minimum Spanning Trees

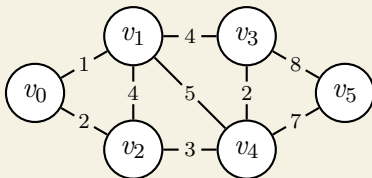
# Minimum Spanning Trees

Consider the following problem:

- Given a connected, weighted graph, find an “MST”, a spanning tree of minimum weight.

## Example

Given:



...the minimum spanning tree has weight 15.

# Minimum Spanning Trees

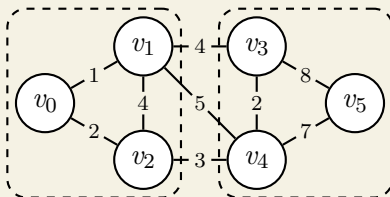
## Definition

A **cut** is a partitioning of every vertex into two disjoint sets.

- Given a cut, its **cut set** is the set of all edges connecting vertices in different sets.

## Example

Given:



...the cut set is  $\{(v_1, v_3), (v_1, v_4), (v_2, v_4)\}$ .

# Minimum Spanning Trees

## The Cut Property

If  $e$  is the edge of strictly minimum weight in a cut set, then  $e$  must be in every MST.

- If lighter edges are considered first, then an edge that would connect previously unreachable vertices is always optimal.

## The Cycle Property

If  $e$  is the edge of strictly maximum weight in a cycle, then  $e$  cannot be in any MST.

- If heavier edges are considered last, then an edge that would connect already reachable vertices is never optimal.



# Kruskal's Algorithm

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MINIMUMSPANNINGTREE( $G = (V, E)$ )

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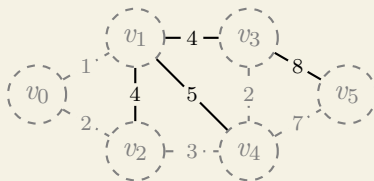
**Input:** A connected, weighted graph  $G$

**Output:** A minimum spanning tree of  $G$

- 1: **let**  $T$  be an empty tree and  $S$  be an empty disjoint set
  - 2: **for** all  $v \in V$  **do**
  - 3:     Make a new subset containing  $v$  in  $S$
  - 4: **let**  $A$  be  $E$ , sorted in increasing order by weight
  - 5: **for** all  $e = (u, v) \in A$  **do**
  - 6:     **if**  $u$  and  $v$  are in different subsets in  $S$  **then**
  - 7:         Union the subsets containing  $u$  and  $v$  in  $S$
  - 8:         Add  $e$  to  $T$
  - 9: **return**  $T$
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# Kruskal's Algorithm

## Example



1 Add  $e = (v_0, v_1)$ ,  $w_e = 1$ .

2 Add  $e = (v_0, v_2)$ ,  $w_e = 2$ .

3 Add  $e = (v_3, v_4)$ ,  $w_e = 2$ .

4 Add  $e = (v_2, v_4)$ ,  $w_e = 3$ .

5 Skip  $e = (v_1, v_2)$ ,  $w_e = 4$ .

6 Skip  $e = (v_1, v_3)$ ,  $w_e = 4$ .

7 Skip  $e = (v_1, v_4)$ ,  $w_e = 5$ .

8 Add  $e = (v_4, v_5)$ ,  $w_e = 7$ .

9 Skip  $e = (v_3, v_5)$ ,  $w_e = 8$ .

# Kruskal's Algorithm

## Example

- Sorting  $E$  has complexity  $O(|E| \log |E|)$ .
- Each vertex appears in the disjoint set exactly once.
- Each edge causes a disjoint set operation at most thrice.

Assuming a tree-based implementation of the disjoint set, `MINIMUMSPANNINGTREE` has complexity  $O(|E| \log |E|)$ .

- Disjoint sets can be implemented as trees: two nodes have the same root if and only if their elements are in the same subset.
- With a hash table implementation of the disjoint set, Kruskal's algorithm has complexity  $O(|V| |E|)$ .

# Exchange Arguments

## Theorem

Suppose `MINIMUMSPANNINGTREE` returns  $T$  and there exists an optimal MST  $\text{OPT}$ .

## Lemma

If  $T$  differs from  $\text{OPT}$  by  $i$  edges, then  $\text{OPT}$  can be transformed into  $T$  without increasing its weight.

Then  $T$  and  $\text{OPT}$  have the same weight.

- Note that  $T$  need not include the same edges as  $\text{OPT}$ ; it need only have the same sum total weight.
- Since the edges of  $\text{OPT}$  can be “exchanged” to equal those of  $T$  without changing their weight,  $T$  has the same weight.



## Activity Selection

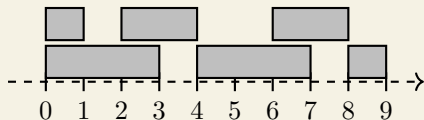
## Activity Selection

Consider the following problem:

- An **activity** is a pair  $a_i = (s_i, f_i)$  of a start time  $s_i$  and a finish time  $f_i$ , taking place during the time  $[s_i, f_i)$ .
- Given a set of activities, find a maximum subset of non-conflicting activities.

### Example

Given  $\{(2, 4), (0, 3), (0, 1), (4, 7), (6, 8), (8, 9)\}$ :



...one of the maximum subsets is  $\{(0, 1), (2, 4), (4, 7), (8, 9)\}$ .

## Activity Selection

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ACTIVITYSELECTION( $S = \{a_0, a_1, \dots, a_{n-1}\}$ )

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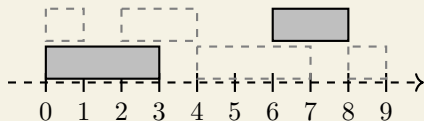
**Input:** A finite, non-empty set  $S$  of  $n$  activities, where  $a_i = (s_i, f_i)$

**Output:** A maximum subset of non-conflicting activities

- 1: **let**  $A$  be  $S$ , sorted in increasing order by finish time
  - 2: **let**  $a_j = (s_j, f_j)$  be the first activity in  $A$  and  $X$  be  $\{a_j\}$
  - 3: **for** all  $a_i = (s_i, f_i) \in A$  **do**
  - 4:     **if**  $s_i \geq f_j$  **then**
  - 5:         **let**  $X$  be  $X \cup \{a_i\}$
  - 6:         **let**  $a_j$  be  $a_i$
  - 7: **return**  $X$
-

# Activity Selection

## Example



- |                  |                  |                  |
|------------------|------------------|------------------|
| 1 Select (0, 1). | 3 Select (2, 4). | 5 Skip (6, 8).   |
| 2 Skip (0, 3).   | 4 Select (4, 7). | 6 Select (8, 9). |

## Example

- Sorting  $S$  has complexity  $O(n \log n)$ .
- Collectively considering activities has complexity  $O(n)$ .

ACTIVITYSELECTION has complexity  $O(n \log n)$ .



# Greedy Stays Ahead

## Theorem

Suppose ACTIVITYSELECTION returns  $X = (x_0, x_1, \dots, x_{n-1})$  and there exists an optimal selection  $\text{OPT} = (y_0, y_1, \dots, y_{m-1})$ , both sorted in increasing order of finish time.

## Lemma

For all  $i < n$ ,  $f_{x_i} \leq f_{y_i}$ .

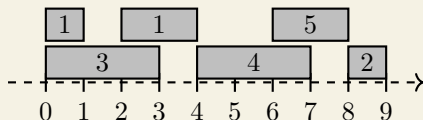
Then  $n = m$ .

- Note that  $X$  need not include the same activities as  $\text{OPT}$ ; it need only have as many activities as  $\text{OPT}$ .
- By “staying ahead” of  $\text{OPT}$ ,  $X$  always has the option of including the same activities.

# Weighted Activity Selection

## Example

Given:



...ACTIVITYSELECTION returns  $\{(0, 1), (2, 4), (4, 7), (8, 9)\}$ , which is a subset of maximum cardinality but not maximum weight.