DSC40B: Theoretical Foundations of Data Science II

Lecture 16: Minimum Spanning Tree, properties, and general greedy algorithms

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Previously

- Given directed or undirected graphs
 - Graph search / traversal strategies (DFS / BFS)
 - Single source shortest paths in weighted graphs
 - Bellman-Ford algorithm for general graphs
 - Dijkstra algorithm for graphs with positive edge weights

- ► Today:
 - ▶ Computing a minimum spanning tree (MST) of an undirected graph



Trees, spanning trees, and minimum spanning tree



Trees

- An undirected graph G = (V, E) is a tree if and only if
 - (i) it is connected; and
 - (ii) it is acyclic (i.e., does not contain any cycle)
- Claim [Tree Edges]:
 - If T = (V, E) is a tree, then we have that |E| = |V| 1

Alternative definition

Alternative definition:

An undirected graph G = (V, E) is a tree if and only if that (i) it is connected; and (ii) |E| = |V| - 1

A tree

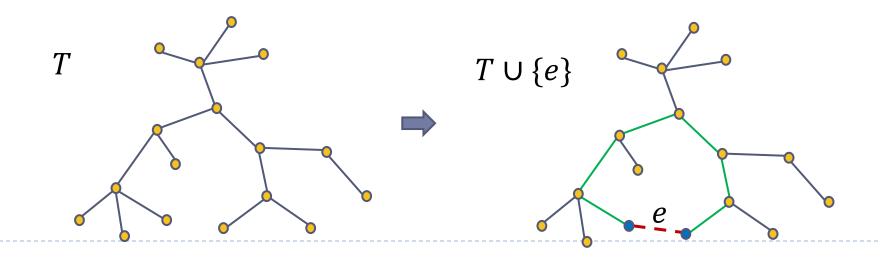
A tree

NOT a tree

NOT a tree

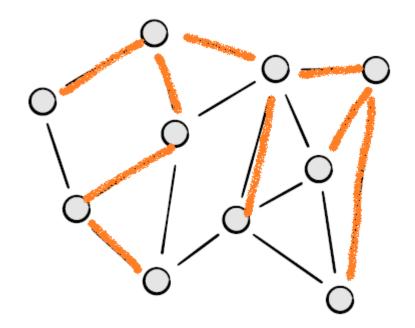
Remarks

- Key properties: If T = (V, E) is a tree,
 - \triangleright there is a unique path between any two nodes in V
 - lacktriangle adding any other edge e to T will create a unique cycle containing e
 - ▶ i.e., $T \cup \{e\}$ contains a cycle for any $e \notin T$
 - removing an edge from T will disconnect it
- Dut of all connected graphs on n nodes, a tree has least number (i.e., n-1) of edges



Spanning Tree

▶ Given an undirected graph G = (V, E), a spanning tree of G is any graph $T = (V, E' \subseteq E)$ that is a tree.



Example of spanning trees for the graph on the right.



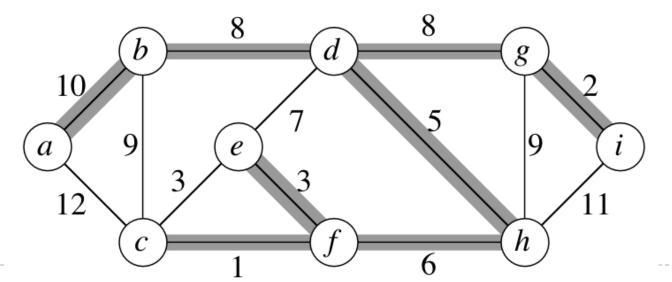
Spanning Tree

- ▶ Given an undirected graph G = (V, E), a spanning tree of G is any graph $T = (V, E' \subseteq E)$ that is a tree.
- Intuitively, a spanning tree of G contains smallest number of edges in E to connect all nodes in G.
- lacktriangle Note that if the input graph G is not connected, then there exists no spanning tree.
 - We can talk about spanning forest, consisting a set of spanning trees, one for each connected component in G.



Minimum spanning tree (MST)

- Weight of spanning tree T of a weighted graph G = (V, E) is
 - the total weights of all edges in T, i.e., $\omega(T) = \sum_{e \in T} \omega(e)$,
 - where $\omega: E \to R$ is the edge weights associated to G.
- A minimum spanning tree (MST) of a weighted graph G = (V, E) is a spanning tree with smallest possible weight.



Weight of this spanning tree: 43

Turns out this is also a minimum spanning tree.

MSTs

- MST may not be unique
- ▶ All MSTs of a given graph G = (V, E) have the same number of edges!
 - They all have |V| 1 number of edges
- If all edges in input graph have the same weight, then how can we find a MST for it?
 - Any spanning tree of it is a minimum spanning tree!

Exercise:

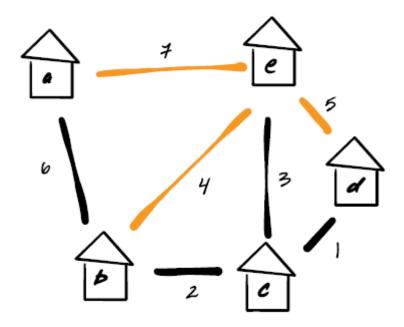
Design an algorithm to compute an MST for a graph where all edges have weight 1



Motivation and properties of MST



Motivation

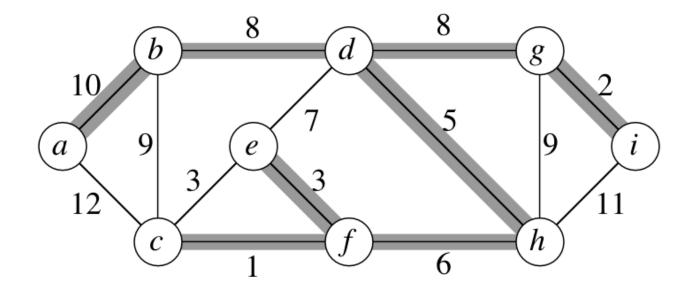


- Among all possible road segment choices, build a set of road segments so that all houses are connected and the total cost is minimized
- Solution: Find the MST of the input weighted graph where edge weight represents the cost of build that road segment.



MST Problem

- Input:
 - ightharpoonup a weighted undirected graph G
- Output:
 - \blacktriangleright the set of edges in a MST of G

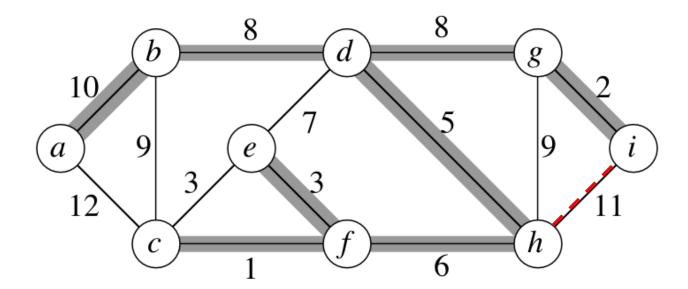




Key property

Key property of MST:

- Given a MST T of G = (V, E), let $e \in E$ be any edge in E but not in T. The following then holds:
 - ▶ there is a unique cycle C containing e in $T \cup \{e\}$.
 - ightharpoonup e has the largest weight among all edges in this cycle C.





Key property

Key property of MST:

- Given a MST T of G = (V, E), let $e \in E$ be any edge in E but not in T. The following then holds:
 - \blacktriangleright there is a unique cycle C containing e in $T \cup \{e\}$.
 - \triangleright e has the largest weight among all edges in this cycle C.

Proof sketch:

- If e does not have largest weight, let $e' \in C$ be an edge with largest weight in C.
 - $T' = T \{e'\} + \{e\}$ is also a spanning tree of G
 - $\blacktriangleright weight(T') \leq weight(T) \Rightarrow T \text{ cannot be MST.}$
 - ▶ Contradiction \Rightarrow *e* must have largest weight in *C*.

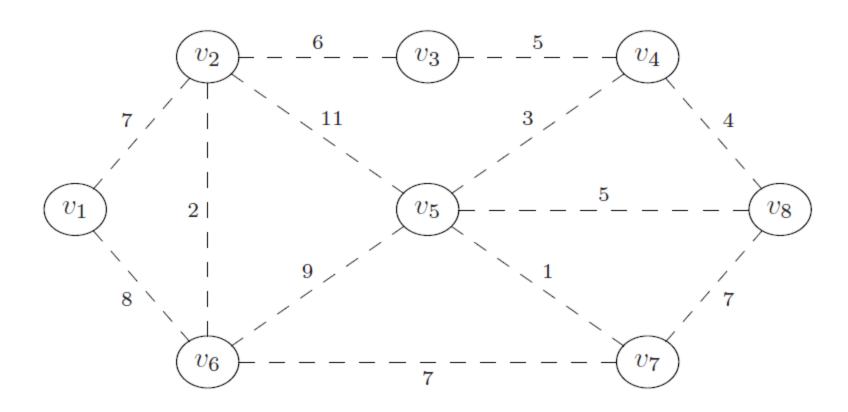
First greedy algorithm for MST: Prim's algorithm



General greedy idea:

- Input:
 - ▶ a weighted undirected graph G = (V, E), with $\omega: E \to R$
- Output:
 - \blacktriangleright the set of edges in a MST T of G
- ▶ A MST T consist of V-1 number of edges that connect all nodes, with no cycle.
- Intuitively, we will grow the tree edge-by-edge, and choose "safe" edges greedily to incrementally build T
 - such that any time, the edges we choose will form a part of some MST

Example



What is a "safe" edge to add first?

Two greedy algorithms

- Two greedy algorithms
 - ▶ Today: Prim's algorithm
 - Next class: Kruskal's algorithm
 - ▶ They differ in the order of edges they visit and thus ``safe' edges they add

Idea for Prim's algorithm

► Input:

- ▶ a weighted undirected graph G = (V, E), with $\omega: E \to R$
- Output:
 - the set of edges in a MST T of G

Intuitively,

- Incrementally grow a partial tree $T(S) \subseteq E$ connecting a subset of nodes $S \subseteq V$
- At the beginning of each iteration, T(S) is a sub-tree of some MST of G
- At each iteration, grow T(S') to include one more vertex $S' = S \cup \{u\}$
 - ▶ such that T(S') is still a sub-tree of some MST of G
 - the new node is reached via a greedy choice of a crossing-edge
 - in particular, the greedy choice is the minimum weight edge connect some node in S to some node in U = V S (i.e., outside S)

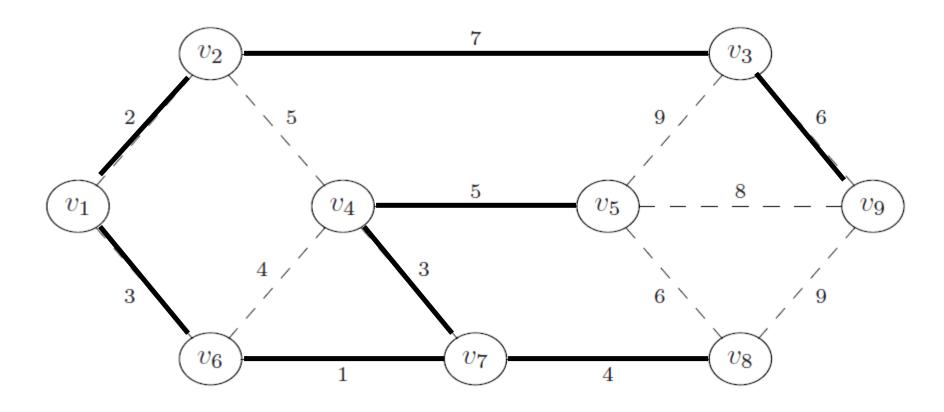


High level outline (not code)

```
procedure PrimMST(G)
1 U \leftarrow V(\mathsf{G}) - \{v_1\}; /* V(\mathsf{G}) = set \ of \ vertices \ of \ graph \ G */
2 v_1. predecessor \leftarrow \mathbf{NULL};
3 while (U \neq \emptyset) and (\exists \text{ edge from } (V(\mathsf{G}) - U) \text{ to } U) do
        (v_i, v_j) \leftarrow \text{minimum weight edge from } V(\mathsf{G}) - U \text{ to } U;
        v_j. predecessor \leftarrow v_i; U \leftarrow U - \{v_j\};
7 end
```

- ▶ *U* : unconnected vertices
- $\triangleright S = V U$: vertices connected by current partial tree

Example



ightharpoonup Suppose we grow the tree starting from v_1



Correctness

MST Theorem:

Let T be a sub-tree of a minimum spanning tree. If e is a minimum weight edge connecting T to some vertex not in T, then $T \cup \{e\}$ is a subtree of a minimum spanning tree.

- ▶ Key to the correctness of PrimMST algorithm.
 - ► Loop invariant:

each time PrimMST() algorithm grows the partial tree (i.e, adds another edge to it), the invariant is that the new tree is still a subtree of <u>some</u> minimum spanning tree of input graph G.

▶ Termination:

when all nodes are connected, we obtain a MST of G.

(or if we cannot reach all nodes, then the input graph is not connected)



Idea for proving loop invariant

- ▶ By the theorem's hypothesis, *T* is a subtree of some MST *A* of *G*.
- If e is not an edge of A, then $A \cup \{e\}$ contains a cycle.
- Let C be this cycle. There must exists some edge $e' \in C$ from T(S) to a vertex not in S (those vertices already connected).
- Since e is a minimum weight edge from vertices in T to vertices not in T, $weight(e) \le weight(e')$.
- ▶ Replacing $e' \in A$ by e gives a new tree $B = A \{e'\} + \{e\}$ such that $weight(B) \le weight(A)$.
- ▶ $T \cup \{e\} \subseteq B$. So $T \cup \{e\}$ is also a subtree of some MST.
- Done.



Implementation of Prim's algorithm



Naïve implementation of Prim's Alg

```
procedure PrimMST(G)
1 U \leftarrow V(\mathsf{G}) - \{v_1\}; /* V(\mathsf{G}) = set \ of \ vertices \ of \ graph \ G */
2 v_1. predecessor \leftarrow \mathbf{NULL};
 3 while (U \neq \emptyset) and (\exists \text{ edge from } (V(\mathsf{G}) - U) \text{ to } U) do
         (v_i, v_j) \leftarrow \text{minimum weight edge from } V(\mathsf{G}) - \overline{U} \text{ to } U;
 \begin{array}{c|c} \mathbf{5} & v_j. \ \text{predecessor} \leftarrow v_i; \\ \mathbf{6} & U \leftarrow U - \{v_j\}; \end{array}
```

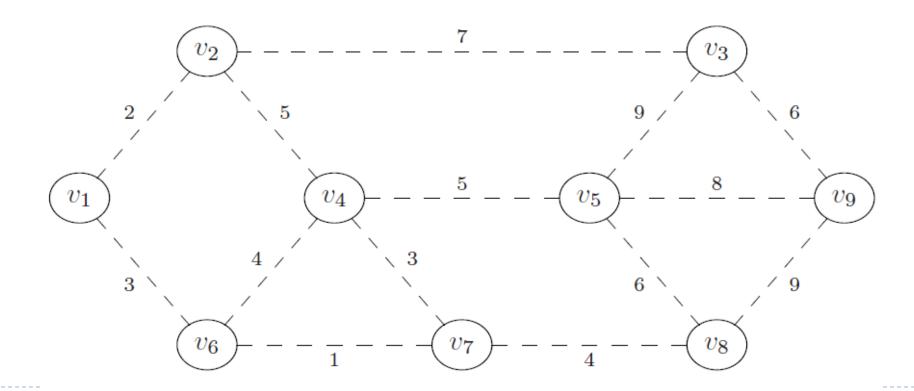
- Naïve implementation: linear scan all edges to identify minweight edge (v_i, v_j) at each iteration
- ▶ Total time complexity: O(VE)



First improvement

Storing costs at nodes

Each unvisted nodes v in U maintain v. cost, which is the smallest weight of any edge from v to visited nodes in S



Outline of first improvement

```
procedure PrimMST(G)
1 U \leftarrow V(\mathsf{G});   /*V(\mathsf{G}) = set\ of\ vertices\ of\ graph\ G\ */
 2 foreach v_i \in V(G) - \{v_1\} do v_i.cost \leftarrow \infty;
 3 v_1.cost \leftarrow 0;
 4 v_1. predecessor \leftarrow NULL;
 5 while (U \neq \emptyset) do
       v_i \leftarrow v_i \in U with minimum v_i.cost;
        U \leftarrow U - \{v_i\};
                                                /* Remove v_i from U
       /*(v_i, v_i) predecessor) is an MST edge
       foreach edge (v_i, v_k) incident on v_i do
            if (v_k \text{ is in } U \text{ and weight}(v_i, v_k) < v_k.\text{cost}) then
                 v_k. predecessor \leftarrow v_i;
10
               v_k.cost \leftarrow weight(v_i, v_k);
11
            end
12
        end
13
14 end
```

- If we use linear scan to find the outside node with minimum cost for Line 6 in the algorithm in previous slide, then the entire algorithm takes $O(V^2)$ time.
 - Line 6 takes O(V) time
 - Lines 8-13 takes $\Theta(\deg(v_i)) = O(V)$
 - Hence each iteration of the while-loop takes O(V) time
 - ▶ The while-loop runs *V* iterations
 - Hence total time complexity is $O(V^2)$

Better implementation

- Similar to Dijkstra algorithm, we can use priority-queue to significantly speed up the time complexity!
- In particular, we need a data structure to maintain the costs of unvisited nodes, which supports:
 - deleting the node with minimum cost (.extract_min!)
 - update (decrease) the cost value stored at a node (change_priority!)
- In our case, a priority queue stores (key, value) pairs, where key refers to identity of some node, while value is the cost of this node.



Recall Heap implementation

- A priority queue can be implemented using a (min) heap
- min-heap implementation of priority queue:
 - PriorityQueue(priorities): takes $\Theta(n)$ time for n = |priorities|
 - .extract_min(): takes $\Theta(\log n)$ time where n is the size of priority queue
 - .change_priority(key, value): takes $\Theta(\log n)$ time where n is the size of priority queue



Final implementation of Prim's Alg

```
def prim(graph, weight):
    tree = UndirectedGraph()
    estimated_predecessor = {node: None for node in graph.nodes}
    cost = {node: float('inf') for node in graph.nodes}
    priority_queue = PriorityQueue(cost)
   while priority_queue:
        u = priority_queue.extract_min()
        if estimated predecessor[u] is not None:
            tree.add_edge(estimated_predecessor[u], u)
        for v in graph.neighbors(u):
            if weight(u, v) < cost[v] and v not in tree.nodes:</pre>
                priority_queue.decrease_priority(v, weight(u, v))
                cost[v] = weight(u, v)
                estimated_predecessor[v] = u
    return tree
```



Time complexity analysis

- We use min-heap to implement the priority queue
- ▶ The maximum size of Q is V
- # iterations of While-loop?
 - V
- # iterations of each call of the inner for-loop?
 - $\rightarrow \deg(v_j)$
- ▶ Total #times lines 7—10 are executed:
 - $\sum_{v_j \in V} \deg(v_j) = 2E$

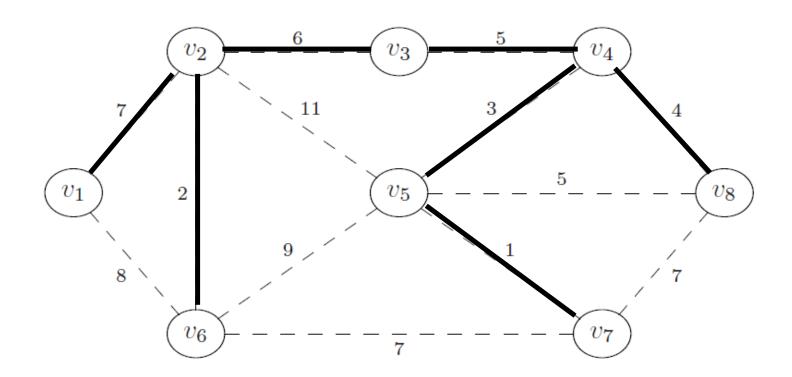
- Initialize priority_queue
 - ▶ Total cost: $\Theta(V)$
- extract_min
 - ▶ Total #:V
 - ▶ Total cost: $\Theta(V \lg V)$

- decrease_priority
 - ▶ Total #: at most 2E
 - ightharpoonup Total cost: $O(E \lg V)$

Total time complexity: $\Theta((V + E) \lg V)$



Example





Comparison with Dijkstra algorithm

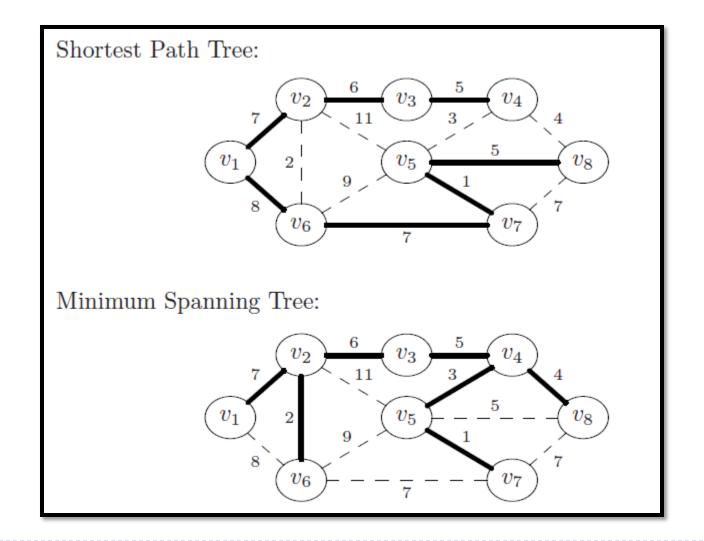
Dijkstra:

- Each node maintains the best distance estimate from source to the current node
 - \blacktriangleright when inspecting a new (crossing) edge (u, v),
 - $\square v.distance = min(v.distnace, u.distance + weight(u,v))$

Prim's:

- Each node (not yet visited) maintains the minimum weight of any edge to reach a visited-node.
 - \blacktriangleright when inspecting a new crossing edge (u, v),
 - $\square v.cost = min(v.cost, weight(u, v))$

Comparison with Dijkstra





Summary and remarks

Prim's algorithm:

- A greedy algorithm which repeatedly choose the minimum-weight edge to reach an unvisited node
- Share similarity to Dijkstra algorithm
- ▶ Runs in $\Theta((V + E) \lg V)$ time using min-heap
- Similar to Dijkstra algorithm, we can further improve the time complexity to $\Theta(E + V \lg V)$ using Fibonacci heap, which is a more efficient implementation of priority queue.
- Next time,
 - Another greedy algorithm, called Kruskal algorithm, which has other properties too.

FIN

