Chapter 3: Greedy Algorithm Design Technique - II

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November 27, 2020

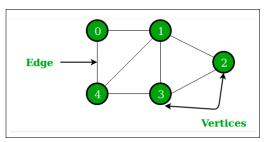
Design and Analysis of Algorithms IIT Jammu, Jammu 1 Review of Essential Graph Basics

Minimum Spanning Trees

Revisiting a Graph

Definition

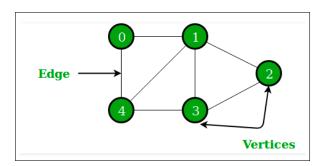
- informally, a set of vertices (nodes) and edges connecting them.
- a graph G = (V, E) where, V is a set of vertices i.e. $V = v_i$ with an edge $e = v_i, v_j$ connecting two vertices with each e ϵ to a set of all such edges between two vertices of the graph viz. $E = (v_i, v_j)$. An example graph G = (V, E) shown here



Graph attributes

A Path

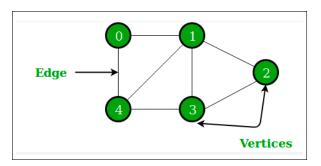
• A Path p of length k, is a sequence of connected vertices p = $< v_0, v_1,, v_k >$ where $(v_i, v_{i+1}) \in E$



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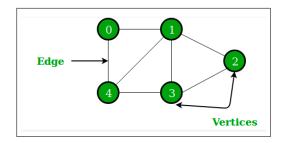
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- in the figure given here various paths are :



Graph attributes: Path, Path-length

A Path and path-length

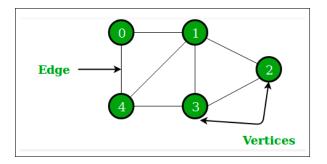
• Path-length is the number of vertices in a graph e.g. the path lengths in this graph are:



Graph attributes: Cycle

A Cycle

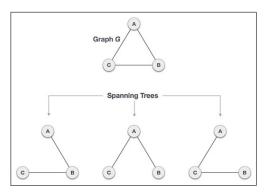
• Cycle: A graph contains no cycles if there is no path $p = \langle v_0, v_1, ..., v_k \rangle$ such that $v_0 = v_k$ e.g. in the graph shown one of the cycles is :



Graph attributes: A Spanning Tree

A Spanning Tree

• A Spanning Tree is a set of |V| - 1 edges that connect all the vertices of a graph.



Graph attributes: A Spanning Tree...

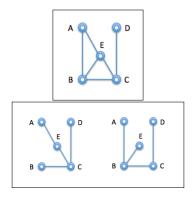


Figure: Another illustration of Spanning Tree

Graph attributes: A Spanning Tree...

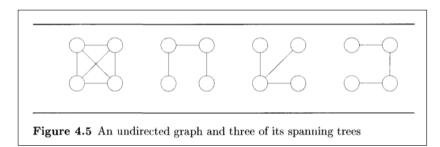


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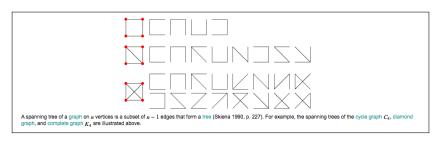


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Properties of a Spanning Tree

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- Spanning trees are a subset of connected Graph G and disconnected graphs do not have spanning tree.

Sparse and Dense Graphs

Dense Graphs

• Dense graphs have a lot of edges compared to the number of vertices.

By comparison, a tree is sparse because |E| = n - 1 = O(n).

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- With n = |V| for the number of vertices, how many edges maximally a dense graph can have ?
- That is $|E| = n^2$

By comparison, a tree is sparse because |E| = n - 1 = O(n).

Algorithm1: Node-centric ST Algorithm

• Pick an arbitrary node and mark it as being in the tree.

- We iterate (n1) times in Step 2, because there are (n1) vertices that have to be added to the tree.
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 - Pick an arbitrary node u in the tree with an edge e to a node w not in the tree.
 - Add e to the spanning tree and mark w as in the tree.
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Algorithm2: Edge-centric ST Algorithm

Start with the collection of singleton trees, each with exactly one node.

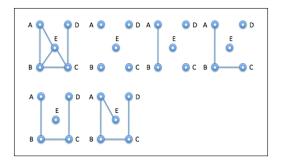


Figure: Another illustration of Spanning Tree

Algorithm2: Edge-centric ST Algorithm

- Start with the collection of singleton trees, each with exactly one node.
- As long as we have more than one tree, connect two trees together with an edge in the graph.

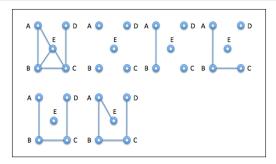


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Minimum Spanning Tree of a Graph

Definition

If a cost c_{ij} is associated with an edge $e_{ij} = (v_i, v_j)$ then the minimum spanning tree is the set of edges E_{span} such that $C = \sum_{for\ all\ e_{ii}\ \epsilon\ E_{span}} c_{ij}$ is a minimum

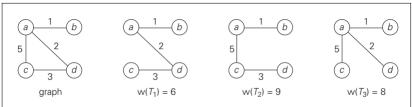


FIGURE 9.2 Graph and its spanning trees, with T_1 being the minimum spanning tree.

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- The approach uses the concepts of safe edges, light edges and a cut of a graph
- Let us investigate these concepts.



Safe edge and Generic MST approach

Safe edge

Given the loop invariant that *Prior to each iteration*, A is a subset of some minimum spanning tree at each step, we determine an edge (u, v) that can be safely added to the set A - we call such an edge a safe edge for A, since we can add it safely to A while maintaining the invariant.

```
Algorithm GENERIC-MST(G,w) 

1 A=\Phi

2 while A does not form the spanning tree

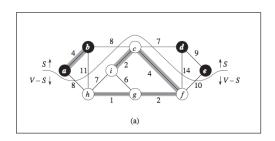
3 find an edge (u,v) that is safe for A

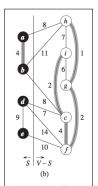
4 A = A U {(u,v)}

5 return A
```

Definitions

Cut of a graph : A cut (S, V S) of an undirected graph G = (V, E) is a partition of the nodes of the graph V in such a way that an edge (u,v) ϵ E crosses the cut (S, V-S) if one of its endpoints is in S and the other endpoint is in V-S. We say, a cut respects a set A of edges if no edge in A crosses the cut.

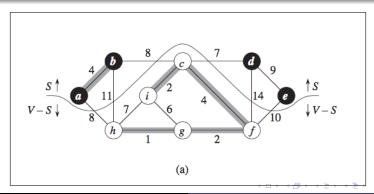




Theorem: Determining Safe edge

Theorem

Let G=(V,E) be a connected, undirected graph with a real-valued weight function w defined on E. Let A be a subset of E that is included in some minimum spanning tree for G, let (S, V-S) be any cut of G that respects A, and let (u,v) be a light edge crossing (S, V-S). Then, edge (u,v) is safe for A.



How to find a safe edge ?

Determining a safe edge

- Calculate the minimum spanning tree
- Put all the vertices into single node trees by themselves
- Put all the edges in a priority queue
- Repeat until we have constructed a spanning tree
- Extract the cheapest edge
- If it forms a cycle, ignore it, else add it to the forest of trees -(it will join two trees into a larger tree)
- Return the spanning tree

Pseudocode: How to find a safe edge?

Cycle detection

- Uses a Union-find structure
- Union-Find data structure is based on partitioning of a set
- How do we view Partition formally ?
- Parition is a set of subsets of elements of a set where
 - every element belongs to one of the sub-sets
 - no element belongs to more than one sub-set
 - that is, given that Set, $S = s_i$,
 - Partition(S)= P_i , where $P_i = s_i$
 - $\forall s_i \in S_i, s_i \in P_i$
 - $\bullet \ \forall j,k,\ P_j \ \cap \ P_k \ = \ \phi$
 - $S = \cup P_j$

Equivalence Class

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- Each element of the set is related to every other element

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- Initially, each vertex is in a class by itself
- As edges are added, more vertices become related and the equivalence classes grow, until finally all the vertices are in a single equivalence class

Representatives of Equivalence Classes

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Now we are ready to apply these concepts for cycle determination

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 If two vertices have the same representative, they are already connected and additing a further connection between them is pointless

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- If two vertices have the same representative, they are already connected and additing a further connection between them is pointless
- Hence, for each end-point of the edge that you are going to add
 - follow the lists and find its representative
 - if the two representatives are equal, then the edge will form a cycle

Let us apply the concept of equivalence classes to determine MST in the given graph

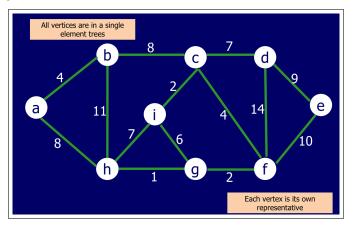


Figure: MST determination (a)

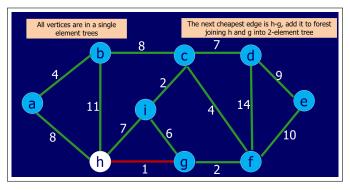


Figure: MST determination (b)

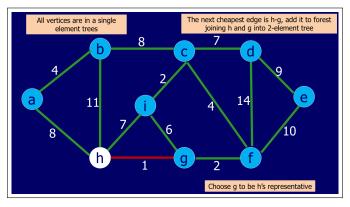


Figure: MST determination (c)

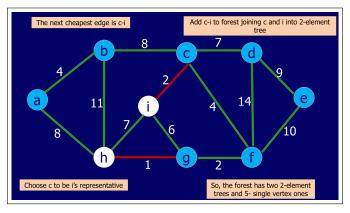


Figure: MST determination (d)

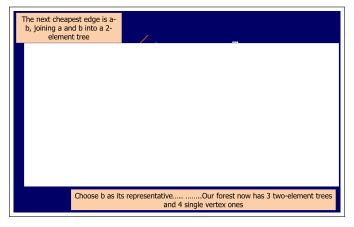


Figure: MST determination (e)

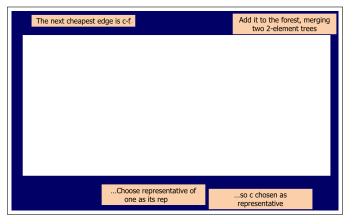


Figure: MST determination (f)

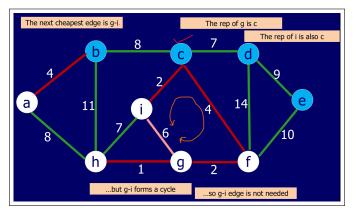


Figure: MST determination (g)

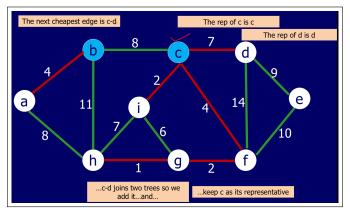


Figure: MST determination (h)

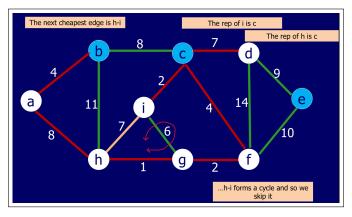


Figure: MST determination (i)

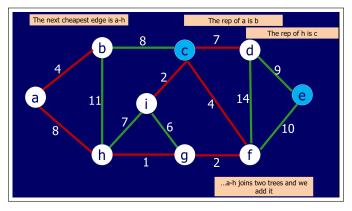


Figure: MST determination (j)

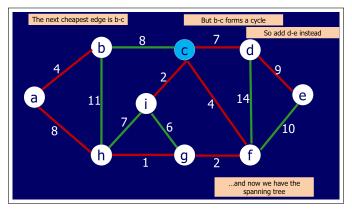


Figure: MST determination (k)

Illustrate the algorithm discussed on the following graph to compute its MST

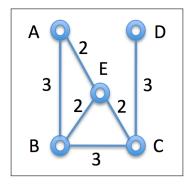


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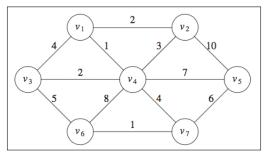
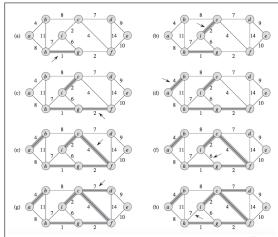
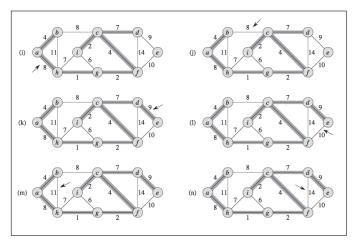


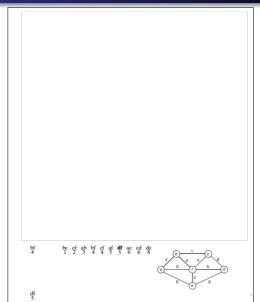
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Kruskal's Algorithm

```
Algorithm Kruskal (Graph G(V, E), Weight_function w)
1. A=\phi
2. for each vertex v \epsilon V[G]
3.
                  do MAKE-SET(v)
4. Sort the edges of E into nondecreasing order
                                     by weight w
   for each edge (u,v) \in E,
6.
                  if FIND-SET(u) \neq FIND-SET(u)
7.
                           A = A \cup \{(u,v)\}
8.
                           UNION(u, v)
9.
         return A
```

Proof by contradiction

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- What would be the cheapest way to do so ?
- The cheapest way to do this is to add e_x because e_x is the cheapest edge
- So we should have added e_x instead of e_z
- This proves that the greedy approach is correct for MST

Theorems Associated

Lemma

Let F be a forest, that is any undirected acyclic graph. Let e=(v,w) be an edge that is NOT in F. Then, there is a cycle in F, consisting of edges of F and edge e, iff v and w are in the same connected component of F.

Theorem

Let G = (V,E) be a connected, undirected graph with a real-valued weight function w defined on the set of edges E. Let A be a subset of E that is included in some MST for G. Let (S, V-S) be a cut of G that respects A and let (u,v) be a light edge crossing (S, V-S). Then, edge (u,v) is safe for A.

Theorems Associated

Theorem

Let G=(V,E) be a connected, undirected graph and G'=(V',E') be a partial graph formed by the nodes of G and the edges in E'. Let there be n nodes in V. Then, prove that the set E' with n or more edges cannot be optimal. Also, prove that E' must have exactly (n-1) edges and so E' must be a tree.

Theorem

Prove that Kruskal's algorithm is correct and it finds its MST.

Tutorial Assignment Proofs

Theorem

- Let T be a MST for a graph G, let e be an edge in T and let T' be T with e removed. Show that e is a minimum weight edge between components of T'.
- Comment on the validity of the statement. If all the weights in G are distinct, distinct spanning trees of G have distinct weights. Give a counterexample.
- Let T and T' be two STs of a connected graph G. Suppose that an edge e is in T but not in T'. Show that there is an edge e' in T', but not in T, such that $(T-e \cup e')$ and $(T'-e' \cup e)$ are STs of G.

Time Complexity

Steps

- Initialise forest O(|V|)
- Sort edges $O(|E|\log|E|)$
 - Check edge for cycles $O(|V|) \times$
 - Number of edges O(|V|) i.e. $O(|V|^2)$
- Total $O(|V| + |E|\log|E| + |V|^2)$
- Since $|E| = O(|V|^2)$ Total = $O(|V|^2 log |V|)$
- Thus, Kruskal's algorithm is tagged as $On^2 \log n$ algorithm for a graph of n vertices
- This is an upper bound, some improvements on this are known...
- Prim's algorithm can be O(|E| + |V|log|V|) using Fibonacci heaps.
- Even better variants are know for restricted cases, such as sparse graphs ($|E| \approx |V|$)

- Follows the natural greedy approach starting with the source vertex to create the spanning tree,
- add an edge to the tree that is attached at exactly one end to the tree & has minimum weight among all such edges.
- Prim's algorithm starts from one vertex and grows the rest of the tree an edge at a time.
- As a greedy algorithm, which edge should we pick?

Points to note

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 - the result is a sort of forest of trees that grow haphazardly and later merge into a tree.
- As compared in Prim's the MST grows in a natural manner, staring from an arbitrary root.
- At each, a new edge is added to a tree already constructed.

Approach

• Consider B - a set of nodes, T - a set of edges

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- then it adds v to B and {u, v} to T.
- in this way, in T at any instant an MST for the nodes in B is formed.

```
Algorithm Prim(Graph\ G(V,E),\ Weight\_function\ w) {initialization}

1. T \epsilon \phi

2. B \leftarrow \{an\ arbitrary\ member\ of\ V\}

3. while B \neq N do

4. find E=\{u,\ v\} of minimum weight such that u \epsilon B and v \epsilon V-B

5. T \leftarrow T \cup \{e\}

6. B \leftarrow B \cup \{v\}

7. return T
```

How to determine the edge with the minimum weight?

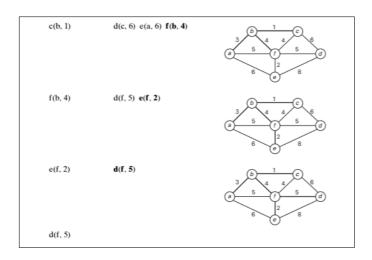
Prim's Algorithm Refined

```
Algorithm Prim (Graph G(V, E), Weight_function w, r)
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\* During execution of the algorithm, all vertices that
are not in the tree reside in a min-priority queue Q
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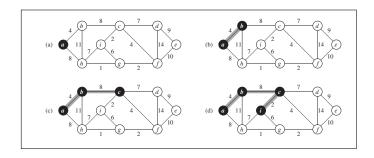
Applying Prim's Algorithm...

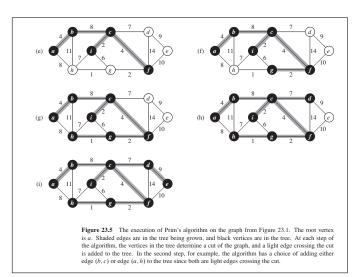
Tree vertices	Remaining vertices	Illustration
a(-, -)	$\mathbf{b}(\mathbf{a}, 3) \ c(-, \infty) \ d(-, \infty)$ $e(\mathbf{a}, 6) \ f(\mathbf{a}, 5)$	a 5 f 5 d
b(a, 3)	$\begin{array}{ll} c(b,1) & d(-,\infty) & e(a,6) \\ f(b,4) & \end{array}$	3 b 1 c 6 6 g 5 f 5 d

Applying Prim's Algorithm...

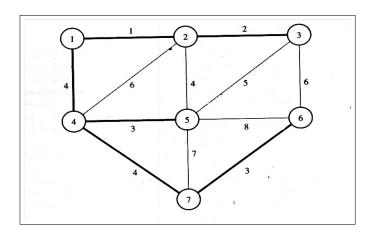


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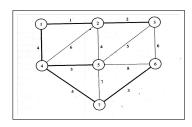




Comparing Kruskal and Prim by applying them

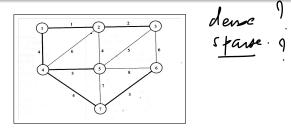


Running Prim's on the given graph



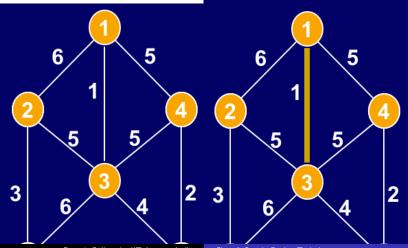
	E 1 C 11 1	
Step	EdgeConsidered	ConnectedComponents
Init	-	{1}
1	{1 2}	{1 2} ∨
2	{2 3}	{1 2 3} ✓
3	{1 4}	{1 2 3 4 } 🗸
4	{4 5}	{1 2 3 4 5 }
5	{4 7}	{1 2 3 4 5 7}
6	{5}	Rejected *
7	{4 7}	{1 2 3 4 5 6 7}

Running Kruskal's on the given graph

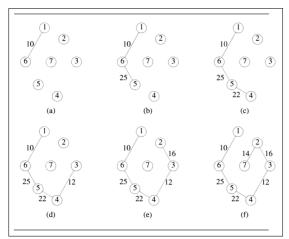


Step	EdgeConsidered	ConnectedComponents]
Init	-	{1} {2} {3} {4} {5} {6} {7}	-fore
1	{1 2} ∨	{1 2} {3} {4} {5} {6} {7}] '
2	{2 3}	{1 2 3} {4} {5} {6} {7}	
3	{4 5}	{1 2 3} {4 5} {6} {7}	
4	{6 7}	{1 2 3} {4 5} {6 7}	
5	{1 4}	{1 2 3 4 5} {6 7}	
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Illustrate the PRim's algorithm in the following graph to compute its $\ensuremath{\mathsf{MST}}$



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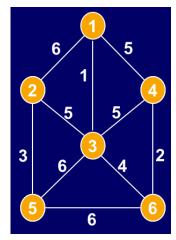


Figure: Various iterations of Prim

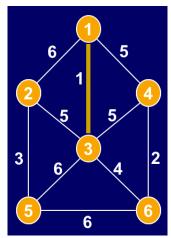


Figure: Various iterations of Prim

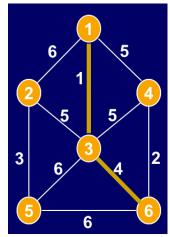


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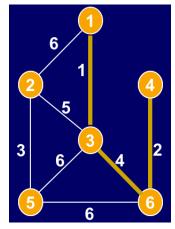


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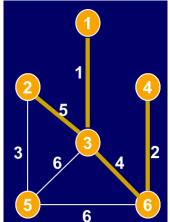


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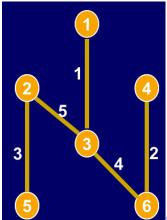
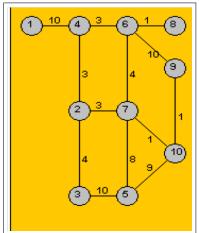
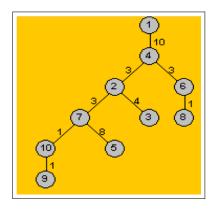


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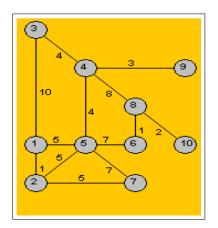
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Answer...



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Prim's Algorithm Refined...repeated here

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- total time of for loop is O(E lg V).
- Therefore, Prim takes $O(V \lg V + E \lg V)$ time.

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