

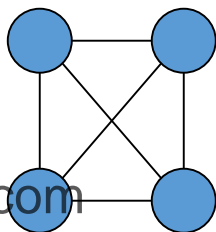
# Minimum Spanning Tree

This slides are adapted from the notes of  
Jonathan Davis & George Bebis, from the Univeristy of Nevada, Reno

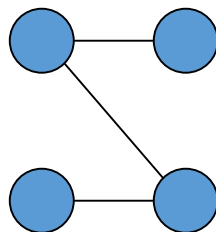
# Spanning Trees

- Greedy algorithm: makes locally best choice/decision ignoring effect on future.
- Tree: connected acyclic graph.
- Spanning tree: a spanning tree of a graph  $G$  is a subset of edges of  $G$  that form a tree and reach all vertices of  $G$ .
- A tree (i.e., connected, acyclic graph) which contains all the vertices of the graph.
- A graph may have many spanning trees.

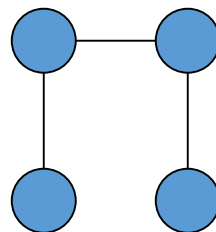
Graph A



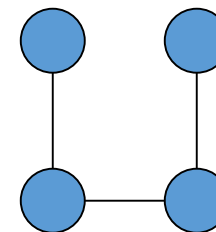
Some Spanning Trees from Graph A



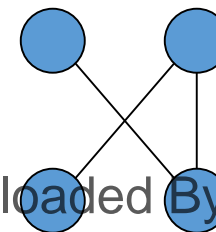
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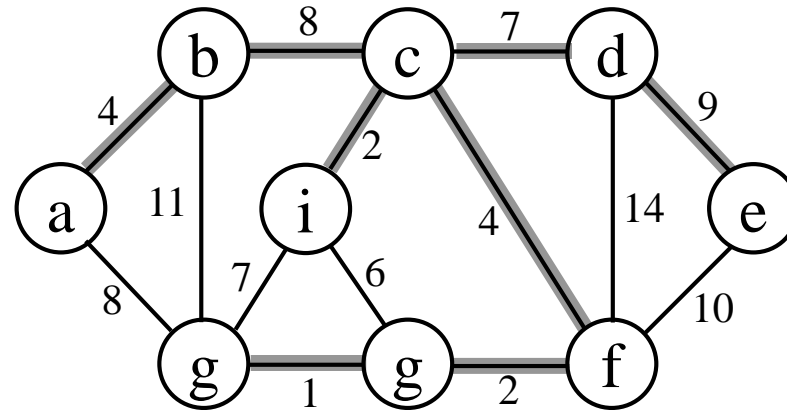
or



# Spanning Trees

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- Minimum Spanning Tree
  - Spanning tree with the **minimum sum of weights**.

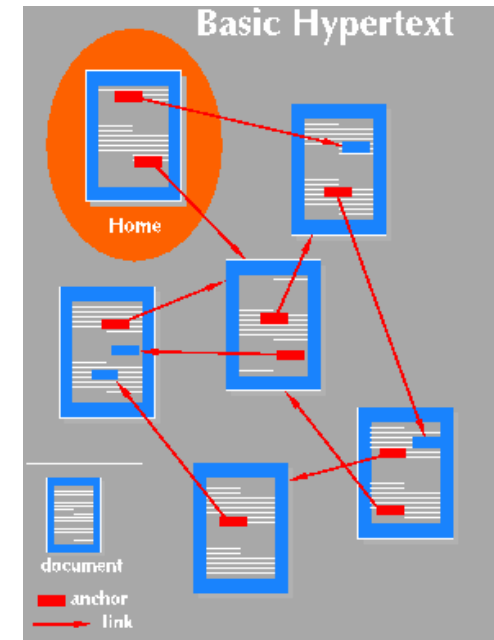
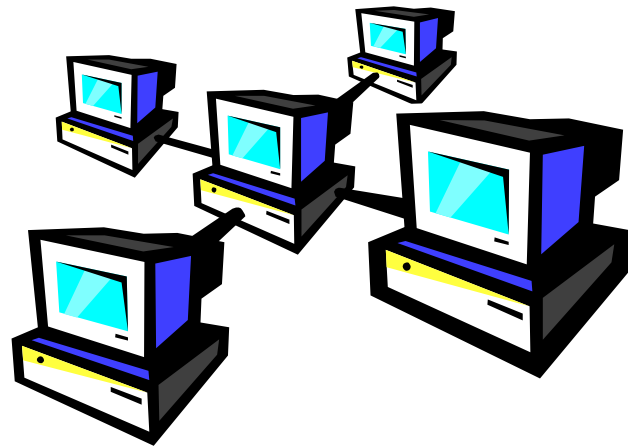


- Spanning forest
  - If a graph is not connected, then there is a spanning tree for each connected component of the graph

# Applications to MST

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Find the least expensive way to connect a set of cities, terminals, computers, etc.



# MST vs. Shortest Path

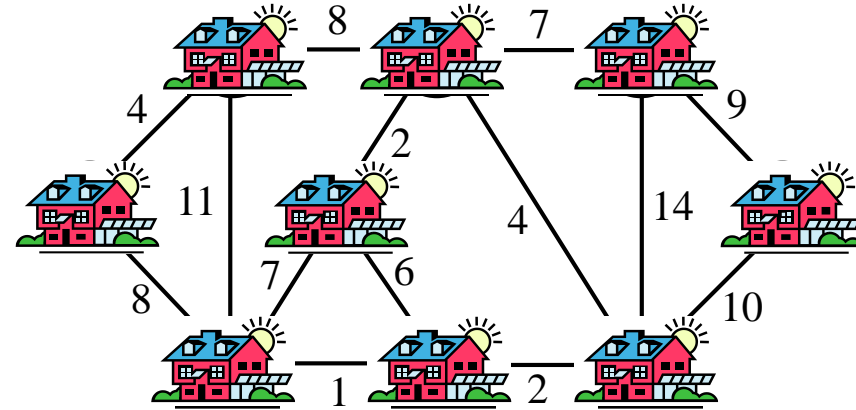
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- MST is a tree in a graph that spans all the vertices and total weight of a tree is minimal
- Shortest path is a shortest path from one vertex to another
- In MST there is no source/destination but the subset (tree) of the graph connects all vertices of the graph without any cycle with the minimum sum of weights

# Example

## Problem

- A town has a set of houses and a set of roads
- A road connects 2 and only 2 houses
- A road connecting houses  $u$  and  $v$  has a repair cost  $w(u, v)$



**Goal:** Repair enough (and no more) roads such that:

1. Everyone stays connected  
i.e., can reach every house from all other houses
2. Total repair cost is minimum

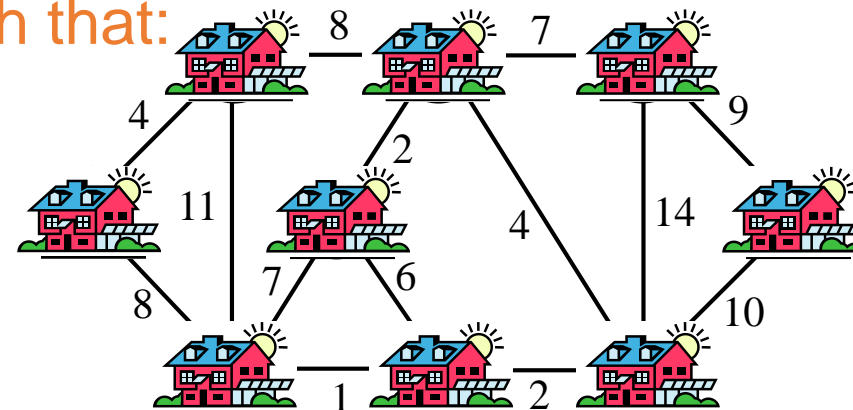
# Minimum Spanning Trees

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- A connected, undirected graph:
  - Vertices = houses,      Edges = roads
- A **weight**  $w(u, v)$  on each edge  $(u, v) \in E$

Find a spanning tree  $T \subseteq E$  such that:

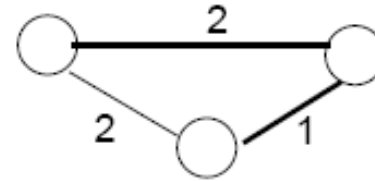
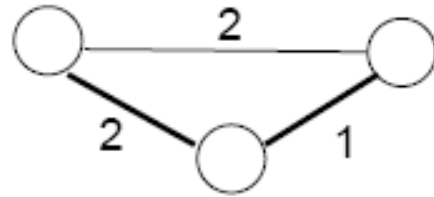
1.  $T$  connects all vertices
2.  $w(T) = \sum_{(u,v) \in T} w(u, v)$  is minimized



# Properties of Minimum Spanning Trees

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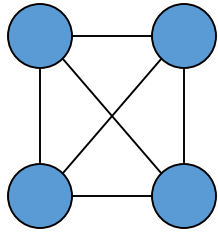
- Minimum spanning tree is **not** unique



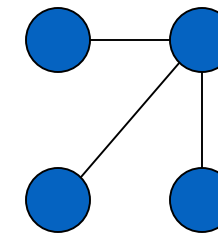
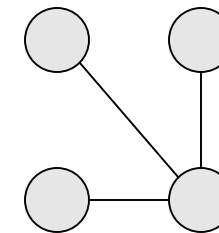
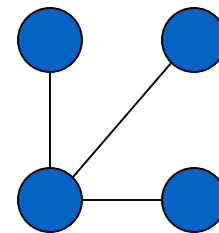
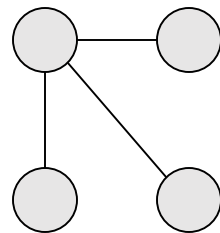
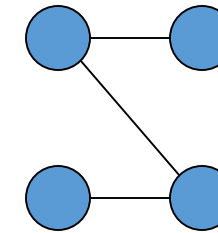
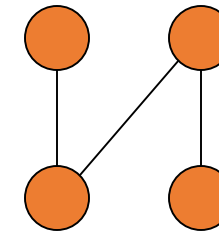
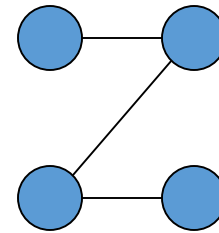
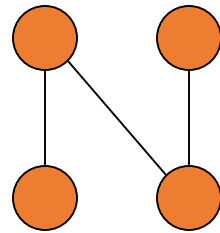
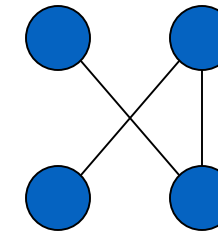
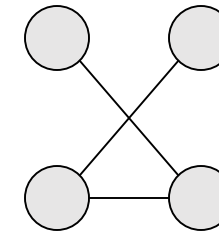
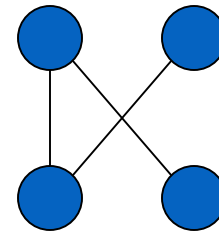
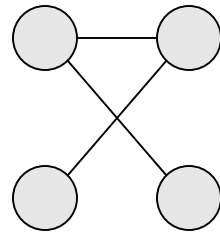
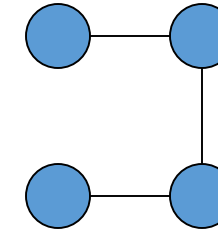
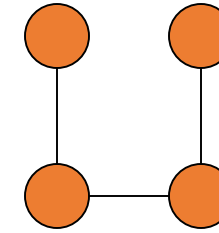
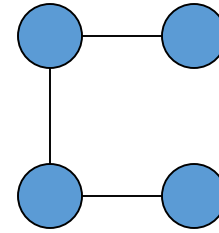
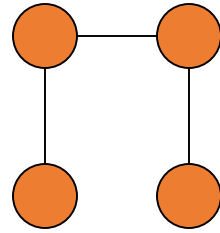
- MST has no cycles
  - We can take out an edge of a cycle, and still have the vertices connected while reducing the cost
- # of edges in a MST is  $|V| - 1$



Complete Graph



All 16 of its Spanning Trees

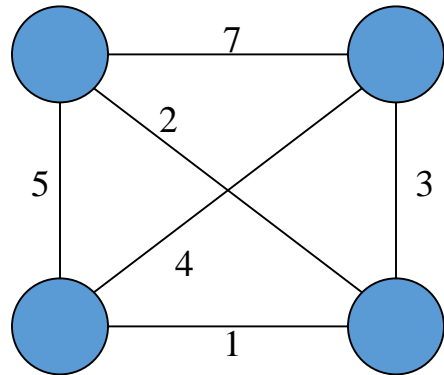


# Minimum Spanning Trees

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A Minimum Spanning Tree (MST) is a subgraph of an undirected graph such that the subgraph spans (includes) all nodes, is connected, is acyclic, and has minimum total edge weight

Complete Graph



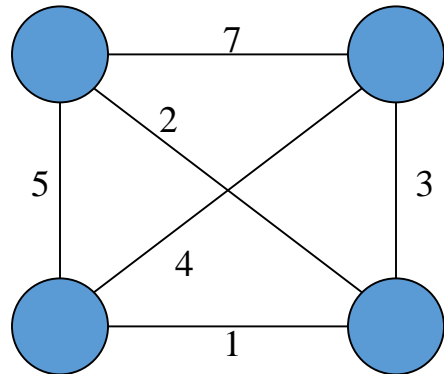
Minimum Spanning Tree

# Minimum Spanning Trees

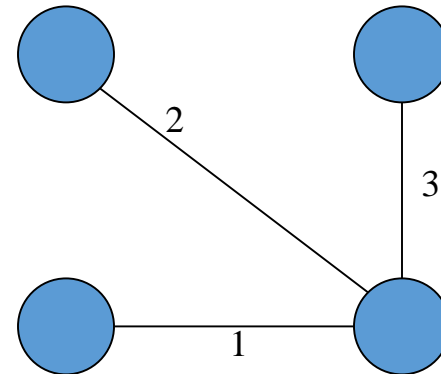
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A Minimum Spanning Tree (MST) is a subgraph of an undirected graph such that the subgraph spans (includes) all nodes, is connected, is acyclic, and has minimum total edge weight

Complete Graph



Minimum Spanning Tree



# Algorithms for Obtaining the Minimum Spanning Tree

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- Prim's Algorithm
- Kruskal's Algorithm

# Prim's Algorithm

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This algorithm starts with one node. It then, one by one, adds a node that is unconnected to the new graph, each time selecting the node whose connecting edge has the smallest weight out of the available nodes' connecting edges.

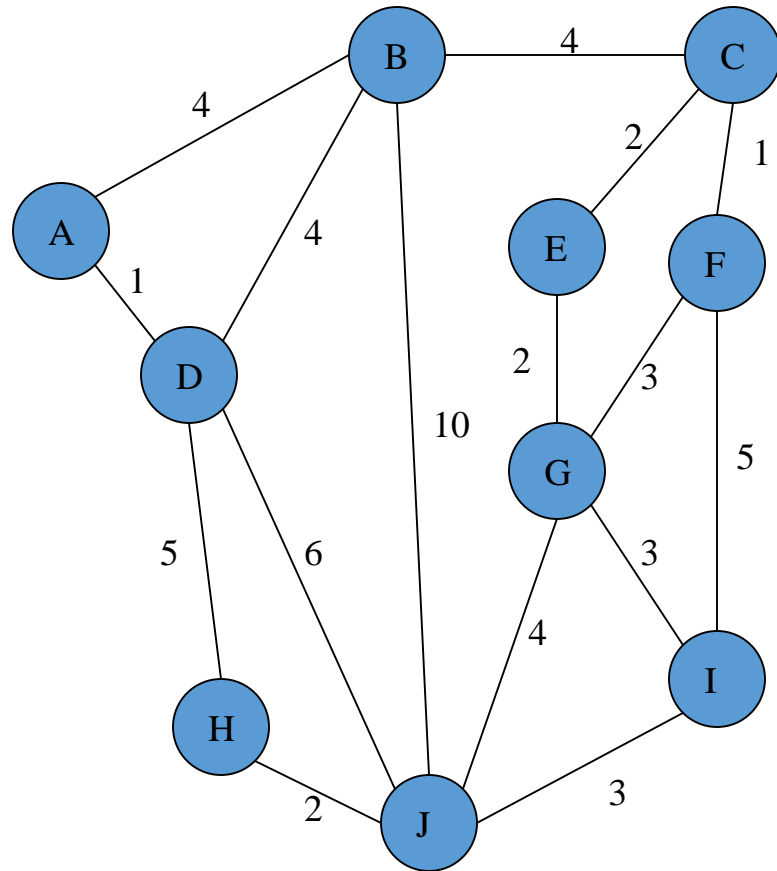
We can think of the implementation of Prim's algorithm to be (almost) identical to Dijkstra's algorithm

Steps:

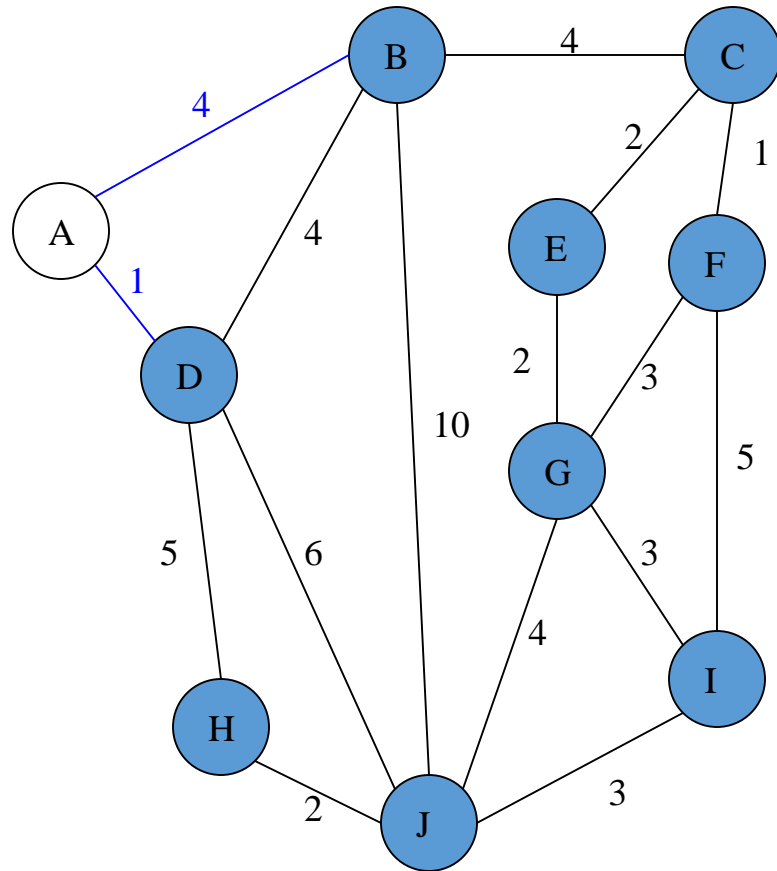
1. The new graph is constructed - with one node from the old graph.
2. While new graph has fewer than  $n$  nodes:
  1. Find the node from the old graph with the smallest connecting edge to the new graph
  2. Add it to the new graph

Every step will have joined one node, so that at the end we will have one graph with all the nodes and it will be a minimum spanning tree of the original graph.

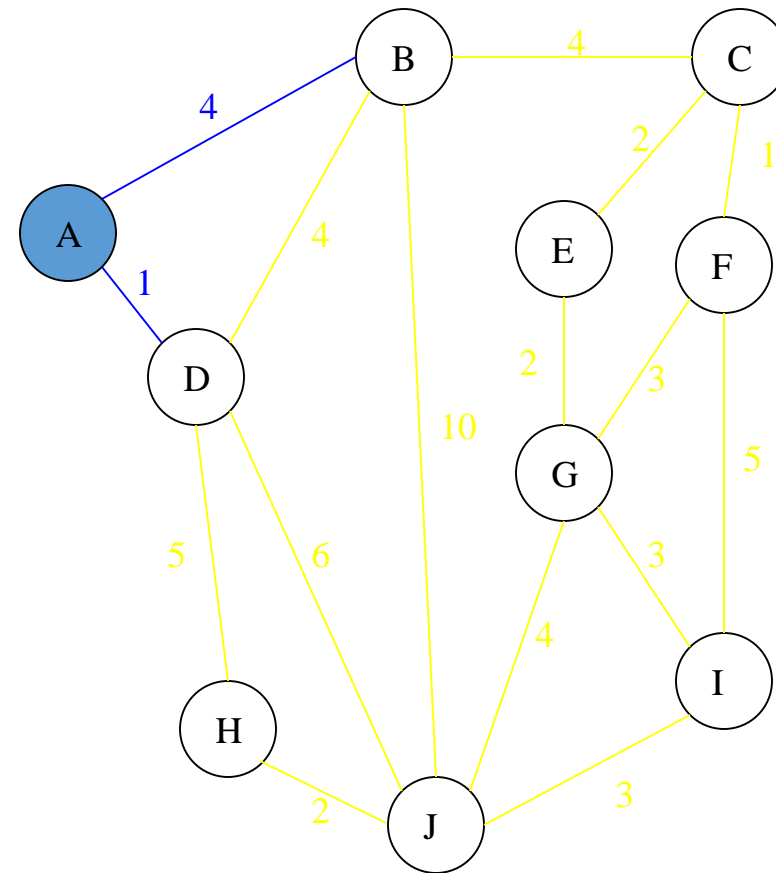
## Complete Graph



### Old Graph

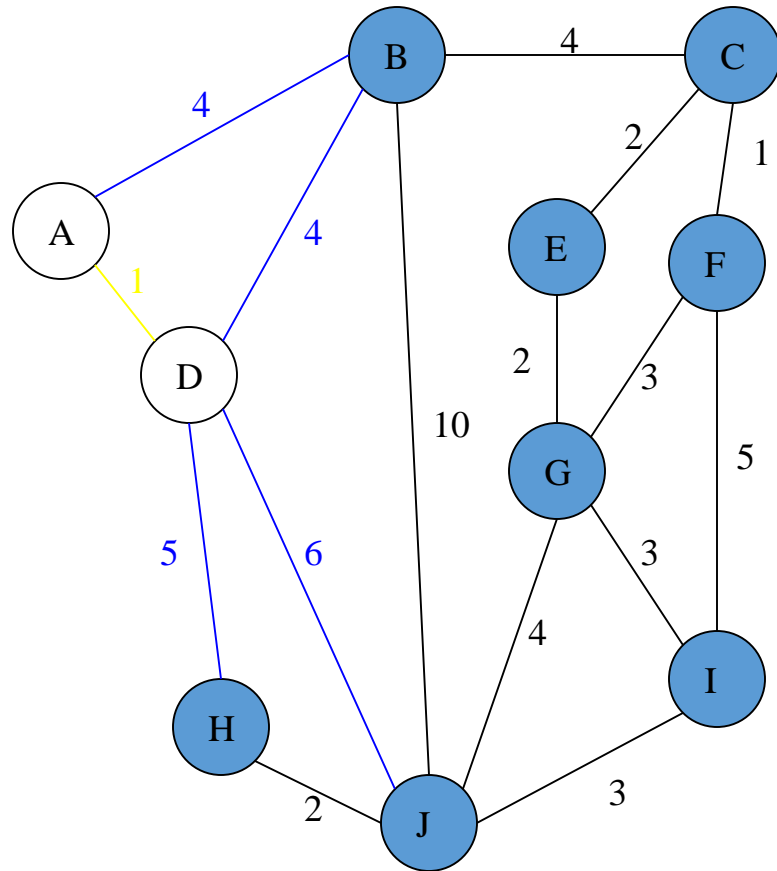


### New Graph

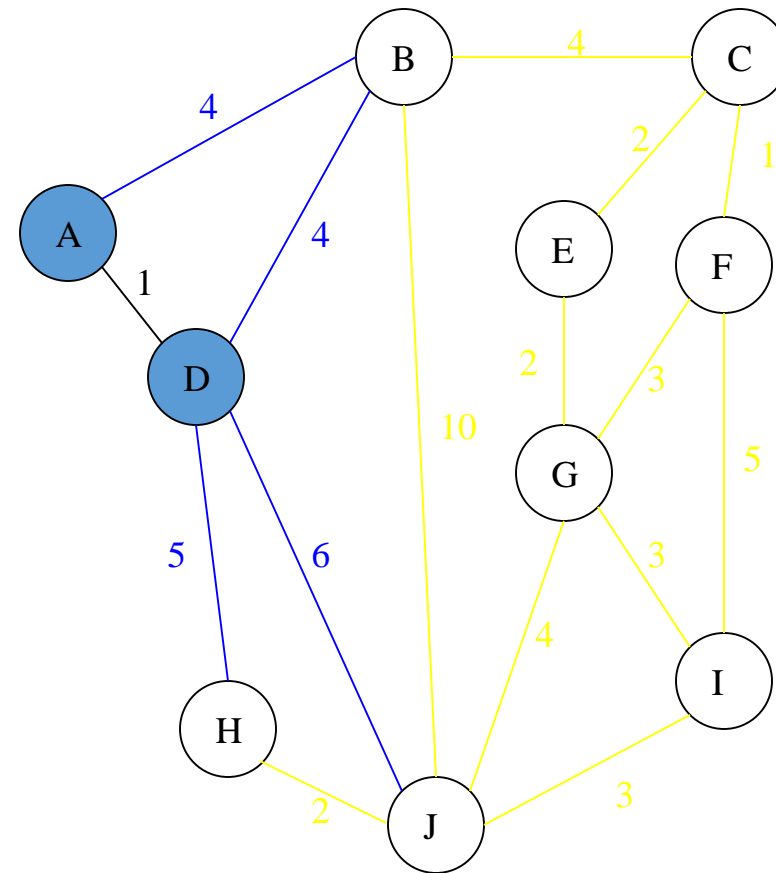




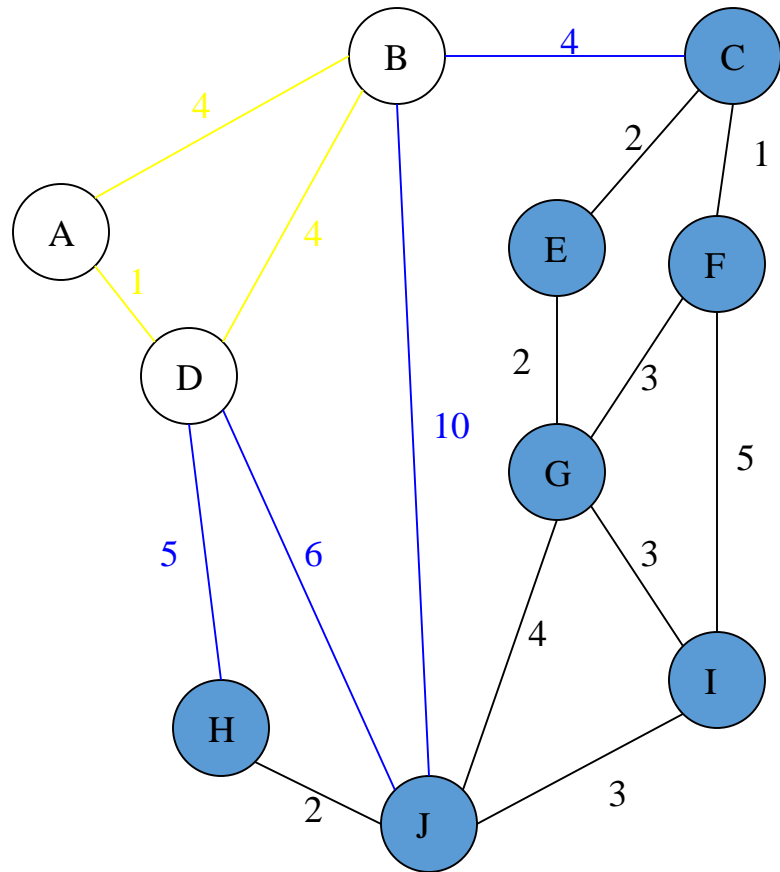
# Old Graph



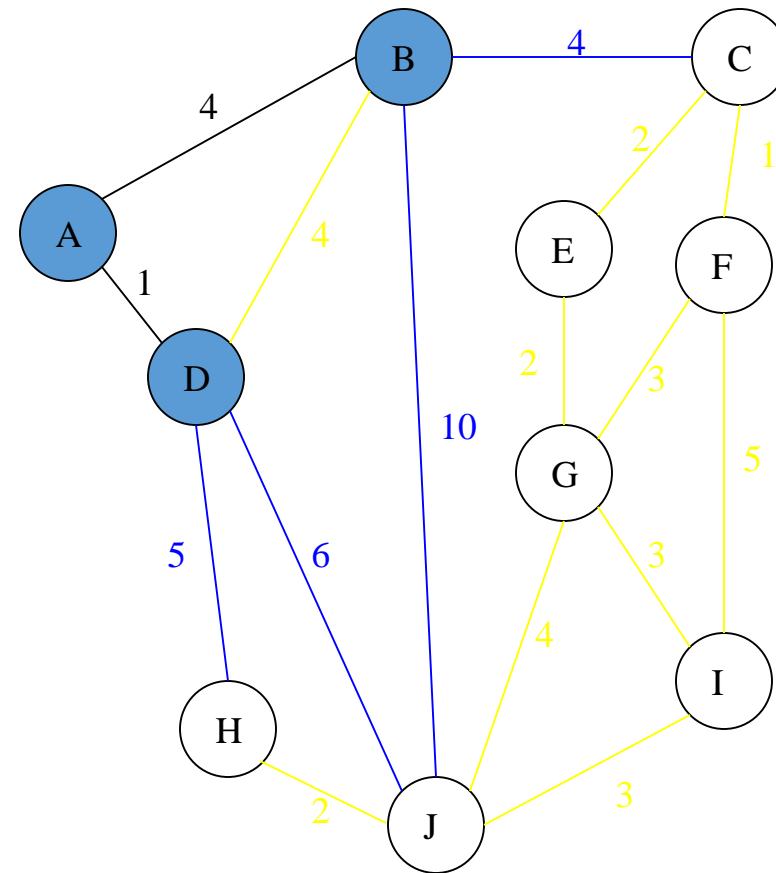
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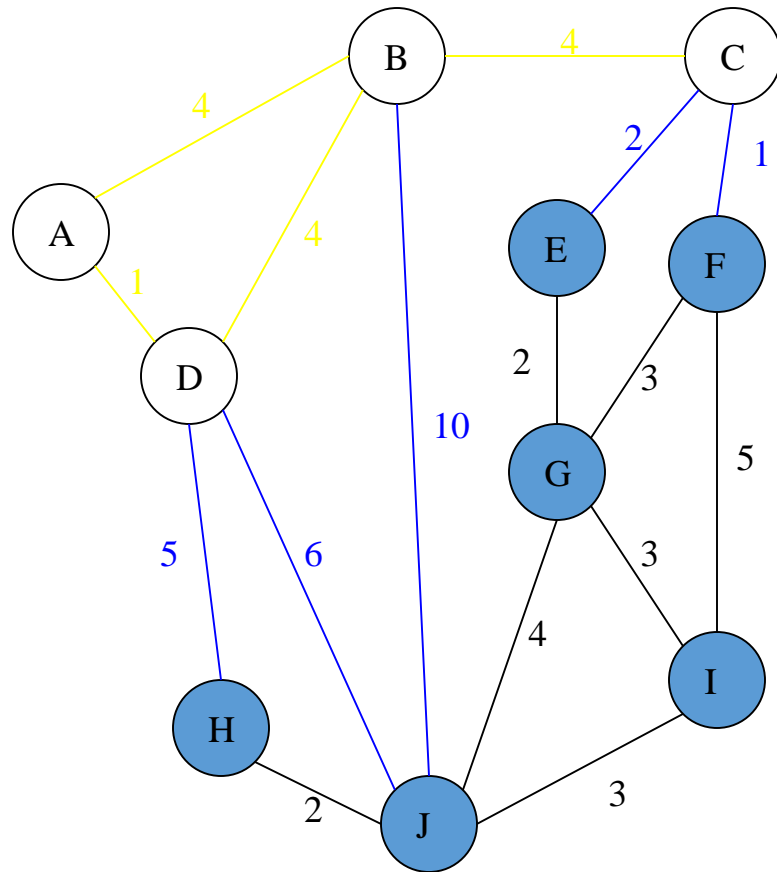
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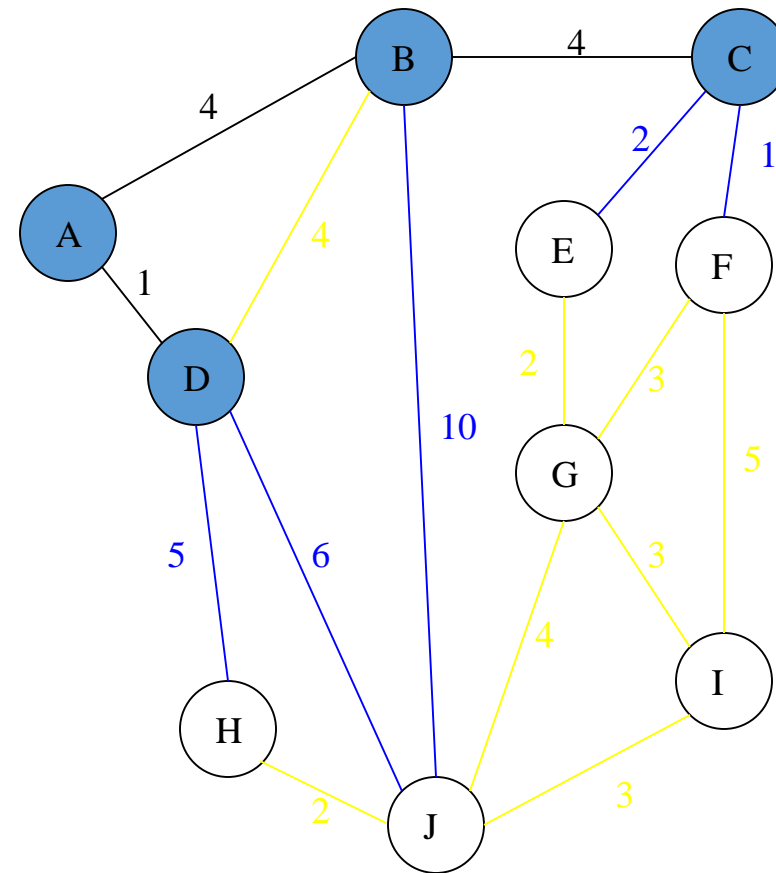
# New Graph



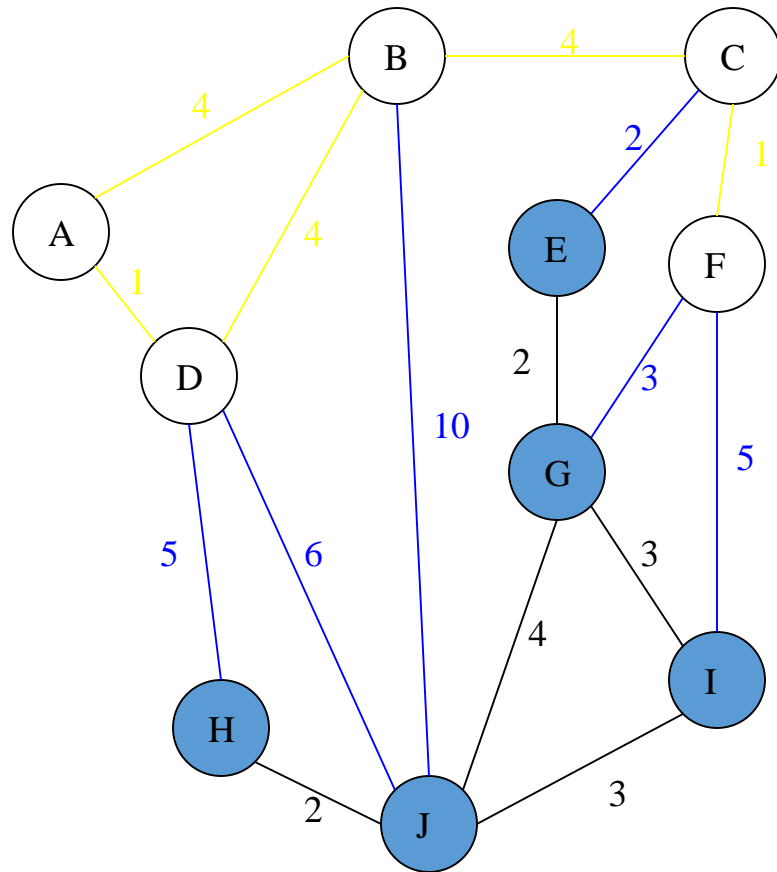
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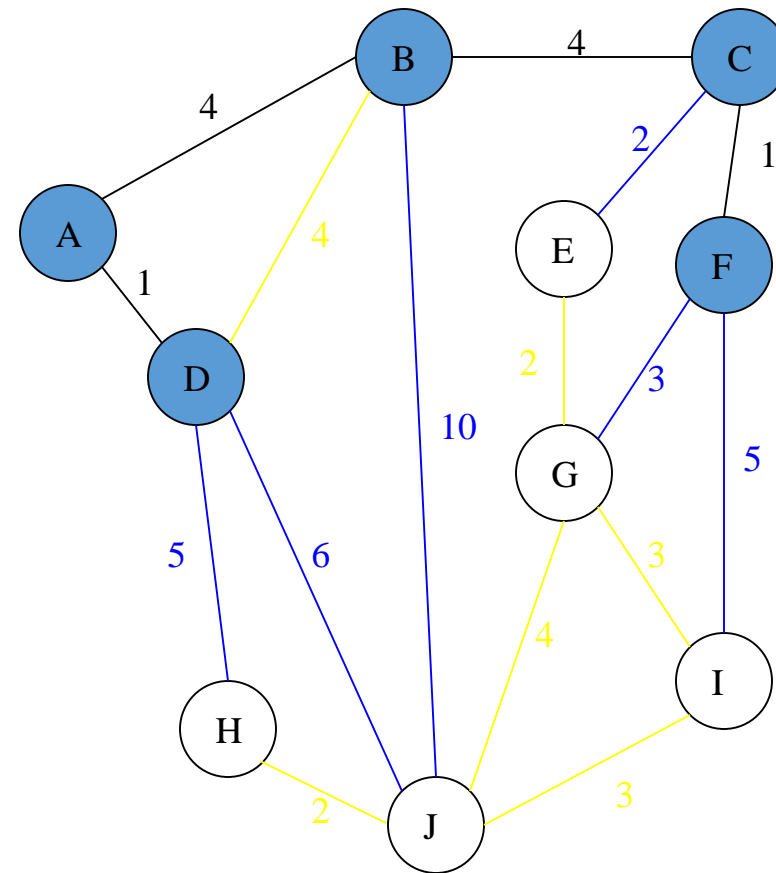
# New Graph



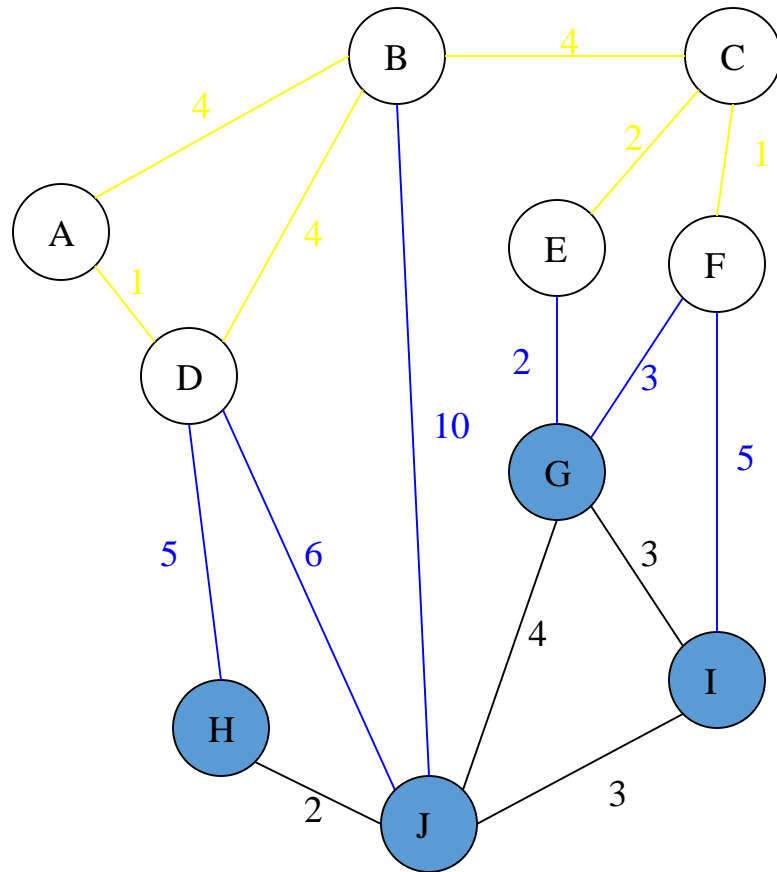
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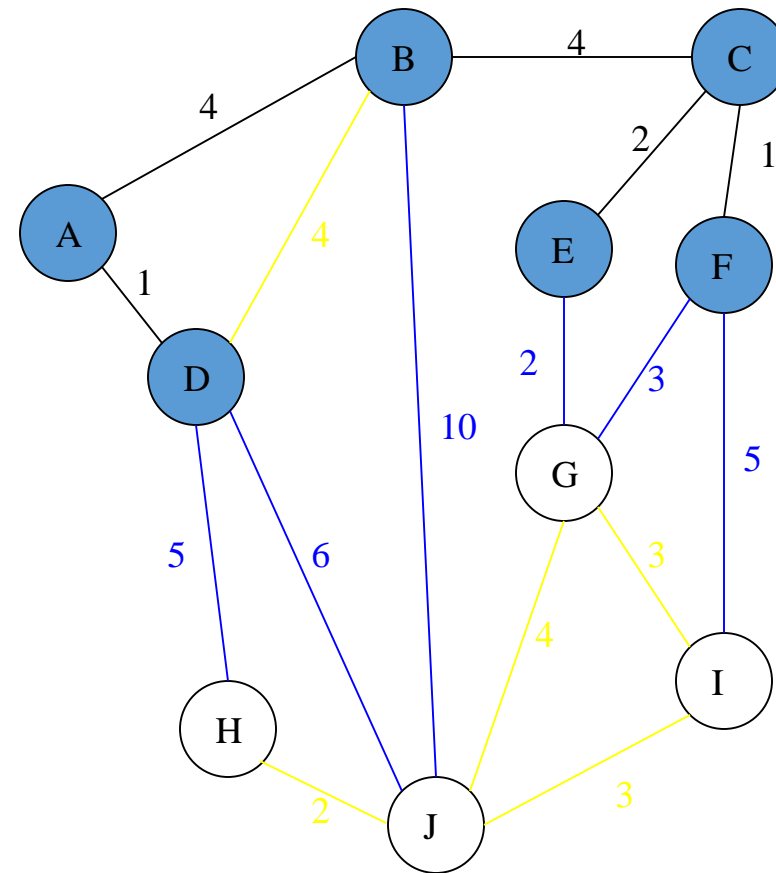
# New Graph



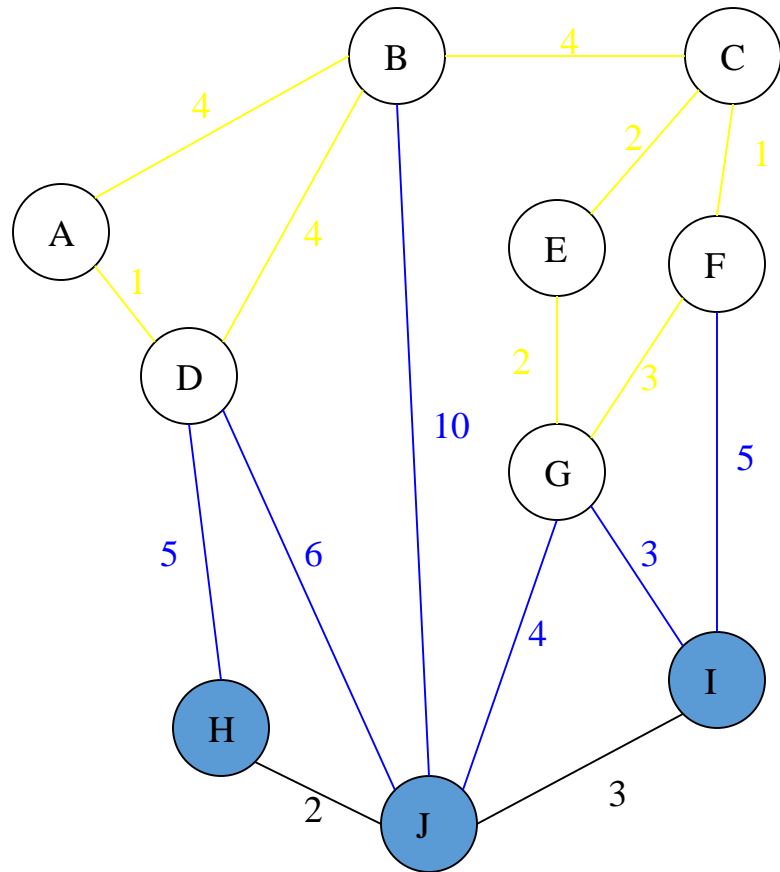
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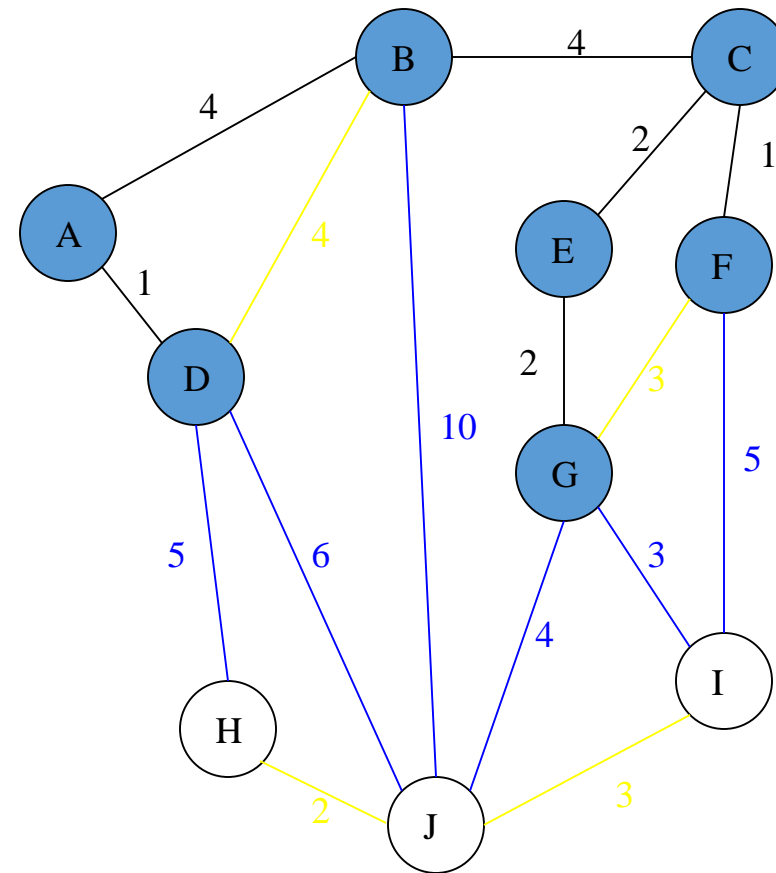
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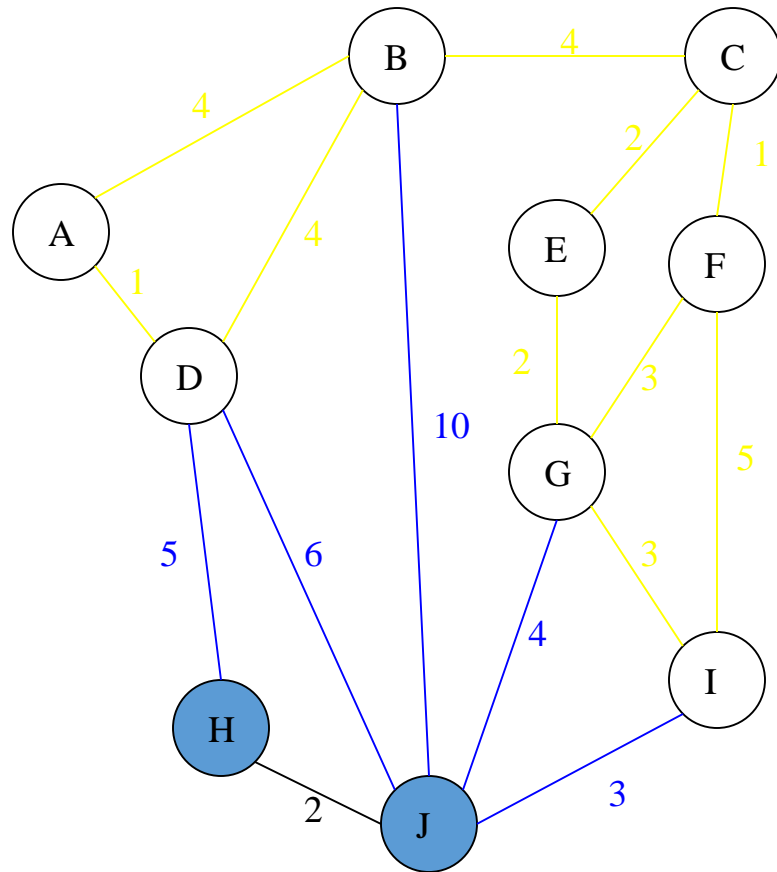
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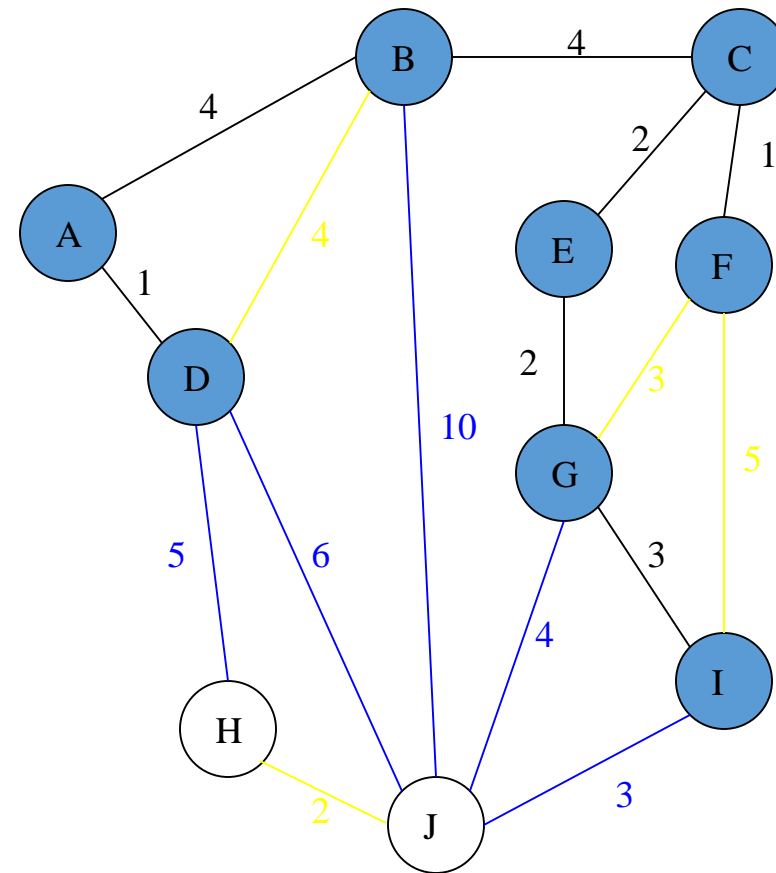
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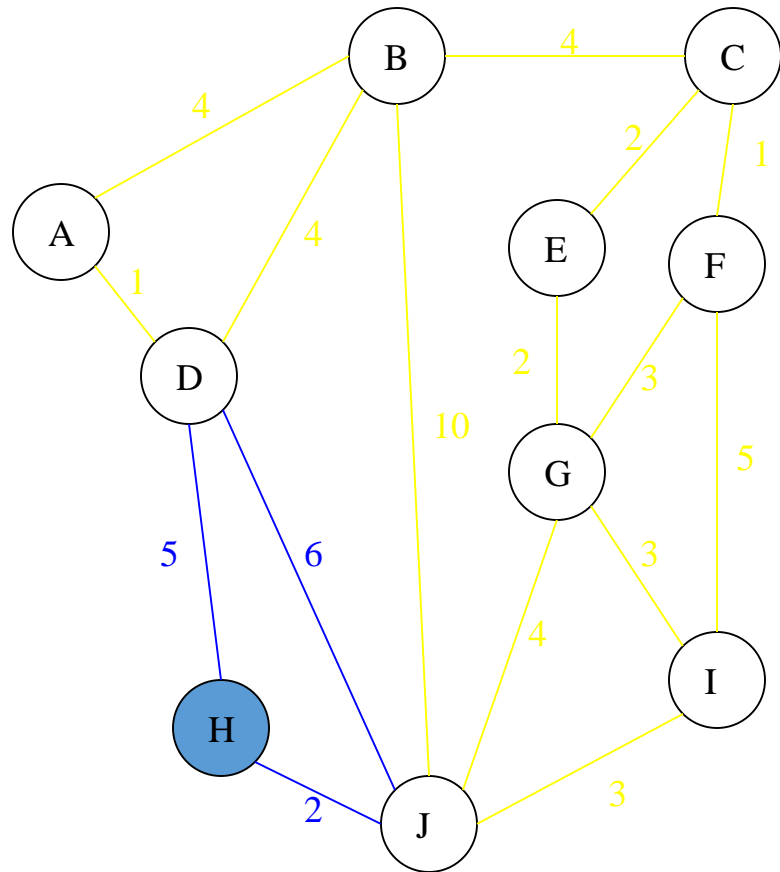
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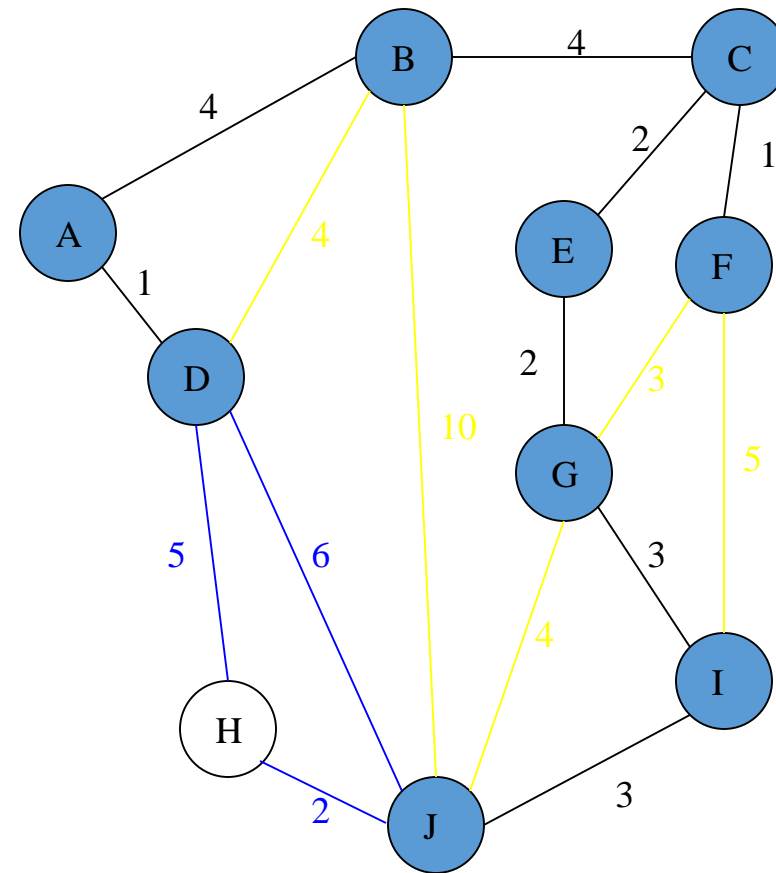
# New Graph



# Old Graph

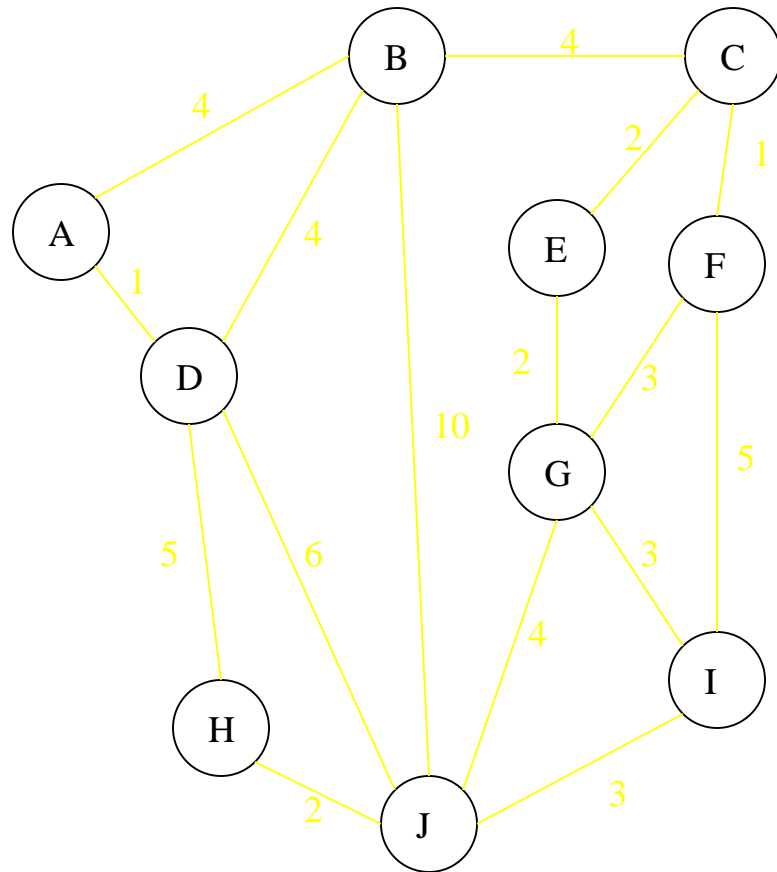


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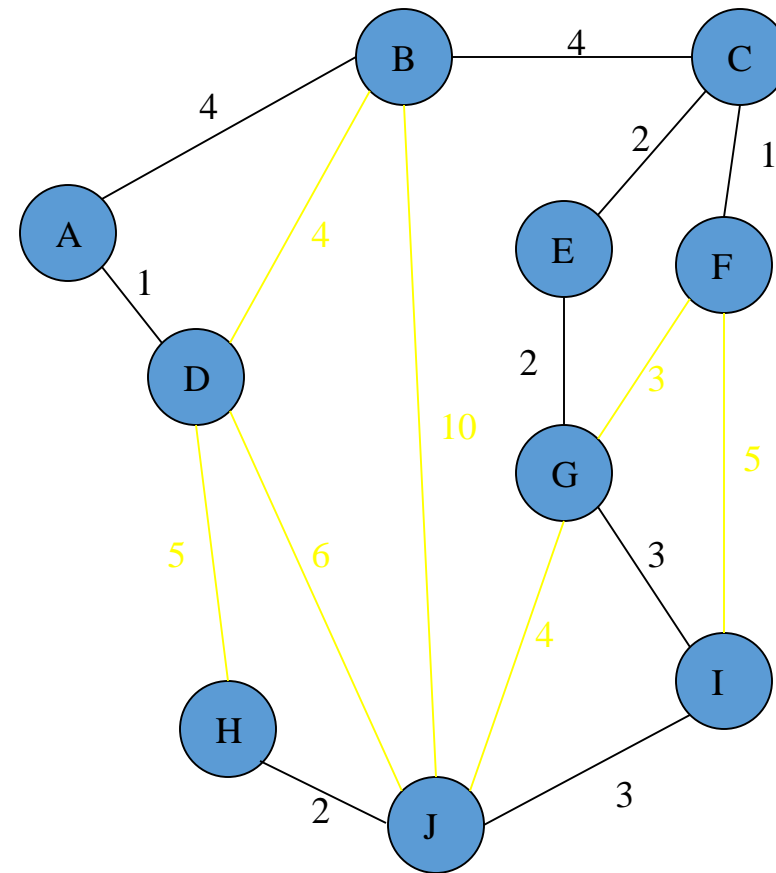




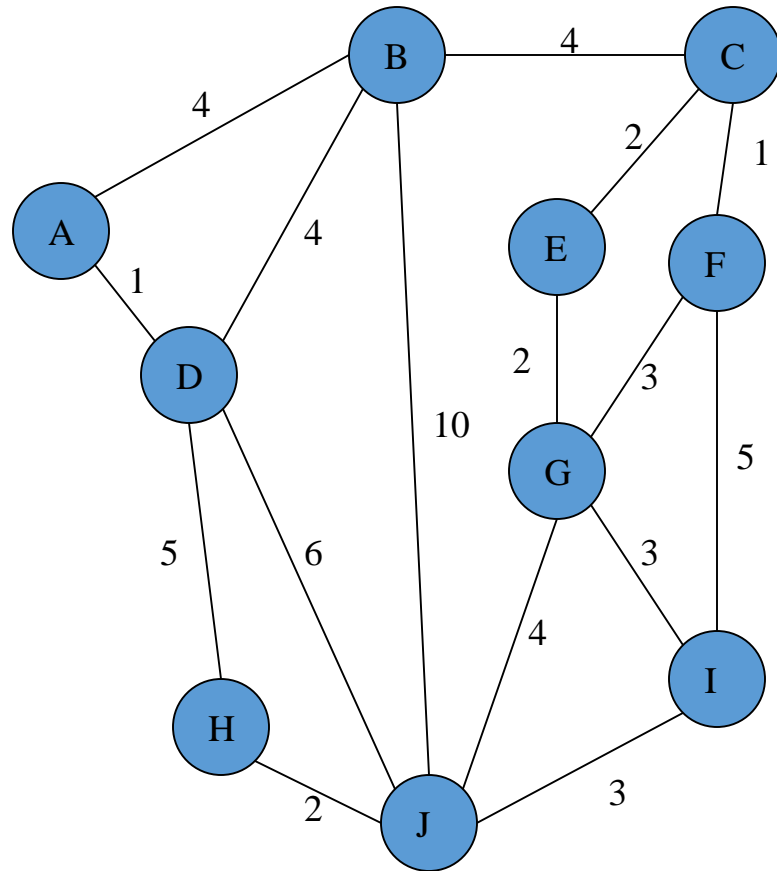
# Old Graph



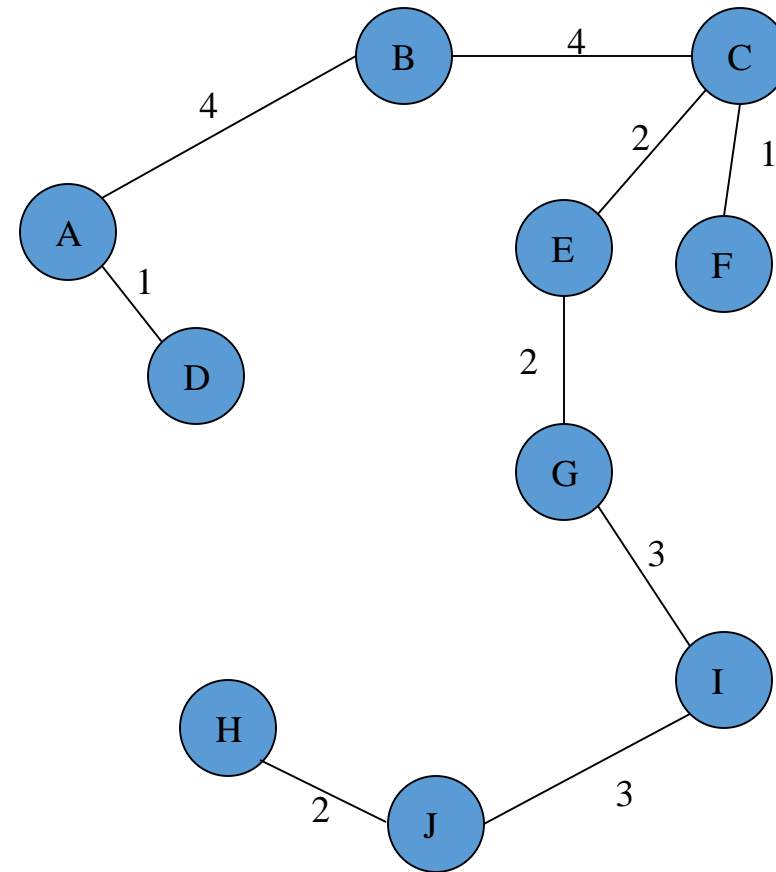
# New Graph



## Complete Graph



## Minimum Spanning Tree



# Analysis of Prim's Algorithm

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Running Time =  $O(e + v \log v)$       ( $e$  = edges,  $v$  = nodes)

If a heap is not used, the run time will be  $O(n^2)$  instead of  $O(e + v \log v)$ . However, using a heap complicates the code since you're complicating the data structure.

Unlike Kruskal's, it doesn't need to see all of the graph at once. It can deal with it one piece at a time. It also doesn't need to worry if adding an edge will create a cycle since this algorithm deals primarily with the nodes, and not the edges.

# Kruskal's Algorithm

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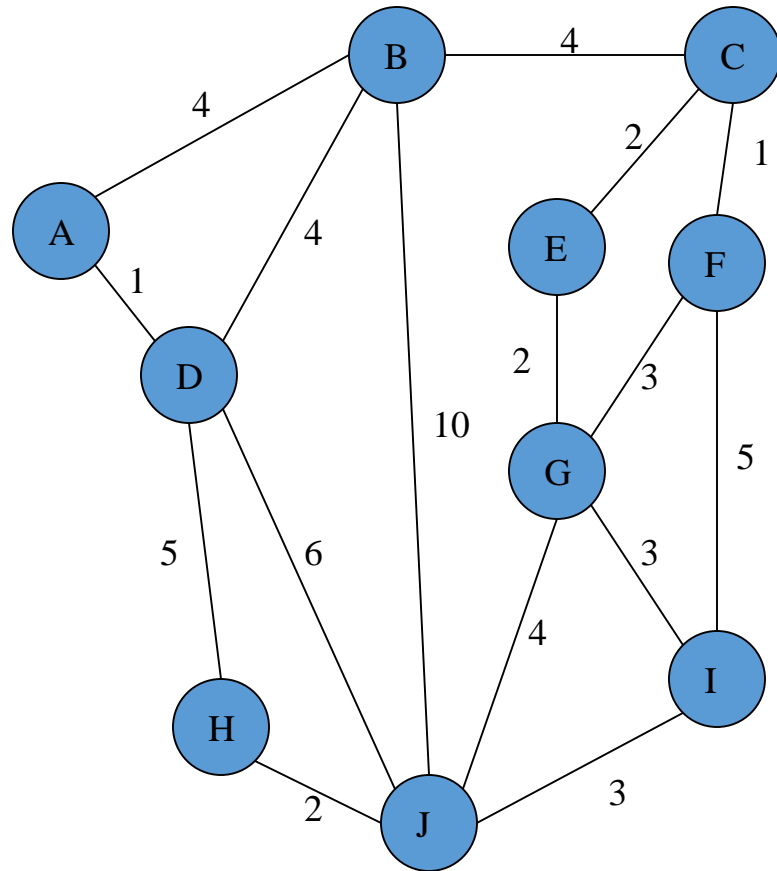
This algorithm creates a forest of trees. Initially the forest consists of  $n$  single node trees (and no edges). At each step, we add one edge (the cheapest one) so that it joins two trees together. If it were to form a cycle, it would simply link two nodes that were already part of a single connected tree, so that this edge would not be needed.

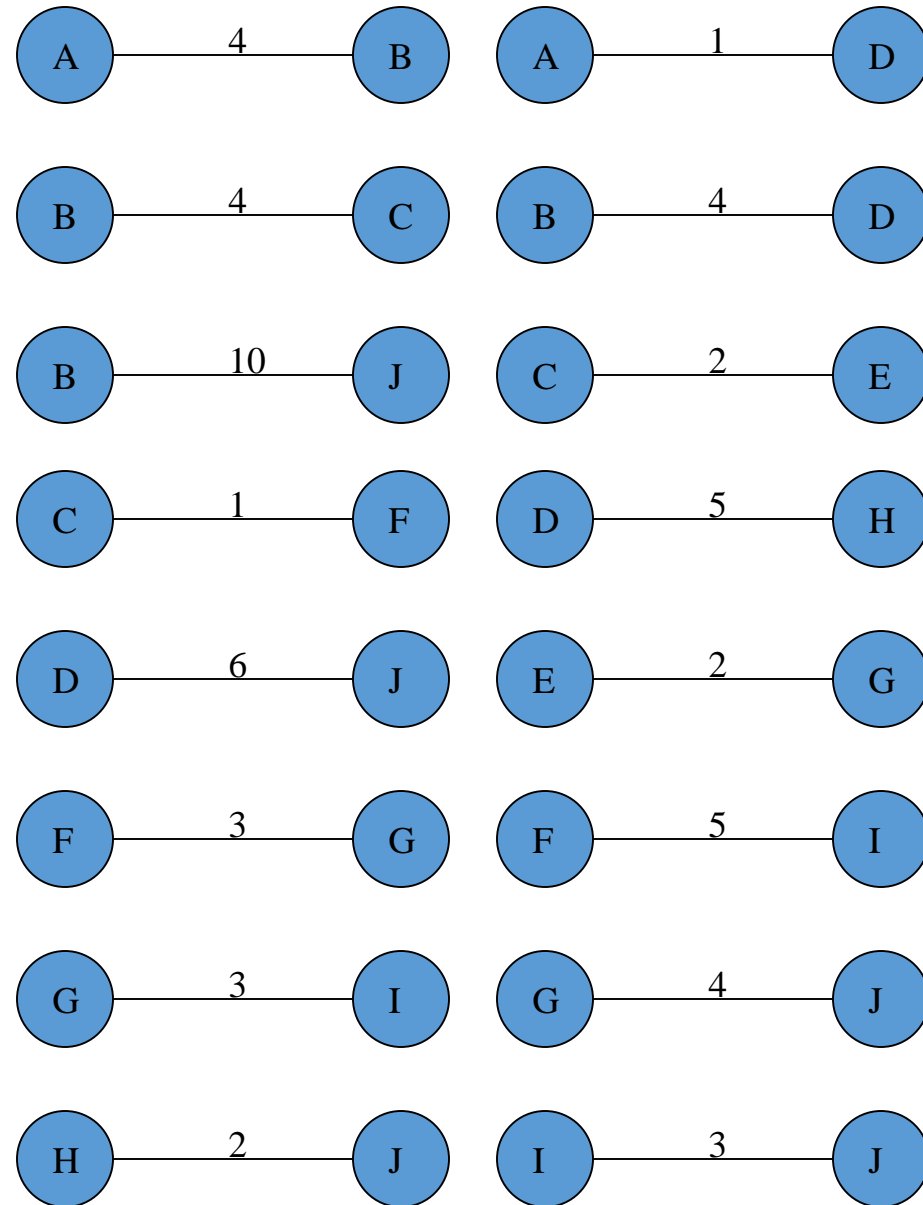
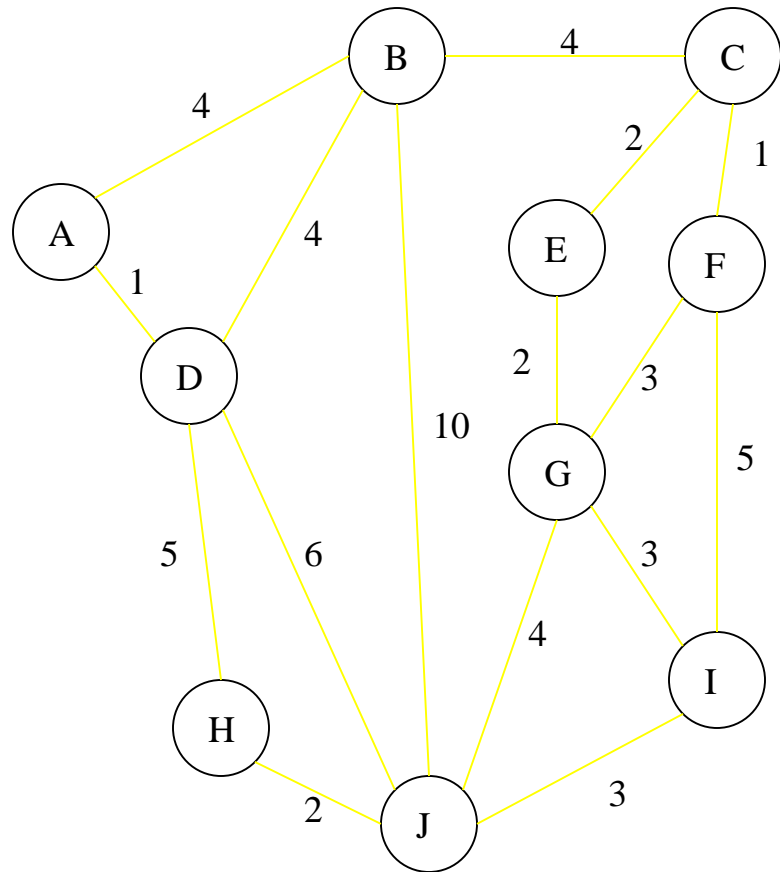
The steps are:

1. The forest is constructed - with each node in a separate tree.
2. The edges are placed in a priority queue.
3. Until we've added  $n-1$  edges
  1. Extract the cheapest edge from the queue
  2. If it forms a cycle, reject it
  3. Else add it to the forest. Adding it to the forest will join two trees together.

Every step will have joined two trees in the forest together, so that at the end, there will only be one tree in T.

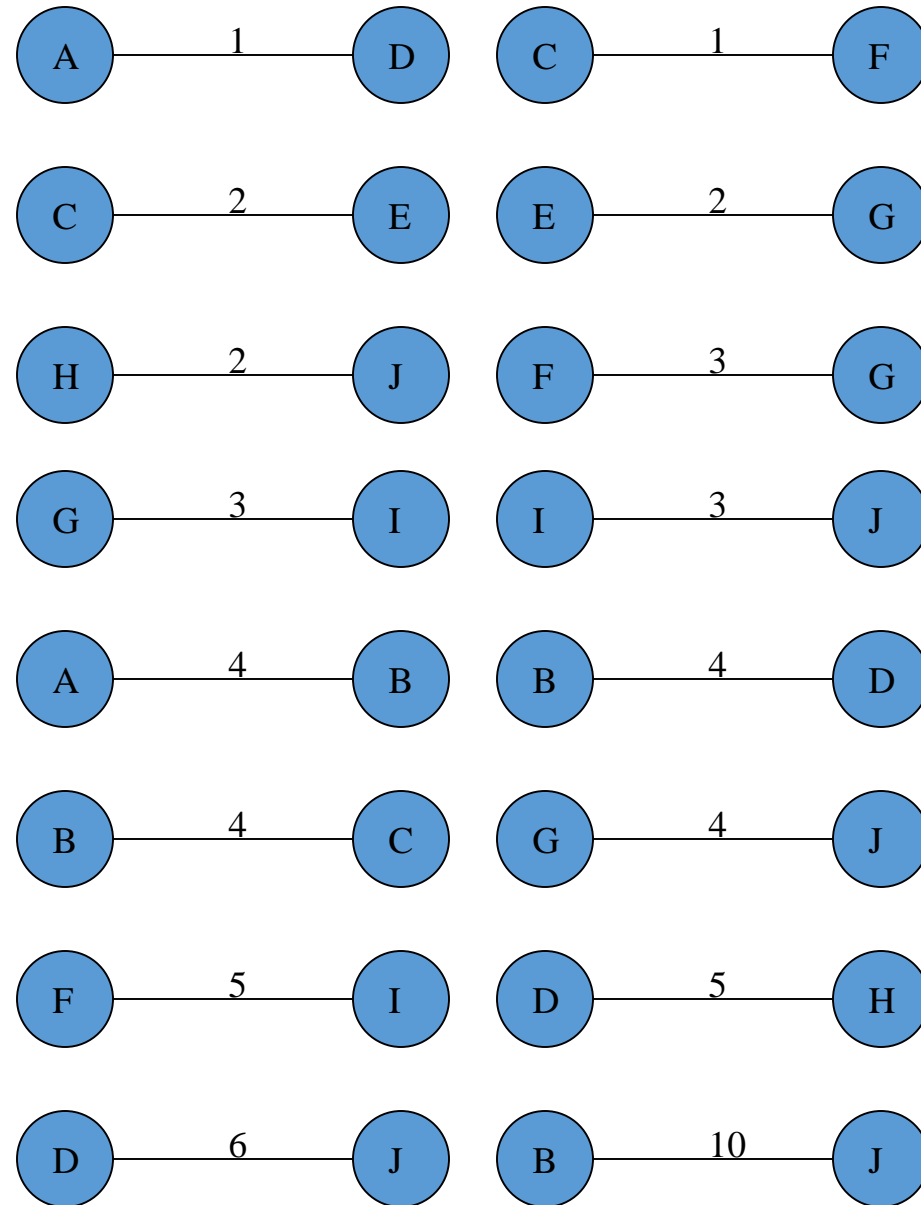
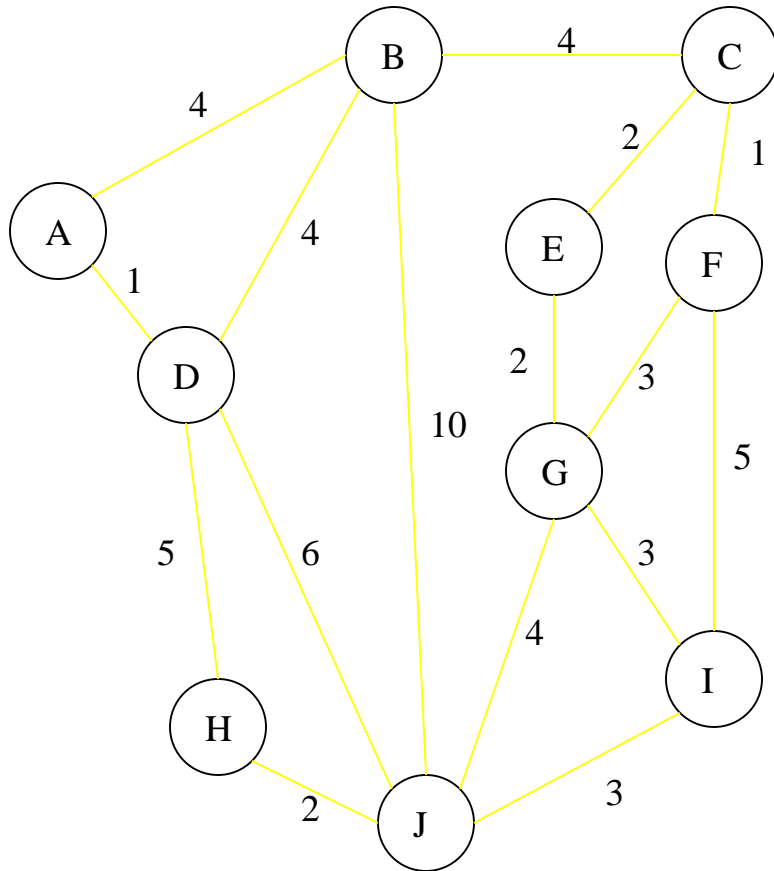
## Complete Graph





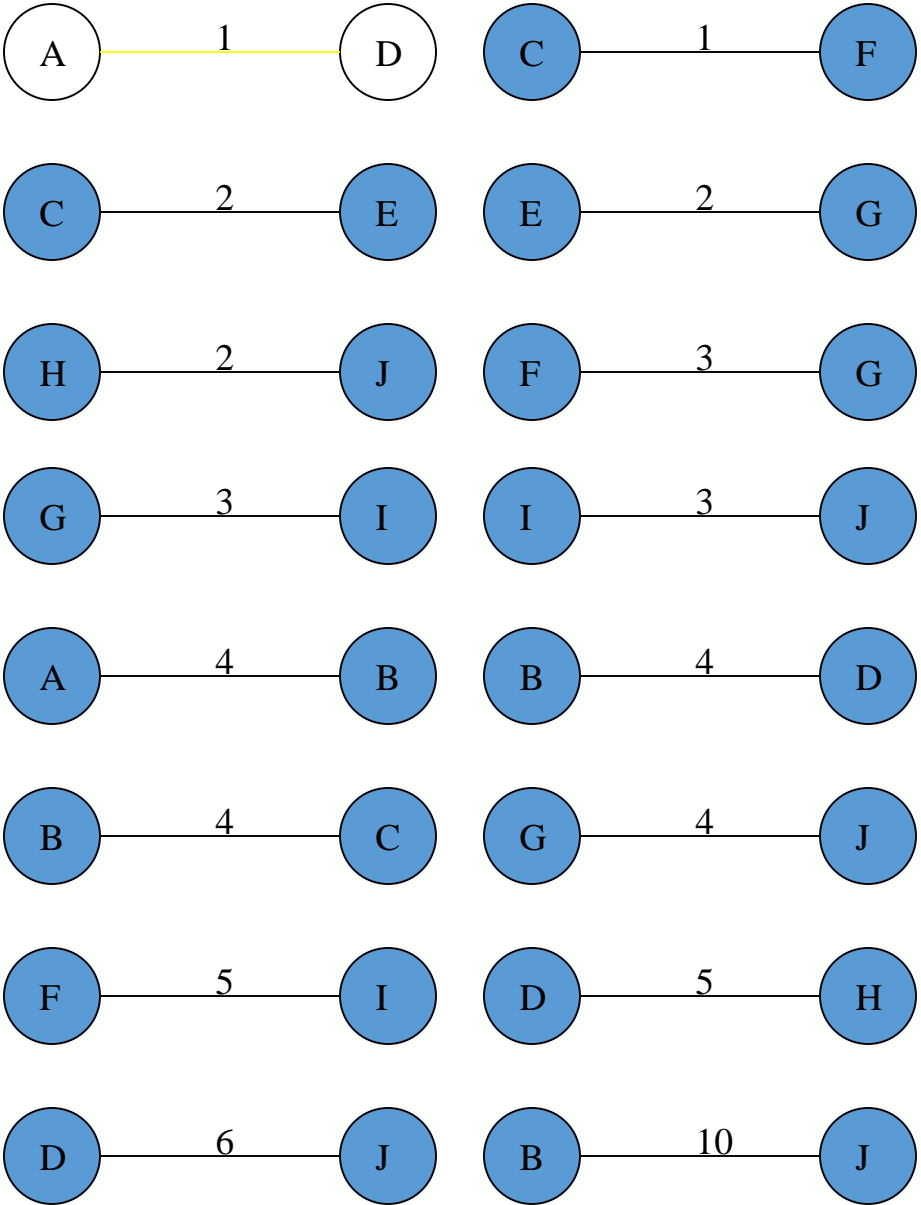
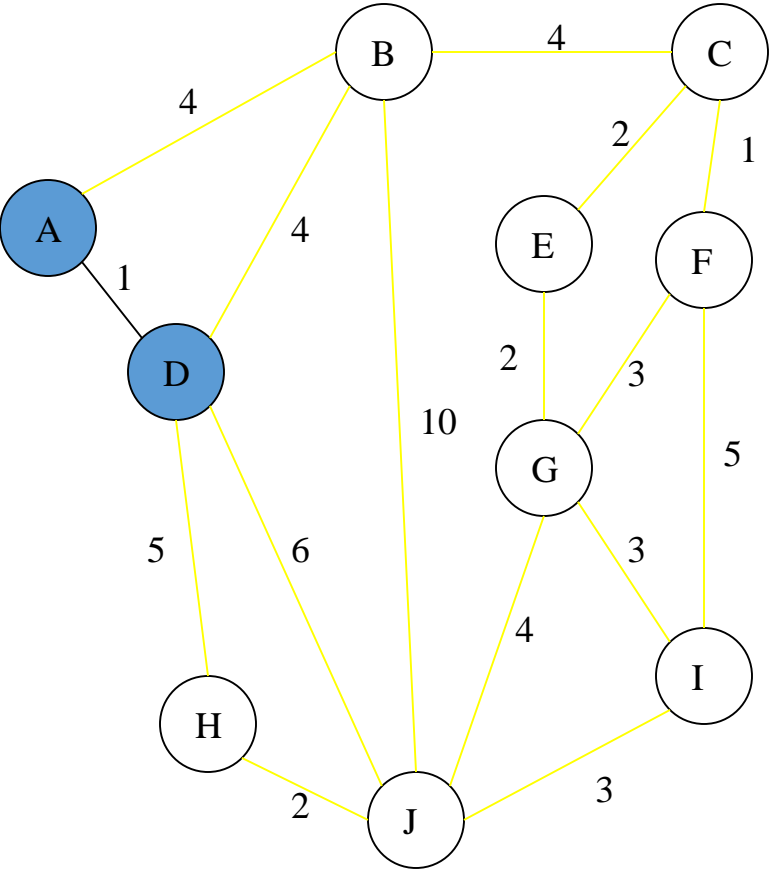
## Sort Edges

(in reality they are placed in a priority queue - not sorted - but sorting them makes the algorithm easier to visualize)

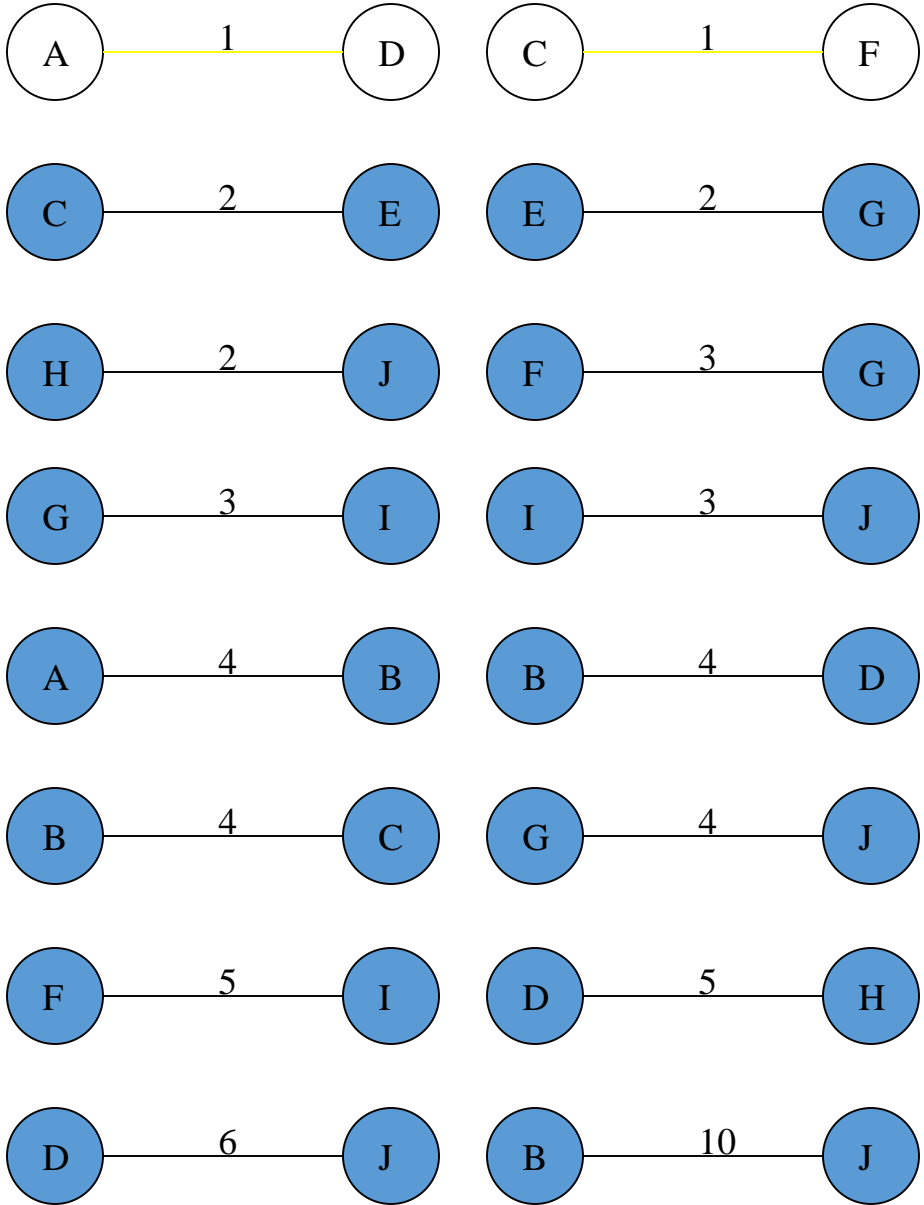
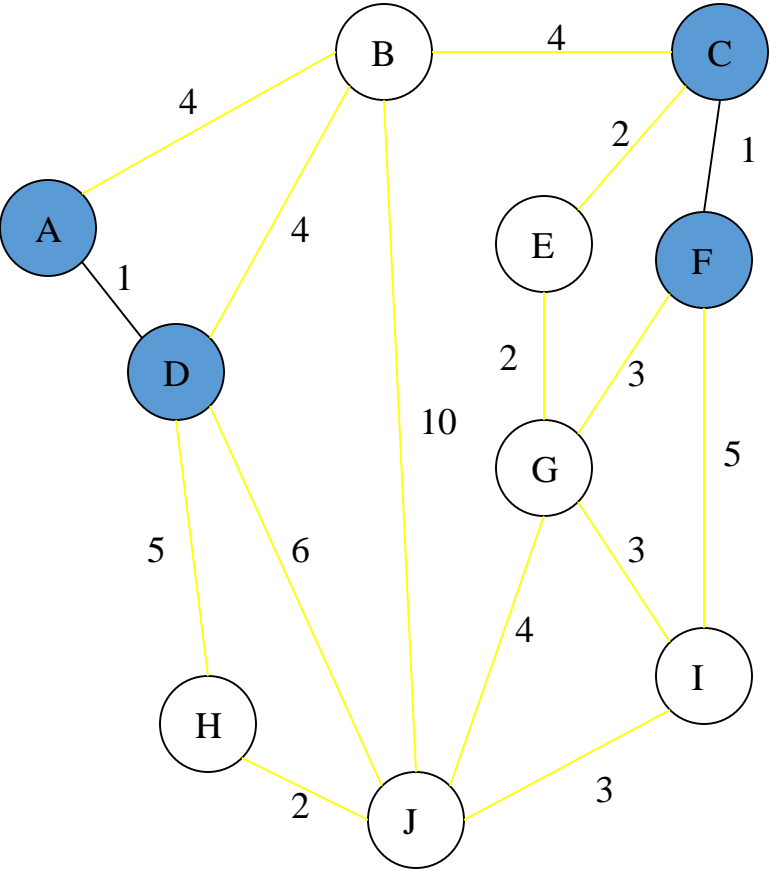




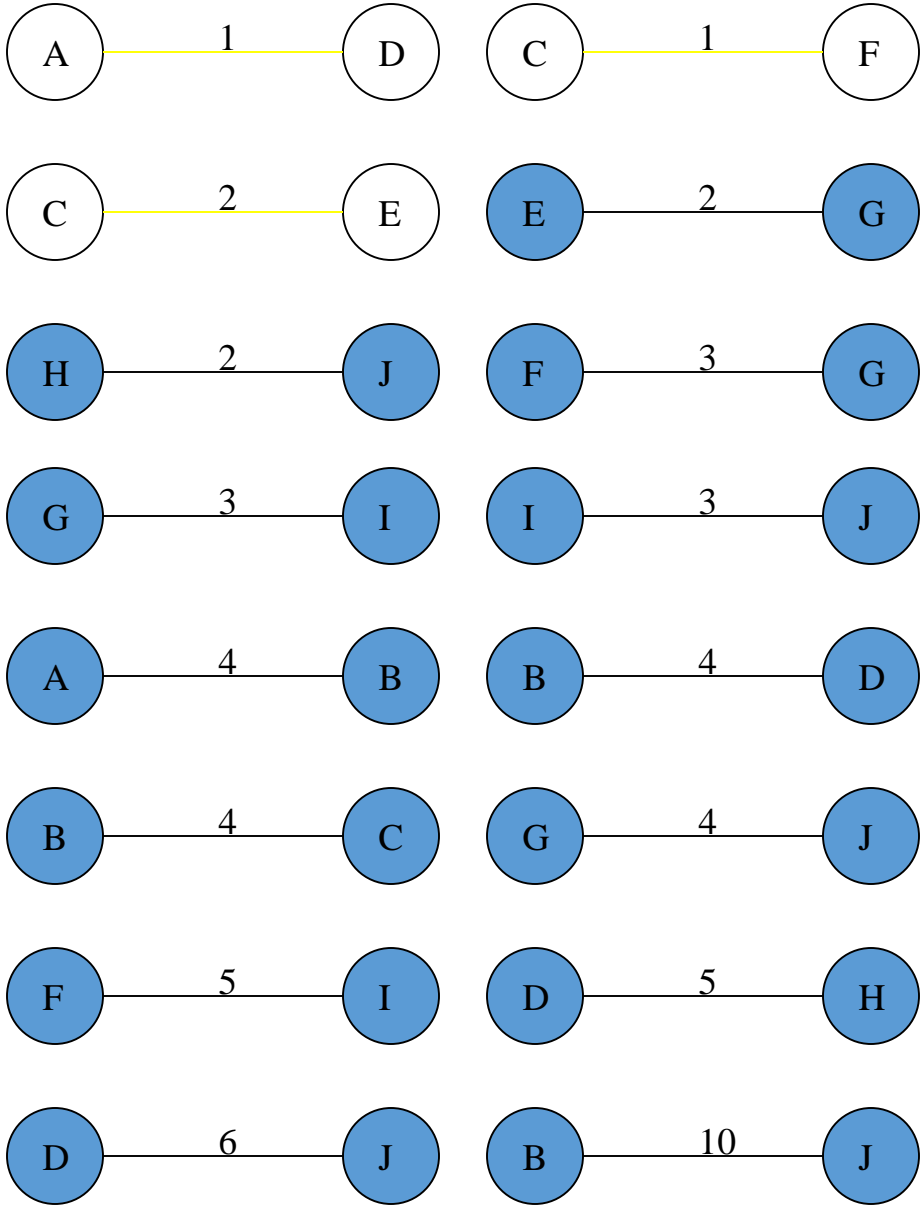
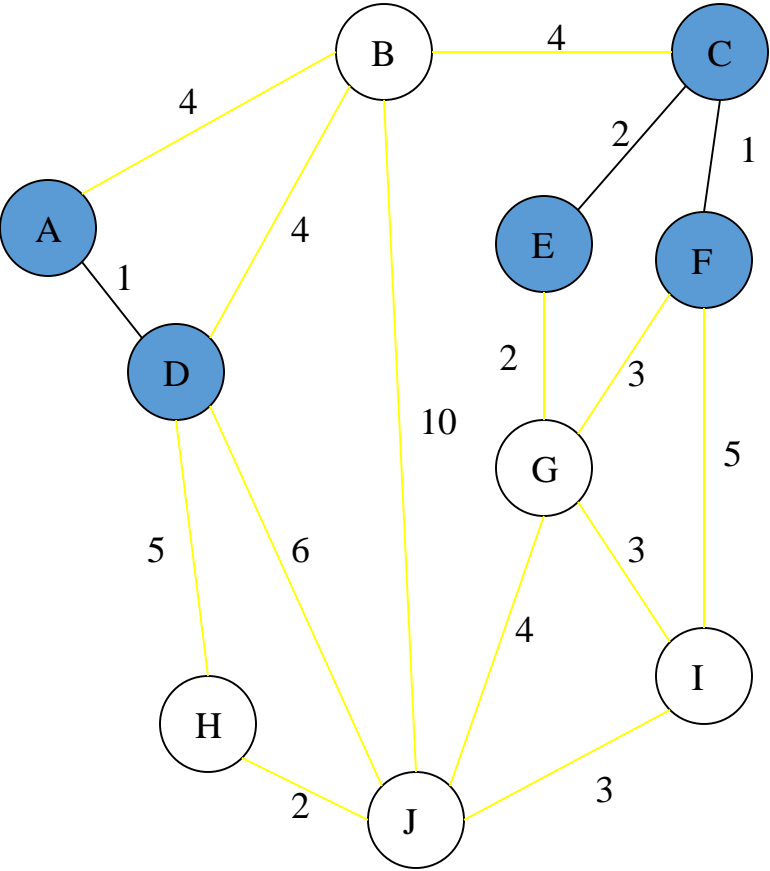
Add Edge



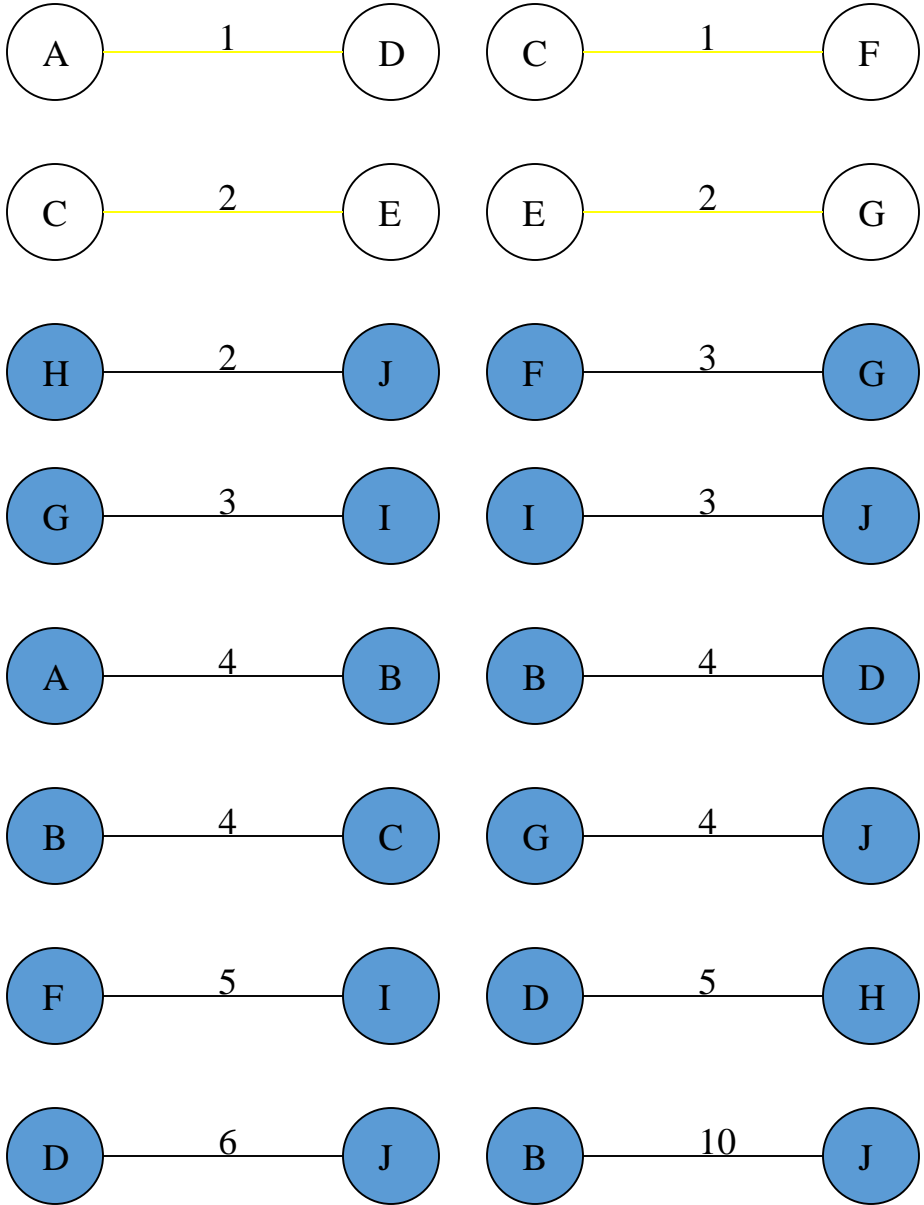
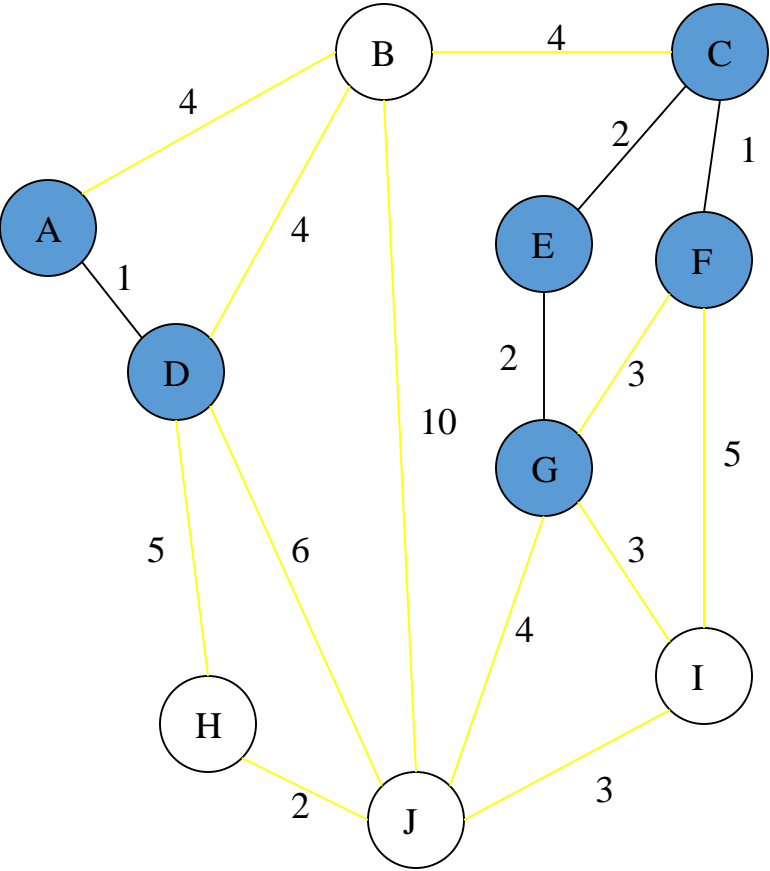
Add Edge



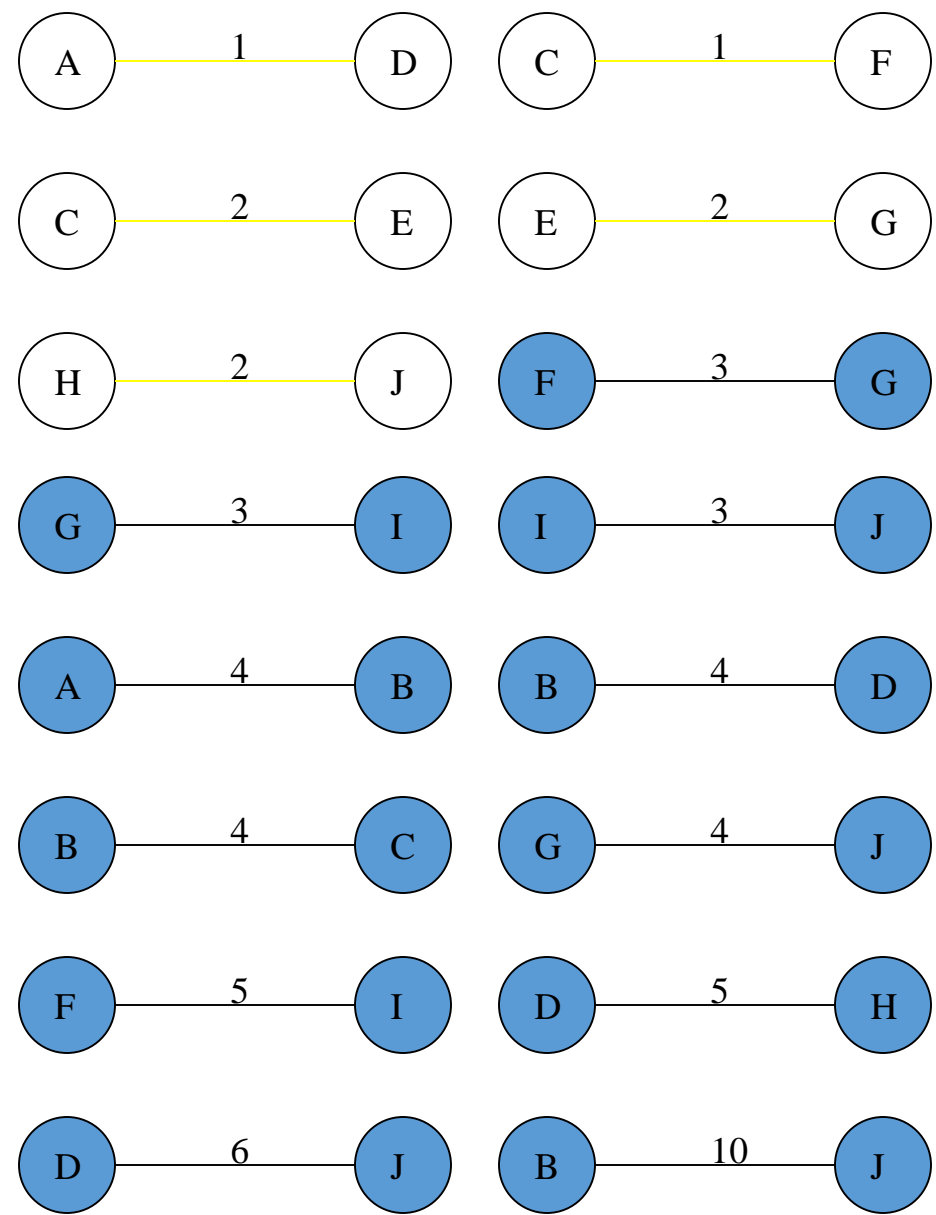
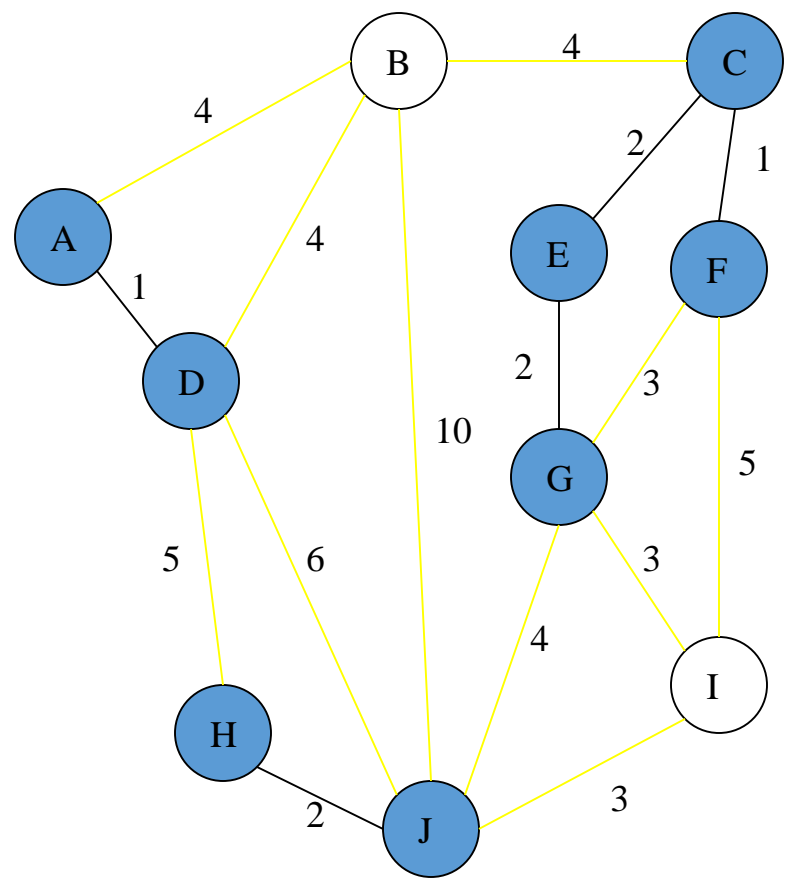
Add Edge



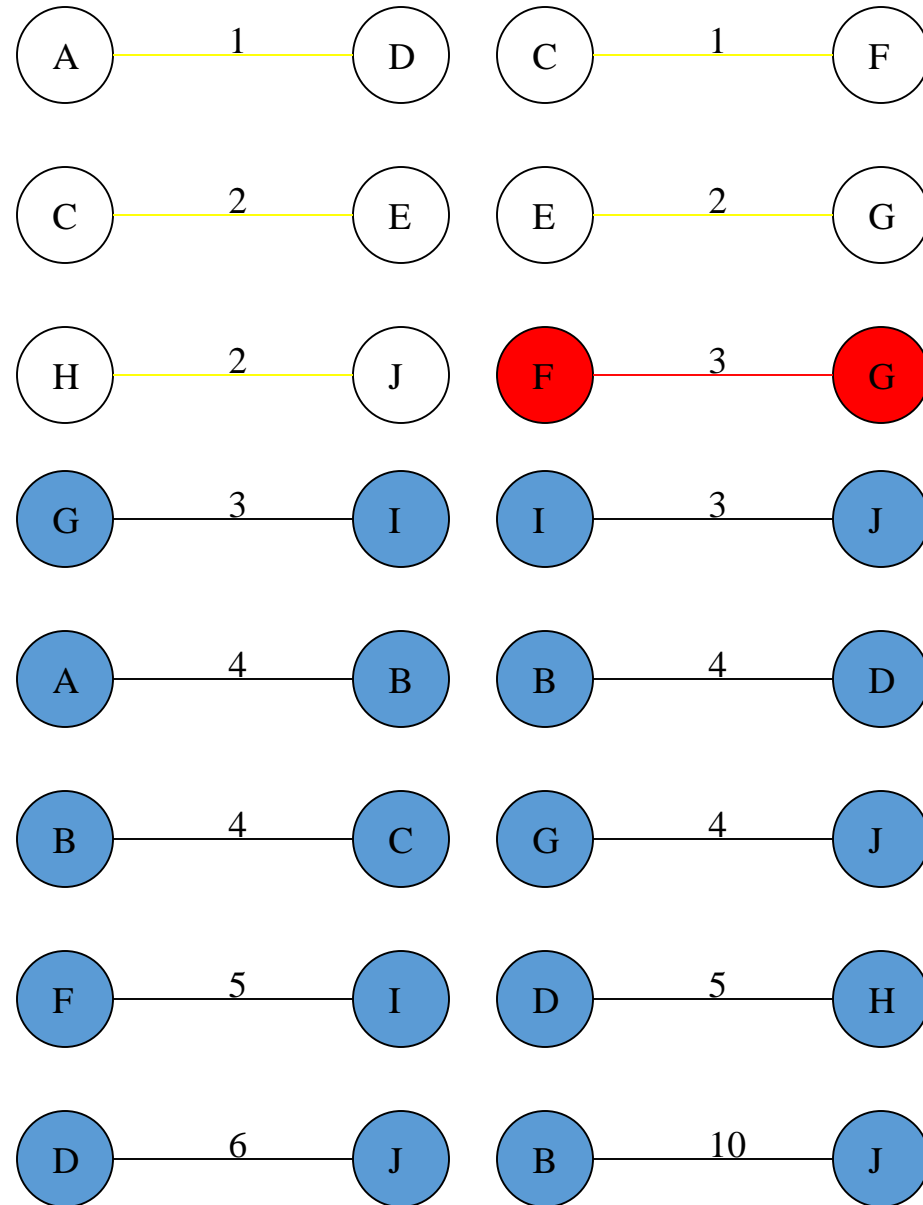
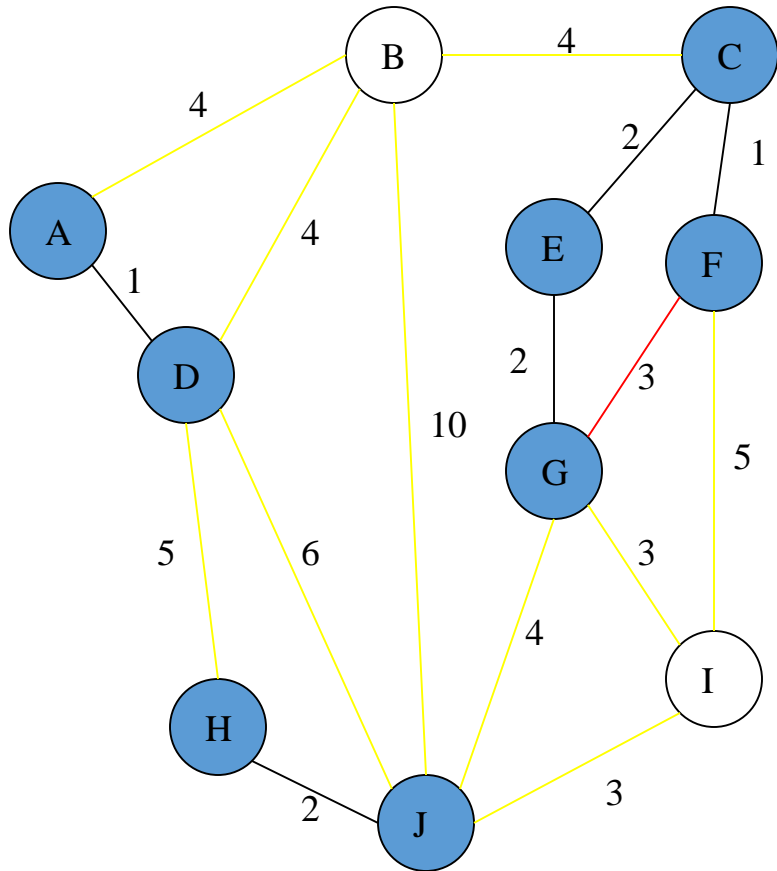
Add Edge



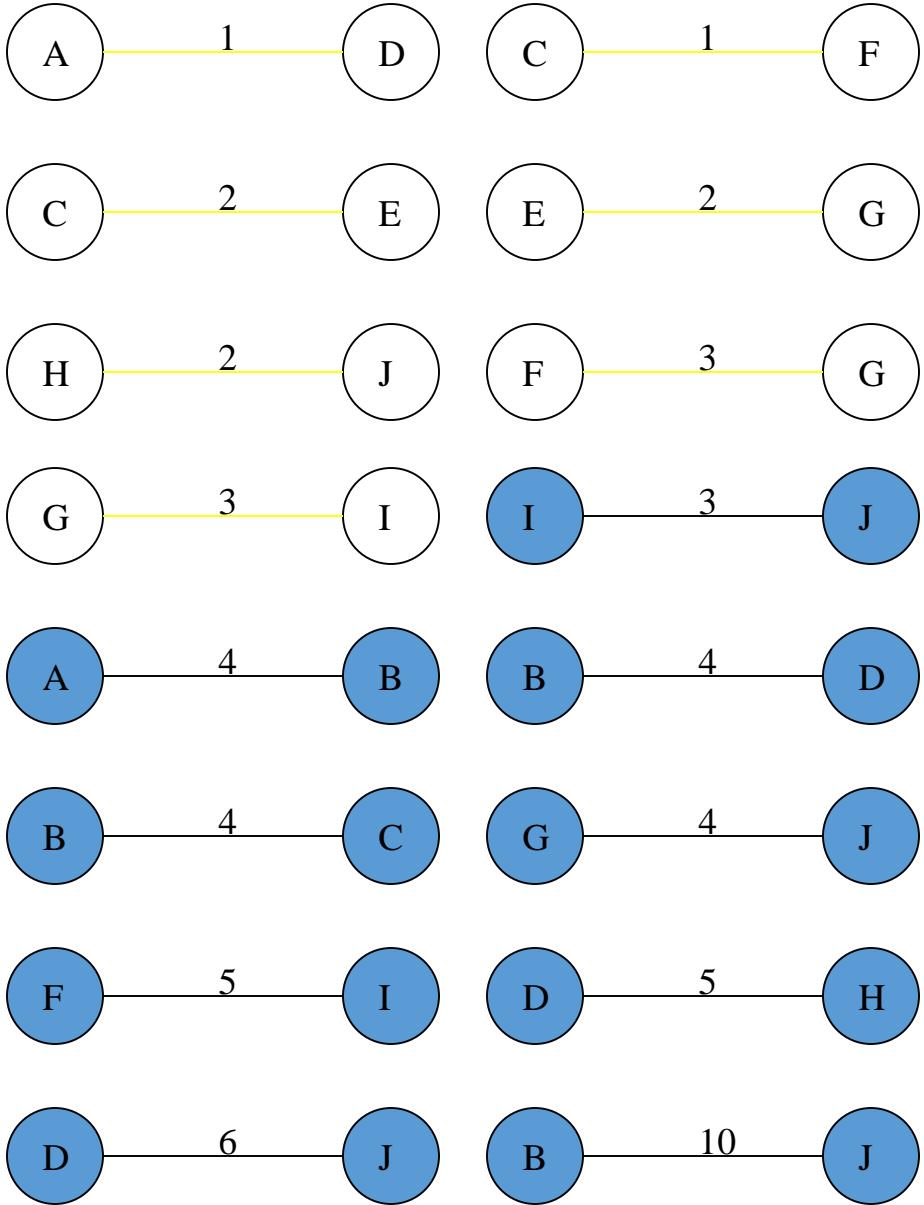
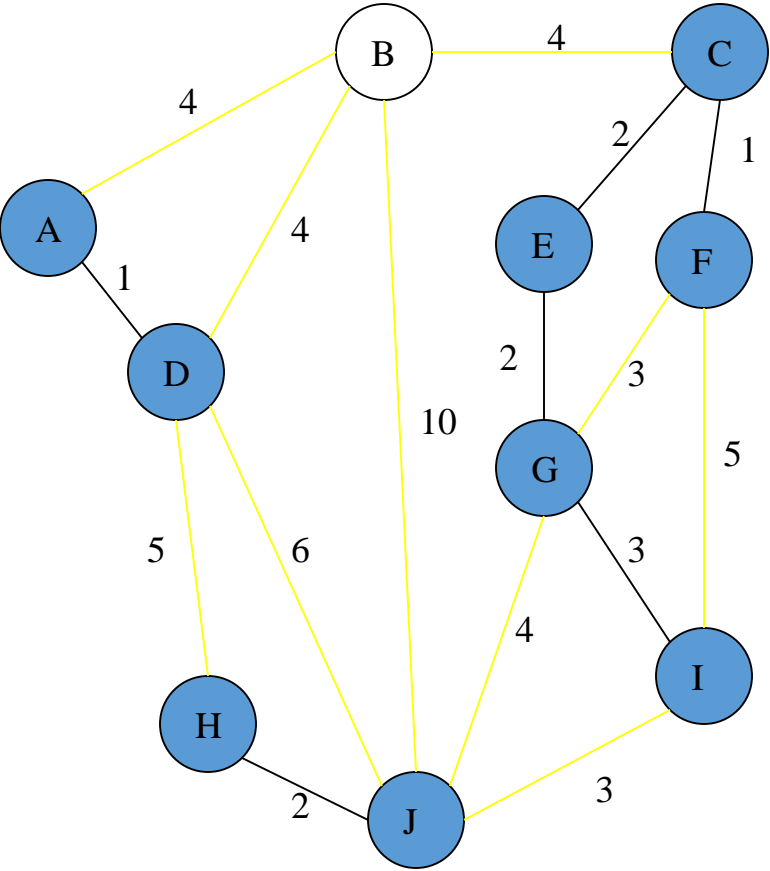
Add Edge



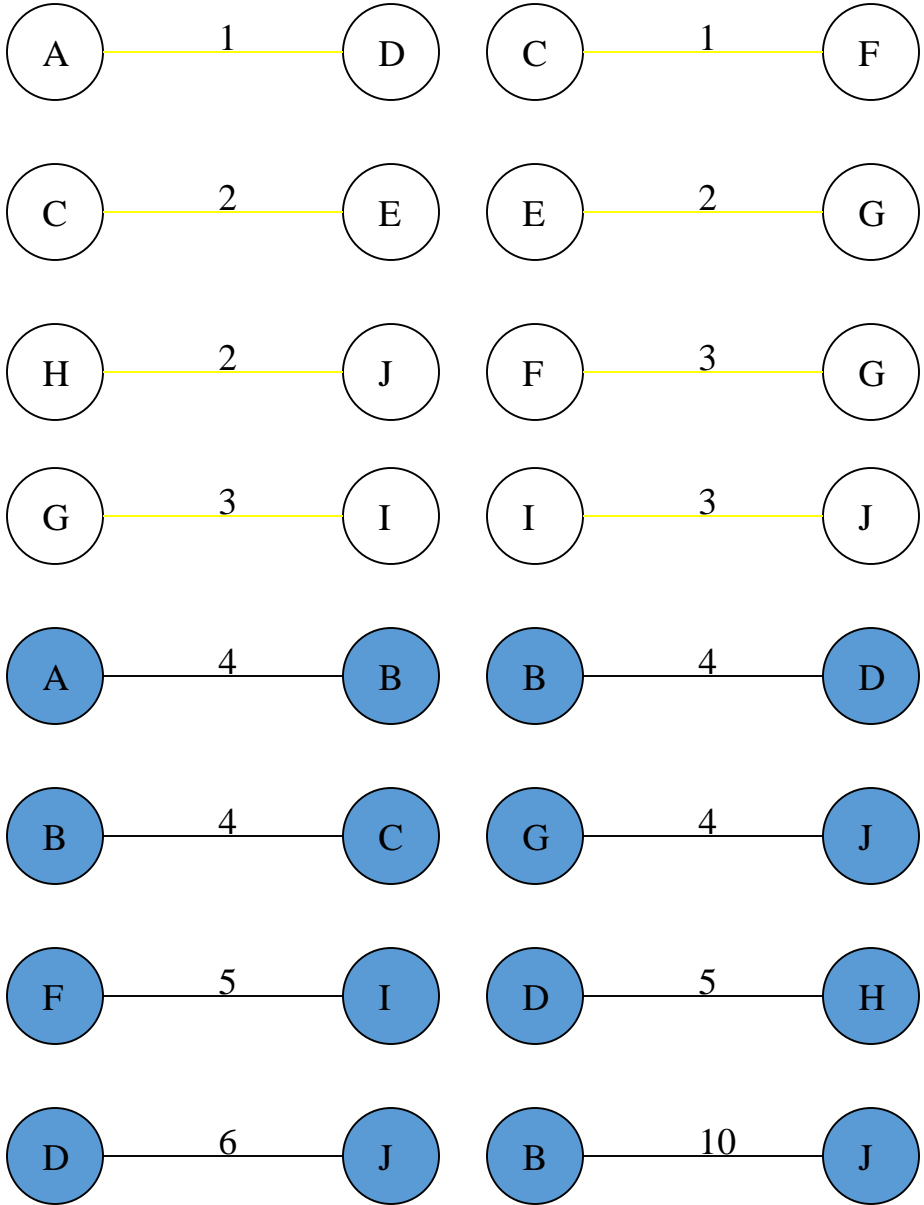
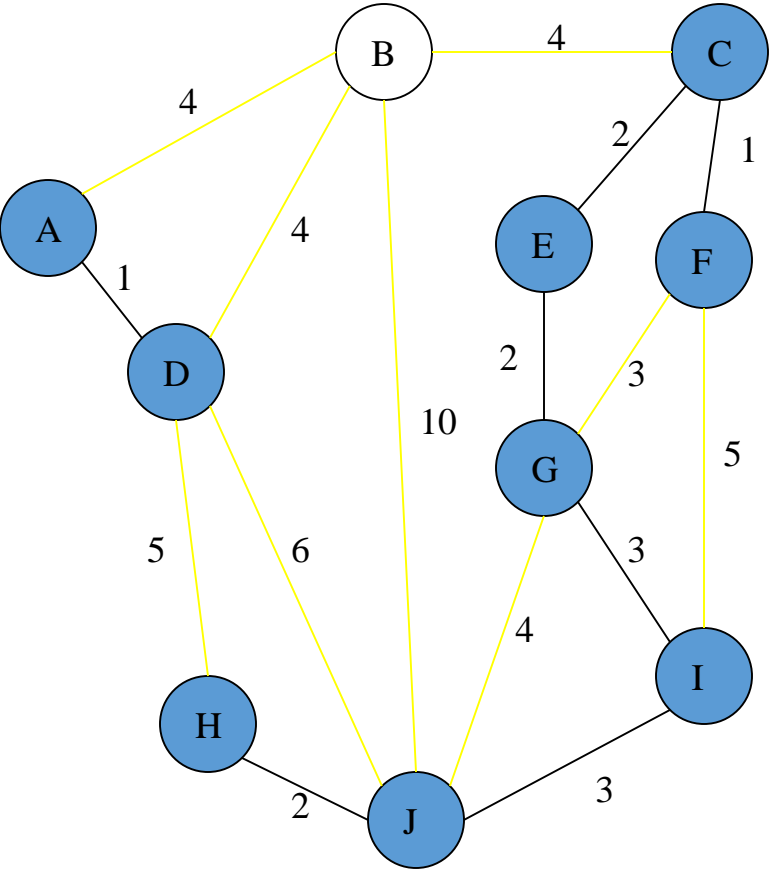
Cycle  
Don't Add Edge



Add Edge

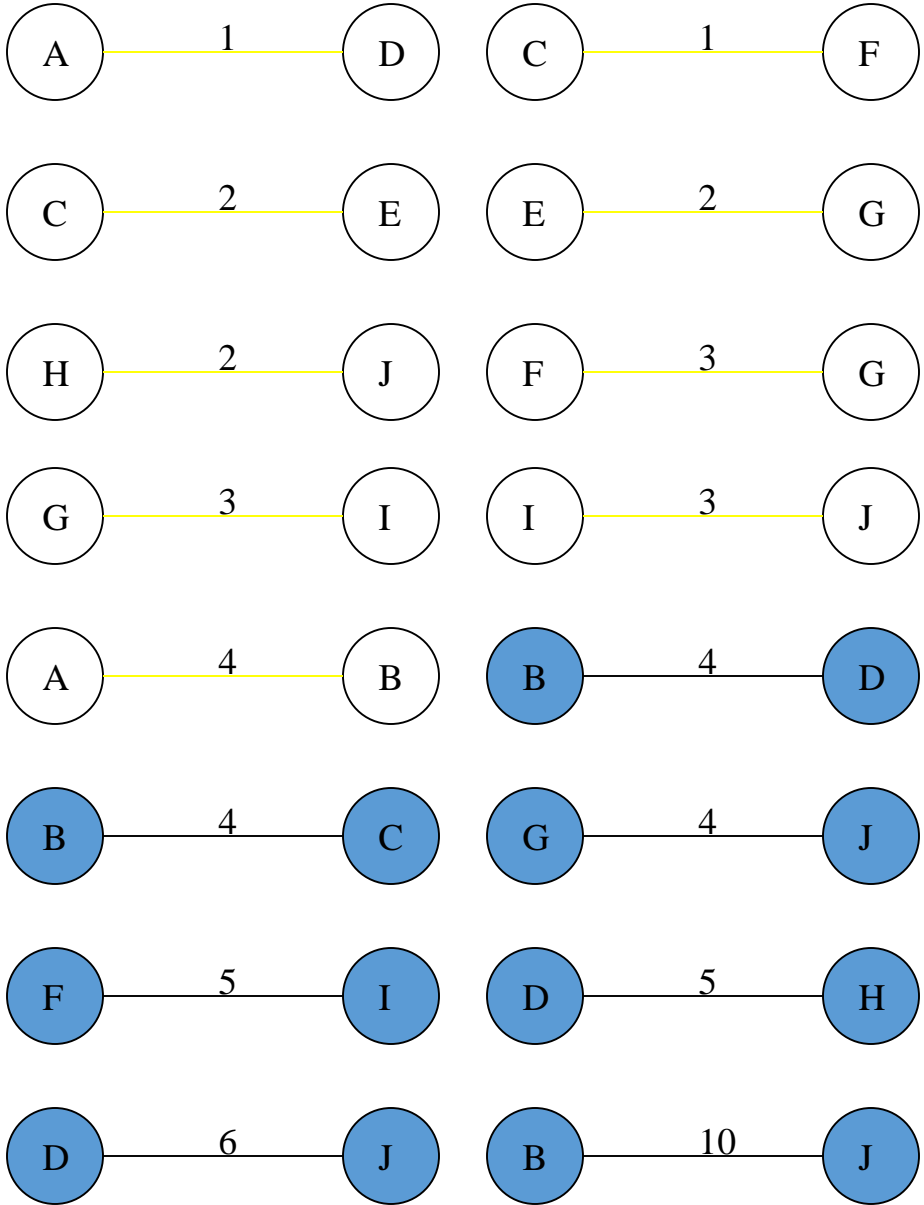
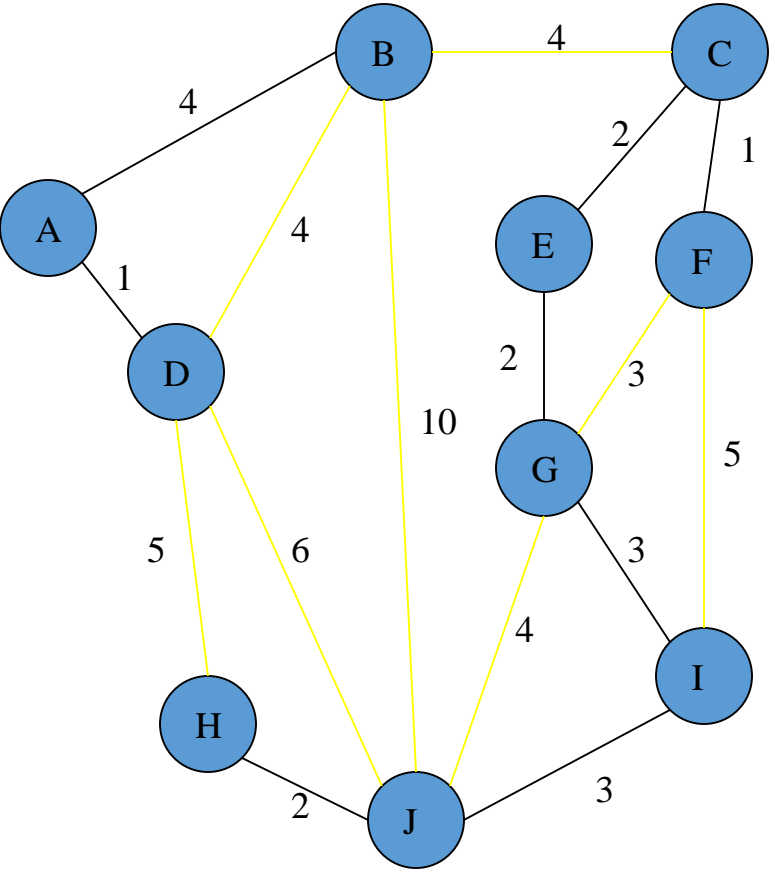


Add Edge

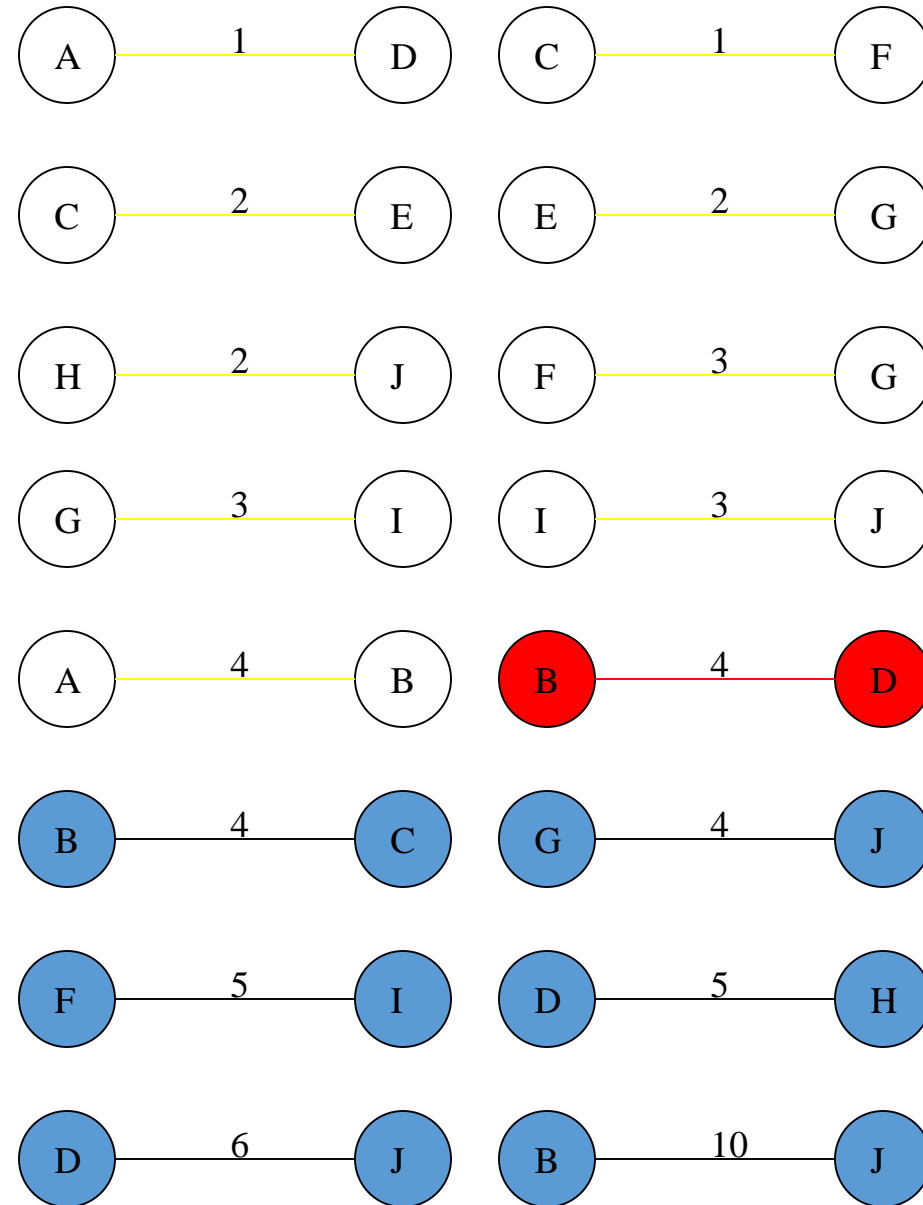
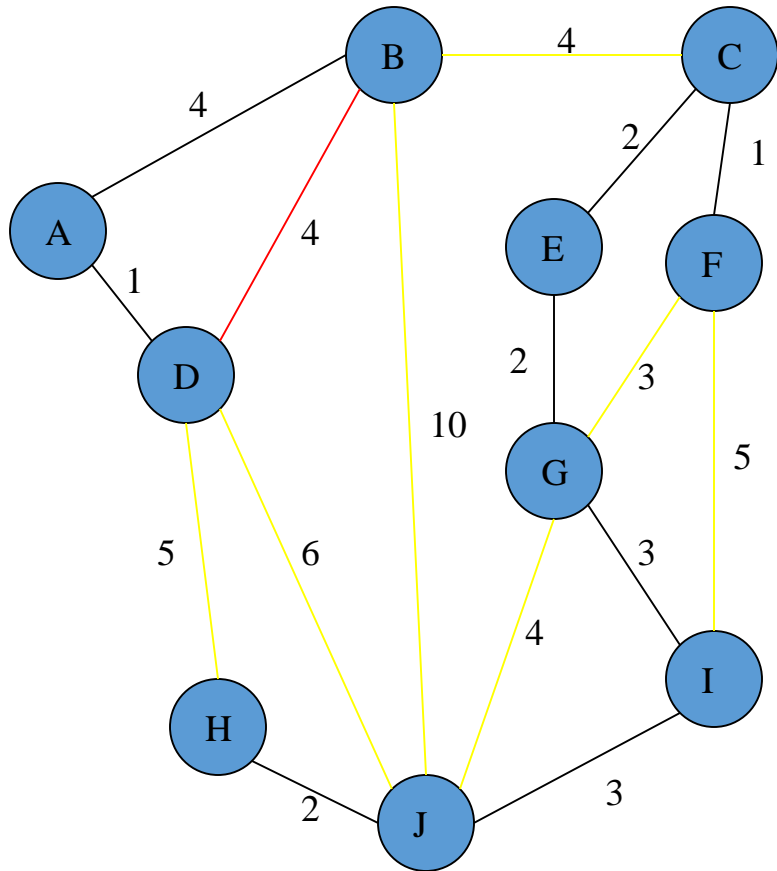




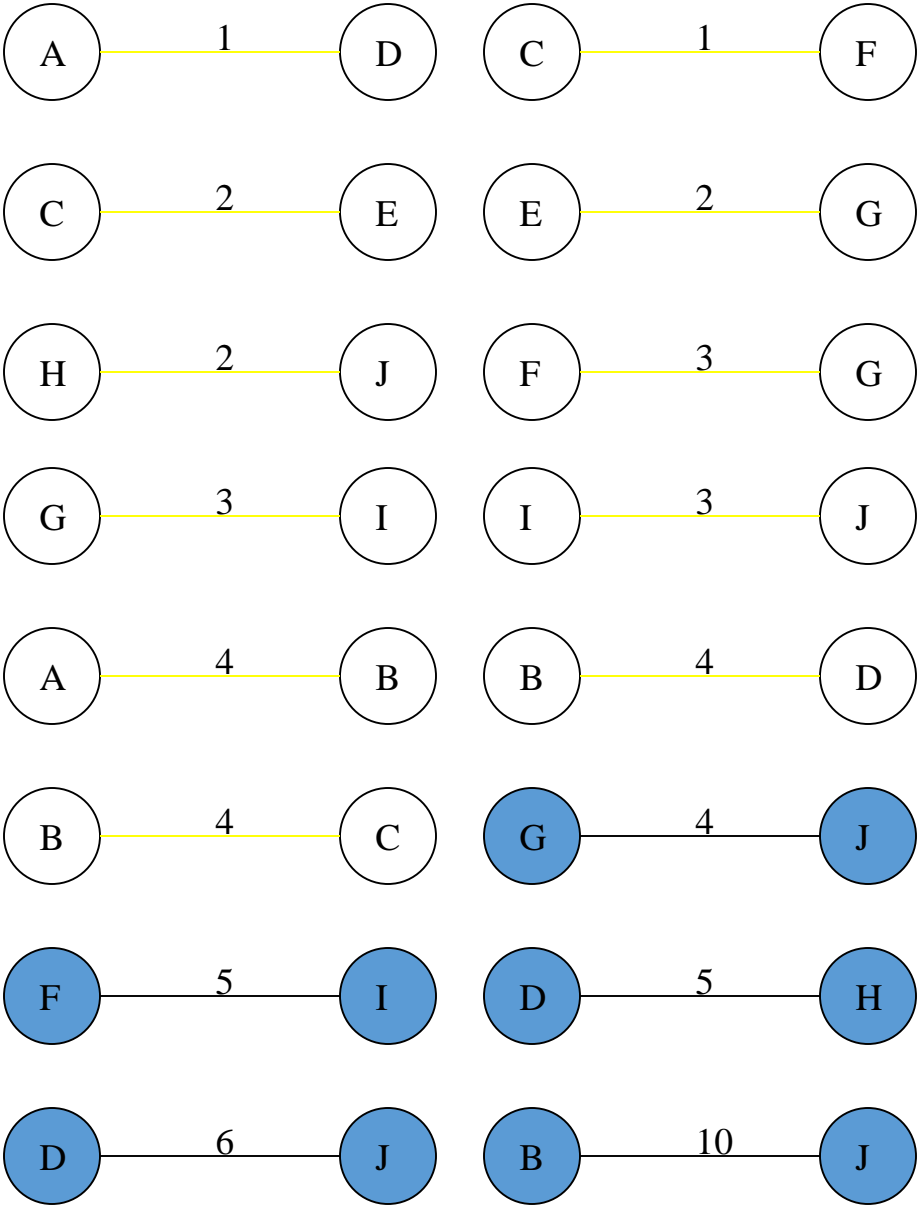
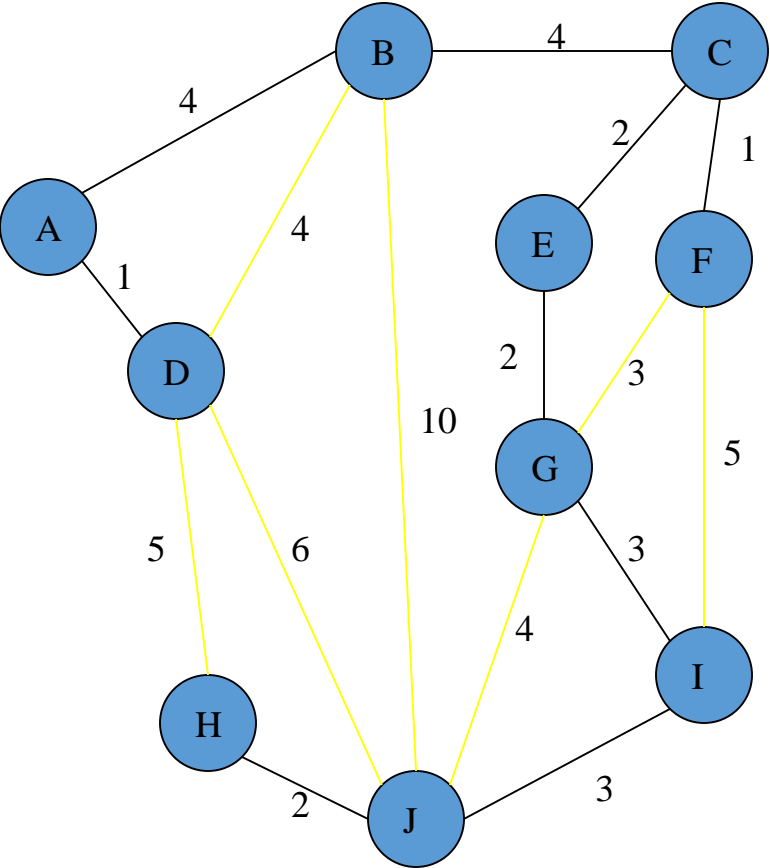
Add Edge



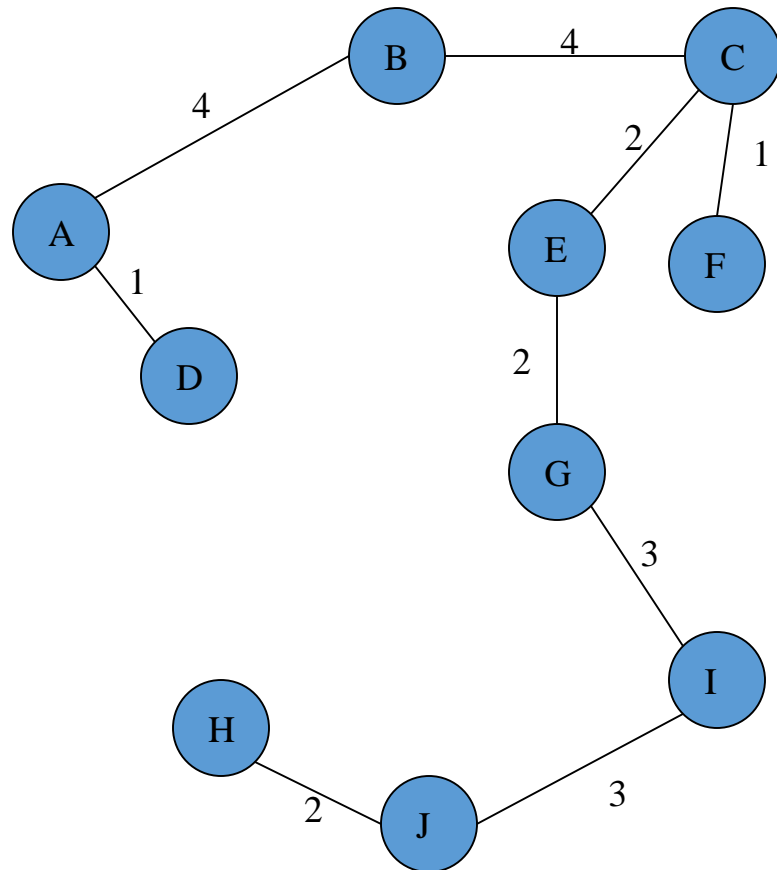
Cycle  
Don't Add Edge



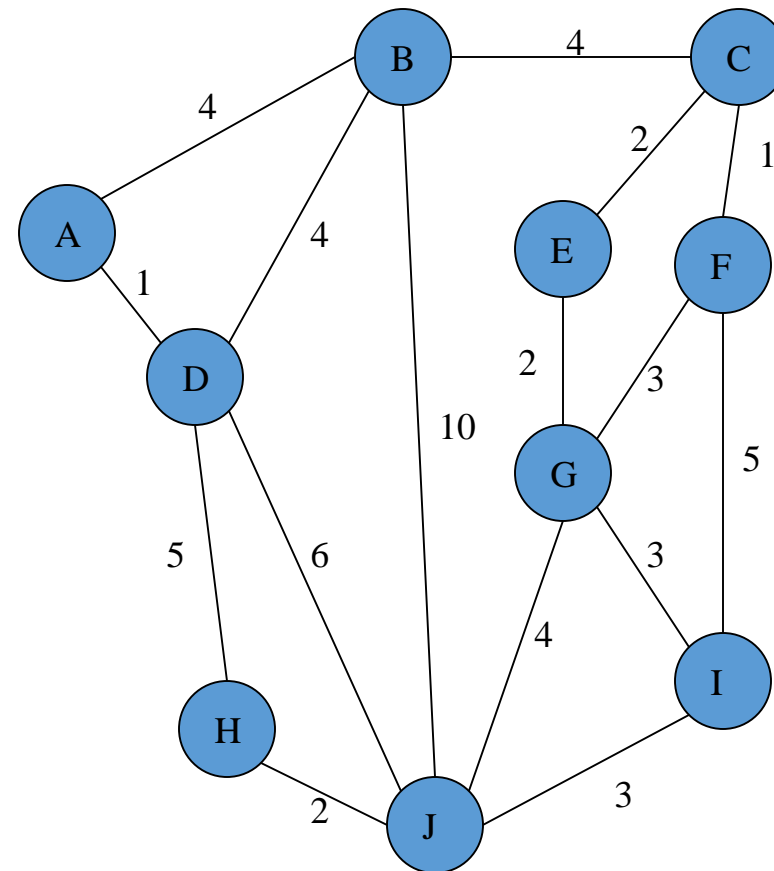
Add Edge



## Minimum Spanning Tree

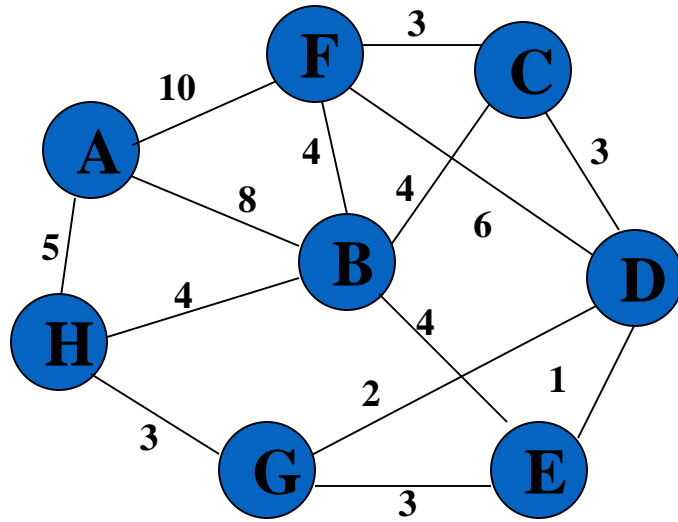


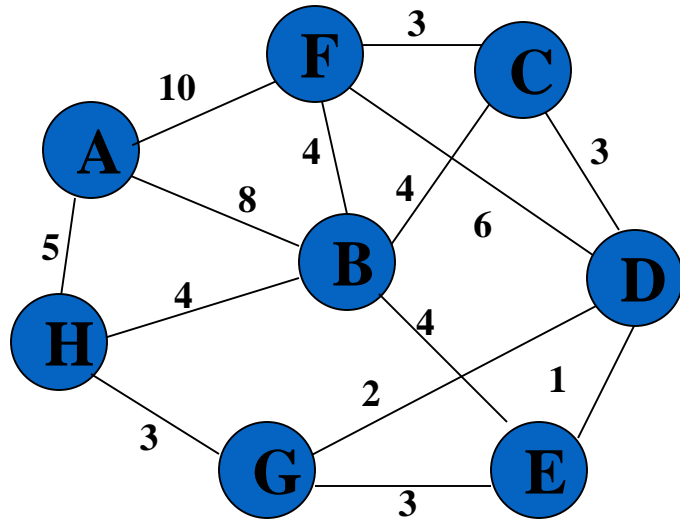
## Complete Graph



# Walk-Through

Consider an undirected, weight graph

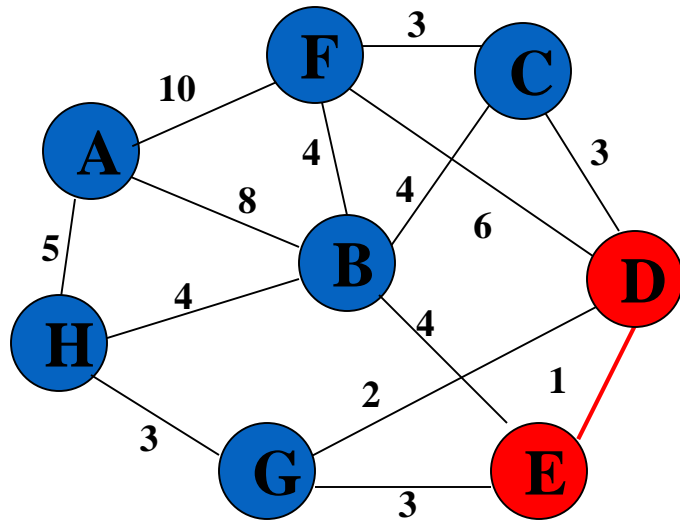




Sort the edges by increasing edge weight

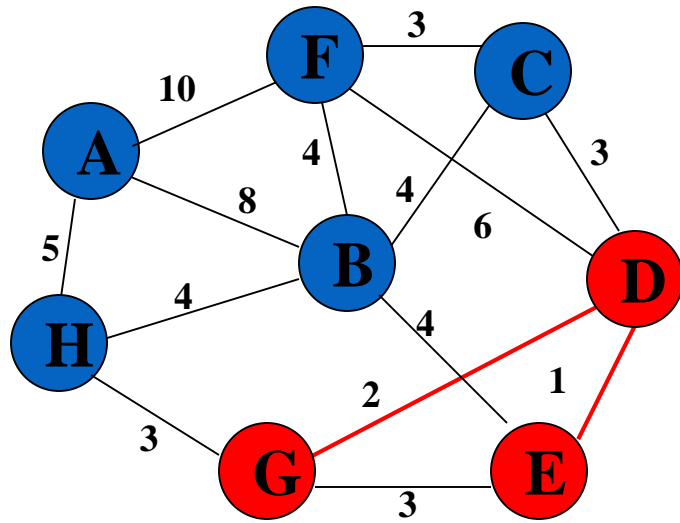
<i>edge</i>	$d_v$	
(D,E)	1	
(D,G)	2	
(E,G)	3	
(C,D)	3	
(G,H)	3	
(C,F)	3	
(B,C)	4	

<i>edge</i>	$d_v$	
(B,E)	4	
(B,F)	4	
(B,H)	4	
(A,H)	5	
(D,F)	6	
(A,B)	8	
(A,F)	10	



<i>edge</i>	$d_v$	
(D,E)	1	✓
(D,G)	2	
(E,G)	3	
(C,D)	3	
(G,H)	3	
(C,F)	3	
(B,C)	4	

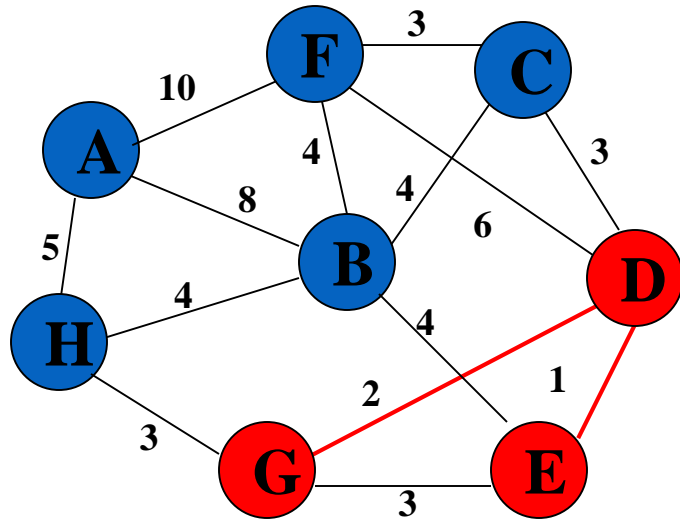
<i>edge</i>	$d_v$	
(B,E)	4	
(B,F)	4	
(B,H)	4	
(A,H)	5	
(D,F)	6	
(A,B)	8	
(A,F)	10	



<i>edge</i>	$d_v$	
(D,E)	1	✓
(D,G)	2	✓
(E,G)	3	
(C,D)	3	
(G,H)	3	
(C,F)	3	
(B,C)	4	

<i>edge</i>	$d_v$	
(B,E)	4	
(B,F)	4	
(B,H)	4	
(A,H)	5	
(D,F)	6	
(A,B)	8	
(A,F)	10	

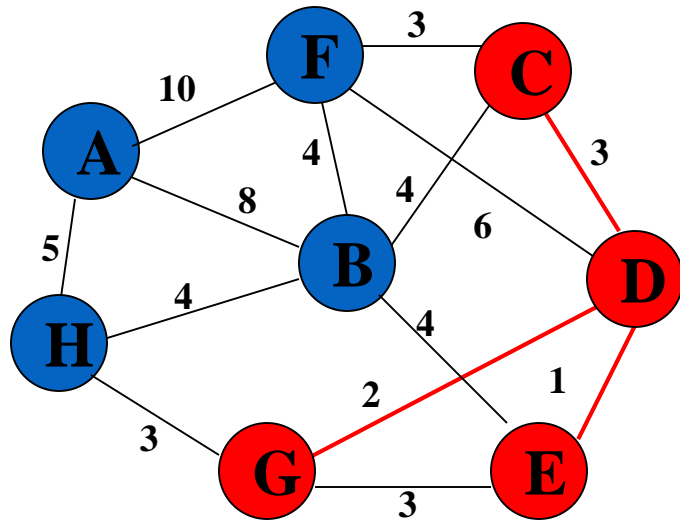




<i>edge</i>	$d_v$	
(D,E)	1	✓
(D,G)	2	✓
(E,G)	3	✗
(C,D)	3	
(G,H)	3	
(C,F)	3	
(B,C)	4	

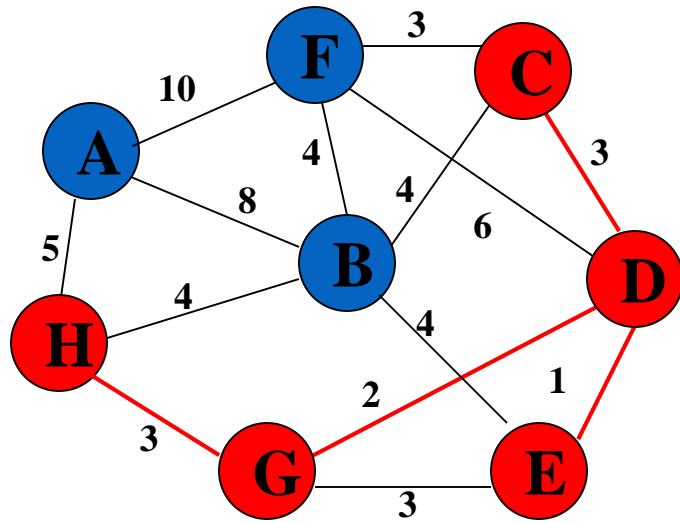
<i>edge</i>	$d_v$	
(B,E)	4	
(B,F)	4	
(B,H)	4	
(A,H)	5	
(D,F)	6	
(A,B)	8	
(A,F)	10	

Accepting edge (E,G) would create a cycle



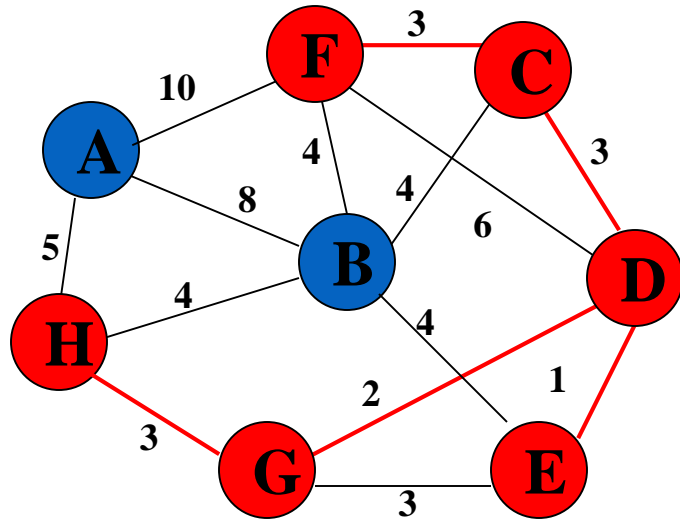
<i>edge</i>	$d_v$	
(D,E)	1	✓
(D,G)	2	✓
(E,G)	3	✗
(C,D)	3	✓
(G,H)	3	
(C,F)	3	
(B,C)	4	

<i>edge</i>	$d_v$	
(B,E)	4	
(B,F)	4	
(B,H)	4	
(A,H)	5	
(D,F)	6	
(A,B)	8	
(A,F)	10	



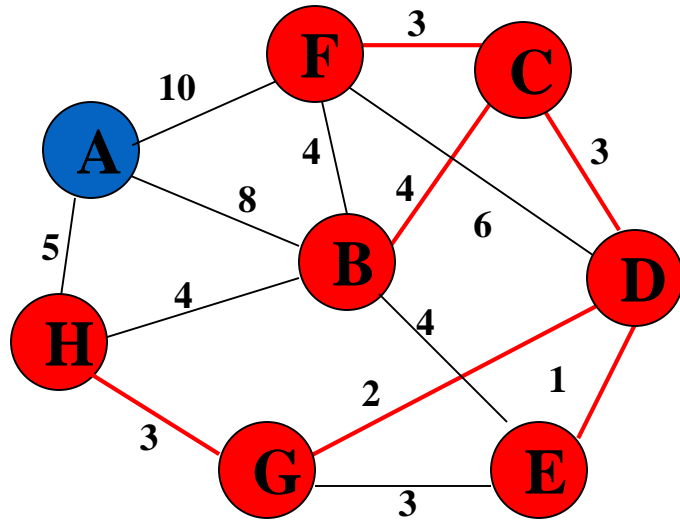
<i>edge</i>	$d_v$	
(D,E)	1	✓
(D,G)	2	✓
(E,G)	3	✗
(C,D)	3	✓
(G,H)	3	✓
(C,F)	3	
(B,C)	4	

<i>edge</i>	$d_v$	
(B,E)	4	
(B,F)	4	
(B,H)	4	
(A,H)	5	
(D,F)	6	
(A,B)	8	
(A,F)	10	



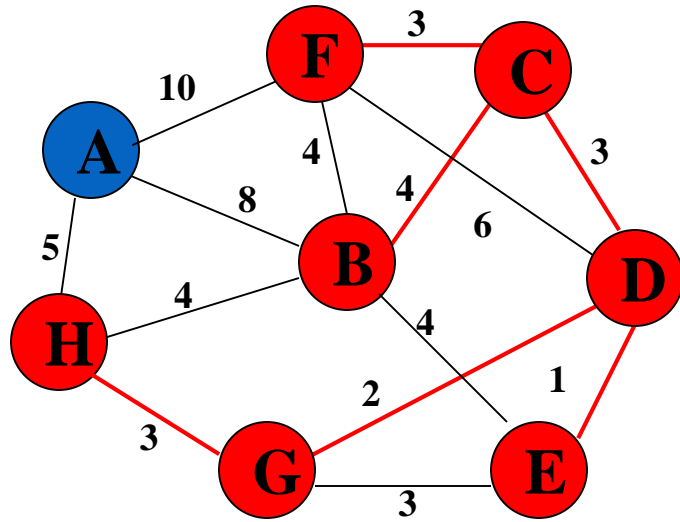
<i>edge</i>	$d_v$	
(D,E)	1	✓
(D,G)	2	✓
(E,G)	3	✗
(C,D)	3	✓
(G,H)	3	✓
(C,F)	3	✓
(B,C)	4	

<i>edge</i>	$d_v$	
(B,E)	4	
(B,F)	4	
(B,H)	4	
(A,H)	5	
(D,F)	6	
(A,B)	8	
(A,F)	10	



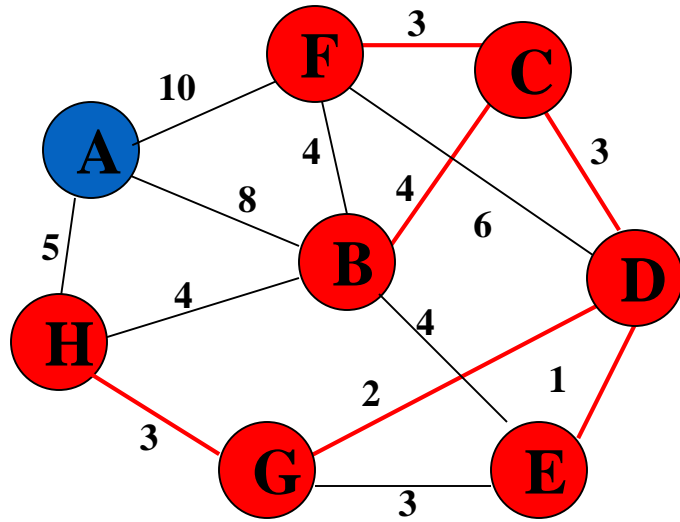
<i>edge</i>	$d_v$	
(D,E)	1	✓
(D,G)	2	✓
(E,G)	3	✗
(C,D)	3	✓
(G,H)	3	✓
(C,F)	3	✓
(B,C)	4	✓

<i>edge</i>	$d_v$	
(B,E)	4	
(B,F)	4	
(B,H)	4	
(A,H)	5	
(D,F)	6	
(A,B)	8	
(A,F)	10	



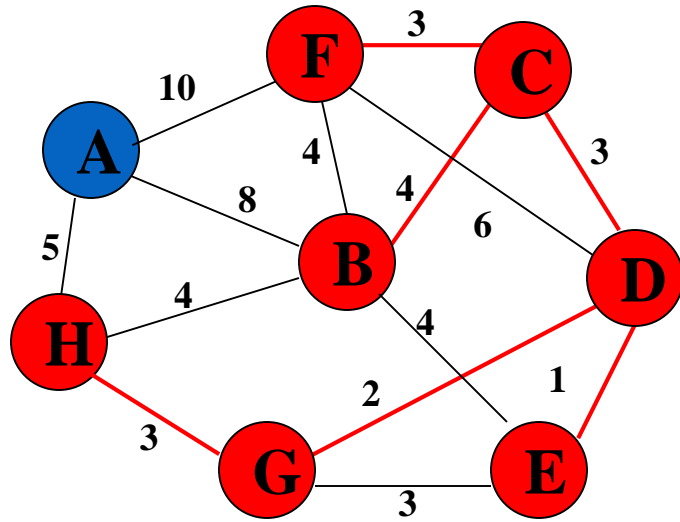
<i>edge</i>	$d_v$	
(D,E)	1	✓
(D,G)	2	✓
(E,G)	3	✗
(C,D)	3	✓
(G,H)	3	✓
(C,F)	3	✓
(B,C)	4	✓

<i>edge</i>	$d_v$	
(B,E)	4	✗
(B,F)	4	
(B,H)	4	
(A,H)	5	
(D,F)	6	
(A,B)	8	
(A,F)	10	



<i>edge</i>	$d_v$	
(D,E)	1	✓
(D,G)	2	✓
(E,G)	3	✗
(C,D)	3	✓
(G,H)	3	✓
(C,F)	3	✓
(B,C)	4	✓

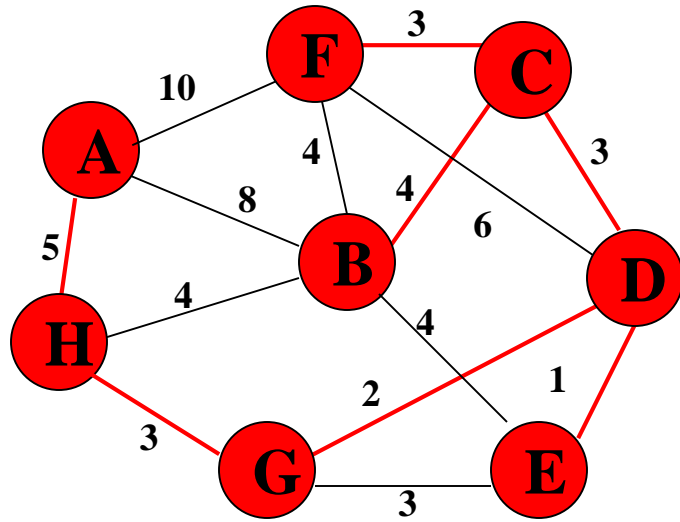
<i>edge</i>	$d_v$	
(B,E)	4	✗
(B,F)	4	✗
(B,H)	4	
(A,H)	5	
(D,F)	6	
(A,B)	8	
(A,F)	10	



<i>edge</i>	$d_v$	
(D,E)	1	✓
(D,G)	2	✓
(E,G)	3	✗
(C,D)	3	✓
(G,H)	3	✓
(C,F)	3	✓
(B,C)	4	✓

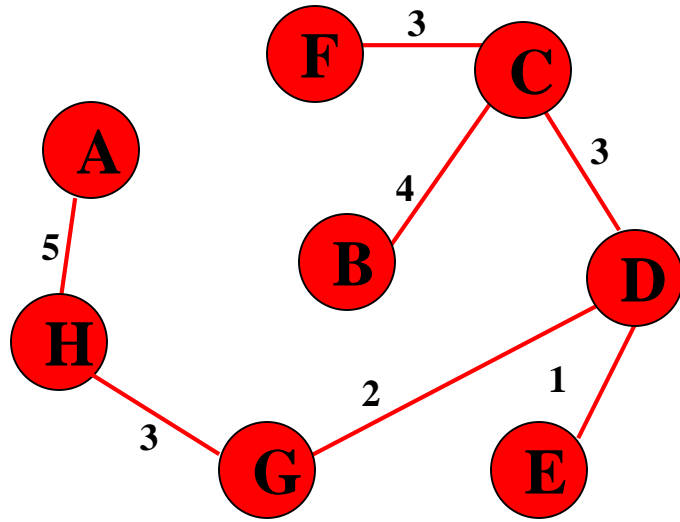
<i>edge</i>	$d_v$	
(B,E)	4	✗
(B,F)	4	✗
(B,H)	4	✗
(A,H)	5	
(D,F)	6	
(A,B)	8	
(A,F)	10	





<i>edge</i>	$d_v$	
(D,E)	1	✓
(D,G)	2	✓
(E,G)	3	✗
(C,D)	3	✓
(G,H)	3	✓
(C,F)	3	✓
(B,C)	4	✓

<i>edge</i>	$d_v$	
(B,E)	4	✗
(B,F)	4	✗
(B,H)	4	✗
(A,H)	5	✓
(D,F)	6	
(A,B)	8	
(A,F)	10	



<i>edge</i>	$d_v$	
(D,E)	1	✓
(D,G)	2	✓
(E,G)	3	✗
(C,D)	3	✓
(G,H)	3	✓
(C,F)	3	✓
(B,C)	4	✓

<i>edge</i>	$d_v$	
(B,E)	4	✗
(B,F)	4	✗
(B,H)	4	✗
(A,H)	5	✓
(D,F)	6	
(A,B)	8	
(A,F)	10	

} not  
considered

**Done**

**Total Cost =  $\Sigma d_v =$   
*21***

# Analysis of Kruskal's Algorithm

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Running Time =  $O(e \log v)$       ( $e$  = edges,  $v$  = nodes)

Testing if an edge creates a cycle can be slow

This algorithm works best if the number of edges is kept to a minimum.

# Conclusion

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Prim's has a better complexity than Kruskal's, so it is better used with dense graphs in which there are many edges (way more edges than vertices) while Kruskal performs better in sparse graphs because it uses simpler data structures

Prim's algorithm grows one tree all the time

Kruskal's algorithm grows multiple trees (i.e., a forest) at the same time

Trees are merged together