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Example

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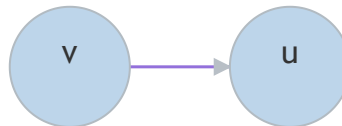
All Pair-Shortest Path

Graph theory

Glossary

- **Graph G (V , E):**
 - **V :** vertices set
 - **E :** edges set
- **Degree:** number of vertices

- **Walk:** series of connected vertices
- **Path:** walk without repeated vertices
- **Closed walk:** walk where $v_0 = v_n$
- **Cycle:** closed walk without repeating vertices
- **Euler path:** visit each edge exactly once
- **Hamilton path:** visit each vertex exactly once
- **Directed graph:** edges are ordered pairs
- **Ancestor:** v , **Successor:** u in



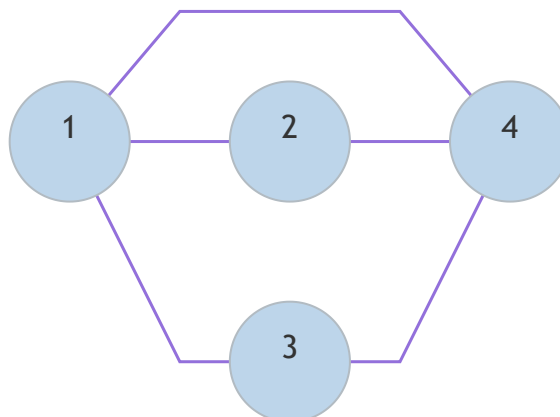
- $\text{deg}_{\text{in}}(v)$: number of incoming edges into v
- $\text{deg}_{\text{out}}(v)$: number of outgoing edges into v

Graph Representation

Adjacency matrix:

matrix where $A_{uv} = \begin{cases} 1 & \text{if } (u, v) \in E \\ 0 & \text{otherwise} \end{cases}$

Graph:



Matrix:

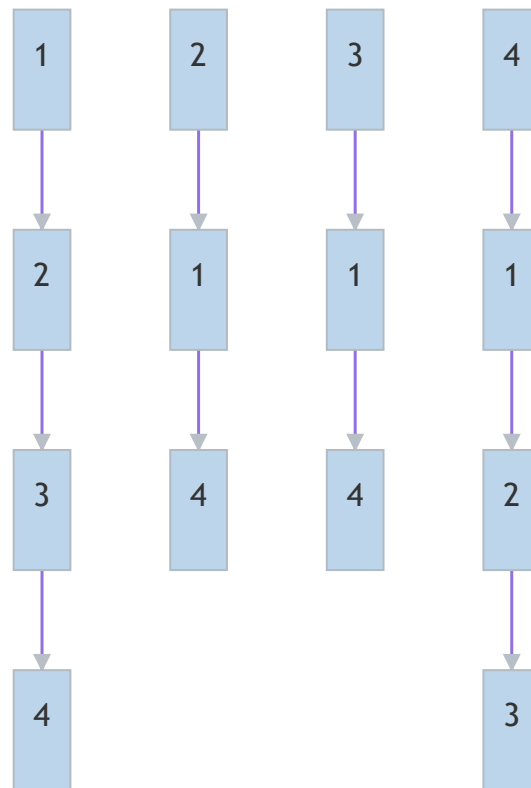
$$\begin{pmatrix} 0 & 1 & 1 & 1 \\ 1 & 0 & 0 & 1 \\ 1 & 0 & 0 & 1 \\ 1 & 1 & 1 & 0 \end{pmatrix}$$

Adjacency list

Array of linked lists, where $\text{Adj}[u]$ contains a list containing all the neighbors of u .

Graph: Same as above

List:



Runtimes

| | Matrix | List |
|--|--------------------------|--|
| Find all neighbors of v | $\mathcal{O}(n)$ | $\mathcal{O}(\text{deg}_{out}(v))$ |
| Find $v \in V$ without neighbors | $\mathcal{O}(n^2)$ | $\mathcal{O}(n)$ |
| Check if $(v, u) \in E$ | $\mathcal{O}(1)$ | $\mathcal{O}(1 + \text{deg}_{out}(v))$ |
| Insert edge | $\mathcal{O}(1)$ | $\mathcal{O}(1)$ |
| Remove edge (v, u) | $\mathcal{O}(1)$ | $\mathcal{O}(\text{deg}_{out}(v))$ |
| Check whether an Eulerian path exists or not | $\mathcal{O}(V * E)$ | $\mathcal{O}(V + E)$ |

Algorithms

Depth-First Search (DFS)

Used mainly to check whether a Graph can be topological sorted or not (\Leftrightarrow has a cycle). A

topological sorting of a graph it's a sequence of all its nodes with the property that a node u comes after a node v **if and only if** either a walk from v to u exists or u cannot be reached starting from v .

Pseudocode

```
1 DFS(G):  
2   t = 1  
3   for (v in V not marked):  
4     DFS-Visit(v)
```

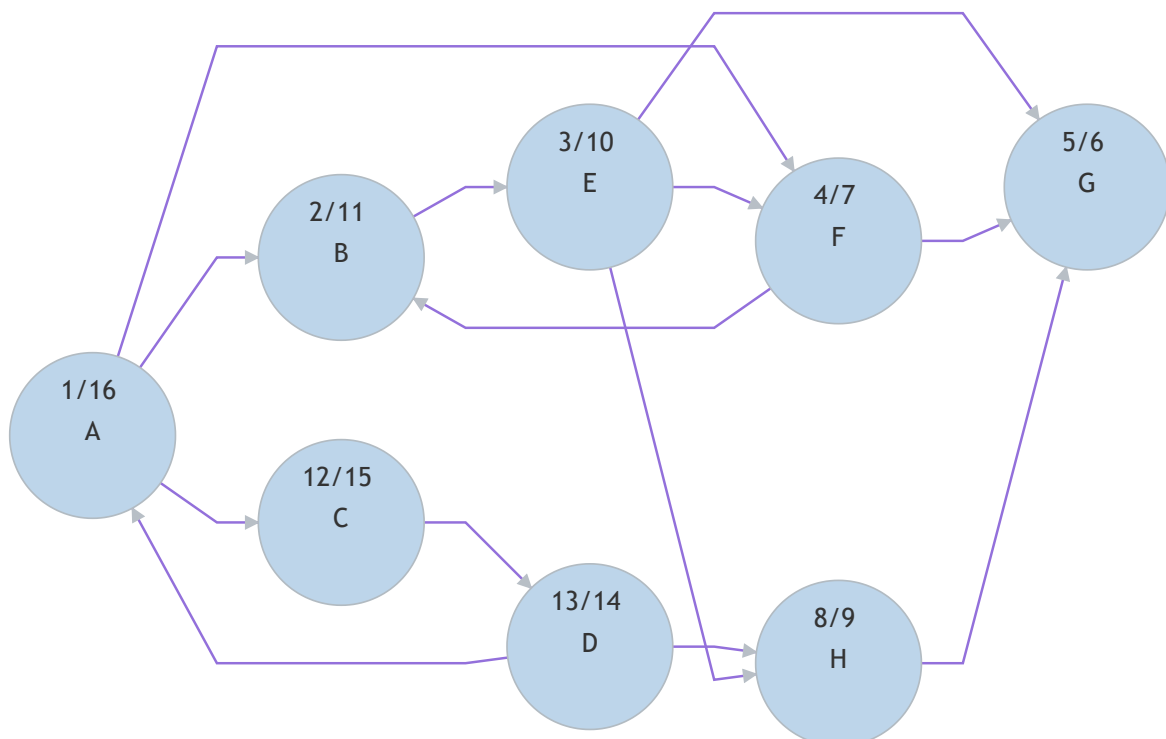
```
1 DFS-Visit(v):  
2   pre[v] = t++  
3   marked[v] = true  
4   for ((v, u) in E, u not marked)  
5     DFS_Visit(u)  
6   post[v] = t++
```

Runtime

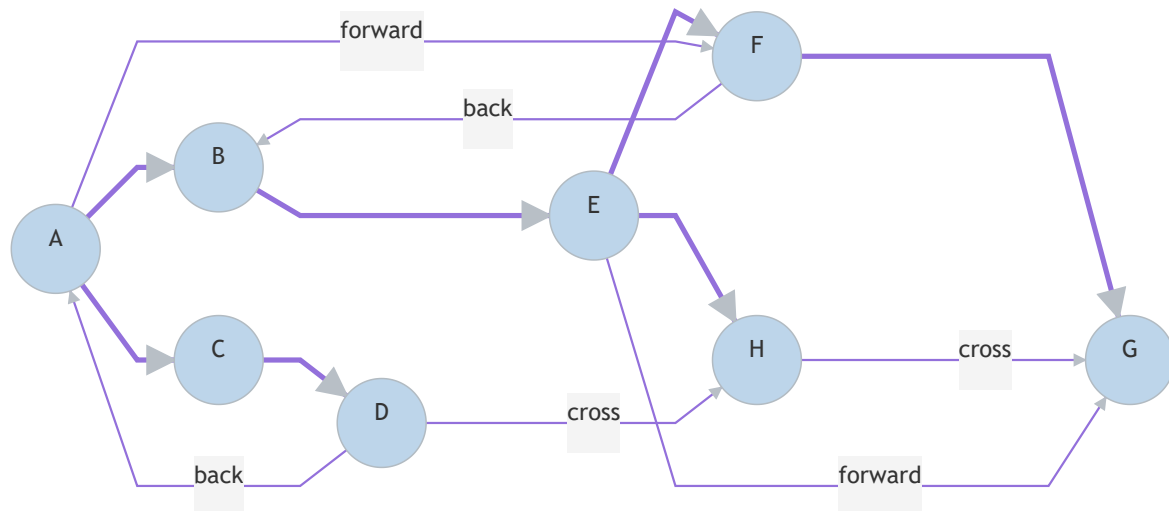
| | |
|------------|------------------------------|
| Operations | $T(n) \in \Theta(E + V)$ |
| Memory | $T(n) \in \Theta(V)$ |

Edge classification (post and pre numbers)

Example: DFS(A) got called



This graph generate the following tree (rotated of 90 degree to save space):



| Pre and post number | Name of the edge $(v, u) \in E$ |
|---|---------------------------------|
| $pre(u) < pre(v)$ and $post(u) < post(v)$ | Not possible |
| $pre(u) < pre(v)$ and $post(u) > post(v)$ | Tree edge |
| $pre(u) < pre(v)$ and $post(u) < post(v)$ but $(u, v) \notin E$ | Forward edge |
| $pre(u) > pre(v)$ and $post(u) > post(v)$ | Back edge |
| $pre(u) > pre(v)$ and $post(u) < post(v)$ | Cross edge |
| $pre(u) < pre(v)$ and $post(u) < post(v)$ | Not possible |

Remark: \exists back edge $\Leftrightarrow \exists$ closed walk (cycle)

Breadth-First Search (BFS)

Instead of searching through the depth of a graph, one can also go first through all the successor of the root with the BFS algorithm.

Pseudocode

```

1  BFS(G):
2      for (v in V not marked):
3          BFS-Visit(v)

```

```

1 BFS-Visit(v):
2     Q = new Queue()
3     active[v] = true //used to check whether a vertex is in the queue or not
4     enqueue(v, Q)
5     while (!isEmpty(Q)):
6         w = dequeue(Q)
7         visited[w] = true
8         for ((w, x) in E):
9             if(!active[x] && !visited[x]):
10                 active[x] = true
11                 enqueue(x, Q)

```

Runtime

| | |
|------------|------------------------------|
| Operations | $T(n) \in \Theta(E + V)$ |
| Memory | $T(n) \in \Theta(V)$ |

Find shortest path in DAG (Directed Acyclic Graph)

We can compute a recurrence following the topological sorting of the graph.

Pseudocode

```

1 ShortestPath(G, s):
2     d[s] = 0, d[v] = inf
3     for (v in V \ {s}, following topological sorting):
4         for ((u, v) in E):
5             d[v] = min(d[v], d[u] + c(u,v))

```

Runtime

$T(n) \in \mathcal{O}(|E| * |V|)$ if adjacency list is given

Dijkstra

Used to find the shortest (cheapest) path between two nodes in a graph.

Remark: The graph must **not** have negative weights

Pseudocode

```

1 Dijkstra(G, s):
2     for (v in V):
3         d[v] = infinity
4         parent[v] = null
5         insert(Q, v, d[v])
6     d[s] = 0
7     Q = new Queue()
8     decreaseKey(Q, s, 0) // decrease the priority of s to 0 (min)
9     while(!Q.isEmpty()):
10         v* = Q.extractMin() // extract from Q the node with minimum priority
11         for ((v*, v) in E):
12             dist = d[v*] + w(v*, v)
13             if (dist < d[v]):
14                 d[v] = dist

```

```

15     parent[v] = v*
16     decreaseKey(Q, v, d[v])

```

Runtime

If implemented with a Heap: $T(n) \in \mathcal{O}((|E| + |V|) * \log(|V|))$

If implemented with a **Fibonacci-Heap**: $T(n) \in \mathcal{O}((|E| + |V| * \log(|V|)))$

Bellman-Ford

Used for graph with general weight (**positive and negative!**)

Pseudocode

```

1  BellmanFord(G, s):
2      for (v in V):
3          distance[v] = infinity
4          parent[v] = null
5      distance[s] = 0
6      for (i = 1, 2, ..., |V| - 1):
7          for ((u, v) in E):
8              if (distance[v] > distance[u] + w(u, v)):
9                  distance[v] = distance[u] + w(u, v)
10                 parent[v] = u
11      for ((u, v) in E):
12          if (distance[u] + w(u, v) < distance[v]):
13              return "negative cycle!"

```

Runtime

$T(n) \in \mathcal{O}(|E| * |V|)$

Boruvka

Used to find a MST in a given graph G

Minimum Spanning Trees (MSTs)

A minimum spanning tree is a subgraph $H = (V, E^*)$ of a graph $G = (V, E)$ with $E^* \subseteq E$, such that every vertex $v \in V$ is connected and that **the sum of all edges' weight is minimal**.

Pseudocode

```

1  Boruvka(G):
2      F = new Set() // Initialize a new forest with every vertex being a tree
                        and 0 edges
3      while (F not SpanningTree): // check that ZHKs of F > 1 or number of
                        edges < |V| - 1
4          ZHKs of F = (S1, ..., Sk)
5          minEdges of S1, ..., Sk = (e1, ..., ek)
6          F = F U (e1, ..., ek)
7      return F

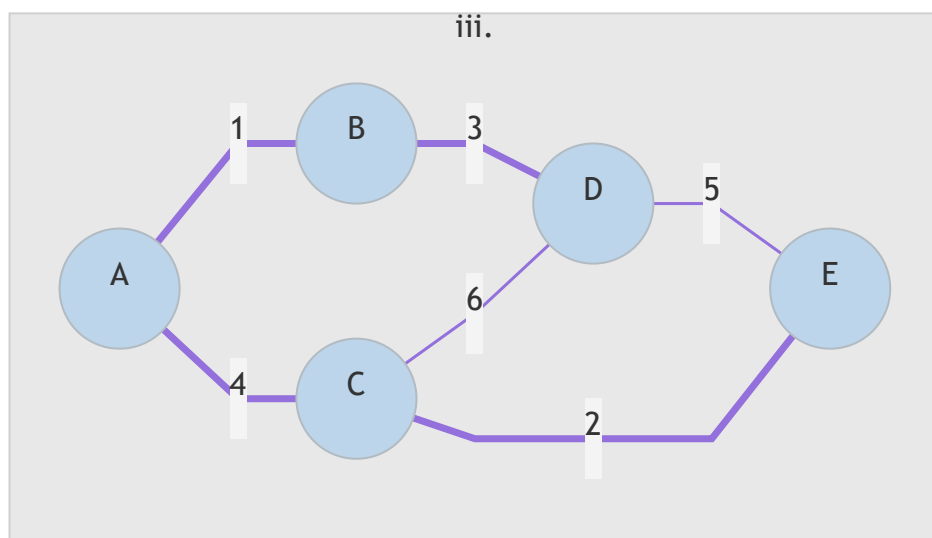
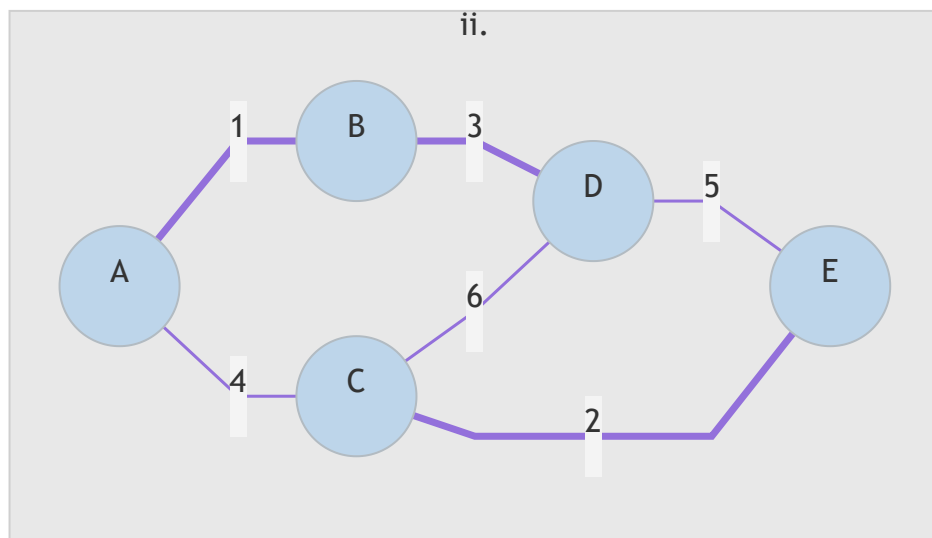
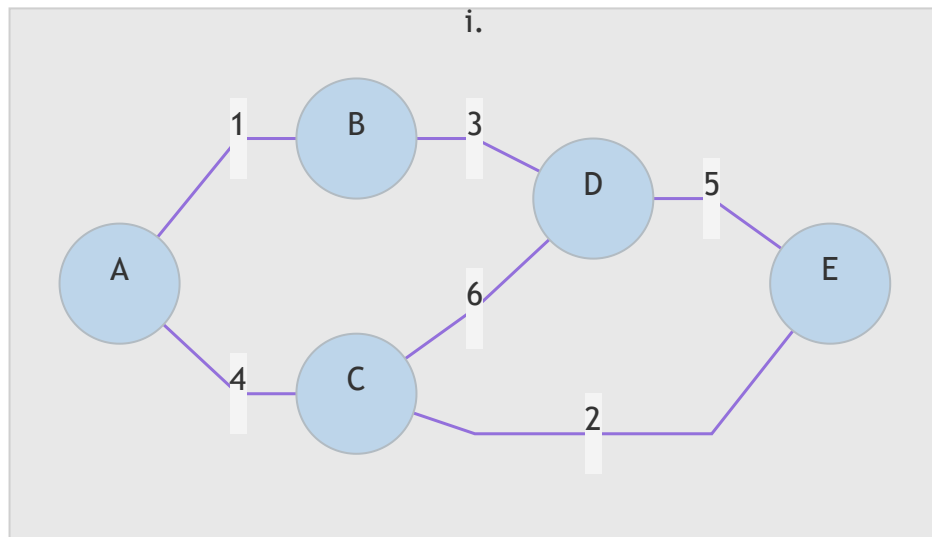
```

Runtime

$$T(n) \in \mathcal{O}((|E| + |V|) * \log(|V|))$$

Example

First choose the minimal edge for every vertex and add them to the new graph. Then repeat for every ZHK (vertices connected with edges) until you have a MST (until there is only 1 ZHK).



Prim

Alternative to Kruskal, it needs a starting vertex as input.

Pseudocode

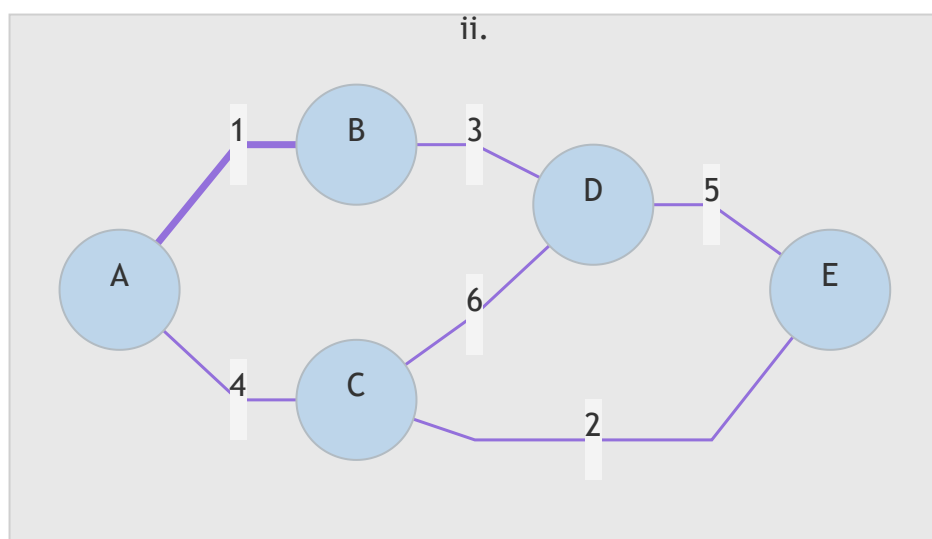
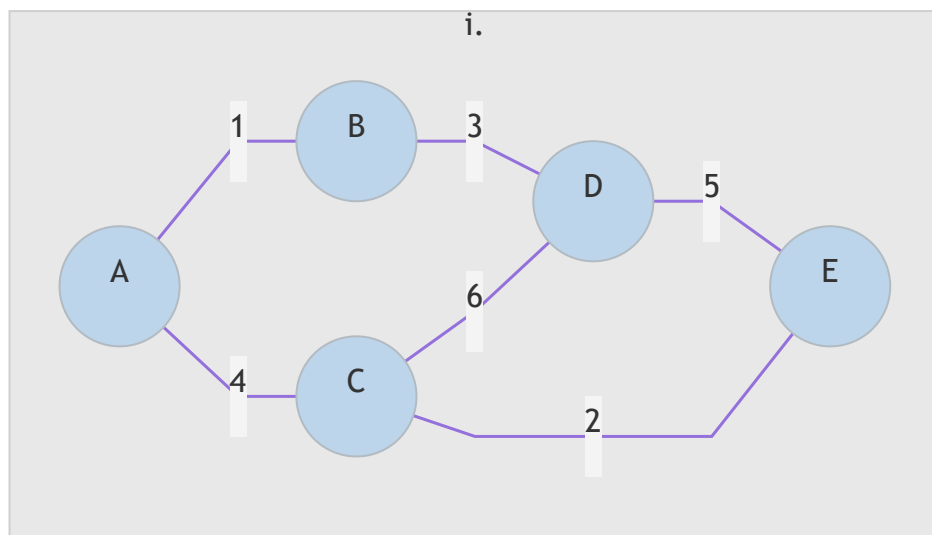
```
1 Prim(G, s):  
2   MST = new Set()  
3   H = new minHeap(V, infinity)  
4   decreaseKey(H, s, 0)  
5   while (!H.isEmpty()):  
6     v = extractMin(H)  
7     MST.add(v)  
8     for ((v, u) in E && u not in MST)  
9       decreaseKey(H, u, w(v, u))
```

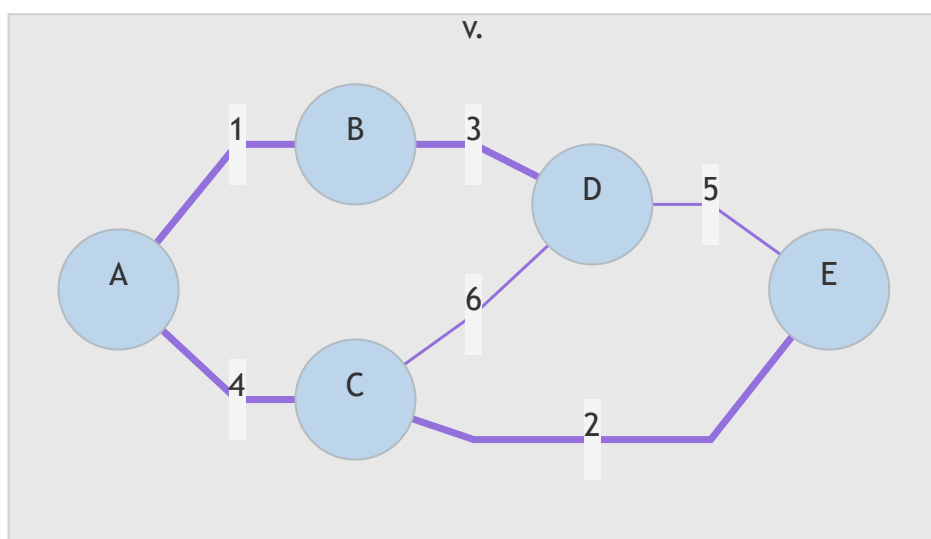
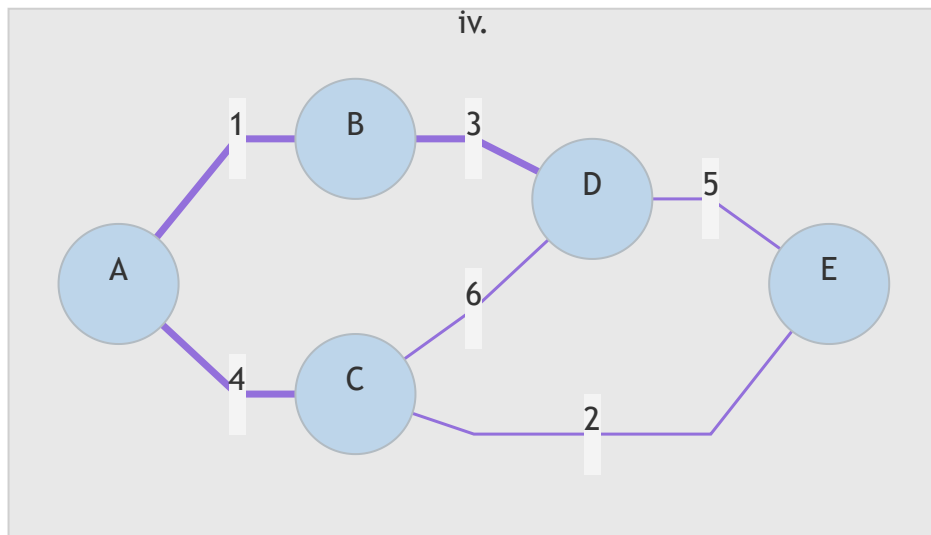
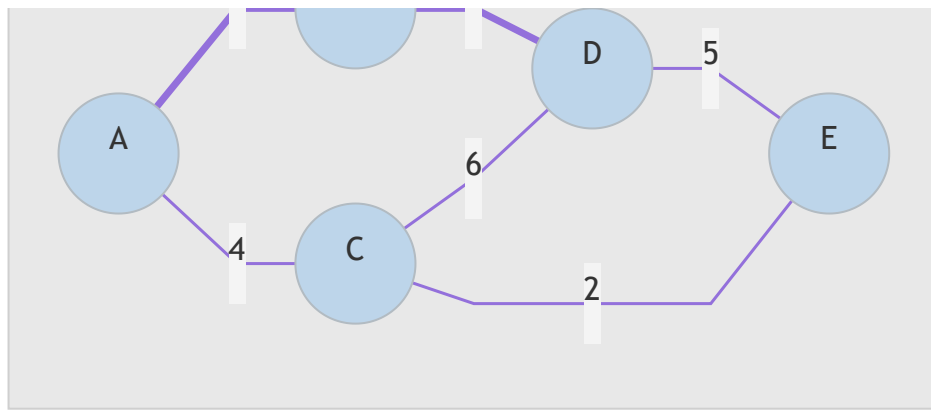
Runtime

$$T(n) \in \mathcal{O}((|E| + |V|) * \log(|V|))$$

Example

Add the minimal edge adjacent to s. Then take the newly created ZHK and add to it its minimal outgoing edge. Proceed like that until you have a spanning tree (all the vertices are connected).





Kruskal

Another algorithm to find a MST in a given graph. It sorts edges by weight and adds them one by one, **unless adding an edge would form a cycle.**

Pseudocode

```

1 kruskal(G):
2   MST = new Set()
3   E.sort() // sort all edges by weight
4   for ((u, v) in E):
5       if (u and v in 2 different ZHKs of MST):
6           MST.add((u, v))

```

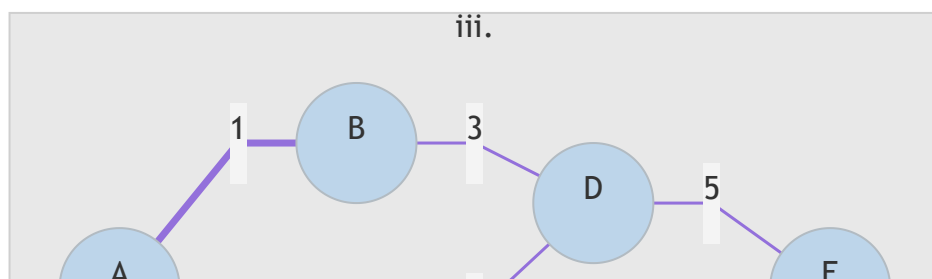
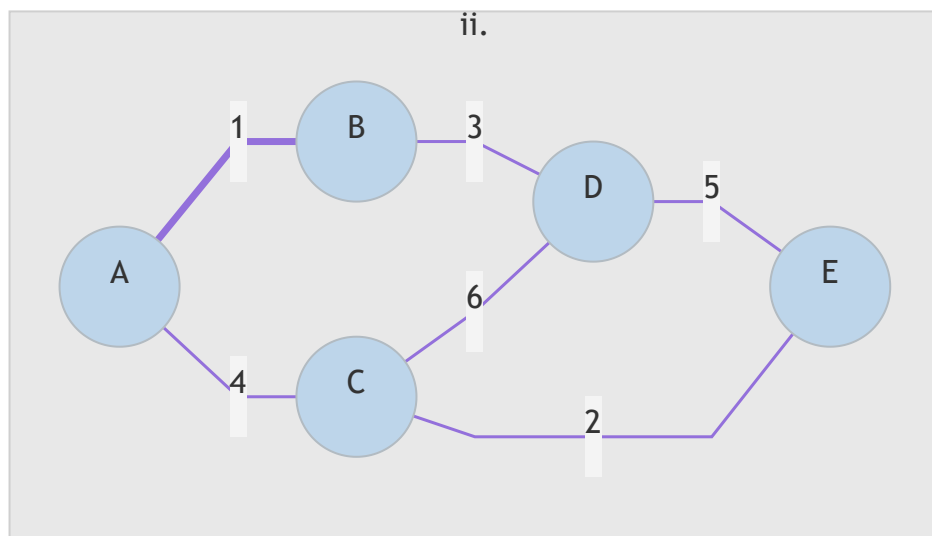
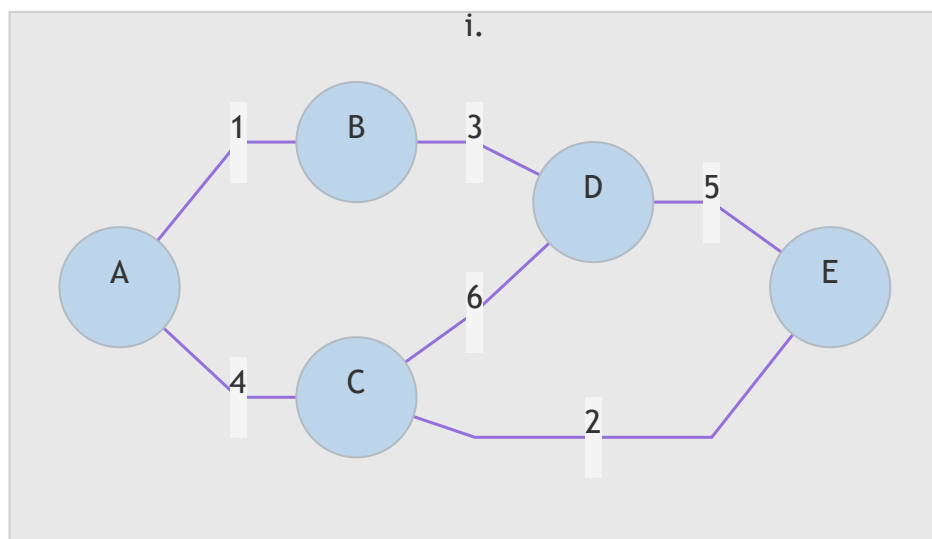
Runtime

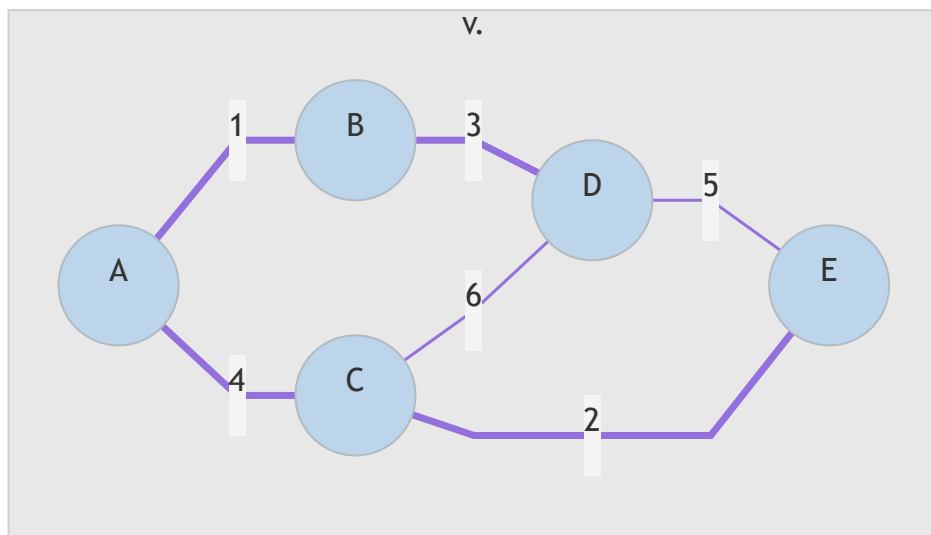
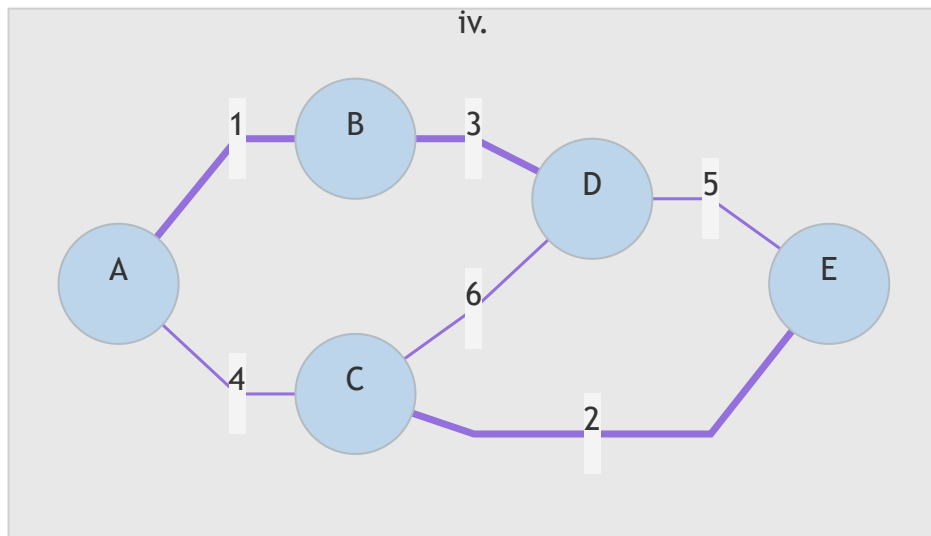
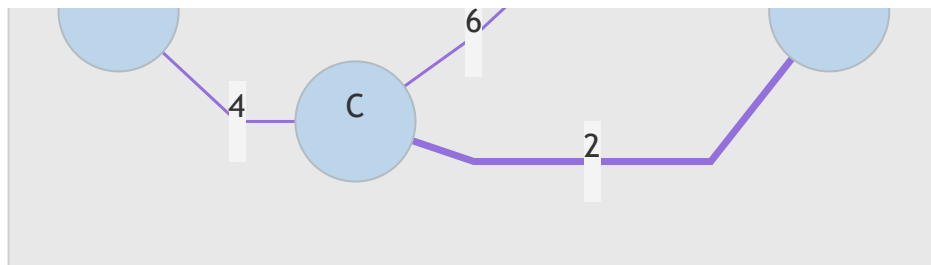
If implemented normally: $T(n) \in \mathcal{O}(|E| * |V| + |E| * \log(|E|))$ (second part to sort)

If implemented with an improved union-find DS: $T(n) \in \mathcal{O}(|V| * \log(|V|) + |E| * \log(|E|))$ (second part to sort)

Example

Add edges one by one following weight-order. If adding an edge would form a cycle, skip it.





Floyd-Warshall

Used to solve the **all-pair shortest path** problem, i.e., to find the shortest distance between **any** two vertices of a given graph G .

It makes use of a 3-Dimensional DP table.

Pseudocode

$d[i][u][v]$ represents the shortest path from u to v passing through $\leq i$ vertices.

```

1 FloydWarshall(G):
2   for (v in V):
3       d[0][v][v] = 0 // layer 0, row v, column v
4   for ((v, u) in E):
5       d[0][v][u] = w(v, u)
6   else: // if u, v isn't in E
7       d[0][v][u] = infinity
8   for (i = 1, ..., |V|):
9       for (u = 1, ..., |V|):
10          for (v = 1, ..., |V|):
11              d[i][u][v] = min(d[i-1][u][v], d[i-1][u][i] + d[i-1][i][v])
12   return d

```

Remarks:

- This algorithm can be implemented **inplace**, it just suffice to leave the indices away.
- The algorithm does **not** work if negative cycles are present.

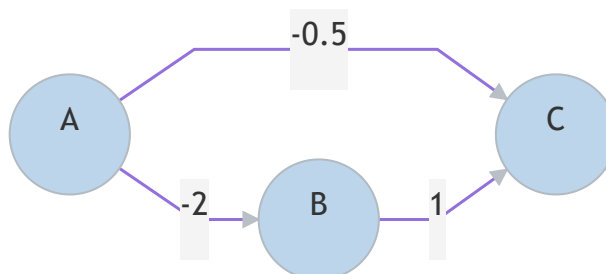
Runtime

$$T(n) \in \mathcal{O}(|V|^3)$$

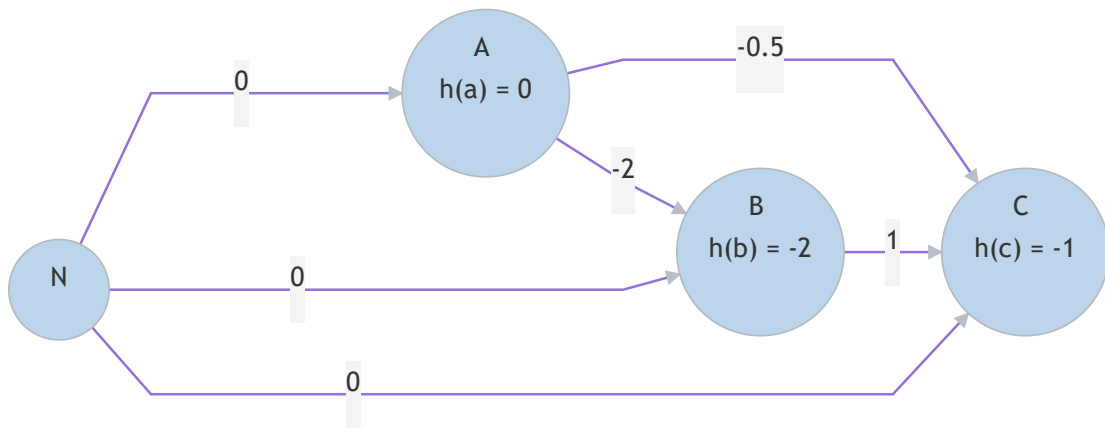
Johnson

Used to solve the all-pair shortest path problem. First one has to make every weight positive, by adding an "external" vertex, and then proceed by using Dijkstra $|V|$ times.

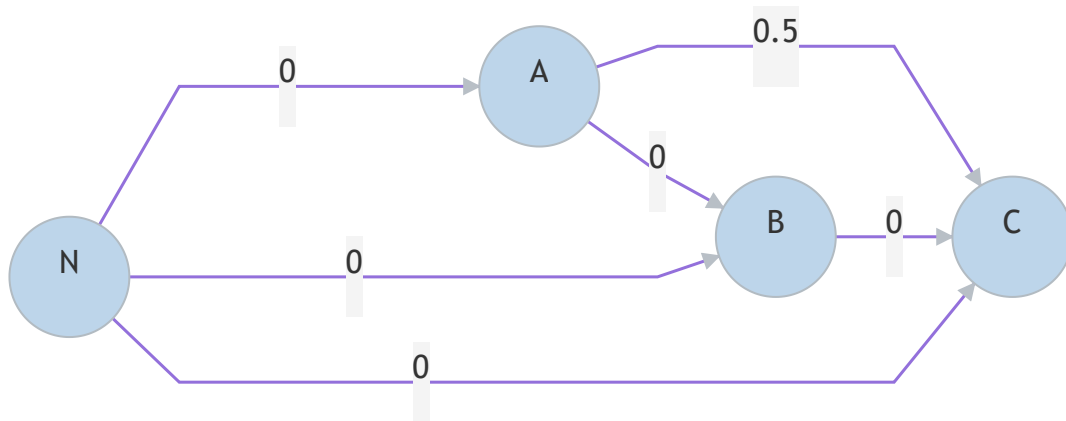
Example



- First, add the new vertex, and connect it to every other vertex with weight 0.
 $h(n)$ is the "height" of the node n , equals to **the shortest path from N to n** , found by applying n times Dijkstra.



- We now can modify each weight $w(u, v)$ of each edge into a new weight $w^*(u, v) = w(u, v) + (h(u) - h(v))$



Runtime

- Create new node and add new edges: $\mathcal{O}(|V|)$
- Assign h-values: Bellman-Ford, $\mathcal{O}(|V| * |E|)$
- $|V|$ times Dijkstra: $\mathcal{O}(|V| * |E| + |V|^2 * \log(|V|))$

All Pair-Shortest Path

All the algorithms we know to solve the APSP problem can be compared in the following way (**top**: less general, **bottom**: more general):

| Graph | Algorithm | Runtime |
|---|---|--|
| $G = (V, E)$ | $ V * BFS$ | $\mathcal{O}(V * E + V ^2)$ |
| $G = (V, E, w)$ $w : E \rightarrow \mathbb{R}^+$ | $ V * Dijkstra$ | $\mathcal{O}(V * E + V ^2 * \log(V))$ |
| $G = (V, E, w)$ $w : E \rightarrow \mathbb{R}$ | $ V * Bellman - Ford$ $Floyd - Warshall$ $Johnson$ | $\mathcal{O}(V ^2 * E)$ $\mathcal{O}(V ^3)$ $\mathcal{O}(V * E + V ^2 * \log(V))$ |