

# Paths in Graphs: Fastest Route

Michael Levin

Higher School of Economics

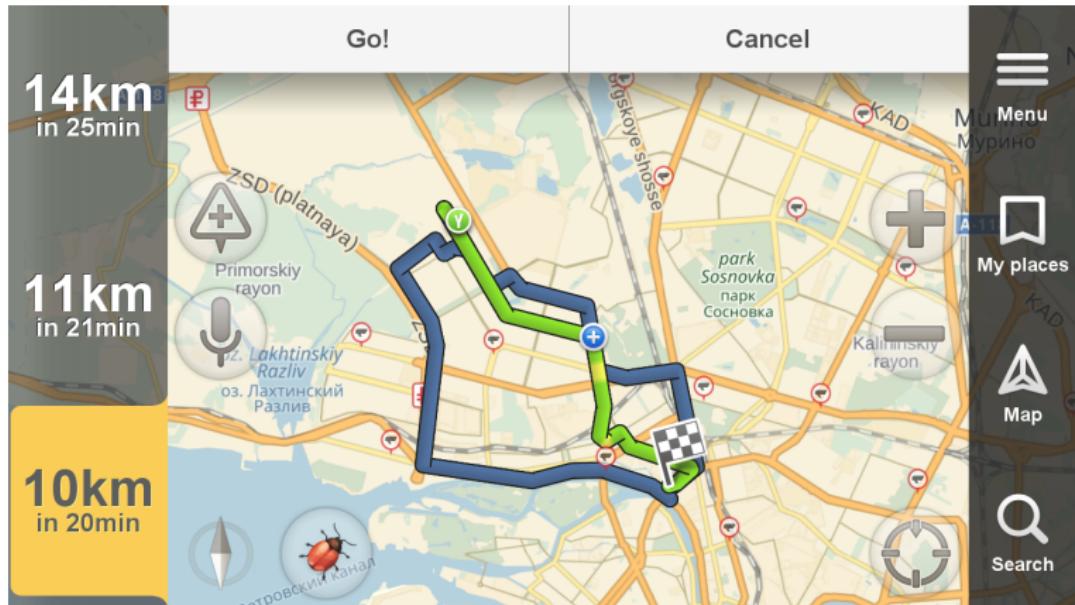
Graph Algorithms  
Data Structures and Algorithms

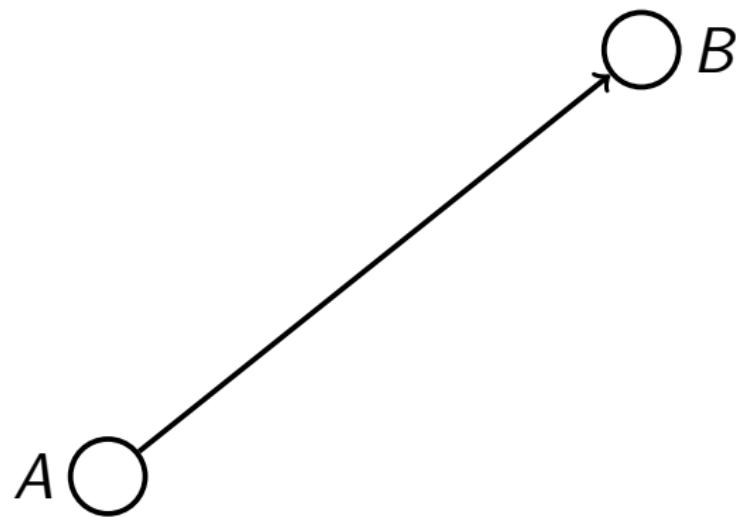
# Outline

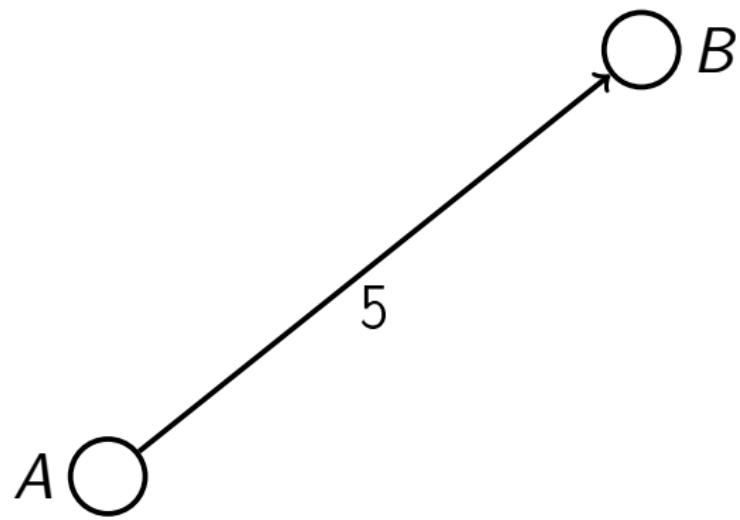
- ① Fastest Route
- ② Naive Algorithm
- ③ Dijkstra's Algorithm

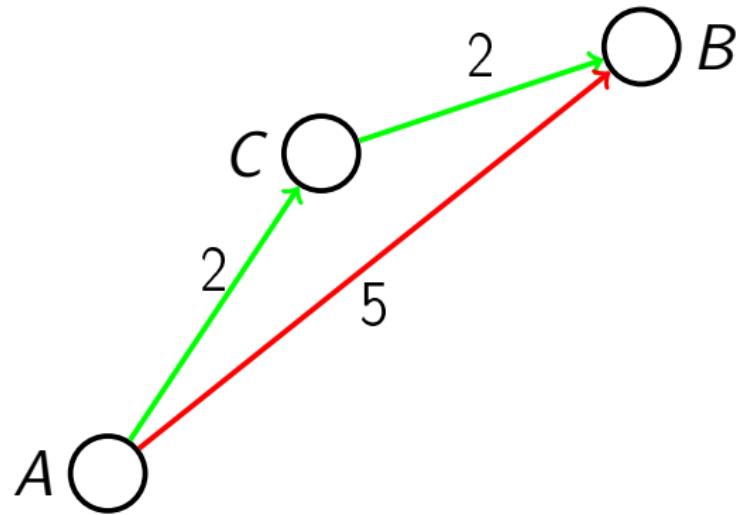
# Fastest Route

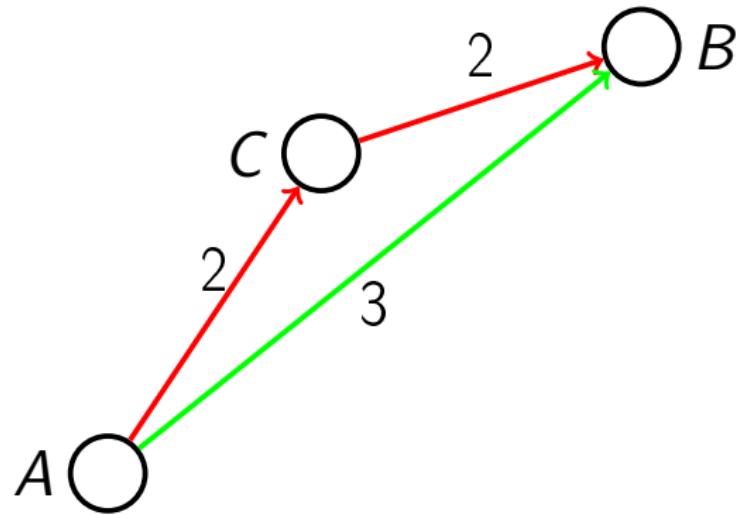
What is the fastest route to get home from work?





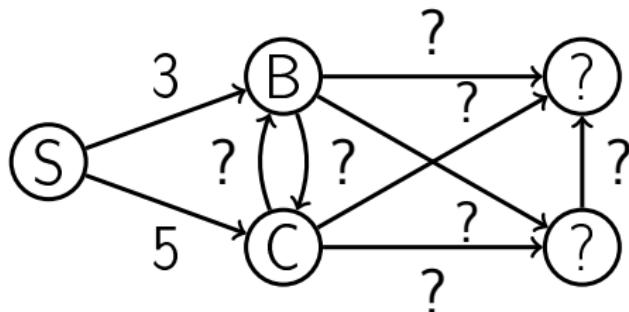






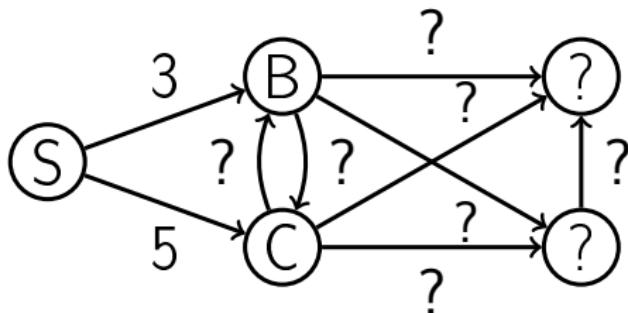
# Intuition

- Assume that we stay at  $S$  and observe two outgoing edges:



# Intuition

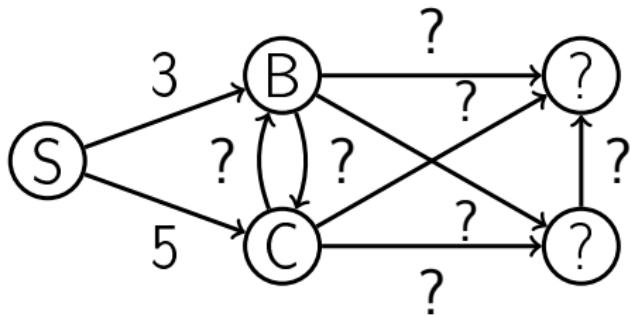
- Assume that we stay at  $S$  and observe two outgoing edges:



- Can we be sure that the distance from  $S$  to  $C$  is 5?

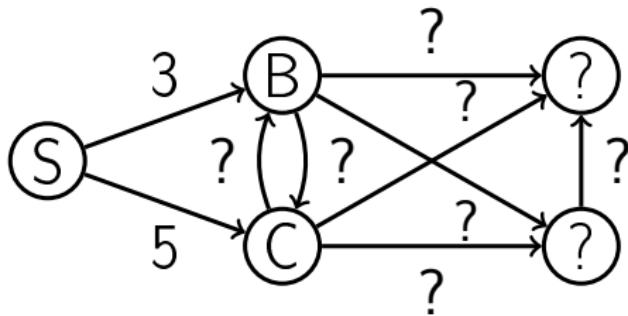
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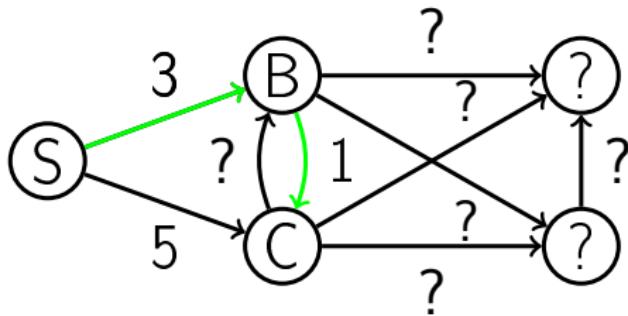
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- No, because the weight of the edge  $(B, C)$  might be equal to, say, 1.

# Intuition

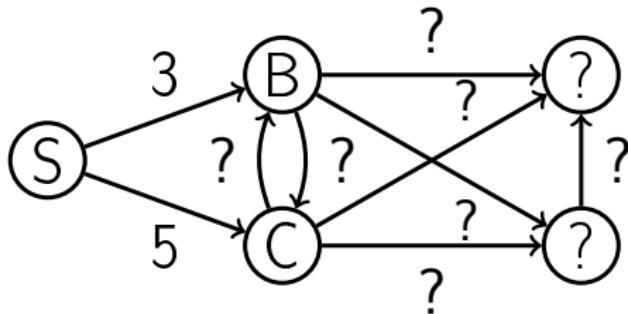
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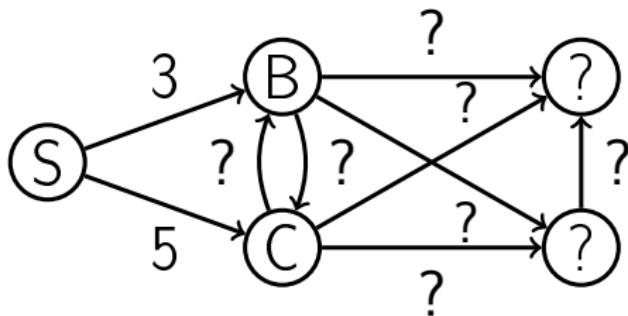
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- Can we be sure that the distance from  $S$  to  $B$  is 3?



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- Yes, because there are no negative weight edges.

# Outline

- ① Fastest Route
- ② Naive Algorithm
- ③ Dijkstra's Algorithm

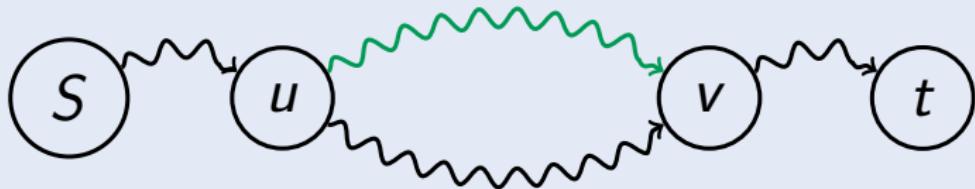
# Optimal substructure

## Observation

Any subpath of an optimal path is also optimal.

## Proof

Consider an optimal path from  $S$  to  $t$  and two vertices  $u$  and  $v$  on this path. If there were a shorter path from  $u$  to  $v$  we would get a shorter path from  $S$  to  $t$ .



# Corollary

If  $S \rightarrow \dots \rightarrow u \rightarrow t$  is a shortest path from  $S$  to  $t$ , then

$$d(S, t) = d(S, u) + w(u, t)$$

# Edge relaxation

- $\text{dist}[v]$  will be an upper bound on the actual distance from  $S$  to  $v$ .

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- $\text{dist}[v]$  will be an upper bound on the actual distance from  $S$  to  $v$ .
- The edge relaxation procedure for an edge  $(u, v)$  just checks whether going from  $S$  to  $v$  through  $u$  improves the current value of  $\text{dist}[v]$ .

$\text{Relax}((u, v) \in E)$

```
if  $dist[v] > dist[u] + w(u, v)$ :  
     $dist[v] \leftarrow dist[u] + w(u, v)$   
     $prev[v] \leftarrow u$ 
```

# Naive approach

Naive( $G, S$ )

for all  $u \in V$ :

$dist[u] \leftarrow \infty$

$prev[u] \leftarrow nil$

$dist[S] \leftarrow 0$

do:

    relax all the edges

while at least one  $dist$  changes

# Correct distances

## Lemma

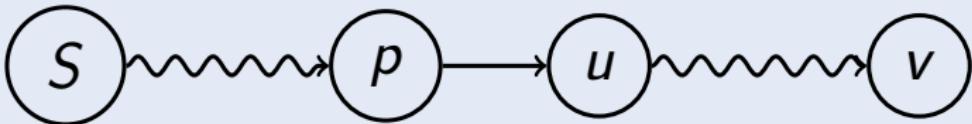
After the call to Naive algorithm all the distances are set correctly.

## Proof

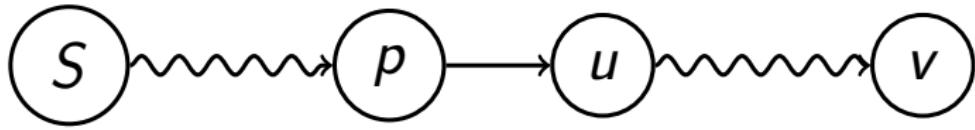
- Assume, for the sake of contradiction, that no edge can be relaxed and there is a vertex  $v$  such that  $\text{dist}[v] > d(S, v)$ .

## Proof

- Assume, for the sake of contradiction, that no edge can be relaxed and there is a vertex  $v$  such that  $\text{dist}[v] > d(S, v)$ .
- Consider a shortest path from  $S$  to  $v$  and let  $u$  be the first vertex on this path with the same property. Let  $p$  be the vertex right before  $u$ .

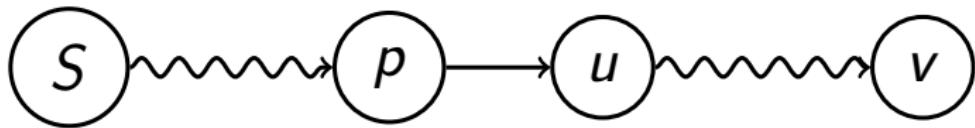


# Proof (continued)



- Then  $d(S, p) = \text{dist}[p]$  and hence  
 $d(S, u) = d(S, p) + w(p, u) =$   
 $\text{dist}[p] + w(p, u)$

# Proof (continued)



- Then  $d(S, p) = \text{dist}[p]$  and hence  
 $d(S, u) = d(S, p) + w(p, u) =$   
 $\text{dist}[p] + w(p, u)$
- $\text{dist}[u] > d(S, u) = \text{dist}[p] + w(p, u) \Rightarrow$   
edge  $(p, u)$  can be relaxed —  
a contradiction.



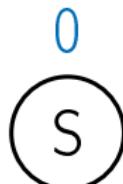
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- ① Fastest Route
- ② Naive Algorithm
- ③ Dijkstra's Algorithm

# Intuition

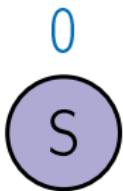
0  
S

# Intuition

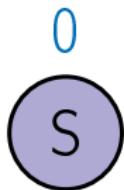


initially, we only know the distance to S

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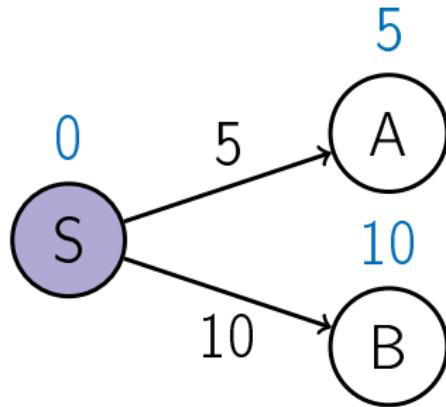


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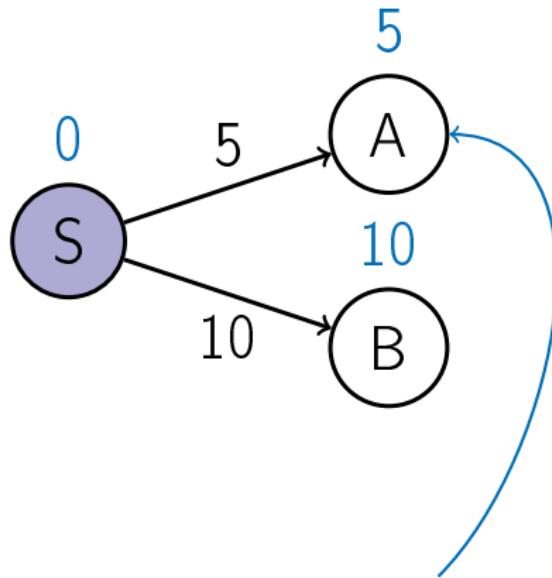
let's relax all the edges from S

# Intuition



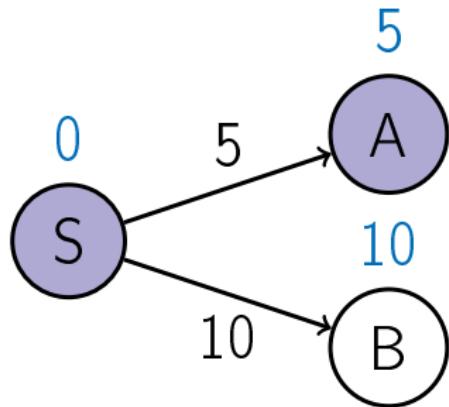
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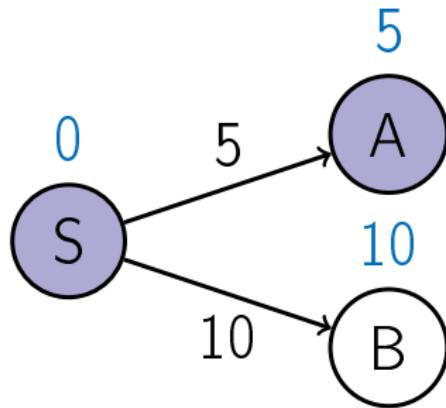


we now know the distance for A

# Intuition

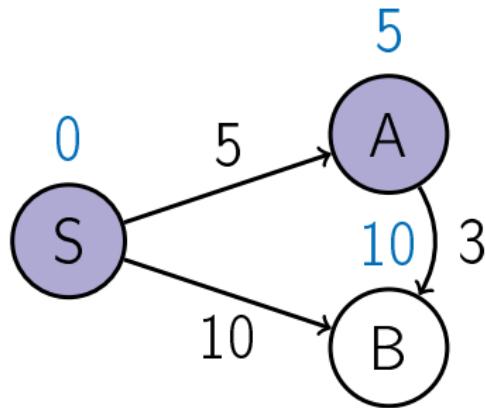


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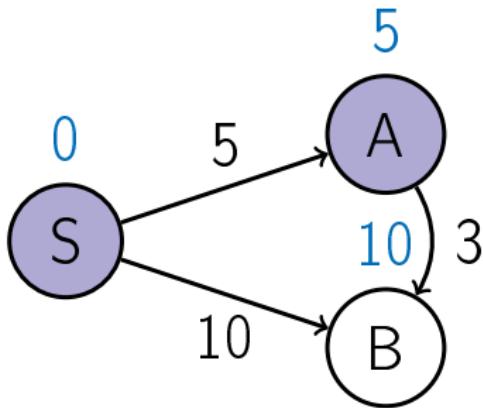
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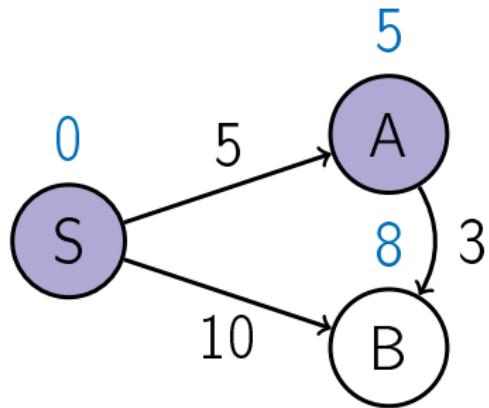
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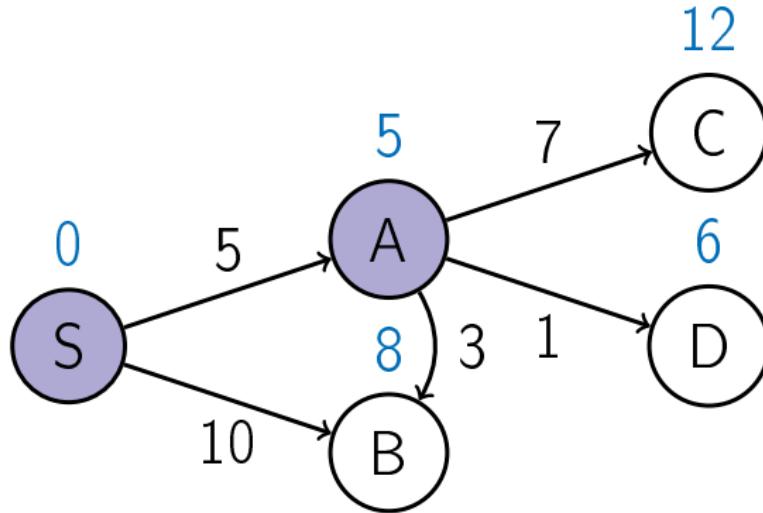


we discover an edge  $(A, B)$  of weight 3  
that updates  $\text{dist}[B]$

# Intuition

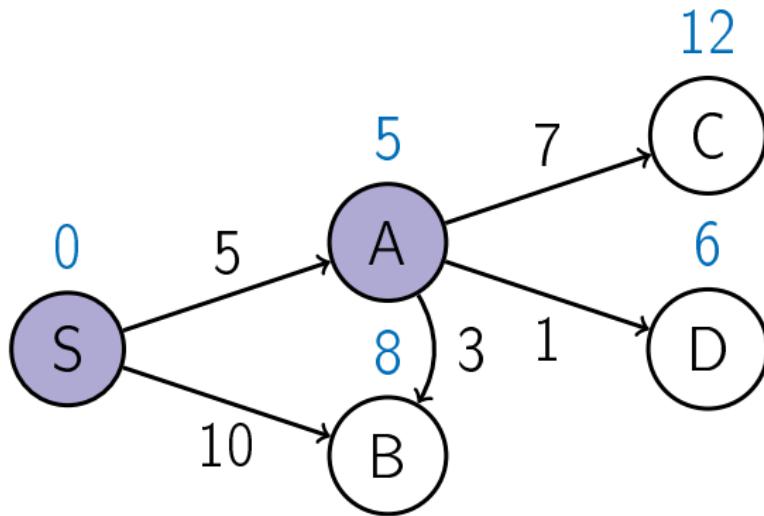


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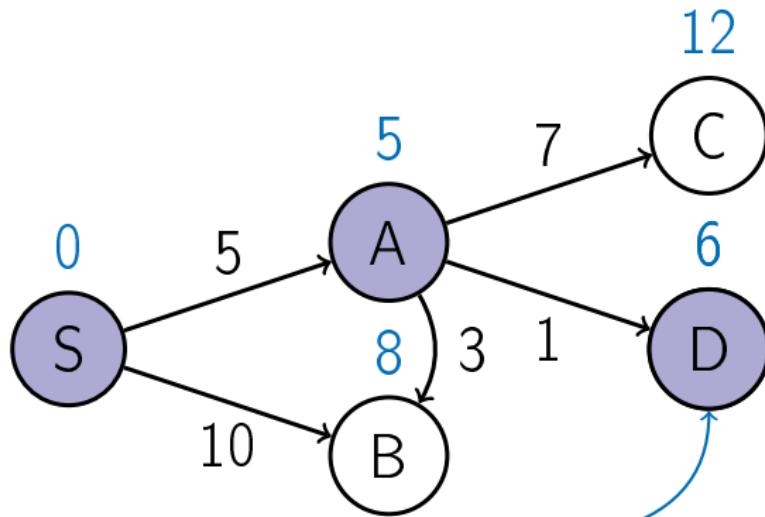
we also discover a few more outgoing edges

# Intuition



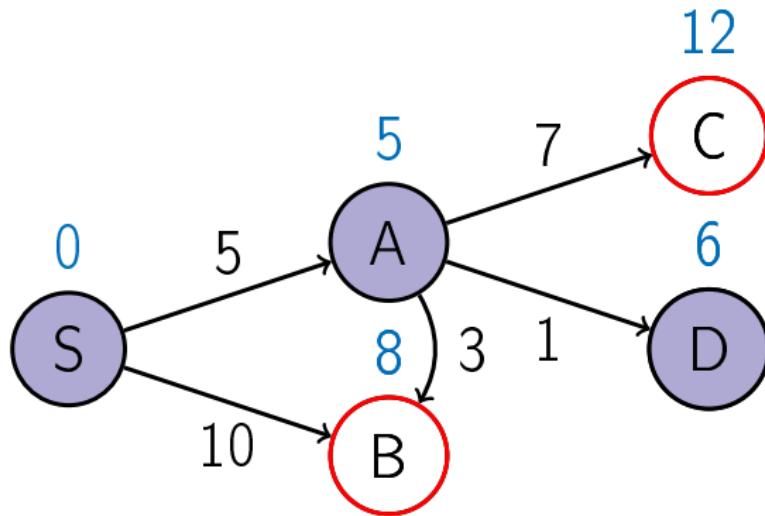
what is the next vertex for which we already know the correct distance?

# Intuition



it is *D*

# Intuition



while for  $B$  and  $C$  it is possible that their dist values are larger than actual distances

# Main ideas of Dijkstra's Algorithm

- We maintain a set  $R$  of vertices for which  $\text{dist}$  is already set correctly (“known region”).

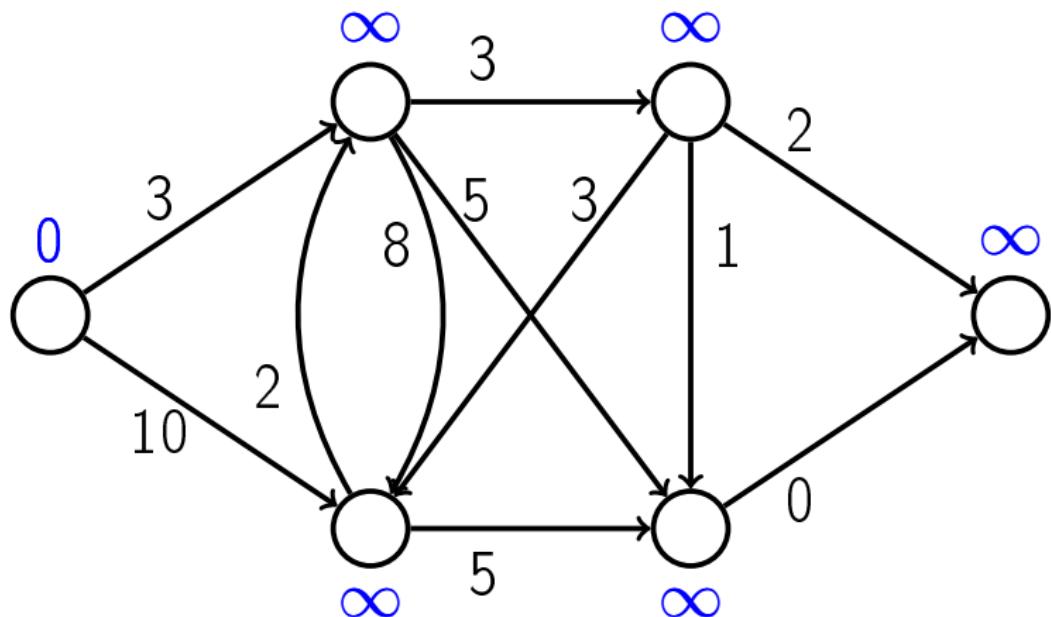
# Main ideas of Dijkstra's Algorithm

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- The first vertex added to  $R$  is  $S$ .

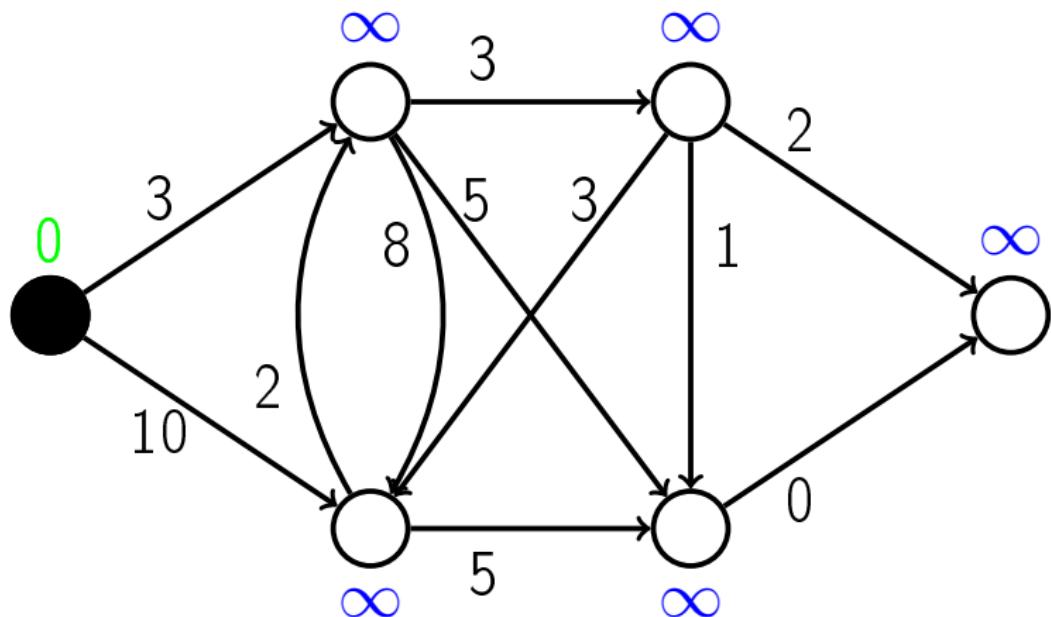
# Main ideas of Dijkstra's Algorithm

- We maintain a set  $R$  of vertices for which  $\text{dist}$  is already set correctly (“known region”).
- The first vertex added to  $R$  is  $S$ .
- On each iteration we take a vertex outside of  $R$  with the minimal  $\text{dist}$ -value, add it to  $R$ , and relax all its outgoing edges.

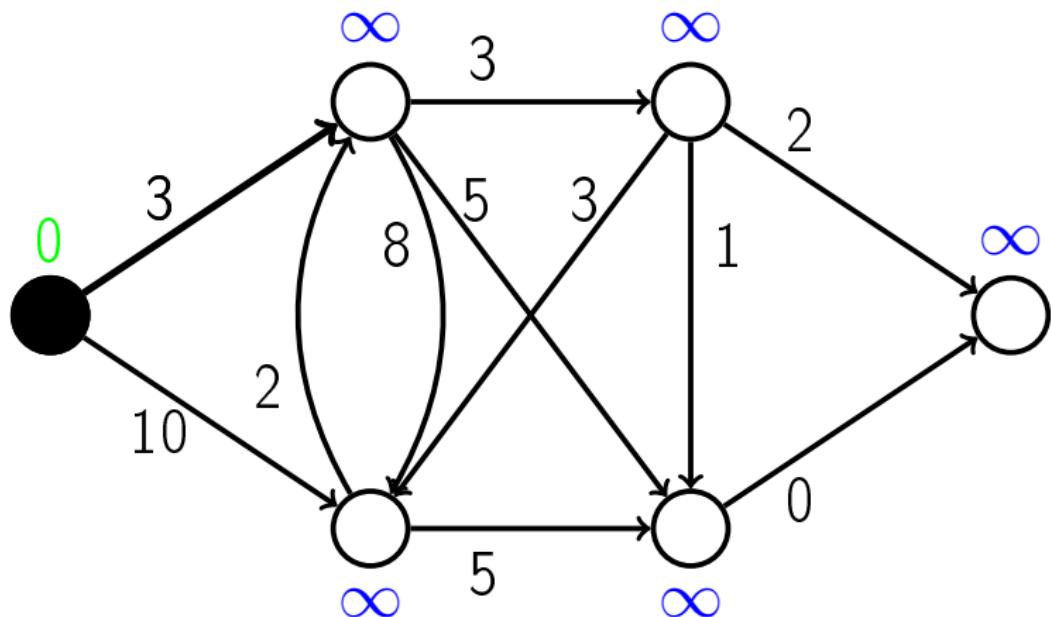
# Example



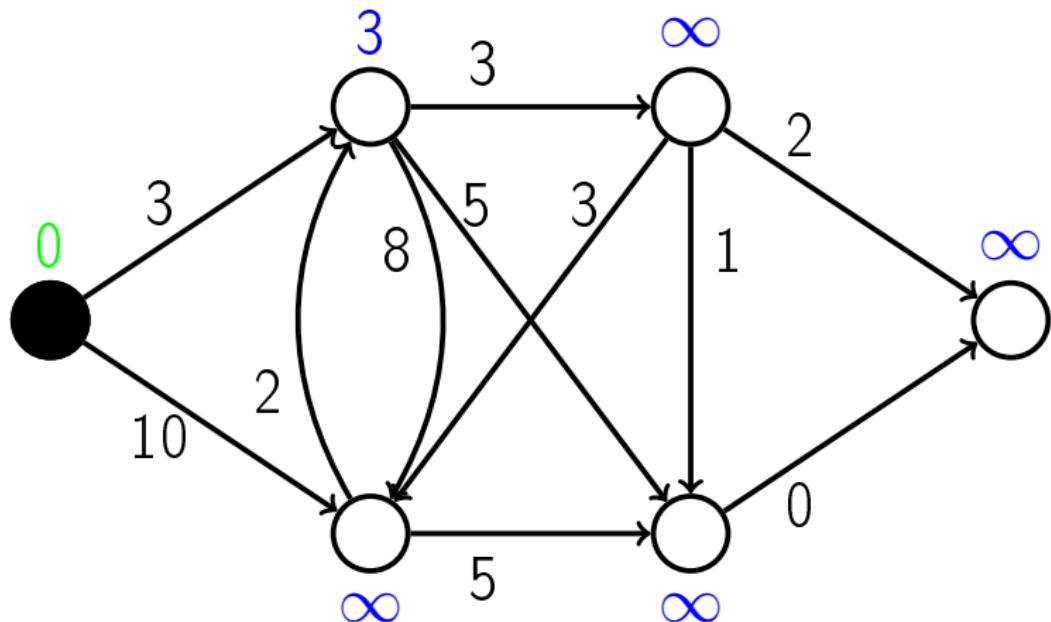
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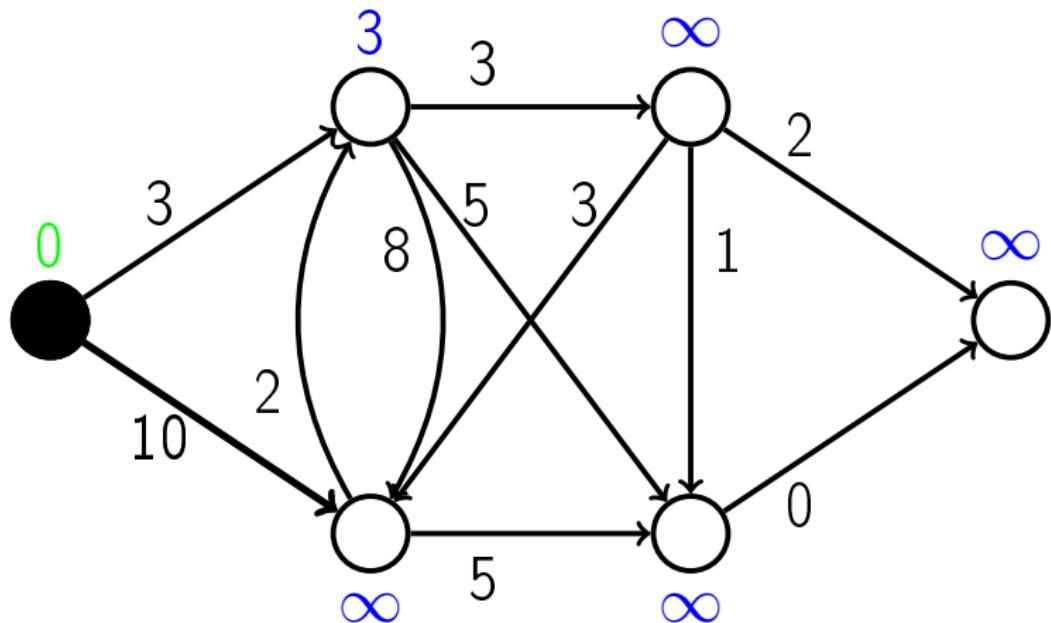
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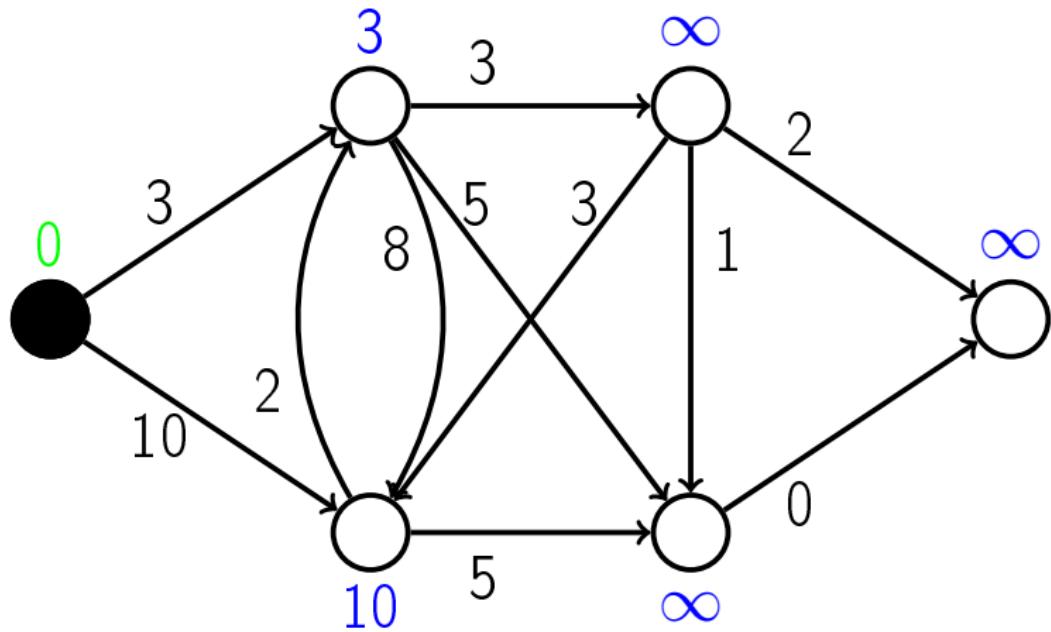
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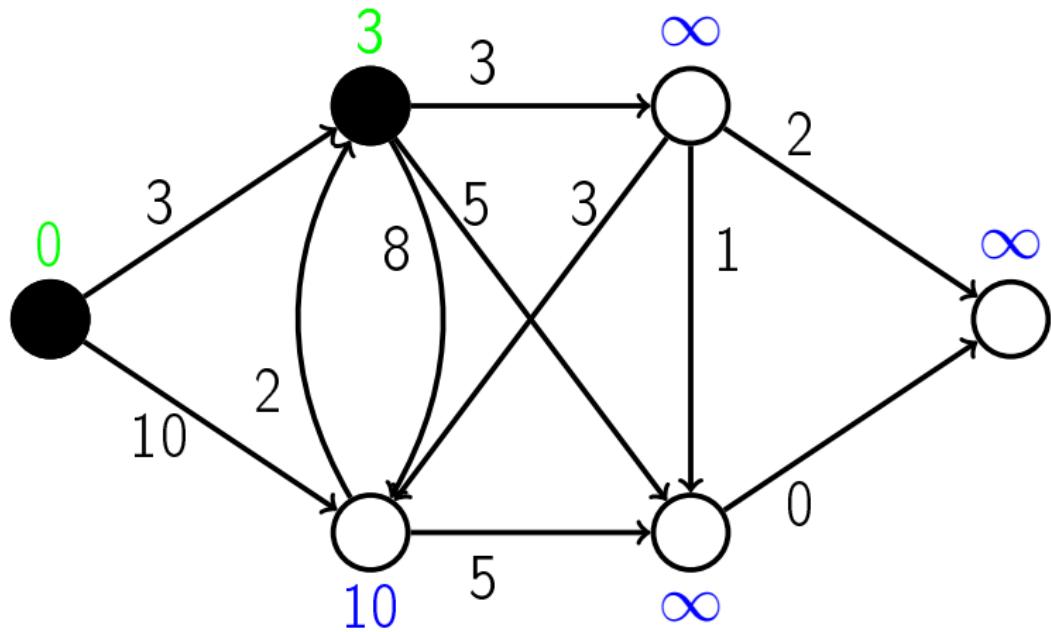
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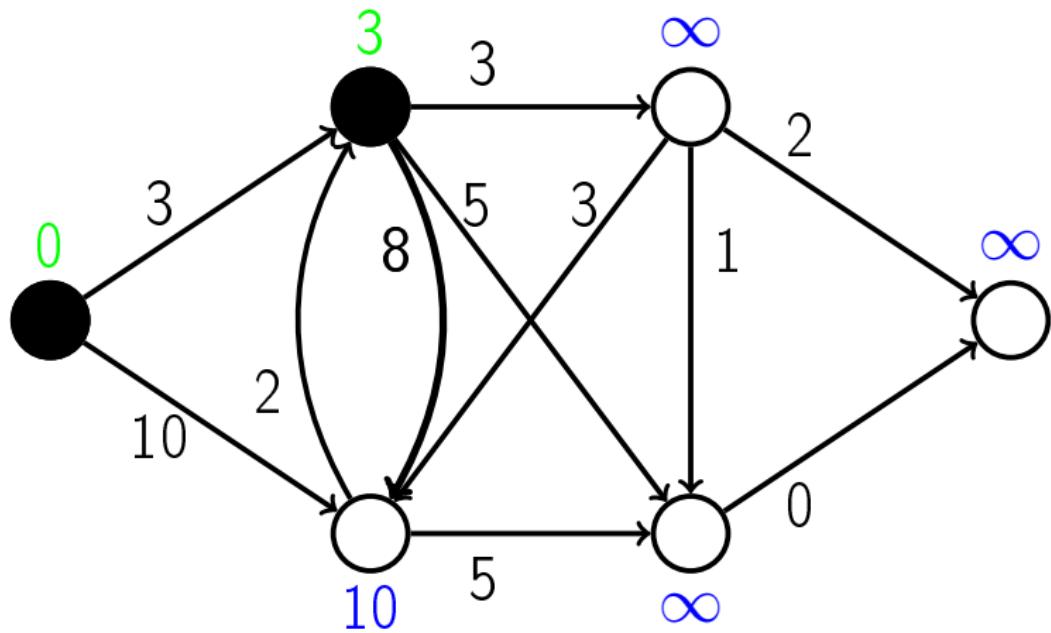
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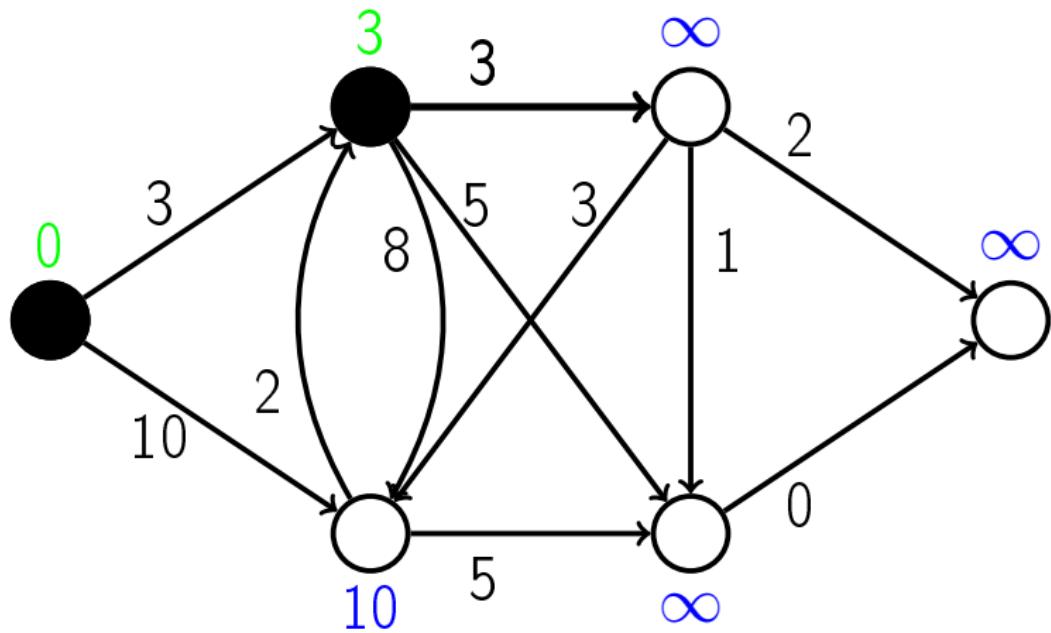
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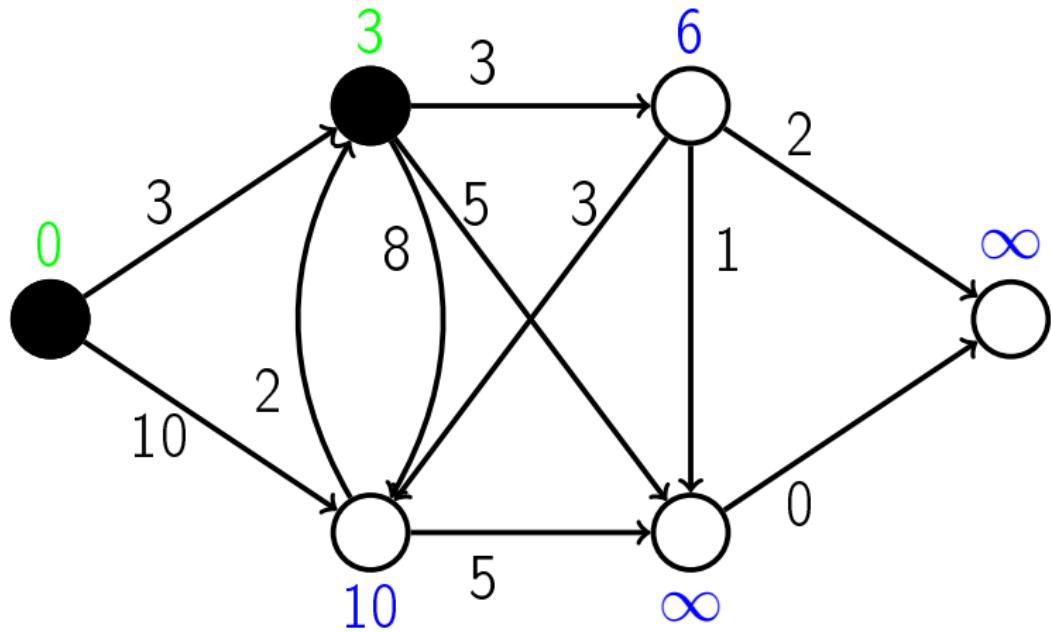
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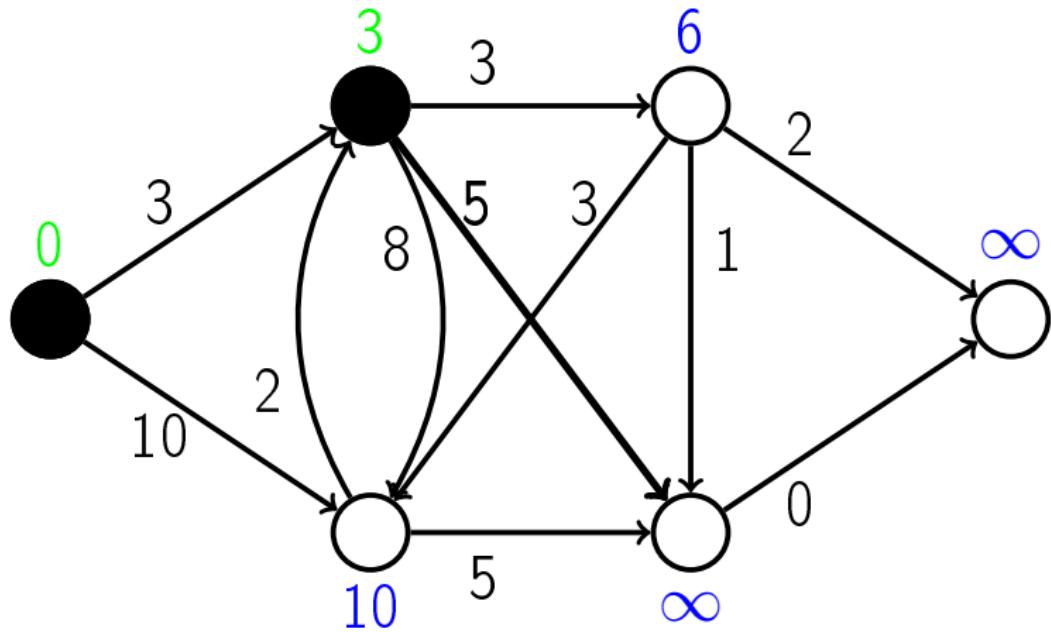
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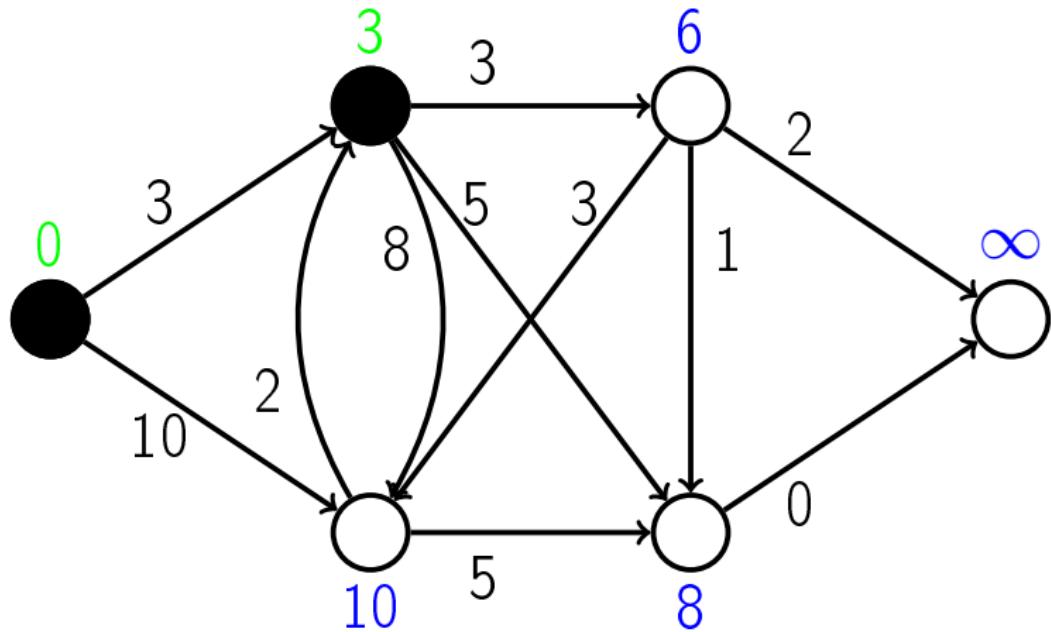
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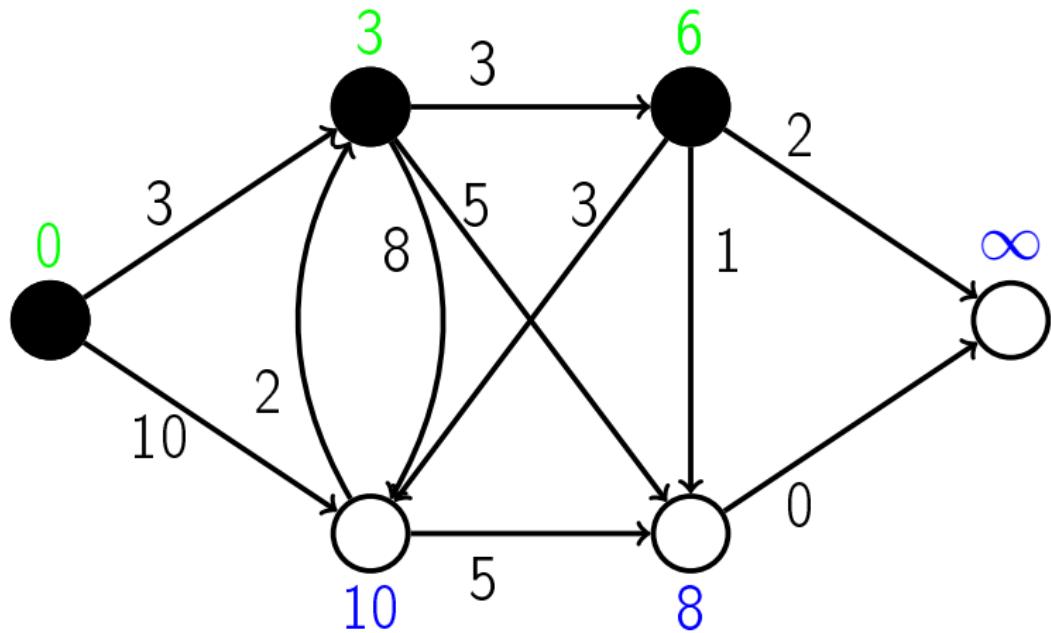
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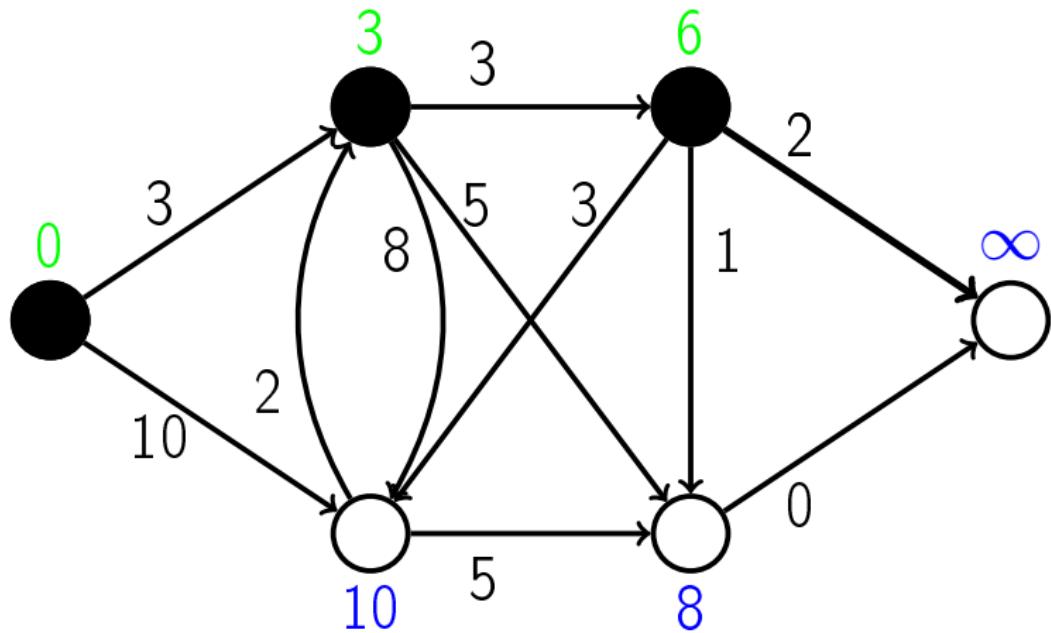
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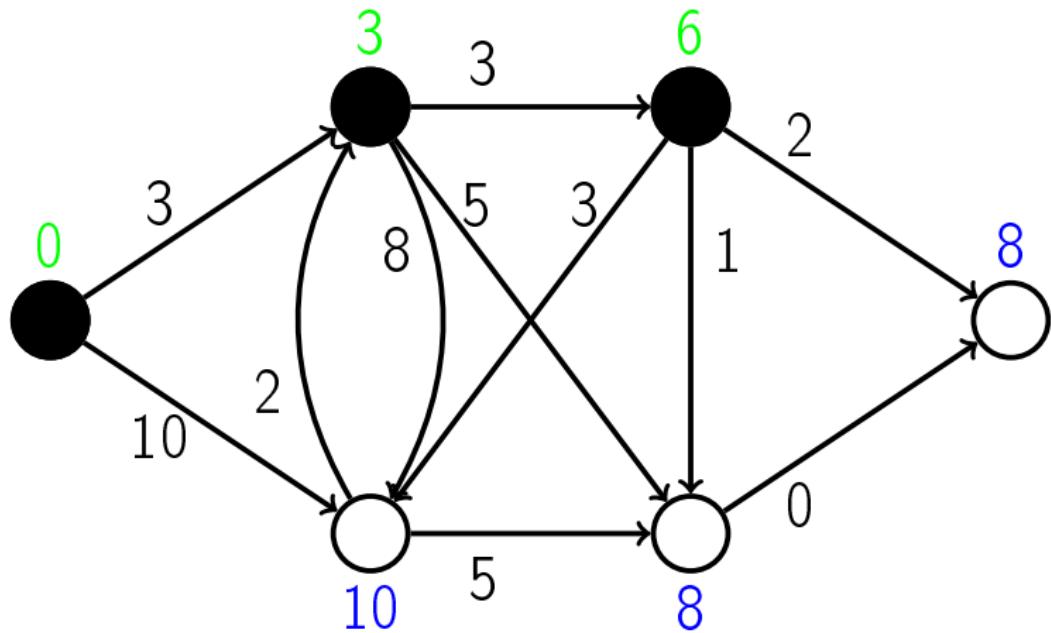
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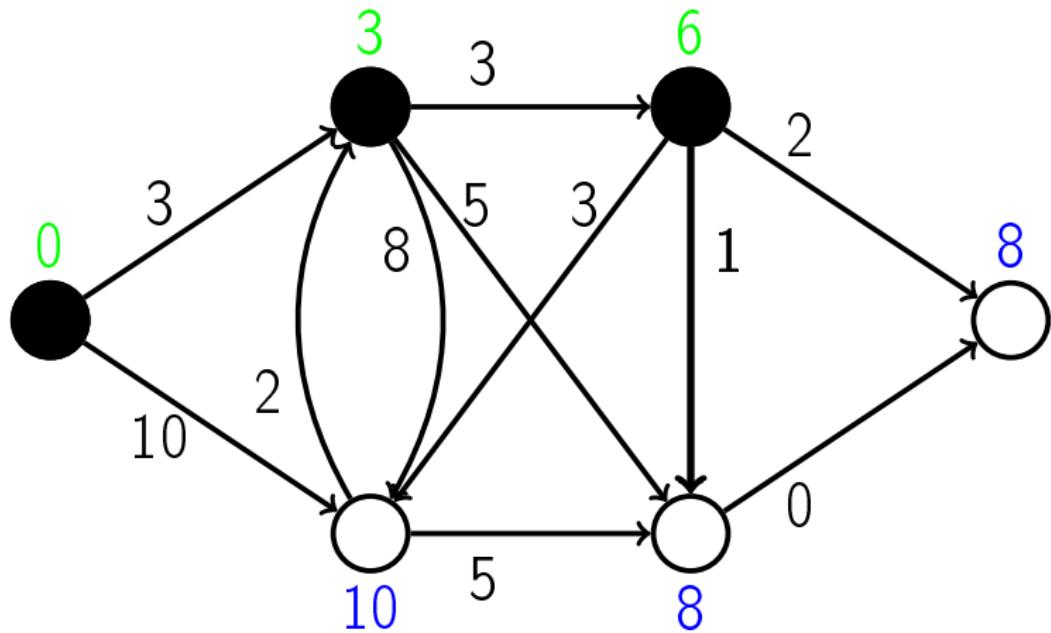
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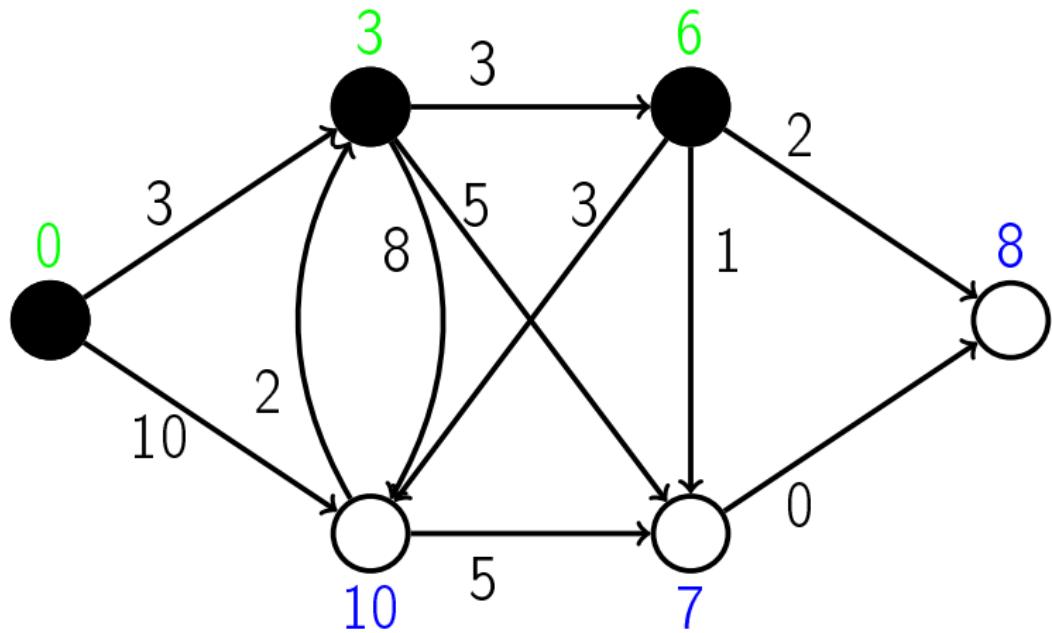
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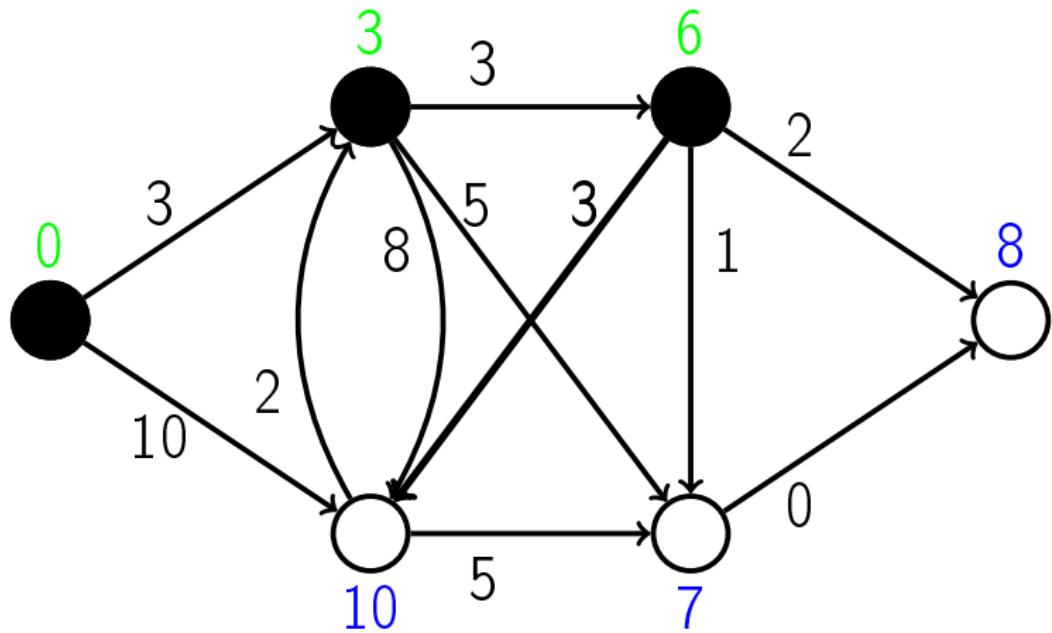
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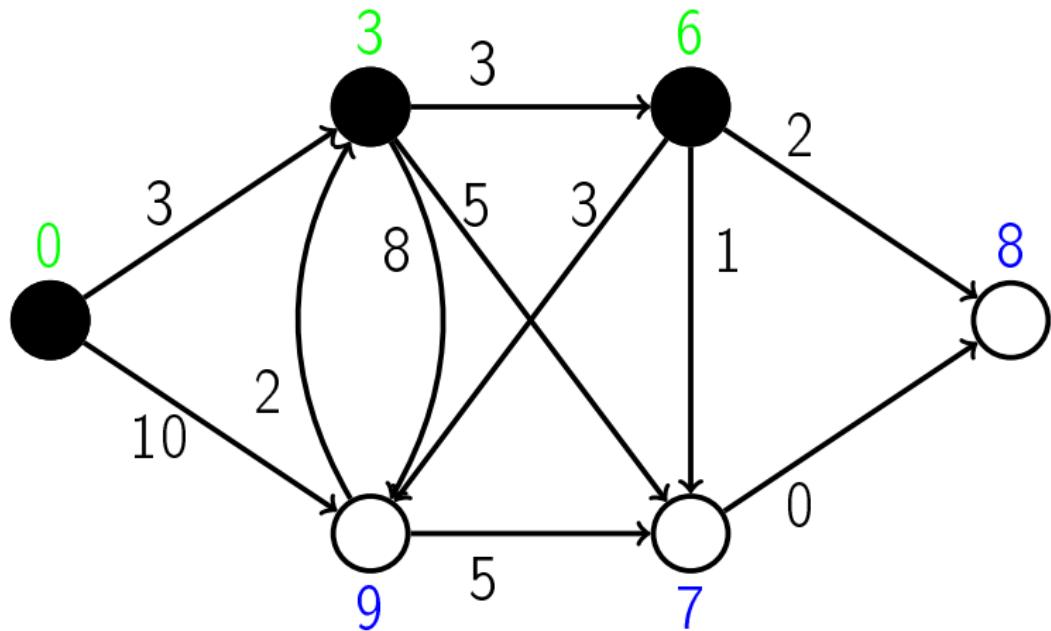
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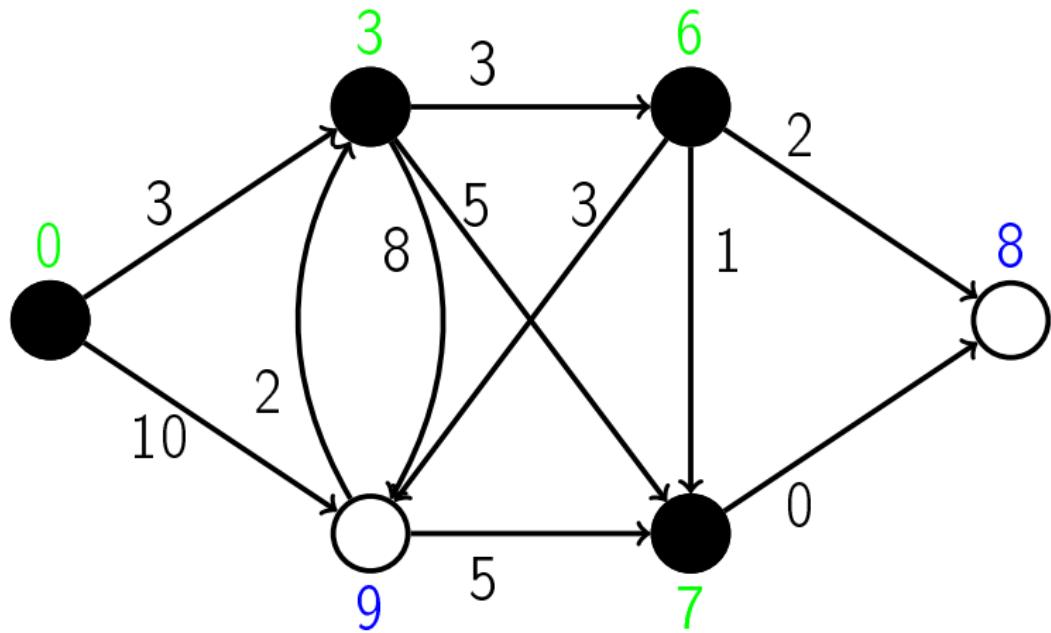
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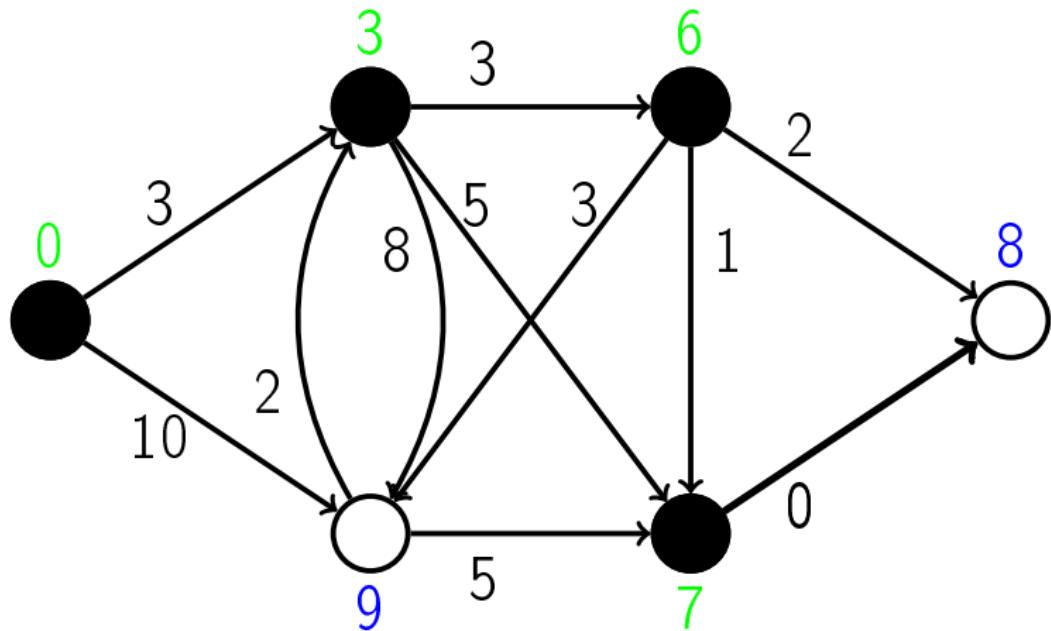
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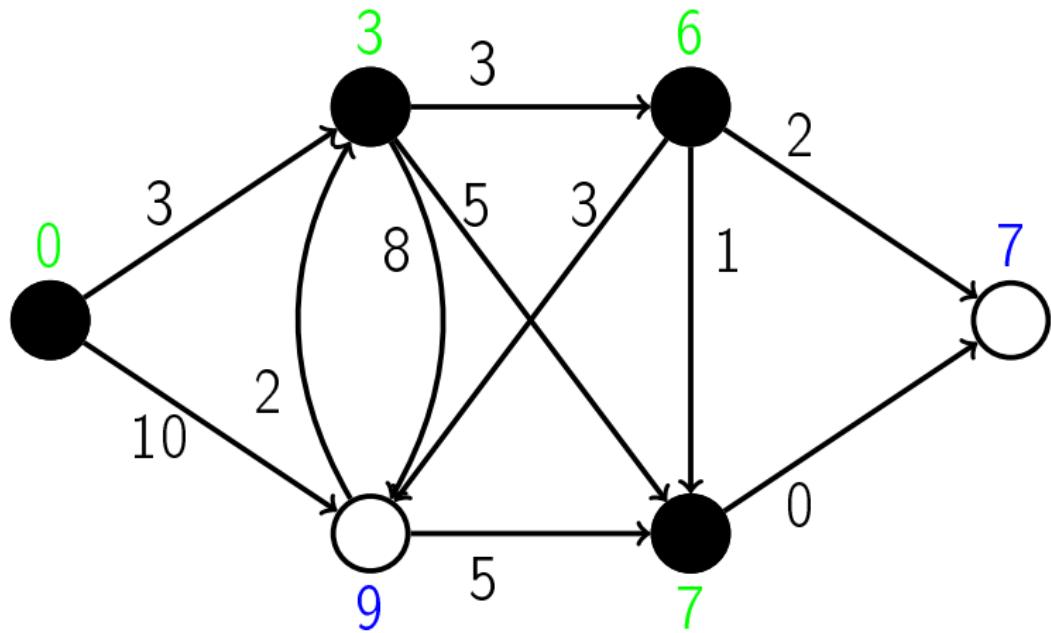
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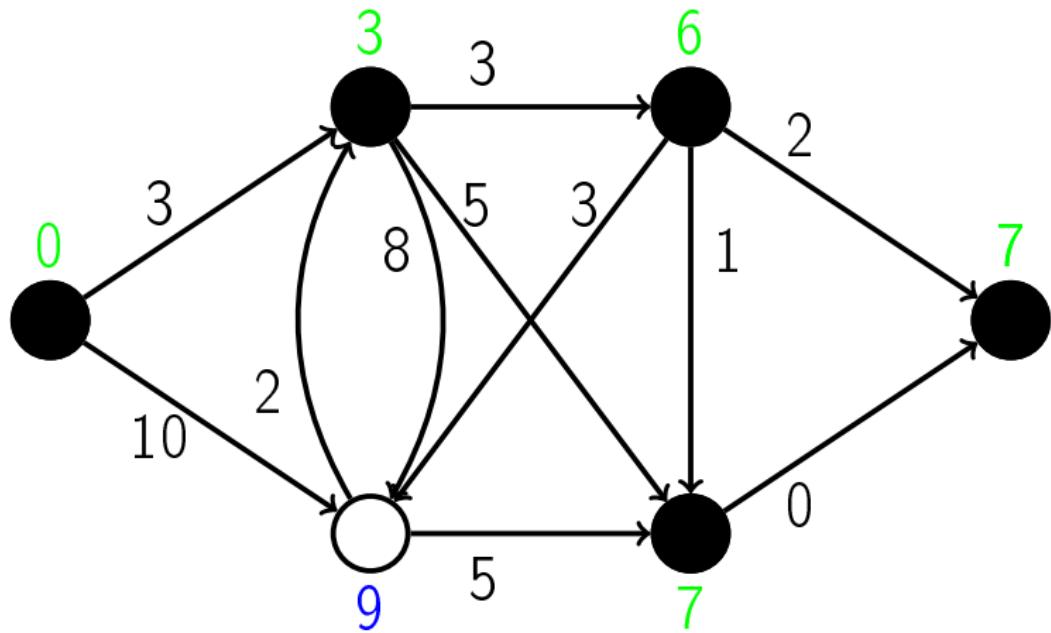
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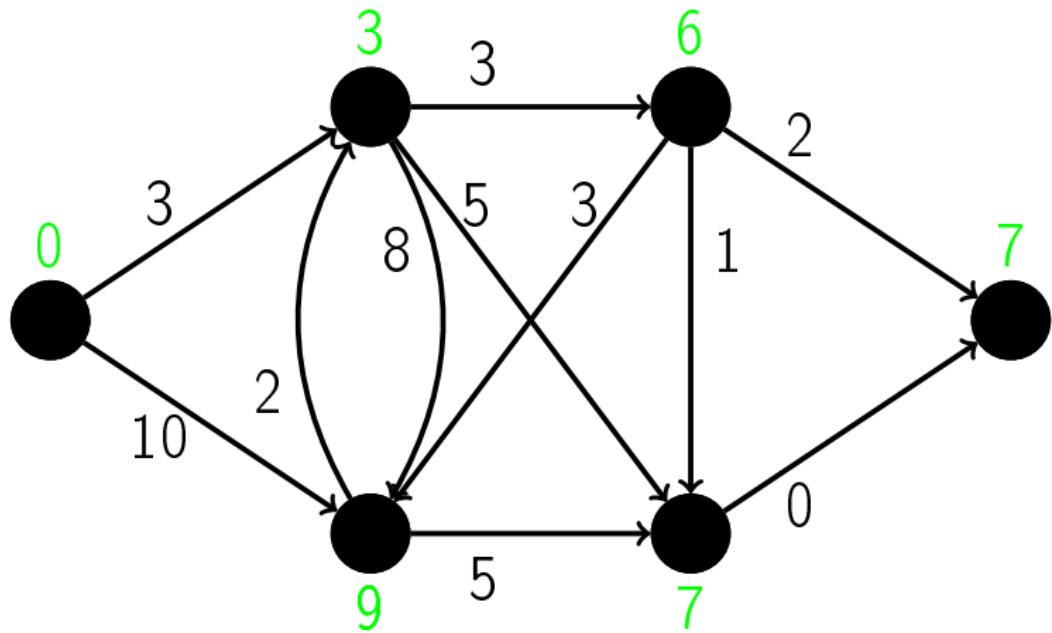
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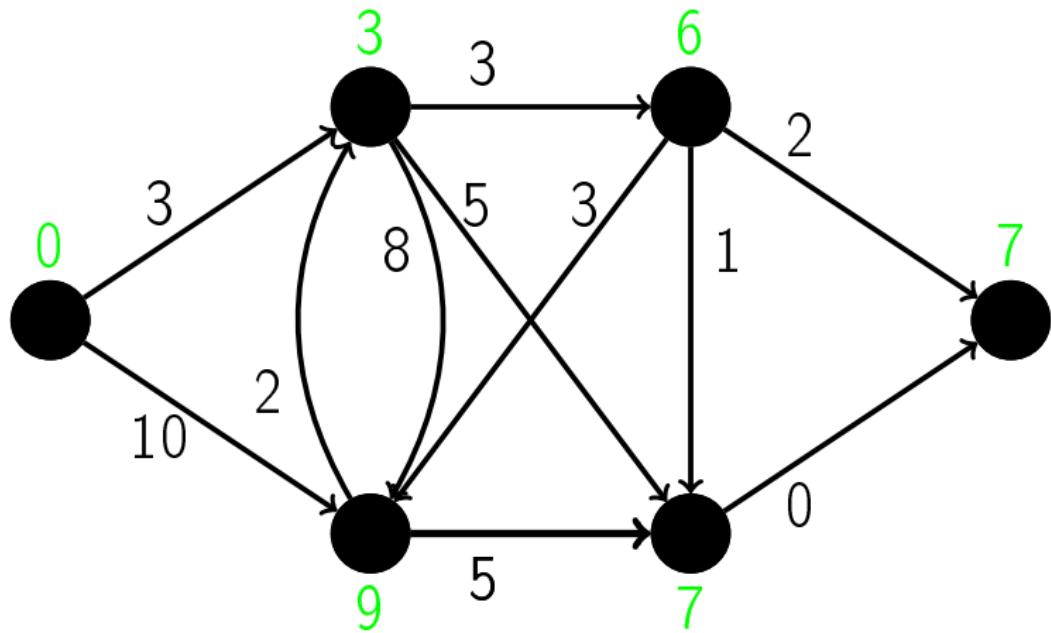
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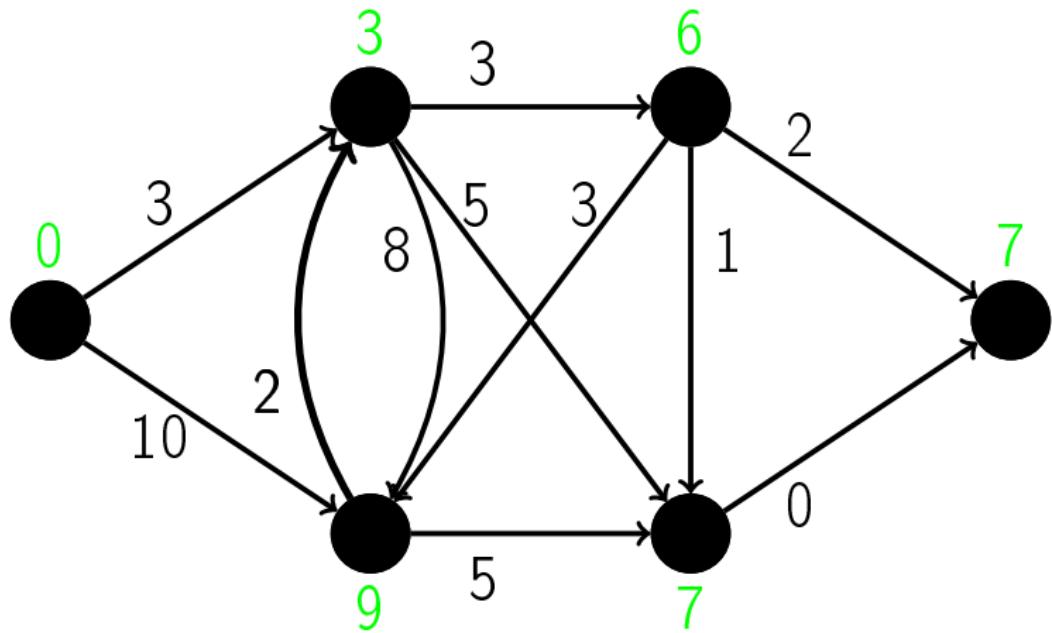
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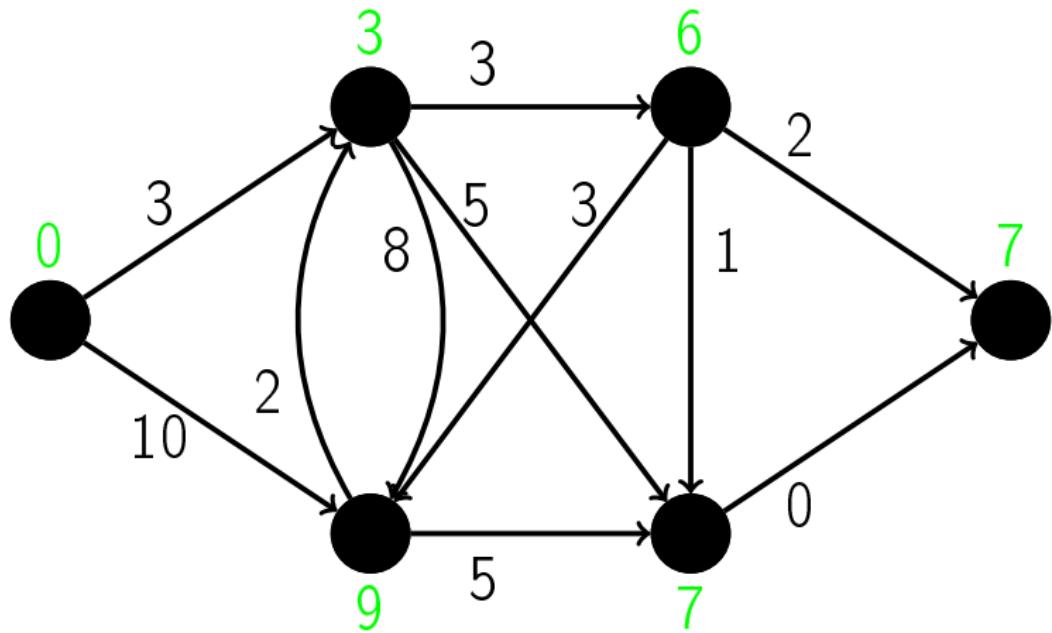
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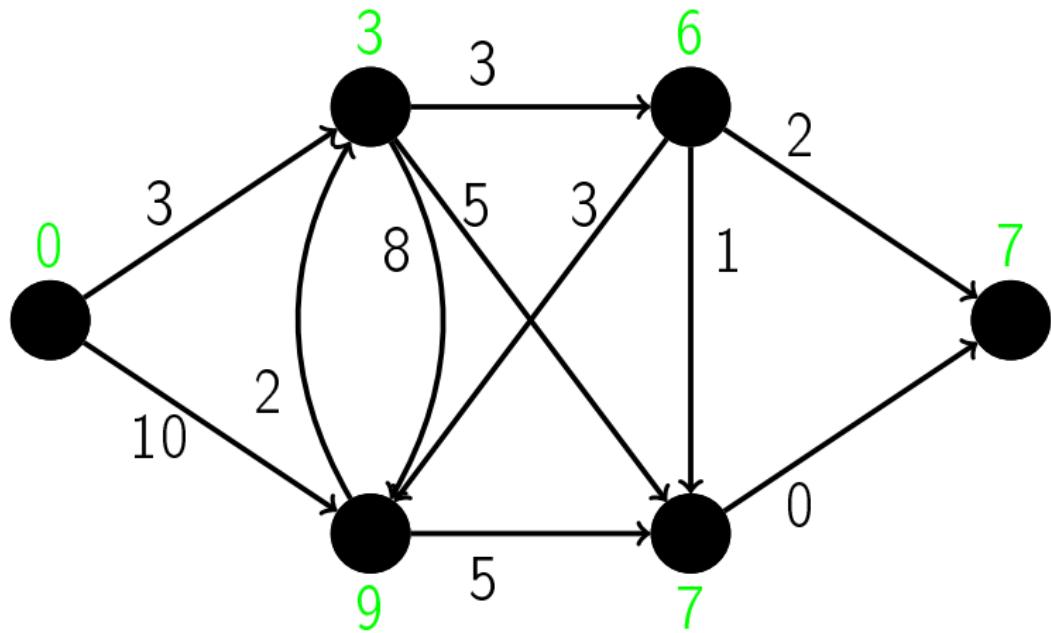
# Example



# Example



# Example



# Pseudocode

## Dijkstra( $G, S$ )

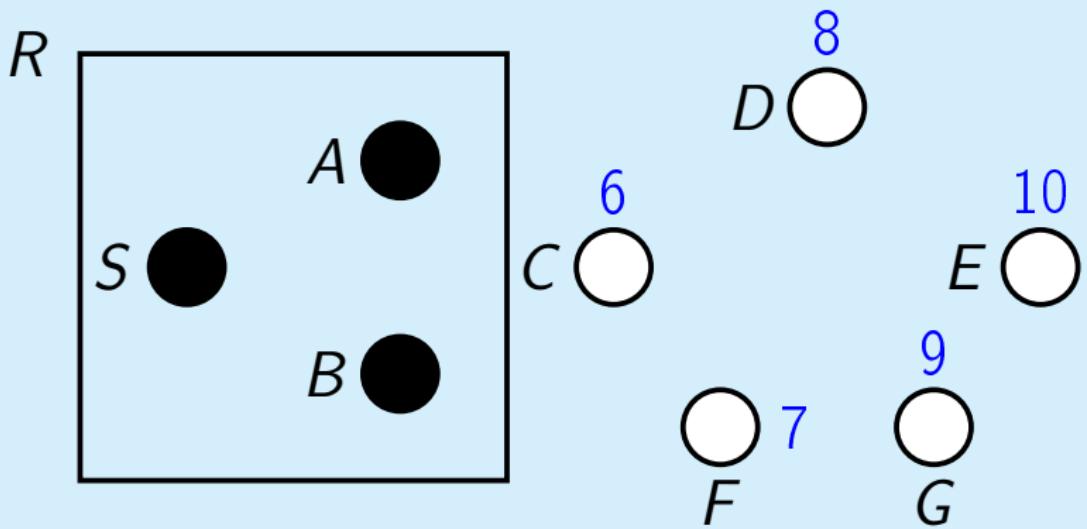
```
for all  $u \in V$ :
     $dist[u] \leftarrow \infty$ ,  $prev[u] \leftarrow \text{nil}$ 
 $dist[S] \leftarrow 0$ 
 $H \leftarrow \text{MakeQueue}(V)$  {dist-values as keys}
while  $H$  is not empty:
     $u \leftarrow \text{ExtractMin}(H)$ 
    for all  $(u, v) \in E$ :
        if  $dist[v] > dist[u] + w(u, v)$ :
             $dist[v] \leftarrow dist[u] + w(u, v)$ 
             $prev[v] \leftarrow u$ 
            ChangePriority( $H, v, dist[v]$ )
```

# Correct distances

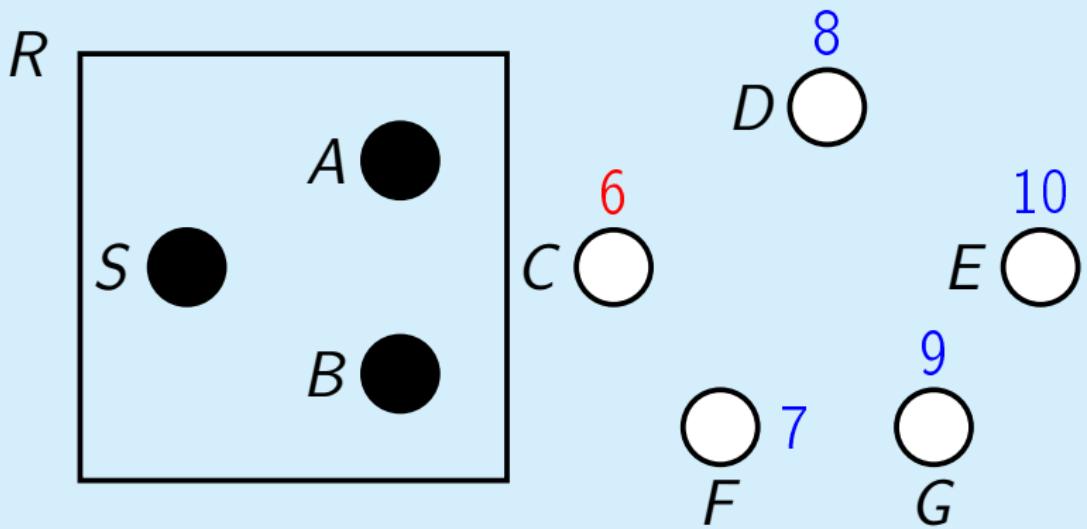
## Lemma

When a node  $u$  is selected via ExtractMin,  
 $\text{dist}[u] = d(S, u)$ .

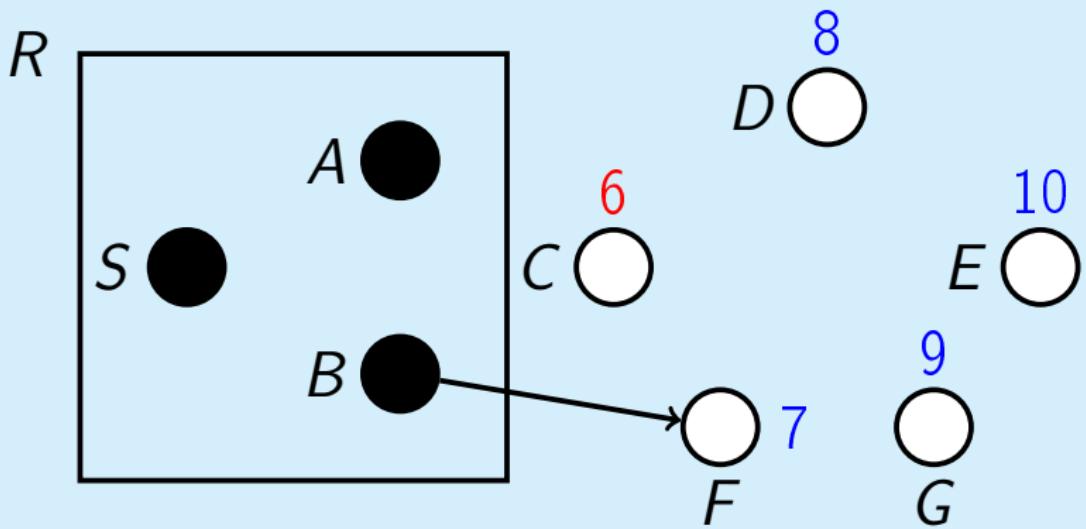
## Proof



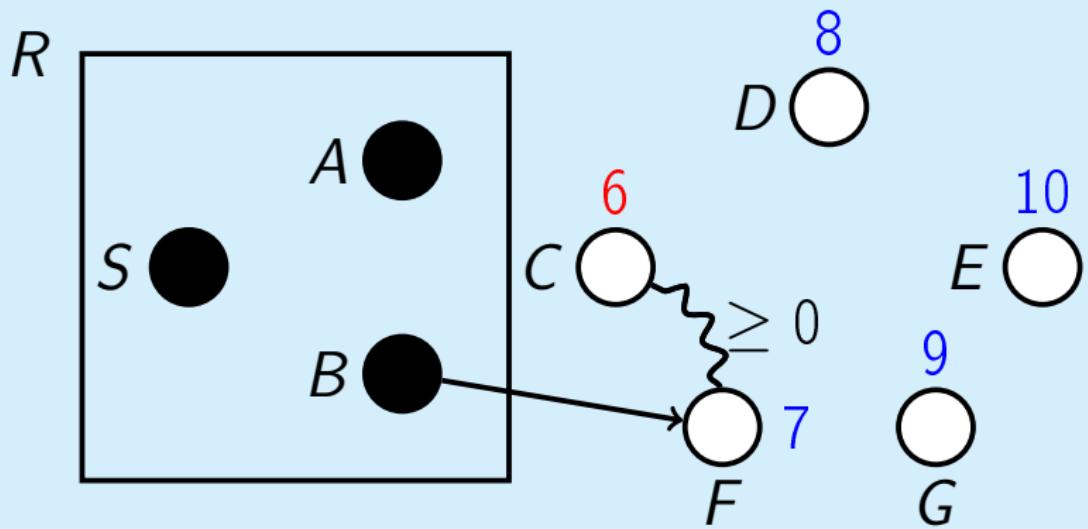
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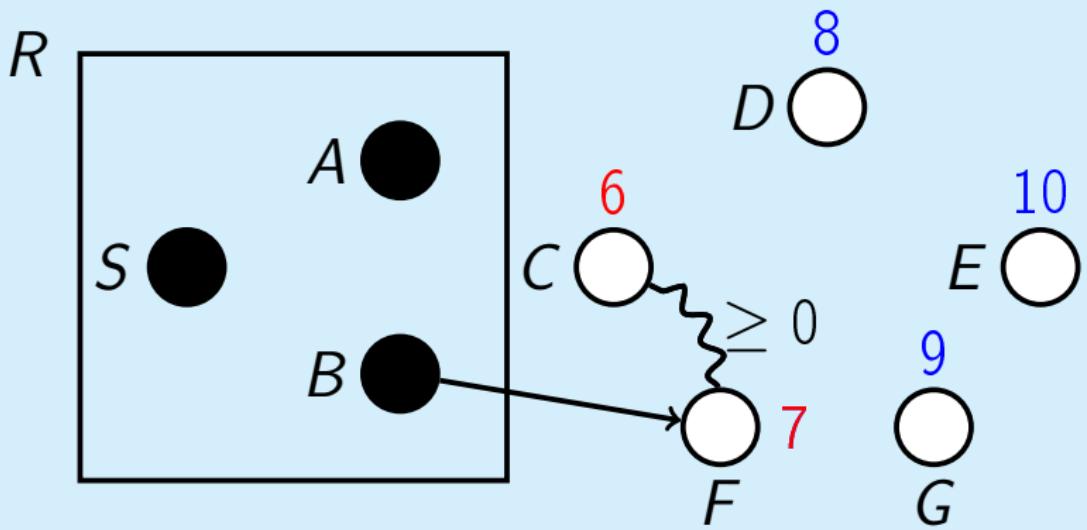
## Proof



## Proof



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# Running time

Total running time:

$$\begin{aligned} T(\text{MakeQueue}) + |V| \cdot T(\text{ExtractMin}) \\ + |E| \cdot T(\text{ChangePriority}) \end{aligned}$$

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Priority queue implementations:

- array:

$$O(|V| + |V|^2 + |E|) = O(|V|^2)$$

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$$T(\text{MakeQueue}) + |V| \cdot T(\text{ExtractMin}) \\ + |E| \cdot T(\text{ChangePriority})$$

Priority queue implementations:

- array:

$$O(|V| + |V|^2 + |E|) = O(|V|^2)$$

- binary heap:

$$O(|V| + |V| \log |V| + |E| \log |V|) = \\ O((|V| + |E|) \log |V|)$$

# Conclusion

- Can find the minimum time to get from work to home
- Can find the fastest route from work to home
- Works for any graph with non-negative edge weights
- Works in  $O(|V|^2)$  or  $O((|V| + |E|) \log(|V|))$  depending on the implementation