

# Paths in Graphs: Fastest Route

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Higher School of Economics

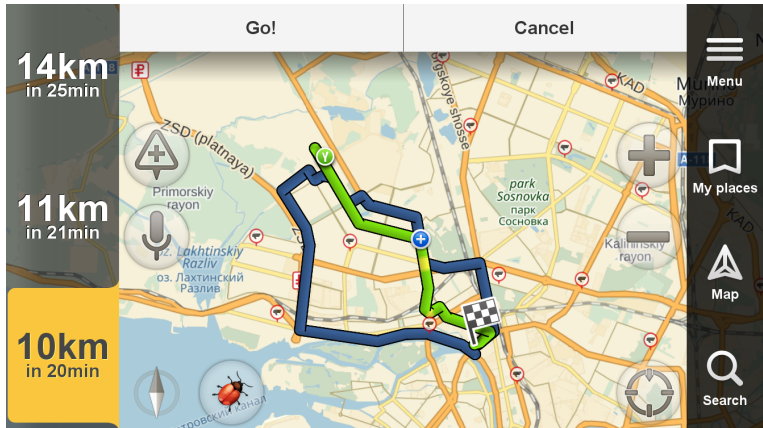
Graph Algorithms  
Data Structures and Algorithms

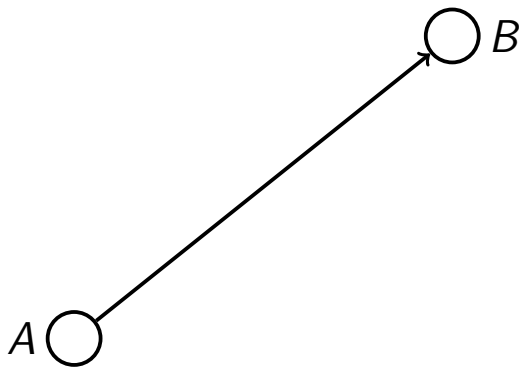
# Outline

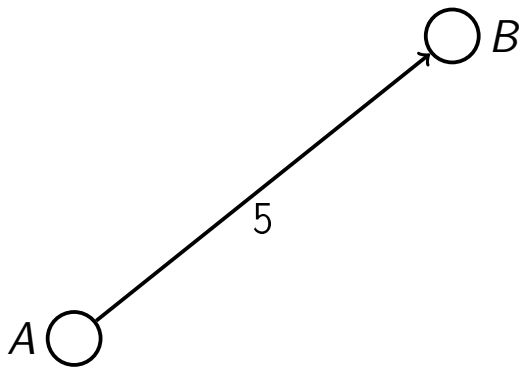
- 1 Fastest Route
- 2 Naive Algorithm
- 3 Dijkstra's Algorithm

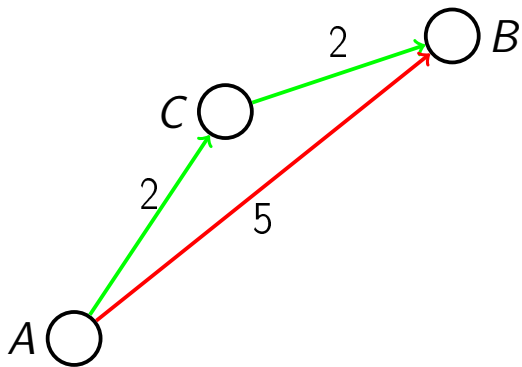
# Fastest Route

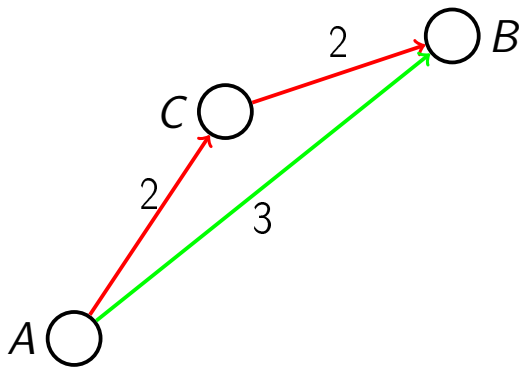
What is the fastest route to get home from work?





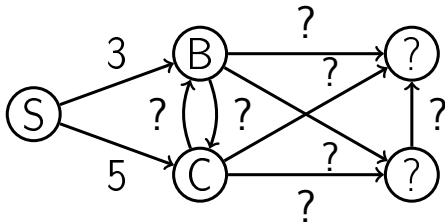






# Intuition

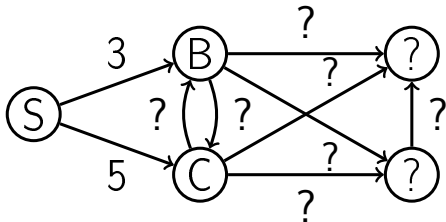
- Assume that we stay at  $S$  and observe two outgoing edges:





# Intuition

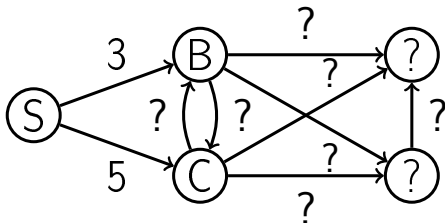
- Assume that we stay at  $S$  and observe two outgoing edges:



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

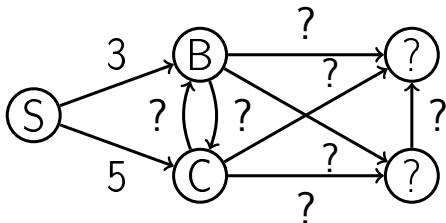
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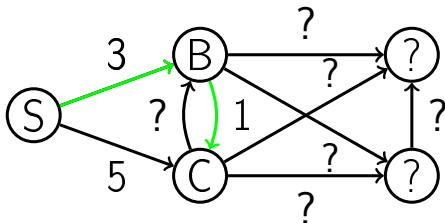
- Can we be sure that the distance from  $S$  to  $C$  is 5?



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

# Intuition

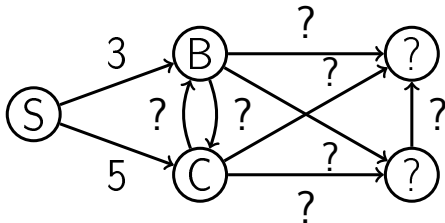
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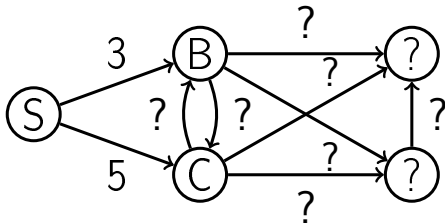
# Intuition

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



# Intuition

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



- Yes, because there are no negative weight edges.

# Outline

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# Optimal substructure

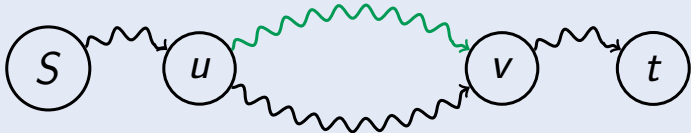
## Observation

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  $\text{dist}[v]$ .

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.

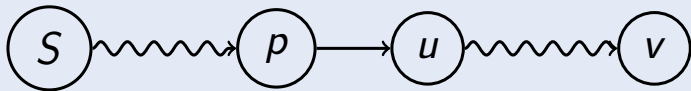
## Proof

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

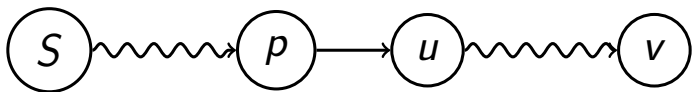


## Proof

- 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  $u$ .

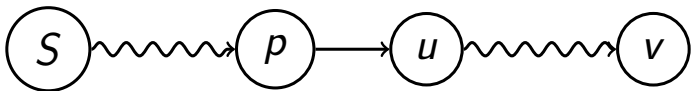


## Proof (continued)



- 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)

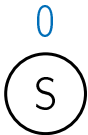


- Then  $d(S, p) = \text{dist}[p]$  and hence  $d(S, u) = d(S, p) + w(p, u) = \text{dist}[p] + w(p, u)$
- $\text{dist}[u] > d(S, u) = \text{dist}[p] + w(p, u) \Rightarrow$  edge  $(p, u)$  can be relaxed — a contradiction. □

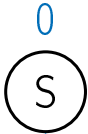
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# Intuition

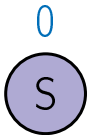


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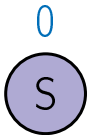


initially, we only know the distance to S

# Intuition



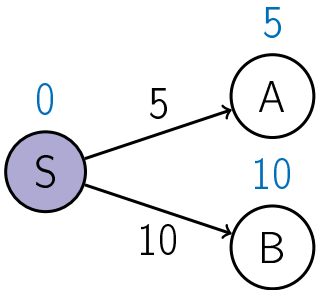
# Intuition



let's relax all the edges from S

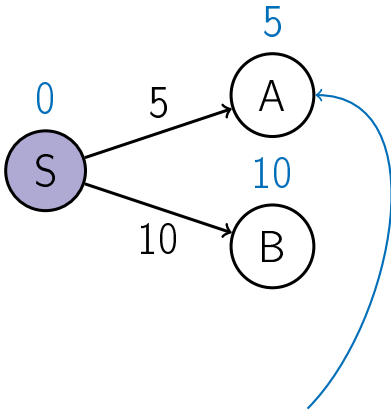


# Intuition



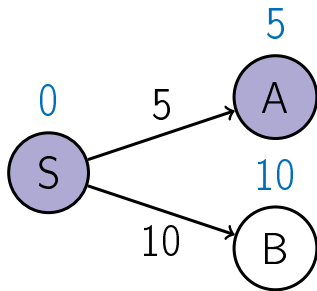
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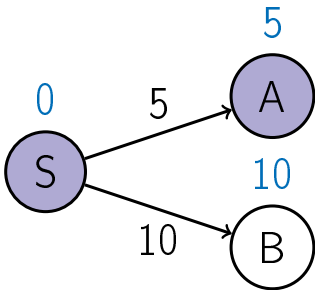


we now know the distance for A

# Intuition

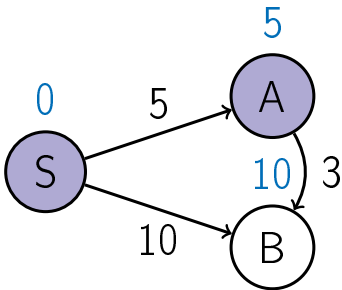


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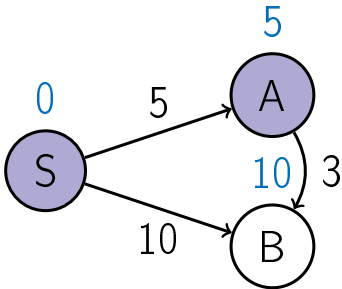
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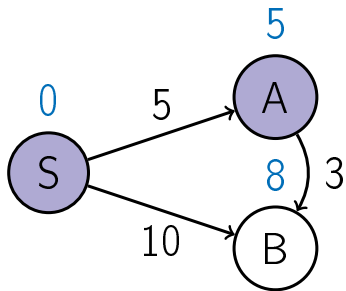
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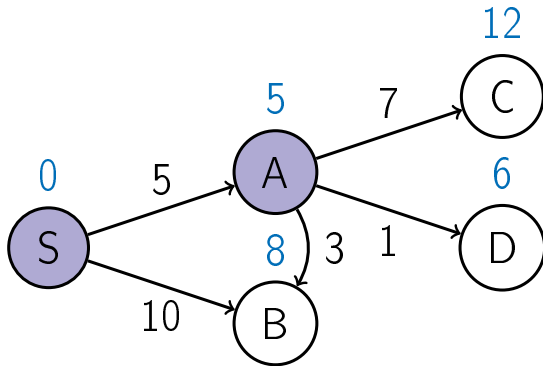


we discover an edge  $(A, B)$  of weight 3  
that updates  $\text{dist}[B]$

# Intuition



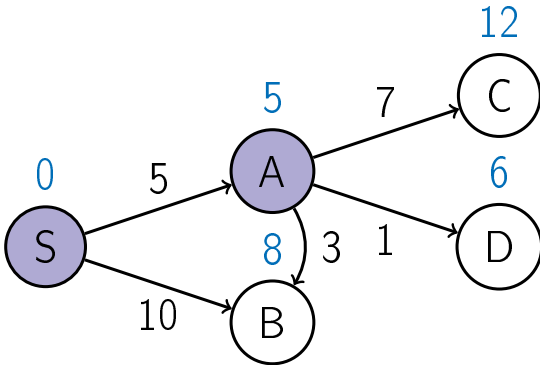
# Intuition



we also discover a few more outgoing edges

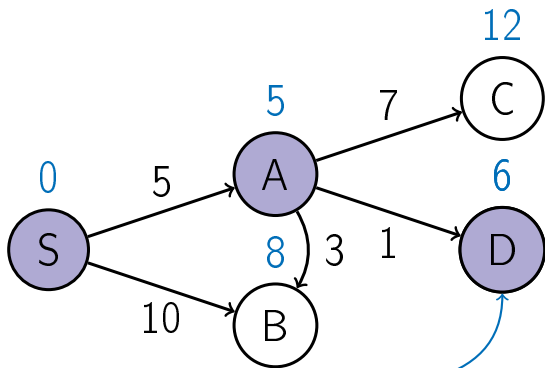


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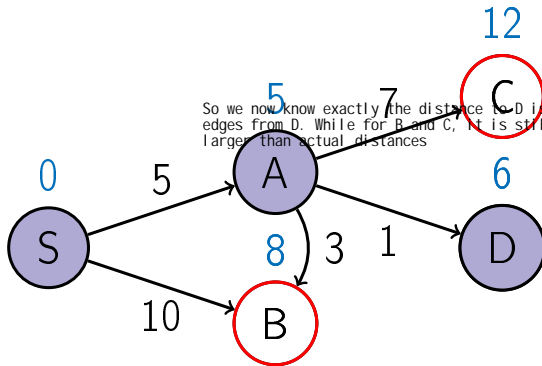
what is the next vertex for which we already know the correct distance?

# Intuition



it is  $D$

# Intuition



So we now know exactly the distance to D is 6, and we can continue by relaxing edges from D. While for B and C, it is still possible that their dist values are larger than actual distances

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 `dist` is already set correctly (“known region”).

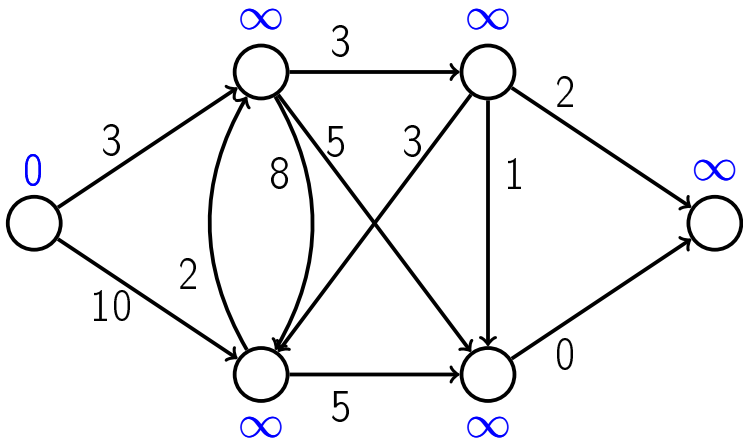
# Main ideas of Dijkstra's Algorithm

- We maintain a set  $R$  of vertices for which `dist` is already set correctly (“known region”).
- The first vertex added to  $R$  is  $S$ .

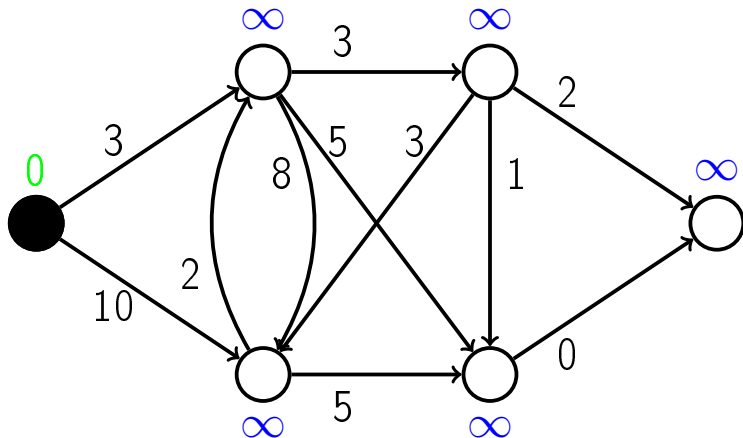
# Main ideas of Dijkstra's Algorithm

- We maintain a set  $R$  of vertices for which `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 `dist`-value, add it to  $R$ , and relax all its outgoing edges.

# Example

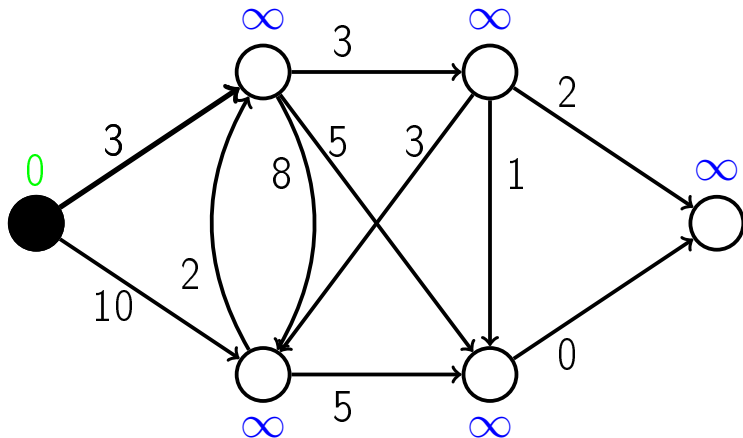


# Example

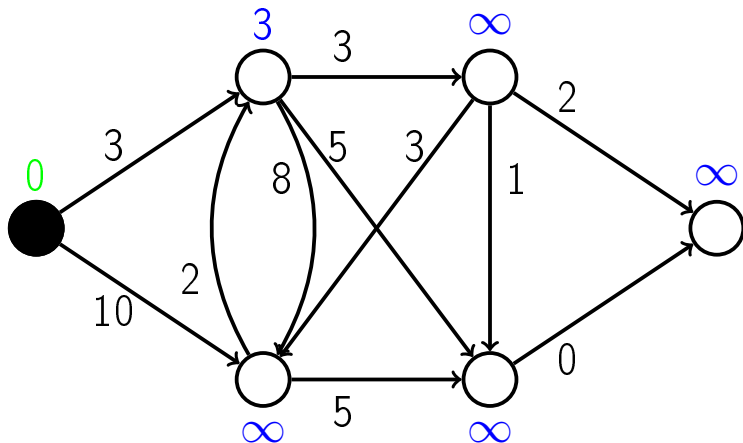




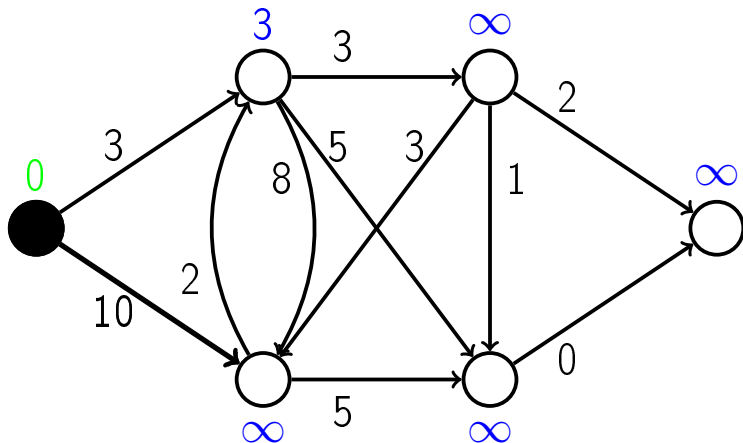
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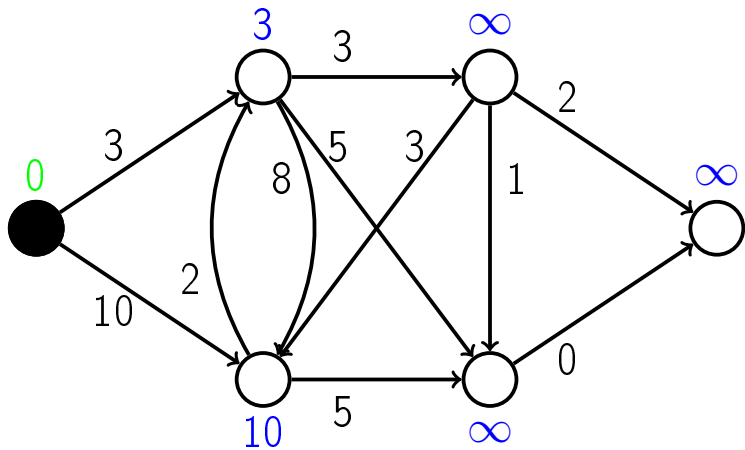
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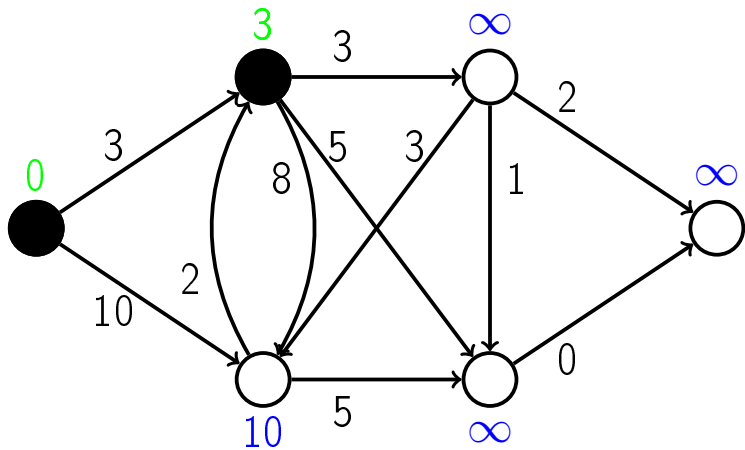
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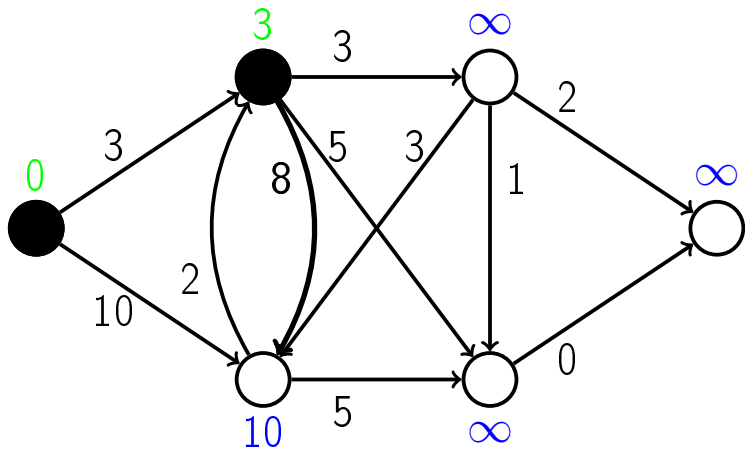
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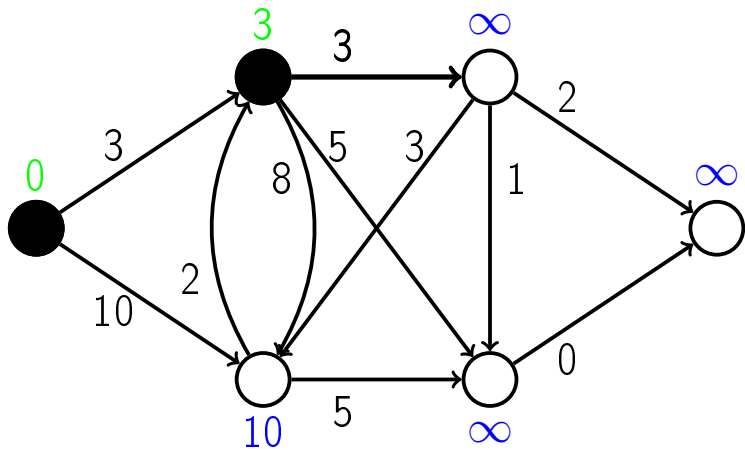
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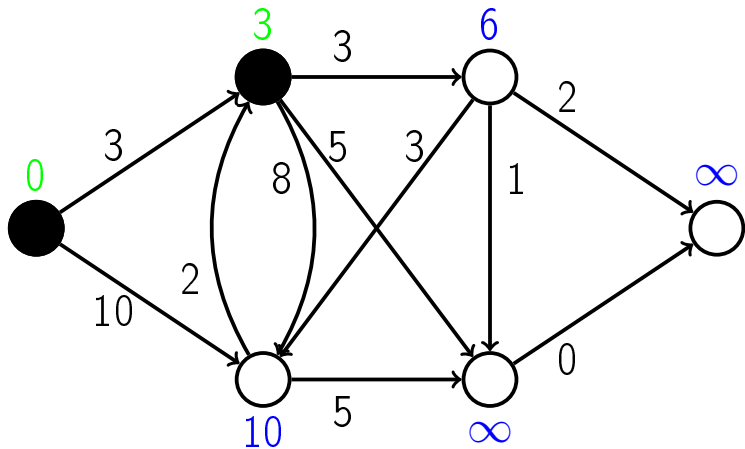
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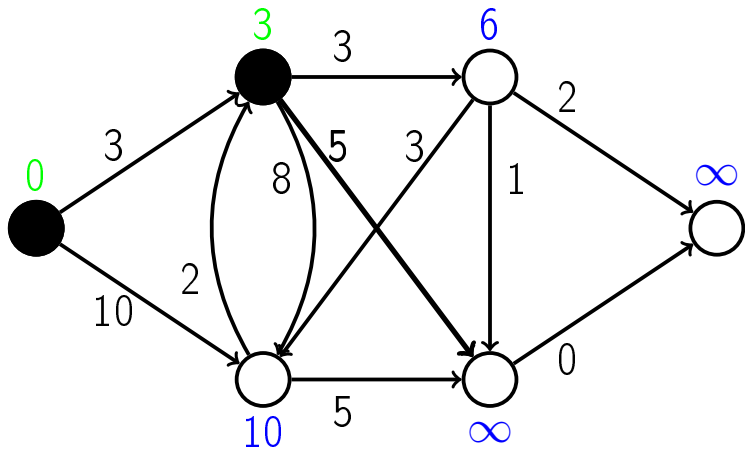


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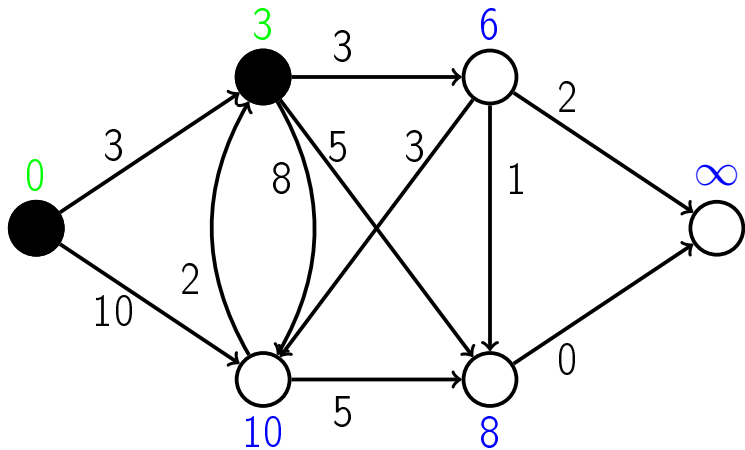




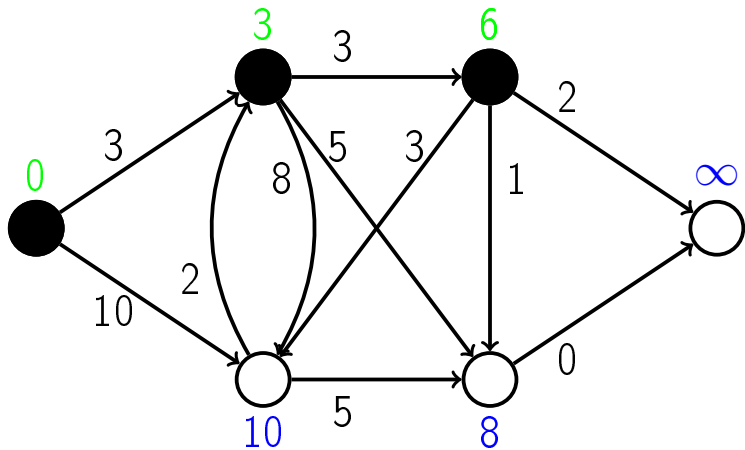
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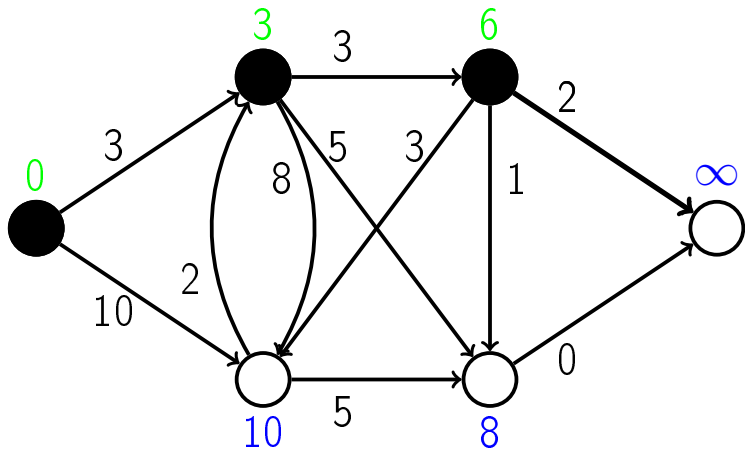
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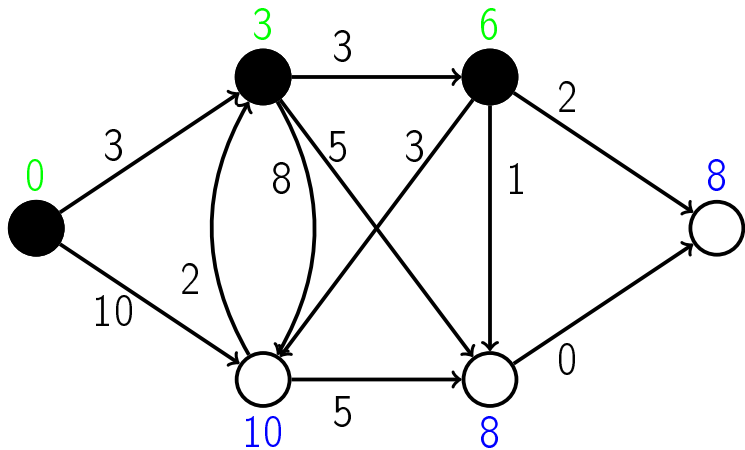
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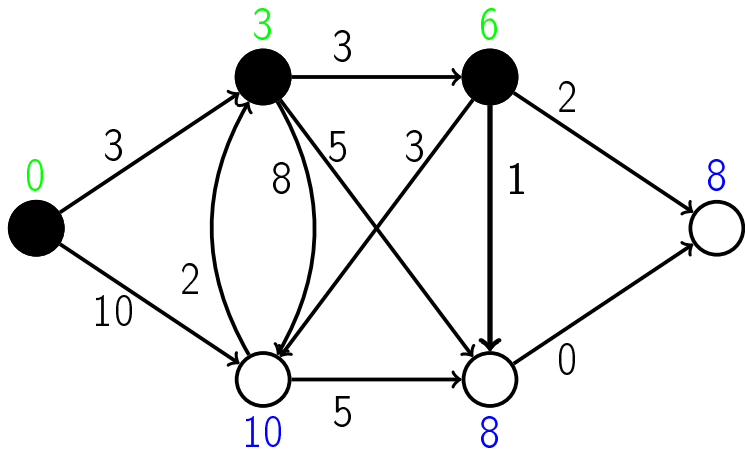
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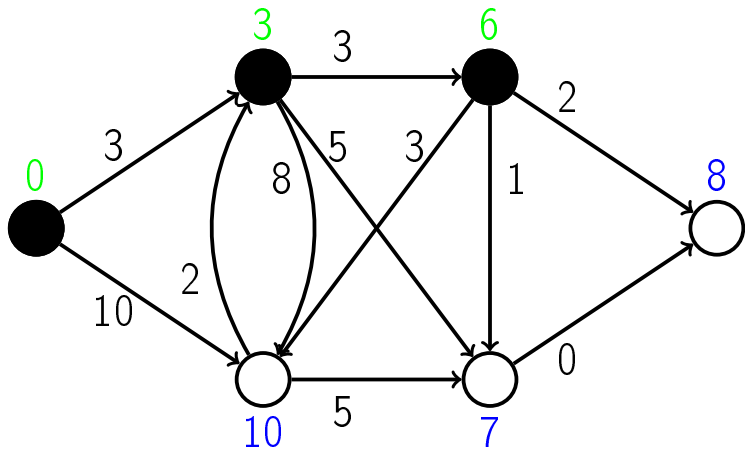
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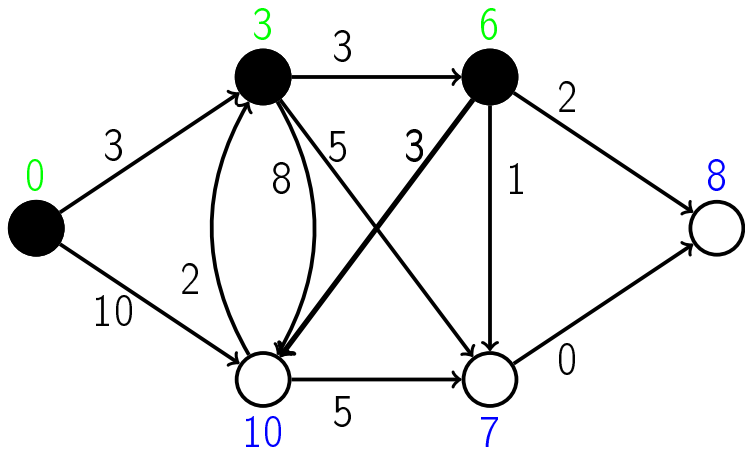
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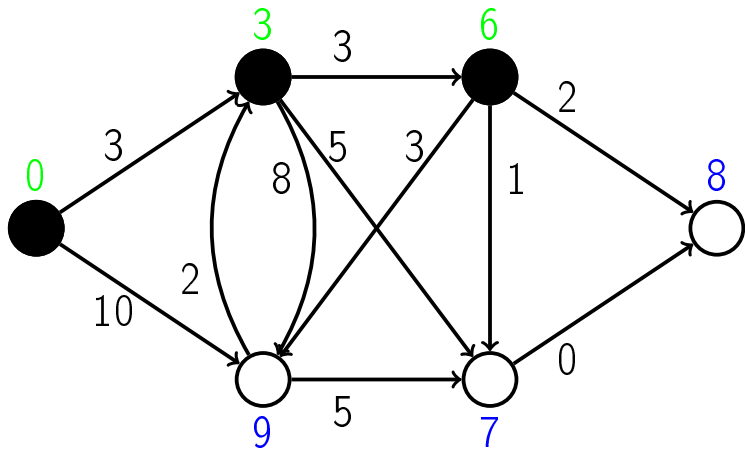


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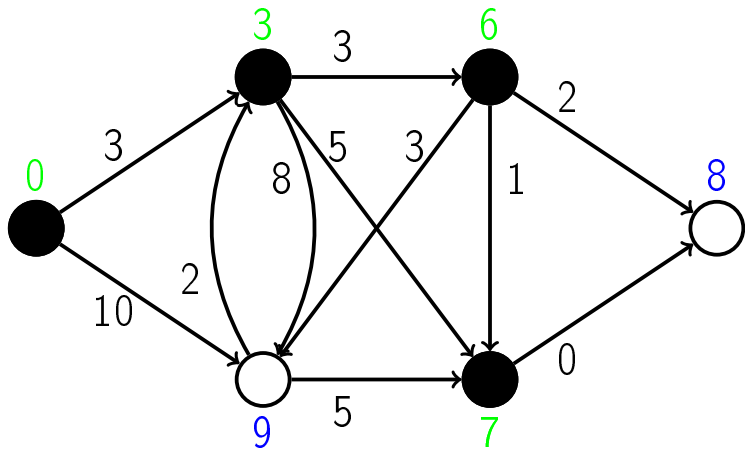




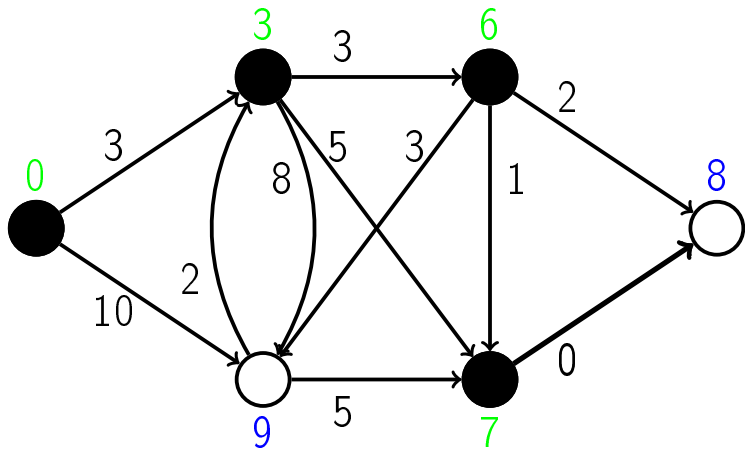
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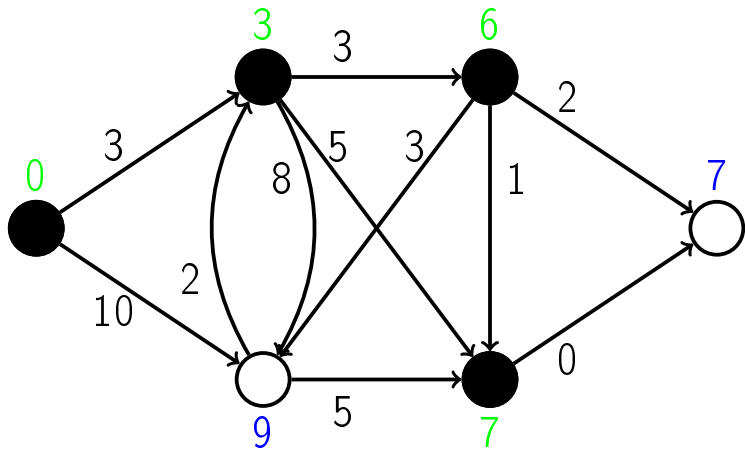
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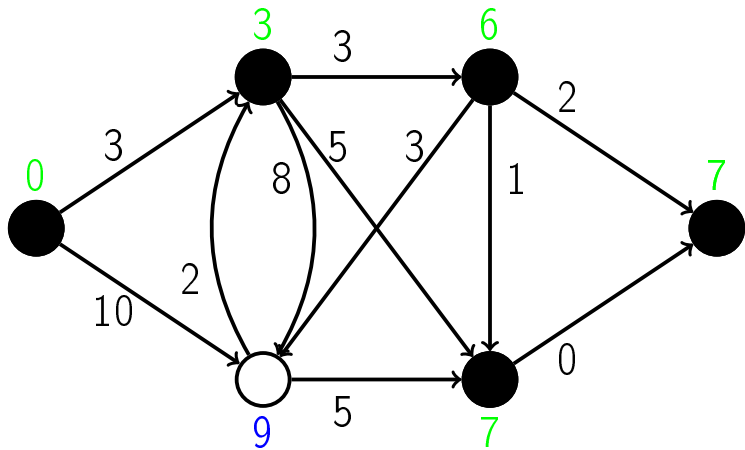
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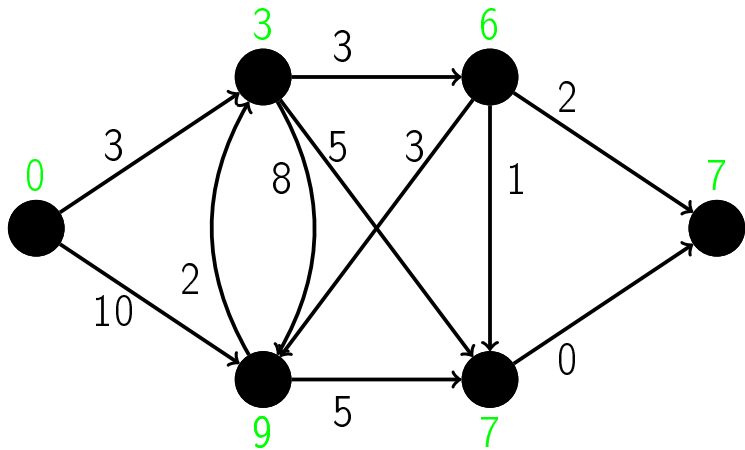
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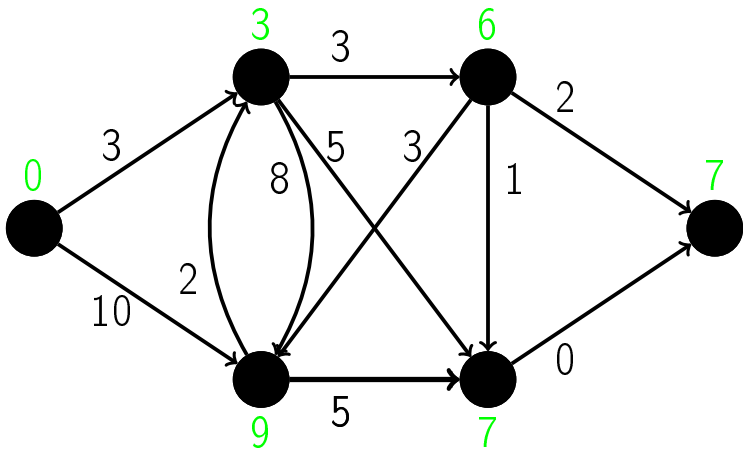
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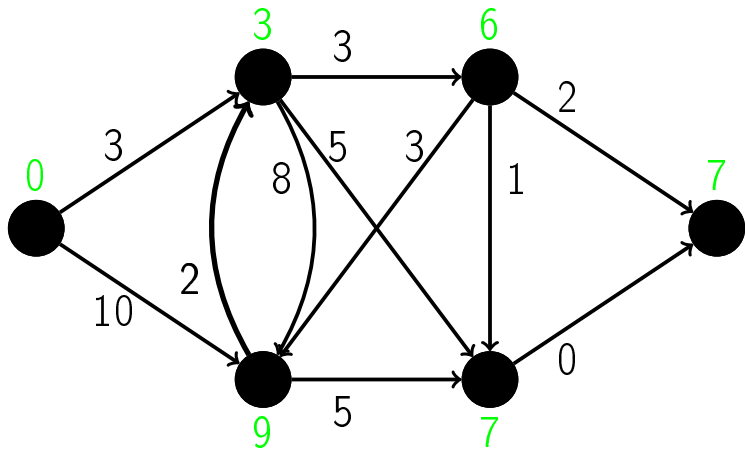
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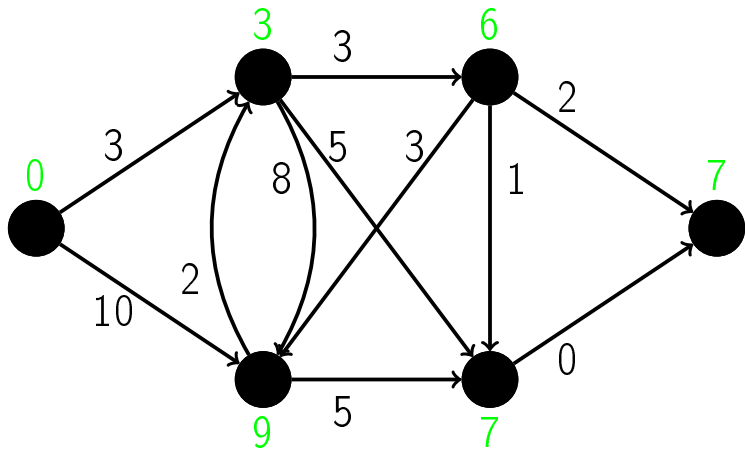


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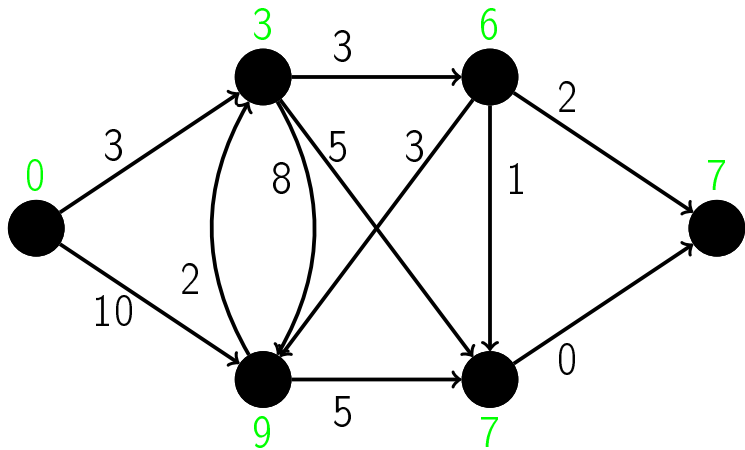




# Example



# Example



# Pseudocode

## Dijkstra( $G, S$ )

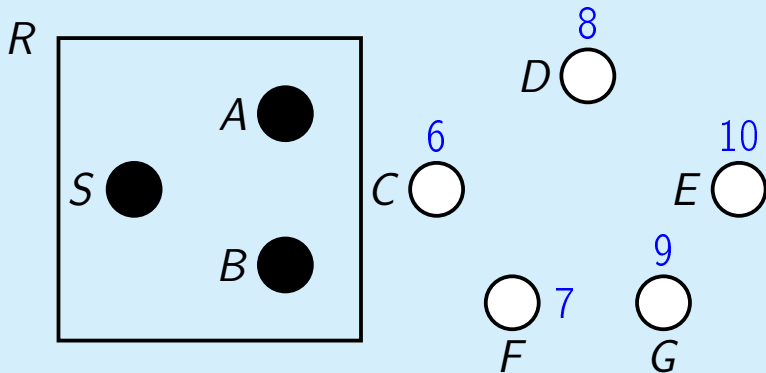
```
for all  $u \in V$ :  
     $\text{dist}[u] \leftarrow \infty, \text{prev}[u] \leftarrow \text{nil}$   
 $\text{dist}[S] \leftarrow 0$   
 $H \leftarrow \text{MakeQueue}(V)$  {dist-values as keys}  
while  $H$  is not empty:  
     $u \leftarrow \text{ExtractMin}(H)$   
    for all  $(u, v) \in E$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

# Correct distances

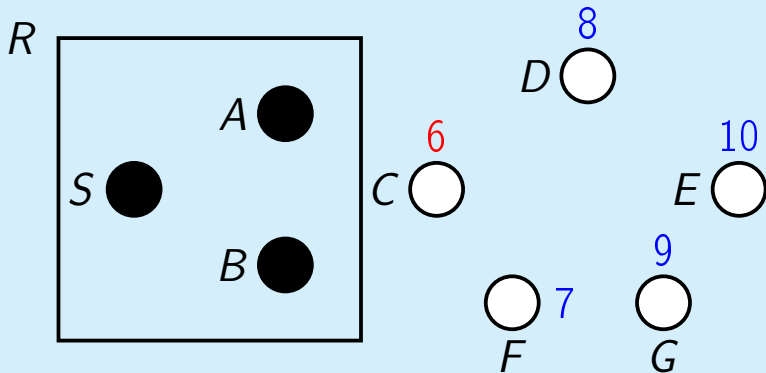
## Lemma

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

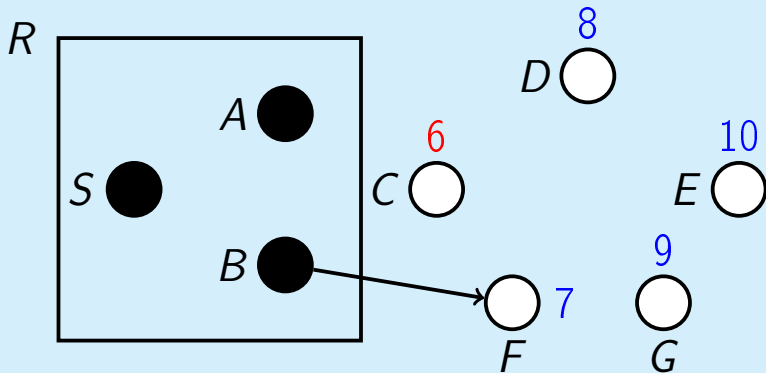
# Proof



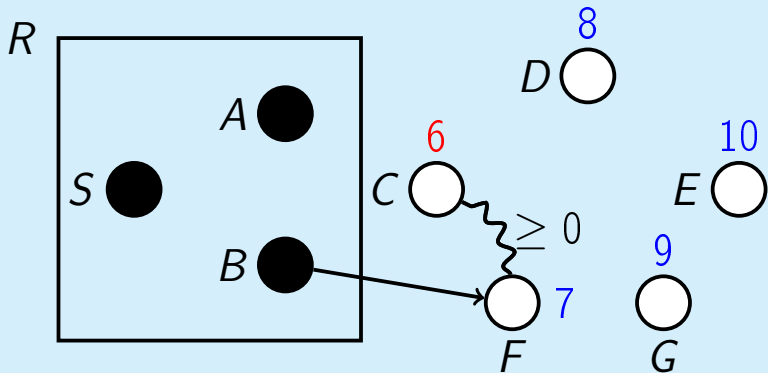
# Proof



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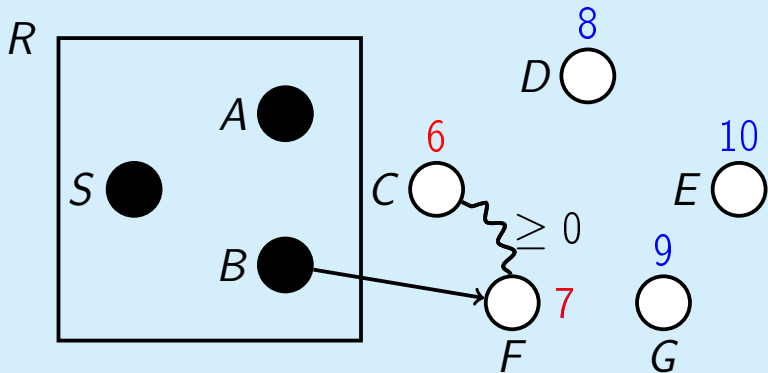


# Proof





# Proof



# 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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Priority queue implementations:

- array:

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

- binary heap:

$$\begin{aligned} O(|V| + |V| \log |V| + |E| \log |V|) = \\ O((|V| + |E|) \log |V|) \end{aligned}$$

# 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