

Dijkstra's Algorithm

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Problem Statement

First Route:



Problem Statement

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First Route:

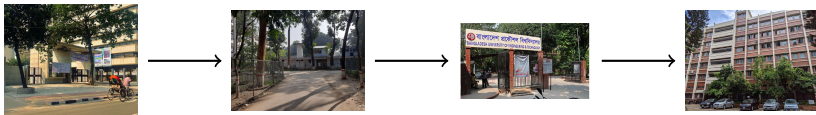


Second Route:



Problem Statement

First Route:



Second Route:



Problem Statement

First Route:

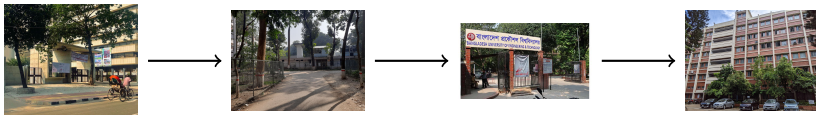


Second Route:



Problem Statement

First Route:



Second Route:



Problem Statement

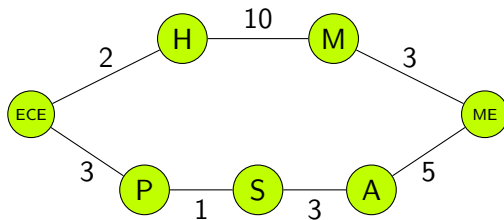
First Route:



Second Route:



Problem Statement



A Graph Representation of the previous problem

Dijkstra's Algorithm

Definition

Dijkstra's algorithm is a greedy algorithm for finding the shortest paths between nodes in a weighted graph.



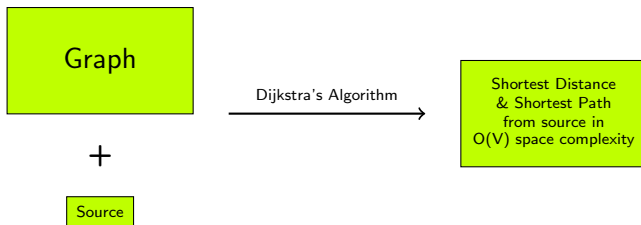
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$O(V^2)$ Algorithm

What is the shortest way to travel from Rotterdam to Groningen, in general: from given city to given city. It is the algorithm for the shortest path, which I designed in about twenty minutes. One morning I was shopping in Amsterdam with my young fiancée, and tired, we sat down on the café terrace to drink a cup of coffee and I was just thinking about whether I could do this, and I then designed the algorithm for the shortest path. [1]

History

$O(V^2)$ Algorithm

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$O(E + V \log V)$ Algorithm

In 1984, Fredman & Tarjan proposed use of fibonacci heap to optimize the running time of the algorithm to $O(\|E\| + \|V\| \log V)$ [2]

Pseudocode

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1      def Dijkstra(Graph, source):
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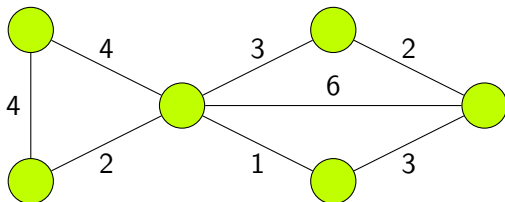
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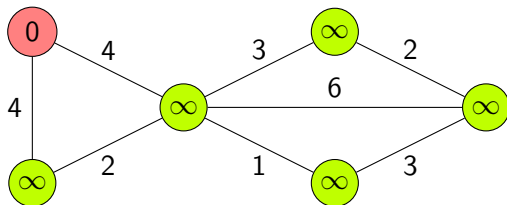
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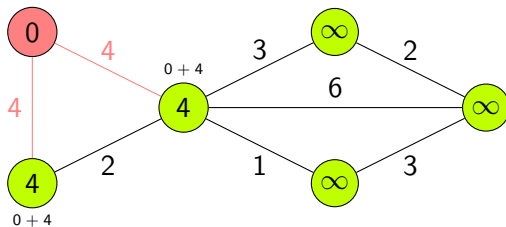
Simulation



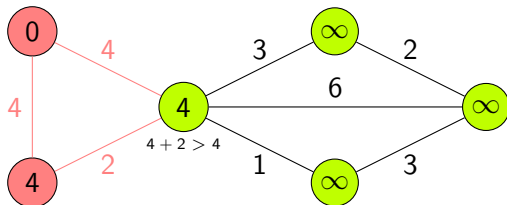
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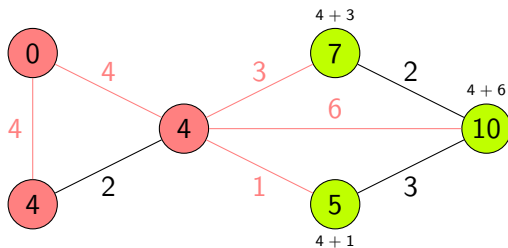
Simulation



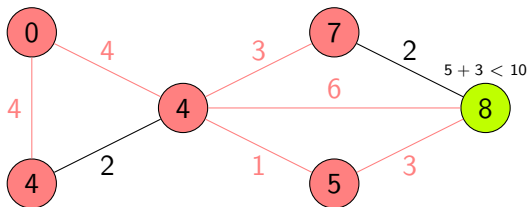
Simulation



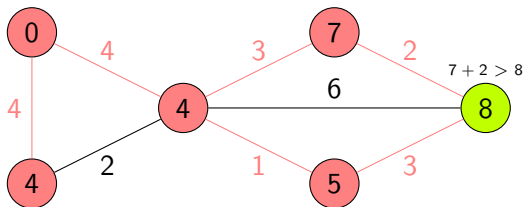
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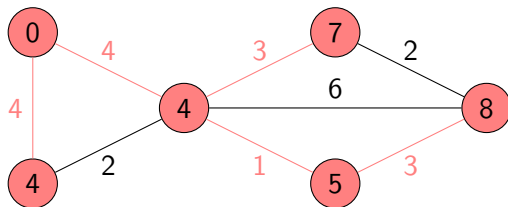
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Running time - revisiting the pseudocode

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Running time

Can you identify the costly operations performed here?

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Running time

Number of calls

Extract min (T_{em}) $\Theta(|V|)$

Decrease priority (T_{dp}) $\Theta(|E|)$

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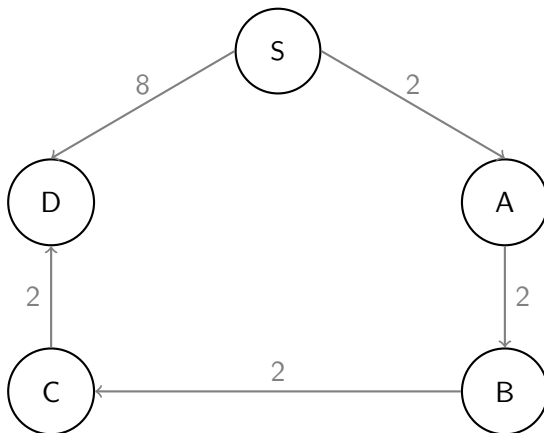
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Binary Heap	$\Theta(\log V)$	$\Theta(\log V)$	$\Theta((E + V)\log V)$
Fibonacci Heap	$\Theta(1)$	$\Theta(\log V)$	$\Theta(E + V\log V)$

Traversing unnecessarily

When we are eager to know the distance from the source to a particular node, Dijkstra's algorithm may result in a longer run-time.

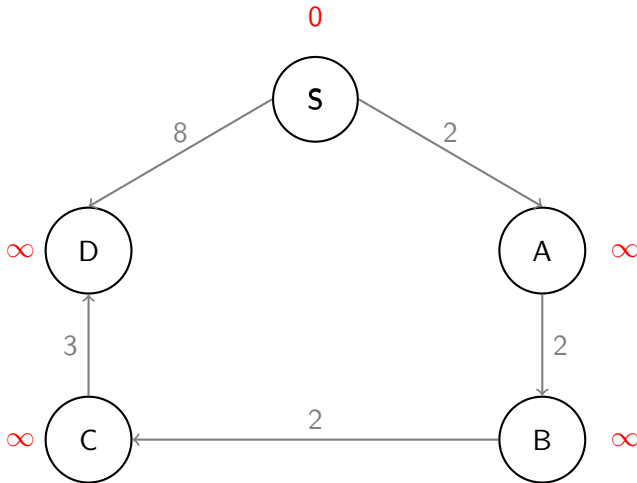
Traversing unnecessarily

- Find the shortest distance from S to D



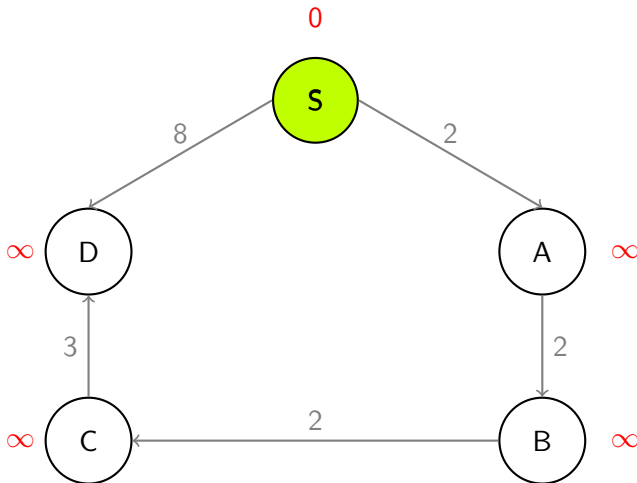
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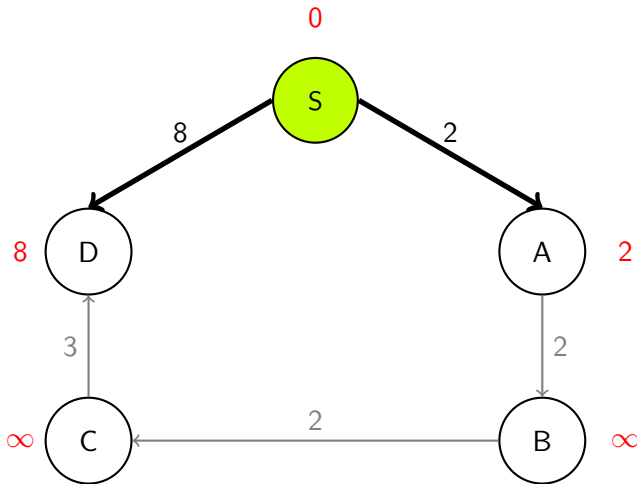
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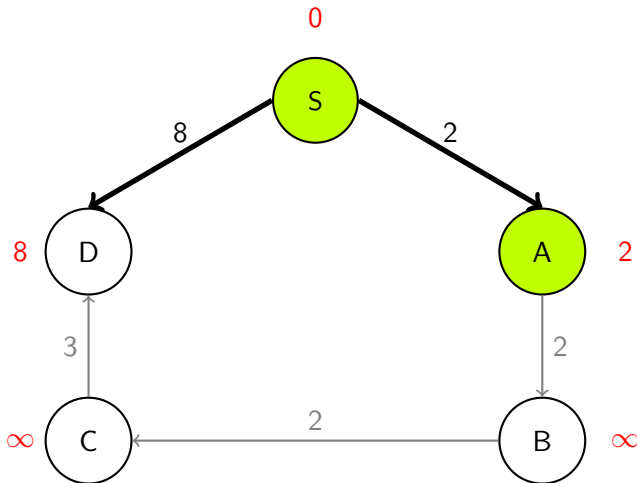
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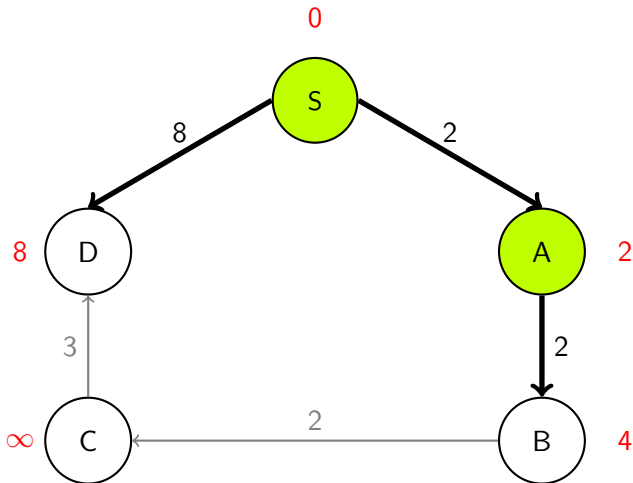
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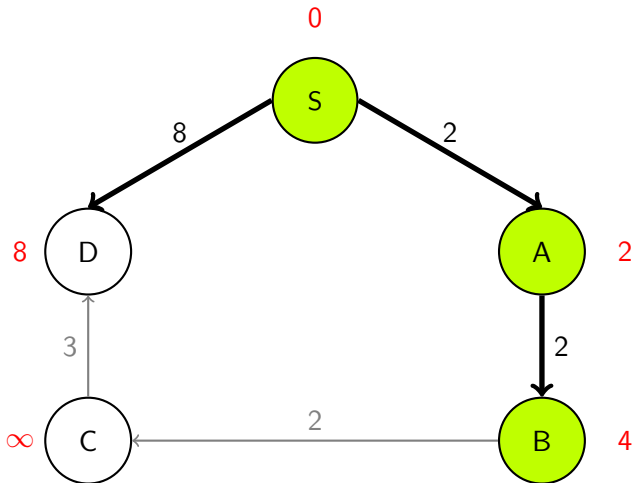
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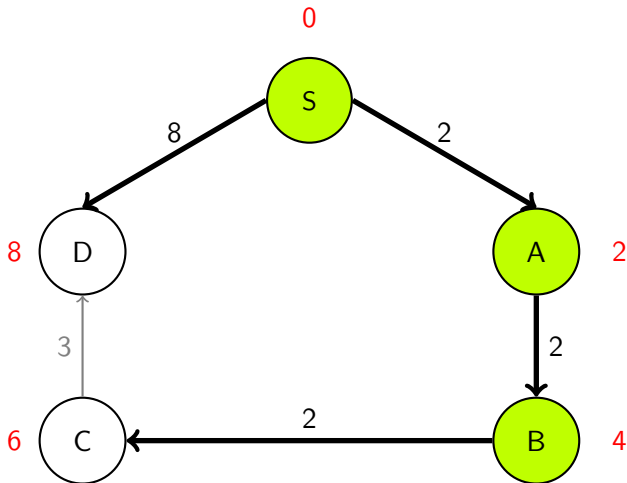
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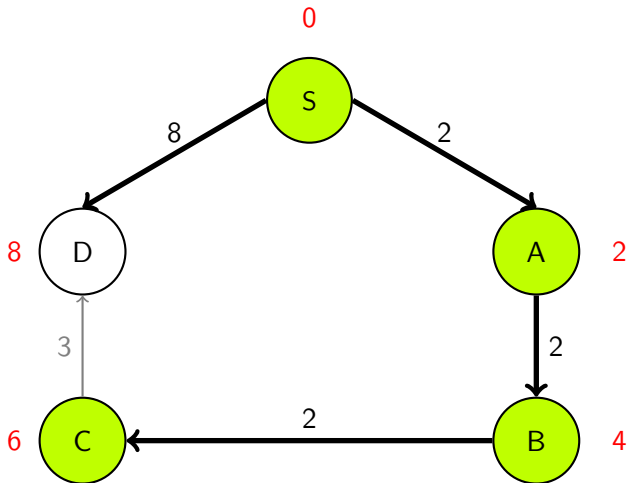
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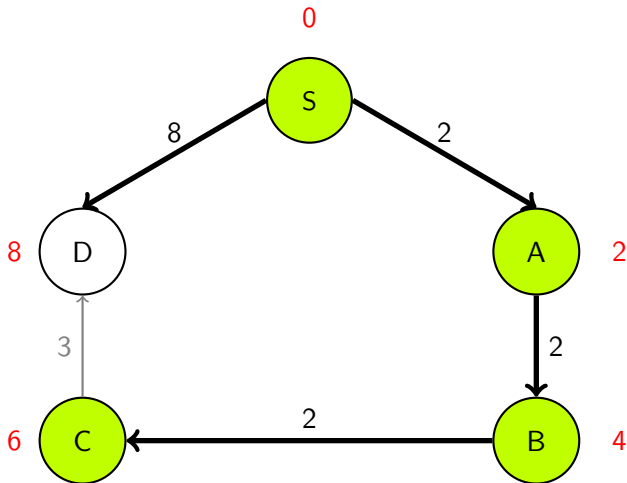
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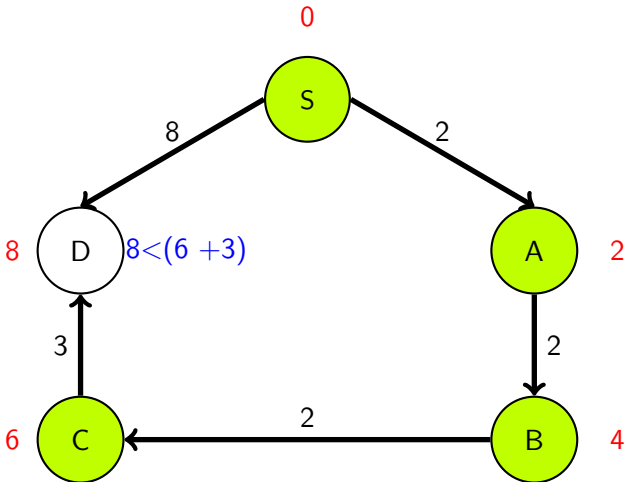
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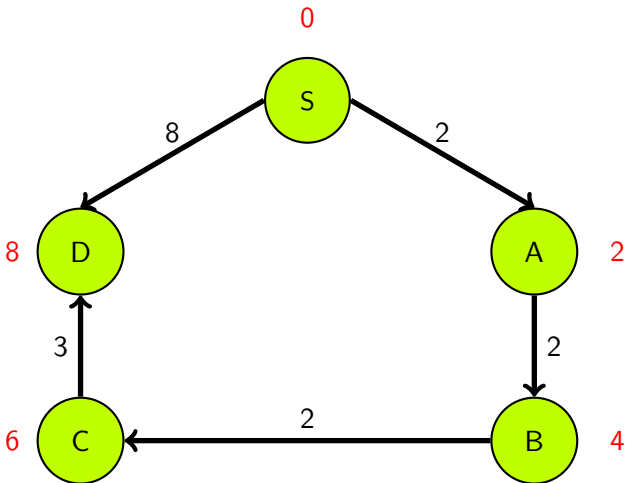
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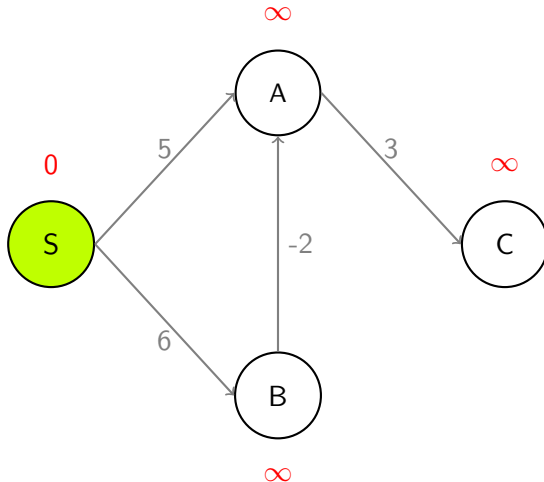


Fails in case of negative weight edge

If a graph contains negative weight edge(s), the algorithm may often produce wrong results.

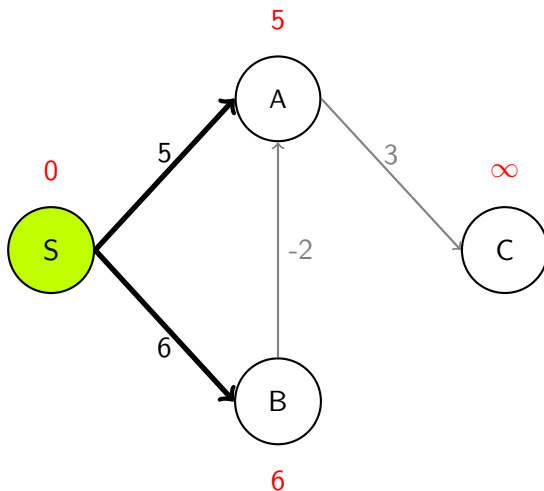
Negative Weight Issue

- Find the shortest distance from S to C



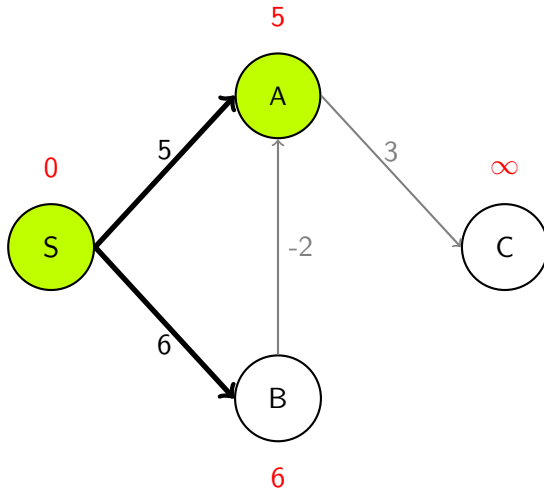
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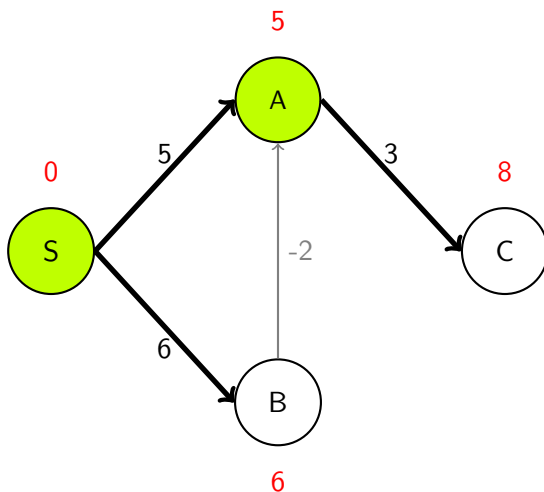
Negative Weight Issue

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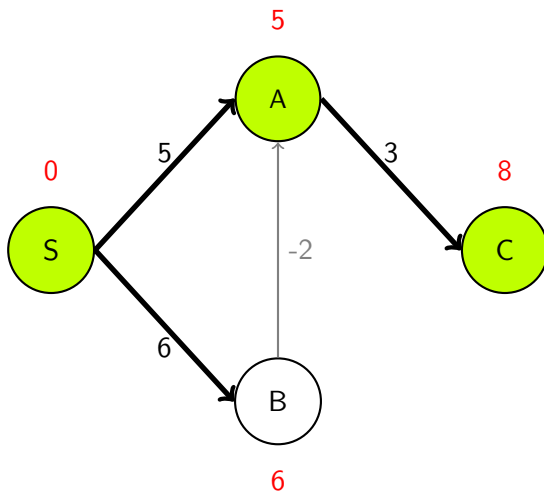
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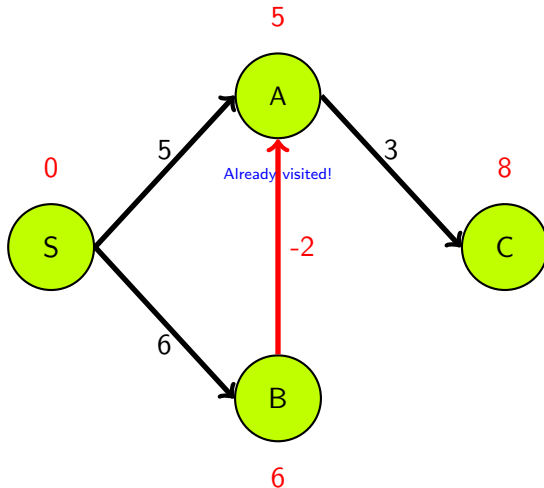
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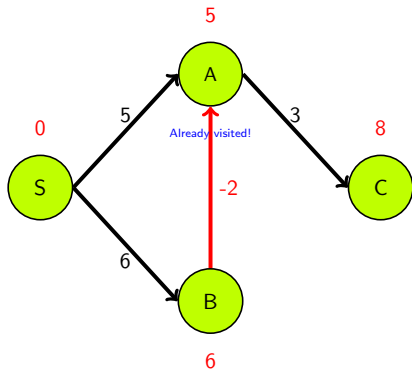


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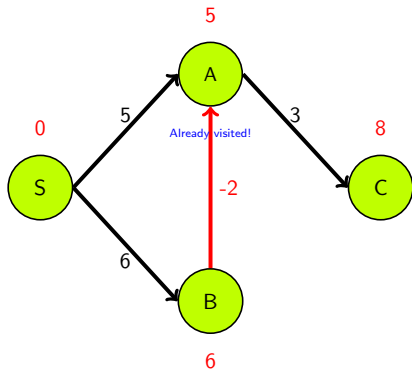


Negative Weight Issue



- Thus, the algorithm gives us the shortest distance of 8 by taking the path $S \rightarrow A \rightarrow C$

Negative Weight Issue



- However, taking the path $S \rightarrow B \rightarrow A$ would result in a shorter distance of $6 - 2 + 3 = 7$

Specialized variants

- When arc weights are small integers (bounded by a parameter C), specialized queues which take advantage of this fact can be used to speed up Dijkstra's algorithm.

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Dial's algorithm	Bucket queue	$O(E + V * C)$

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Ahuja <i>et al.</i>	Fibonacci Heap	$O(E + V \sqrt{\log C})$

- **Routing Protocols in Computer Networks:**

Dijkstra's algorithm is often used in routing protocols such as OSPF (Open Shortest Path First) and IS-IS (Intermediate System to Intermediate System)

- **GPS Navigation Systems:**

In GPS navigation, Dijkstra's algorithm is employed to find the shortest route between two locations, considering factors like distance, traffic conditions, and road types.

- **Game Development:**

In video games, especially those involving maps and navigation, Dijkstra's algorithm can be employed to create intelligent and efficient path-finding for characters or objects.

- **Resource Management in Operating Systems:**

Dijkstra's algorithm is applied in resource management to allocate resources optimally.

Thank You

References



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