# Design and Analysis of Algorithms Part IV: Graph Algorithms

**Lecture 12: Single-Source Shortest Paths Problem** 



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### Outline

- Review to Part IV
- Single-Source Shortest Paths Problem
- Dijkstra's Algorithm
  - The idea
  - The algorithm
  - Analysis of Dijkstra's algorithm
- The Bellman-Ford Algorithm
  - The algorithm
  - Analysis of Bellman-Ford algorithm

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## Introduction to Part IV

- In Part IV, we will illustrate several graph algorithm problems using several examples:
  - Basic Concepts of Graphs (图的基本概念)
  - Breadth-First Search [BFS] (广度优先搜索)
  - Depth-First Search [DFS] (深度优先搜索)
  - Topological Sort (拓扑排序)
  - Strongly Connected Components (强联通分量)
  - Minimum Spanning Trees (最小生成树)
  - Single-Source Shortest Paths (单源最短路径)
  - All-Pairs Shortest Paths (所有结点对的最短路径)
  - Maximum/Network Flows (最大流/网络流)

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# Single-Source Shortest Paths Problem



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#### Definition

The length of a path  $p = \langle v_0, v_1, ..., v_k \rangle$  is the sum of the weights of its constituent edges:

$$length(p) = \sum_{i=1}^{n} w(v_{i-1}, v_i).$$

#### Definition

The distance from u to v, denoted  $\delta(u, v)$ , is the length of the minimum length path if there is a path from u to v;

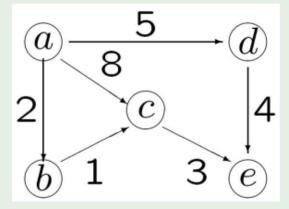
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#### Example

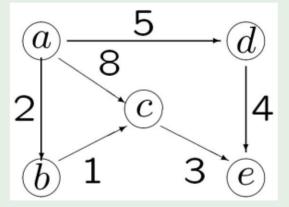


• length( $\langle a, b, c, e \rangle$ ) =

#### **Definition**

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#### Example



• length( $\langle a, b, c, e \rangle$ ) = 6; distance from a to e is 6

## Single-Source Shortest-Paths Problem

- Definition: Single-source shortest-paths problem
  - Given a digraph G = (V, E) with no-negative edge weights w and a designated source vertex  $s \in V$ , determine the distance and a shortest path from the source vertex to every vertex in the digraph.

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#### Question

How do you design an efficient algorithm for this problem?

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  - d[s] = 0 and d[v] = ∞ for all others vertices v.
- One by one we select vertices from V \ S to add to S.
- Questions to answer at each step:
  - Which vertex do we select?
  - How do we update the distance upper bounds after a vertex is added to S?

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- The next vertex processed is always a vertex u ∈ V \ S for which d[u] is minimum

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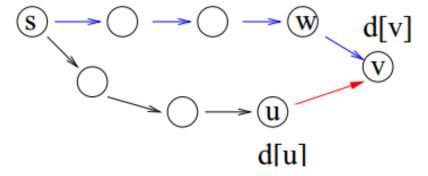
**Answer**: We use a greedy algorithm.

- For each vertex in  $u \in V \setminus S$ , we have computed a distance upper bound d[u].
- The next vertex processed is always a vertex u ∈ V \ S for which d[u] is minimum
  - that is, we take the unprocessed vertex that is closest (by our estimate) to s.

• Current distance upper bound for v: d[v].

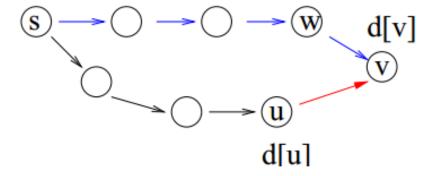
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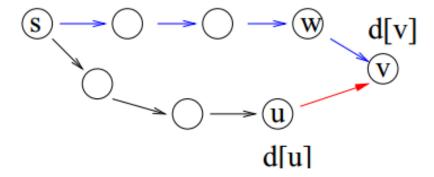
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- Current shortest path to v: <s, ..., w, v>, length d[v].
- New path to v: <s, ..., u, v>, length d[u] + w(u,v).

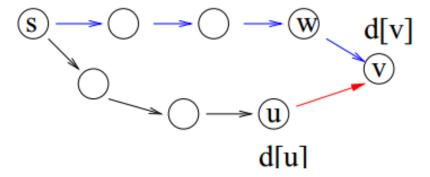
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- Current shortest path to v: <s, ..., w, v>, length d[v].
- New path to v: <s, ..., u, v>, length d[u] + w(u,v).
- If new path is shorter(d[u] + w(u,v) < d[v]), update d[v] d[v] = d[u] + w(u,v)

## **Updating Distance Estimates**

- Current distance upper bound for v: d[v].
- Vertex u just added to S. Edge (u, v) with weight w(u, v).
- How do we update d[v]?



- Current shortest path to v: <s, ..., w, v>, length d[v].
- New path to v: <s, ..., u, v>, length d[u] + w(u,v).
- If new path is shorter(d[u] + w(u,v) < d[v]), update d[v] d[v] = d[u] + w(u,v)

Now we have a better (tighter) upper bound for d[v]. This is called relaxing the edge (u, v).

```
Input: Update estimation of u according to distance of v
Output: None
if d[u] + w(u, v) < d[v] then
end
```

```
Input: Update estimation of u according to distance of v
Output: None
if d[u] + w(u, v) < d[v] then
d[v] \leftarrow
end
```

```
Input: Update estimation of u according to distance of v

Output: None

if d[u] + w(u, v) < d[v] then

d[v] \leftarrow d[u] + w(u, v);
pred[v] \leftarrow
end
```

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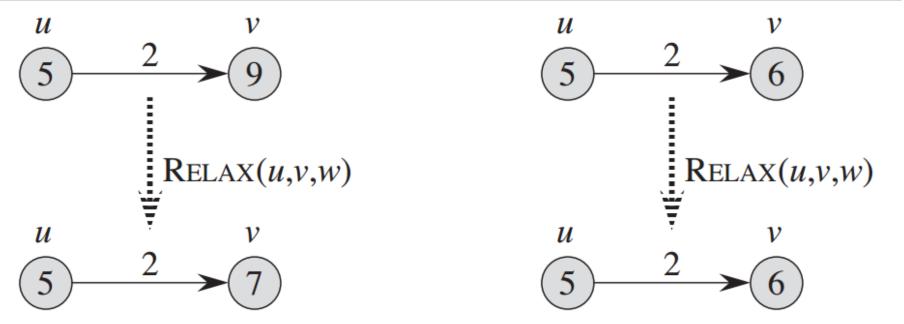
if d[u] + w(u, v) < d[v] then
d[v] \leftarrow d[u] + w(u, v);
pred[v] \leftarrow u;
end
```

# The Algorithm for Relaxing an Edge: An Example

```
Input: Update estimation of \boldsymbol{u} according to distance of \boldsymbol{v}
Output: None

if d[u] + w(u, v) < d[v] then

d[v] \leftarrow d[u] + w(u, v);
pred[v] \leftarrow u;
end
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Input: Update estimation of u according to distance of v
Output: None

if d[u] + w(u, v) < d[v] then
d[v] \leftarrow d[u] + w(u, v);
pred[v] \leftarrow u;
end
```

- Remark 1: The predecessor pointer pred[] is for determining the shortest paths.
- Remark 2: After edge (u, v) is relaxed, we have  $d[v] \le d[u] + w(u, v)$

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 Note: if we implement the priority queue using a heap, we can perform the operations Insert(), Extract-Min(), and Decrease-Key(), each in O( ) time.

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 Note: if we implement the priority queue using a heap, we can perform the operations Insert(), Extract-Min(), and Decrease-Key(), each in O(log n) time.

Dijkstra(G,w,s)

**Input:** A graph G, a matrix w representing the weights between vertices in G, source vertex s

Output: None for  $u \in V$  do  $| d[u] \leftarrow$ 

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 $\begin{array}{ll} \textbf{Output:} \ \text{None} \\ \textbf{for} \ u \in V \ \textbf{do} \\ | \ d[u] \leftarrow \infty, color[u] \leftarrow \end{array}$ 

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Input: A graph G, a matrix w representing the weights between vertices
         in G, source vertex s
Output: None
for u \in V do
    d[u] \leftarrow \infty, color[u] \leftarrow \text{WHITE}; // \text{Initialize}
end
d[s] \leftarrow
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    Q \leftarrow queue with all vertices;
    while Non-Empty(Q) do
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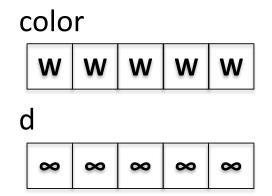
Dijkstra(G, w, s)Input: A graph G, a matrix w representing the weights between vertices

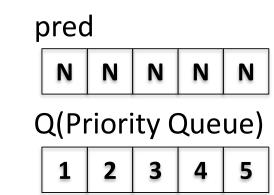
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            pred[v] \leftarrow u;
         end
    end
    color[u] \leftarrow
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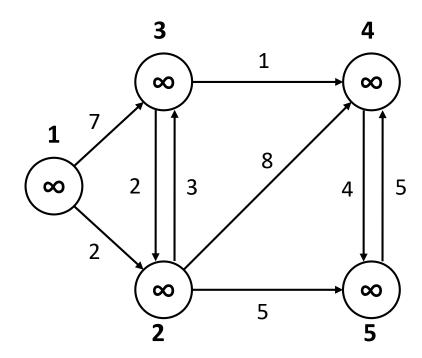
 $color[u] \leftarrow BLACK;$ 

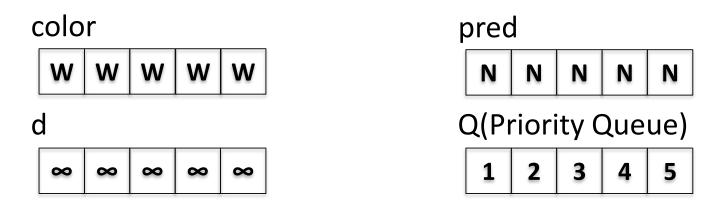
end

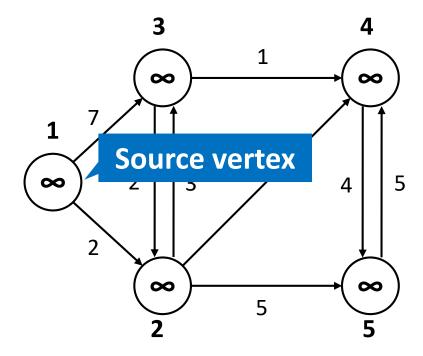
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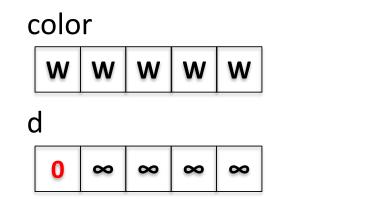


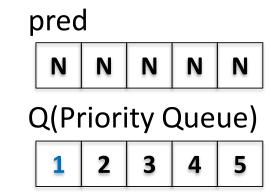


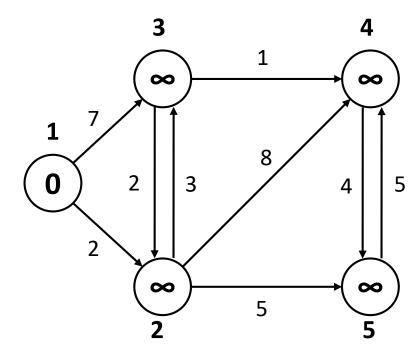




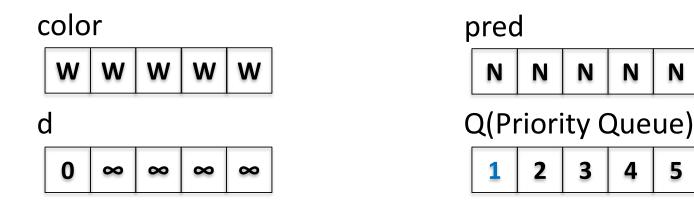


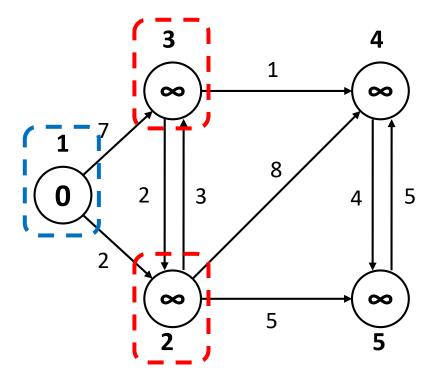


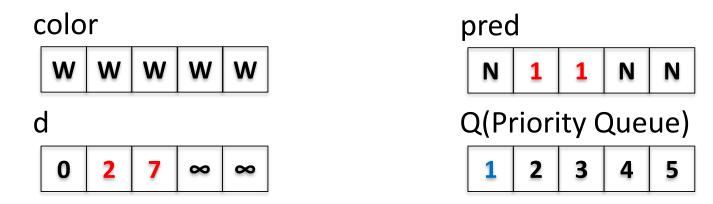


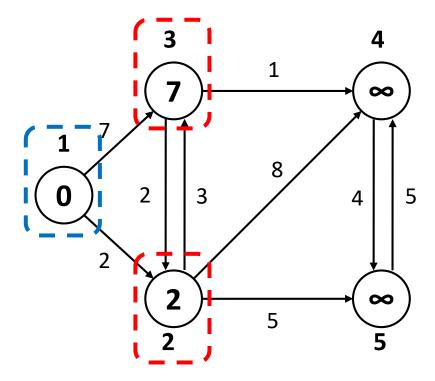


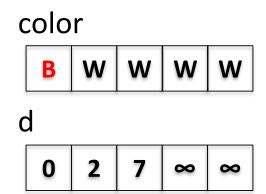
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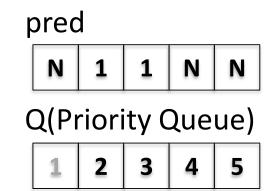


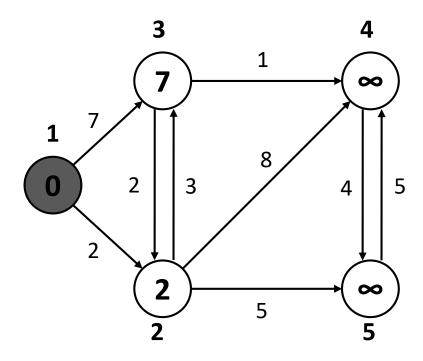


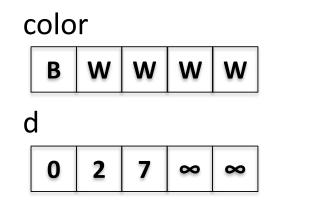


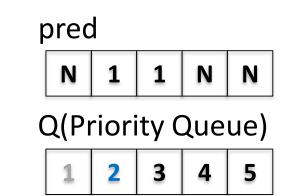


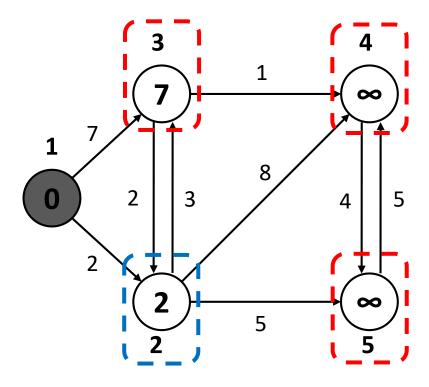


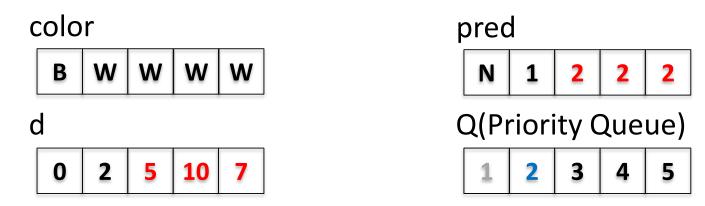


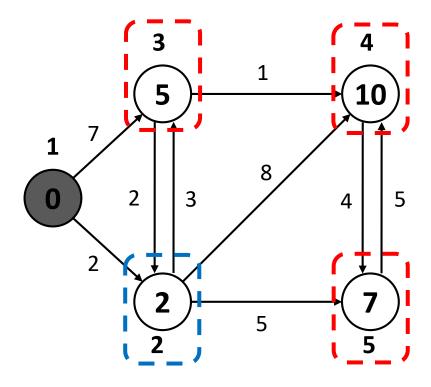


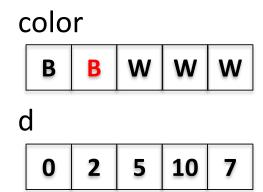


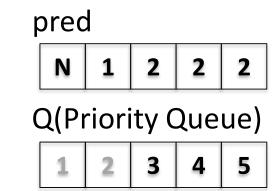


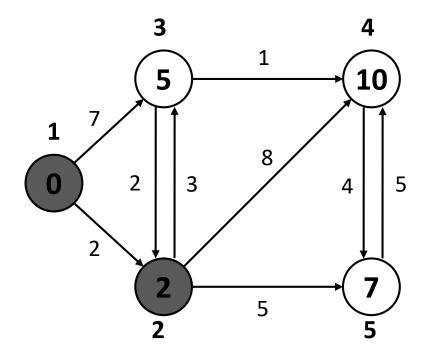


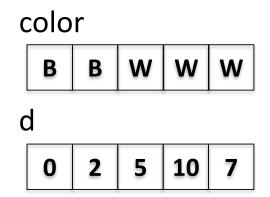


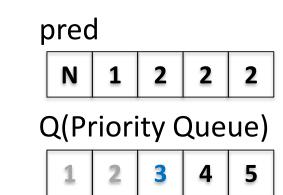


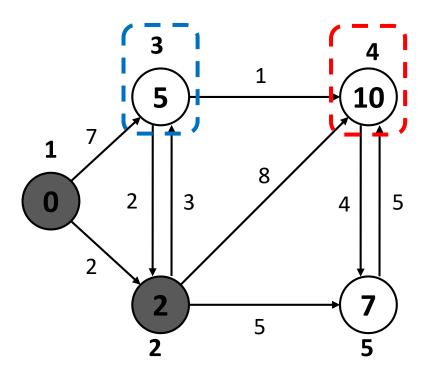


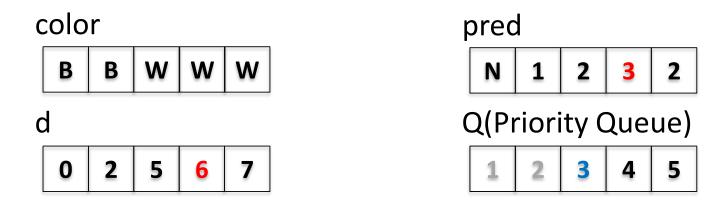


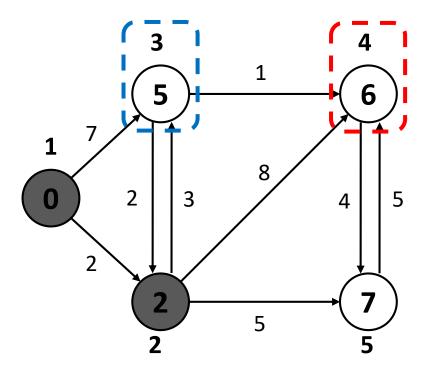


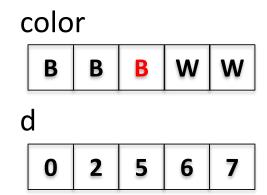


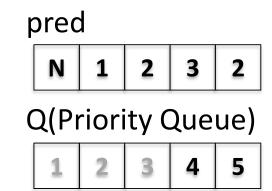


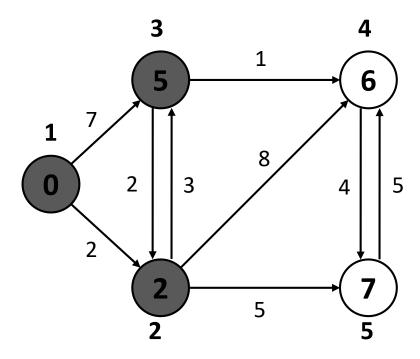


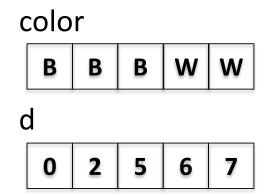


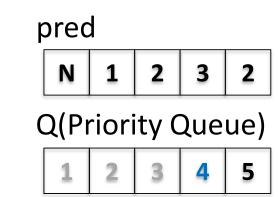


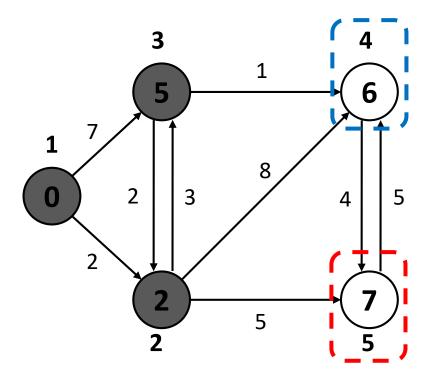


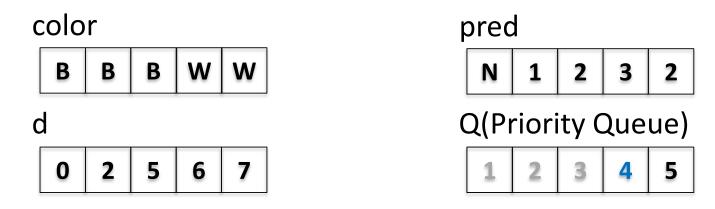


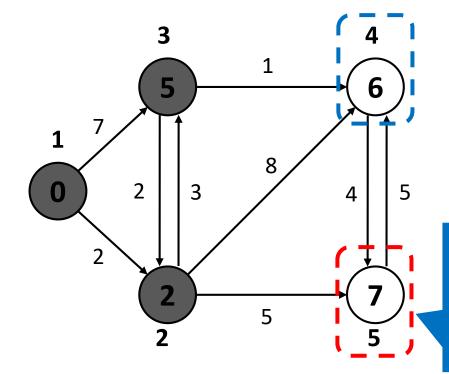




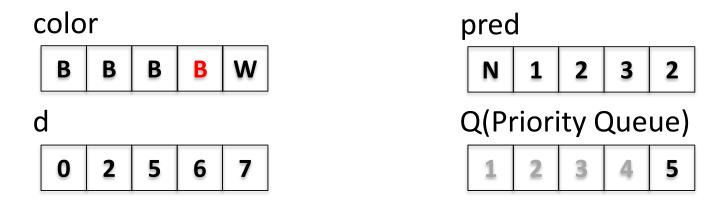


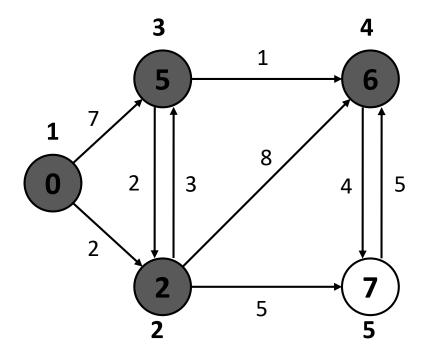


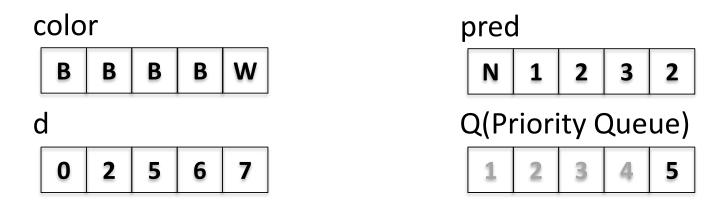


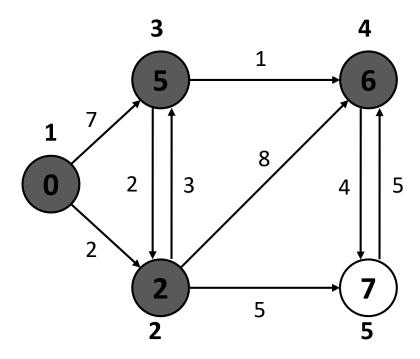


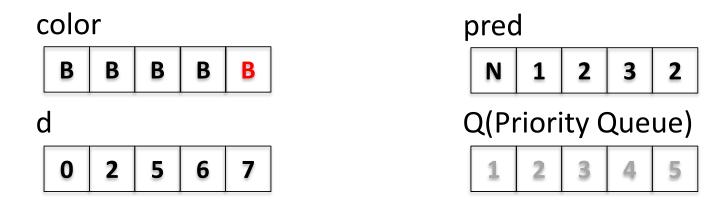
d[5] is not needed to be updated because d[4]+4 > d[5]

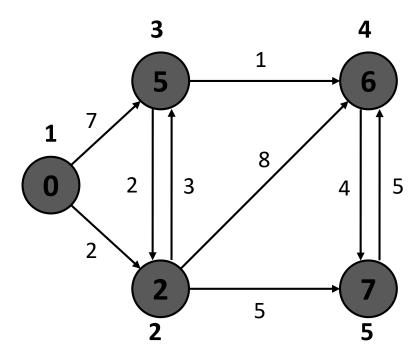


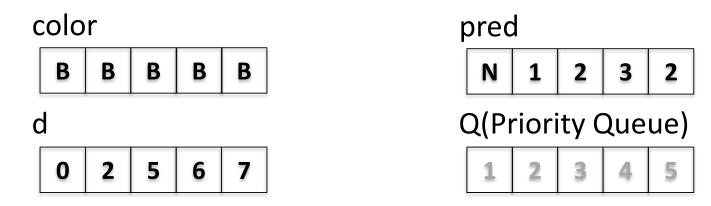


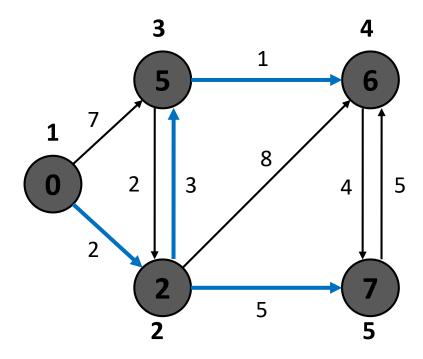












# Shortest Path Tree for Dijkstra's Algorithm

```
Shortest Path Tree: T = (V ; A), where A = \{(pred[v], v) | v \in V \setminus \{s\}\}
```

## Shortest Path Tree for Dijkstra's Algorithm

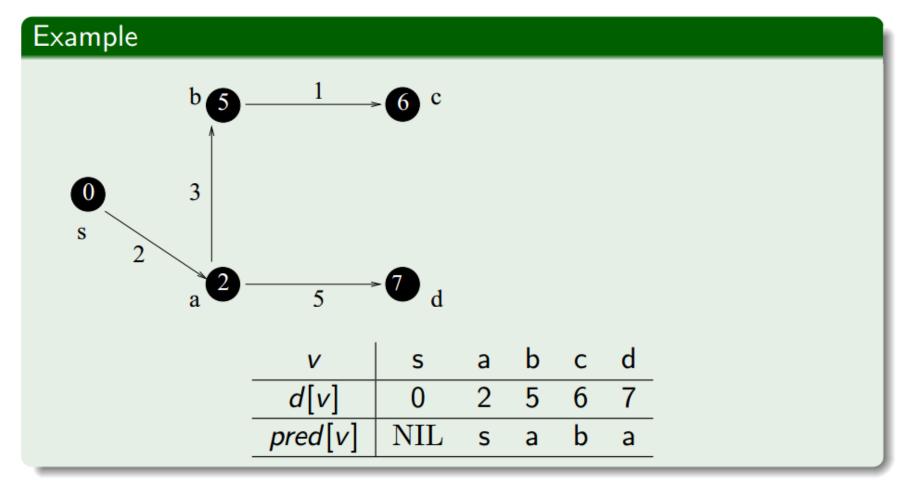
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### Outline

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  - The algorithm
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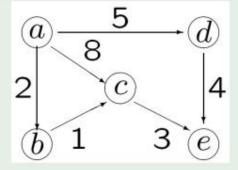
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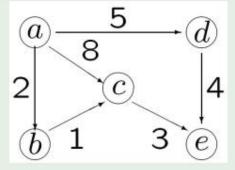


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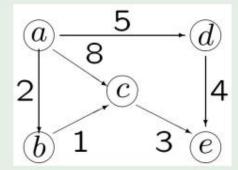


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### Question

Why?

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- Consider the true shortest path from s to u: <s,..., x, y,..., u>.
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Thus y should have been added to S before u, in contradiction to our assumption that u is the next vertex to be added to S.

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time.

## Description of Dijkstra's Algorithm

```
Dijkstra(G,w,s)
     Input: A graph G, a matrix w representing the weights between vertices
                in G, source vertex s
     Output: None
     for u \in V do O(|V|)

|d[u] \leftarrow \infty, color[u] \leftarrow \text{Willie}, // \text{Initialize}
     end
     d[s] \leftarrow 0;
     pred[s] \leftarrow \text{NULL};
     Q \leftarrow queue with all vertices;
     while Non-Empty(Q) do
         // Process all vertices
         u \leftarrow \text{Extract-Min}(Q); // \text{ Find new vertex}
         for v \in Adj[u] do
              if d[u] + w(u, v) < d[v] then
                 // If estimate improves
                  d[v] \leftarrow d[u] + w(u, v); // \text{ relax}
Decrease-Key(Q, v, d[v]); O(\log |V|) \cdot |E|
                  pred[v] \leftarrow u;
              end
         end
         color[u] \leftarrow BLACK;
     end
```

# Prim's Algorithm vs. Dijkstra's Algorithm

Dijkstra's algorithm looks similar to Prim's algorithm.

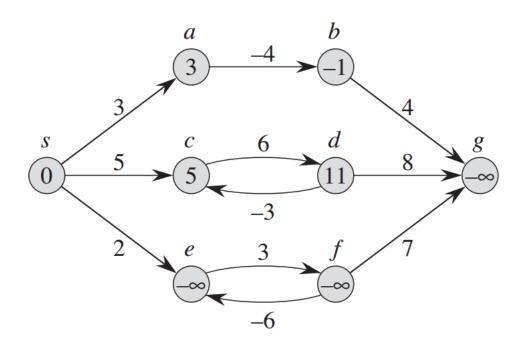
 To understand the differences clearly, try them both on some example.

#### Outline

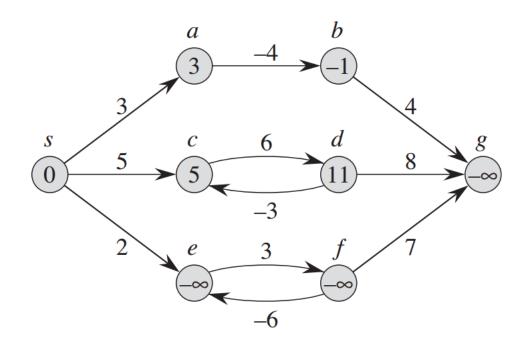
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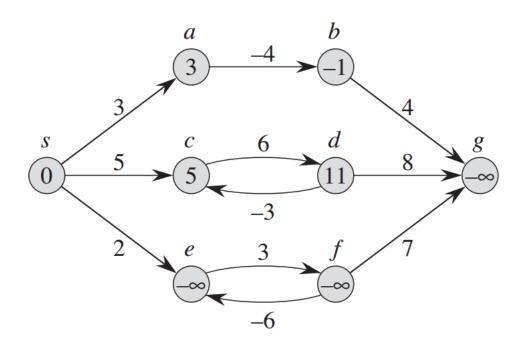
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- By following the proposed "shortest" path and then traversing the negative-weight cycle, we can always find a path with lower weight.



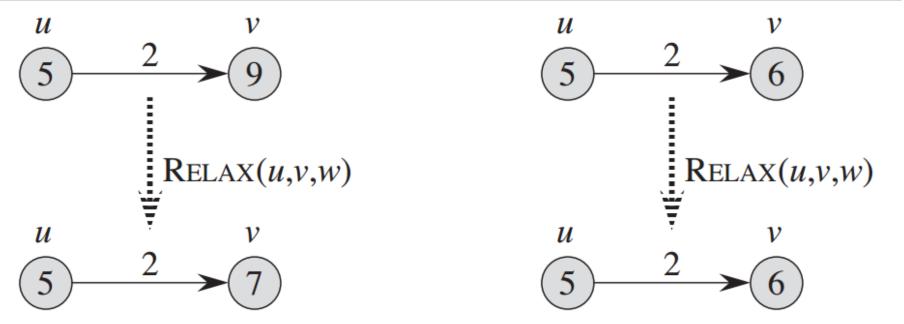
## Review of The Algorithm for Relaxing an Edge

#### Relax(u,v)

```
Input: Update estimation of u according to distance of v
Output: None

if d[u] + w(u, v) < d[v] then

d[v] \leftarrow d[u] + w(u, v);
pred[v] \leftarrow u;
end
```



• The algorithm relaxes edges, progressively decreasing an estimate v.d on the weight of a shortest path from the source s to each vertex  $v \in V$  until it achieves the actual shortest-path weight  $\delta(s,v)$ . Bellman-Ford(G,w,s)

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for  $u \in V$  do  $| d[u] \leftarrow \infty, pred[u] \leftarrow \text{NIL}; // \text{Initialize}$ end

for  $i \leftarrow 1$  to |V| - 1 do  $| \text{for } e \in E \text{ do}$  | RELAX(u, v, w); | endend

#### The Bellman-Ford Algorithm

• The algorithm relaxes edges, progressively decreasing an estimate v.d on the weight of a shortest path from the source s to each vertex  $v \in V$  until it achieves the actual shortest-path weight  $\delta(s,v)$ .

Bellman-Ford(G, w, s)

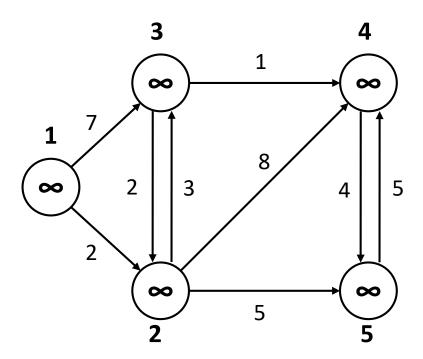
```
Input: A directed graph G, weights w, and the source vertex s
Output: Return FALSE if G contains negative cycle, return TRUE if
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for u \in V do
  d[u] \leftarrow \infty, pred[u] \leftarrow \text{NIL}; // \text{Initialize}
end
for i \leftarrow 1 to |V| - 1 do
   for e \in E do
     RELAX(u, v, w);
   end
end
for e \in E do
   if d[v] > d[u] + w(u, v) then
    return FALSE;
   end
end
return
```

• The algorithm relaxes edges, progressively decreasing an estimate v.d on the weight of a shortest path from the source s to each vertex  $v \in V$  until it achieves the actual shortest-path weight  $\delta(s,v)$ . Bellman-Ford(G,w,s)

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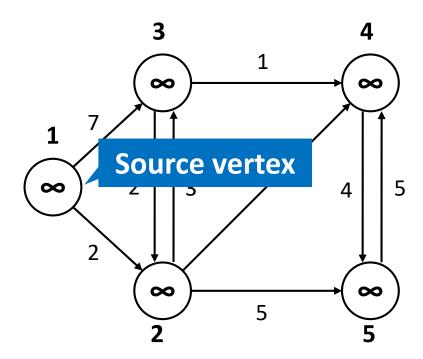
#### **Initialization**





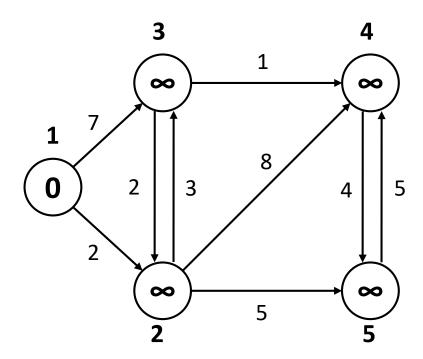
#### **Initialization**





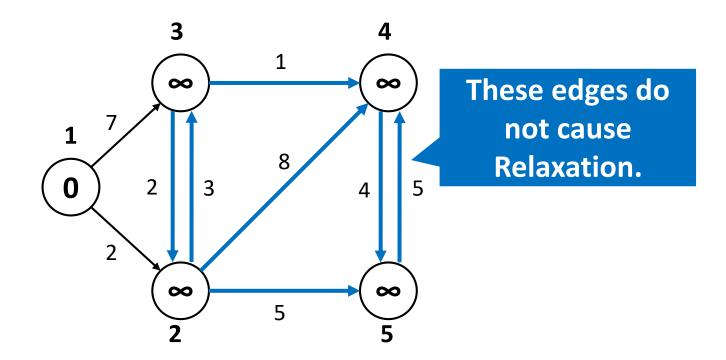
#### **Initialization**





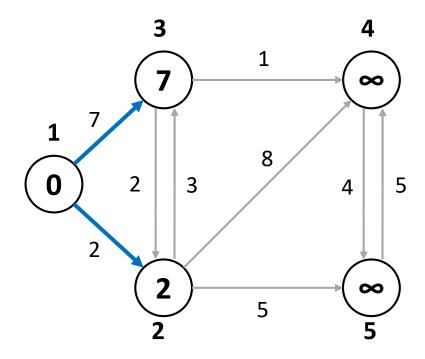
#### 1st round



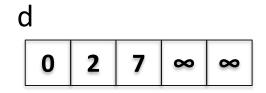


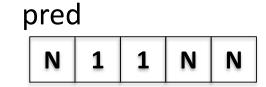
#### 1st round

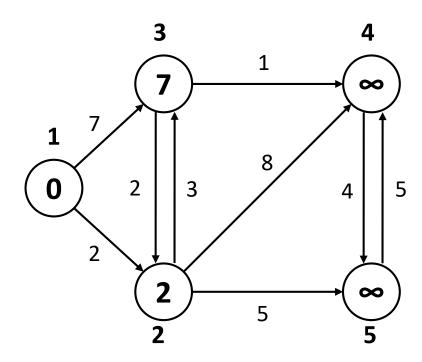


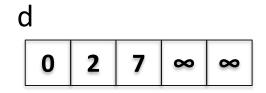


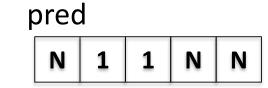
#### 2<sup>nd</sup> round

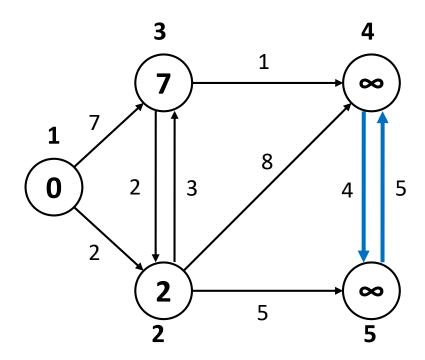


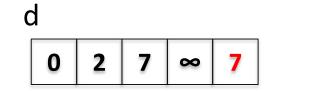


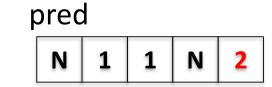


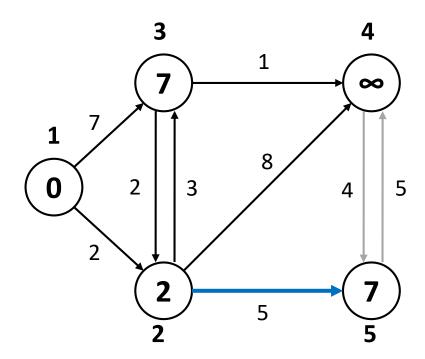


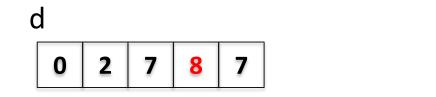


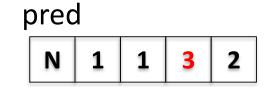


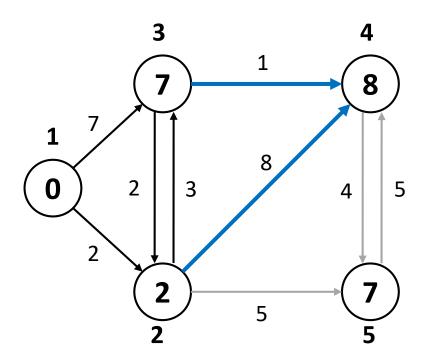






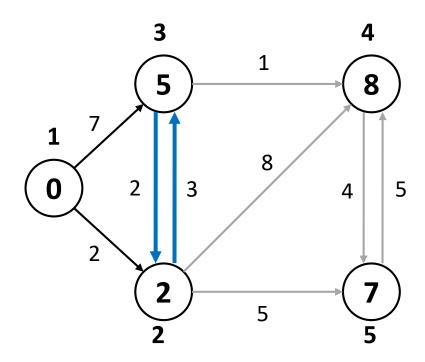


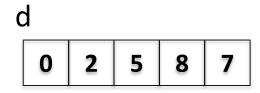


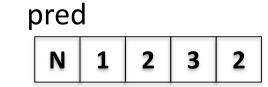


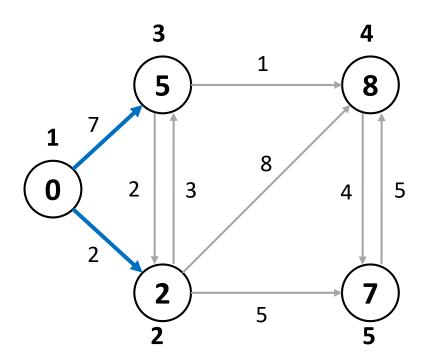


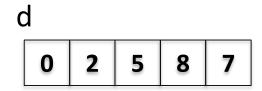


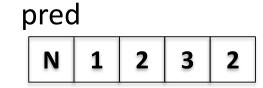


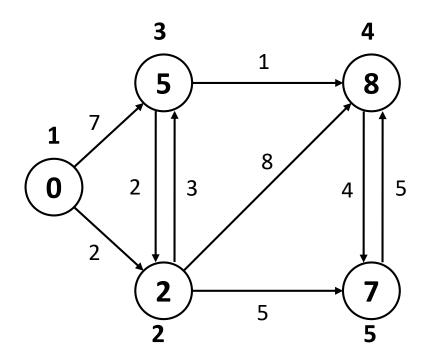


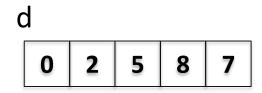


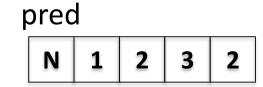


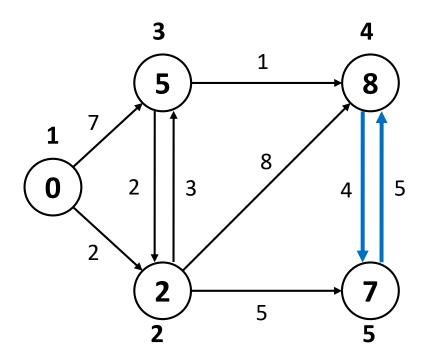


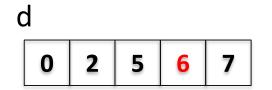


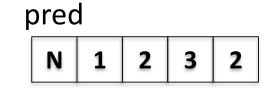


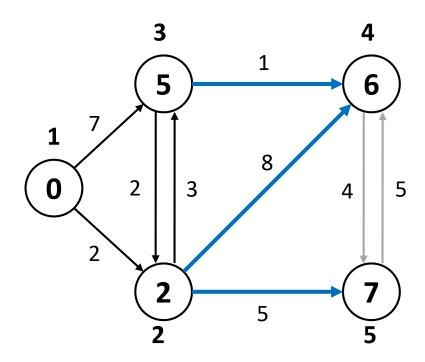


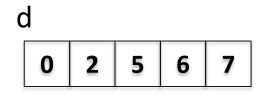


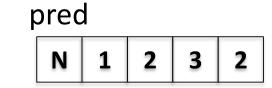


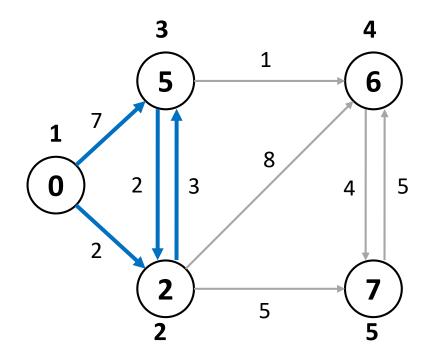




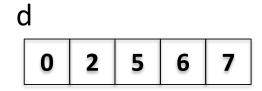


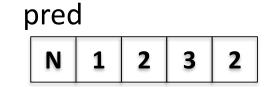


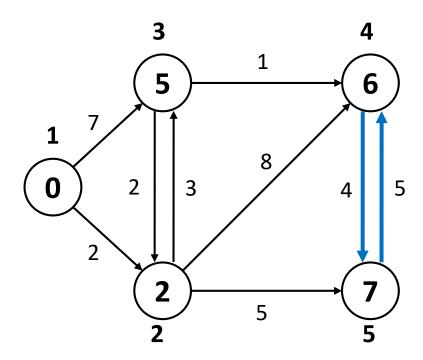




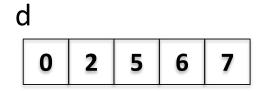
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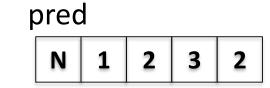


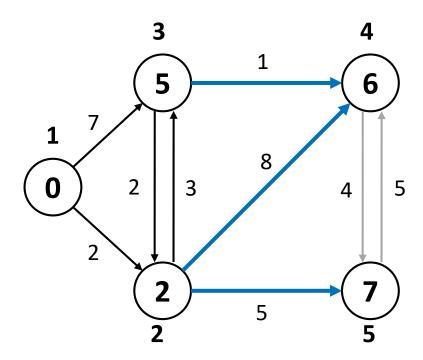




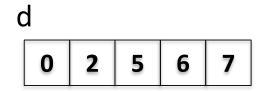
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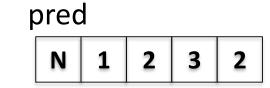


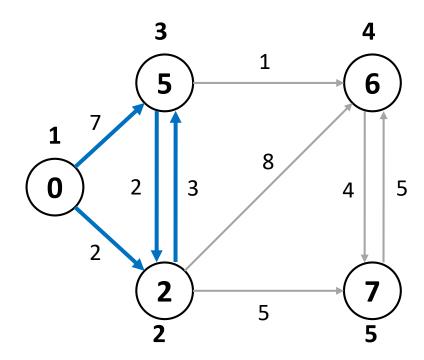




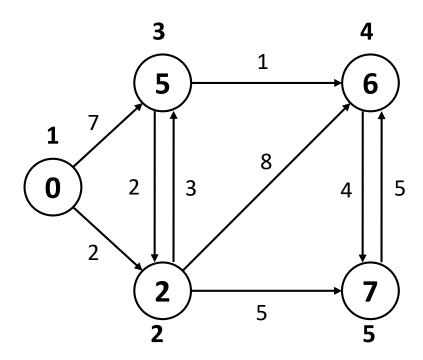
#### 4th round











### Outline

- Review to Part IV
- Single-Source Shortest Paths Problem
- Dijkstra's Algorithm
  - The idea
  - The algorithm
  - Analysis of Dijkstra's algorithm
- The Bellman-Ford Algorithm
  - The algorithm
  - Analysis of Bellman-Ford algorithm

### Analysis of Bellman-Ford Algorithm

• The Bellman-Ford algorithm runs in time  $O(|V| \cdot |E|)$  since the initialization takes O(|V|) time, each of the |V| - 1 passes over the edges takes O(|E|) time, and the **for** loop takes O(|E|) time.

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dank u Tack ju faleminderit Asante ipi Tak mulţumesc

Salamat! Gracias
Terima kasih Aliquam

Merci Dankie Obrigado
köszönöm Grazie

Aliquam Go raibh maith agat
děkuji Thank you

gam