

OME Algorithms

Single-source shortest paths

input: graph G , source vertex s ▷

```
1  $d[s] \leftarrow 0$ 
2  $\text{PARENT}[s] \leftarrow \text{None}$ 
3 for  $v \leftarrow V - \{s\}$ :
4    $d[v] \leftarrow \infty$ 
5    $\text{PARENT}[v] \leftarrow \text{None}$ 
6 end
```

Algorithm 1: Relaxation Technique Initialisation
 $O(n) - n$ vertices

input: from vertex u , to vertex v , weights w ▷

```
1 if  $d[v] > d[u] + w(u, v)$  then
2    $d[v] \leftarrow d[u] + w(u, v)$ 
3    $\text{PARENT}[v] = u$ 
4 end
```

Algorithm 2: Relaxation Technique Relax Edge
 $O(1)$ except if we perform additional steps

input: graph G , weights w , source vertex s
output: cycle present c

```
1 relaxation_init( $G, s$ ) ▷ Algorithm 1
2  $(V, E) \leftarrow G$ 
3  $n \leftarrow |V|$ 
4 for  $i \leftarrow 1$  to  $(n - 1)$  do ▷ nodes indexed from 1,  $s$  will already be correct
5   for  $u, v \leftarrow E$  do ▷ consider in any arbitrary order
6     relaxation_relax( $u, v, w$ ) ▷ Algorithm 2
7   end
8 end
9  $c \leftarrow \perp$ 
10 for  $u, v \leftarrow E$  do
11   if  $d[v] > d[u] + w(u, v)$  then
12      $c \leftarrow \top$  ▷ negative cycle reachable from  $s$ 
13   end
14 end
```

Algorithm 3: Bellman-Ford Algorithm
 $\Theta(nm) - n$ vertices, m edges

```

input: graph  $G$ , weights  $w$ , source vertex  $s$ 
output: cycle present  $c$ 
1 relaxation_init( $G, s$ )
2  $(V, E, \text{Adj}) \leftarrow G$ 
3  $Q \leftarrow \text{Queue}()$ 
4  $Q \leftarrow s$ 
5  $c \leftarrow \perp$ 
6 while not  $Q.\text{empty}$  do
7    $u \leftarrow Q.\text{dequeue}()$ 
8   for  $v \leftarrow \text{Adj}[u]$  do
9     relaxation_relax( $u, v, w$ )
10    if [edge was relaxed] then
11      if  $v$  not in  $Q$  then
12         $Q \leftarrow v$ 
13      end
14    end
15  end
16  if [negative cycle present] do
17     $c \leftarrow \top$ 
18    break
19  end
20 end

```

▷ Algorithm 1

▷ create a new empty queue

▷ Algorithm 2

▷ algorithm has been omitted

Algorithm 4: Bellman-Ford Algorithm with FIFO Queue
 $O(nm)$ — n vertices, m edges

input: graph G , weights w , source vertex s
assumption: all edge weights are non-negative

```

1   $(V, E, \text{Adj}) \leftarrow G$ 
2  relaxation_init( $G, s$ ) ▷ Algorithm 1
3   $S \leftarrow \emptyset$  ▷ keep track of nodes that we know  $d[v]$  is  $\delta(s, v)$ 
4   $Q \leftarrow \text{PriorityQueue}(V)$  ▷ write all nodes excluding  $s$  into a priority queue
5  while not  $Q.\text{empty}$  do
6       $u \leftarrow Q.\text{extract\_min}()$  ▷ find  $u$  with minimum  $d[\cdot]$  value in  $Q$  and remove it from the queue
7       $S \leftarrow S \cup \{u\}$  ▷ mark this vertex as visited
8      for  $v \leftarrow \text{Adj}[u]$  do
9          relaxation_relax( $u, v, w$ ) ▷ Algorithm 2
10         if [ $d[v]$  has reduced] then for modified Dijkstra's algorithm
11              $Q \leftarrow Q \cup v$ 
12              $S \leftarrow S - \{u\}$  ▷ mark this vertex as unvisited
13         end
14     end
15 end

```

Algorithm 5: Dijkstra's Shortest-paths Algorithm

$O(\min\{n^2, m \log n\})$ — n vertices, m edges

Unordered List: $O(n^2)$ — n vertices

Ordered List: $O(mn)$ — n vertices, m edges

Heap: $O(m \log n)$ — n vertices, m edges

Directed acyclic graph shortest paths

input: graph G , weights w , source vertex s

```
1 (Adj)  $\leftarrow G$ 
2  $V \leftarrow \text{topological\_sort}(G)$ 
3  $\text{relaxation\_init}(G, s)$ 
4 for  $u \leftarrow V$  do
5   for  $v \leftarrow \text{Adj}[u]$  do
6      $\text{relaxation\_relax}(u, v, w)$ 
7   end
8 end
```

Algorithm 6: Directed Acyclic Graph Shortest Paths Algorithm

$O(n + m)$ — n vertices, m edges

All-pairs shortest paths

If all edge weights are non-negative, we repeatedly use Dijkstra's algorithm. No method with better (worst-case) running time is known.

For negative edge weights, we can use:

- Bellman-Ford, $O(n^2m)$, $\Theta(n^4)$ if $m = \Theta(n^2)$ (Algorithm 3)
- Floyd-Warshall algorithm, $\Theta(n^3)$
- Johnson's algorithm, $O(nm \log n)$

```
input: graph  $G$ , weights  $w$ 
output: shortest paths  $M$ 
1   $(V, E) \leftarrow G$ 
2   $G' \leftarrow \text{copy}(G)$                                 ▷ create a clone of the graph
3   $(V', E') \leftarrow G'$ 
4   $V' \leftarrow V \mid \{s\}$                                 ▷ create a new node  $s$ 
5   $E' \leftarrow E \mid \{(s, v) \mid v \in V\}$                 ▷ create new edge from  $s$  to  $v$ 
6  for  $v \leftarrow V$  do
7       $w(s, v) \leftarrow 0$                                 ▷ set the edge weights to 0
8  end
9   $c \leftarrow \text{bellman\_ford}(G', w, s)$                     ▷ Algorithm 3
10 if  $c$  then
11     terminate                                          ▷ there is a negative cycle so we do not continue
12 end
13 for  $v \leftarrow V$  do
14      $h[v] \leftarrow \delta(s, v)$                             ▷ copy computed  $\delta(s, v)$  from Bellman Ford algorithm to our distance array
15 end
16  $w' \leftarrow w$                                           ▷ clone the weights
17 for  $u, v \leftarrow E$  do
18      $w'(u, v) = w(u, v) + h[u] - h[v]$ 
19 end
20  $M \leftarrow ()$                                           ▷ empty matrix
21 for  $u \leftarrow V$  do
22      $\text{dijkstra}(G, w', u)$                                 ▷ Algorithm 5, compute  $\delta(u, v)$  for all  $v \in V$ 
23     for  $v \leftarrow V$  do
24          $M_{uv} \leftarrow \delta(u, v) - [h[u] - h[v]]$ 
25     end
26 end
```

Algorithm 7: Johnson's Algorithm

$$\underbrace{O(\min\{n^3, nm \log n\})}_{\text{bellman-ford}} - \underbrace{n \text{ vertices}, m \text{ vertices}}_{n \text{ dijkstra's}} + \underbrace{O(n^2)}_{\text{other computation}}$$

Maximum-flow problem

input: graph $G(V, E, c, s, t)$
output: maximum flow f

```
1  $f \leftarrow$  zero flow ▷  $f(u, v) = 0$  for each  $(u, v)$  in  $E$ 
2 while  $\top$  do
3    $G_f \leftarrow$  [construct residual network]
4    $P \leftarrow$  [find augmenting path in  $G_f$  using depth-first search]
5   if  $P$  then
6      $f \leftarrow f_p^f$  ▷ path augmentation
7   else
8     break
9   end
10 end
```

Algorithm 8: Ford-Fulkerson Method (DFS)

input: graph $G(V, E, l, u, s, t)$
output: maximum flow f

```
1  $f \leftarrow$  zero flow ▷  $f(u, v) = 0$  for each  $(u, v)$  in  $E$ 
2 while  $\top$  do
3    $G_f \leftarrow$  [construct residual network]
4    $P \leftarrow$  [find shortest augmenting path in  $G_f$  using breadth-first search]
5   if  $P$  then
6      $f \leftarrow f_p^f$  ▷ path augmentation
7   else
8     break
9   end
10 end
```

Algorithm 9: Edmonds-Karp Method (BFS)