

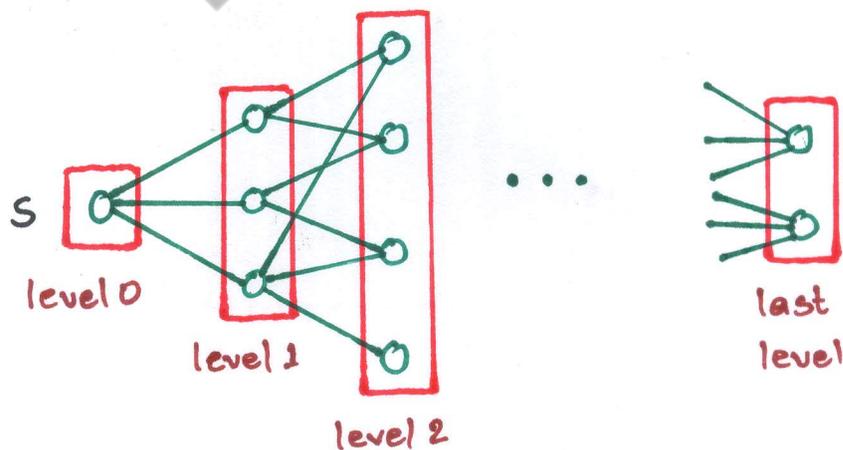
Week 9: Lecture Notes

Topics: BFS and DFS
Shortest path problem
Dijkstra's Algorithm
Example of Dijkstra
Bellman Ford

Breadth-First Search

Explore graph level by level from s (start vertex)

- level 0 = $\{s\}$
- level i = vertices reachable by path of i edges but not fewer.
- build level $i > 0$ from $i-1$ by trying all outgoing edges, but ignoring vertices from previous levels

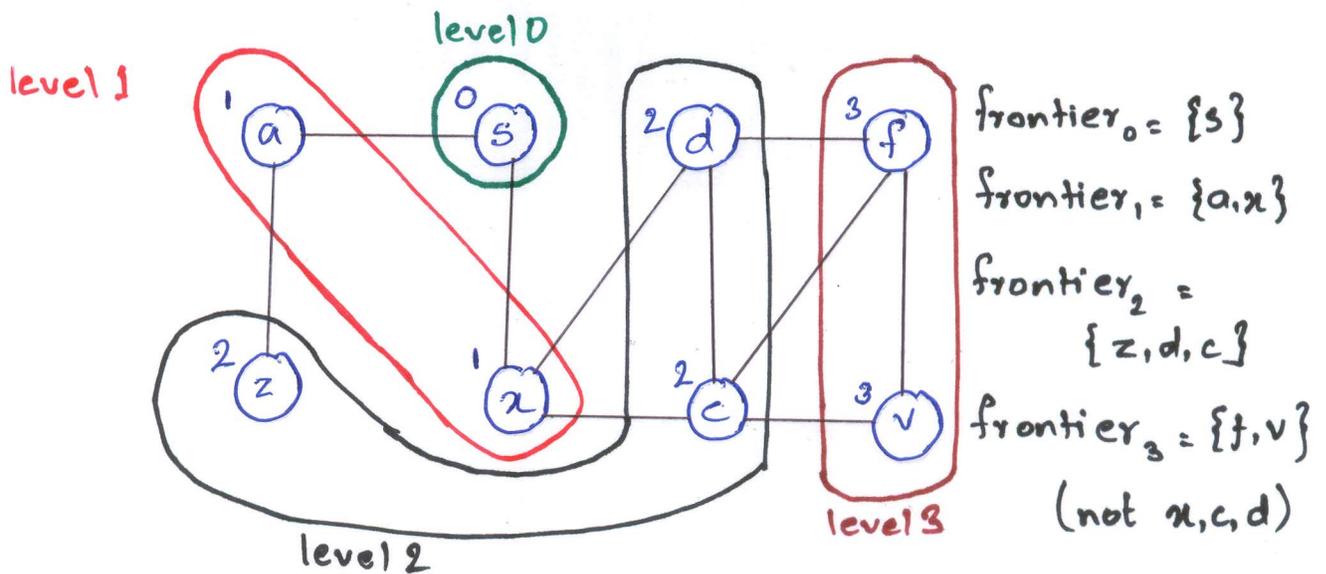


Breadth-First-Search Algorithm

BFS(v, Adj, s):

1. level = {s: 0}
2. parent = {s: None}
3. i = 1
4. frontier = [s]
5. while frontier
6. next = []
7. for u in frontier:
8. for v in Adj[u]:
9. if u not in level:
10. level[v] = i
11. parent[v] = u
12. next.append(v)
13. frontier = next
14. i = i + 1

Example



Analysis

- vertex v enters next (and then frontier) only once (because level[v] then set)

base case: $v = s$

- \Rightarrow Adj[v] looped through only once.

$$\text{time} = \sum_{v \in V} |\text{Adj}[v]| = \begin{cases} |E| & \text{for directed graphs} \\ 2|E| & \text{for undirected graphs} \end{cases}$$

- $\Rightarrow O(E)$ time

- $O(V+E)$ ("linear time") to also list vertices unreachable from v (those still not assigned level)

Shortest Paths

- for every vertex v , fewest edges to get from s to v is
$$\begin{cases} \text{level}[v] & \text{if } v \text{ is assigned level} \\ \infty & \text{else (no path)} \end{cases}$$

- parent pointers from shortest-path tree
= union of such shortest path for each v .

\Rightarrow to find shortest path, take v , parent[v], parent[parent[v]], etc., until s (or None)

Depth - First Search (DFS)

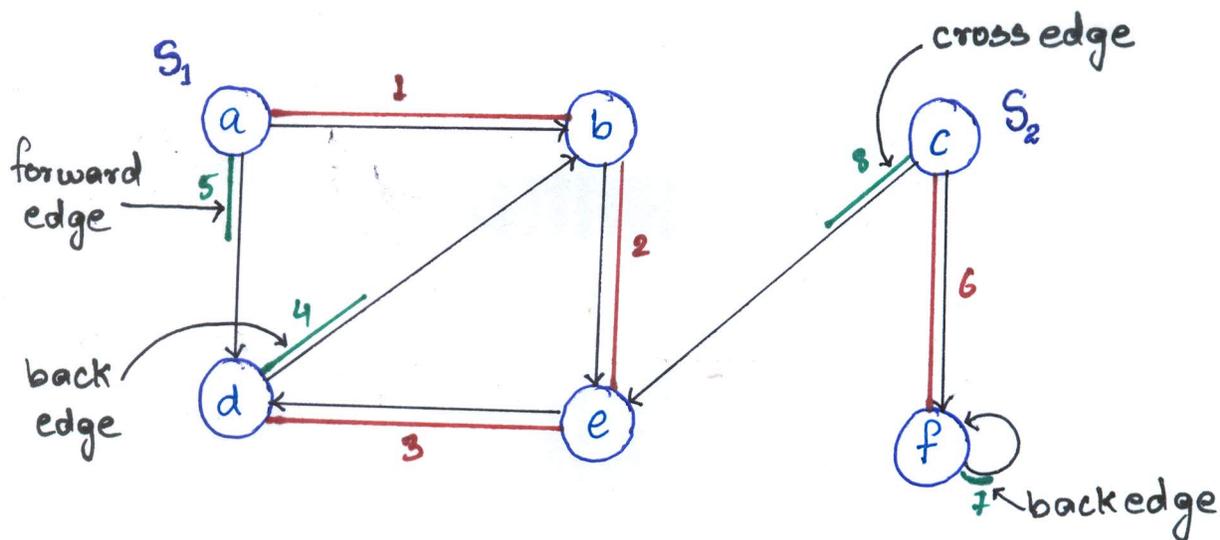
This is like exploring a maze.

- follow a path until you get stuck.
- backtrack along breadcrumbs until reach unexplored neighbour
- recursively explore
- careful not to repeat a vertex.

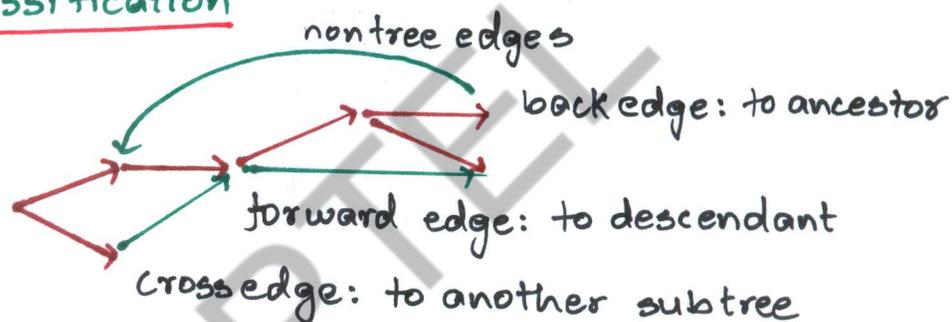
Depth-First-Search Algorithm

1. parent = {s: None}
 2. DFS-visit (v, Adj, s):
 3. for v in Adj[s]:
 4. if v not in parent:
 5. parent[v] = s
 6. DFS-visit (v, Adj, v)
 7. DFS (v, Adj)
 8. parent = {}
 9. for s in V:
 10. if s not in parent:
 11. parent[s] = None
 12. DFS-visit (v, Adj, s)
- search from start vertex s
(only see stuff reachable from s)
- explore entire graph
- (could do same to extend BFS)

Example of DFS



Edge Classification



- to compute this classification (**back or not**), mark nodes for duration they are "**on the stack**"
- only tree and back edges in undirected graph.

Analysis of DFS

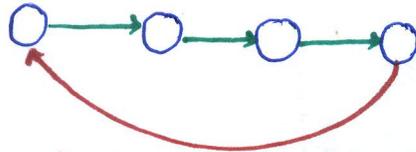
- DFS-visit gets called with a vertex s only once
⇒ time in DFS-visit = $\sum_{s \in V} |\text{Adj}[s]| = O(E)$
- DFS outer loop adds just $O(V)$
⇒ $O(V+E)$ time (linear time)

Cycle Detection

Graph G has a cycle \Leftrightarrow DFS has a back edge

Proof:

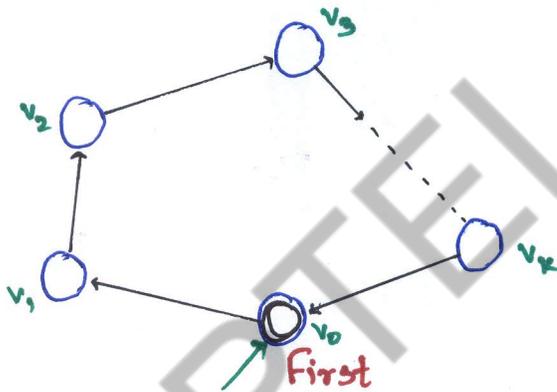
(\Leftarrow) tree edges



is a cycle

back edge: to tree ancestor

(\Rightarrow) consider first visit to cycle:



- before visit to v_i finishes, will visit v_{i+1} (and finish): will consider edge $(v_i, v_{i+1}) \Rightarrow$ visit v_{i+1} now or already did
- \Rightarrow before visit to v_0 finishes will visit v_k (and didn't before)
- \Rightarrow before visit to v_k (or v_0) finishes will see (v_k, v_0) as back edge.

Job Scheduling

Given Directed Acyclic Graph (DAG), where vertices represents task and edges represent dependencies, order tasks without violating dependencies.

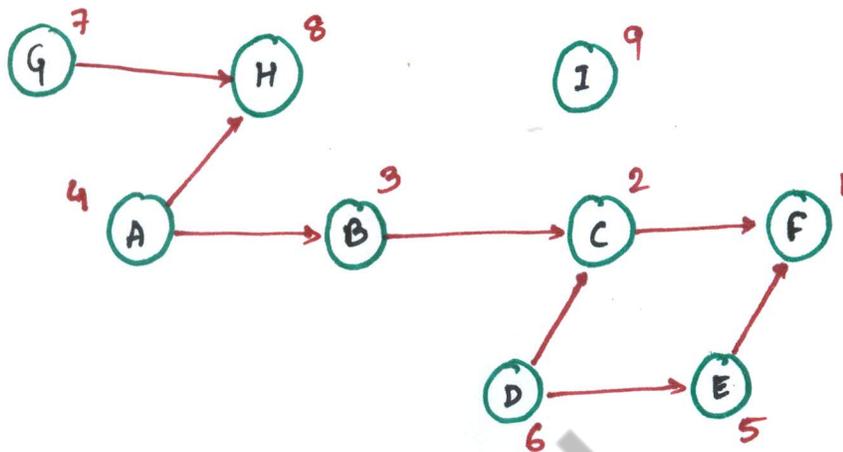


Fig: Dependence graph: DFS Finishing times.

Source:

Source = vertex with no incoming edges
= scheduling at beginning (A, G, I).

Attempt:

BFS from each source

- from A finds A, B, H, C, F
- from D finds D, B, E, C, F ← slow... and wrong!
- from G finds G, H
- from I finds I.

Topological Sort

Reverse of DFS finishing times
(time at which DFS-visit(v) finishes)

DFS-visit(v)
...
order.append(v)
order.reverse()

Correctness

For any edge (u, v) - u ordered before v , i.e. v finished before u



- if u visited before v :
 - before visit to u finishes, will visit v (via (u, v) or otherwise)
- $\Rightarrow v$ finishes before u

if v visited before u

graph is cyclic

$\Rightarrow u$ cannot be reached from v

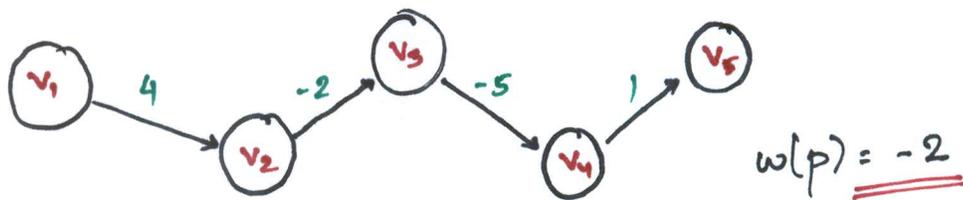
\Rightarrow visit to v finishes before visiting u .

Paths in graphs

Consider a digraph $G = (V, E)$ with edge-weight function $w: E \rightarrow \mathbb{R}$. The **weight** of path $p = v_1 \rightarrow v_2 \rightarrow \dots \rightarrow v_k$ is defined to be

$$w(p) = \sum_{i=1}^{k-1} w(v_i, v_{i+1})$$

Example:



Shortest Paths

A "shortest path" from u to v is a path of minimum weight from u to v .

The "shortest path weight" from u to v is defined as:

$$\delta(u, v) = \min \{ w(p) : p \text{ is a path from } u \text{ to } v \}$$

Note: $\delta(u, v) = \infty$ if no path from u to v exists.

Optimal Substructure

Theorem:

A subpath of a shortest path is a shortest path.

Triangle Inequality

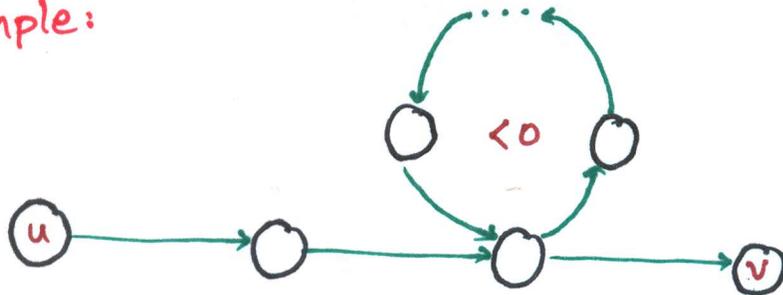
For all $u, v, x \in V$, we have

$$\delta(u, v) \leq \delta(u, x) + \delta(x, v)$$

Well-definedness of shortest paths

If a graph G contains a negative-weight cycle, then some shortest path may not exist.

Example:



Single-source shortest paths

Problem: From a given source vertex $s \in V$, find the shortest-path weights $\delta(u, v)$ for all $v \in V$.

If all edge weights $w(u, v)$ are non-negative, all shortest-path weights must exist.

Idea: Greedy

1. Maintain a set S of vertices whose shortest-path distance from s are known.
2. At each step add to S the vertex $v \in V - S$ whose distance estimate from s is minimal.
3. Update the distance estimates of vertices adjacent to v .

Dijkstra's Algorithm

$d[s] \leftarrow 0$

for each $v \in V - \{s\}$

do $d[v] \leftarrow \infty$

$S \leftarrow \emptyset$

$Q \leftarrow V$ \blacktriangleright Q is a priority queue maintaining $V - S$

while $Q \neq \emptyset$

do $u \leftarrow \text{EXTRACT-MIN}(Q)$

$S \leftarrow S \cup \{u\}$

for each $v \in \text{Adj}[u]$

do if $d[v] > d[u] + w(u,v)$

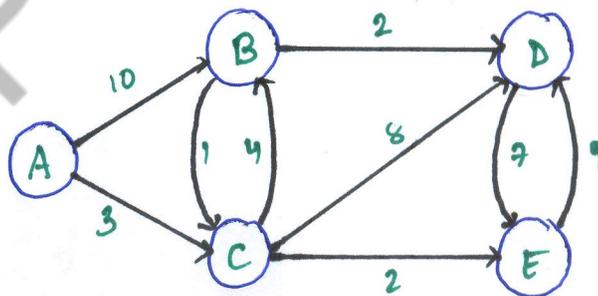
then $d[v] \leftarrow d[u] + w(u,v)$

} relaxation step

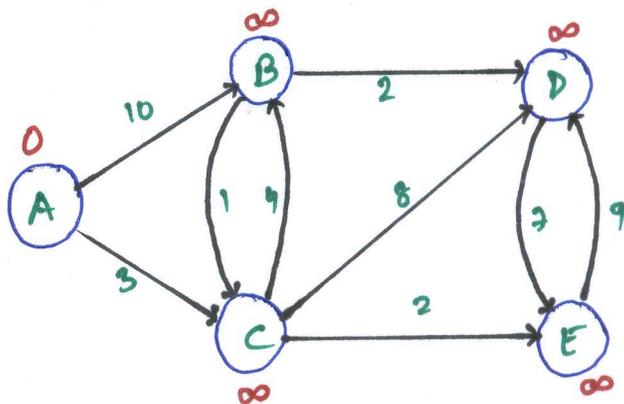
Implicit DECREASE-KEY

Example of Dijkstra's Algorithm

Graph with non-negative edge weights



Initialize

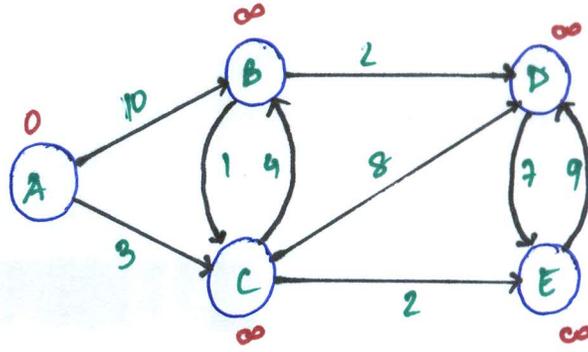


Q: A B C D E
 0 ∞ ∞ ∞ ∞

$A \leftarrow \text{EXTRACT-MIN}(Q)$

Q:	A	B	C	D	E
	0	∞	∞	∞	∞

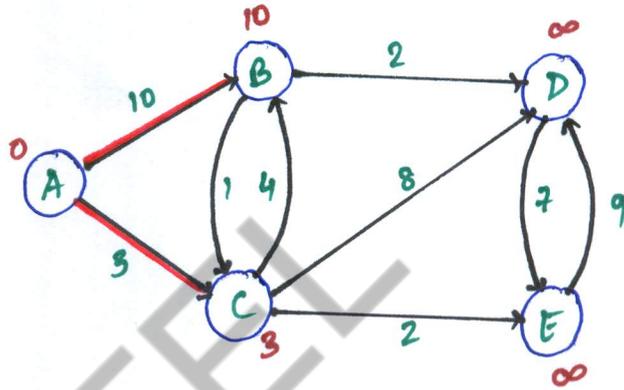
S: {A}



Relax all edges leaving A

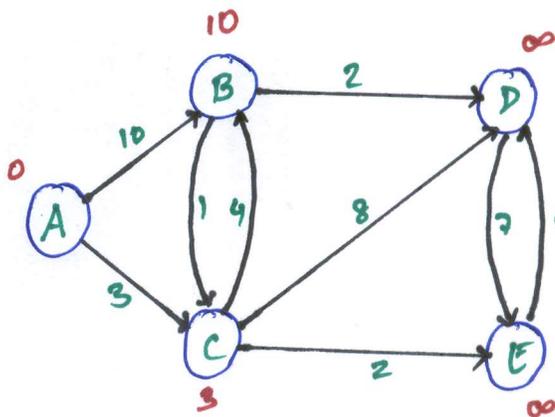
Q:	A	B	C	D	E
	0	∞	∞	∞	∞
		10	3	-	-

S: {A}



$C \leftarrow \text{EXTRACT-MIN}(Q)$

Q:	A	B	C	D	E
	0	∞	∞	∞	∞
		10	3	-	-

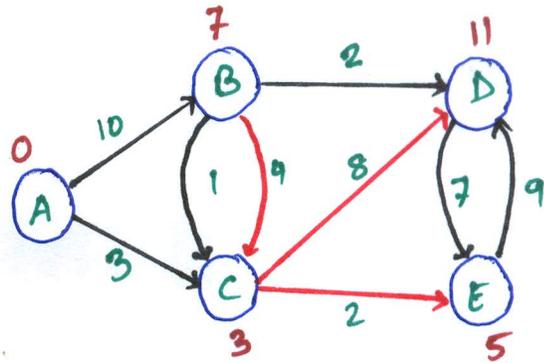


S: {A, C}

Relax all edges
leaving C

Q:

A	B	C	D	E
0	∞	∞	∞	∞
	10	3	-	-
	7		11	5

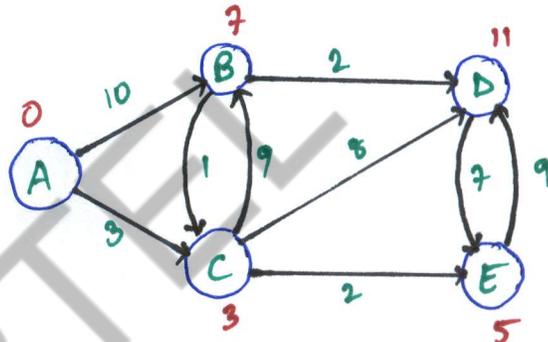


S: {A, C}

E ← EXTRACT-MIN(Q)

Q:

A	B	C	D	E
0	∞	∞	∞	∞
	10	3	-	-
	7		11	5

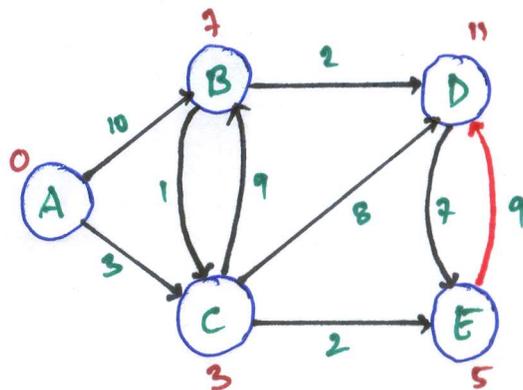


S: {A, C, E}

Relax all edges
leaving E

Q:

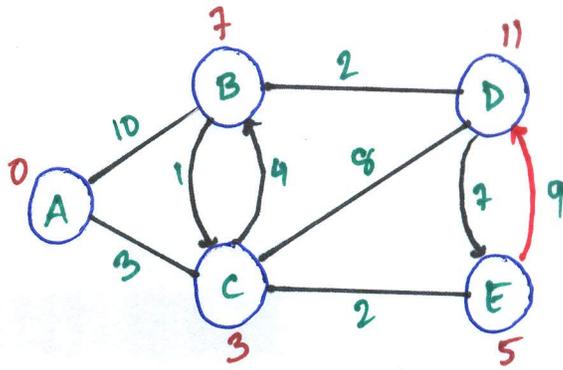
A	B	C	D	E
0	∞	∞	∞	∞
	10	3	-	-
	7		11	5
	7		11	



S: {A, C, E}

$B \leftarrow \text{EXTRACT-MIN}(Q)$:

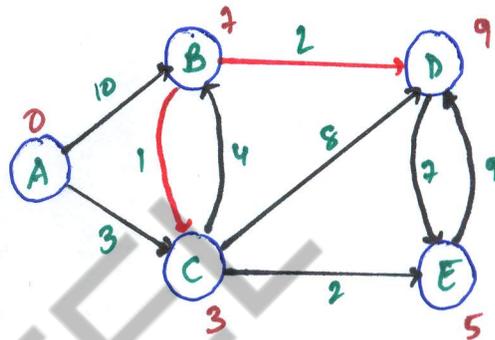
Q:	A	B	C	D	E
0	∞	∞	∞	∞	∞
	10	3	∞	∞	
	7			11	5
	7			11	



$S: \{A, C, E, B\}$

Relax all edges leaving B

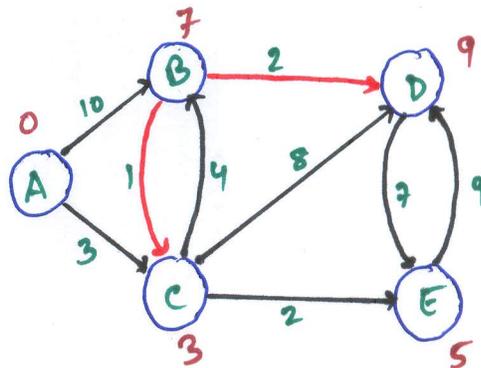
Q:	A	B	C	D	E
0	∞	∞	∞	∞	∞
	10	3	∞	∞	
	7			11	5
	7			11	
				9	



$S: \{A, C, E, B\}$

$D \leftarrow \text{EXTRACT-MIN}(Q)$

Q:	A	B	C	D	E
0	∞	∞	∞	∞	∞
	10	3	∞	∞	
	7			11	5
	7			11	
				9	



$S: \{A, C, E, B, D\}$

Correctness - Part I

Lemma: Initializing $d[s] \leftarrow 0$ and $d[v] \leftarrow \infty$ for all $v \in V - \{s\}$ establishes $d[v] \geq \delta(s, v)$ for all $v \in V$, and this invariant is maintained over any sequence of relaxation steps.

Proof:

Suppose **not**. Let v be the first vertex for which $d[v] < \delta(s, v)$, and let u be the vertex that caused $d[v]$ to change: $d[v] = d[u] + w(u, v)$. Then

$$d[v] < \delta(s, v) \quad \text{supposition}$$

$$\leq \delta(s, u) + \delta(u, v) \quad \text{triangle inequality}$$

$$\leq \delta(s, u) + w(u, v) \quad \text{sh. path} \leq \text{specific path}$$

$$\leq d[u] + w(u, v) \quad v \text{ is first violation.}$$

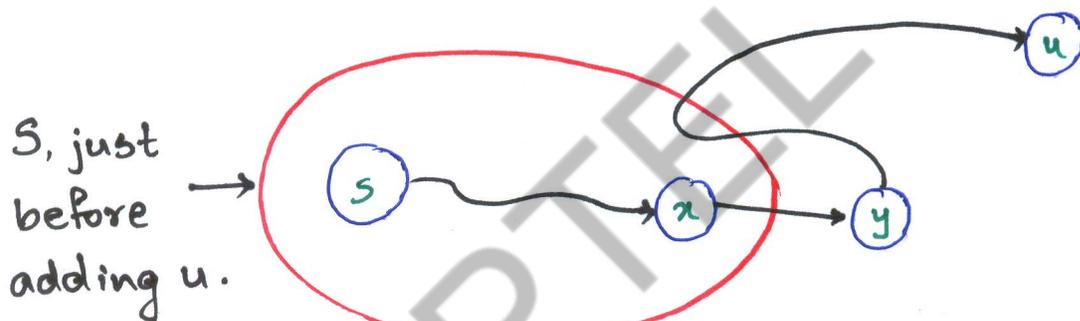
which is a contradiction.

Correctness - Part II

Theorem: Dijkstra's algorithm terminates with $d[v] = \delta(s, v)$ for all $v \in V$

Proof:

It suffices to show that $d[v] = \delta(s, v)$ for every $v \in V$ when v is added to S . Suppose u is the first vertex added to S for which $d[u] \neq \delta(s, u)$. Let y be the first vertex in $V - S$ along a shortest path from s to u , and x be its predecessor:



Since u is the first vertex violating the claimed invariant, we have $d[x] = \delta(s, x)$. Since subpaths of shortest paths are shortest paths, it follows that $d[y]$ was set to $\delta(s, x) + w(x, y) = \delta(s, y)$ when (x, y) was relaxed just after x was added to S .

Consequently, we have $d[y] = \delta(x, y) \leq \delta(s, u) \leq d[u]$.

But $d[u] \leq d[y]$ by our choice of u , and hence

$$\underline{d[y] = \delta(x, y) = \delta(s, u) = d[u]}$$

which is a contradiction.

Analysis of Dijkstra

```
while  $Q \neq \emptyset$ 
do  $u \leftarrow \text{EXTRACT-MIN}(Q)$ 
    $S \leftarrow S \cup \{u\}$ 
   for each  $v \in \text{Adj}[u]$ 
     do if  $d[v] > d[u] + w(u,v)$ 
        then  $d[v] \leftarrow d[u] + w(u,v)$ 
```

Handshaking Lemma

$\Rightarrow \Theta(E)$ implicit DECREASE-KEYS

Time = $\Theta(V) \cdot T_{\text{EXTRACT-MIN}} + \Theta(E) \cdot T_{\text{DECREASE-KEY}}$

Note: Same formula as in the analysis of Prim's minimum spanning tree algorithm.

<u>Q</u>	<u>$T_{\text{EXTRACT-MIN}}$</u>	<u>$T_{\text{DECREASE-KEY}}$</u>	<u>Total</u>
array	$O(V)$	$O(1)$	$O(V^2)$
binary heap	$O(\log V)$	$O(\log V)$	$O(E \log V)$
Fibonacci heap	$O(\log V)$	$O(1)$	$O(E + V \log V)$
	amortized	amortized	worst case

Negative Weight Cycle

Recall: If a graph $G = (V, E)$ contains a negative-weight cycle, then some shortest paths may not exist.

Bellman-Ford Algorithm:

Finds all shortest-path lengths from a source $s \in V$ to all $v \in V$ or determines that a negative cycle exists.

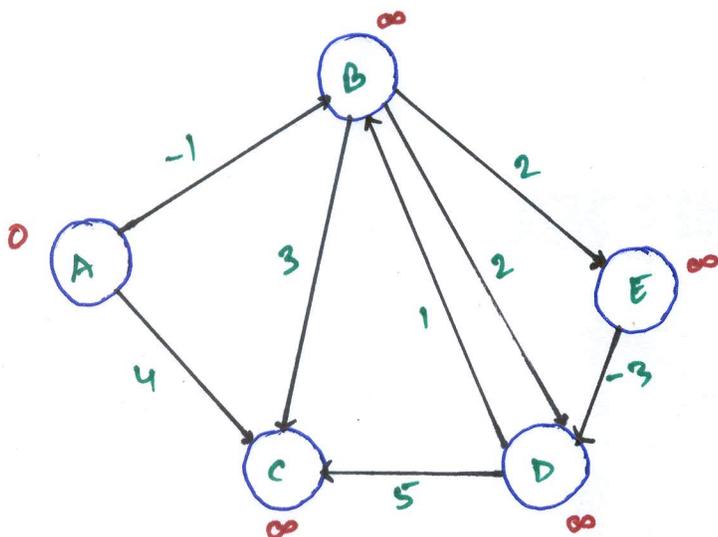
Bellman-Ford Algorithm

1. $d[s] \leftarrow 0$
 2. for each $v \in V - \{s\}$
 3. do $d[v] \leftarrow \infty$
 4. for $i \leftarrow 1$ to $|V| - 1$
 5. do for each edge $(u, v) \in E$
 6. do if $d[v] > d[u] + w[u, v]$
 7. then $d[v] \leftarrow d[u] + w[u, v]$
 8. for each edge $(u, v) \in E$
 9. do if $d[v] > d[u] + w[u, v]$
 10. then report that a negative-weight cycle exists.
- } initialization
} relaxation step

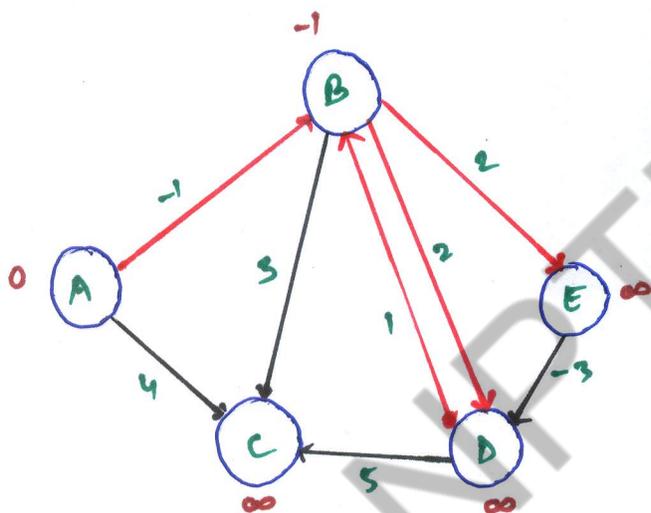
At the end, $d[v] = \delta(s, v)$

Time = $O(VE)$

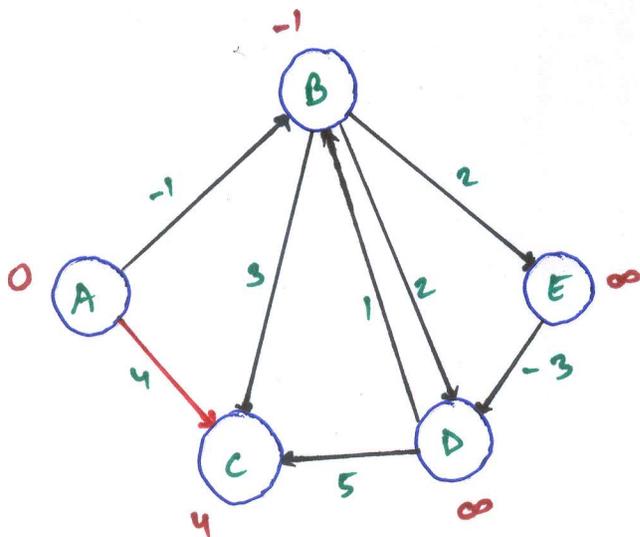
Example of Bellman-Ford



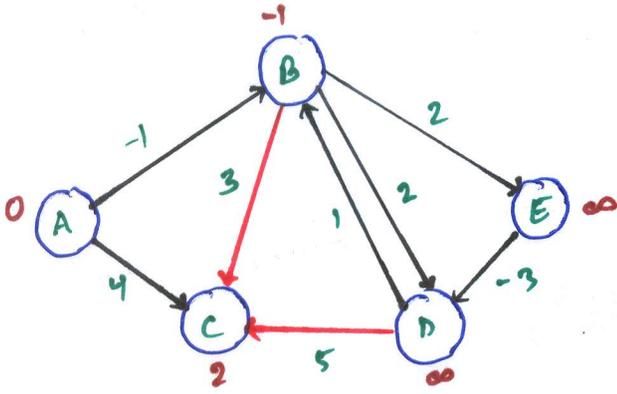
A	B	C	D	E
0	∞	∞	∞	∞



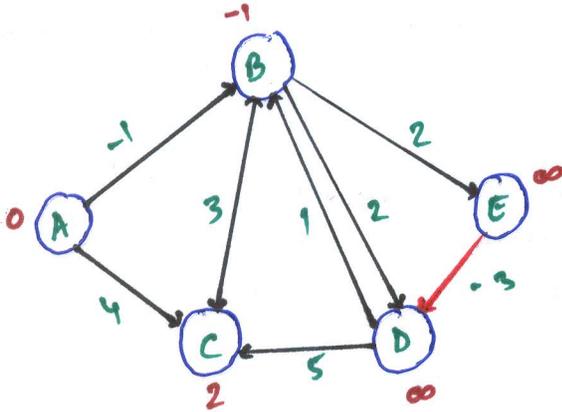
A	B	C	D	E
0	∞	∞	∞	∞
0	-1	∞	∞	∞



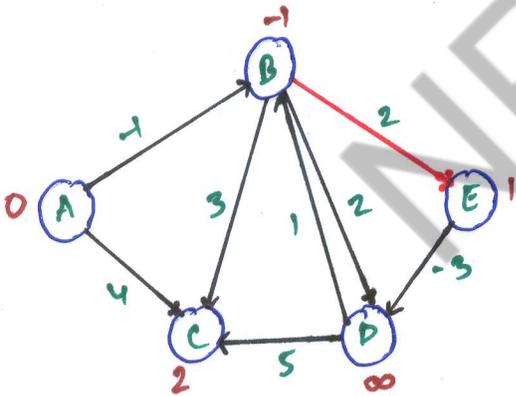
A	B	C	D	E
0	∞	∞	∞	∞
0	-1	∞	∞	∞
0	-1	4	∞	∞



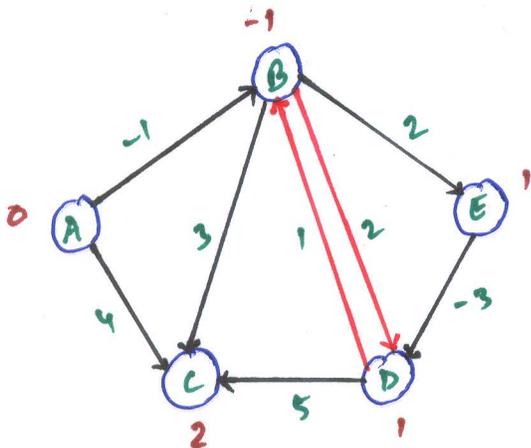
A	B	C	D	E
0	∞	∞	∞	∞
0	-1	∞	∞	∞
0	-1	4	∞	∞
0	-1	2	∞	∞



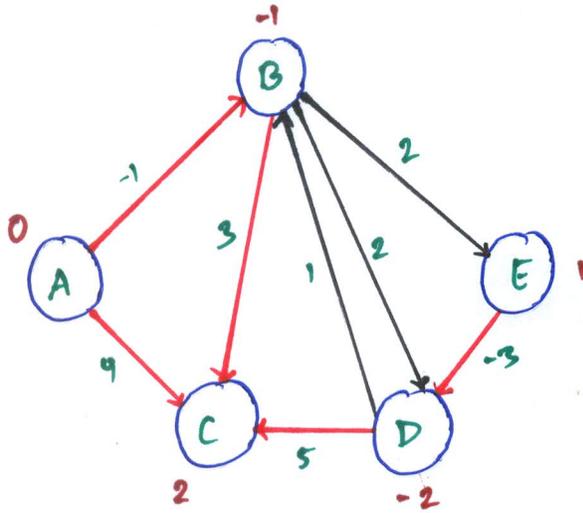
A	B	C	D	E
0	∞	∞	∞	∞
0	-1	∞	∞	∞
0	-1	4	∞	∞
0	-1	2	∞	∞



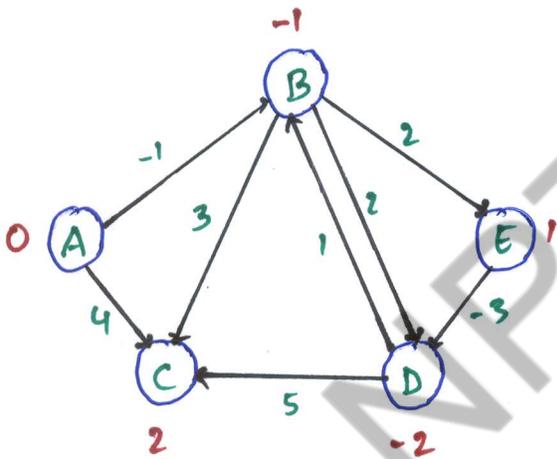
A	B	C	D	E
0	∞	∞	∞	∞
0	-1	∞	∞	∞
0	-1	4	∞	∞
0	-1	2	∞	∞
0	-1	2	∞	1



A	B	C	D	E
0	∞	∞	∞	∞
0	-1	∞	∞	∞
0	-1	4	∞	∞
0	-1	2	∞	∞
0	-1	2	∞	1
0	-1	2	1	1



<u>A</u>	<u>B</u>	<u>C</u>	<u>D</u>	<u>E</u>
0	∞	∞	∞	∞
0	-1	∞	∞	∞
0	-1	4	∞	∞
0	-1	2	∞	∞
0	-1	2	∞	1
0	-1	2	1	1
0	-1	2	-2	1



<u>A</u>	<u>B</u>	<u>C</u>	<u>D</u>	<u>E</u>
0	∞	∞	∞	∞
0	-1	∞	∞	∞
0	-1	4	∞	∞
0	-1	2	∞	∞
0	-1	2	∞	1
0	-1	2	1	1
0	-1	2	-2	1