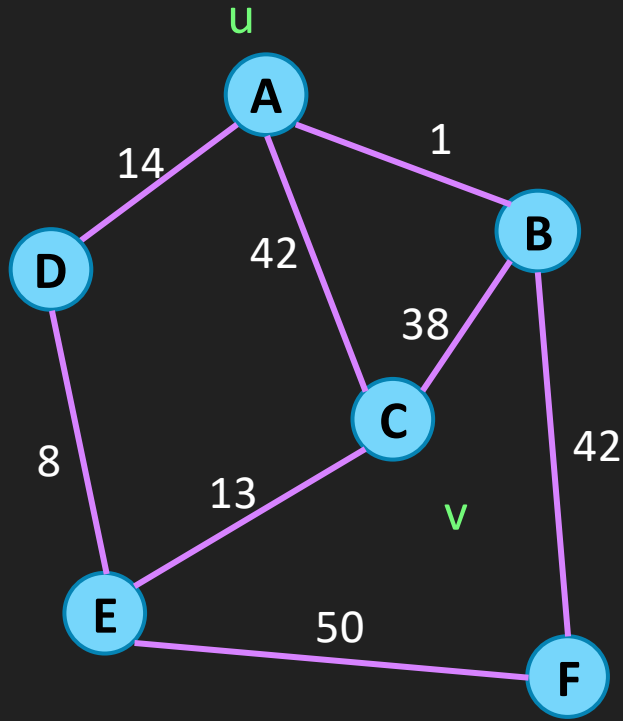


# Shortest Path in Graph

## The Problem



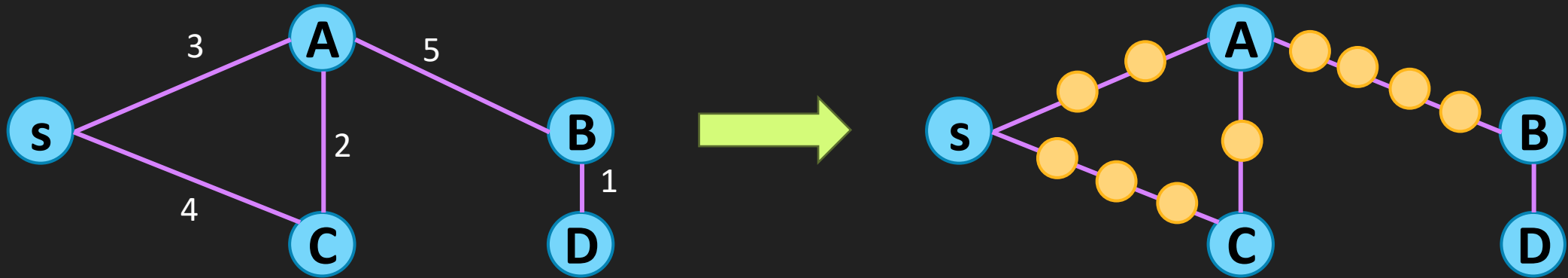
- Finding a path from **u** to **v** with minimum summation of weight of the path
  - May not be the path with minimum number of edges
- [A,C] 42
- [A,B,C] 39
- [A,D,E,C] 35

## Single Source Shortest Path Problem

- **Problem:** Given a graph  $G$  with a starting node  $s$ , find the shortest path from  $s$  to every node in the graph
- **Input:**
  - A graph  $G = (V, E)$
  - A starting node  $s$
  - A weight function  $w$  where  $w(a, b)$  returns a weight of an edge  $(a, b)$ 
    - For unweighted graph, we can set  $w(a, b) = 1$  for any edge  $(a, b)$
- **Output:**
  - $dist[x]$  the shortest distance from  $s$  to  $x$ 
    - If a path is needed,  $prev[x]$  is used just like MST

## Direct solution using BFS

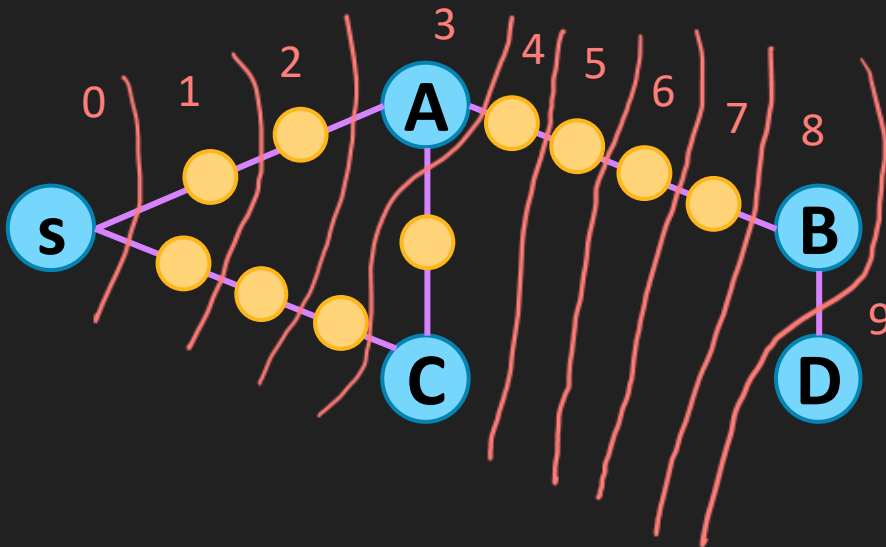
- BFS can give us the shortest path on unweighted graph
- Convert weighted to unweighted by adding **auxiliary nodes**



- However, this is very slow. BFS is  $O(n+e)$  but now we have  $n$  approximately as  $\sum(w^*)$ 
  - Imagine when a single edge has length  $10^6$

## Alarm Clock Analogy

- No need to calculate the distance to the auxiliary node
  - We want to skip to the actual node
- Since BFS consider nodes in order of distance, we want to skip to the distance when actual node is reached



no need to calculate  
distance of 1, 2, 5, 6, 7 we  
can skip to 3, 4, 8 and 9

## Alarm Clock Algorithm

- Each node has its alarm clock, initially these alarm is not set.
- Set an alarm clock for node  $s$  at time 0
- Repeat until there are no more alarms:
  - Let the next alarm goes off at time  $T$ , at node  $u$ . Then:
    - The distance from  $s$  to  $u$  is  $T$  (i.e. set  $\text{dist}[u] = T$ )
    - For  $v$  in  $G.\text{adj}(u)$ 
      - Consider the edge  $(u,v)$  having weight as  $w(u,v)$ .
      - If we do BFS on this edge with auxiliary nodes, we will reach  $v$  at time  $T + w(u, v)$
      - If there is no alarm yet for  $v$  set it to the time  $T + w(u, v)$ .
      - If  $v$ 's alarm is set for later than  $T + w(u, v)$ , then reset it to this earlier time.

# Dijkstra's Algorithm

```
def dijkstra(G,w,s)
  for u in G.V
    dist[u] =  $\infty$ 
    prev[u] = -1
  end
  S = new Set
  S.insert( (0,s) )
  while S is not empty
    (T,u) = S.min
    for v in G.adj(u)
      if dist[v] > T + w(u,v)
        # change the value of v to the new cost[v]
        S.remove_if_exist( (dist[v],v) )
        dist[v] = T + w(u,v)
        prev[v] = u
        S.insert( (dist[v],v) )
      end
    end
  end
  return dist
end
```

This is the same as  
prim's algorithm except  
that we use  $T + w(u,v)$   
instead of  $w(u,v)$

- Use **set** to store alarm of each node.
- The next alarm goes off at the minimum value in the set
- Adjust alarm by remove and re-insert alarm of that node

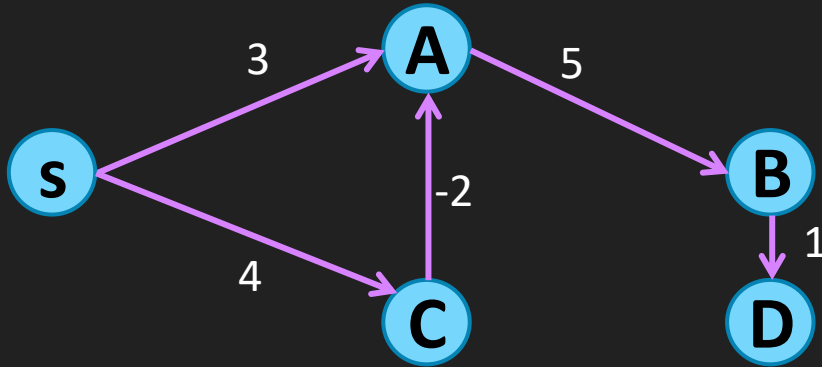


## Analysis

- Exactly the same as prim's algorithm
  - $O( n (\lg n) + e (\lg n) ) = O( (n+e) \lg n )$
  - Depends on underlying datastructure that store the alarm
    - Fibonacci Heap gives the best complexity
- Unlike minimum spanning tree, Dijkstra's algorithm works with directed graph as well
  - Because we do not care about going back from  $v$  to  $s$ , we consider only from  $s$  to other nodes



## Graph with Negative Edge



Shortest path to **A** pass through **C** which is, in BFS sense, is further than **A**

- Dijkstra's works because a shortest path to **v** must pass through a node **u** which is closer to **s** than **v**
  - Dijkstra's algorithm says that when an alarm goes off, that is when BFS has reached that node and we know the shortest distance.
  - The alarm at the closer node goes off before **v** so that we can calculate the alarm at **v** correctly
  - This assumption is not correct when the graph has negative edges

## Directed Graph and Negative Edges

- An undirected graph with a negative edge has no shortest path
  - The shortest path makes no sense because we can go through the negative edge repeatedly and the distance always decrease
- Unlike minimum spanning tree, Dijkstra's algorithm works with directed graph as well
  - Because we do not care about going back from  $v$  to  $s$ , we consider only from  $s$  to other nodes
- For directed graph, negative edges are possible if they do not make a negative cycle

| Directed | Negative Edge | Shortest Path possible?               | Dijkstra's Algorithm Work Correctly? |
|----------|---------------|---------------------------------------|--------------------------------------|
| No       | No            | Yes                                   | Yes                                  |
| No       | Yes           | No                                    | No                                   |
| Yes      | No            | Yes                                   | Yes                                  |
| Yes      | Yes           | Yes (graph can't have negative cycle) | No                                   |

# Bellman-Ford Algorithm

Dealing with directed graph with negative edge

## Key Idea in Shortest Path

T is the current dist[u]

```
...  
while S is not empty  
    (T,u) = S.min  
    for v in G.adj(u)  
        if dist[v] > T + w(u,v)  
            # change the value of v to the new cost[v]  
            S.remove_if_exist( (dist[v],v) )  
            dist[v] = T + w(u,v)  
            prev[v] = u  
            S.insert( (dist[v],v) )  
        end  
    end  
end
```

dist[u] is T, the distance  
of the node whose alarm  
just go off

- Update the distance

$\text{if } \text{dist}[v] > \text{dist}[u] + w(u,v)$

$\text{dist}[v] = \text{dist}[u] + w(u,v)$

dist[v] is the alarm of  
neighbor of u

## Another Approach to Shortest Path

- A shortest path can have at most  $n - 1$  edges
- Writing a recurrence relation on number of edges
  - $D(a, v)$  = shortest distance from  $s$  to  $v$  using at most  $a$  edges
- Initial condition  $D(*, s) = 0$ ,  $D(*, *) = \infty$  for any other cases

- $D(a, v) = \min$

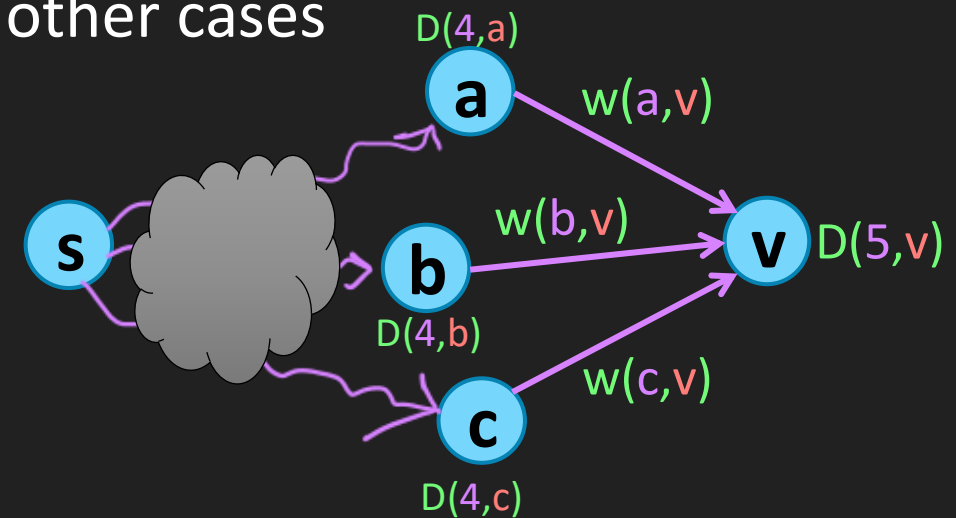
- $D(a - 1, v)$

- $\min( D(a - 1, u) + w(u, v) )$

- For any  $u$  that has edges go to  $v$

Remember the shortest path using lesser number of edges

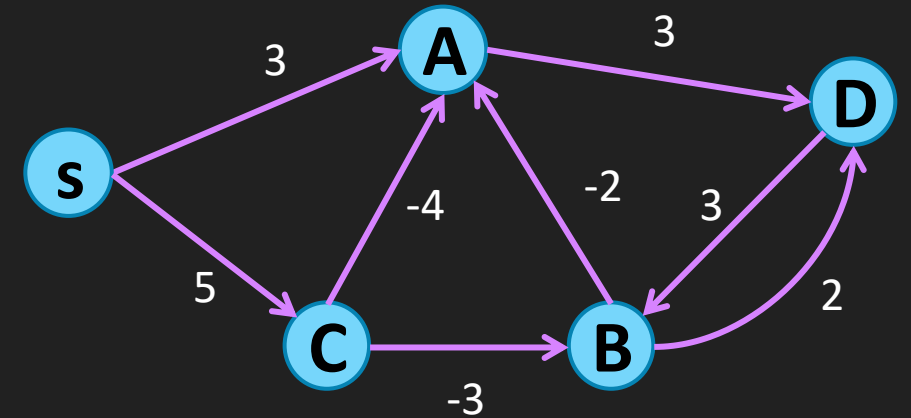
We use  $a-1$  edges to reach a node that has a path to  $v$  and use the edge  $(u, v)$  to reach  $v$



# Bellman-Ford Algorithm

- Dynamic Programming that calculates  $D(*,*)$

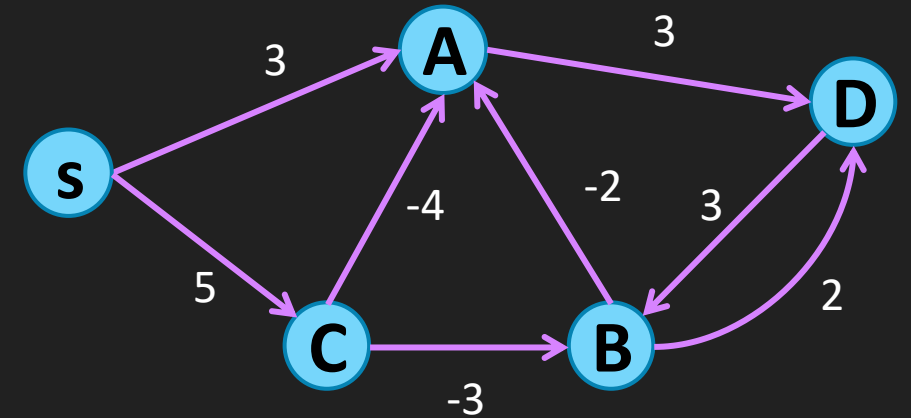
|   | <b>S</b> | <b>A</b> | <b>B</b> | <b>C</b> | <b>D</b> |
|---|----------|----------|----------|----------|----------|
| 0 | 0        | $\infty$ | $\infty$ | $\infty$ | $\infty$ |
| 1 | 0        | 3        | $\infty$ | 5        | $\infty$ |
| 2 | 0        |          |          |          |          |
| 3 | 0        |          |          |          |          |
| 4 | 0        |          |          |          |          |



# Bellman-Ford Algorithm

- Dynamic Programming that calculates  $D(*,*)$

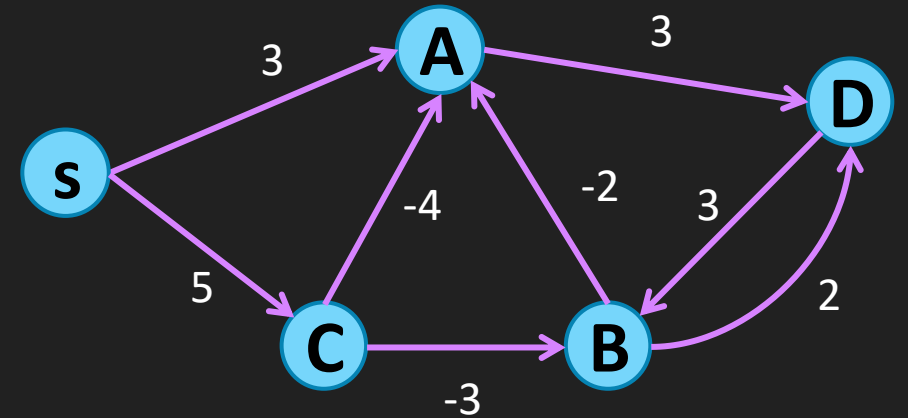
|   | <b>s</b> | <b>A</b> | <b>B</b> | <b>C</b> | <b>D</b> |
|---|----------|----------|----------|----------|----------|
| 0 | 0        | $\infty$ | $\infty$ | $\infty$ | $\infty$ |
| 1 | 0        | 3        | $\infty$ | 5        | $\infty$ |
| 2 | 0        | 1        | 2        | 5        | 6        |
| 3 | 0        |          |          |          |          |
| 4 | 0        |          |          |          |          |



## Bellman-Ford Algorithm

- Dynamic Programming that calculates  $D(*,*)$

|   | <b>S</b> | <b>A</b> | <b>B</b> | <b>C</b> | <b>D</b> |
|---|----------|----------|----------|----------|----------|
| 0 | 0        | $\infty$ | $\infty$ | $\infty$ | $\infty$ |
| 1 | 0        | 3        | $\infty$ | 5        | $\infty$ |
| 2 | 0        | 1        | 2        | 5        | 6        |
| 3 | 0        | 0        | 2        | 5        | 4        |
| 4 | 0        |          |          |          |          |

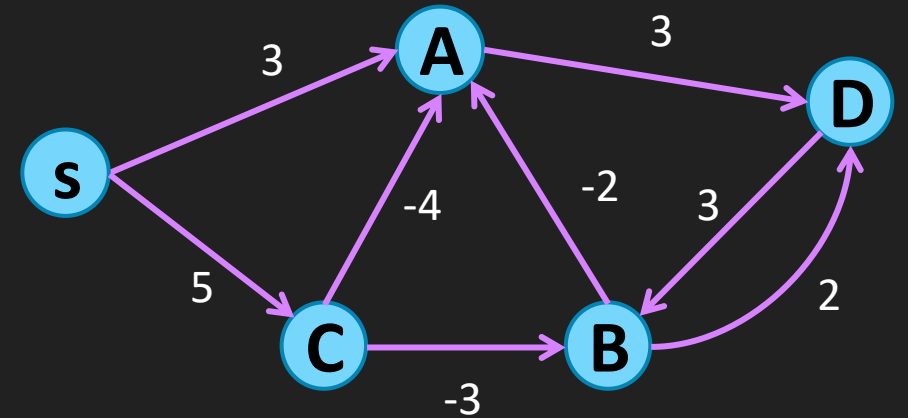




# Bellman-Ford Algorithm

- Dynamic Programming that calculates  $D(*,*)$

|   | <b>S</b> | <b>A</b> | <b>B</b> | <b>C</b> | <b>D</b> |
|---|----------|----------|----------|----------|----------|
| 0 | 0        | $\infty$ | $\infty$ | $\infty$ | $\infty$ |
| 1 | 0        | 3        | $\infty$ | 5        | $\infty$ |
| 2 | 0        | 1        | 2        | 5        | 6        |
| 3 | 0        | 0        | 2        | 5        | 4        |
| 4 | 0        | 0        | 2        | 5        | 3        |

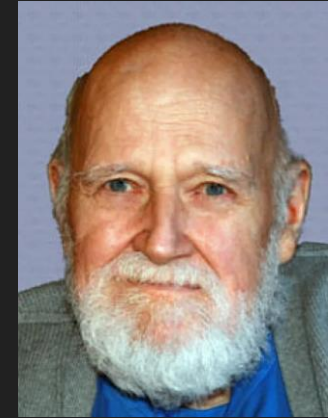


# Bellman-Ford Algorithm (Direct Implementation)

```
def bellman_ford(G, w, s)
    for u in G.V
        D[0][u] =  $\infty$ 
        prev[u] = -1
    end

    D[0][s] = 0
    for i from 1 to n - 1
        for u in G.V
            D[i][u] = D[i-1][u]
            for all edges (a,b) in G.E
                if D[i][b] > D[i-1][a] + w(a,b)
                    D[i][b] = D[i-1][a] + w(a,b)
                    prev[b] = a
                end
            end
        end
    end

    for u in G.V
        dist[u] = D[n-1][u]
    end
end
```



## Bellman-Ford Algorithm (Actual)

- Observe that  $D[a][*]$  uses  $D[a-1][*]$ 
  - When computing  $D[a]$ , we don't need  $D[a-2][*]$ ,  $D[a-3][*]$ , ...
  - No need to keep every row of  $D$ , just remember the previous row
- In fact, Bellman-ford proposed that we just **use only one row** and compute the recurrence relation directly on that row
  - This does not strictly compute the recurrence relation faithfully in each iteration (it might “compute ahead”) but at the final round, it is guaranteed that  $D[*]$  is the same as the recurrence relation, or lower because no path has more than  $n-1$  edges

```
def bellman_ford(G, w, s)
    for u in G.V
        dist[u] = ∞
        prev[u] = -1
    end

    dist[s] = 0
    for i from 1 to n - 1
        for all edges (a,b) in G.E
            if dist[b] > dist[a] + w(a,b)
                dist[b] = dist[a] + w(a,b)
                prev[b] = a
            end
        end
    end
    return dist
end
```

## Detecting Negative Cycle

- After repeating the loop  $n-1$  times
- If there is a case when
$$\text{dist}[v] > \text{dist}[u] + w(u,v)$$
- Then, there is a negative cycle
  - Because there is a shorter path that use  $n$  edges which indicates that some node are repeated in the path, hence, a cycle

```
def bellman_ford(G, w, s)
    for u in G.V
        dist[u] = ∞
        prev[u] = -1
    end
    dist[s] = 0
    for i from 1 to n - 1
        for all edges (a,b) in G.E
            if dist[b] > dist[a] + w(a,b)
                dist[b] = dist[a] + w(a,b)
                prev[b] = a
            end
        end
    end
    for all edges (a,b) in G.E
        if dist[b] > dist[a] + w(a,b)
            return "negative cycle"
        end
    end
    return dist
end
```

## Analysis

- Very simple
- Loop  $n-1$  times (plus another round for detect a negative cycle)
  - Each loop takes  $e$  iterations
- $O(ne)$ 
  - Comparing to Dijkstra's  $O(e \lg n)$
- Dense graph
  - Bellman-Ford  $O(n^3)$
  - Dijkstra's  $O(n^2 \lg n)$
- Bellman-Ford is slower but can work with negative edge

# All Pair Shortest Path

Shortest path between every pair of nodes

## All Pair Shortest Path Problem

- **Problem:** Given a graph  $G$  find the shortest path for every pair of nodes
- **Input:**
  - A graph  $G = (V, E)$
  - A weight function  $w$  where  $w(a, b)$  returns a weight of an edge  $(a, b)$ 
    - For unweighted graph, we can set  $w(a, b) = 1$  for any edge  $(a, b)$
- **Output:**
  - $dist[a][b]$  the shortest distance from  $a$  to  $b$ 
    - If a path is needed,  $prev[a][b]$  is used just like MST, starting with  $a$

## Approach

- Standard shortest path gives shortest path from a given vertex  $s$  to every vertex
  - Repeat this for every starting vertex  $s$
  - Dijkstra  $O(n (e \lg n) )$
  - Bellman-Ford  $O(n (e n) )$
- Floyd-Warshall use Dynamic Programming Approach that gives  $O(n^3)$





## Floyd-Warshall

- Use another recurrent
- $d_k(a,b)$  is a shortest distance from  $a$  to  $b$  that the path can visit node only in the set  $\{1,2,3,\dots,k\}$ 
  - I.e., the shortest path can be  $[a,p_1,p_2,\dots,p_m,b]$  where  $1 \leq p_i \leq k$  and we can use any number of nodes in  $p_1,\dots,p_m$
- $d_0(a,b) = w(a,b)$  because it can not go through any nodes

## Recurrent Relation

- $d_k(a,b) = \min$  of

- $d_{k-1}(a,b)$

- $d_{k-1}(a,k) + d_{k-1}(k,b)$

There are exactly three dependency (comparing to Bellman-Ford where each value depends on P other entries where P is the in-degree)

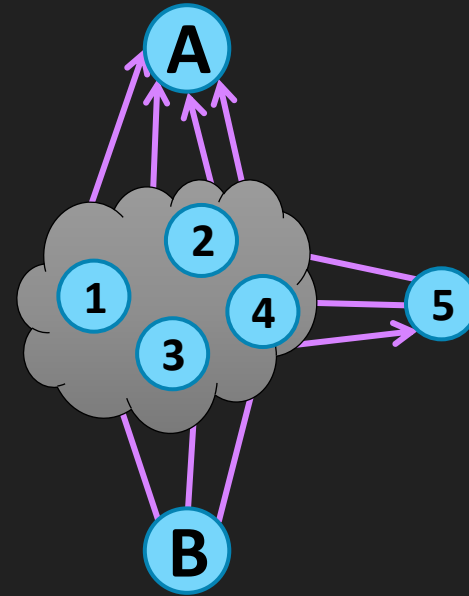
- Initial condition

- $d_0(a,b) = w(a,b)$

- $d_0(a,a) = 0$

Even though D is now 3 dimension,  $d_k$  depends on  $d_{k-1}$ . There is no need to remember multiple layer. So, 2D dimension array can be used.

- The answer to the problem is  $d_n(*,*)$



# Floyd-Warshall Implementation

```
def floyd_warshall(G,w)
  for i = 1 to n
    for j = 1 to n
      dist[i][j] = w(i,j)

  for k = 1 to n
    for i = 1 to n
      for j = 1 to n
        dist[i][j] =
          min (dist[i][j],
              dist[i][k] + dist[k][j])
    for i = 1 to n
      if dist[i][i] < 0
        return "negative cycle"
  return dist
end
```

Also assume that  $w(a,a)$  is 0

The node  $k$  must be the outermost loop

- Analysis
  - Very simple
  - 3 nested loops of  $n$
  - $O(n^3)$
- Detecting negative cycle by checking if  $\text{dist}[a][a]$  is negative

## Comparing with Bellman-Ford

- For dense graph (where  $e$  approach  $n^2$ )
  - Both Bellman-Ford and Floyd-Warshall has the same time complexity  $O(n^3)$ 
    - May choose Floyd-Warshall over Bellman-Ford
- For sparse graph
  - Bellman-ford is  $O(n e)$
  - Floyd-Warshall is still  $O(n^3)$