COMP319 Algorithms 1 Lecture 14 Shortest Paths

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Textbook Chapters 25 and 26

Slide credits: 홍석원, 명지대학교; 김한준, 서울시립대학교; J.

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- Introduction to Graph Algorithms
- Definition of shortest path problem
- Bellman-Ford Algorithm
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Shortest Paths

- 조건
 - 간선 가중치가 있는 유향 그래프
 - 무향 그래프는 각 간선에 대해 양쪽으로 유향 간선이 있는 유향 그래프로 생각할 수 있다
 - o 즉, 무향 간선 (u, v)는 유향 간선 (u, v)와 (v, u) 두 개가 존재함을 의미한다고 가정하면 된다
- 두 정점 사이의 최단경로
 - 가능한 경로들 중 간선 가중치 합이 최소인 경로
 - 간선 가중치의 합이 음인 싸이클이 있으면 문제가 정의되지 않는다
 - ο 무한반복된다

Shortest Paths

- 단일 시작점 최단경로
 - 주어진 단일 시작점으로부터 각 정점에 이르는 최단경로를 구한다
 - 싸이클이 없는 그래프의 최단경로
 - 벨만-포드(Bellman-Ford) 알고리즘
 - o 음의 가중치를 허용하는 최단경로
 - 다익스트라(Dijkstra's) 알고리즘
 - o 음의 가중치를 허용하지 않는 최단경로
- 모든 쌍 최단경로
 - 모든 정점 쌍 사이의 최단경로를 모두 구한다
 - 플로이드-워샬(Floyd-Warshall) 알고리즘

Single-source shorted path BELLMAN-FORD ALGORITHM

Single-Source Shortest Path

- Problem: given a weighted directed graph G, find the minimum-weight path from a given source vertex s to another vertex v
 - "Shortest-path" = minimum weight
 - Weight of path is sum of edges
 - E.g., a road map: what is the shortest path from Chapel Hill to Charlottesville?

Shortest Path Properties

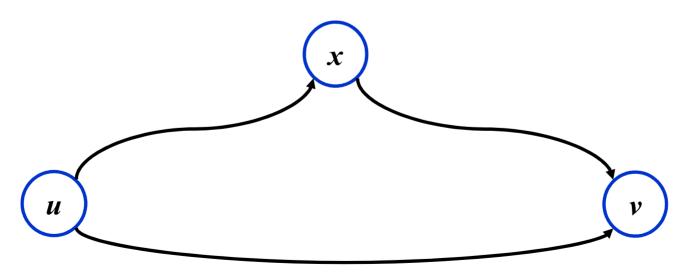
 Again, we have optimal substructure: the shortest path consists of shortest subpaths:



- Proof: suppose some subpath is not a shortest path
 - o There must then exist a shorter subpath
 - o Could substitute the shorter subpath for a shorter path
 - But then overall path is not shortest path. Contradiction

Shortest Path Properties

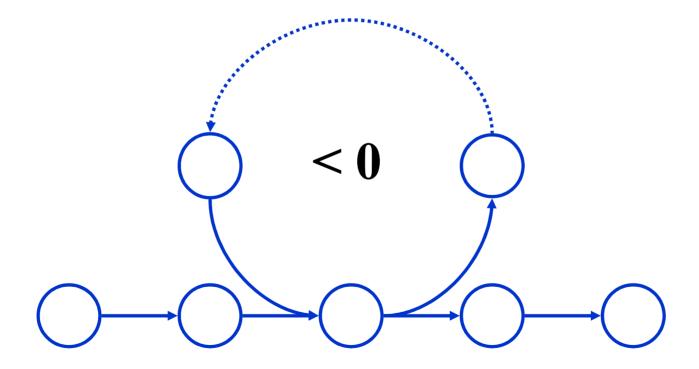
- Define $\delta(u,v)$ to be the weight of the shortest path from u to v
- Shortest paths satisfy the triangular inequality:
 - $\delta(u,v) \leq \delta(u,x) + \delta(x,v)$
- "Proof":



This path is no longer than any other path

Shortest Path Properties

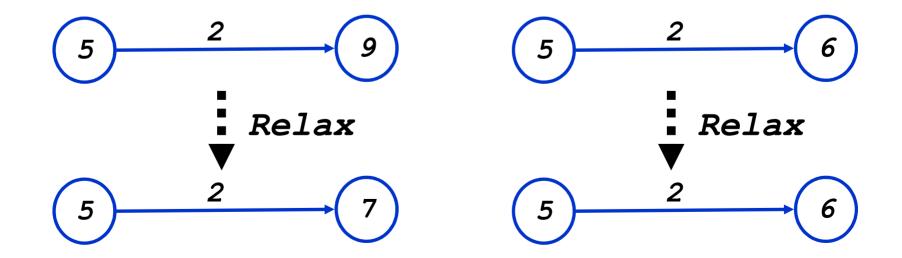
• In graphs with negative weight cycles, some shortest paths will not exist (Why?):



Relaxation

- A key technique in shortest path is relaxation
 - Idea: for all v, maintain upper bound d[v] on $\delta(s,v)$

```
Relax(u,v,w) {
    if (d[v] > d[u]+w) then d[v]=d[u]+w;
}
```



Bellman-Ford by Dynamic Programming

- d_t^k : 중간에 최대 k 개의 간선을 거쳐 정점 r로부터 정점 t에 이르는 최단거리
- 목표: d_tⁿ⁻¹

$$\sqrt{M} \mathcal{A} \stackrel{\mathcal{A}}{\mathcal{A}} \stackrel{\mathcal{A}}{\mathcal{A}}$$

$$d_{v}^{k} = \min_{\text{for } \mathcal{A} \stackrel{\mathcal{A}}{\mathcal{A}} \stackrel{\mathcal{A}}{\mathcal{A}} \stackrel{\mathcal{A}}{\mathcal{A}} = 0$$

$$d_{r}^{0} = 0$$

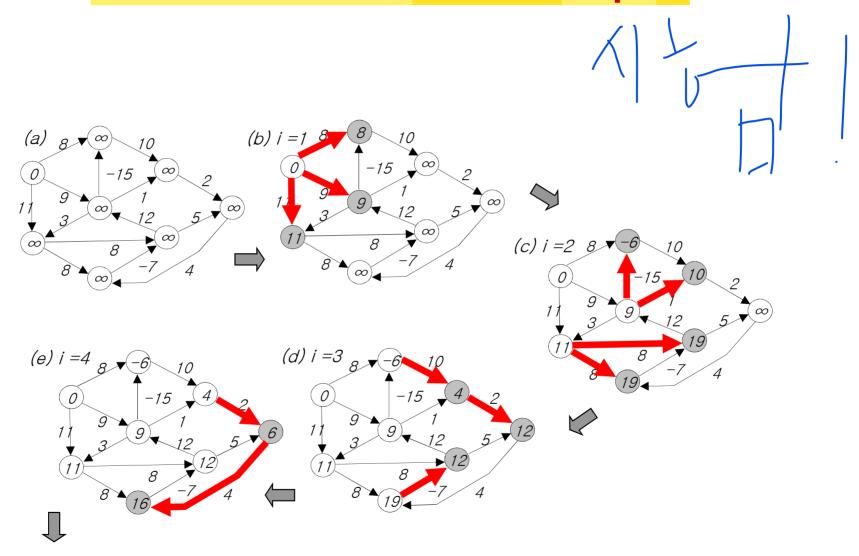
$$d_{t}^{0} = \infty, \quad t \neq r$$

Bellman-Ford Algorithm

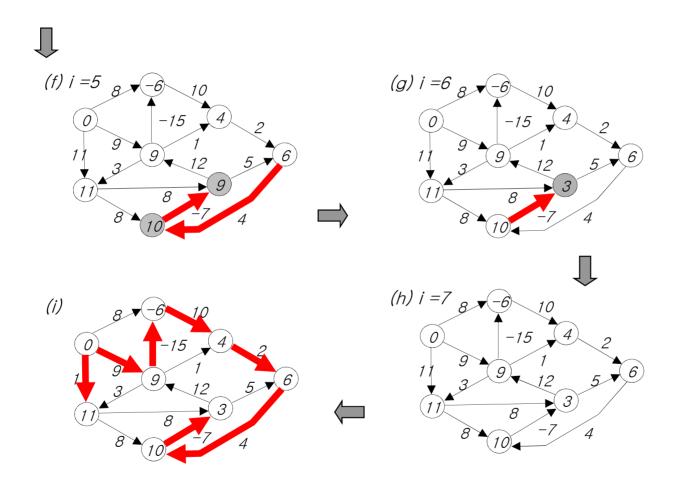
```
BellmanFord()
                                            Initialize d[], which will converge to shortest-path value \delta
    for each v \in V
       d[v] = \infty;
   d[s] = 0;
                                            Relaxation:
                                            Make |V|-1 passes,
       for each edge (u,v) \in E
                                            relaxing each edge
           Relax(u,v, w(u,v));
    for each edge (u,v) \in E
                                            Test for solution
       if (d[v] > d[u] + w(u,v))
                                            Under what condition
                                            do we get a solution?
              return "no solution";
```

```
Relax(u,v,w): if (d[v] > d[u]+w) then d[v]=d[u]+w
```

Bellman-Ford Example



Bellman-Ford Example



Bellman-Ford Algorithm

```
BellmanFord()
   for each v \in V
      d[v] = \infty;
   d[s] = 0;
   for i=1 to |V|-1
      for each edge (u,v) \in E
         Relax(u,v, w(u,v));
   for each edge (u,v) \in E
      if (d[v] > d[u] + w(u,v))
            return "no solution";
                                        Ex: work on board
```

```
Relax(u,v,w): if (d[v] > d[u]+w) then d[v]=d[u]+w
```

Bellman-Ford Algorithm

```
BellmanFord()
   for each v \in V
                                      What will be the
      d[v] = \infty;
                                      running time?
   d[s] = 0;
                                      A: O(VE)
   for i=1 to |V|-1
      for each edge (u,v) \in E
         Relax(u,v, w(u,v));
   for each edge (u,v) \in E
      if (d[v] > d[u] + w(u,v))
            return "no solution";
```

Relax(u,v,w): if (d[v] > d[u]+w) then d[v]=d[u]+w

Correctness of Bellman-Ford

- If G=(V,E) contains no negative-weight cycles, then after then after the Bellman-Ford algorithm executes, $d[v] = \delta(s,v)$ for all $v \in V$.
- Proof. Let v ∈ V be any vertex, and consider a shortest path p from s to v with the minimum number of edges.

$$s \rightarrow V_1 \rightarrow V_2 \rightarrow V_3 \rightarrow V_4 \rightarrow V$$

Since p is a shortest path, we have

$$\delta(s, v_i) = \delta(s, v_{i-1}) + w(v_{i-1}, v_i) .$$

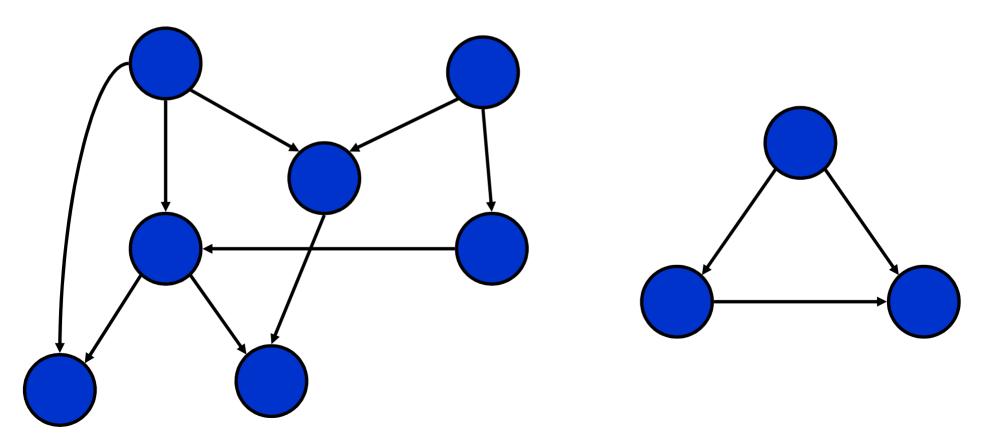
Correctness of Bellman-Ford

$$s \rightarrow V_1 \rightarrow V_2 \rightarrow V_3 \rightarrow V_4 \rightarrow V$$

- Initially, $d(v_0) = 0 = \delta(s, v_0)$, and d[s] is unchanged by subsequent relaxations (because $d[v] \ge \delta(s, v)$).
 - After 1 pass through E, we have $d(v_1) = \delta(s, v_1)$.
 - After 2 passes through *E*, we have $d(v_2) = \delta(s, v_2)$.
 - • •
 - After k passes through E, we have $d(v_k) = \delta(s, v_k)$.
- Since G contains no negative-weight cycles, p is a longest simple path that has ≤ |V| – 1 edges.

Directed Acyclic Graphs

 A directed acyclic graph or DAG is a directed graph with no directed cycles:

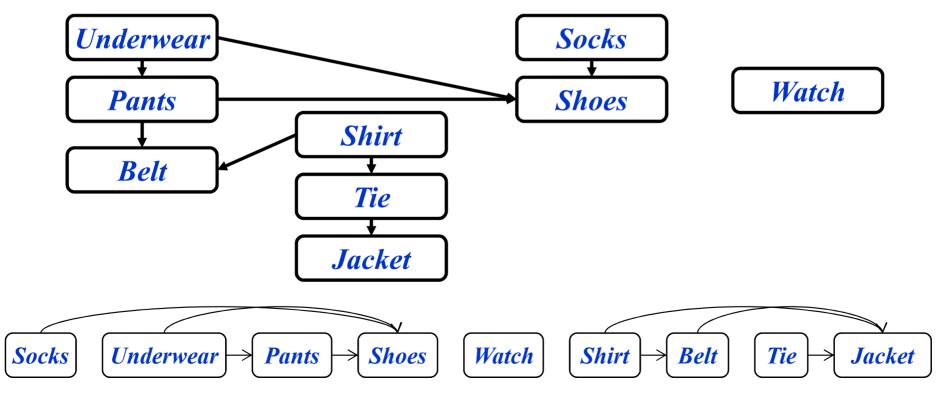


DAG Shortest Paths

- Problem: finding shortest paths in DAG
 - Bellman-Ford takes O(VE) time.
 - How can we do better?
 - Idea: use topological sort
 - If were lucky and processes vertices on each shortest path from left to right, would be done in one pass
 - Every path in a directed acyclic graph is a subsequence of topologically sorted vertex order, so processing vertexes in that order, we will do each path in forward order (will never relax edges out of vertex before doing all edges into vertex).
 - o Thus: just one pass. What will be the running time?

Review: Topological Sort

- Topological sort of a DAG (directed acyclic graph):
 - Linear ordering of all vertices in graph G such that vertex u comes before vertex v if edge $(u, v) \in G$
- Real-world example: getting dressed



DAG Shortest Paths

```
DAG-SHORTEST-PATHS(V,E,w,s)

topologically sort the vertices

INIT-SINGLE-SOURCE(V,s)

for each vertex u,

take in topologically sorted order

do for each vertex v <- Adj[u]

do RELAX(u,v,w)
```

Time complexity: O(V+E)

Source: https://zoomkoding.github.io/algorithm/2019/05/09/shortest-paths-1.html

Review: Bellman-Ford

- Running time: O(VE)
 - Not so good for large dense graphs
 - But a very practical algorithm in many ways
- Note that order in which edges are processed affects how quickly it converges (show example)
 - Using topological sort: O(V+E)

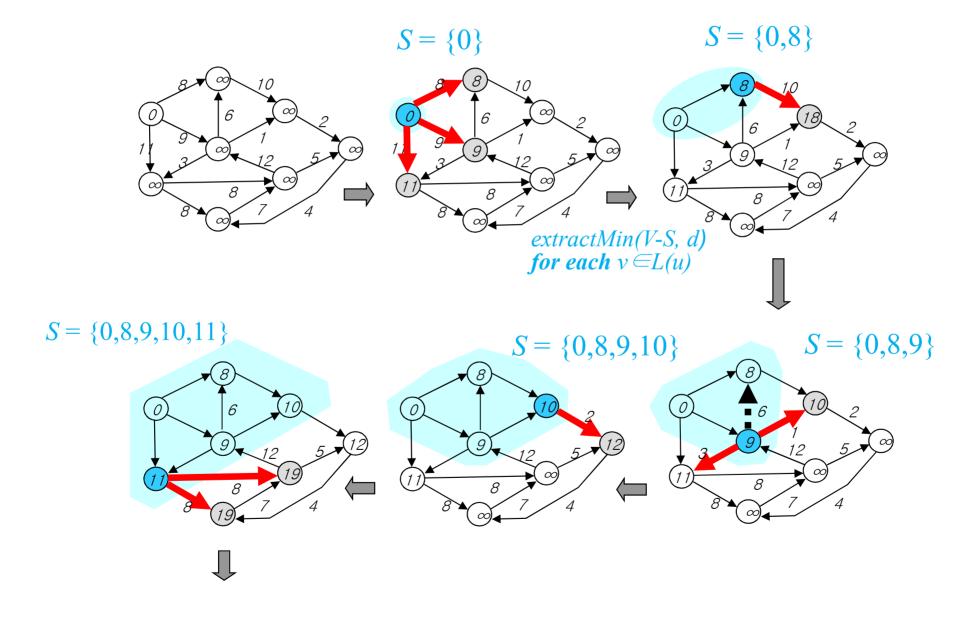
Single-Source Shortest Path without negative edge weight

DIJKSTRA'S ALGORITHM

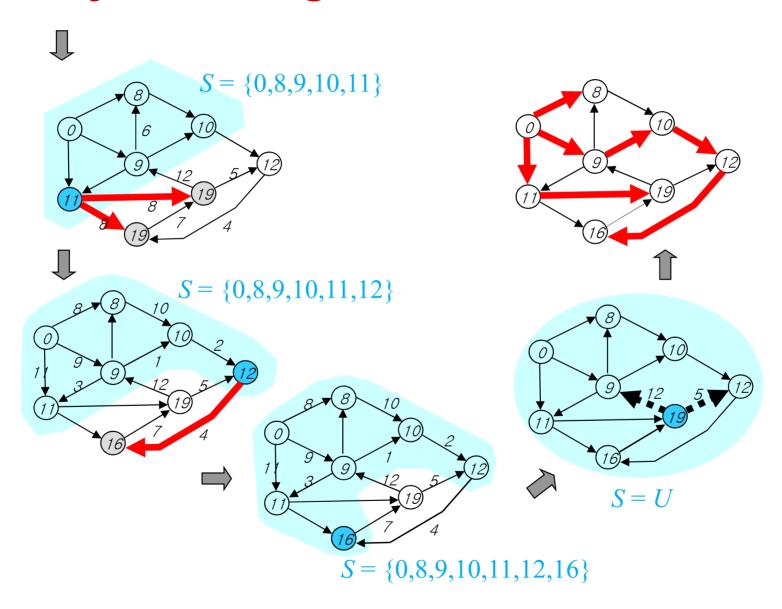
Dijkstra Algorithm

```
Dijkstra(G, r)
▷ G=(V, E): 주어진 그래프
▷ r: 시작으로 삼을 정점
                                                        모든 간선의 가중치는 음이 아니어야 함
                        ▷ S : 정점 집합
     S \leftarrow \Phi:
     for each u∈V
          d_{ij} \leftarrow \infty;
     d_r \leftarrow 0;
                      ⊳ n회 순환된다
     while (S \neq V){
          u \leftarrow extractMin(V-S, d);
          S \leftarrow S \cup \{u\};
          for each v \in L(u) \triangleright L(u) : u로부터 연결된 정점들의 집합
               if (v \subseteq V-S \text{ and } d_v < d_u + w_{u,v}) \text{ then } d_v \leftarrow d_u + w_{u,v};
                           이완(relaxation)
extractMin(Q, d)
                                                               ✔ 수행시간: O(|E|log|V|)
     집합 Q에서 d값이 가장 작은 정점 u를 리턴한다;
                                                                                  힙 이용
}
```

Dijkstra Algorithm의 작동 예



Dijkstra Algorithm의 작동 예



Dijkstra's Algorithm

```
Dijkstra(G)
     for each v \in V
                                        10
        d[v] = \infty;
    d[s] = 0; S = \emptyset; Q = V;
    while (Q \neq \emptyset)
        u = ExtractMin(Q);
                                Ex: run the algorithm
        S = S \cup \{u\};
        for each v \in u-\lambda j[]
            if (d[v] > d[u]+w(u,v))
                d[v] = d[u]+w(u,v);
Note: this
is really a
call to Q->DecreaseKey()
```

Dijkstra's Algorithm

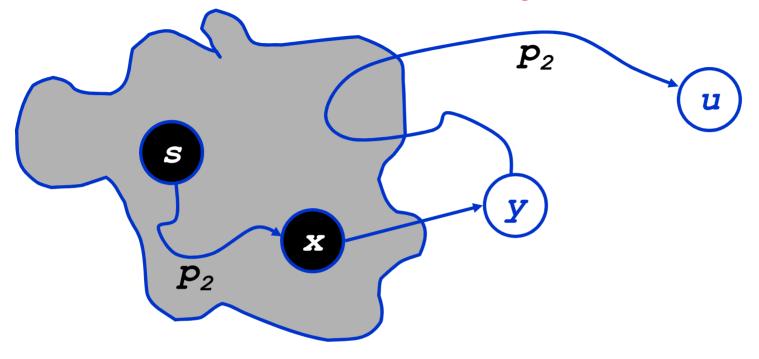
```
Dijkstra(G)
    for each v \in V
                            How many times is
       d[v] = \infty;
                            ExtractMin() called?
   d[s] = 0; S = \emptyset; Q = V;
   while (Q \neq \emptyset)
                            How many times is
       u = ExtractMin(Q); DecraseKey() called?
       S = S \cup \{u\};
       for each v \in u-Adj[]
          if (d[v] > d[u]+w(u,v))
              d[v] = d[u]+w(u,v);
What will be the total running time?
A: O(E lg V) using binary heap for Q
Can achieve O(V, lg, V, + E) with Fibonacci
```

Dijkstra's Algorithm

```
Dijkstra(G)
   for each v \in V
       d[v] = \infty;
   d[s] = 0; S = \emptyset; Q = V;
   while (Q \neq \emptyset)
       u = ExtractMin(Q);
       S = S \cup \{u\};
       for each v \in u-Adj[]
           if (d[v] > d[u]+w(u,v))
               d[v] = d[u]+w(u,v);
```

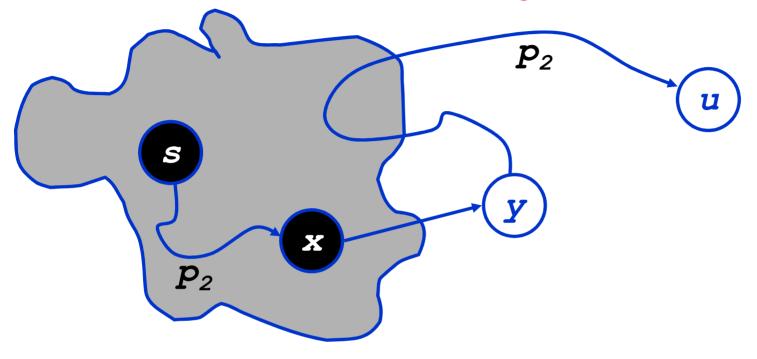
Correctness: we must show that when u is removed from Q, it has already converged
Slide credit: 홍석원, 명지대학교; 김한준, 서울시립대학교; J. Lillis, UIC

Correctness Of Dijkstra's Algorithm



- Note that $d[v] \ge \delta(s,v) \ \forall v$
- Let u be first vertex picked s.t. \exists shorter path than d[u] \Rightarrow d[u] > δ (s,u)
- Let y be first vertex \in V-S on actual shortest path from s \rightarrow u \Rightarrow d[y] = δ (s,y)
 - Because d[x] is set correctly for y's predecessor $x \in S$ on the shortest path, and
 - When we put x into S, we relaxed (x,y), giving d[y] the correct value

Correctness Of Dijkstra's Algorithm



- Note that $d[v] \ge \delta(s,v) \ \forall v$
- Let u be first vertex picked s.t. \exists shorter path than d[u] \Rightarrow d[u] > δ (s,u)
- Let y be first vertex \in V-S on actual shortest path from s \rightarrow u \Rightarrow d[y] = δ (s,y)

•
$$d[u] > \delta(s,u)$$

= $\delta(s,y) + \delta(y,u)$ (*Why?*)
= $d[y] + \delta(y,u)$
 $\geq d[y]$ But if $d[u] > d[y]$, wouldn't have chosen u. Contradiction.

문병로, 쉽게 배우는 알고리즘

Cevdet Aykanat and Mustafa Ozdal, CS473, Bilkent Univ

Andreas Klappenecker

THE FLOYD-WARSHALL ALGORITHM

All-Pairs Shortest Path Problem

- Suppose we are given a directed graph G and a weight function w:
 - G=(V,E); ω : E->R
- We assume that G does not contain cycles of weight 0 or less.
- The All-Pairs Shortest Path Problem asks to find the length of the shortest path between <u>any pair of</u> <u>vertices</u> in G.
- Applications
 - Network communications

All Pairs Shortest Paths (APSP)

given: directed graph G = (V, E), weight function $\omega: E \to R$, |V| = n

goal : create an $n \times n$ matrix $D = (d_{ij})$ of shortest path distances

i.e.,
$$d_{ij} = \delta(v_i, v_j)$$

trivial solution: run a SSSP algorithm *n* times, one for each vertex as the source.

All Pairs Shortest Paths (APSP)

- ▶ all edge weights are nonnegative : use Dijkstra's algorithm
 - PQ = linear array : O ($V^3 + VE$) = O (V^3)
 - PQ = binary heap : O ($V^2 lgV + EV lgV$) = O ($V^3 lgV$) for dense graphs
 - o better only for sparse graphs
 - PQ = fibonacci heap : O ($V^2 \lg V + EV$) = O (V^3) for dense graphs
 - o better only for sparse graphs
- ▶ negative edge weights : use Bellman-Ford algorithm
 - O (V^2E) = O (V^4) on dense graphs

Floyd-Warshall Algorithm

- A dynamic programming solution that solved the APSP problem with time complexity $O(n^3)$ for a graph with n vertices.
 - Dynamic programming

Adjacency Matrix Representation of Graphs

 $\triangleright n \times n$ matrix $\mathbf{W} = (\omega_{ij})$ of edge weights :

$$\omega(v_i, v_j) \text{ if } (v_i, v_j) \in E$$

$$\omega_{ij} = \begin{cases} \infty & \text{if } (v_i, v_j) \notin E \end{cases}$$

- ightharpoonup assume $\omega_{ii} = 0$ for all $v_i \in V$, because
 - no neg-weight cycle
 - \Rightarrow shortest path to itself has no edge,

i.e.,
$$\delta (v_i, v_i) = 0$$

Intermediate Vertices

- Without loss of generality, we will assume that $V=\{1,2,...,n\}$, i.e., that the vertices of the graph are numbered from 1 to n.
- Given a path $p=(v_1, v_2,..., v_m)$ in the graph, we will call the vertices v_k with index k in $\{2,...,m-1\}$ the intermediate vertices of path p.
- If k is an intermediate vertex of path p, then we break p down into:
 - $i \rightarrow k \rightarrow j$ (p1 and p2)
 - p1 is a shortest path from i to k with all intermediate vertices in the set {1,2,...,k-1}
 - p2 is a shortest path from k to j with all intermediate vertices in the set {1,2,...,k-1}

All intermediate vertices in {1, 2, ..., *k*-1}

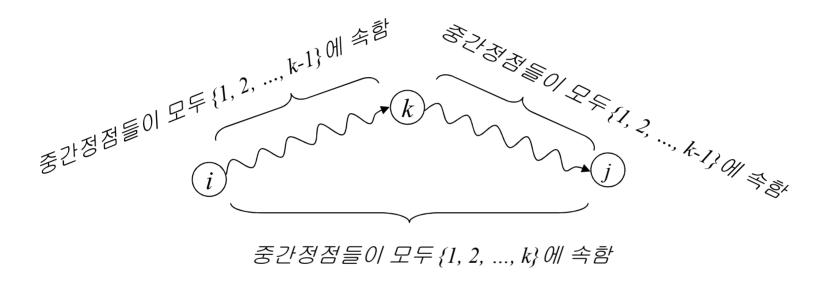


Figure 2. Path p is a shortest path from vertex i to vertex j, and k is the highest-numbered intermediate vertex of p. Path p1, the portion of path p from vertex i to vertex k, has all intermediate vertices in the set $\{1, 2, ..., k-1\}$. The same holds for path p2 from vertex k to vertex j.

Key Idea

- Let $d_{ij}^{(k)}$ denote the length of the shortest path from i to j such that all intermediate vertices are contained in the set $\{1,...,k\}$.
- Consider a shortest path *p* from *i* to j such that the intermediate vertices are from the set {1,...,*k*}.
 - If the vertex k is not an intermediate vertex on p, then $d_{ij}^{(k)} = d_{ij}^{(k-1)}$.
 - If the vertex k is an intermediate vertex on p, then $d_{ij}^{(k)} = d_{ik}^{(k-1)} + d_{ki}^{(k-1)}$.
 - Interestingly, in either case, the subpaths contain merely nodes from $\{1,...,k-1\}$.

Recursive Formulation

If we do not use intermediate nodes, i.e., when k=0, then

$$d_{ij}^{(0)} = W_{ij}$$

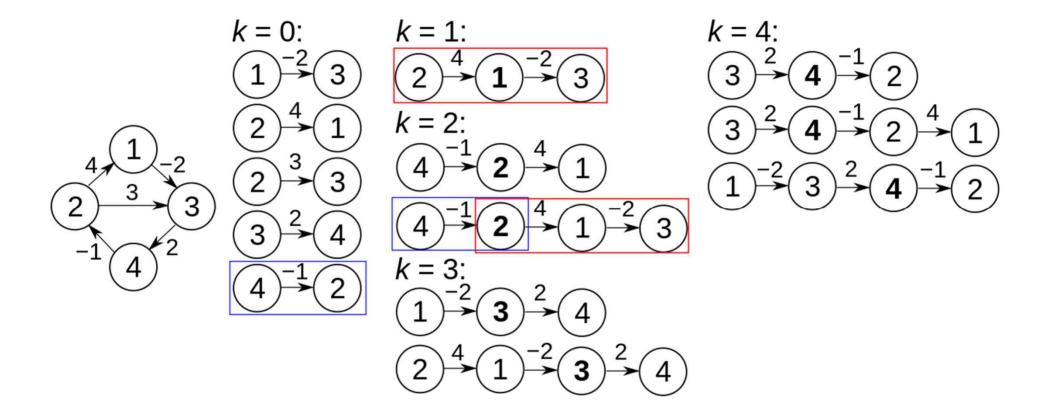
If k>0, then

$$d_{ij}^{(k)} = \min\{d_{ij}^{(k-1)}, d_{ik}^{(k-1)} + d_{kj}^{(k-1)}\}$$

The Floyd-Warshall Algorithm

```
Floyd-Warshall(W)
n = # of rows of W;
D^{(0)} = W; /* initialization, equivalent to the next nested loop */
/* for (i,j) = (1, 1) to (n, n) do; d_{ii}^{(k)} = w_{ij}; od; */
for k = 1 to n do
    for i = 1 to n do
          for j = 1 to n do
                     d_{ii}^{(k)} = \min\{d_{ii}^{(k-1)}, d_{ik}^{(k-1)} + d_{ki}^{(k-1)}\};
          od;
    od;
                              ✔ d^{(k)}_{ii} : 중간 정점으로 정점 집합 {1, 2, ..., k} 만을
                                  사용하여 정점 i 에서 정점 j 에 이르는 최단경로
od;
return D<sup>(n)</sup>;
```

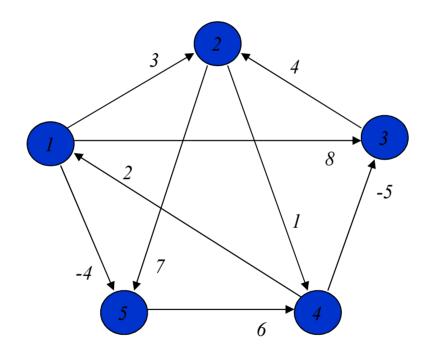
Example: Floyd-Warshall Algorithm

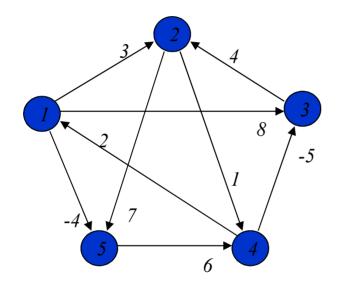


Alternate Implementation of the Floyd-Warshall

```
EXTEND (D, W)

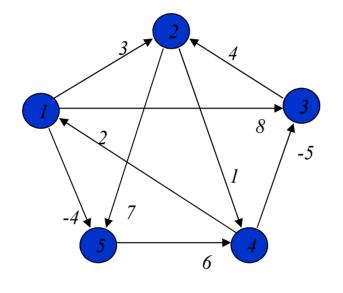
 D = (d_{ij}) \text{ is an n x n matrix} 
 \text{for } i \leftarrow 1 \text{ to } n \text{ do} 
 \text{for } j \leftarrow 1 \text{ to } n \text{ do} 
 d_{ij} \leftarrow \infty 
 \text{for } k \leftarrow 1 \text{ to } n \text{ do} 
 d_{ij} \leftarrow \min\{d_{ij}, d_{ik} + \omega_{kj}\} 
 \text{return D}
```





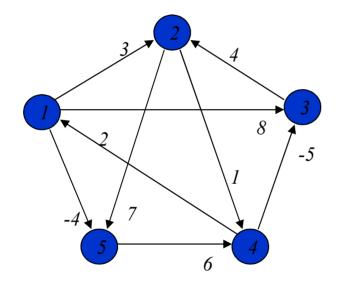
	1	2	3	4	5
1	0	3	8	8	-4
2	8	0	8	1	7
3	8	4	0	8	8
4	2	8	-5	0	∞
5	∞	∞	∞	6	0

$$D^{l} = D^{0}W$$



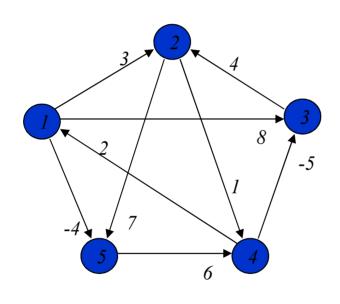
	1	2	3	4	5
1	0	3	8	2	-4
2	3	0	-4	1	7
3	8	4	0	5	11
4	2	-1	-5	0	-2
5	8	∞	1	6	0

$$D^2 = D^I W$$



	1	2	3	4	5
1	0	3	-3	2	-4
2	3	0	-4	1	-1
3	7	4	0	5	11
4	2	-1	-5	0	-2
5	8	5	1	6	0

$$D^3 = D^2 W$$



	1	2	3	4	5
1	0	1	-3	2	-4
2	3	0	-4	1	-1
3	7	4	0	5	3
4	2	-1	-5	0	-2
5	8	5	1	6	0

$$D^4 = D^3 W$$

Time and Space Requirements

The running time is obviously $O(n^3)$.

However, in this version, the space requirements are high. One can reduce the space from O(n³) to O(n²) by using a single array d.

Next topic: Minimum Spanning Trees

END OF LECTURE 14