

# CMSC 451 - Algorithm Design

## Lecture 11 - All-pairs Shortest Paths + Floyd-Warshall

Earlier, we covered the single-source shortest path problem for edge-weighted digraphs.

Dijkstra -  $O(m + n \log n)$   $n = |V|$   $m = |E|$

Bellman-Ford -  $O(n \cdot m)$  [allows neg. edge weights]

### Notation:

- The cost of a path is sum of edge weights
- The distance between two vertices  $u$  +  $v$  is min. cost of any path from  $u$  to  $v$ .

### All-pairs shortest paths:

Given digraph  $G = (V, E)$ . Each edge  $(u, v) \in E$  has weight  $w(u, v)$ , may be negative (but no neg. cost cycles), compute distance between all pairs of vertices  $u, v \in V$ .

We'll present Floyd-Warshall algorithm.  
DP-based,  $O(n^3)$  time.

## Notes:

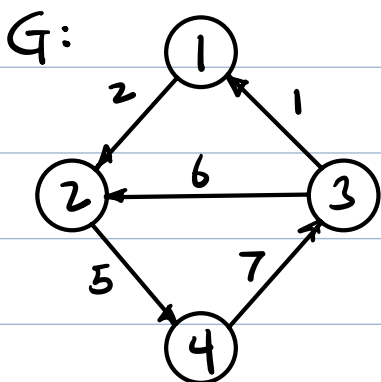
- Discovered independently by Robert Floyd + Stephen Warshall (mid 1960's).
- First discovered (but not credited) by Bernard Roy.
- General DP structure works for many reachability problems:
  - Transitive closure of a binary relation
  - Convert finite state automaton to regular expression.
  - Compute max capacity paths

## Input/Output Representation:

Input: (Augmented) adjacency matrix:

$$V = \{1, 2, \dots, n\}$$

$$w_{ij} = \begin{cases} 0 & \text{if } i=j \\ w(i,j) & \text{if } i \neq j, (i,j) \in E \\ +\infty & \text{if } i \neq j, (i,j) \notin E \end{cases}$$



w:

	1	2	3	4
1	0	2	$\infty$	$\infty$
2	$\infty$	0	$\infty$	5
3	1	6	0	$\infty$
4	$\infty$	$\infty$	7	0

## Output: Distance matrix

$d_{ij}$  = distance from  $i$  to  $j$   
( $\infty$  if no path)

$d$ :

	1	2	3	4
1	0	2	14	7
2	13	0	12	5
3	1	3	0	8
4	8	10	7	0

→ Space =  $O(n^2)$

Paths?  $O(n^2)$  paths  $\Rightarrow O(n^3)$  storage?

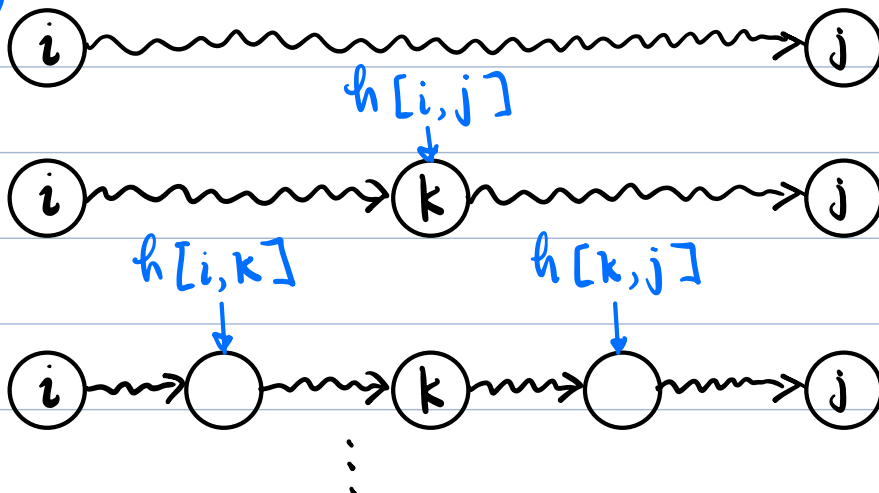
Clever trick - Reduces storage to  $O(n^2)$

- "Hook" matrix  $H = h[1..n, 1..n]$

- For each  $i, j \in V$

$h[i, j] = \begin{cases} \emptyset & \text{if shortest path is direct (edge } (i, j)) \\ k & k = \text{any vertex along the shortest path} \end{cases}$

Fill in the path recursively



## Floyd-Warshall Algorithm:

- How to reduce shortest paths to smaller subproblems?

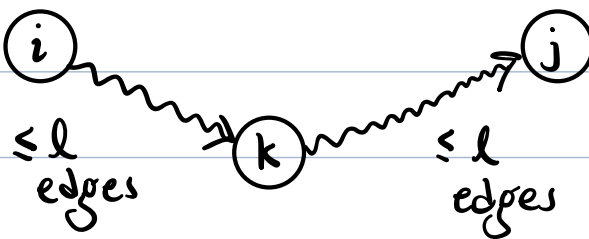
- Obvious? (But not best!)

- Restrict path length

e.g.  $d_{ij}^{(l)}$  = shortest path from  $i$  to  $j$  using  $\leq l$  edges

- Build by doubling:

$$d_{ij}^{(2l)} = \min_k (d_{ik}^{(l)} + d_{kj}^{(l)})$$



- This leads to a slower algorithm.

- Floyd-Warshall insight:

- Don't restrict length, restrict which vertices you can go through.

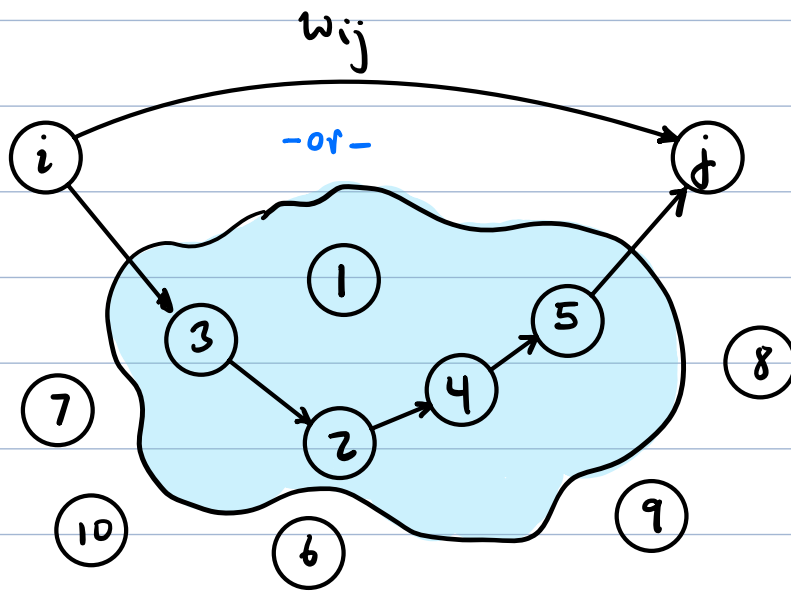
- Given a path  $\langle v_1, v_2, \dots, v_{l-1}, v_l \rangle$   
 $v_2 \dots v_{l-1}$  are the intermediate vertices.

- for  $1 \leq i, j \leq n$  &  $0 \leq k \leq n$ ,  
define:

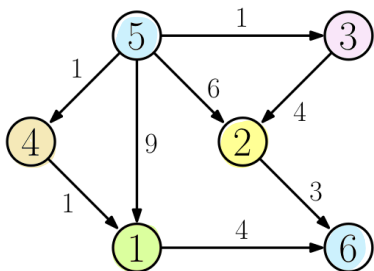
$d_{ij}^{(k)}$  = cost of shortest path  
from  $i$  to  $j$  with intermediate  
vertices from  $\{1, \dots, k\}$ .

Eg.  $k=5$

$d_{ij}^{(5)}$ :



Approach -  $d_{ij}^{(0)} = w_{ij}$  (no intermediates  $\Rightarrow$  edge)  
for  $k=1$  to  $n$   
compute  $d_{ij}^{(k)}$  for all  $1 \leq i, j \leq n$ .



$$d_{5,6}^{(0)} = \infty \text{ (no path)}$$

$$d_{5,6}^{(1)} = 13 \langle 5, 1, 6 \rangle$$

$$d_{5,6}^{(2)} = 9 \langle 5, 2, 6 \rangle$$

$$d_{5,6}^{(3)} = 8 \langle 5, 3, 2, 6 \rangle$$

$$d_{5,6}^{(4)} = 6 \langle 5, 4, 1, 6 \rangle$$

$$d_{5,6}^{(5)} = d_{5,6}^{(6)} = 6 \text{ (no change)}$$

## DP Formulation:

Basis:  $d_{ij}^{(0)} = w_{ij}$  (no intermediates  $\Rightarrow$  edge weight)

$k \geq 1$ : Cases:

- Shortest path does not pass through  $k$ :

$\Rightarrow$  only intermediates are  $\{1, \dots, k-1\}$

$$\Rightarrow d_{ij}^{(k)} \leftarrow d_{ij}^{(k-1)}$$

- Shortest path passes through  $k$ :

- Since no neg. cost cycles, visits  $k$  once

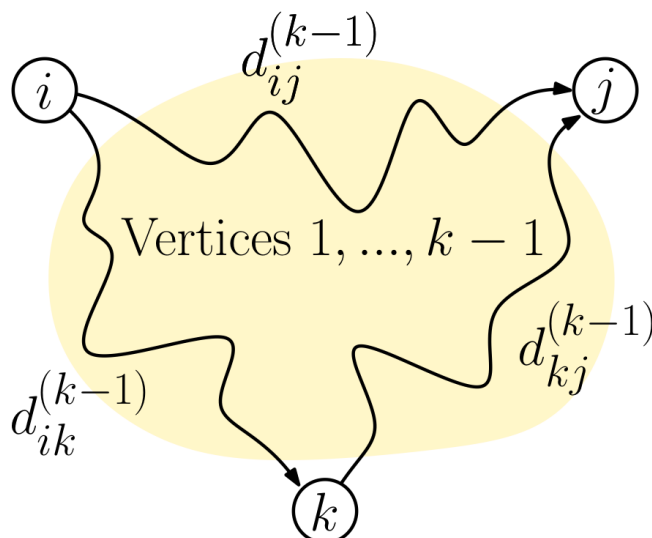
- Path goes  $i \rightsquigarrow k$  then  $k \rightsquigarrow j$

- These paths use intermediates  $\{1, \dots, k-1\}$

$$\Rightarrow d_{ij}^{(k)} \leftarrow d_{ik}^{(k-1)} + d_{kj}^{(k-1)}$$

Principle of optimality holds

$k$  may be equal to  $i$  or  $j$ . Check that this is still correct



DP Credo - Try all options + take best

$$d_{ij}^{(k)} = \begin{cases} w_{ij} & \text{if } k=0 \\ \min \begin{cases} d_{ij}^{(k-1)} \\ d_{ik}^{(k-1)} + d_{kj}^{(k-1)} \end{cases} & \text{if } k \geq 1 \end{cases}$$

### Bottom-Up Implementation

- Build matrix  $d[1..n, 1..n]$
- Matrix of "hooks" to save decisions

floyd-warshall ( $w[1..n, 1..n]$ )

```
for ( $1 \leq i, j \leq n$ ) // init  $d^{(0)}$ 
   $d[i, j] \leftarrow w[i, j]$ 
   $h[i, j] \leftarrow \emptyset$ 
for ( $k = 1$  to  $n$ ) // compute  $d^{(k)}$ 
  for ( $1 \leq i, j \leq n$ )
     $newCost \leftarrow d[i, k] + d[k, j]$ 
    if ( $newCost < d[i, j]$ ) // better to go
       $d[i, j] \leftarrow newCost$  // through  $k$ 
       $h[i, j] \leftarrow k$  // update hook
return  $d + h$  matrices // return dists + hooks
```

Running time:  $O(n^3)$

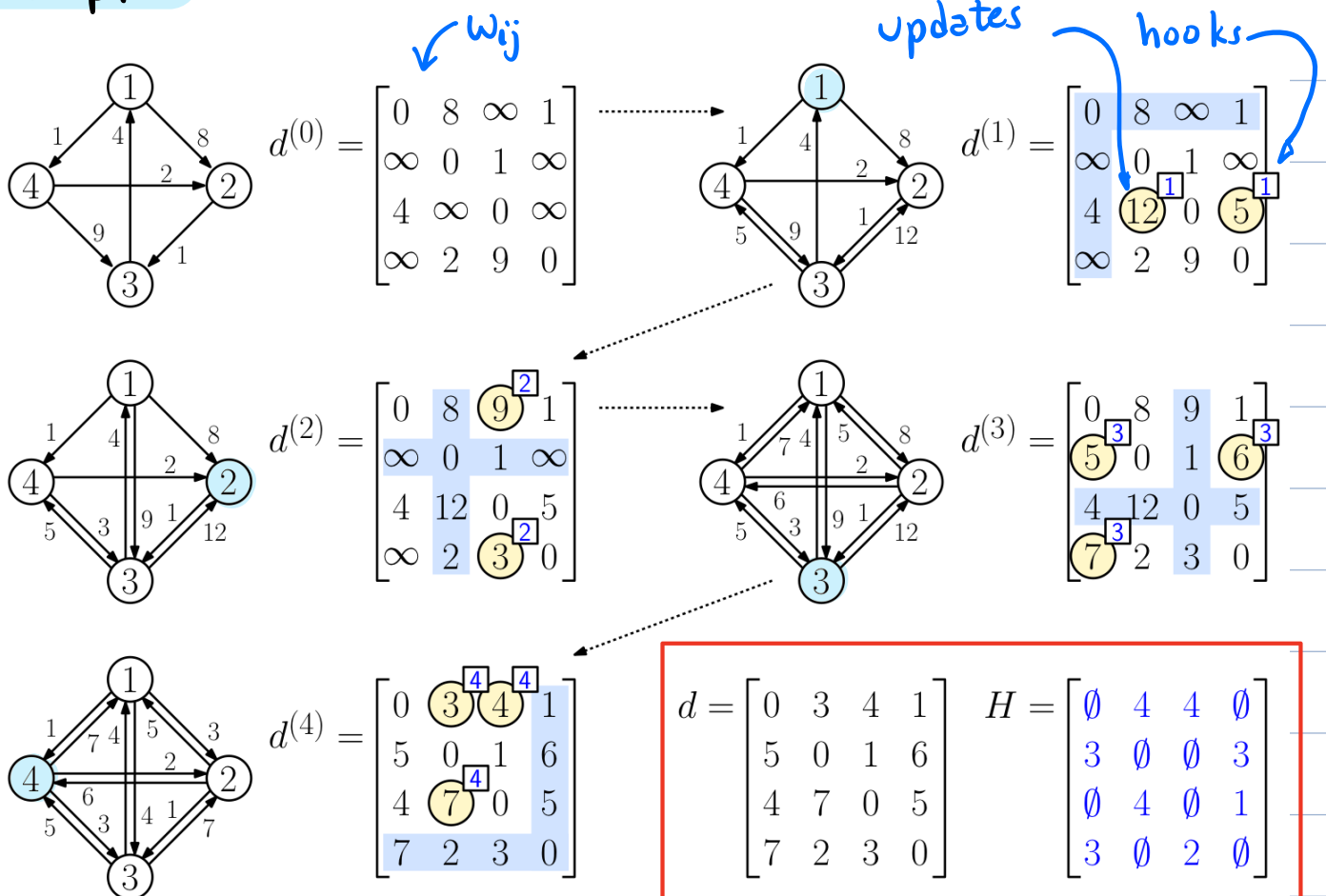


Huh?



- DP formulation had three parameters:  $i, j, k$  but algorithm does not use  $k$ .
- We only use  $k + k - 1$ , so we really only need two copies of  $d_{ij}$  matrix
  - current + previous
- In fact, we can show that only one matrix suffices (see pdf notes for details)

Example:





## Extracting the Shortest Path:

- As mentioned above, we use  $h[i,j]$  to recursively extract path
- If  $d[i,j] = \infty \Rightarrow$  no path otherwise  $\text{get-path}(i,j)$  yields edges on shortest path.

```
get-path(i, j) // path from i to j
if (h[i,j] = ∅) // direct path
  return (i, j) // path is single edge
else
  mid ← h[i,j] // i → mid → j
  return get-path(i, mid) ⊕ get-path(mid, j)
```

concatenate

## Summary:

- Floyd-Warshall all-pairs shortest paths
- $O(n^3)$  time /  $O(n^2)$  space
- General structure can be applied to many similar path-related problems.