



香港中文大學 (深圳)  
The Chinese University of Hong Kong

# CSC3100 Data Structures

## Lecture 18: Graph shortest path

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# Outline

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- ▶ Single-Source Shortest Path Algorithm
  - Non-negative weights: Dijkstra's algorithm
  - Non-negative and negative weights: Bellman-Ford algorithm
- ▶ All-Pair Shortest Path Algorithm
  - Floyd's algorithm



# Bellman-Ford Algorithm

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- ▶ Single-source shortest path problem
  - Computes  $\delta(s, v)$  and  $p[v]$  for all  $v \in V$
- ▶ Allows negative edge weights - can detect negative cycles
  - Returns **TRUE** if no negative-weight cycles are reachable from the source  $s$
  - Returns **FALSE** otherwise  $\Rightarrow$  no solution exists

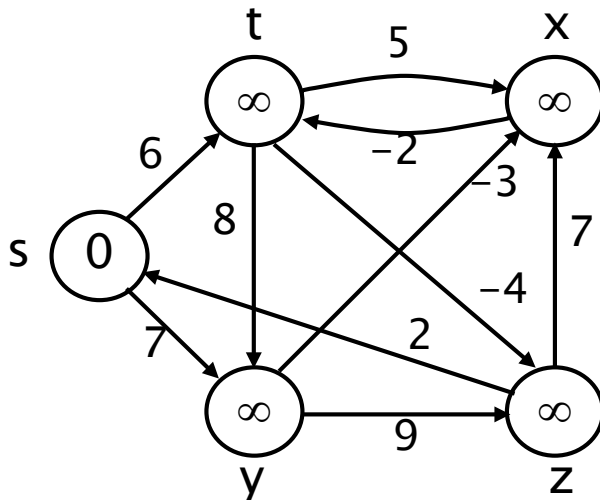


# Bellman-Ford Algorithm (cont'd)

## ► Idea:

- Each edge is relaxed  $|V|-1$  times by making  $|V|-1$  passes over the whole edge set
- Any path will contain at most  $|V|-1$  edges

Edge order: (t, x), (t, y), (t, z), (x, t), (y, x), (y, z), (z, x), (z, s), (s, t), (s, y)

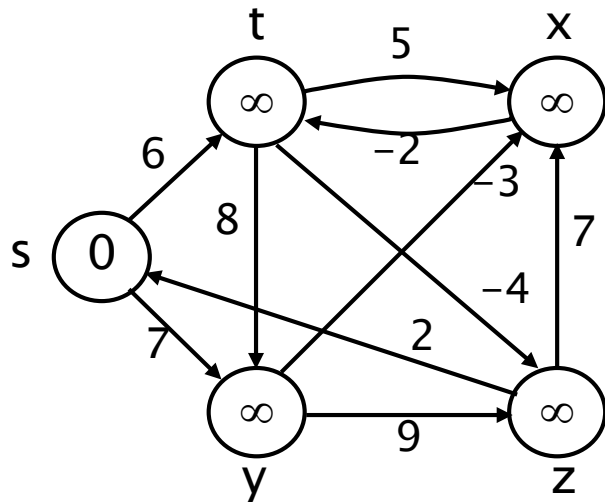


### Relaxation:

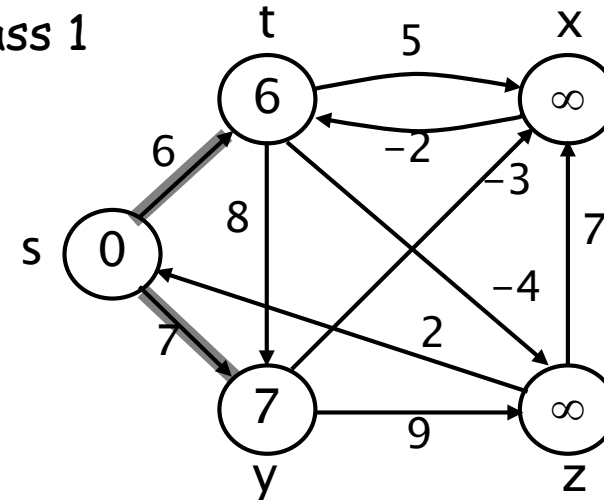
If  $d[v] > d[u] + w(u, v)$   
 $\Rightarrow d[v] = d[u] + w(u, v)$



# BELLMAN-FORD( $V, E, w, s$ )



Pass 1

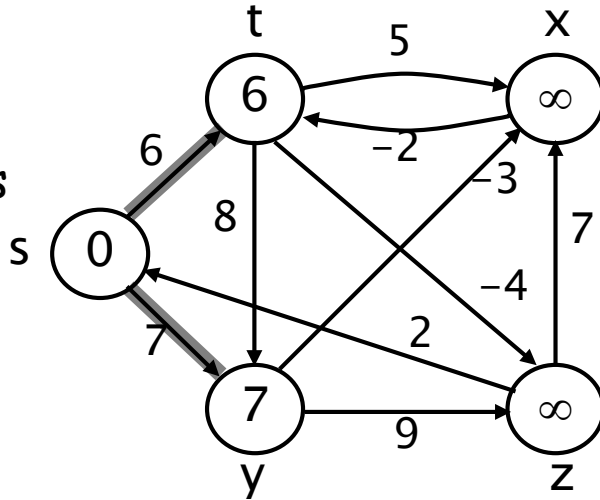


Edge order:  $(t, x), (t, y), (t, z), (x, t), (y, x), (y, z), (z, x), (z, s), (s, t), (s, y)$

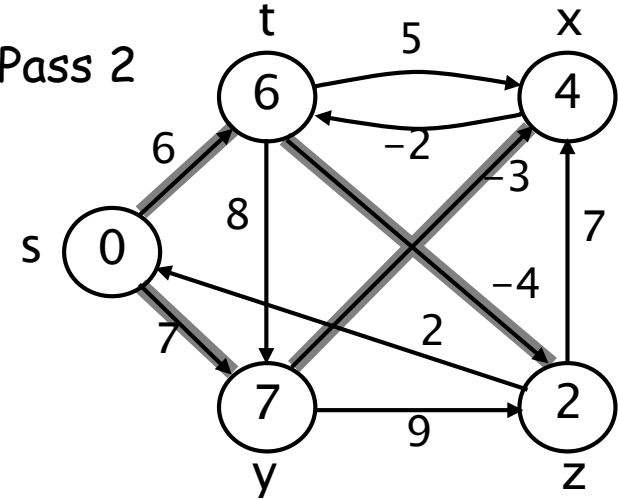


# Example

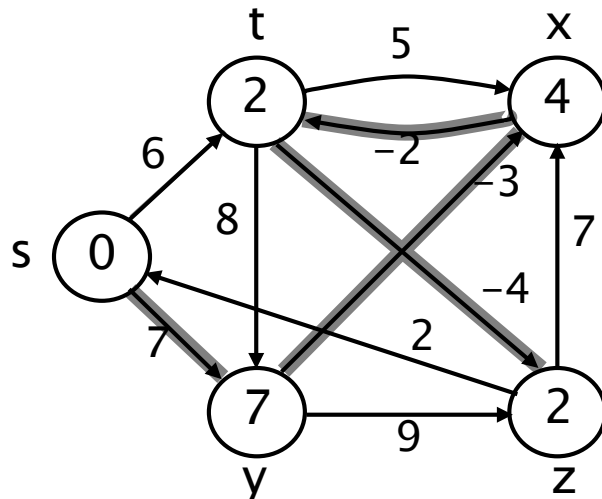
Pass 1  
(from  
previous  
slide)



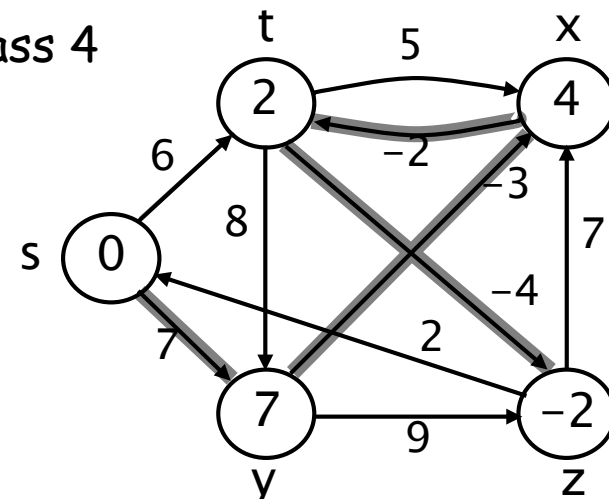
Pass 2



Pass 3



Pass 4

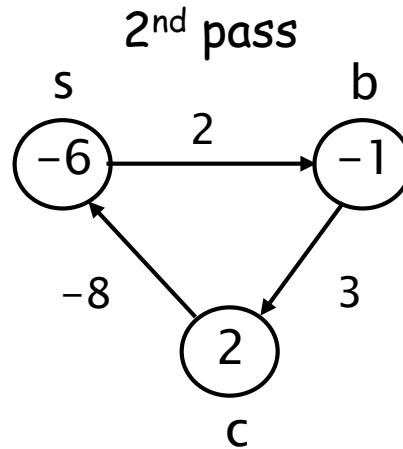
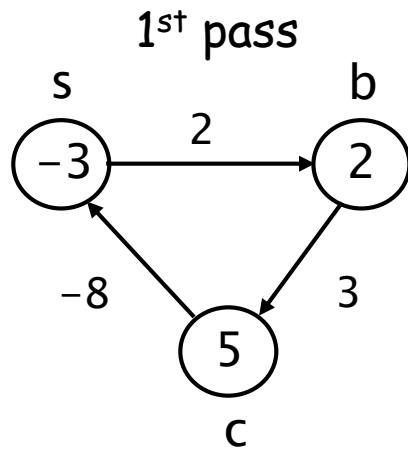
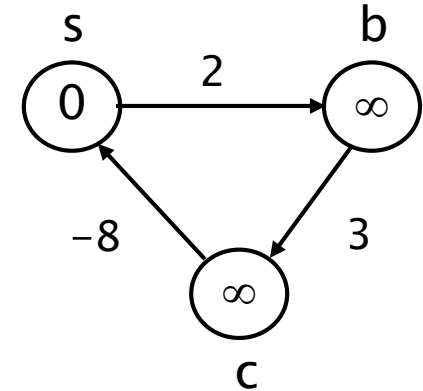


Edge order: (t, x), (t, y), (t, z), (x, t), (y, x), (y, z), (z, x), (z, s), (s, t), (s, y)



# Detecting Negative Cycles (perform extra test after $V-1$ iterations)

- ▶ for each edge  $(u, v) \in E$
- ▶     do if  $d[v] > d[u] + w(u, v)$
- ▶     then return FALSE
- ▶ return TRUE



$(s,b) (b,c) (c,s)$

Look at edge  $(s, b)$ :

$$d[b] = -1$$

$$d[s] + w(s, b) = -4$$

$$\Rightarrow d[b] > d[s] + w(s, b)$$



# BELLMAN-FORD( $V, E, w, s$ )

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1. INITIALIZE-SINGLE-SOURCE( $V, s$ )  $\leftarrow \Theta(|V|)$
  2. **for**  $i \leftarrow 1$  **to**  $|V| - 1$   $\leftarrow O(|V|)$
  3.     **do for** each edge  $(u, v) \in E$   $\leftarrow O(|E|)$
  4.         **do** RELAX( $u, v, w$ )
  5. **for** each edge  $(u, v) \in E$   $\leftarrow O(|E|)$
  6.     **do if**  $d[v] > d[u] + w(u, v)$
  7.         **then return** FALSE
  8. **return** TRUE
- $\left. \begin{array}{l} \leftarrow O(|V|) \\ \leftarrow O(|E|) \end{array} \right\} O(|V||E|)$

Running time:  $O(|V| + |V||E| + |E|) = O(|V||E|)$





# Key points of BELLMAN-FORD

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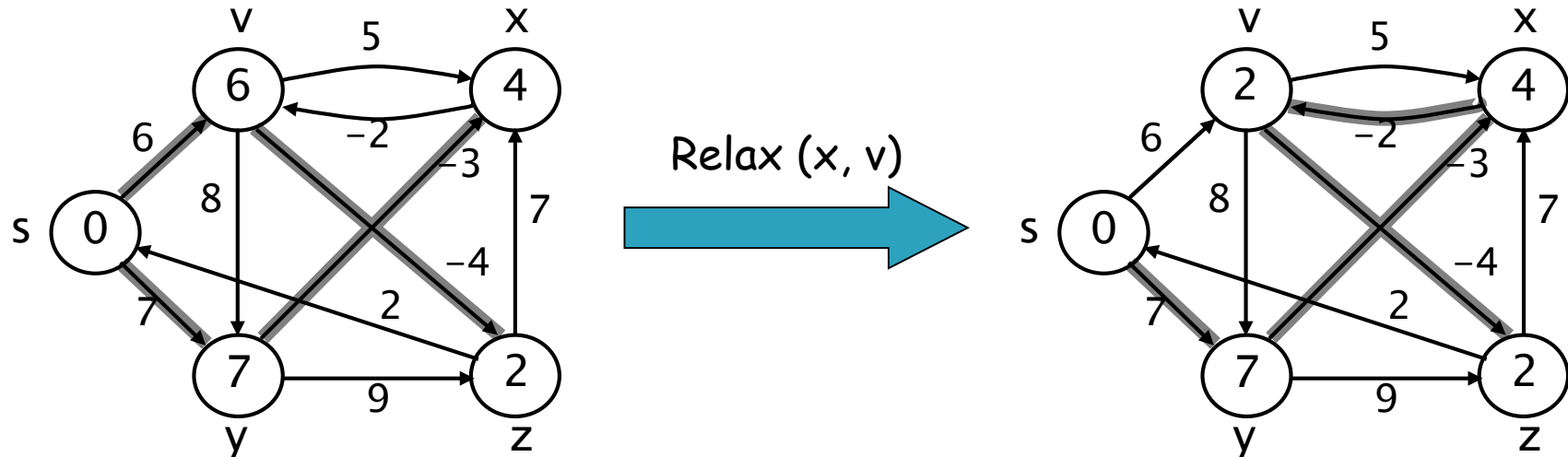
- ▶ After  $|V|-1$  iterations,  $d$  values will not be updated or can't be lower any more, and  $d$  values store the measure of the shortest path. Why?
  - Using a counter example to help you do the analysis
  - How to prove its correctness?



# Shortest Path Properties

## ► Upper-bound property

- We always have  $d[v] \geq \delta(s, v)$  for all  $v$
- The estimate never goes up - relaxation only lowers the estimate

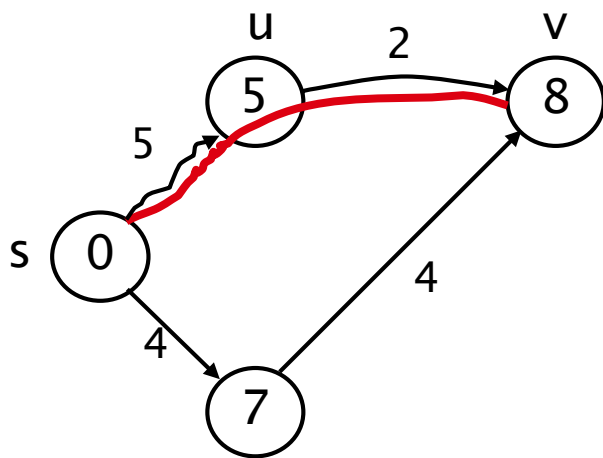




# Shortest Path Properties

## ► Convergence property

If  $s \rightsquigarrow u \rightarrow v$  is a shortest path, and if  $d[u] = \delta(s, u)$  at any time prior to relaxing edge  $(u, v)$ , then  $d[v] = \delta(s, v)$  at all times after relaxing  $(u, v)$



- If  $d[v] > \delta(s, v) \Rightarrow$  after relaxation:  
 $d[v] = d[u] + w(u, v)$   
 $d[v] = 5 + 2 = 7$
- Otherwise, the value remains unchanged, because it must have been the shortest path value

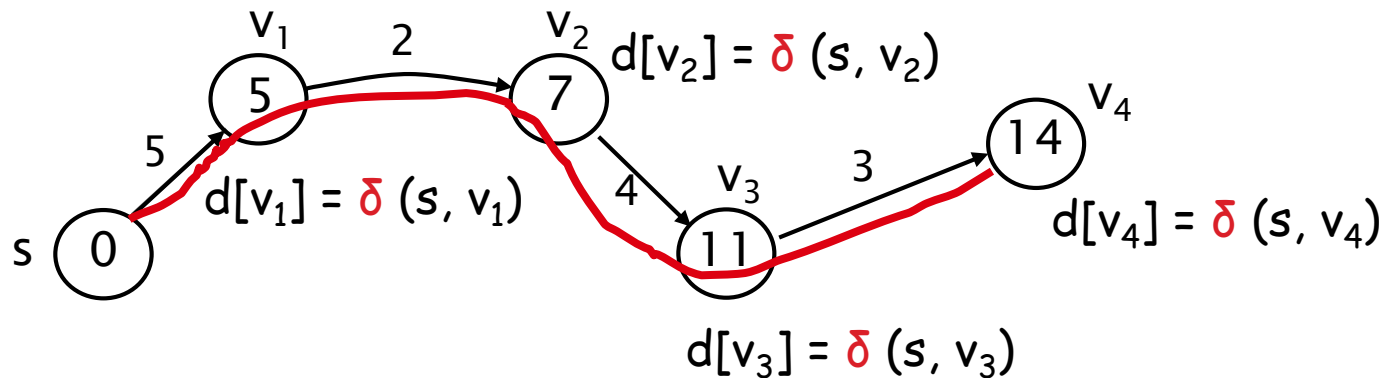


# Shortest Path Properties

## ► Path relaxation property

Let  $p = \langle v_0, v_1, \dots, v_k \rangle$  be a shortest path from  $s = v_0$  to  $v_k$

If we relax, in order,  $(v_0, v_1), (v_1, v_2), \dots, (v_{k-1}, v_k)$ , even intermixed with other relaxations, then  $d[v_k] = \delta(s, v_k)$





# Correctness of Belman-Ford Algorithm

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- ▶ **Theorem:** Show that  $d[v] = \delta(s, v)$ , for every  $v$ , after  $|V| - 1$  passes

Case 1:  $G$  does not contain negative cycles which are reachable from  $s$

- Assume that the shortest path from  $s$  to  $v$  is  $p = \langle v_0, v_1, \dots, v_k \rangle$ , where  $s = v_0$  and  $v = v_k$ ,  $k \leq |V| - 1$
- Use mathematical induction on the number of passes  $i$  to show that:
$$d[v_i] = \delta(s, v_i) , i=0,1,\dots,k$$

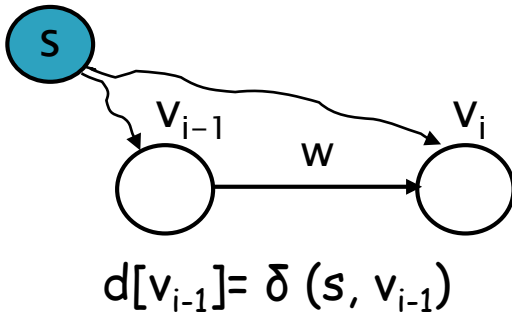


# Correctness of Belman-Ford Algorithm (cont.)

**Base Case:**  $i=0, d[v_0] = \delta(s, v_0) = \delta(s, s) = 0$

**Inductive Hypothesis:**  $d[v_{i-1}] = \delta(s, v_{i-1})$

**Inductive Step:**  $d[v_i] = \delta(s, v_i)$



After relaxing  $(v_{i-1}, v_i)$  (convergence property) :

$$d[v_i] \leq d[v_{i-1}] + w = \delta(s, v_{i-1}) + w = \delta(s, v_i)$$

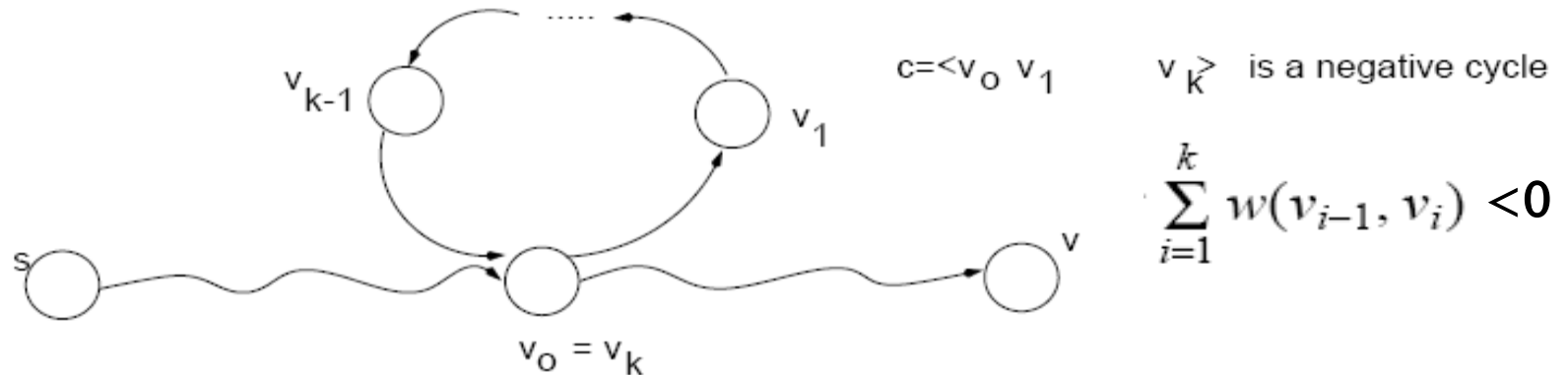
From the upper bound property:  $d[v_i] \geq \delta(s, v_i)$

Therefore,  $d[v_i] = \delta(s, v_i)$



# Correctness of Belman-Ford Algorithm (cont.)

- ▶ Case 2:  $G$  contains a negative cycle which is reachable from  $s$



**Proof by Contradiction:**  
suppose the algorithm returns a solution

After relaxing  $(v_{i-1}, v_i)$ :  $dist[v_i] \leq dist[v_{i-1}] + w(v_{i-1}, v_i)$

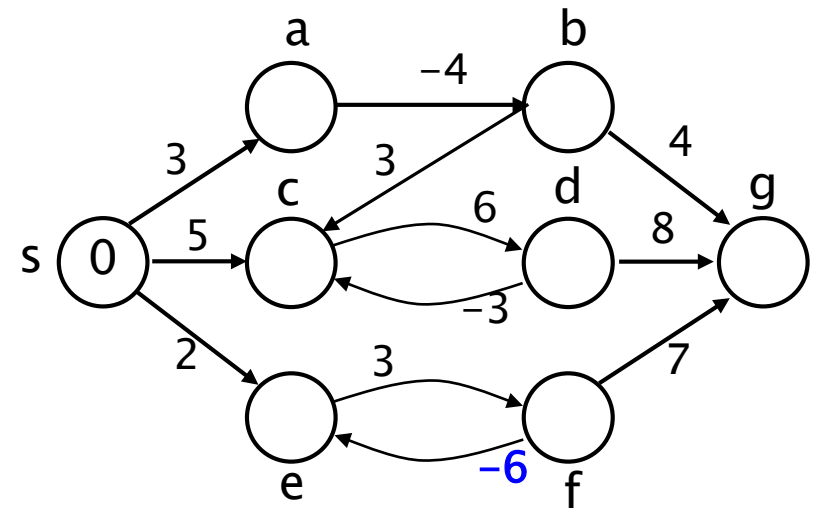
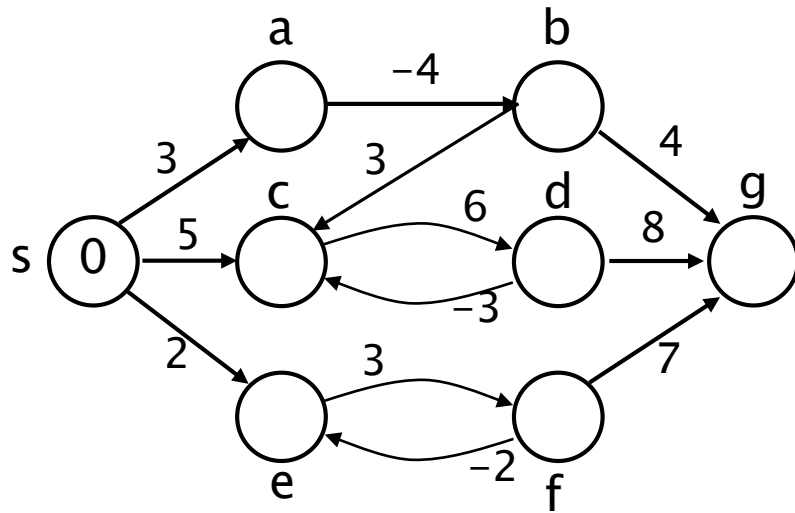
$$\Rightarrow \sum_{i=1}^k dist[v_i] \leq \sum_{i=1}^k dist[v_{i-1}] + \sum_{i=1}^k w(v_{i-1}, v_i)$$

$$\Rightarrow \sum_{i=1}^k w(v_{i-1}, v_i) \geq 0 \quad (\sum_{i=1}^k dist[v_i] = \sum_{i=1}^k dist[v_{i-1}])$$

**Contradiction!**



# Exercise 1







# Floyd's Algorithm

(all pairs shortest paths)



# All pairs shortest path

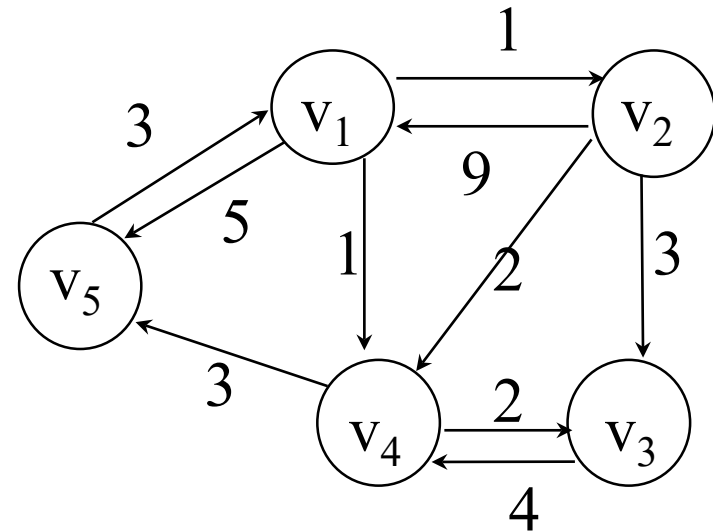
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- ▶ The **graph**: may contain negative edges but no negative cycles
- ▶ A **representation**: a weight matrix where
  - $W(i,j)=0$  if  $i=j$
  - $W(i,j)=\infty$  if there is no edge between  $i$  and  $j$
  - $W(i,j)$ ="weight of edge"
- ▶ The **problem**: find the shortest path between every pair of vertices of a graph
- ▶ Note: we have shown principle of optimality applies to shortest path problems



# The weight matrix and the graph

	1	2	3	4	5
1	0	1	$\infty$	1	5
2	9	0	3	2	$\infty$
3	$\infty$	$\infty$	0	4	$\infty$
4	$\infty$	$\infty$	2	0	3
5	3	$\infty$	$\infty$	$\infty$	0





# A straightforward method

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- ▶ A naive method is to run a single-source shortest path algorithm for each vertex
  - Run Dijkstra's algorithm  $|V|$  times
  - Dijkstra's algorithm's time complexity:  $O(|E| \times \lg|V|)$
  - Total time cost:  $O(|V| \times |E| \times \lg|V|)$
- ▶ Floyd's algorithm
  - Total time cost:  $O(|V|^3)$
  - For dense subgraphs, Floyd's algorithm is faster
  - It is easier to implement



# The subproblems

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- ▶ How can we define the shortest distance  $d_{ij}$  in terms of "smaller" problems?
- ▶ One way is to restrict the paths to only include vertices from a restricted subset
- ▶ Initially, the subset is empty
- ▶ Then, it is incrementally increased until it includes all the vertices



# The subproblems

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- ▶ Let  $D^{(k)}[i,j]$  = weight of a shortest path from  $v_i$  to  $v_j$  using only vertices from  $\{v_1, v_2, \dots, v_k\}$  as intermediate vertices in the path
  - $D^{(0)} = W$
  - $D^{(n)} = D$  which is the goal matrix
- ▶ How do we compute  $D^{(k)}$  from  $D^{(k-1)}$  ?

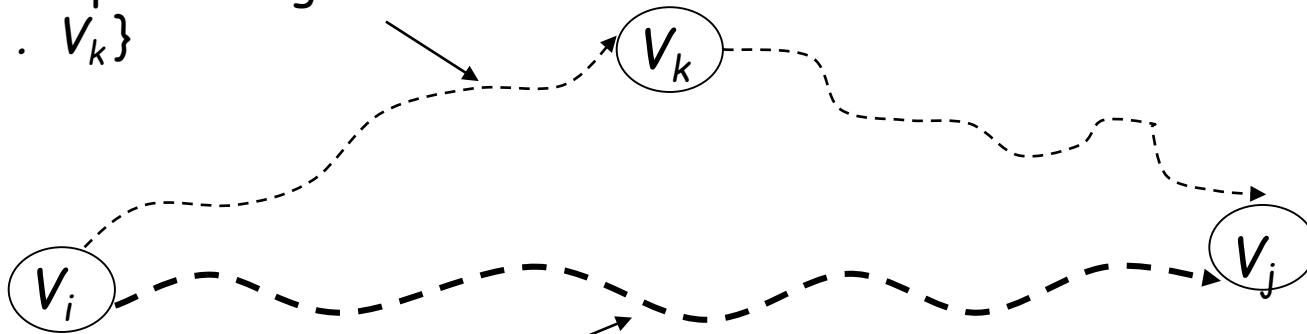


# The Recursive Definition:

Case 1: A shortest path from  $v_i$  to  $v_j$  restricted to using only vertices from  $\{v_1, v_2, \dots, v_k\}$  as intermediate vertices does not use  $v_k$  Then  $D^{(k)}[i,j] = D^{(k-1)}[i,j]$

Case 2: A shortest path from  $v_i$  to  $v_j$  restricted to using only vertices from  $\{v_1, v_2, \dots, v_k\}$  as intermediate vertices does use  $v_k$  Then  $D^{(k)}[i,j] = D^{(k-1)}[i,k] + D^{(k-1)}[k,j]$

Shortest path using intermediate vertices  $\{V_1, \dots, V_k\}$



Shortest path using intermediate vertices  $\{V_1, \dots, V_{k-1}\}$



# The recursive definition

► Since

$$D^{(k)}[i,j] = D^{(k-1)}[i,j] \text{ or}$$

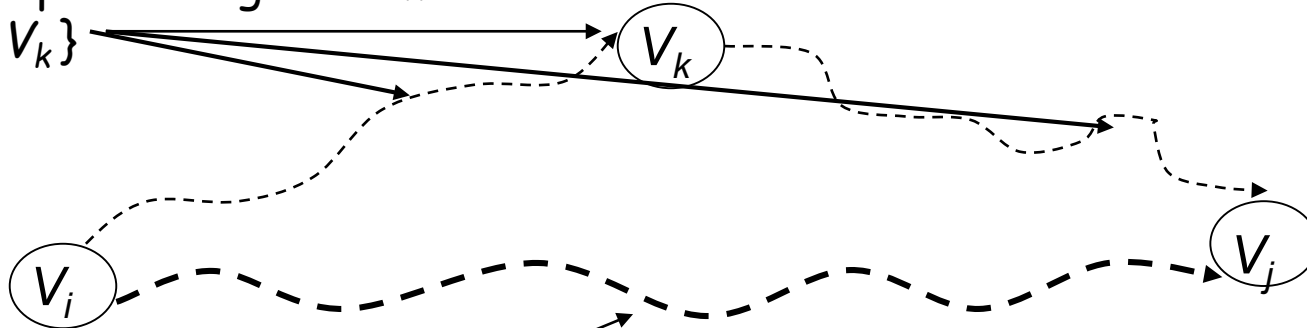
$$D^{(k)}[i,j] = D^{(k-1)}[i,k] + D^{(k-1)}[k,j]$$

We conclude:

$$D^{(k)}[i,j] = \min\{ D^{(k-1)}[i,j], D^{(k-1)}[i,k] + D^{(k-1)}[k,j] \}$$

Shortest path using intermediate vertices

$\{V_1, \dots, V_k\}$



Shortest Path using intermediate vertices  $\{V_1, \dots, V_{k-1}\}$





# The pointer array P

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- ▶ Used to enable finding a shortest path
- ▶ Initially the array contains 0
- ▶ Each time that a shorter path from  $i$  to  $j$  is found the  $k$  that provided the minimum is saved (highest index node on the path from  $i$  to  $j$ )
- ▶ To print the intermediate nodes on the shortest path a recursive procedure that print the shortest paths from  $i$  and  $k$ , and from  $k$  to  $j$  can be used



# Floyd's Algorithm Using $n+1$ $D$ matrices

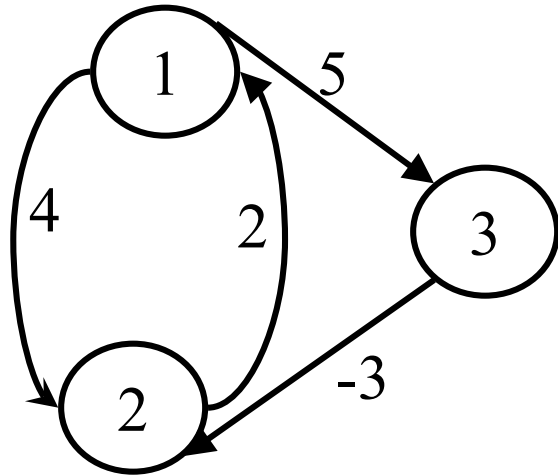
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Floyd//Computes shortest distance between all pairs of  
//nodes, and saves  $P$  to enable finding shortest paths

1.  $D^0 \leftarrow W$  // initialize  $D$  array to  $W[ ]$
2.  $P \leftarrow 0$  // initialize  $P$  array to  $[0]$
3. for  $k \leftarrow 1$  to  $n$
4.     do for  $i \leftarrow 1$  to  $n$
5.         do for  $j \leftarrow 1$  to  $n$
6.             if ( $D^{k-1}[i, j] > D^{k-1}[i, k] + D^{k-1}[k, j]$ )
7.                 then  $D^k[i, j] \leftarrow D^{k-1}[i, k] + D^{k-1}[k, j]$
8.                  $P[i, j] \leftarrow k$ ;
9.             else  $D^k[i, j] \leftarrow D^{k-1}[i, j]$



# Example



$$W = D^0 =$$

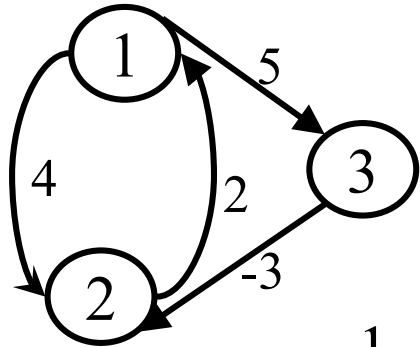
	1	2	3
1	0	4	5
2	2	0	$\infty$
3	$\infty$	-3	0

$$P =$$

	1	2	3
1	0	0	0
2	0	0	0
3	0	0	0



# Example



$D^0 =$

	1	2	3
1	0	4	5
2	2	0	$\infty$
3	$\infty$	-3	0

$k = 1$   
Vertex 1 can be  
intermediate node

$D^1 =$

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

$$\begin{aligned} D^1[2,3] &= \min( D^0[2,3], D^0[2,1]+D^0[1,3] ) \\ &= \min( \infty, 7 ) \\ &= 7 \end{aligned}$$

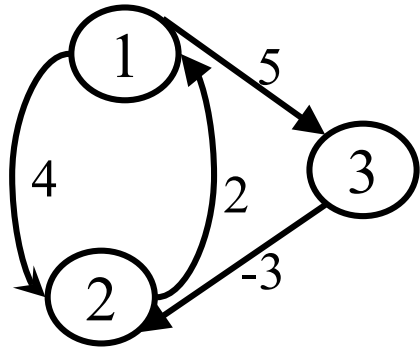
$P =$

	1	2	3
1	0	0	0
2	0	0	1
3	0	0	0

$$\begin{aligned} D^1[3,2] &= \min( D^0[3,2], D^0[3,1]+D^0[1,2] ) \\ &= \min( -3, \infty ) \\ &= -3 \end{aligned}$$



# Example



$D^1 =$

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

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

$k = 2$

Vertices 1, 2 can be intermediate

$D^2 =$

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

$$\begin{aligned}
 D^2[1,3] &= \min( D^1[1,3], D^1[1,2]+D^1[2,3] ) \\
 &= \min( 5, 4+7 ) \\
 &= 5
 \end{aligned}$$

$P =$

	1	2	3
1	0	0	0
2	0	0	1
3	2	0	0

$$\begin{aligned}
 D^2[3,1] &= \min( D^1[3,1], D^1[3,2]+D^1[2,1] ) \\
 &= \min( \infty, -3+2 ) \\
 &= -1
 \end{aligned}$$



# Floyd's Algorithm: Using 2 D matrices

---

Floyd

1.  $D \leftarrow W$  // initialize  $D$  array to  $W[ ]$
2.  $P \leftarrow 0$  // initialize  $P$  array to  $[0]$
3. for  $k \leftarrow 1$  to  $n$   
    *// Computing  $D'$  from  $D$*
4.     do for  $i \leftarrow 1$  to  $n$
5.         do for  $j \leftarrow 1$  to  $n$
6.             if  $(D[i, j] > D[i, k] + D[k, j])$
7.                 then  $D'[i, j] \leftarrow D[i, k] + D[k, j]$
8.                  $P[i, j] \leftarrow k;$
9.             else  $D'[i, j] \leftarrow D[i, j]$
10. Move  $D'$  to  $D$



# Can we use only one $D$ matrix?

---

- ▶  $D[i,j]$  depends only on elements in the  $k$ th column and row of the distance matrix
- ▶ We will show that the  $k$ th row and the  $k$ th column of the distance matrix are unchanged when  $D^k$  is computed
- ▶ This means  $D$  can be calculated *in-place*



# The main diagonal values

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- ▶ Before we show that  $k$ th row and column of  $D$  remain unchanged we show that the main diagonal remains 0
- ▶ 
$$\begin{aligned} D^{(k)}[j,j] &= \min\{ D^{(k-1)}[j,j], D^{(k-1)}[j,k] + D^{(k-1)}[k,j] \} \\ &= \min\{ 0, D^{(k-1)}[j,k] + D^{(k-1)}[k,j] \} \\ &= 0 \end{aligned}$$
- ▶ Based on which assumption?





# The $k$ th column

---

- ▶  $k$ th column of  $D^k$  is equal to the  $k$ th column of  $D^{k-1}$
- ▶ *Intuitively true* - a path from  $i$  to  $k$  will not become shorter by adding  $k$  to the allowed subset of intermediate vertices
- ▶ For all  $i$ ,  $D^{(k)}[i,k] =$ 
  - $= \min\{ D^{(k-1)}[i,k], D^{(k-1)}[i,k] + D^{(k-1)}[k,k] \}$
  - $= \min\{ D^{(k-1)}[i,k], D^{(k-1)}[i,k] + 0 \}$
  - $= D^{(k-1)}[i,k]$



# The $k$ th row

---

- ▶  $k$ th row of  $D^k$  is equal to the  $k$ th row of  $D^{k-1}$

$$\begin{aligned}\text{For all } j, D^{(k)}[k,j] &= \\ &= \min\{ D^{(k-1)}[k,j], D^{(k-1)}[k,k] + D^{(k-1)}[k,j] \} \\ &= \min\{ D^{(k-1)}[k,j], 0 + D^{(k-1)}[k,j] \} \\ &= D^{(k-1)}[k,j]\end{aligned}$$



# Question

---

- ▶ Can we claim that  $D^k$  equals to  $D^{k-1}$ ,  $D^{k-2}$  ?
  - No, we can only claim that
    - The 1-st row and 1-st column of  $D^1$  equal to the 1-st row and 1-st column of  $D^0$ , respectively
    - The 2-nd row and 2-nd column of  $D^2$  equal to the 2-nd row and 2-nd column of  $D^1$ , respectively
    - .....



# Floyd's Algorithm using a single $D$

---

Floyd

```
1.  $D \leftarrow W$  // initialize  $D$  array to  $W[ ]$ 
2.  $P \leftarrow 0$  // initialize  $P$  array to  $[0]$ 
3. for  $k \leftarrow 1$  to  $n$ 
4.     do for  $i \leftarrow 1$  to  $n$ 
5.         do for  $j \leftarrow 1$  to  $n$ 
6.             if ( $D[i, j] > D[i, k] + D[k, j]$ )
7.                 then  $D[i, j] \leftarrow D[i, k] + D[k, j]$ 
8.                  $P[i, j] \leftarrow k;$ 
```

}  $O(|V|^3)$

Total time cost:  $O(|V|^3)$



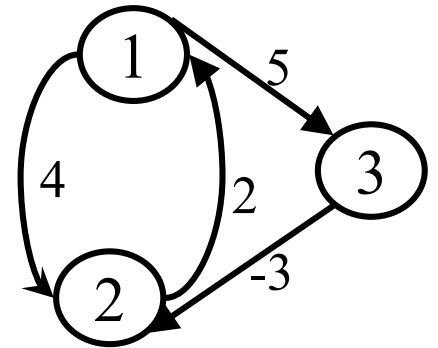
# Printing intermediate nodes on shortest path from $q$ to $r$

```
path(index  $q$ ,  $r$ )
  if ( $P[q, r] \neq 0$ )
    path( $q$ ,  $P[q, r]$ )
    println( " $v$ " +  $P[q, r]$ )
    path( $P[q, r]$ ,  $r$ )
  return;
//no intermediate nodes
else return
```

Before calling path check  $D[q, r] < \infty$ ,  
and print node  $q$ , after the call to  
path print node  $r$

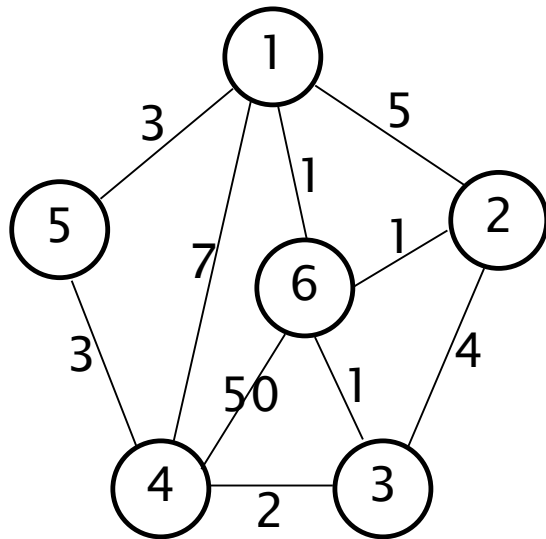
$P =$

	1	2	3
1	0	3	0
2	0	0	1
3	2	0	0





# Example

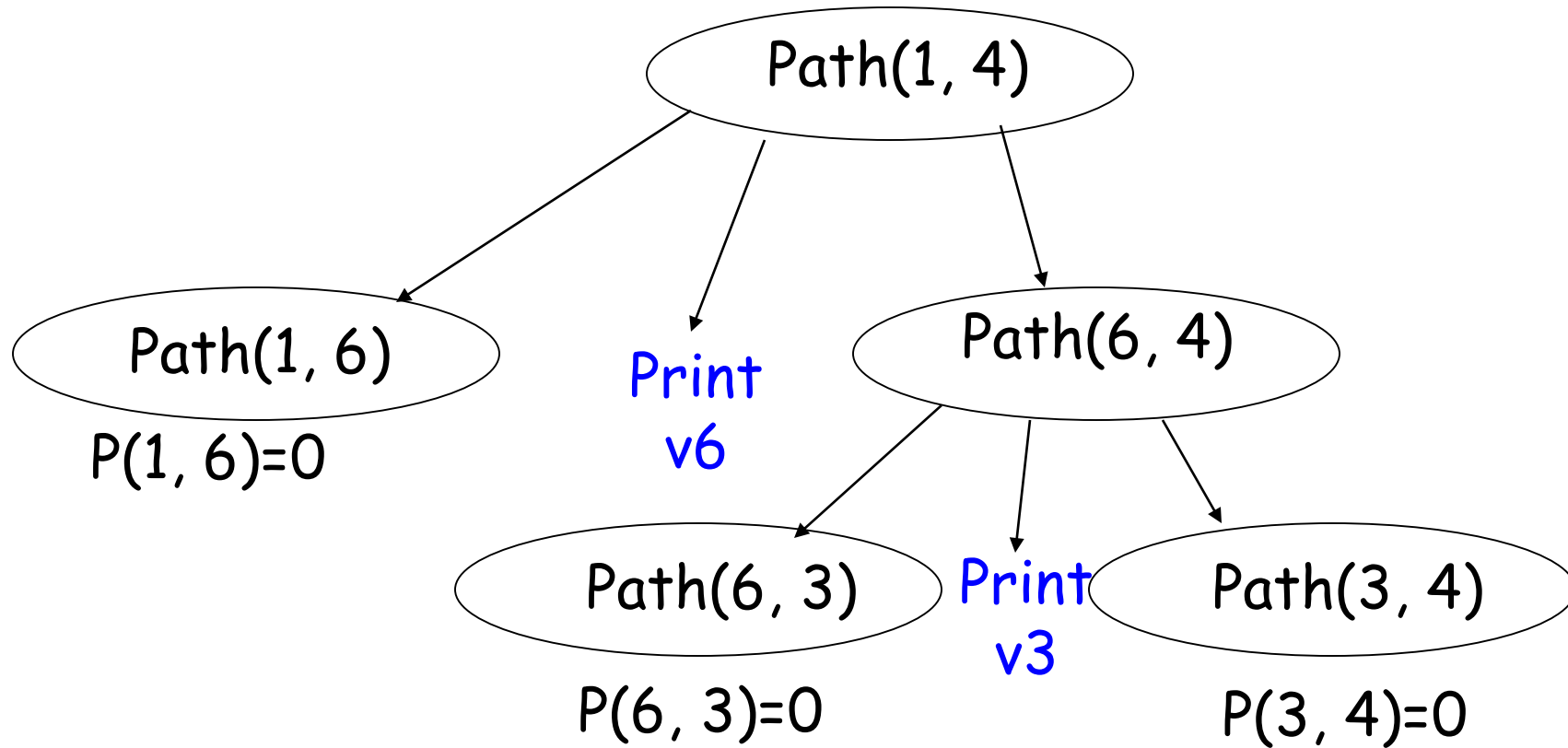


	1	2	3	4	5	6
1	0	2(6)	2(6)	4(6)	3	1
2	2(6)	0	2(6)	4(6)	5(6)	1
$D^6 = 3$	2(6)	2(6)	0	2	5(4)	1
4	4(6)	4(6)	2	0	3	3(3)
5	3	5(6)	5(4)	3	0	4(1)
6	1	1	1	3(3)	4(1)	0

The values in parenthesis are the non zero P values



# The call tree for Path(1, 4)



The intermediate nodes on the shortest path from 1 to 4 are v6, v3.  
The shortest path is v1, v6, v3, v4.



# Recommended Reading

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- ▶ Reading this week
  - Textbook Chapters 24-25
- ▶ Next Week
  - DAG checking and topological sort, Chapter 22