

## General DP Remarks

### Optimal Substructure

- Create optimal solution to problem using optimal solutions to subproblems.
- Can't use DP if optimal solution to a problem does not require subproblem solutions to be optimal.  
→ Often happens when subproblems are *not independent* of each other.

### Overlapping Subproblems

- For DP to be useful, recursive algorithm should require us to compute optimal solutions to the *same subproblems* over and over again.
- In total, there should be a small number of distinct subproblems (i.e. polynomial in input size).

## LCS

$$LCS[i, j] = \begin{cases} 1 + LCS[i - 1, j - 1] & \text{if } x_i = y_i \\ \max(LCS[i - 1, j], LCS[i, j - 1]) & \text{otherwise} \end{cases}$$

		j=0	j=1	j=2	j=3	j=4	j=5	j=6	j=7	j=8
			p	r	i	n	t	i	n	g
i=0		0	0	0	0	0	0	0	0	0
i=1	s	0	0	0	0	0	0	0	0	0
i=2	p	0	1↖	1←	1←	1←	1←	1←	1←	1←
i=3	r	0	1↑	2↖	2←	2←	2←	2←	2←	2←
i=4	i	0	1↑	2↑	3↖	3←	3←	3←	3←	3←
i=5	n	0	1↑	2↑	3↑	4↖	4←	4←	4←	4←
i=6	g	0	1↑	2↑	3↑	4↑	4↑	4↑	4↑	5↖
i=7	t	0	1↑	2↑	3↑	4↑	5↖	5←	5←	5←
i=8	i	0	1↑	2↑	3↖	4↑	5↑	6↖	6←	6←
i=9	m	0	1↑	2↑	3↑	4↑	5↑	6↑	6↑	6↑
i=10	e	0	1↑	2↑	3↑	4↑	5↑	6↑	6↑	6↑

## OBST

$$e[i, j] = \begin{cases} 0 & \text{if } i = j - 1 \\ \min_{i \leq r \leq j} \{e[i, r - 1] + e[r + 1, j] + w(i, j)\} & \text{if } i \leq j \end{cases}$$

$root[i, j]$  = root of subtree with keys  $k_i, \dots, k_j$  for  $1 \leq i \leq j \leq n$

$w[1, \dots, n + 1, 0, \dots, n]$  = sum of probabilities

$w[i, i - 1] = 0$  for  $1 \leq i \leq n$

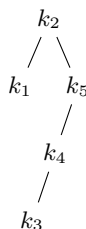
$w[i, j] = w[i, j - 1] + p_j$  for  $1 \leq i \leq j \leq n$

Consider 5 keys with search probabilities  $p_1 = 0.25, p_2 = 0.2, p_3 = 0.05, p_4 = 0.2, p_5 = 0.3$

w	0	1	2	3	4	5
1	0	0.25	0.45	0.5	0.7	1.0
2		0	0.2	0.25	0.45	0.55
3			0	0.05	0.25	0.55
4				0	0.2	0.5
5					0	0.3
6						0

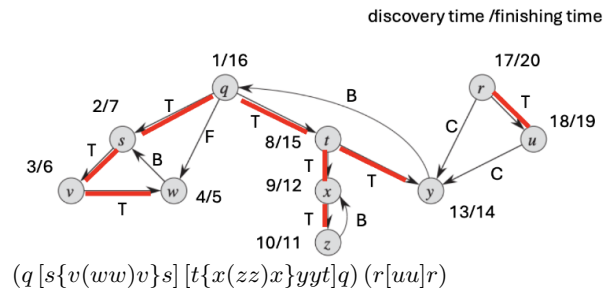
r	0	1	2	3	4	5
1		1	1	1	2	2
2			2	2	2	4
3				3	4	5
4					4	5
5						5
6						

e	0	1	2	3	4	5
1	0	0.25	0.65	0.8	1.25	2.1
2		0	0.2	0.3	0.75	1.35
3			0	0.05	0.3	0.85
4				0	0.2	0.7
5					0	0.3
6						0



## DFS

Tree edges: T  
Back edges: B  
Forward edges: F  
Cross edges: C



**Tree Edges** Are edges in depth-first forest  $G_\pi$ . Edge  $(u, v)$  is a tree edge if  $v$  was first discovered by exploring edge  $(u, v)$

**Back Edges** Are edges  $(u, v)$  connecting a vertex  $u$  to an ancestor in a depth-first tree. We consider self-loops, which may occur in directed graphs, to be back edges.

**Forward Edges** Are nontree edges  $(u, v)$  connecting a vertex  $u$  to a descendant  $v$  in a depth-first tree.

**Cross Edges** Are all other edges. They can go between vertices in the same depth-first tree, as long as one vertex is not an ancestor of the other, or they can go between vertices in different depth-first trees.

## Knapsack

$$c[i, w] = \begin{cases} 0 & \text{if } i = 0 \text{ or } w = 0 \\ c[i - 1, w] & \text{if } w_i > w \\ \max(c[i - 1, w], c[i - 1, w - w_i] + v_i) & \text{if } i > 0 \text{ and } w \geq w_i \end{cases}$$

## Graphs

### Handshaking Lemma

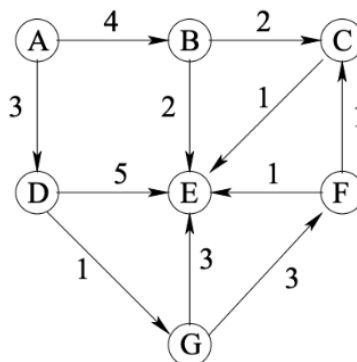
$$\sum_{v \in V} \deg(v) = 2|E|$$

## BFS vs DFS

- DFS is usually for finding relationship among vertices.
- BFS is usually for finding shortest path from a given source.

## Dijkstra

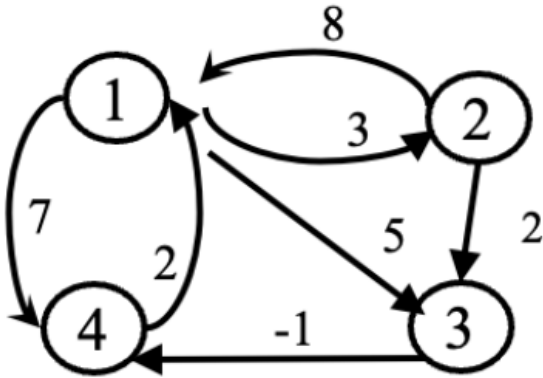
Execute Dijkstra's algorithm on the graph below starting at A. If there are ties, the vertex with the lowest letter comes first.



A: 0 D: 3 B: 4 G: 4 C: 6 E: 6  
B: ∞ B: 4 G: 4 C: 6 E: 6 F: 7  
C: ∞ C: ∞ E: 8 E: 6 F: 7  
D: ∞ E: ∞ C: ∞ F: ∞  
E: ∞ F: ∞ F: ∞  
F: ∞ G: ∞  
G: ∞

Floyd-Warshall

Execute the Floyd-Warshall algorithm on the graph below, provide the  $D^k$  and  $P$  matrixes on each step.



$D^0 =$

	1	2	3	4
1	0	3	5	7
2	8	0	2	$\infty$
3	$\infty$	$\infty$	0	-1
4	2	$\infty$	$\infty$	0

$D^1 =$

	1	2	3	4
1	0	3	5	7
2	8	0	2	15
3	$\infty$	$\infty$	0	-1
4	2	5	7	0

$D^2 =$

	1	2	3	4
1	0	3	5	7
2	8	0	2	15
3	$\infty$	$\infty$	0	-1
4	2	5	7	0

$D^3 =$

	1	2	3	4
1	0	3	5	4
2	8	0	2	1
3	$\infty$	$\infty$	0	-1
4	2	5	7	0

$D^4 =$

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

$P =$

	1	2	3	4
1	0	0	0	0
2	0	0	0	0
3	0	0	0	0
4	0	0	0	0

$P =$

	1	2	3	4
1	0	0	0	0
2	0	0	0	1
3	0	0	0	0
4	0	1	1	0

$P =$

	1	2	3	4
1	0	0	0	0
2	0	0	0	1
3	0	0	0	0
4	0	1	1	0

$P =$

	1	2	3	4
1	0	0	0	3
2	0	0	0	3
3	0	0	0	0
4	0	1	1	0

$P =$

	1	2	3	4
1	0	0	0	3
2	4	0	0	3
3	4	4	0	0
4	0	1	1	0

Knapsack

$$c[i, w] = \begin{cases} 0 & \text{if } i = 0 \text{ or } w = 0 \\ c[i - 1, w] & \text{if } w_i > w \\ \max(c[i - 1, w], c[i - 1, w - w_i] + v_i) & \text{otherwise} \end{cases}$$

**SINGLE-SOURCE SHORTEST PATH ALGORITHMS**

The **shortest path problem** is to identify a minimum cost path in a graph from a source node to a destination.

**Dijkstra's algorithm** uses a priority queue to greedily select the closest vertex that has not yet been processed, and **relaxes** all outgoing edges. Performance is  $O(|E| + |V| \log(|V|))$ . Does not support negative-weight cycles.

**A\* algorithm** uses edge weights plus a heuristic (often Euclidean distance) to relax edges in **best-first** order. Dijkstra's is a special case. Performance is  $O(|E|)$ . Does not support negative-weight cycles.

**Bellman-Ford algorithm** relaxes all  $|E|$  edges at most  $|V| - 1$  times. Edge selection order matters. Performance is  $O(|V||E|)$ . Detects negative-weight cycles.

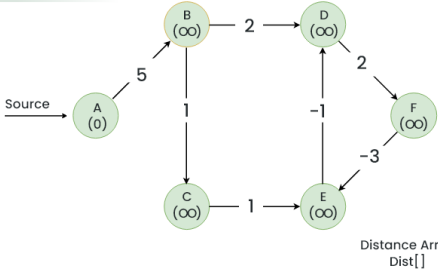
**Legend:**

- $c$  = node (vertex) with source-to-here cost  $c$
- $\bullet$  = source node
- $\bullet$  = destination node
- $\bullet$  = node with all outgoing edges searched
- $w$  = edge with weight  $w$
- $\text{---}$  = edge in minimum cost (shortest) path
- $\text{---}$  = edge considered for next search

Bellman-Ford

Bellman-Ford is a single source shortest path algorithm that determines the shortest path between a given source vertex and every other vertex in a graph. This algorithm can be used on both weighted and unweighted graphs.

Initialize The Distance Array



Bellman-Ford To Detect A Negative Cycle In A Graph

1. Initialize distance array to store shortest dist for each vertex. Initialize source as 0, and all others to be  $\infty$ .
2. Relax all edges  $|V| - 1$  times.
3. Relax all edges one more time to detect negative cycles.

**Case 1** For any edge(u, v, weight), if  $\text{dist}[u] + \text{weight} < \text{dist}[v]$ , then there is a negative cycle.

**Case 2** Case 1 fails for all edges

A	B	C	D	E	F
0	$\infty$	$\infty$	$\infty$	$\infty$	$\infty$
0	5	$\infty$	$\infty$	$\infty$	$\infty$
0	5	6	7	$\infty$	$\infty$
0	5	6	7	7	9
0	5	6	6	6	9
0	5	6	5	6	8

1. Start relaxing edges, 1st relaxation:

$$\begin{aligned} \text{Dist}(B) &> \text{Dist}(A) + w(A, B) \\ \infty &> 0 + 5 \Rightarrow \text{Dist}(B) = 5 \end{aligned}$$

2. 2nd relaxation:

$$\begin{aligned} \text{Dist}(D) &> \text{Dist}(B) + w(B, D) \\ \infty &> 5 + 2 \Rightarrow \text{Dist}(D) = 7 \end{aligned}$$

$$\begin{aligned} \text{Dist}(C) &> \text{Dist}(B) + w(B, C) \\ \infty &> 5 + 1 \Rightarrow \text{Dist}(C) = 6 \end{aligned}$$

3. 3rd relaxation:

$$\begin{aligned} \text{Dist}(F) &> \text{Dist}(D) + w(D, F) \\ \infty &> 7 + 2 \Rightarrow \text{Dist}(F) = 9 \end{aligned}$$

$$\begin{aligned} \text{Dist}(E) &> \text{Dist}(C) + w(C, E) \\ \infty &> 6 + 1 \Rightarrow \text{Dist}(E) = 7 \end{aligned}$$

4. 4th relaxation:

$$\begin{aligned} \text{Dist}(D) &> \text{Dist}(E) + w(E, D) \\ 7 &> 7 + (-1) \Rightarrow \text{Dist}(D) = 6 \end{aligned}$$

$$\begin{aligned} \text{Dist}(E) &> \text{Dist}(F) + w(F, E) \\ 7 &> 9 + (-3) \Rightarrow \text{Dist}(E) = 6 \end{aligned}$$

5. 5th relaxation:

$$\begin{aligned} \text{Dist}(F) &> \text{Dist}(D) + w(D, F) \\ 9 &> 6 + 2 \Rightarrow \text{Dist}(F) = 8 \end{aligned}$$

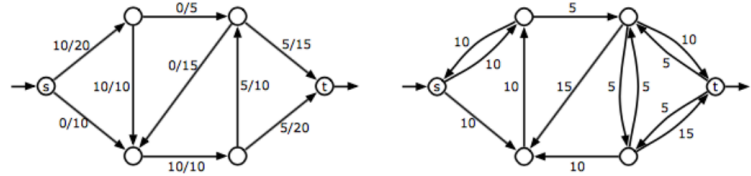
$$\begin{aligned} \text{Dist}(D) &> \text{Dist}(E) + w(E, D) \\ 6 &> 6 + (-1) \Rightarrow \text{Dist}(D) = 5 \end{aligned}$$

6. 6th relaxation (final):

$$\begin{aligned} \text{Dist}(E) &> \text{Dist}(F) + w(F, E) \\ 6 &> 8 + (-3) \Rightarrow \text{Dist}(E) = 5 \end{aligned}$$

$$\begin{aligned} \text{Dist}(F) &> \text{Dist}(D) + w(D, F) \\ 8 &> 5 + 2 \Rightarrow \text{Dist}(F) = 7 \end{aligned}$$

# Network Flow



- Flow** Amount of material that can be transported between two nodes.
- Capacity** Maximum amount of material that can be transported between two nodes.
- Residual Capacity** Amount of material that can still be transported between two nodes.
- Augmenting Path** Path from source to sink where all edges have residual capacity.
- Residual Graph** Graph where edges have residual capacity.