

Example: Find Common Subsequence

	1	2	3	4	5	6
A	b	a	c	d	c	a
B	a	d	b	c	d	a

	0	1	2	3	4	5	6
0	0 ⊥	0 ⊥	0 ⊥	0 ⊥	0 ⊥	0 ⊥	0 ⊥
1	0 ⊥	0 ←	0 ←	1 ↖	1 ←	1 ←	1 ←
2	0 ⊥	1 ↘	1 ←	1 ←	1 ←	1 ←	2 ↘
3	0 ⊥	1 ↑	1 ←	1 ←	2 ↘	2 ←	2 ←
4	0 ⊥	1 ↑	2 ↘	2 ←	2 ←	3 ↘	3 ←
5	0 ⊥	1 ↑	2 ↑	2 ←	3 ↘	3 ←	3 ←
6	0 ⊥	1 ↘	2 ↑	2 ←	3 ↑	3 ←	4 ↘

Find Common Subsequence

```
1:  $i \leftarrow n, j \leftarrow m, S \leftarrow ()$ 
2: while  $i > 0$  and  $j > 0$  do
3:   if  $\pi[i, j] = "\searrow"$  then
4:     add  $A[i]$  to beginning of  $S, i \leftarrow i - 1, j \leftarrow j - 1$ 
5:   else if  $\pi[i, j] = "\uparrow"$  then
6:      $i \leftarrow i - 1$ 
7:   else
8:      $j \leftarrow j - 1$ 
9: return  $S$ 
```

Variants of Problem

Edit Distance with Insertions and Deletions

Input: a string A and a string B

each time we can delete a letter from A or insert a letter to A

Output: minimum number of operations (insertions or deletions) we need to change A to B ?

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Obs. $\#OPs = \text{length}(A) + \text{length}(B) - 2 \cdot \text{length}(LCS(A, B))$

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Example:

- $A = \text{ocurrance}$, $B = \text{occurrence}$.
- 2 operations: insert 'c', change 'a' to 'e'
- Not related to LCS any more

Edit Distance with Replacing: Reduction to a Variant of LCS

- Need to match letters in A and B , every letter is matched at most once and there should be no crosses.
- However, we can **match two different letters**: Matching a same letter gives score 2, matching two different letters gives score 1.
- Need to maximize the score.
- DP recursion for the case $i > 0$ and $j > 0$:

$$opt[i, j] = \begin{cases} opt[i - 1, j - 1] + 2 & \text{if } A[i] = B[j] \\ \max \begin{cases} opt[i - 1, j] \\ opt[i, j - 1] \\ opt[i - 1, j - 1] + 1 \end{cases} & \text{if } A[i] \neq B[j] \end{cases}$$

- Relation : $\#OPs = \text{length}(A) + \text{length}(B) - \text{max_score}$

Edit Distance (with Replacing): using DP directly

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Outline

- 1 Weighted Interval Scheduling
- 2 Subset Sum Problem
- 3 Knapsack Problem
- 4 Longest Common Subsequence**
 - Longest Common Subsequence in Linear Space
- 5 Shortest Paths in Directed Acyclic Graphs
- 6 Matrix Chain Multiplication
- 7 Optimum Binary Search Tree
- 8 Summary

Computing the Length of LCS

```
1: for  $j \leftarrow 0$  to  $m$  do
2:    $opt[0, j] \leftarrow 0$ 
3: for  $i \leftarrow 1$  to  $n$  do
4:    $opt[i, 0] \leftarrow 0$ 
5:   for  $j \leftarrow 1$  to  $m$  do
6:     if  $A[i] = B[j]$  then
7:        $opt[i, j] \leftarrow opt[i - 1, j - 1] + 1$ 
8:     else if  $opt[i, j - 1] \geq opt[i - 1, j]$  then
9:        $opt[i, j] \leftarrow opt[i, j - 1]$ 
10:    else
11:       $opt[i, j] \leftarrow opt[i - 1, j]$ 
```

Obs. The i -th row of table only depends on $(i - 1)$ -th row.

Reducing Space to $O(n + m)$

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Q: How to use this observation to reduce space?

A: We only keep two rows: the $(i - 1)$ -th row and the i -th row.

Linear Space Algorithm to Compute Length of LCS

```
1: for  $j \leftarrow 0$  to  $m$  do
2:    $opt[0, j] \leftarrow 0$ 
3: for  $i \leftarrow 1$  to  $n$  do
4:    $opt[i \bmod 2, 0] \leftarrow 0$ 
5:   for  $j \leftarrow 1$  to  $m$  do
6:     if  $A[i] = B[j]$  then
7:        $opt[i \bmod 2, j] \leftarrow opt[i - 1 \bmod 2, j - 1] + 1$ 
8:     else if  $opt[i \bmod 2, j - 1] \geq opt[i - 1 \bmod 2, j]$  then
9:        $opt[i \bmod 2, j] \leftarrow opt[i \bmod 2, j - 1]$ 
10:    else
11:       $opt[i \bmod 2, j] \leftarrow opt[i - 1 \bmod 2, j]$ 
12: return  $opt[n \bmod 2, m]$ 
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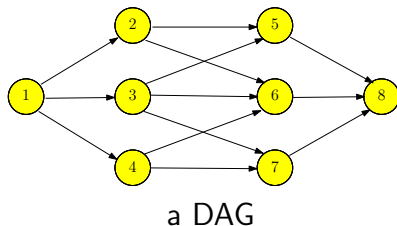
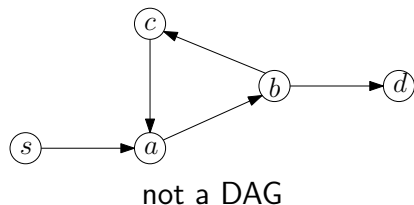
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- Using **Divide and Conquer** + Dynamic Programming:
 - Space: $O(m + n)$
 - Time: $O(nm)$

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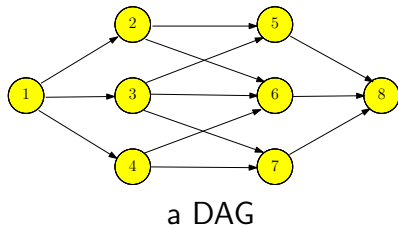
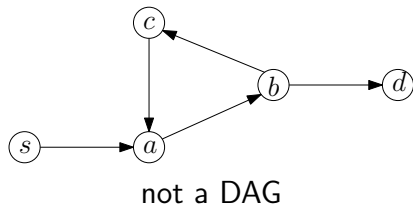
Directed Acyclic Graphs

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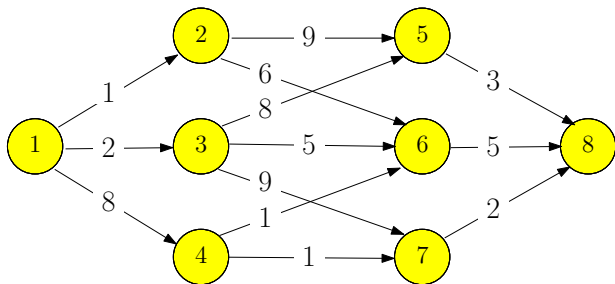
Lemma A directed graph is a DAG if and only if its vertices can be topologically sorted.

Shortest Paths in DAG

Input: directed acyclic graph $G = (V, E)$ and $w : E \rightarrow \mathbb{R}$.

Assume $V = \{1, 2, 3, \dots, n\}$ is topologically sorted: if $(i, j) \in E$, then $i < j$

Output: the shortest path from 1 to i , for every $i \in V$

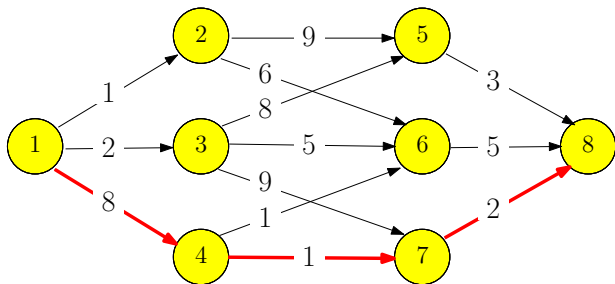


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- $f[i]$: length of the shortest path from 1 to i

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- $f[i]$: length of the shortest path from 1 to i

$$f[i] = \begin{cases} 0 & i = 1 \\ \min_{j:(j,i) \in E} \{f(j) + w(j,i)\} & i = 2, 3, \dots, n \end{cases}$$