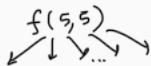


Pre-lecture brain teaser

What is the running time of the following algorithm:

Consider computing $f(x, y)$ by recursive function + memoization.



$$f(x, y) = \sum_{i=1}^{\min(x, y)} x * f(x + y - i, i - 1),$$

$f(0, y) = y$ $f(x, 0) = x.$

Work per subproblem = $O(n) \dots (c)$

$\max_i x + y - i = x + y - 1$
 $\downarrow \quad \downarrow$
 $O(n) \quad O(n)$
 $\Rightarrow O(n) \dots (a)$

$\max_i i - 1 = \min(x, y) - 1$
 \downarrow
 $O(n)$
 $\Rightarrow O(n) \dots (b)$

The resulting algorithm when computing $f(n, n)$ would take: (a) and (b)

(a) $O(n^2)$

(b) $O(n^3)$

(c) $O(2^n)$

(d) $O(n^n)$

(e) The function is ill defined - it can not be computed.

(a), (b), and (c) $\Rightarrow T(n) = O(n^2) \cdot O(n)$
 $= O(n^3)$

$\Rightarrow O(n^2)$
 subproblems

ECE-374-B: Lecture 13 - Dynamic Programming II

Instructor: Abhishek Kumar Umrawal

March 5, 2024

University of Illinois at Urbana-Champaign

Pre-lecture brain teaser

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Consider computing $f(x, y)$ by recursive function + memoization.

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The resulting algorithm when computing $f(n, n)$ would take:

- (a) $O(n^2)$
- (b) $O(n^3)$
- (c) $O(2^n)$
- (d) $O(n^n)$
- (e) The function is ill defined - it can not be computed.

Recipe for Dynamic Programming

1. Develop a recursive backtracking style algorithm \mathcal{A} for given problem.
2. Identify structure of subproblems generated by \mathcal{A} on an instance I of size n
 - 2.1 Estimate number of different subproblems generated as a function of n . Is it polynomial or exponential in n ?
 - 2.2 If the number of problems is “small” (polynomial) then they typically have some “clean” structure.
3. Rewrite subproblems in a compact fashion.
4. Rewrite recursive algorithm in terms of notation for subproblems.
5. Convert to iterative algorithm by bottom up evaluation in an appropriate order.
6. Optimize further with data structures and/or additional ideas.

Edit Distance and Sequence Alignment

Spell Checking Problem

Given a string “exponen” that is not in the dictionary, how should a spell checker suggest a nearby string?

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What does nearness mean?

Question: Given two strings $x_1x_2 \dots x_n$ and $y_1y_2 \dots y_m$ what is a distance between them?

Spell Checking Problem

Given a string “**exponen**” that is not in the dictionary, how should a spell checker suggest a nearby string?

What does nearness mean?

Question: Given two strings $x_1x_2 \dots x_n$ and $y_1y_2 \dots y_m$ what is a distance between them?

Edit Distance: **minimum number of “edits”** to transform x into y .

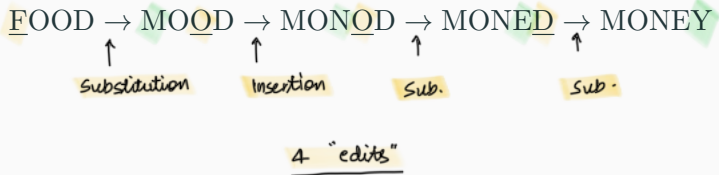
Edit Distance

Definition

Edit distance between two words X and Y is the number of letter insertions, letter deletions and letter substitutions required to obtain Y from X .

Example

The edit distance between FOOD and MONEY is at least 4:



Edit Distance: Alternate View

Alignment

Place words one on top of the other, with gaps in the first word indicating insertions, and gaps in the second word indicating deletions.



Mismatched letters mean substitution (same as deletion followed by an insertion).

Edit Distance: Alternate View

Alignment

Place words one on top of the other, with gaps in the first word indicating insertions, and gaps in the second word indicating deletions.

	F	O	O		D		
	M	O	N	E	Y	(A1)	<u>"4"</u>

Formally, an **alignment** is a set M of pairs (i, j) such that each index appears at most once, and there is no "crossing": $i < i'$ and i is matched to j implies i' is matched to $j' > j$. In the above example, this is $M = \{(1, 1), (2, 2), (3, 3), (4, 5)\}$.



(B1Y)

F	O	O	D				
		M	O	N	E	Y	

(A2)
"7"

Edit Distance: Alternate View

Alignment

Place words one on top of the other, with gaps in the first word indicating insertions, and gaps in the second word indicating deletions.

F	O	O		D
M	O	N	E	Y

Formally, an **alignment** is a set M of pairs (i, j) such that each index appears at most once, and there is no “crossing”: $i < i'$ and i is matched to j implies i' is matched to $j' > j$. In the above example, this is $M = \{(1, 1), (2, 2), (3, 3), (4, 5)\}$. Cost of an alignment is the number of mismatched columns plus number of unmatched indices in both strings.

Edit Distance Problem

Problem

Given two words, find the edit distance between them, i.e., an alignment of smallest cost.

Applications

- Spell-checkers and Dictionaries
- Unix diff
- DNA sequence alignment ... but, we need a new metric

Sequence alignment problem - Similarity Metric

Definition

For two strings X and Y , the cost of alignment M is

- [Gap penalty] For each gap in the alignment, we incur a cost δ . $\leftarrow 1$
- [Mismatch cost] For each pair p and q that have been matched in M , we incur cost α_{pq} ; typically $\alpha_{pp} = 0$.
 \uparrow
 $|$

Sequence alignment problem - Similarity Metric

Definition

For two strings X and Y , the cost of alignment M is

- [Gap penalty] For each gap in the alignment, we incur a cost δ .
- [Mismatch cost] For each pair p and q that have been matched in M , we incur cost α_{pq} ; typically $\alpha_{pp} = 0$.

Edit distance is special case when $\delta = \alpha_{pq} = 1$.

Edit distance as alignment

An Example

Example

(1)

→	o	◆	c	u	r	r	a	n	c	e
→	o	c	c	u	r	r	e	n	c	e

Cost = $\underline{\delta} + \underline{\alpha_{ae}}$

Alternative:

(2)

	o	◆	c	u	r	r	◆	a	n	c	e
	o	c	c	u	r	r	e	◆	n	c	e

Cost = $\underline{3\delta}$

Or a really stupid solution (delete string, insert other string):

(3)

	o	c	u	r	r	a	n	c	e	◆	◆	◆	◆	◆	◆	◆	◆	◆	◆
◆	◆	◆	◆	◆	◆	◆	◆	◆	◆	o	c	c	u	r	r	e	n	c	e

Cost = $\underline{19\delta}$.

What is the edit distance between...

What is the **minimum edit distance** for the following two strings, if insertion/deletion/change of a single character cost 1 unit?

374

473

- (a) 1
- (b) 2**
- (c) 3
- (d) 4
- (e) 5

What is the edit distance between...

What is the minimum edit distance for the following two strings, if insertion/deletion/change of a single character cost 1 unit?

373

473

- (a) 1
- (b) 2
- (c) 3
- (d) 4
- (e) 5

What is the edit distance between...

What is the **minimum edit distance** for the following two strings, if insertion/deletion/change of a single character cost 1 unit?

37

473

- (a) 1
- (b) 2
- (c) 3
- (d) 4
- (e) 5

Sequence Alignment

Input Given two words X and Y , and gap penalty δ and mismatch costs α_{pq}

Goal Find alignment of minimum cost

Edit distance: The algorithm

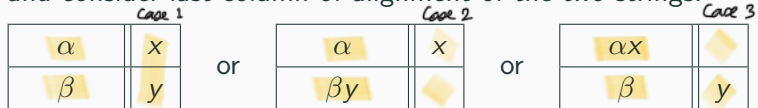
Edit distance - Basic observation

$x_1 x_2 \dots x_M$ $y_1 y_2 \dots y_N$
Let $X = \alpha x$ and $Y = \beta y$

α, β : strings.

x and y single characters.

Think about optimal edit distance between X and Y as alignment, and consider last column of alignment of the two strings:



Prefixes must have optimal alignment!

Problem Structure

Let $X = x_1x_2 \cdots x_m$ and $Y = y_1y_2 \cdots y_n$. If (m, n) are not matched then either the m^{th} position of X remains unmatched or the n^{th} position of Y remains unmatched.

- **Case** x_m and y_n are matched.
 - Pay mismatch cost $\alpha_{x_my_n}$ plus cost of aligning strings $x_1 \cdots x_{m-1}$ and $y_1 \cdots y_{n-1}$
- **Case** x_m is unmatched.
 - Pay gap penalty plus cost of aligning $x_1 \cdots x_{m-1}$ and $y_1 \cdots y_n$
- **Case** y_n is unmatched.
 - Pay gap penalty plus cost of aligning $x_1 \cdots x_m$ and $y_1 \cdots y_{n-1}$

Subproblems and Recurrence

$x_1 \dots x_{i-1}$	x_i	or	$x_1 \dots x_{i-1}$	x	or	$x_1 \dots x_{i-1}x_i$	
$y_1 \dots y_{j-1}$	y_j		$y_1 \dots y_{j-1}y_j$			$y_1 \dots y_{j-1}$	y_j

Optimal Costs

Let $\text{Opt}(i, j)$ be optimal cost of aligning $x_1 \dots x_i$ and $y_1 \dots y_j$.

Then

$$\text{Opt}(i, j) = \min \begin{cases} \alpha_{x_i y_j} + \text{Opt}(i-1, j-1), \\ \delta + \text{Opt}(i-1, j), \\ \delta + \text{Opt}(i, j-1) \end{cases}$$

Subproblems and Recurrence

$x_1 \dots x_{i-1}$	x_i	or	$x_1 \dots x_{i-1}$	x	or	$x_1 \dots x_{i-1}x_i$	
$y_1 \dots y_{j-1}$	y_j		$y_1 \dots y_{j-1}y_j$			$y_1 \dots y_{j-1}$	y_j

Optimal Costs

Let $\text{Opt}(i, j)$ be optimal cost of aligning $x_1 \dots x_i$ and $y_1 \dots y_j$.

Then

$$\text{Opt}(i, j) = \min \begin{cases} \alpha_{x_i y_j} + \text{Opt}(i - 1, j - 1), \\ \delta + \text{Opt}(i - 1, j), \\ \delta + \text{Opt}(i, j - 1) \end{cases}$$

Base Cases: $\text{Opt}(i, 0) = \delta \cdot i$ and $\text{Opt}(0, j) = \delta \cdot j$

Recursive Algorithm

Assume X is stored in array $A[1..m]$ and Y is stored in $B[1..n]$
Array $COST$ stores cost of matching two chars. Thus $COST[a, b]$
give the cost of matching character a to character b .

```
EDIST(A[1..m], B[1..n])
  If (m = 0) return nδ
  If (n = 0) return mδ
  m1 = δ + EDIST(A[1..(m - 1)], B[1..n])
  m2 = δ + EDIST(A[1..m], B[1..(n - 1)])
  m3 = COST[A[m], B[n]] + EDIST(A[1..(m - 1)], B[1..(n - 1)])
  return min(m1, m2, m3)
```

(R1Y)

Example: DEED and DREAD

	<u>ε</u>	<u>D</u>	<u>R</u>	<u>E</u>	<u>A</u>	<u>D</u>
<u>ε</u>	0*	1*	2	3	4	5
<u>D</u>	$\frac{1}{+1}$	0	1			
<u>E</u>	2					
<u>E</u>	3					
<u>D</u>	4					

The table above shows the dynamic programming table for the sequence alignment problem. The top row and left column are labeled with the characters ϵ , D, E, E, D. The cells contain the minimum edit distance values. Handwritten annotations include:

- Red arrows pointing from (row ϵ , col D) to (row D, col ϵ), (row D, col D) to (row D, col R), and (row E, col D) to (row E, col R).
- Black arrows pointing from (row D, col D) to (row D, col ϵ), (row D, col D) to (row E, col D), and (row E, col D) to (row E, col ϵ).
- A circle around the cell (row D, col D) containing the value 0.
- Blue shaded cells: (row D, col D), (row D, col R), (row E, col D), (row E, col E), (row D, col D), (row D, col D).
- Yellow shaded cells: (row ϵ , col D), (row D, col ϵ), (row E, col ϵ), (row E, col D), (row D, col ϵ).

$\text{Opt}(i, j) =$

$$\min \begin{cases} \alpha_{x_i y_j} + \text{Opt}(i-1, j-1), \\ \delta + \text{Opt}(i-1, j), \\ \delta + \text{Opt}(i, j-1) \end{cases}$$

Base Cases:

- $\text{Opt}(i, 0) = \delta \cdot i$
- $\text{Opt}(0, j) = \delta \cdot j$

Example: DEED and DREAD

	ϵ	D	R	E	A	D
ϵ	0	1	2	3	4	5
D	1					
E	2					
E	3					
D	4					

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E	3					
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	ϵ	D	R	E	A	D
ϵ	0	1	2	3	4	5
D	1	0	1	2	3	4
E	2	1	1	1	2	3
E	3					
D	4					

$\text{Opt}(i, j) =$

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Example: DEED and DREAD

	ϵ	D	R	E	A	D
ϵ	0	1	2	3	4	5
D	1	0	1	2	3	4
E	2	1	1	1	2	3
E	3	2	2	1	2	3
D	4					

$\text{Opt}(i, j) =$

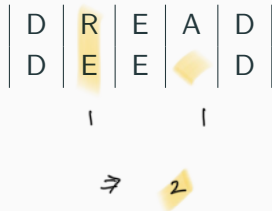
$$\min \begin{cases} \alpha_{x_i y_j} + \text{Opt}(i-1, j-1), \\ \delta + \text{Opt}(i-1, j), \\ \delta + \text{Opt}(i, j-1) \end{cases}$$

Base Cases:

- $\text{Opt}(i, 0) = \delta \cdot i$
- $\text{Opt}(0, j) = \delta \cdot j$

Example: DEED and DREAD

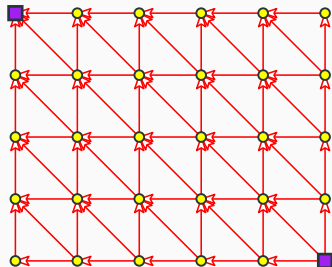
	ϵ	<i>D</i>	<i>R</i>	<i>E</i>	<i>A</i>	<i>D</i>
ϵ	0	1	2	3	4	5
<i>D</i>	1	0	1	2	3	4
<i>E</i>	2	1	1	1	2	3
<i>E</i>	3	2	2	1	2	3
<i>D</i>	4	3	3	2	2	2



Example: DEED and DREAD

	ϵ	<i>D</i>	<i>R</i>	<i>E</i>	<i>A</i>	<i>D</i>
ϵ	0	1	2	3	4	5
<i>D</i>	1	0	1	2	3	4
<i>E</i>	2	1	1	1	2	3
<i>E</i>	3	2	2	1	2	3
<i>D</i>	4	3	3	2	2	2

<i>D</i>	<i>R</i>	<i>E</i>	<i>A</i>	<i>D</i>
<i>D</i>	<i>E</i>	<i>E</i>		<i>D</i>



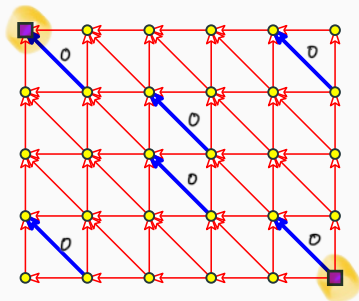
Example: DEED and DREAD

	ϵ	D	R	E	A	D
ϵ	0	1	2	3	4	5
D	1	0	1	2	3	4
E	2	1	1	1	2	3
E	3	2	2	1	2	3
D	4	3	3	2	2	-2

chosen
cost of the \uparrow path = 2

D	R	E	A	D
D	E	E		D

Cost = 2

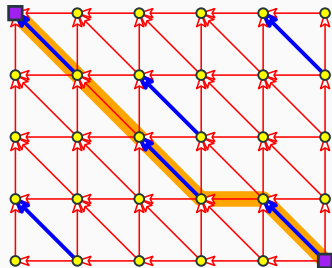


Shortest path problem.

Example: DEED and DREAD

	ϵ	<i>D</i>	<i>R</i>	<i>E</i>	<i>A</i>	<i>D</i>
ϵ	0	1	2	3	4	5
<i>D</i>	1	0	1	2	3	4
<i>E</i>	2	1	1	1	2	3
<i>E</i>	3	2	2	1	2	3
<i>D</i>	4	3	3	2	2	2

<i>D</i>	<i>R</i>	<i>E</i>	<i>A</i>	<i>D</i>
<i>D</i>	<i>E</i>	<i>E</i>		<i>D</i>



Dynamic programming algorithm for edit-distance

As part of the input...

The cost of aligning a character against another character

Σ : Alphabet

We are given a cost function (in a table):

$$\forall b, c \in \Sigma \quad \text{COST}[b][c] = \text{cost of aligning } b \text{ with } c.$$

$$\forall b \in \Sigma \quad \text{COST}[b][b] = 0$$

δ : price of deletion ~~of~~ insertion of a single character
or

Dynamic program for edit distance

EDIST(*A*[1..*m*], *B*[1..*n*])

int *M*[0..*m*][0..*n*]

for *i* = 1 to *m* **do** *M*[*i*, 0] = *i*δ

for *j* = 1 to *n* **do** *M*[0, *j*] = *j*δ

for *i* = 1 to *m* **do**

for *j* = 1 to *n* **do**

$$M[i][j] = \min \begin{cases} \text{COST}[A[i]][B[j]] + M[i-1][j-1], \\ \delta + M[i-1][j], \\ \delta + M[i][j-1] \end{cases}$$

(R1Y)

Dynamic program for edit distance

$EDIST(A[1..m], B[1..n])$

int $M[0..m][0..n]$

for $i = 1$ to m do $M[i, 0] = i\delta$

for $j = 1$ to n do $M[0, j] = j\delta$

for $i = 1$ to m do

for $j = 1$ to n do

$$M[i][j] = \min \begin{cases} COST[A[i]][B[j]] + M[i-1][j-1], \\ \delta + M[i-1][j], \\ \delta + M[i][j-1] \end{cases}$$

Analysis

- Running time is $O(mn)$.
- Space used is $O(mn)$.

of subproblems : $O(mn)$
cost for each subproblem : $O(1)$
 $\rightarrow O(mn)$

Reducing space for edit distance

Matrix and DAG of computation of edit distance

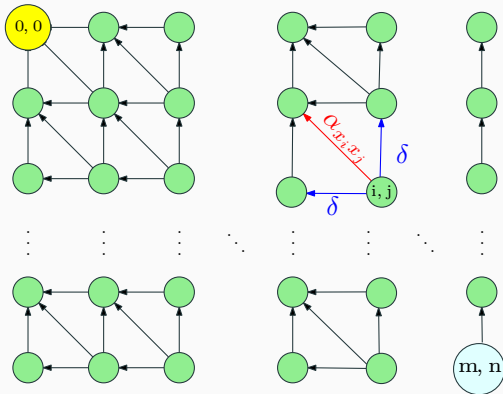


Figure 1: Iterative algorithm in previous slide computes values in row order.

Optimizing Space

- Recall

$$M(i, j) = \min \begin{cases} \alpha_{x_i y_j} + M(i - 1, j - 1), \\ \delta + M(i - 1, j), \\ \delta + M(i, j - 1) \end{cases}$$

- Entries in j^{th} column only depend on $(j - 1)^{\text{st}}$ column and earlier entries in j^{th} column
- Only store the current column and the previous column reusing space; $N(i, 0)$ stores $M(i, j - 1)$ and $N(i, 1)$ stores $M(i, j)$

Example: DEED vs. DREAD filled by column

	ϵ	D	R	E	A	D
ϵ						
D						
E						
E						
D						

Example: DEED vs. DREAD filled by column

	ϵ	<i>D</i>	<i>R</i>	<i>E</i>	<i>A</i>	<i>D</i>
ϵ	0	1	2	3	4	5
<i>D</i>	1					
<i>E</i>	2					
<i>E</i>	3					
<i>D</i>	3					

Example: DEED vs. DREAD filled by column

	ϵ	D	R	E	A	D	n
ϵ	0	1	2	3	4	5	
D	1	0					
E	2	1					
E	3	2					
D	3	3					

m

or
 2 columns = $2m \rightarrow O(m)$
 2 rows = $2n \rightarrow O(n)$

$\rightarrow O(\min(m, n))$

Example: DEED vs. DREAD filled by column

	ϵ	<i>D</i>	<i>R</i>	<i>E</i>	<i>A</i>	<i>D</i>
ϵ	0	1	2	3	4	5
<i>D</i>	1	0	1			
<i>E</i>	2	1	1			
<i>E</i>	3	2	2			
<i>D</i>	3	3	3			

Example: DEED vs. DREAD filled by column

	ϵ	<i>D</i>	<i>R</i>	<i>E</i>	<i>A</i>	<i>D</i>
ϵ	0	1	2	3	4	5
<i>D</i>	1	0	1	2		
<i>E</i>	2	1	1	1		
<i>E</i>	3	2	2	1		
<i>D</i>	3	3	3	2		

Example: DEED vs. DREAD filled by column

	ϵ	<i>D</i>	<i>R</i>	<i>E</i>	<i>A</i>	<i>D</i>
ϵ	0	1	2	3	4	5
<i>D</i>	1	0	1	2	3	
<i>E</i>	2	1	1	1	2	
<i>E</i>	3	2	2	1	2	
<i>D</i>	3	3	3	2	2	

Example: DEED vs. DREAD filled by column

	ϵ	<i>D</i>	<i>R</i>	<i>E</i>	<i>A</i>	<i>D</i>
ϵ	0	1	2	3	4	5
<i>D</i>	1	0	1	2	3	4
<i>E</i>	2	1	1	1	2	3
<i>E</i>	3	2	2	1	2	3
<i>D</i>	3	3	3	2	2	2

Computing in column order to save space

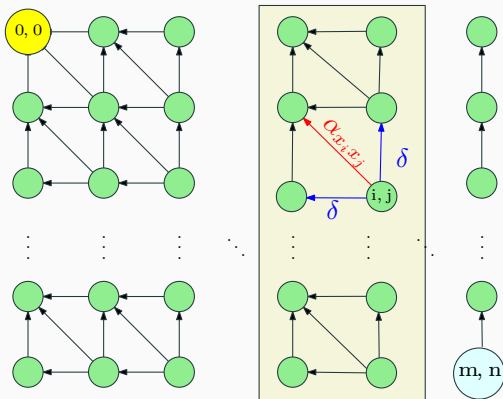


Figure 2: $M(i, j)$ only depends on previous column values. Keep only two columns and compute in column order.

Space Efficient Algorithm

```
for all  $i$  do  $N[i, 0] = i\delta$ 
for  $j = 1$  to  $n$  do
   $N[0, 1] = j\delta$  (* corresponds to  $M(0, j)$  *)
  for  $i = 1$  to  $m$  do
    
$$N[i, 1] = \min \begin{cases} \alpha_{x_i y_j} + N[i - 1, 0] \\ \delta + N[i - 1, 1] \\ \delta + N[i, 0] \end{cases}$$

  for  $i = 1$  to  $m$  do
    Copy  $N[i, 0] = N[i, 1]$ 
```

Analysis

Running time is $O(mn)$ and space used is $O(2m) = O(m)$

Analyzing Space Efficiency

- From the $m \times n$ matrix M we can construct the actual alignment (exercise)
- Matrix N computes cost of optimal alignment but no way to construct the actual alignment
- Space efficient computation of alignment? More complicated algorithm — see notes and Kleinberg-Tardos book.

Longest Common Subsequence Problem

LCS Problem

Definition

LCS between two strings X and Y is the length of longest common subsequence between X and Y .

ABAZDC → AB

BACBAD → AB

AB: Common Subsequence (CS)

ABAZDC

BACBAD

ABAD: Longest CS (LCS)

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Example

LCS between ABAZDC and BACBAD is 4 via ABAD

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Derive a **dynamic programming algorithm** for the problem.

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 $\max(LCS(A[1\dots m-1], B[1\dots n]), LCS(A[1\dots m], B[1\dots n-1]))$

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- Base Case: **A is empty** or **B is empty**

LCS recursive definition

$A[1..n], B[1..m]$: Input strings.

$$LCS(i, j) = \begin{cases} 0 & i = 0 \text{ or } j = 0 \\ \max \begin{pmatrix} LCS(i-1, j), \\ LCS(i, j-1) \end{pmatrix} & A[i] \neq B[j] \\ 1 + LCS(i-1, j-1) & A[i] = B[j] \end{cases}$$

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Running time: Similar to edit distance... $O(nm)$

Space: $O(m)$ space.

Longest common subsequence is just edit distance for the two sequences...

A, B : input sequences, Σ : "alphabet" all the different values in A and B

$$\forall b, c \in \Sigma : b \neq c$$

$$COST[\underline{b}][\underline{c}] = \underline{+\infty}.$$

$$\forall b \in \Sigma$$

$$COST[\underline{b}][\underline{b}] = \underline{1}$$

1 : price of deletion of insertion of a single character
or

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1 : price of deletion or insertion of a single character

LCS = 3

						ED	LCS
Maximum ED	D	R	E	A	D		
Min LCS						D	E E D
Sub-opt ED	D	R	E	A	D		
Sub-opt LCS					D	E	E D
Min ED	D	R	E	A	D		
Max LCS	D		E		E	D	
						9 + 0 = 9	
						8 + 1 = 9	$\begin{matrix} 5+4 \\ m+n \end{matrix}$
						6 + 3 = 9	

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$$\underline{\text{Length of longest common sub-sequence}} = \underline{m + n} - \underline{\text{ed}(A, B)}$$