

# Dynamic Programming Part 2

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## Contents

<b>1</b>	<b>Longest Increasing Subsequence</b>	<b>1</b>
1.1	Idea . . . . .	1
<b>2</b>	<b>Longest Common Subsequence</b>	<b>2</b>
2.1	Bivariate Recurrence . . . . .	2

## 1 Longest Increasing Subsequence

**Input:** an array  $A$  of  $n$  integers

**Output:** a longest increasing subsequence of  $A$  that does not need to be contiguous

**Remark:** there are  $2^n$  subsequences

### 1.1 Idea

A longest increasing subsequence  $S$  ending at  $A[i]$  looks like  $S = [..., A[j], A[i]] = s' + [A[i]]$ .

$S'$  is a longest increasing subsequence ending at  $A[j]$  or empty.

Don't know  $j$  but can try all  $j < i$  for which  $A[j] < A[i]$ .

`LongestIncreasingSubsequence(A[1..n]):`

```
L[1] = 1
for i in range(2, n):
    L[i] = 1
    for j in range(1, i-1):
```

```

        if A[j] < A[i]:
            L[i] = max(L[i], L[j] + 1)
    return max entry in L

```

**Runtime:**  $\Theta(n^2)$

**Remark:** the algorithm does not return the sequence itself, but could be modified to do so

## 2 Longest Common Subsequence

**Input:** arrays  $A$  and  $B$  of length  $n$  and  $m$  characters respectively

**Output:** the max length  $k$  of a common subsequence to  $A$  and  $B$  (no need to be contiguous)

**Remark:** there are  $2^n$  subsequences in  $A$  and  $2^m$  subsequences in  $B$

### 2.1 Bivariate Recurrence

**Definition:** let  $M[i, j]$  be the longest subsequence between  $A[1..i]$  and  $B[1..j]$

- $M[0, j] = 0$  for all  $j$
- $M[i, 0] = 0$  for all  $i$
- $M[i, j]$  is the max of up to 3 values:
  - $M[i, j - 1]$  (don't use  $B[j]$ )
  - $M[i - 1, j]$  (don't use  $A[i]$ )
  - $1 + M[i - 1, j - 1]$  if  $A[i] = B[j]$

The algorithm computes all  $M[i, j]$  using 2 nested loops so runtime  $\Theta(mn)$