**Dynamic Programming**

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[Leetcode discussion](https://leetcode.com/discuss/post/5190444/dynamic-programming-problems-category-wi-7xpk/)

# **Theory**

## What is Dynamic Programming?

Dynamic programming (DP) is an optimization technique used to solve complex problems by breaking them down into simpler subproblems and storing the results of these subproblems to avoid redundant computations. It is particularly useful for problems with overlapping subproblems and optimal substructure properties.

**Key Concepts of Dynamic Programming:**

1. **Overlapping Subproblems**: Solving the same subproblems multiple times.
2. **Optimal Substructure**: The optimal solution to a problem can be constructed from optimal solutions of its subproblems.
3. **Memoization**: Storing results of subproblems to reuse them, thus saving computation time. Also called top-down approach.
4. **Tabulation**: Building a table in a bottom-up manner to store solutions of subproblems.

**Steps to approach Dynamic Programming problems:**

1. Find the recurrence relationship
2. Find the base case
3. Find way to store solutions of subproblems

# **Sample Code**

Fibonacci Series – Recursion:

def fib(n):

    if n<=1:

        return n

    else:

        return fib(n-1) + fib(n-2)

# fib : 0,1,1,2,3,5,8

print(fib(6)) #8

Here we see, our solution finding solution of same subproblems several time, which cause slowness. We can prevent this using DP, by saving solution of subproblems in an array.

Fibonacci Series –Top down:

Here we built solution from n to i, normally via recursion.

def fib\_top\_down(n, dp):

    if n<=1:

        return n

    if dp[n]!= -1:

        return dp[n]

    else:

        dp[n]= fib\_top\_down(n-1, dp) + fib\_top\_down(n-2, dp)

        return dp[n]

n=6

dp = [-1]\*(n+1)

print(fib\_top\_down(n, dp)) #8

Fibonacci Series – Bottom Up:

Here we built solution from i to n, normally via loop.

def fib\_bottom\_up(n):

    dp = [-1]\*(n+1)

    dp[0], dp[1] = 0, 1

    for i in range(2,n+1):

        dp[i] = dp[i-1] + dp[i-2]

    return dp[n]

print(fib\_bottom\_up(6)) #8

Fibonacci Series – Bottom Up – Space Optimized:

In place of dp array we can use 2 variables for same purpose

def fib\_bottom\_up(n):

    prev = 1

    prev1 = 0

    curr = 0

    for i in range(2,n+1):

        curr = prev + prev1

        prev1 = prev

        prev = curr

    return prev

print(fib\_bottom\_up(6)) #8

# LEVEL 1: **Linear DP**

For such questions you need to find the repetitive part of solution and improve it by saving result in array or somewhere, classic example is Fibonacci series. Also use when there is multiple way to do something and we need optimal way, min-max scenarios.

### Climbing Stairs

Link: <https://leetcode.com/problems/climbing-stairs/description/>

### House Robber

Link: <https://www.geeksforgeeks.org/problems/maximum-money2855/1>

### House Robber 2

Link: <https://leetcode.com/problems/house-robber-ii/>

### Reach to one

Given a number x, you can do 3 different operations on x: #1. Subtract 1 from it. #2 If it is divisible by 2, divide by 2. #3 If it is divisible by 3, divide by 3 .Find the minimum number of steps that it takes to get to 1 using only the above operations.

### Jump Game

Link: <https://leetcode.com/problems/jump-game/>

### Stones

Link: <https://atcoder.jp/contests/dp/tasks/dp_k>

# LEVEL 2: **Multi-Dimensional DP**

### Ninja training

Link: <https://www.naukri.com/code360/problems/ninja-s-training_3621003>

# LEVEL 3: **String DP**

These problems focus on operations over strings, such as finding subsequences or transformations.

### Longest Increasing Subsequence(LIS)

Link: <https://leetcode.com/problems/longest-increasing-subsequence/description/>

### Longest common subsequence

Link: <https://leetcode.com/problems/longest-common-subsequence/description/>

### Longest repeating subsequence

Link: [https://www.naukri.com/code360/problems/longest-repeating-subsequence](https://www.geeksforgeeks.org/problems/longest-repeating-subsequence2004/1)

# LEVEL 4: **DP on Grid**

These problems involve navigating a 2D grid, making optimal decisions at each cell.

### Unique Paths

Link: <https://leetcode.com/problems/unique-paths/description/>

### Minimum path sum

Link: <https://leetcode.com/problems/minimum-path-sum/>

# Type 5: **Knapsack**

These problems revolve around selecting items with given weights and values to maximize or minimize a certain criterion.

### Knapsack-1

Link: <https://atcoder.jp/contests/dp/tasks/dp_d>

### Coin Change

Link: <https://leetcode.com/problems/coin-change/>

# Type 6: **Kadane’s Algorithm**

Kadane’s Algorithm is a dynamic programming approach used to solve the Maximum Subarray Sum Problem — i.e., to find the contiguous subarray within a 1D array of numbers that has the largest sum.

### Maximum Subarray

Link: <https://leetcode.com/problems/maximum-subarray/description/>

### Best time to Buy or Sell Stock

Link: <https://leetcode.com/problems/best-time-to-buy-and-sell-stock/description/>

### Best time to Buy or Sell Stock II

Link: <https://leetcode.com/problems/best-time-to-buy-and-sell-stock-ii/>

# LEVEL 7: **Difficult**

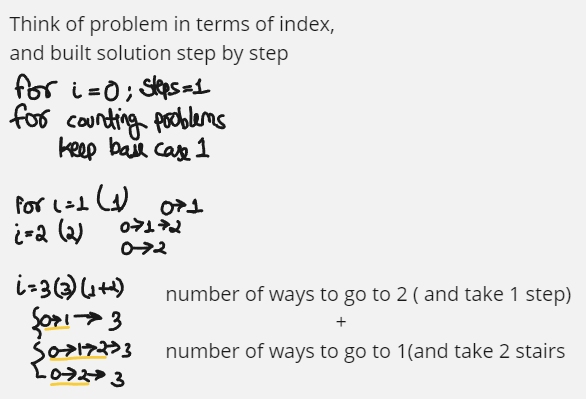
### K-ordered longest common subsequence

A k-ordered LCS is defined to be the LCS of two sequences if you are allowed to change at most k elements in the first sequence to any value you wish to. You are given 2 integer sequences and a number k. You can make max k changes in sequence 1 to get maximum LCS, find the max length of LCS.

# **SOLUTIONS:**

## **LEVEL 1: Linear DP**

1. **Climbing Stairs**



# for n = number of ways to go to n-1 + number of ways to go to n-2 (Fibonacci series)

# 0->1 for making code easy

# 1->1

# 2 -> 2  (0->2 , 0->1->2)

class Solution:

    def climbStairs(self, n: int) -> int:

        dp = [1]\*(n+1)

        for i in range(2,n+1):

            dp[i] = dp[i-1] + dp[i-2]

        return dp[n]

class Solution:

    def climbStairs(self, n: int) -> int:

        def helper(n):

            if n==0: return 1

            if n==1: return 1

            if dp[n]!=-1 : return dp[n]

            dp[n] = helper(n-1) + helper(n-2)

            return dp[n]

        dp = [-1]\*(n+1)

        return helper(n)

1. **House Robber**

F[i] = max loot done till ith house, so F[i] = max ( arr[i] + F[i-2] , F[i-1] )

*#Loot HOUSE*

def lootBU(n,arr):

    dp=[0]\*(n)

    dp[0],dp[1] = arr[0],max(arr[0],arr[1])

    for i in range(2,n):

        dp[i] = max(arr[i]+dp[i-2] ,dp[i-1])

    print(dp)

    return dp[n-1]

arr = [6,2,3,9]

print(lootBU(len(arr),arr))

1. **House Robber 2**

Just like previous question,

Only difference is first and last house can not be looted together. So we will run the code to both subarrays, one without first house and one without last house and will return maximum from wither case.

def helper(*house*):

    def maxLoot(*n*,*dp*):

*if* *n*<0: *return* 0

*if* *n*==0: *return* *house*[0]

*if* *dp*[*n*]!=-1: *return* *dp*[*n*]

*dp*[*n*] = max(maxLoot(*n*-1,*dp*) , maxLoot(*n*-2,*dp*)+*house*[*n*])

*return* *dp*[*n*]

    n = len(*house*)

    dp=[-1]\*n

*return* maxLoot(n-1,dp)

def houseRobber(*valueInHouse*):

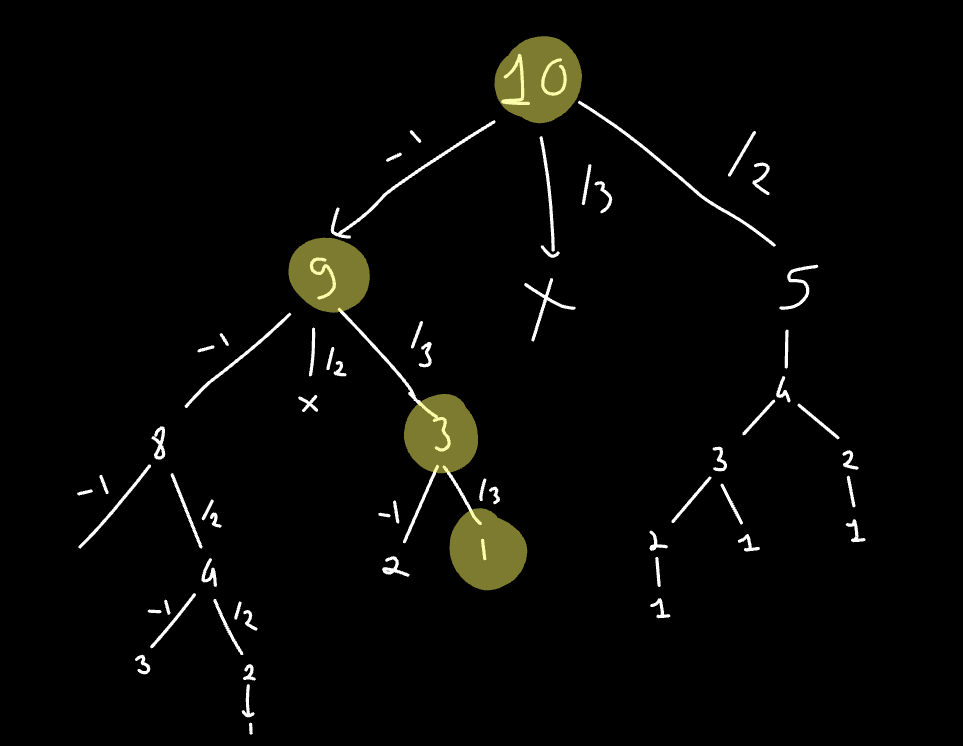
*if* len(*valueInHouse*)==1:

*return* *valueInHouse*[0]

*return* max(helper(*valueInHouse*[:-1]) , helper(*valueInHouse*[1:]))

1. **Minimum steps to reach one**

If see local minimum, going from 10 to 5 is preferred than 10 to 9, however 5 takes more time to go to 1 than 9. Therefore in global way going to 9 is preferred.



*#Top down*

def minStepsToOneTD(n,dp):

    if n==1: return 0

    if n==2 or n==3: return 1

    if dp[n]!=0: return dp[n]

    div\_by\_3, div\_by\_2, less\_by\_1 = float('inf'),float('inf'),float('inf')

    if(n%3==0):

        div\_by\_3 = 1+minStepsToOneTD(n//3,dp)

    if(n%2==0):

        div\_by\_2 = 1+minStepsToOneTD(n//2,dp)

    less\_by\_1 = 1+minStepsToOneTD(n-1,dp)

    dp[n]=min(div\_by\_3, div\_by\_2, less\_by\_1)

    return dp[n]

n=7

dp=[0]\*(n+1)

print(minStepsToOneTD(n,dp))

*#Bottom Up*

def min\_steps\_to\_one(x):

    dp = [1]\*(x+1)

    dp[1]=0

    dp[2],dp[3]=1,1

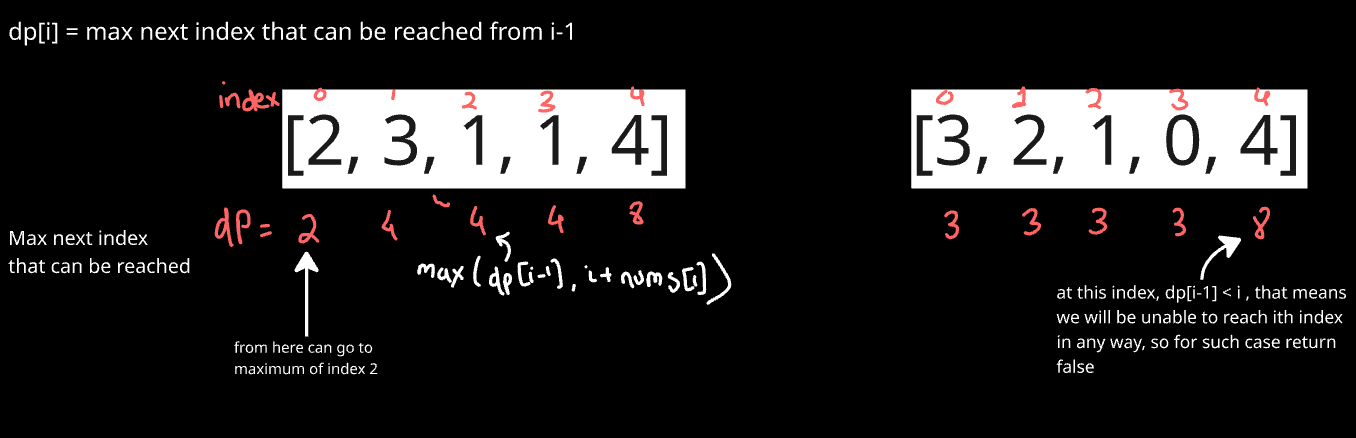
    for i in range(4,x+1):

        dp[i]=1+min(dp[i-1], dp[i//2] if i%2==0 else x, dp[i//3] if i%3==0 else x)

    return dp[x]

print(min\_steps\_to\_one(10))

1. **Jump Game**



class Solution:

    def canJump(self, nums: List[int]) -> bool:

        n = len(nums)

        dp=[0]\*(n)

        dp[0] = nums[0]

        for i in range(1,n):

            if dp[i-1]<i:

                return False

            dp[i] = max(i+nums[i], dp[i-1])

        return True

**6. Stone**

If k==0 : any player who reaches this state will lose

If k<min(a): here also if any player reaches this state will lose

So winning or losing depends on state and independent of who plays. So if state=k is winning state, player one wins else player 2.

State K is winning state if any state ( k-a[i] ) is losing state for all a[i] in array a. Meaning if first player can push second player to any losing state then first player can win. But if all ( k-a[i] ) states are winning states, then kth state is losing state.

n , k = map(int,input().split())

a = list(map(int,input().split()))

dp = [-1]\*(k+1)

#at state 0 all will lose

dp[0]=0

#if a=[2,3] so at 2 and 3, player takes all stone and next player won't have any stone to pick.

#so that will be winning state.

for i in range(1,k+1):

    flag=0

    for j in a:

        if j<=i and dp[i-j]==0: #can send next player to any losing state

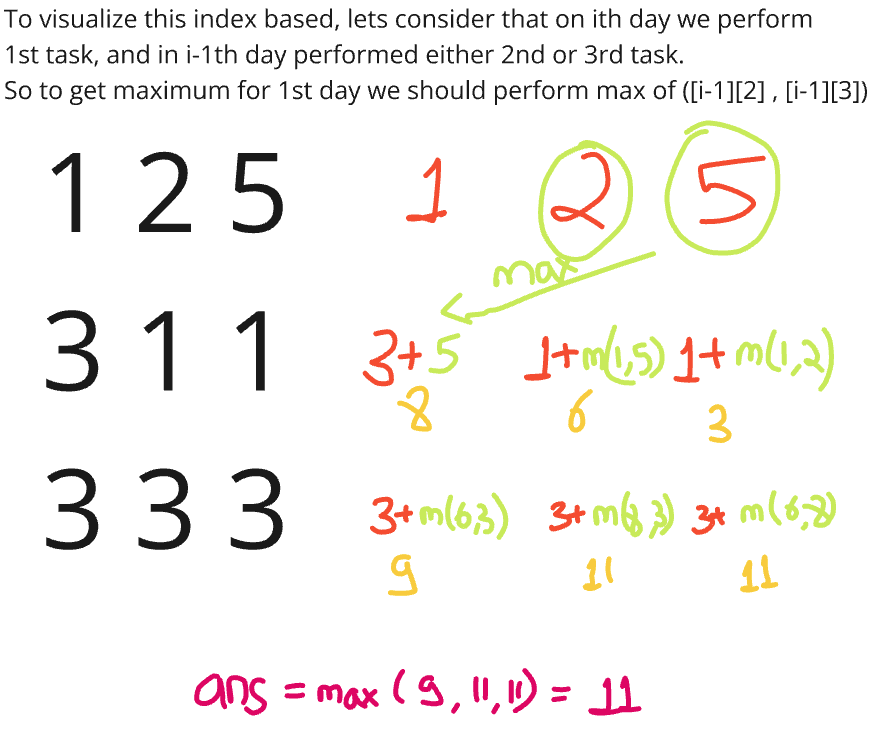
            flag=1

    dp[i]=flag

print("First" if dp[k]==1 else "Second")

## **TYPE 2: Multi-Dimensional DP**

1. **Ninja Training**



Can further do space optimization as in each step we just need to get the previous state. So in place of saving whole dp, just save the previous state.

def ninjaTraining(n: int, points: List[List[int]]) -> int:

    # Write your code here.

    dp = [[0]\*3 for i in range(n)]

    dp[0][0] = points[0][0]

    dp[0][1] = points[0][1]

    dp[0][2] = points[0][2]

    for i in range(1,n):

        dp[i][0] = points[i][0] + max(dp[i-1][1] , dp[i-1][2])

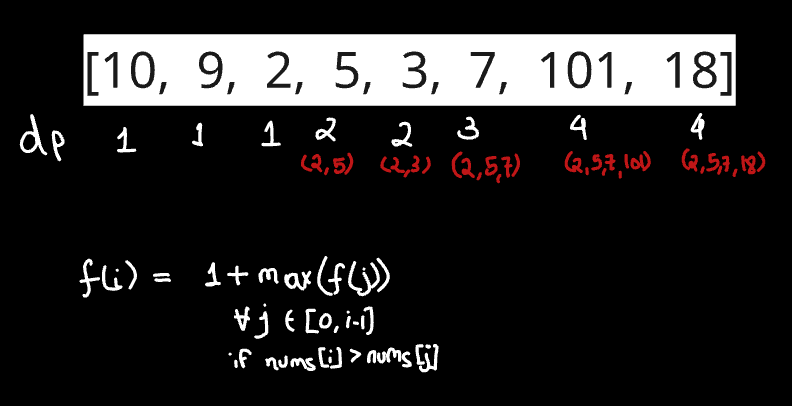
        dp[i][1] = points[i][1] + max(dp[i-1][0] , dp[i-1][2])

        dp[i][2] = points[i][2] + max(dp[i-1][0] , dp[i-1][1])

    return max(dp[n-1])

## **TYPE 3: String DP**

1. **Longest Increasing subsequence**



class Solution:

    def lengthOfLIS(self, nums: List[int]) -> int:

        n = len(nums)

        dp =[1]\*n

        for i in range(n):

            for j in range(i):

                if nums[i]>nums[j]:

                    dp[i] = max(dp[i] , 1+dp[j])

        return max(dp)

**2. Longest common subsequence**

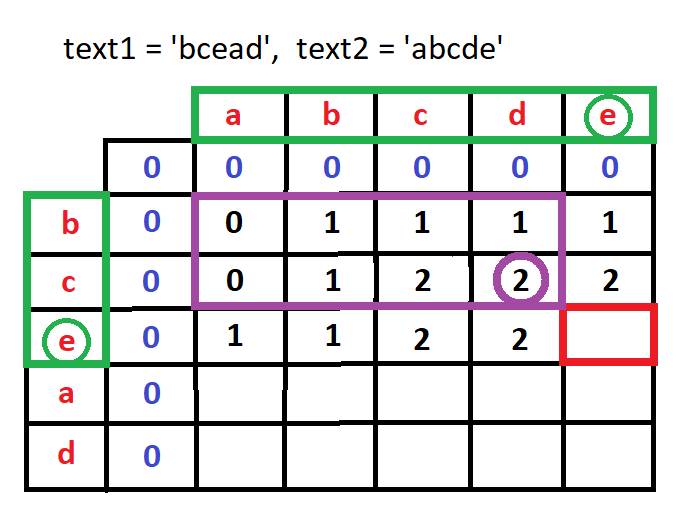
Here recursive relation is:

Eg:- st1= “abc” , st2=“adb” , I know st1[0] == st2[0],

So to get lcs, do **lcs(“abc”, “adb”) = 1 + lcs(“bc”,”db”)**

Eg2:- st1= “pqrs” , st2= “xqor” , I know st1[0] != st2[0]

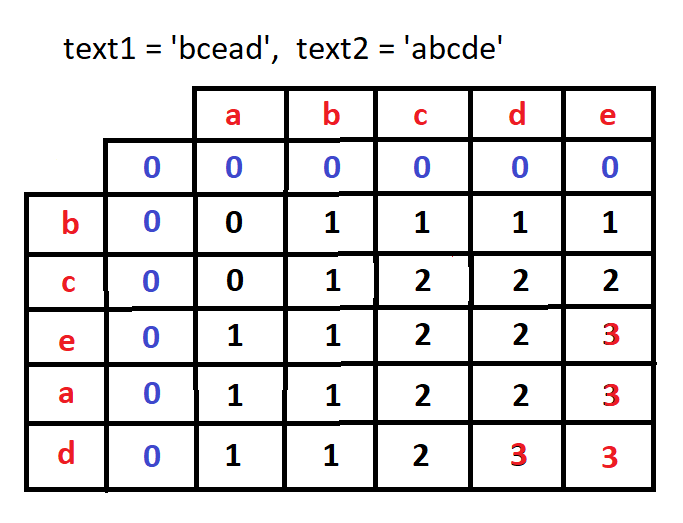
So to get lcs, do **lcs(“pqrs”, “xqor”) = max( lcs(“qrs” , “xqor”) , lcs(“pqrs”, “qor”) )**



For this red box:

Till now, text1=”bce” and text2=”abcde”

Now since last values are equal, it is equal to 1+lcs(“bc”, “abcd”) =3



class Solution:

    def longestCommonSubsequence(self, text1: str, text2: str) -> int:

        n,m=len(text1),len(text2)

        dp = [[0]\*(m+1) for i in range(n+1)] #m+1 rows and n+1 columns

        for i in range(1,n+1):

            for j in range(1,m+1):

                if text1[i-1]==text2[j-1]:

                    dp[i][j] = 1+dp[i-1][j-1]

                else:

                    dp[i][j] = max(dp[i-1][j] ,dp[i][j-1])

        return dp[n][m]

**## Demo run**

       ''  a  c  e

 ''    [0, 0, 0, 0]

  a    [0, 1, 1, 1]

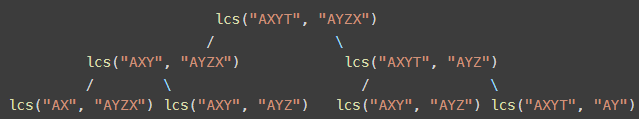
  b    [0, 1, 1, 1]

  c    [0, 1, 2, 2]

  d    [0, 1, 2, 2]

  e    [0, 1, 2, 3]

**For top down:** Will start looking from last values str[-1] and str[-2]



#Normal Recursion

class Solution:

    def longestCommonSubsequence(self, text1: str, text2: str) -> int:

        n = len(text1)

        m = len(text2)

        def helper(i, j):

            if i==0 or j==0: return 0

            if text1[i-1] == text2[j-1]:

                return helper(i-1,j-1)+1

            else:

                return  max(helper(i-1, j), helper(i, j-1))

        return helper(n,m)

#Recursion + memorization = DP

class Solution:

    def longestCommonSubsequence(self, text1: str, text2: str) -> int:

        n, m = len(text1), len(text2)

        dp = [[-1]\*(m+1) for \_ in range(n+1)]

        def helper(i,j):

            if i==0 or j==0:

                return 0

            if dp[i][j]!=-1:

                return dp[i][j]

            if text1[i-1]==text2[j-1]:

                dp[i][j] =  1+helper(i-1,j-1)

            else:

                dp[i][j] = max(helper(i-1,j),helper(i,j-1))

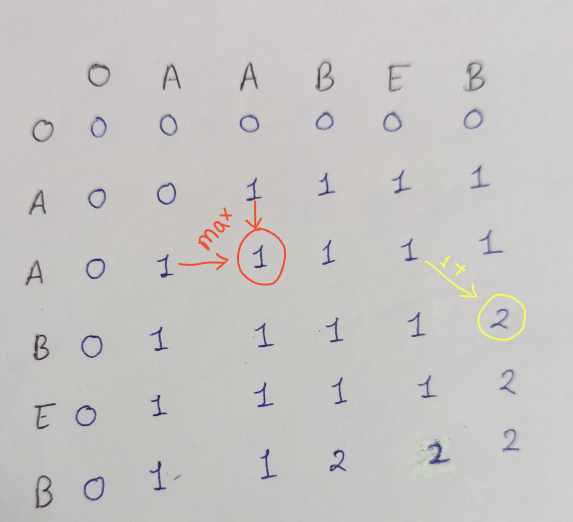
            return dp[i][j]

        return helper(n,m)

**3. Longest repeating subsequence**

Treat it as same as LCS on same string

But equal condition will have extra check that i != j,



class Solution:

    def LongestRepeatingSubsequence(self, s):

        n = len(s)

        dp = [[0]\*(n+1) for \_ in range(n+1)]

        for i in range(1,n+1):

            for j in range(1,n+1):

                if i!=j and s[i-1]==s[j-1]:

                    dp[i][j] = 1+dp[i-1][j-1]

                else:

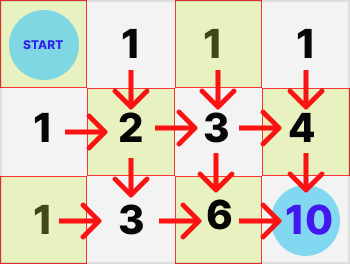
                    dp[i][j] = max(dp[i-1][j], dp[i][j-1])

        return dp[n][n]

## **TYPE 4: DP on Grids**

1. **Unique Paths**

To go to m,n. we have 2 options. (1) go down from n-,m (2)go right from n,m-1. By this logic. For all m=0 or n=0 there is only one way. So value=1.



class Solution:

    def uniquePaths(self, m: int, n: int) -> int:

        dp = [[1] \* n for i in range(m)]

        for i in range(1, m):

            for j in range(1, n):

                dp[i][j] = dp[i - 1][j] + dp[i][j - 1]

        return dp[m - 1][n - 1]

**Using combination**

For any given M x N grid, each unique path (no matter which one it is) requires you to move right from the starting point N - 1 times and move down from the starting point M - 1 times. Hence, regardless of the order you choose to move right or down, you need to make a total of (M - 1) + (N - 1) = M + N - 2 moves.

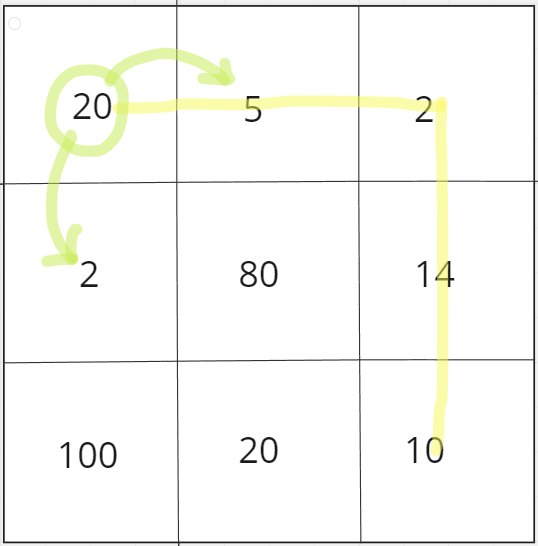
Then, out of the M + N - 2 moves, we need to select M - 1 moves to move right and the remaining N - 1 moves to move down. This essentially is why this problem boils down to combinatorics, because we need to calculate how many different ways we can select M - 1 moves from M + N - 2 moves (or equivalently, N - 1 moves from M + N - 2 moves).

class Solution:

    def uniquePaths(self, m: int, n: int) -> int:

        return math.comb(m+n-2, m-1)  # or math.comb(m+n-2, n-1)

**2. Minimum Path Sum**



If we use greedy strategy here. Then we must be going to 2, instead of 5.

But 5 gives us best results globally. So for this case as we have global consideration to get best results, we use dp to check all possible paths and get best results.

class Solution:

    def minPathSum(self, grid: List[List[int]]) -> int:

        n = len(grid)

        m = len(grid[0])

        dp = [[0]\*m for \_ in range(n)]

        for i in range(n):

            for j in range(m):

                if i==0 and j==0:

                    dp[i][j] = grid[i][j]

                else:

                    up,left = float('inf'),float('inf')

                    if i>0:

                        up = grid[i][j] + dp[i-1][j]

                    if j>0:

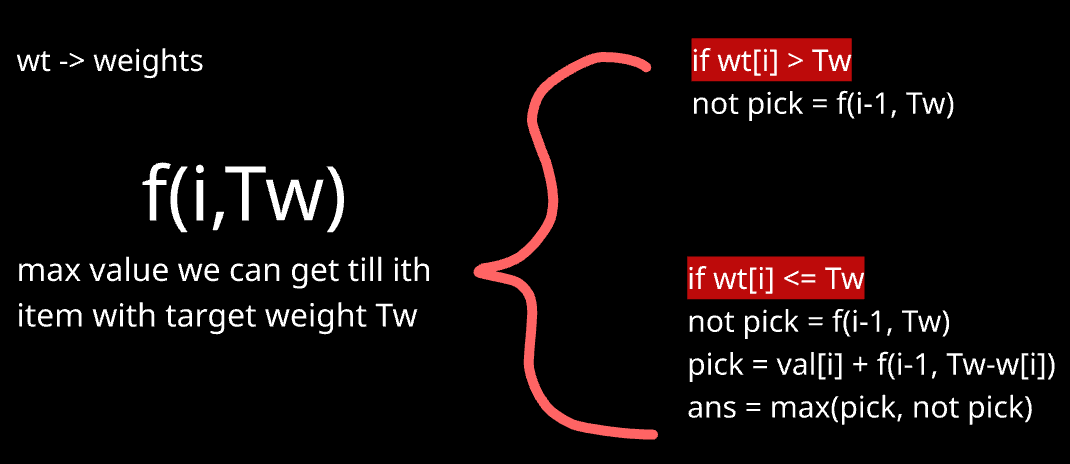
                        left = grid[i][j] + dp[i][j-1]

                    dp[i][j] = min(up,left)

        return dp[n-1][m-1]

## **TYPE 5: Knapsack**

1. **Knapsack-1**



iin=lambda:map(int,input().split())

def knapsack(i, Tw, dp, wt, val):

    if i == 0:

        if wt[0] <= Tw:

            return val[0]

        return 0

    if dp[i][Tw] != -1:

        return dp[i][Tw]

    not\_pick = knapsack(i-1, Tw, dp, wt, val)

    pick = 0

    if wt[i] <= Tw:

        pick = val[i] + knapsack(i-1, Tw-wt[i], dp, wt, val)

    dp[i][Tw] = max(pick, not\_pick)

    return dp[i][Tw]

#main method

N,Tw=iin()

wt,val = [],[]

for \_ in range(N):

    w,v = iin()

    wt.append(w)

    val.append(v)

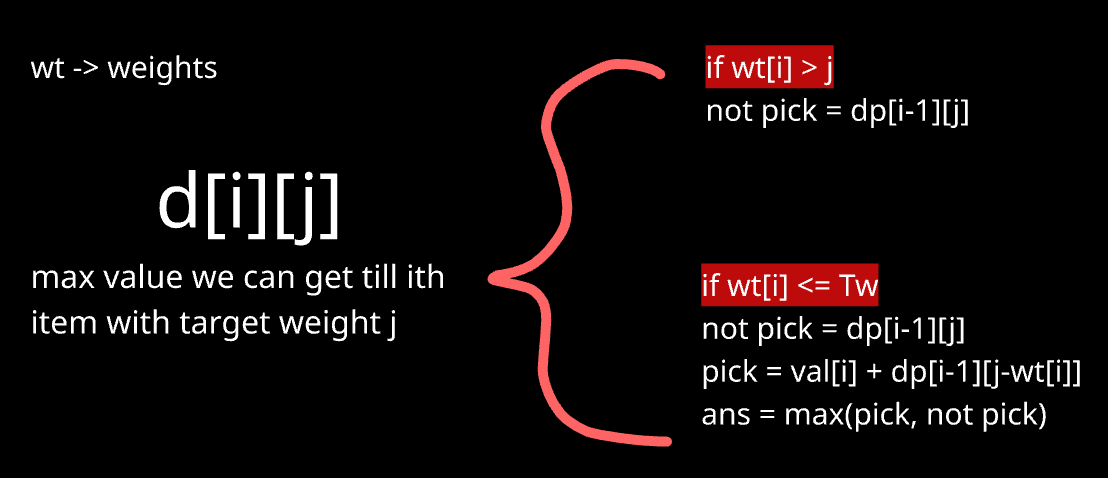
# Initialize dp with -1

dp = [[-1] \* (Tw + 1) for \_ in range(N)]

ans= knapsack(N-1,Tw,dp,wt,val)

print(ans)

Bottom up



iin=lambda:map(int,input().split())

#main method

N,Tw=iin()

wt,val = [],[]

for \_ in range(N):

    w,v = iin()

    wt.append(w)

    val.append(v)

# Initialize dp with -1

dp = [[0] \* (Tw + 1) for \_ in range(N+1)]

#fill row wise (0->tw+1)

for i in range(1,N+1):

    for j in range(1,Tw+1):

        if(wt[i-1]<=j):

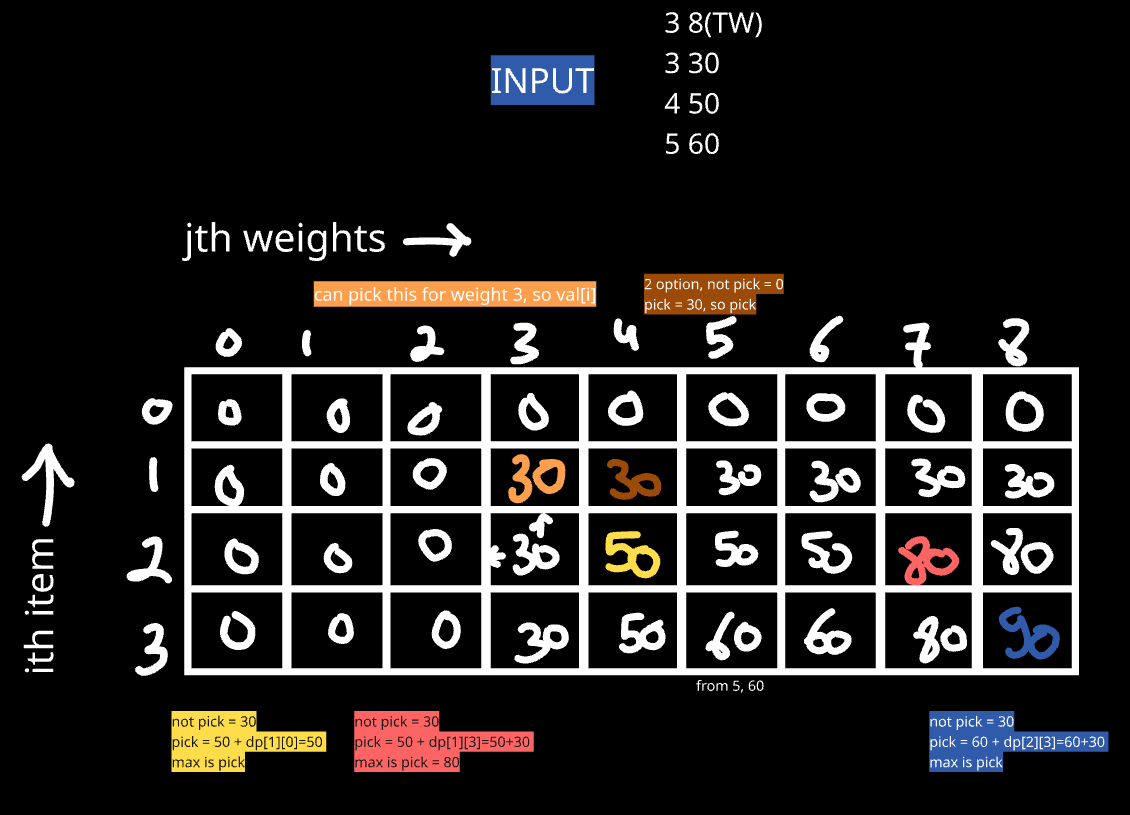
            dp[i][j]=max(dp[i-1][j], val[i-1] + dp[i-1][j-wt[i-1]])

        else:

            dp[i][j]=dp[i-1][j]

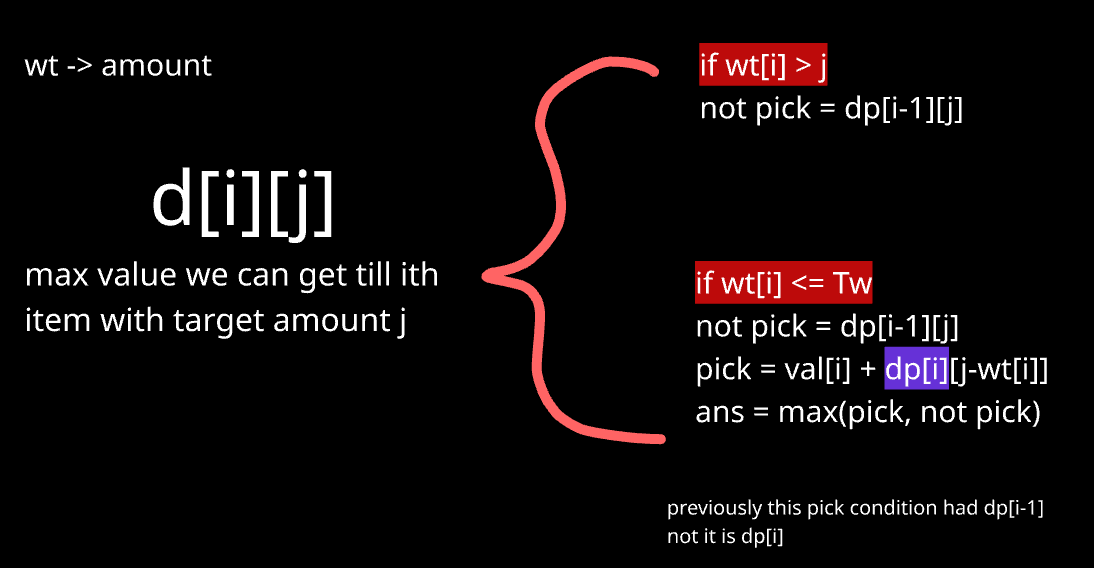
print(dp)

print(dp[N][Tw])



1. **Coin Change**

**Approach 1 :** Just like previous one, but there since same value can’t be pick twice, so in pick case we were doing dp[i-1], but here same I can be picked again, so in pick condition don’t change to i-1 and check again for ith



For base condition

Since we iterate row wise, we have to fill 0th row for base condition. now if we have target and only coins[0] as coin, so maximum coins we can put is target//coins[0]. So will do that.

class Solution:

    def coinChange(self, coins: List[int], amount: int) -> int:

        n = len(coins)

        dp = [[0]\*(amount+1) for \_ in range(n)]

        for t in range(amount+1):

            if t%coins[0]==0 :

                dp[0][t] = t//coins[0]

            else:

                dp[0][t] = float('inf')

        #moving row wise

        for i in range(1,n):

            for j in range(amount+1):

                not\_pick = dp[i-1][j]

                pick = float('inf')

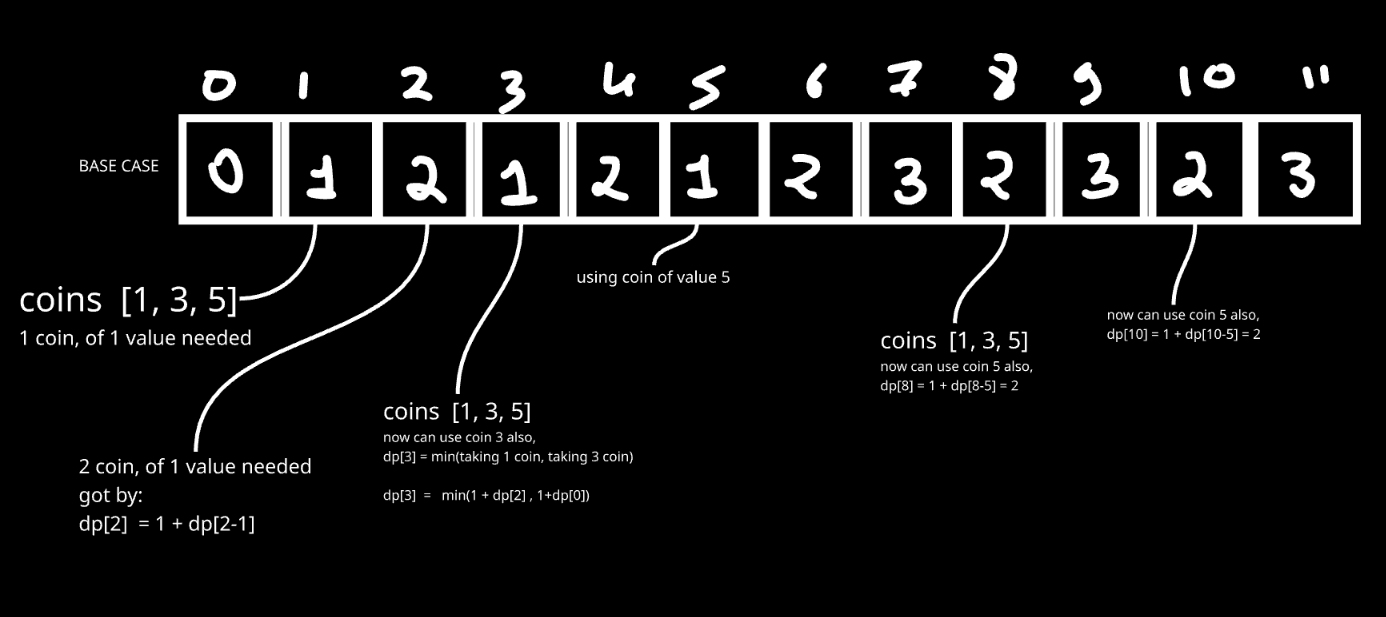
                if coins[i] <= j:

                    pick = 1 + dp[i][j-coins[i]]

                dp[i][j] = min(pick,not\_pick)

        return dp[n-1][amount] if dp[n-1][amount]!=float('inf') else -1

**Approach 2 :** Just loop from 0 to amount, and for each value, check which coin from coins give minimum ans and put that in dp



#Bottom up  (performs better both space and time wise)

class Solution:

    def coinChange(self, coins: List[int], amount: int) -> int:

        dp = [float('inf')]\*(amount+1)

        dp[0]=0

        for i in range(1,amount+1):

            for j in coins:

                if j<=i:

                    dp[i] = min(dp[i], 1+dp[i-j])

        return dp[amount] if dp[amount]!=float('inf') else -1

#Top down

class Solution:

    def coinChange(self, coins: List[int], amount: int) -> int:

        def helper\_td(n , coins , dp):

            if n==0: return 0

            if dp[n]!=-1:return dp[n]

            temp = float('inf')

            for i in coins:

                if i <= n:

                    temp = min(temp,1+helper\_td(n-i,coins,dp))

            dp[n] = temp

            return dp[n]

        dp = [-1]\*(amount+1)

        ans=helper\_td(amount , coins, dp)

## **TYPE 6: Kadane’s Algorithm**

1. **Maximum Subarray**

class Solution:

    def maxSubArray(self, nums: List[int]) -> int:

        curr=0

        maxSum=nums[0]

        for num in nums:

            if curr<0:  #when sum becomes negative, start from 0

                curr = 0

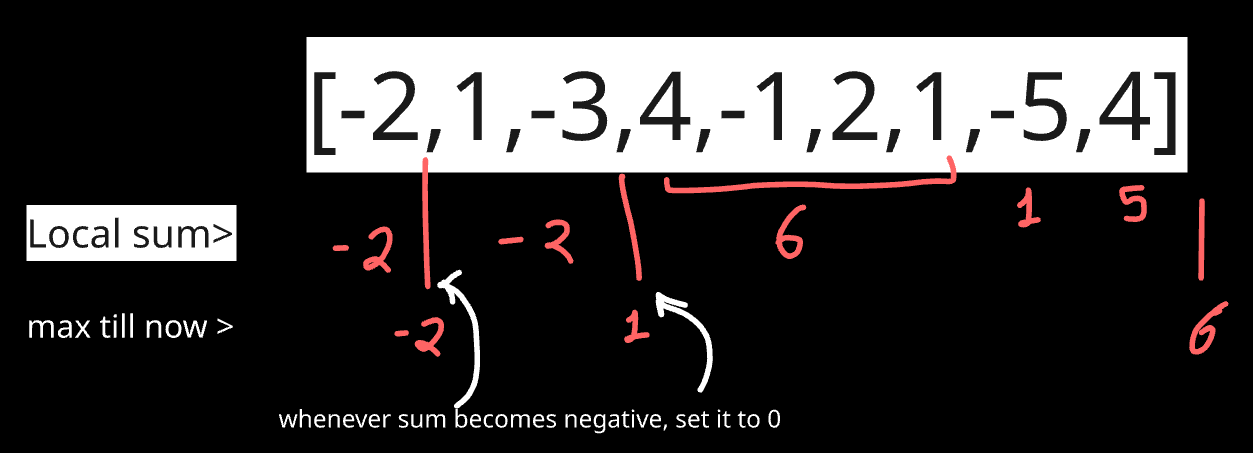
            curr+=num

            maxSum = max(maxSum,curr)

        return maxSum

At each position i, we have two choices:

* Extend the previous subarray: max\_ending\_here + arr[i]
* Start a new subarray from current element: arr[i]



**Why is Kadane’s Algorithm Dynamic Programming?**

Because it builds the solution using:

* **Optimal substructure**: Solution at index i depends on solution at i-1.
* **Overlapping sub problems**: Repeated computation of subarray sums.
* But it uses **space optimization** (constant space), which is why it may not seem like "typical" DP.

1. **Best time to buy and sell stock**

To get maximum profit on ith day, you have to buy that at minimum price till 1->i-1th day. So we will keep track of minimum value.

class Solution:

    def maxProfit(self, prices: List[int]) -> int:

        min\_price = prices[0]

        profit=0

        for price in prices[1:]:

            profit = max(profit, price-min\_price)

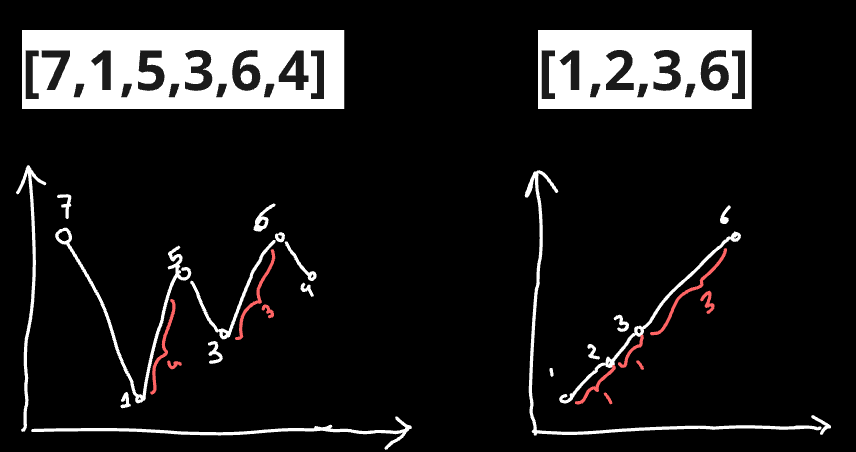
            min\_price = min(min\_price, price)

        return profit

1. **Best time to buy and sell stock II**

Here we can buy and sell multiple times, so better is that buy every time possible if next one is having more price.

Check from graph, whenever there is rise.



class Solution:

    def maxProfit(self, prices: List[int]) -> int:

        profit=0

        for i in range(1,len(prices)):

            if prices[i]-prices[i-1]>0:

                profit += prices[i]-prices[i-1]

        return profit

## **LEVEL 7:**

1. **K-ordered LCS**

Same as like LCS, but one new case.

New case. Now we can change k values in seq1 to make it’s ith value equal to jth value of seq2.

So when seq1[i] != seq2[j], then 2 cases, (1) we can use k and make both values equal, (2) don’t use k and proceed as normal. Final ans is maximum of both.

def KOrderedLCS(seq1, seq2, k):

    def helper(i, j, k, dp):

        if i==0 or j==0: return 0

        if dp[i][j][k]!=-1:

            return dp[i][j][k]

        if seq1[i-1] == seq2[j-1]:

            dp[i][j][k] = 1+ helper(i-1,j-1,k,dp)

        else:

            if k>0:

                #we replacing a value in seq1,so will act as seq1[i]==seq2[j]

                temp = 1+ helper(i-1 ,j-1 , k-1, dp)

                #Case we not replacing any value

                temp2 =  max(helper(i-1,j ,k ,dp), helper(i, j-1, k, dp))

                dp[i][j][k] = max(temp, temp2)

            else:

                dp[i][j][k] = max(helper(i-1, j, k, dp), helper(i, j-1, k, dp))

        return dp[i][j][k]

    n,m=len(seq1),len(seq2)

    # dp of size n\*m\*k

    dp = [[[-1 for \_ in range(k+1)] for \_ in range(m+1)] for \_ in range(n+1)]

    return helper(n,m,k,dp)

seq1 = [1, 2, 3, 4, 5]

seq2 = [5, 3, 1, 4, 2]

k=1

print(KOrderedLCS(seq1, seq2, k))

**## Demo run**

eg:-

seq1 = [1, 2, 3, 4, 5]

seq2 = [5, 3, 1, 4, 2]

k = 1

Ans = 3

You can change the first element of the first sequence to 5 to get the LCS comprising of the sequence (5, 3, 4)

Bottom Up

def KOrderedLCS(seq1, seq2, k):

    n, m = len(seq1), len(seq2)

*# Initialize the dp table with zero values*

    dp = [[[0 for \_ in range(k+1)] for \_ in range(m+1)] for \_ in range(n+1)]

*# Iterate through all positions of seq1 and seq2*

    for i in range(1, n+1):

        for j in range(1, m+1):

            for z in range(k+1):

                if seq1[i-1] == seq2[j-1]:

                    dp[i][j][z] = dp[i-1][j-1][z] + 1

                else:

                    if z > 0:

*# Option 1: Replace seq1[i-1] with seq2[j-1]*

                        replace = dp[i-1][j-1][z-1] + 1

*# Option 2: Do not replace, just move in one of the sequences*

                        dont\_replace = max(dp[i-1][j][z], dp[i][j-1][z])

                        dp[i][j][z] = max(replace, dont\_replace)

                    else:

*# When no replacements are allowed*

                        dp[i][j][z] = max(dp[i-1][j][z], dp[i][j-1][z])

    return dp[n][m][k]

*# Read input*

seq1 = [1, 2, 3, 4, 5]

seq2 = [5, 3, 1, 4, 2]

k=1

*# Print the result*

print(KOrderedLCS(seq1, seq2, k))