Dynamic Programming, Graph Algorithm Problems and Their Solutions In Code

compiled by D. Gueorguiev, 8/21/2024

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# Problems Solvable by Dynamic Programming

## The N-th Fibonacci Number

//TODO: Finish the section on the N-th Fibonacci number

## Rod Cutting Problem

Given a rod of length inches (or centimeters) and a table of prices for find the max obtainable revenue by cutting the rod and selling the pieces.

The problem must exhibit optimal substructure if in order to solve the original problem of size we solve similar problems of the same type but smaller sizes.

(1)

Execution time complexity:

(2)

The solution of (2) is given by

(3)

(3) represents exponential time complexity.

Let us assume we are given an integer n and a list p with len(p) >= n. We want to find the max revenue and a partition of n which achieves it.

in code:

def cut\_rod(n: int, p: list[int]):

if n == 0:

return 0

//TODO: Finish the section on the rod cutting problem

## Min Cost Path

Given a cost matrix cost[][] and a position (M,N) in cost[][], write a function that returns cost of minimum cost path to reach (M,N) from (0,0). Each cell of the matrix represents a cost to traverse through that cell. The total cost of a path to reach (M,N) is the sum of all the costs on that path (including both source and destination). You can only traverse down , right, and diagonally lower cells from a given cell, i.e. from a given cell (i,j), cells (i+1,j), (i,j+1), and (i+1,j+1) can be traversed.

Example:

1, 2, 3

4, 8, 2

1, 5, 3

The path with min cost is (0,0)->(0,1)->(1,2)->(2,2). The cost of the path is 1+2+2+3=8.

Naïve implementation of the solution using recursive algorithm

import sys

def minCost(cost, m, n):

if n < 0 or m < 0:

return sys.maxsize

elif m == 0 and n == 0:

return cost[m][n]

else:

return cost[m][n] + min(minCost(cost, m-1, n-1),

minCost(cost, m-1, n),

minCost(cost, m, n-1))

Slightly better implementation using recursive algorithm with memoization

import sys

def minCostMemoized(cost, m, n):

if n < 0 or m < 0:

return sys.maxsize

elif m == 0 and n == 0:

return cost[m][n]

if memo[m][n] != -1:

return memo[m][n]

memo[m][n] = cost[m][n] + min(

minCostMemoized(cost, m-1, n-1),

minCostMemoized(cost, m-1, n),

minCostMemoized(cost, m, n-1))

Here memo is a list of lists with sizes equal to the row length and col length of the cost matrix initialized originally to -1:

memo = [[-1] \* C for \_ in range(R)] # R = len(cost) and C = len(cost[0])

Dynamic Programming formulation and implementation of the min cost problem:

minCost(1,1)

minCost(0,0)

minCost(0,1)

minCost(1,0)

minCost(1,2)

minCost(0,1)

minCost(0,2)

minCost(1,1)

minCost(2,1)

minCost(1,0)

minCost(1,1)

minCost(2,0)

minCost(2,2)

def minCost(cost, m, n):

tot\_cost = [[0 for x in range(n+1)] for x in range(m+1)]

tot\_cost[0][0] = cost[0][0]

# Initialize first column of tot\_cost array

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

tot\_cost[i][0] = tot\_cost[i-1][0] + cost[i][0]

# Initialize first row of tot\_cost array

for j in range(1, n+1):

tot\_cost[0][j] = tot\_cost[0][j-1] + cost[0][j]

# Construct rest of the tot\_cost array

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

for j in range(1, n+1):

tot\_cost[i][j] = min(tot\_cost[i-1][j-1], tot\_cost[i-1][j], tot\_cost[i][j-1]) + cost[i][j]

return tot\_cost[m][n]

Dynamic Programming formulation without the additional space complexity of tot\_cost

def minCost(cost, row, col):

# For 1st column

for i in range(1, row):

cost[i][0] += cost[i - 1][0]

# For 1st row

for j in range(1, col):

cost[0][j] += cost[0][j - 1]

# For rest of the 2d matrix

for i in range(1, row):

for j in range(1, col):

cost[i][j] += (min(cost[i - 1][j - 1],

min(cost[i - 1][j],

cost[i][j - 1])))

# Returning the value in

# last cell

return cost[row - 1][col - 1]

//TODO: Finish the section on the Min Cost Path problem

## Moving in An Array in an Optimal Way (Denardo)

//TODO: Finish the section on the problem of Moving In An Array in an Optimal Way

## Weighted Job Scheduling

Given are N jobs where each job is represented by following three elements of it.

1 ) Start Time

2 ) Finish Time

3 ) Profit / Associated Value (>= 0)

Find the maximum profit subset of jobs such that no two jobs in the subset overlap

Example:

Input:

Number of jobs n = 4

Job Details {Start Time, Finish Time, Profit}

Job 1: {1, 2, 50}

Job 2: {3, 5, 20}

Job 3: {6, 19, 100}

Job 4: {2, 100, 200}

Output:

Max profit is 250

Max profit can be attained by scheduling jobs 1 and 4.

Recursive solution for the job scheduling problem:

1 ) Sort jobs according to finish time

2 ) Apply the following recursive process

# here `arr` is array of n jobs

def findMaximumProfit(arr, n):

a) if n == 1 return arr[0]

b) return the maximum of the following two profits

(i) maximum profit by excluding current job : findMaximumProfit(arr, n-1)

(ii) maximum profit including the current job

How to find the profit including the current job?

The idea is to find the latest job before the current job in the sorted array that does not conflict with the current job arr[n-1]. Once we find such a job, we recur for all jobs until that job and add profit of current job to result.

Naïve recursion approach:

from functools import cmp\_to\_key

# A job has start time, finish time and profit

class Job:

def \_\_init\_\_(self, start, finish, profit):

self.start = start

self.finish = finish

self.profit = profit

def jobComparator(s1, s2):

return s1.finish < s2.finish

# Find the latest job (in sorted array) that

# doesn't conflict with the job[i]. If there

# is no compatible job, then it returns -1

def latestNonConflict(arr, i):

for j in range(i - 1, -1, -1):

if arr[j].finish <= arr[i - 1].start:

return j

return -1

# A recursive function that returns the

# maximum possible profit from given

# array of jobs. The array of jobs must

# be sorted according to finish time

def findMaxProfitRec(arr, n):

# Base case

if n == 1:

return arr[n - 1].profit

# Find profit when current job is included

inclProf = arr[n - 1].profit

i = latestNonConflict(arr, n)

if i != -1:

inclProf += findMaxProfitRec(arr, i + 1)

# Find profit when current job is excluded

exclProf = findMaxProfitRec(arr, n - 1)

return max(inclProf, exclProf)

# The main function that returns the maximum

# possible profit from given array of jobs

def findMaxProfit(arr, n):

# Sort jobs according to finish time

arr = sorted(arr, key = cmp\_to\_key(jobComparator))

return findMaxProfitRec(arr, n)

Dynamic Programming Implementation

Uses the earlier defined class Job, jobComparator and latestNonConflict functions

def findMaxProfit(arr, n):

# Sort jobs according to finish time

arr = sorted(arr, key=cmp\_to\_key(jobComparator))

# Create an array to store solutions of subproblems.

# table[i] stores the profit for jobs till arr[i]

# (including arr[i])

table = [None] \* n

table[0] = arr[0].profit

# Fill entries in M[] using recursive property

for i in range(1, n):

# Find profit including the current job

inclProf = arr[i].profit

l = latestNonConflict(arr, i)

if l != -1:

inclProf += table[l]

# Store maximum of including and excluding

table[i] = max(inclProf, table[i - 1])

# Store result and free dynamic memory

# allocated for table[]

result = table[n - 1]

return result

//TODO: Finish the section on the Weighted Job Scheduling problem

## Longest Common Sequence

Let us consider strings in the form where are strings or characters. We say that the tuple forms a sequence in if appears on the left of for each pair such that . Given two strings and find the length of the largest common sequence .

#### Naïve Implementation

def longest\_common\_sequence(s1: str, s2: str, m: int, n: int):

“““

Determine the length of the longest common sequence.

:param s1: str

:param s2: str

:param m: int

:param n: int

“““

if m == 0 or n == 0:

return 0

elif s1[m-1] == s2[n-1]:

return 1 + longest\_common\_sequence(s1, s2, m-1, n-1)

else:

return max(longest\_common\_sequence(s1, s2, m, n-1),

longest\_common\_sequence(s1, s2, m-1, n))

**Excerpt 7**: naïve implementation for the longest common sequence

#### Dynamic Programming implementation

Optimal Subproblem formulation

Initially,

//TODO: Finish the section on the LCS problem

## 0/1 Knapsack Problem

Given are items where each item has some weight and profit associated with it. Also, it is given a bag with capacity - that is, the bag can hold at most weight in it. The task is to put such combination of items in the bag so that the profit is maximized. The constraint is that we can put an item in the bag, or we cannot put it at all, it is not possible to put only a fraction of it.

Input: N = 3, W = 4, profit = [1, 3, 4], weight = [4, 5, 1]

Let us denote with

Naïve recursive algorithm:

# A naive recursive implementation

# of 0-1 Knapsack Problem

def knapSack(W, wt, val, n):

# Base Case

if n == 0 or W == 0:

return 0

# If weight of the nth item is

# more than Knapsack of capacity W,

# then this item cannot be included

# in the optimal solution

if (wt[n-1] > W):

return knapSack(W, wt, val, n-1)

# return the maximum of two cases:

# (1) nth item included

# (2) not included

else:

return max(

val[n-1] + knapSack(

W-wt[n-1], wt, val, n-1),

knapSack(W, wt, val, n-1))

1 ) find the subset of elements , each with a weight less than . That is,

2 )

//TODO: Finish the section on the 0/1 Knapsack problem

## Floyd-Warshall Algorithm

//TODO: Finish the section on the Floyd-Warshall algorithm

## Bellman-Ford Algorithm

//TODO: Finish the section on the Bellman-Ford algorithm

## Vertex Cover Problem

//TODO: Finish the section on the Vertex Cover problem

## Travelling Salesman Problem

//TODO: Finish the section on the TSP problem

## Longest Palindromic Subsequence

//TODO: Finish the section on the Longest Palindromic Subsequence problem

## Minimum Insertions to Form a Palindrome

//TODO: Finish the section on the Min Inserts to Form A Palindrome problem

## Matrix Chain Multiplication

//TODO: Finish the section on the Matrix Chain Multiplication

# Problems Solvable By Recursive Backtracking

## N Queens Problem

Place queens on board so that no pair of Queens attack each other.

For example, when N = 4 we have:

.Q..

...Q

Q...

..Q.

//TODO: Finish the section on the N Queens problem

## 2 x 2 Sudoku Solver

//TODO: Finish the section on the 2 x 2 sudoku solver algorithm

# Related Graph Algorithms

## Dijkstra’s Shortest Path Algorithm

//TODO: Finish the section on the Dijkstra’s Shortest Path algorithm

## Breath-First Search Algorithm

//TODO: Finish the section on the Breath-First Search algorithm

## Depth-First Search Algorithm

//TODO: Finish the section on the Depth-First Search algorithm

# References

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