

# CSCE 5150 – Analysis of Computer Algorithms

## Homework No. 3 – Dynamic Programming

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Q1. [2 points] What does dynamic programming have in common with divide-and-conquer? What is the principal difference between them?

1. Dynamic programming: It is a concept that divides a given problem and break it into a reasonable number of subproblems, so that optimal solutions can be used on subproblems to provide optimal solutions on the bigger problem. Ex: LCS, Knapsack, etc.

Similarities with Divide & Conquer:

→ Both techniques divide a larger sub-problem into smaller sub-problems and finally integrated to form a final solution.

Differences between Dynamic Programming & Divide and Conquer:

→ Problem with multiple overlapping of subproblems are solved using dynamic programming. It uses memorization and tabulation to remember and reuse the subproblem results thereby improving the performance.

Q2. [5 points] Consider a modification of the rod-cutting problem in which, in addition to a price  $p_i$  for each rod, each cut incurs a fixed cost of  $c$ . The revenue associated with a solution is now the sum of the prices of the pieces minus the costs of making the cuts. Give a dynamic-programming algorithm to solve this modified problem. (Ex. 15.1-3)

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2. Solution is to modify Bottom-up-cut-rod algorithm.

Modified-cut-rod( $p, n, c$ )

let  $r[0..n]$  be a new array

$r[0] = 0$

for  $i = 1$  to  $n$

$q = p[i]$

    for  $j = 1$  to  $i-1$

$q = \max(q, p[j] + r[i-j] - c)$

$r[i] = q$

return  $r[n]$

We need to account for cost  $c$  of every iteration but at last iteration where  $j=1$ , we make loop run to  $i-1$  instead of  $i$  iterations to make sure that  $c$  is subtracted from the candidate revenue.

Q3. [5 points] We say that a problem exhibits the optimal substructure property when optimal solutions to a problem incorporate optimal solutions to related subproblems, which we may solve independently. Suppose that in the rod-cutting problem, we also had limit  $l_i$  on the number of pieces of length  $i$  that we are allowed to produce, for  $i = 1, 2, \dots, n$ . Show that the optimal-substructure property described no longer holds. (Ex. 15.3-5)

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3. If the subproblem solutions are optimal implies optimal solution to a problem, which we may solve independently. When a limit  $l_i$  is imposed on pieces of size  $i$ , the subproblems cannot be solved independently.

length $i$	1	2	3	4
price $p_i$	15	20	33	36
limit $l$	2	1	1	1

Here, only 3 instances do not violate the limits.

→ length 4 with price 36

→ length 1 & 3 with price 48

→ length 1, 1, 2, 2 with price 50.

\* Here the optimal solution is to cut into 1, 1, 2, 2.

Now, for subproblem of length 2, the solutions are

→ 2 with price 20

→ 1 & 1 with price 30

\* Here the optimal solution is to cut into 1 & 1

But we cannot use this solution for actual problem because it violates limit of two length-rods, as it results in 4 rods of length 1.

Q4. [5 points] Give pseudocode to reconstruct an LCS from the completed  $c$  table and the original sequences  $X = (x_1, x_2, \dots, x_m)$  and  $Y = (y_1, y_2, \dots, y_n)$  in  $O(m+n)$  time, without using the  $b$  table. (Ex. 15.4-2)

(Hint: Try to benefit from the PRINT-LCS procedure)

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We can also use the below pseudo code:

```
def compute_lcs(C,m,n):  
    i = m  
    j = n  
    x = C[i][j]  
    res = "" * (x+1)  
    while i > 0 and j > 0:  
        if X[i-1] == Y[j-1]:  
            res[x-1] = X[i-1]  
            i -= 1  
            j -= 1  
            x -= 1  
        elif C[i-1][j] > C[i][j-1]:  
            i -= 1  
        else:  
            j -= 1  
    return res
```

Q4. [5 points] Give pseudocode to reconstruct an LCS from the completed c table and the original sequences  $X = (x_1, x_2, \dots, x_m)$  and  $Y = (y_1, y_2, \dots, y_n)$  in  $O(m+n)$  time, without using the b table. (Ex. 15.4-2)



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4. LCS pseudocode to print solution in  $O(m+n)$  times

```
print_lcs(a, x, y, x, y)
    if a[x, y] == 0
        return
    if x[x] == y[y]
        print_lcs(a, x, y, x-1, y-1)
    elseif a[x-1, y] > a[x, y-1]
        print_lcs(a, x, y, x-1, y)
    else
        print_lcs(a, x, y, x, y-1)
```

Q5. Consider two teams, A and B, playing a series of games until one of the teams wins  $n$  games. Assume that the probability of A winning a game is the same for each game and equal to  $p$  and the probability of A losing a game is  $q = 1 - p$ . (Hence, there are no ties.) Let  $P(i, j)$  be the probability of A winning the series if A needs  $i$  more games to win the series and B needs  $j$  more games to win the series.

a) Set up a recurrence relation for  $P(i, j)$  that can be used by a dynamic programming algorithm. [2 points]

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5 a) Recurrence relation for  $P(i, j)$   
The probability for A winning a game is  $p$  and probability of A losing game is  $q = 1 - p$ .  
If team A wins, still  $i-1$  more wins are needed by A, where B will still need  $j$  wins. If A loses game, A will still need  $i$  wins while B need  $j-1$  wins.

Recurrence  
$$P(i, j) = pP(i-1, j) + qP(i, j-1) \quad (i, j > 0)$$
  
$$P(0, j) = 1 \quad (j > 0)$$
  
$$P(i, 0) = 0 \quad (i > 0)$$

b) Find the probability of team A winning a seven-game series if the probability of it winning a game is 0.4. [2 points]

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b) Dynamic programming table

Using above recurrence equation

Input	0	1	2	3	4
	0	1	1	1	1
1	0		0.40	0.64	0.78
2	0		0.16	0.35	0.52
3	0		0.06	0.18	0.32
4	0		0.03	0.09	0.18

Hence  $P[4,4] \approx 0.29$

c) Write the pseudocode of the dynamic programming algorithm for solving this problem and analyze its running time. [4 points]

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C.) Algorithm World Series ( $n, p$ )

$q \leftarrow 1 - p$

for  $j \leftarrow 1$  to  $n$  do

$P[0, j] \leftarrow 1.0$

for  $i \leftarrow 1$  to  $n$  do

$P[i, 0] \leftarrow 0.0$

for  $j \leftarrow 1$  to  $n$  do

$P[i, j] \leftarrow p * P[i-1, j] + q * P[i, j-1]$

return  $P[n, n]$

Input : The number of victories  $n$  required to win the series &  $p$  probability of one team winning.

Output : The probability of team winning the series

→ Each entry of table is computed in  $O(1)$  time

→ Hence, time & space efficiency is  $\Theta(n^2)$ .