

# Esercizi su Dynamic Programming

**Exercise 1.1** Write an algorithm to find the maximum value that can be obtained with a full parenthesization of the expression

$$x_1/x_2/x_3/\dots/x_{n-1}/x_n,$$

where  $x_1, x_2, \dots, x_n$  are positive rational numbers and “/” denotes division.

**Exercise 1.2** Give an algorithm that uses the table of additional information  $S[\cdot, \cdot]$  (computed by the Matrix-Chain Multiplication dynamic programming algorithm seen in class) to print the optimal parenthesization for the matrix chain.

**Exercise 1.3** Given the string  $A = \langle a_1, a_2, \dots, a_n \rangle$ , we say that  $A_{i..j} = \langle a_i, a_{i+1}, \dots, a_j \rangle$  is a *palindrome substring* of  $A$  if  $a_{i+h} = a_{j-h}$ , for  $0 \leq h \leq j - i$ . (Intuitively, a palindrome substring is one which is identical to its “mirror” image. For example, if  $A = accaba$ , then both  $A_{1..4} = acca$  and  $A_{4..6} = aba$  are palindrome substrings of  $A$ .)

- (a) Design a dynamic programming algorithm that determines the length of a longest palindrome substring of a string  $A$  in  $O(n^2)$  time and  $O(n^2)$  space.
- (b) Modify your algorithm so that it uses only  $O(n)$  space, while the running time remains unaffected.

**Exercise 1.4** Design and analyze a dynamic programming algorithm which, on input a string  $X$ , determines the minimum number  $p$  of palindrome substrings of  $X$ ,  $Y_1, Y_2, \dots, Y_p$  such that  $X = \langle Y_1, Y_2, \dots, Y_p \rangle$ .

**Exercise 1.5** Design and analyze a dynamic programming algorithm that, given in input a string  $X$ , returns the maximum length of a palindrome *subsequence* of  $X$ . The algorithm must run in time and space  $O(n^2)$ .

**Exercise 1.6** Given a string of *arbitrary* integers  $Z = \langle z_1, z_2, \dots, z_k \rangle$  let  $\text{weight}(Z) = \sum_{i=1}^k z_i$  (note that  $\text{weight}(\epsilon) = 0$ ). Given two integer strings  $X = \langle x_1, x_2, \dots, x_m \rangle$  e  $Y = \langle y_1, y_2, \dots, y_n \rangle$ , design a dynamic programming algorithm to determine a Maximum-Weight Common Subsequence (MWCS)  $Z$  of  $X$  and  $Y$ .

**Exercise 1.7** Design and analyze a dynamic programming algorithm which, on input two nonnegative integers  $n$  and  $k$ , with  $n > 0$  and  $0 \leq k \leq n$ , outputs  $\binom{n}{k}$  by performing  $\Theta(nk)$  sums. (*Hint*: Prove that for  $0 < k < n$ ,  $\binom{n}{k} = \binom{n-1}{k} + \binom{n-1}{k-1}$ .)

**Exercise 1.8** Given two strings  $X$  and  $Y$ , a third string  $Z$  is a *common superstring* of  $X$  and  $Y$ , if  $X$  and  $Y$  are both subsequences of  $Z$ . (*Example*: if  $X = \text{sos}$  and  $Y = \text{soia}$ , then  $Z = \text{sosia}$  is a common superstring of  $X$  and  $Y$ .) Design and analyze a dynamic programming algorithm which, given as input two strings  $X$  and  $Y$ , returns the length of the *Shortest Common Superstring* (SCS) of  $X$  and  $Y$  and additional information needed to print the SCS. The algorithm must run in time  $\Theta(|X||Y|)$ . (*Hint*: Use an approach similar to the one used to compute the LCS of two strings.)

**Exercise 1.9** Given two strings of integers  $Z^1$  and  $Z^2$ , with  $|Z^1| = |Z^2| = k$ , we define their *discrepancy* as  $d(Z^1, Z^2) = \sum_{i=1}^k |Z_i^1 - Z_i^2|$ . Design and analyze a dynamic programming algorithm which, on input two (arbitrary) strings of integers  $X$  e  $Y$ , computes the maximum discrepancy obtainable by a subsequence of  $X$  and a subsequence of  $Y$  of equal length by performing  $\Theta(|X||Y|)$  comparisons and sums between integers.

**Exercise 1.10** Let  $n > 0$ . Given a string of  $n$  integers  $A = \langle a_1, a_2, \dots, a_n \rangle$ , consider the following recurrence, defined for all pairs  $(i, j)$ , with  $1 \leq i \leq j \leq n$ :

$$B(i, j) = \begin{cases} a_i & 1 \leq i = j \leq n, \\ \max\{B(i, k) \cdot B(k+1, j) : i \leq k \leq j-1\} & 1 \leq i < j \leq n. \end{cases}$$

Design and analyze an iterative bottom-up algorithm that, on input  $A$ , returns  $B(1, n)$  by performing  $O(n^3)$  sums.

**Exercise 1.11** Let  $n > 0$ . Assume that a given dynamic programming strategy leads to the following recurrence, defined for all values of  $i$  and  $j$  with  $1 \leq i \leq j \leq n$ :

$$C(i, j) = \begin{cases} 1 & (i = 1) \text{ and } (j = n), \\ \sum_{r=1}^{i-1} C(r, j) + \sum_{s=j+1}^n C(i, s) & \text{altrimenti.} \end{cases}$$

Design and analyze an iterative bottom-up algorithm that computes all values  $C(i, j)$ ,  $1 \leq i \leq j \leq n$ .

**Exercise 1.12** Given the following bottom-up code:

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DP_SUM( $n$ )
  for  $i \leftarrow 1$  to  $n$  do  $A[i, i] \leftarrow i$ 
  for  $\ell \leftarrow 2$  to  $n$  do
    for  $i \leftarrow 1$  to  $n - \ell + 1$  do
       $j \leftarrow i + \ell - 1$ 
       $A[i, j] \leftarrow A[i, j - 1] + A[i + 1, j]$ 
  return  $A[1, n]$ 

```

write an equivalent memoized code and analyze its running time in terms of sums between integers.

**Exercise 1.13** Given a string  $X = \langle x_1, x_2, \dots, x_n \rangle$ , consider the following recurrence  $\ell(i, j)$ , defined for  $1 \leq i \leq j \leq n$ :

$$\ell(i, j) = \begin{cases} 1 & i = j, \\ 2 & i = j - 1 \\ 2 + \ell(i + 1, j - 1) & (i < j - 1) \wedge (x_i = x_j) \\ \sum_{k=i}^{j-1} (\ell(i, k) + \ell(k + 1, j)) & (i < j - 1) \wedge (x_i \neq x_j). \end{cases}$$

Design memoized code to return the value  $\ell(1, n)$  and analyze the code both in the worst case and in the best case, assuming that the only unit-cost operations are character comparisons.