

Algorithms and Data Structures 2015/16

Week 6 tutorial sheet (Tues 23rd - Fri 26th February)

All questions from Q2 onwards depend on the “Dynamic Programming” material from Monday 22nd February. You may want to read the notes for this topic in advance of the 22nd, in order to prepare for the tutorial.

1. Show how to sort n integers in the range $\{1, \dots, n^2\}$ in $O(n)$ time.
2. Find an optimal parenthesization of a matrix-chain product whose sequence of dimensions is $\langle 5, 10, 12, 5, 50, 6 \rangle$.

Simplified version of *Ex. 15.2-1 of [CLRS] (2nd and 3rd eds)*

3. Consider the problem of taking a set of n items with sizes s_1, \dots, s_n , and values v_1, \dots, v_n respectively. We assume $s_i, v_i \in \mathbb{N}$ for all $1 \leq i \leq n$. Suppose we are also given a “knapsack capacity” $C \in \mathbb{N}$. The *knapsack problem* is the problem of finding a subset $S \subseteq \{1, \dots, n\}$ such that $\sum_{i \in S} s_i \leq C$ and such that $\sum_{i \in S} v_i$ is maximized subject to the first constraint.

We write $kp_{n,C}$ to denote the value $\sum_{i \in S} v_i$ of the maximum-value knapsack on the set of all items. For any $k \leq n$, and any $\hat{C} \leq C, \hat{C} \in \mathbb{N}$, we can consider the same problem on the first k items in regard to capacity \hat{C} . We denote the maximum-value knapsack for such a subproblem by $kp_{k,\hat{C}}$.

- (a) Prove that the following recurrence holds:

$$kp_{k,\hat{C}} = \begin{cases} 0 & \text{if } k = 0 \\ kp_{k-1,\hat{C}} & \text{if } k > 0 \text{ but } s_k > \hat{C} \\ \max\{kp_{k-1,\hat{C}}, kp_{k-1,\hat{C}-s_k} + v_k\} & \text{otherwise.} \end{cases}$$

- (b) Use the recurrence in (a) to develop a $\Theta(n \cdot C)$ dynamic programming algorithm to compute the optimal knapsack wrt the original n items and capacity C .
4. On slide 9 of Lectures 10.11 we claim that the number of possible parenthesizations of a Matrix-chain sequence containing n matrices is $\Omega(3^n)$.

Set up a recurrence for the number of parenthesizations.

Prove the *Induction step* wrt $c \cdot 3^n$ (you do not need to prove the base cases - there are too many of these, as n_0 is high. This is why the proof is mentioned as “ugly” in the slides).

5. *Longest Common Subsequence* A subsequence of a given sequence is just the given sequence with some elements (possibly none) left out. Given a sequence $s = s_1 s_2 \dots s_n$, we say another sequence $r = r_1 \dots r_k$ is a subsequence of s if there is a strictly increasing sequence i_1, i_2, \dots, i_k of indices such that for all $j = 1 \dots k$ we have $r_j = s_{i_j}$.

Given two sequences x and y we say that a sequence r is a *common subsequence* if r is a subsequence of both x and y . In the *longest common subsequence* problem, we are given two sequences $x = x_1 \dots x_n$ and $y = y_1 \dots y_m$ and wish to find a maximum-length common subsequence of x and y .

Give a $O(mn)$ -time dynamic programming algorithm to solve the longest common subsequence problem.

Mary Cryan