

# Lab 08

I. El-Shaarawy & M. Touny

December 9, 2023

## Contents

<b>Exercises</b>	<b>2</b>
8.1.3 . . . . .	2
8.1.5 . . . . .	2
8.1.6 . . . . .	3
8.2.2 . . . . .	3
8.2.3 . . . . .	3
8.2.5 . . . . .	4
8.3.4 . . . . .	4
8.3.8 . . . . .	5

## Exercises

### 8.1.3

Homework.

### 8.1.5

Homework.

### 8.1.6

Hints

- Explain why is the formulation  $F(n) = F(n-1) + p_1$  is wrong. Derive a counter example.
- The optimal solution may be  $F(n) = p_n$ . Modify it so that it is in terms of  $F(k)$  for some  $k < n$ .
- Generalize.

**Solution**

Recursive formulation.

$$F(0) = 0$$
$$F(n) = \max_{1 \leq j \leq n} \{p_j + F(n - j)\}$$

Algorithm.

```
# input: Length n, and values of pieces of length i, P[i]
# output: Maximum value of all possible cuts on a rod of length n
def dynamicRodCut(n, P[0..n])

    # a rod of length zero contributes nothing to revenue
    P[0] = 0

    # Initialize an array of size n
    F = [] * n

    # Set the base case
    F[0] = 0

    # Compute bottom-up F[i]
    for i in 1..n

        maxVal = 0

        # Compute the maximum among all js
```

```

    for j in 0..i
      # call memoized subinstances
      # update if found a greater value
      maxVal = max( maxVal, P[j] + F(n-j) )

    # memoize
    F[i] = maxVal

    # return max value of cuts, on given length n
    return F[n]

```

Complexity. Time is  $1 + \dots + n = n(n + 1)/2$ . Additional space is  $n + 1$ .

### 8.2.2

**Homework.**

### 8.2.3

**Homework.**

### 8.2.5

**Hints**

- Recall for the algorithm given in the book, at each step, either we take or leave the  $i$ th item.
- For our case what if we at each step, either leave all items, or take 1st item, or take 2nd item, ..., or  $n$ th item. Modify the formulation.

**Solution**

Recursive Formulation.

$$\begin{aligned}
 F(W) &= 0 && \text{if } W < w_j, 1 \leq j \leq n \\
 F(W) &= \max_{j: W \geq w_j} F(W - w_j) + v_j && \text{otherwise}
 \end{aligned}$$

Algorithm

```

# memory function
# input: i indicating selecting a multiset of size at most i, from all n items
        j capacity
# output: optimal value of feasible multiset of size at most i
def MFKnapsack(i, j, weight, value, F)

    # only if not memoized, compute and cache it
    if F[i,j] < 0

```

```

    # find the maximum value among all cases

    # case, no additional item is taken
    # this value sustains only if there is no capacity for any item
    # recall values are positive integers
    maxVal = MFKnapsack(i-1, j)

    # for each item out of total n items,
    # if there is a capacity for it,
    # compute total value, and update if greater.
    for i in 1..n
        maxVal = max( maxVal, value[i] + MFKnapsack(i-1, j-weight[i]) )

    # memoize
    F[i,j] = maxVal

return F[i,j]

# input: weight of ith item, value of ith item, total capacity W
# output: max value of a multiset of size at most n,
#         from all n items, constrained by capacity W
def dynamicKnapsack(weight[1..n], value[1..n], W)

    # memoization table
    # all cells -1, indicating no value is computed
    F[0..n, 0..W] = -1
    # except row 0 and column 0, values are 0, by definition of base case
    for i in 0..n, F[i,0] = 0
    for i in 0..W, F[0,i] = 0

    # compute memoization table F, and read F[n, W]
    sol = MFKnapsack(n, W, weight, value, F)

    # problem solution is F[n, W]
    return sol

```

### 8.3.4

**Homework.**

### 8.3.8

**Hints.**

- Recall you have table  $R$ , where  $R[i, j]$  contains the root of the tree of nodes  $i, \dots, j$ .
- Recall how the optimal solution of knapsack was constructed.

## Solution.

```
# global variables: table R of roots indices
                    keys A
# input: root node of a tree, and indices i and j of keys covered
# output: None. Tree is modified so the root points to its children
# initialize with i = 1 and j = n
# def optimalBST(root, i, j)

    # base case
    if root = NULL
        return

    # index of the root of subtree of keys A_i, ..., A_j
    k = R[i,j]

    # left child
    root.left = A[ R[i, k-1] ]
    # right child
    root.right = A[ R[k+1, j] ]

    # Recursively, Call the child
    optimalBST(root.left, i, k-1)
    optimalBST(root.right, k+1, j)
```

P.S. Anything by Donald Knuth is worthwhile studying, however for our pragmatic purposes we omit the analysis bounding  $\mathcal{O}(n^2)$ .