

Problem 1: (3 points) Rod Cutting: (from the text CLRS) 15.1-2 – Many possible solutions

Here is a counterexample for the “greedy” strategy:

length i	1	2	3	4
price p_i	1	20	33	36
p_i/i	1	10	11	1

Let the given rod length be 4. According to a greedy strategy, we first cut out a rod of length 3 for a price of 33, which leaves us with a rod of length 1 of price 1. The total price for the rod is 34. The optimal way is to cut it into two rods of length 2 each fetching us 40 dollars.

Problem 2: (3 points) Modified Rod Cutting: (from the text CLRS) 15.1-3

MODIFIED-CUT-ROD(p, n, c)

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

$r[0] = 0$

for $j = 1$ **to** n

$q = p[j]$

for $i = 1$ **to** $j - 1$

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

$r[j] = q$

return $r[n]$

The major modification required is in the body of the inner **for** loop, which now reads $q = \max(q, p[i] + r[j - i] - c)$. This change reflects the fixed cost of making the cut, which is deducted from the revenue. We also have to handle the case in which we make no cuts (when i equals j); the total revenue in this case is simply $p[j]$. Thus, we modify the inner **for** loop to run from i to $j - 1$ instead of to j . The assignment $q = p[j]$ takes care of the case of no cuts. If we did not make these modifications, then even in the case of no cuts, we would be deducting c from the total revenue.

Problem 3: (6 points) Making Change: Given coins of denominations (value) $1 = v_1 < v_2 < \dots < v_n$, we wish to make change for an amount A using as few coins as possible. Assume that v_i 's and A are integers. Since $v_1 = 1$ there will always be a solution.

- a) Describe **(1 pt)** and give pseudocode for a dynamic programming algorithm to find the minimum number of coins to make change for A .

Sample pseudocode example **(3 pts)**

$\text{coins}[k]$ is the number of coins used to make change for k cents
 $\text{sol}[k]$ is the index of last denomination $V[\text{sol}[k]]$ used to obtain change for k cents

Formula

```
coins[i] = inf  if i < 0
coins[0] = 0
coins[1] = 1
coins[j] = min_{1 ≤ i ≤ n} {1 + coins[j - v_i]}, sol[j] = i
```

```
minCoins(A, V[], n)
    coins[0]=0; coins[1] = 1
    for j = 2 to A do {
        min = inf
        for i = 1 to n do {
            if ( j >= V[i] )
                if (coins[j-V[i]] < min )
                {
                    min = coins[j-V[i]]
                    index = i
                }
        }
        coins[j] = min + 1
        sol[j] = index
    }
    return coins[A], sol[A]
```

To reconstruct the series of coins used to obtain change for A with minimum number of coins.
 Call $\text{MakeChange}(\text{sol}, V, A)$.

```
MakeChange(sol[], V[], C[], m)
    While m > 0 {
        C[sol[m]] = C[sol[m]] + 1
        Print V[sol[m]]
        m = m - V[sol[m]]
    }
```

b) What is the theoretical running time of your algorithm? (2 pts)

The running time of MakeChange is $O(A)$

The running time of minCoins is $\Theta(nA)$ since the outer loop is $j = 2 \dots A$ and inner loop is $i = 1 \dots n$.

Overall $\Theta(nA)$

Problem 4:

a) The Shopping Spree problem is modified version of the Knapsack problem. First we find the family member who can carry the most weight and call that weight M_{\max} . We then use the DP algorithm to solve the knapsack problem with n items of price P_i , weight W_i and a knapsack with capacity of M_{\max} . The running time of the DP knapsack algorithm is $\Theta(n M_{\max})$. The maximum total price of goods that family member k can carry is in cell (n, M_k) where M_k is the maximum weight that member k can carry. The maximum value for each family member can be obtained from the table in constant time. The time to backtrack in the DP table to determine the items a single family member takes is $O(n)$ and for all family member this takes $O(nF)$.

b) The overall running time is $\Theta(n M_{\max}) + O(nF)$ or $O(n (M_{\max} + F))$ or $\Theta(nM_{\max})$ if $M_{\max} \gg F$