



Answer any 6 out of the following 8 questions ( $6 \times 15 = 90$ ).

1. (a) Perform **worst case running-time analysis** of the following algorithms and express in  $\mathcal{O}$  notation. [6 + 4]

```
function1(m, n):  
  for(i=1; i <= m; ++i):  
    for(j=1; j <= n; j*=3):  
      if (i%2 == 0):  
        print i  
  for(i=1; i<=n; i*=4):  
    for(j=1; j <= m; ++j):  
      if(j%2 == 0):  
        print j
```

```
function2(n):  
  for(i=1; i<=n; ++i):  
    for(j=1; j<=i; ++j):  
      print i+j
```

- (b) What do you understand by **constant-time costs**? Provide examples of a code fragment with constant-time cost, and another not having a constant-time cost. [2+3]
2. Using the **recursion tree** method, determine a good asymptotic upper bound on the following recurrence:  $T(n) = 2T(\frac{n}{3}) + \mathcal{O}(n)$ . Show details of your calculation. [15]
3. (a) A Dynamic Programming algorithm for the classical **Coin Change problem** is provided at **figure 1**, where given an amount  $n$  and  $m$  types of coins  $C_1, C_2, \dots, C_m$ , you have to minimize the number of coins to build up the amount  $n$ . Now, suppose that, each coin type  $C_i$  has a weight  $W_i$ . Modify the algorithm such that now you have to **minimize the total weight of coins** to build the amount  $n$ . [9]
- (b) Suppose that you have got a set of the following activities, with the given start and end times for the **Activity Selection** problem:

$\{[2, 4], [1, 5], [9, 12], [6, 10], [1, 16], [11, 15], [7, 8]\}$

What will be the solution sets returned by two **greedy algorithms**: one using the end times as the choosing criteria, another using activity lengths? Which one is optimal here? [2.5 + 2.5 + 1]

4. (a) Provide a **Divide-and-Conquer** algorithm to find the count of non-zero elements in an input array. What would be the costs of each of the three steps of **divide-and-conquer**? [6+3]
- (b) What is the basic difference between the algorithmic paradigms - **greedy algorithms** and **dynamic programming**? [3]
- (c) For the classical **Rod cutting** problem, design an example case where a greedy algorithm fails to find an optimal solution. What are the optimal and the greedy solutions for your example? [3]
5. (a) Suppose that you are the owner of an Internet service provider company. The maximum bandwidth capacity that you can supply is *max\_capacity*. There are  $n$  customers who are willing to acquire your Internet service. Each customer has two attributes: the bandwidth  $B_i$  that he/she wants, and the payment  $P_i$  that he/she will pay for that. Provide a **Dynamic Programming** algorithm to find out the maximum total profit that you can make by providing bandwidths within your *max\_capacity*. [12]

- (b) Design a min-heap with height 3 and the minimum possible number of elements. All the elements must be negative here. [3]
6. (a) A dynamic programming (DP) algorithm to the classical **Rod-Cutting problem** is provided at **figure 2**. Simulate the steps of tabulation of the **DP memoization table**, along with the **recursion tree** generated by the DP algorithm, for initial rod length 6 and the profit table  $P = \{1, 2, 3, 3, 4, 5\}$ . [9]
- (b) Propose an algorithm to convert a provided **min-heap** to a **max-heap**. [6]

```

let M be the memoization table, initially filled with NULL
COIN-CHANGE(n)
    if n = 0
        return 0
    if M[n] ≠ NULL
        return M[n]
    bestSoln = infinity
    for i = 1 to m
        if n ≤ C[i]
            soln = 1 + COIN-CHANGE(n - C[i])
            if bestSoln > soln
                bestSoln = soln
    M[n] = bestSoln
    return M[n]

```

Figure 1: **COIN-CHANGE**

```

let M be the memoization table, initially filled with NULL
CUT-ROD(length)
    if length = 0
        return 0
    if M[length] ≠ NULL
        return M[length]
    bestSoln = P[length]
    for i = 1 to length - 1
        soln = P[i] + CUT-ROD(length - i)
        if bestSoln < soln
            bestSoln = soln
    M[length] = bestSoln
    return M[length]

```

Figure 2: **CUT-ROD**

7. (a) Provide a recursive equation for the running-time  $T(n)$  for the **Divide-and-Conquer** algorithm provided at **figure 3**, in terms of the input array ( $A$ ) size  $n$ , where  $n = r - l + 1$ . You must mention the costs of each of the step of this divide-and-conquer algorithm. [6]
- (b) Design examples with  $|A| = n = 6$  items of both a **worst-case** and a **best-case** for the algorithm provided at **figure 4**. Also mention the runtime complexities for both cases in  $\mathcal{O}$  notation. [3 + 3 + 3]
8. (a) What is the depth of the **Merge-Sort** algorithm's recursion tree, when sorting an  $n$ -element array in *non-increasing* order? What is the space complexity of **Merge-Sort**? Why? [2.5 + 2.5]
- (b) At a **Min-Priority-Queue** of size  $n$  and unique elements, what are the possible indices of the maximum element? [3]
- (c) Suppose that you have a **Max-Heap** with  $n$  elements. Is it guaranteed to remain a max-heap if you perform the following changes on it? Explain briefly. [2 + 2 + 3]
- (i) increase the value of the root; (ii) decrease the value of a leaf; (iii) decrease the value at index 2, and  $n \geq 5$ .

```

DUMMY(A, l, r)
1: if l = r then
2:   return A[l]
3: mid = ⌊ (low + high) / 2 ⌋
4: r1 = DUMMY(low, mid)
5: r2 = DUMMY(mid + 1, r)
6: sum = 0
7: for i = l to r do
8:   sum = sum + i
9: return sum - (r1 + r2)

```

Figure 3: **DUMMY**

```

Search(A, l, r, key):
if l <= r:
    m = (l+r)/2
    if A[m] == key:
        return m
    else if A[m] > key:
        return Search(A, l, m-1, key)
    else:
        return Search(A, m+1, r, key)
return -1

```

Figure 4: **SEARCH**