



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

- Derive the **exact-cost equation** for the running-time of the algorithm at **Fig. 1**, and express it in the big-Oh ( $\mathcal{O}$ ) notation. [6]
  - Demonstrate the full working process of the **Divide-and-Conquer** algorithm for the **Find Minimum Subarray** problem, on the following array:  $[-2, 1, 0, -5, 3, -4, 7, -5]$ . [9]

```

Algorithm1(n, m):
  for(i = 1; i <= n; i = i + 1)
    print(i)

  for(j = 1; j <= m; j = j + 1)
    for(i = 1; i <= n; i = i + 2)
      print(i * j)
  
```

Figure 1: **Q. 1a**

```

Algorithm2(n):
  sum = 0

  for(i = 1; i <= n/2; i = i + 1)
    for(j = 1; j <= 100; j = j + 1)
      sum = sum + j * i

  return sum
  
```

Figure 2: **Q. 4c**

- Suppose you are developing a greedy algorithm for the **Rod Cutting Problem** where each cut is **greedily** made using the **maximum price**. Now design a suitable price array for rod length 6 on which the algorithm will **not** provide an optimal solution. What is the optimal solution for your example? [3]
  - You are going to build a new computer system that requires  $t$  watts of power to run. But there might not be power chips available at the market with exactly  $t$  watts. There are  $n$  types of power chips available, with the following capacities:  $P_1, P_2, \dots, P_n$ . You want to assemble the  $t$  watts using the **minimum number** of power chips. Propose a **Dynamic Programming** algorithm to solve this problem. [12]
- Design an algorithm that provided two parameters  $n$  and  $m$ , runs in time  $\mathcal{O}(n + m \times \log m)$ . [7]
  - Add **memoization** to the algorithm in **Fig. 3** to get its **Dynamic Programming** version. Then calculate  $CATALAN(3)$  using the memoized algorithm. You must demonstrate the recursion-tree generated. [3+5]

```

CATALAN(n):
  if n = 0
    return 1

  c = 0
  for i = 0 to n - 1
    c = c + [CATALAN(i) * CATALAN(n - i - 1)]

  return c
  
```

Figure 3: **Q. 3b**

$$T(n) = \begin{cases} \mathcal{O}(1), & \text{if } n = 1; \\ 3T(\frac{n}{6}) + \mathcal{O}(n), & \text{otherwise.} \end{cases}$$

Figure 4: **Q. 7a**

- Trump coins* are used by the people of the *Trump land* for everyday transactions. Only the following coins are available at this system: 1, 7, 12, 25. Provide an example of an amount where the **greedy** strategy for the **Coin-Change** problem does not provide an optimal solution. You must mention the optimal solution, and the greedy solution. [3]
  - Suppose that you run a server lending business, where you have got the the following server-use requests for the next month. Each request is at the format:  $[start\ date, end\ date]$ . A server can be used by **at most one** user at a time. Using a **greedy algorithm**, find out the **minimum** number of servers required to satisfy all the requests.  $\{[6, 10], [3, 5], [3, 8], [4, 9], [1, 7], [1, 2], [3, 7], [9, 11]\}$  [6]

- (c) Derive the **exact-cost equation** for the running-time of the algorithm at **Fig. 2**, and express it in the big-Oh ( $\mathcal{O}$ ) notation. [6]
5. (a) Consider this modified version of the **Merge sort** algorithm as follows: divide the provided array of size  $n$  into **three** subarrays of sizes roughly  $\frac{n}{3}$ , sort each of these three subarrays recursively, and then combine the three sorted subarrays in time  $\mathcal{O}(n)$ . Answer the following:
- Design a **recurrence relation** for the running-time  $T(n)$  of this algorithm. [3]
  - Using the **recursion-tree method**, solve the recurrence and derive its running-time in the big-Oh ( $\mathcal{O}$ ) notation. [7]
- (b) Given the arrival and the departure times of 6 trains for a railway platform, find out the **maximum** number of trains that can use that platform without any collision, using a **greedy algorithm**. There **must exist at least 10 minutes of safety break** between the departure of one train and arrival of a next one.  $\{[1000, 1030], [840, 1030], [850, 1040], [1700, 2000], [800, 835], 1300, 1800\}$  [5]
6. (a) Take a look at the algorithm at **Fig. 5**. Now provide **both the best-case and worst-case examples** of the arrays  $A$  and  $B$  for  $|A| = n = 4$  and  $|B| = m = 5$ , and  $val = 10$ . Also derive the **running-time complexities** for both the cases in the big-Oh ( $\mathcal{O}$ ) notation.  $[3 + 3 + 5]$
- (b) For the **Divide-and-Conquer** algorithm for the **Find Maximum Subarray** problem, if the cost for finding the *maximum crossing sum* is  $\mathcal{O}(n^2)$ , then what will be the **recursive equation** for the running-time of the algorithm? [4]

Algorithm3(A, B, val):

```

n = A.length
m = B.length

for(i = 1; i <= n; i = i + 1)
    print(A[i])

for(j = 1; j <= n; j = 2 * j)
    for(k = 1; k <= m; k = 2 * k)
        if(A[j] * B[k] == val)
            return j

return -1

```

Figure 5: Q. 6a

```

n: total number of items
V: array of values for the items
W: array of weights for the items
M: DP memoization table,
    initially filled with -1

KNAPSACK(index, maxWeight):
    if i > n
        return 0
    if M[index][maxWeight] != -1
        return M[index][maxWeight]
    if W[index] > maxWeight
        val = KNAPSACK(index + 1, maxWeight)
    else
        val1 = KNAPSACK(index + 1, maxWeight)
        val2 = V[index] + KNAPSACK(index + 1, maxWeight - W[index])
        val = max(val1, val2)
    M[index][maxWeight] = val
    return val

```

Figure 6: Q. 8a

7. (a) Using the **recursion-tree method**, find out an asymptotic upper bound in the big-Oh ( $\mathcal{O}$ ) for the recurrence in **Fig. 4**. [12]
- (b) What is the fundamental difference between **Greedy Strategies** and **Dynamic Programming**? [3]
8. (a) A **Dynamic Programming** algorithm for the classical **0/1 Knapsack problem** is provided in **Fig. 6**. Consider the following added restrictions to the problem: if you do not take the  $i^{th}$  item, you have to pay a penalty  $P_i$ , and if you take it, you have to pay a transportation fee  $T_i$ . Update the algorithm such that it can solve this modified problem. [7]
- (b) Provide a **Divide-and-Conquer** algorithm to find the count of negative ( $< 0$ ) elements in an input array  $A$ . Also, mention the running-times of each of the **three steps** of the divide-and-conquer strategy in your algorithm.  $[5 + 3]$