

DELHI SKILL AND ENTREPRENEURSHIP UNIVERSITY

B.Tech Semester-V (CSE)
End-Semester Examination (Set 3)

Time: 3 Hours

Instructions to Candidates:

1. This paper consists of two sections: Section A and Section B.
 2. Section A is compulsory. Attempt all questions.
 3. Section B contains descriptive questions.
 4. Assume necessary data if not given.
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SECTION A – Short Answer Questions (25 Marks)

Attempt all questions. Answers must be concise.

Q1. Define 'Little-oh' notation (o). How does it differ from Big-Oh (O)?

Answer: $f(n) = o(g(n))$ means $f(n)$ grows strictly slower than $g(n)$. Formally, for **any** positive constant $c > 0$, there exists a constant $n_0 > 0$ such that $0 \leq f(n) < c \cdot g(n)$ for all $n \geq n_0$. Unlike O , the bound is not tight; the limit $\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = 0$.

Q2. What is the theoretical lower bound for any comparison-based sorting algorithm?

Answer: Any comparison-based sorting algorithm must perform $\Omega(n \log n)$ comparisons in the worst case. This is derived from the decision tree model, where a tree with $n!$ leaves (permutations) must have a height of at least $\log(n!)$.

Q3. Define the 'Selection Problem' (finding the i^{th} order statistic).

Answer: The Selection Problem involves finding the i^{th} smallest element in a set of n unsorted elements.

- $i = 1$: Minimum.
- $i = n$: Maximum.
- $i = \lceil n/2 \rceil$: Median.

Q4. What is the 'Hamiltonian Cycle' problem? Is it in P or NP?

Answer: A Hamiltonian Cycle is a cycle in a graph that visits every vertex exactly once and returns to the starting vertex.

- It is an ****NP-Complete**** problem. There is no known polynomial-time algorithm to solve it for general graphs, but a solution can be verified in polynomial time.

Q5. Why might a Greedy Algorithm fail for the 'Coin Change Problem' with arbitrary denominations?

Answer: Greedy works for standard currencies (like 1, 5, 10), but fails for arbitrary sets because it lacks lookahead.

- Example: Coins {1, 3, 4}, Target 6.
- Greedy: Pick 4, then 1, then 1 (Total 3 coins).
- Optimal: Pick 3, then 3 (Total 2 coins).

Q6. Explain the transition function used in 'String Matching with Finite Automata'.

Answer: The transition function $\delta(q, a)$ determines the next state given current state q and input character a .

$$\delta(q, a) = \sigma(P_q a)$$

It maps to the length of the longest prefix of pattern P that is a suffix of $P_q a$ (the pattern prefix of length q concatenated with character a).

Q7. What is 'Memorization' (Memoization) in Dynamic Programming?

Answer: Memoization is a top-down optimization technique where we cache the results of expensive function calls (sub-problems) and return the cached result when the same inputs occur again, preventing redundant computations in recursive algorithms.

Q8. State the time complexity of 'Job Sequencing with Deadlines' if we use a Disjoint Set data structure.

Answer:

- Sorting jobs by profit: $O(n \log n)$.
- Using Disjoint Sets (Union-Find) to find the available slot nearest to the deadline: $O(n \cdot \alpha(n))$, where α is the inverse Ackermann function (nearly constant).
- Total: $O(n \log n)$ (dominated by sorting).

Q9. What is a 'Clique' in an undirected graph?

Answer: A **Clique** is a subset of vertices in an undirected graph such that every two distinct vertices in the clique are adjacent (complete subgraph). The 'Maximum Clique Problem' (finding the largest clique) is NP-Hard.

Q10. Identify the algorithmic approach used for the '15-Puzzle Problem'.

Answer: The **Branch and Bound** technique (specifically Least Cost Search) is typically used. It uses a heuristic cost function (like Manhattan distance of tiles from target positions) to explore the most promising moves first.

SECTION B – Descriptive Questions (75 Marks)

Detailed answers required. Draw diagrams wherever necessary.

Q11. Explain the **Radix Sort** algorithm. How does it sort integers without direct comparison? Analyze its time complexity.

Answer: Working Principle: Radix Sort avoids comparison by creating and distributing elements into buckets according to their radix (base). It typically uses a stable sort (like Counting Sort) as a subroutine to sort digits from the Least Significant Digit (LSD) to the Most Significant Digit (MSD).

Steps (LSD method): 1. Find the maximum number to know the number of digits, d . 2. Do the following for each digit i from 1 to d :

- Sort the array elements using a stable sorting algorithm (Counting Sort) according to the i^{th} digit.

Complexity: Let n be the number of elements, b be the base (usually 10), and d be the max number of digits.

- Counting sort takes $O(n + b)$.
- We repeat this d times.
- Total Time: $O(d(n + b))$. Since b and d are often small constants, it runs in linear time $O(n)$.

Q12. Using the **Recursion Tree Method**, solve the recurrence: $T(n) = 3T(n/4) + n^2$.

Answer: We expand the tree for $T(n) = 3T(n/4) + n^2$.

- ****Level 0:**** Cost n^2 . (1 node).
- ****Level 1:**** 3 children, each size $n/4$. Cost $3 \times (n/4)^2 = 3/16n^2$.
- ****Level 2:**** 9 children, each size $n/16$. Cost $9 \times (n/16)^2 = 9/256n^2 = (3/16)^2n^2$.

Summing costs:

$$Total = n^2 + \frac{3}{16}n^2 + \left(\frac{3}{16}\right)^2 n^2 + \dots$$

This is a geometric series with ratio $r = 3/16 < 1$. The sum converges to $\frac{n^2}{1-r} = \frac{n^2}{1-3/16} = \frac{16}{13}n^2$.

Conclusion: $T(n) = \Theta(n^2)$.

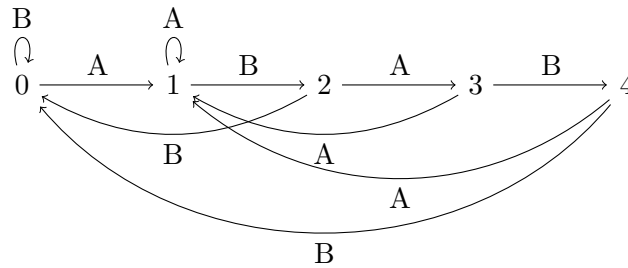
Q13. Discuss the **Finite Automata** approach for String Matching. Draw the transition diagram for the pattern $P = ABAB$ over alphabet $\Sigma = \{A, B\}$.

Answer: Concept: We build a Deterministic Finite Automaton (DFA) that accepts strings ending with Pattern P . We process the text T through the DFA. If we reach the final state, a match is found. Time complexity is $O(n)$ for matching.

Pattern $P = ABAB$. States: 0, 1, 2, 3, 4 (Final).

- State 0: Start. Input A \rightarrow 1. Input B \rightarrow 0.
- State 1 (A): Input B \rightarrow 2. Input A \rightarrow 1.
- State 2 (AB): Input A \rightarrow 3. Input B \rightarrow 0.
- State 3 (ABA): Input B \rightarrow 4. Input A \rightarrow 1.

- State 4 (ABAB): Match found. Next $A \rightarrow 1$, Next $B \rightarrow 0$ (overlaps).



Q14. Explain the **N-Queens Problem**. How do we check if placing a queen at position (r, c) is valid in $O(1)$ time given previous placements?

Answer: Problem: Place N queens on an $N \times N$ chessboard such that no two queens attack each other (no shared row, column, or diagonal).

Validation Strategy: We place queens row by row. When placing at (r, c) : 1. ****Column:**** Check if 'col[c]' is occupied. 2. ****Main Diagonal:**** Constant $(r - c)$. Check 'diag1[r-c + offset]' is occupied. 3. ****Anti-Diagonal:**** Constant $(r + c)$. Check 'diag2[r+c]' is occupied. By maintaining boolean arrays for columns and diagonals, validity is checked in $O(1)$ instead of iterating $O(N)$.

Q15. Explain the **Job Sequencing with Deadlines** problem. Solve for the given jobs (Profit, Deadline): $J_1(20, 2)$, $J_2(15, 2)$, $J_3(10, 1)$, $J_4(5, 3)$, $J_5(1, 3)$.

Answer: Goal: Maximize profit by scheduling jobs within their deadlines. Each job takes 1 unit of time.

Greedy Strategy: Sort jobs by Profit (Descending). 1. J_1 (20, D=2): Assign to slot [1-2]. (Status: 1-2 Filled). 2. J_2 (15, D=2): Slot [1-2] full. Check [0-1]. Empty. Assign [0-1]. (Status: 0-1, 1-2 Filled). 3. J_3 (10, D=1): Deadline 1. Slot [0-1] full. Reject. 4. J_4 (5, D=3): Deadline 3. Slot [2-3] empty. Assign [2-3]. 5. J_5 (1, D=3): Slot [2-3] full. Earlier slots full. Reject.

Selected Jobs: J_2, J_1, J_4 . **Total Profit:** $15 + 20 + 5 = 40$.

Q16. (a) Design an algorithm to find both the **Minimum and Maximum** elements in an array using the **Divide and Conquer** strategy.

(b) Analyze the number of comparisons made. How does it compare to the naive linear scan?

Answer: (a) Algorithm (MinMax): Function 'FindMinMax(arr, low, high)':

1. If 'low == high': Return '(arr[low], arr[low])'.
2. If 'high == low + 1': Compare 'arr[low]' and 'arr[high]', return pair '(min, max)'.
3. Else:
 - 'mid = (low + high) / 2'
 - '(min1, max1) = FindMinMax(arr, low, mid)'
 - '(min2, max2) = FindMinMax(arr, mid+1, high)'
 - Return '(min(min1, min2), max(max1, max2))'

(b) Complexity Analysis: Recurrence for number of comparisons $C(n)$:

$$C(n) = 2C(n/2) + 2$$

Base case: $C(2) = 1$. Solving this:

$$C(n) = \frac{3n}{2} - 2$$

Comparison:

- **Naive Scan:** $2n - 2$ comparisons (worst case: verify every element for min and max independently).
- **Divide and Conquer:** $1.5n - 2$ comparisons.
- **Improvement:** D&C performs roughly 25% fewer comparisons.

Q17. (a) Define the **Longest Common Subsequence (LCS)** problem.

(b) Find the LCS of sequences $X = \text{BDCABA}$ and $Y = \text{ABCBDAB}$ using Dynamic Programming. Show the DP table.

Answer: (a) Problem: Given two sequences, find the length of the longest subsequence present in both of them. A subsequence is a sequence that appears in the same relative order, but not necessarily contiguously.

(b) DP Solution: Formula: If $X[i] == Y[j]$, $c[i, j] = 1 + c[i - 1, j - 1]$. Else $c[i, j] = \max(c[i - 1, j], c[i, j - 1])$.

$X: \text{BDCABA}$ ($m = 6$), $Y: \text{ABCBDAB}$ ($n = 7$).

	\emptyset	A	B	C	B	D	A	B
\emptyset	0	0	0	0	0	0	0	0
B	0	0	1	1	1	1	1	1
D	0	0	1	1	1	2	2	2
C	0	0	1	2	2	2	2	2
A	0	1	1	2	2	2	3	3
B	0	1	2	2	3	3	3	4
A	0	1	2	2	3	3	4	4

Traceback: $(6, 7) = 4$ (Match 'A'? No, from Left). $(6, 6) = 4$ (Match 'B'? No, from Top).
 ... Common chars traced: ****BCBA****. (LCS Length = 4). *Note: Other valid LCS include BDAB, BCAB.*

Q18. Explain the **Bellman-Ford Algorithm** for single-source shortest paths. How does it handle negative edge weights? What is the condition to detect a negative weight cycle?

Answer: Algorithm Concept: Bellman-Ford computes shortest paths from a source vertex to all other vertices in a weighted digraph. Unlike Dijkstra, it can handle negative weights. It is based on the **Relaxation** principle applied repeatedly.

Steps:

1. Initialize distance to source = 0, others = ∞ .
2. Relax all edges $|V| - 1$ times.
3. For each edge (u, v) with weight w :

$$\text{If } dist[u] + w < dist[v], \text{ then } dist[v] = dist[u] + w$$

Negative Cycle Detection: After $|V| - 1$ iterations, the shortest paths are guaranteed to be found if no negative cycle exists. We run the relaxation loop one more time (the $|V|^{th}$ time).

- If any distance value changes (i.e., $dist[u] + w < dist[v]$), then a ****Negative Weight Cycle**** exists in the graph. The algorithm returns False (no solution).

Q19. (a) Explain the **0/1 Knapsack Problem** using the **Branch and Bound** technique.

(b) How is the Upper Bound (UB) calculated for a node in the state space tree? (Hint: Relationship with Fractional Knapsack).

Answer: (a) Branch and Bound Approach: Instead of exploring all subsets (Backtracking) or building a table (DP), we organize the search in a State Space Tree.

- We explore nodes based on the "best possible" outcome (Upper Bound) they can lead to.
- We maintain a global variable 'maxProfit'.
- If a node's Upper Bound is less than 'maxProfit', we prune it.

(b) Calculating Upper Bound: To check how "promising" a node is, we relax the integer constraint. We calculate the profit obtainable by solving the ****Fractional Knapsack**** problem for the remaining capacity.

- $UB = \text{CurrentProfit} + \text{FractionalBound}(\text{RemainingItems}, \text{RemainingCap})$
- Since Fractional Knapsack is greedy and optimal, this value represents the maximum possible profit achievable from this branch. If this value \leq current best solution, the branch is dead.

Q20. Explain the concept of **Polynomial Time Reducibility**. Show that if Problem A is NP-Complete and Problem A reduces to Problem B in polynomial time ($A \leq_p B$), then Problem B is NP-Hard.

Answer: Polynomial Time Reducibility ($A \leq_p B$): This means there exists a function f computable in polynomial time that transforms any instance I_A of problem A into an instance I_B of problem B, such that:

$$I_A \text{ is YES instance} \iff f(I_A) \text{ is YES instance of B}$$

This implies B is at least as hard as A.

Proof Logic: 1. We know A is NP-Complete. This means every problem $X \in NP$ can be reduced to A ($X \leq_p A$). 2. We are given $A \leq_p B$. 3. By transitivity of reduction: If $X \leq_p A$ and $A \leq_p B$, then $X \leq_p B$. 4. Therefore, every problem in NP is reducible to B. 5. By definition, if every problem in NP is reducible to B, then ****B is NP-Hard****. (Note: If B is also in NP, then B is NP-Complete).