

## Program 1: Huffman Coding (Greedy Algorithm)

### Code Summary:

This program implements **Huffman Coding**, a lossless data compression technique.

It uses a **greedy algorithm** to assign shorter binary codes to more frequent characters.

---

### Viva Questions & Answers (15)

#### 1. Q1. What is Huffman coding?

A: Huffman coding is a data compression algorithm that assigns variable-length binary codes to input characters — shorter codes for more frequent characters.

#### 2. Q2. Which algorithmic strategy is used in Huffman coding?

A: It uses the **Greedy strategy**, selecting the two least frequent nodes repeatedly.

#### 3. Q3. What is the role of a MinHeap in this code?

A: A MinHeap helps efficiently extract the two smallest frequency nodes during each iteration.

#### 4. Q4. What is the purpose of the MinHeapNode class?

A: It represents each character node with its frequency, left child, and right child for tree construction.

#### 5. Q5. Why do we override the \_\_lt\_\_ method?

A: To define custom comparison logic based on frequency, required by the heapq module.

#### 6. Q6. What does the \$ symbol represent in this program?

A: It represents an **internal node** in the Huffman tree (not an actual data character).

#### 7. Q7. How does the algorithm combine nodes?

A: It repeatedly removes the two nodes with the smallest frequency, creates a new node with their combined frequency, and re-inserts it into the heap.

#### 8. Q8. What is the time complexity of Huffman coding?

A:  $O(n \log n)$  due to heap operations.

#### 9. Q9. What is the space complexity?

A:  $O(n)$ , as it stores all nodes and the Huffman tree.

#### 10. Q10. What is printed as output?

A: Each character and its corresponding Huffman binary code.

**11. Q11. What happens when all characters have equal frequency?**

**A:** All will have codes of equal length since no character has higher priority.

**12. Q12. Why is Huffman coding called “lossless”?**

**A:** Because no data is lost during compression — the original data can be perfectly reconstructed.

**13. Q13. Where is Huffman coding used in real life?**

**A:** In JPEG image compression, ZIP file compression, and MP3 audio encoding.

**14. Q14. What Python module is used for heap implementation?**

**A:** The built-in heapq module.

**15. Q15. What happens in the print\_codes function?**

**A:** It traverses the Huffman tree recursively to print binary codes for each character.

**16. • Q16. Why do we use a priority queue (MinHeap) instead of a normal list?**

**A:** Because MinHeap gives efficient access to the smallest frequency element in **O(log n)** time.

**17. • Q17. What happens if there's only one character in the input?**

**A:** The algorithm assigns a single code, usually “0”, since there's no need for a tree.

**18. • Q18. How is Huffman tree different from a Binary Search Tree (BST)?**

**A:** A Huffman tree is built based on frequency, not key order. It's a **full binary tree**, not ordered like BST.

**19. • Q19. Why are Huffman codes prefix-free?**

**A:** No code is a prefix of another, ensuring unambiguous decoding.

**20. • Q20. What is the output if all characters have same frequency?**

**A:** Each character gets a code of the same length, since priority doesn't change.

**21. • Q21. How do you calculate total bits saved after compression?**

**A:** Compare total bits before and after using Huffman codes based on character frequencies.

**22. • Q22. Why do we use recursion in print\_codes()?**

**A:** To traverse the Huffman tree using depth-first traversal and generate codes for each leaf node.

**23. • Q23. What type of tree is generated by Huffman coding?**

**A:** A **binary tree** where each leaf node represents a character.

24. • **Q24. Can Huffman coding handle non-text data (like images)?**

**A:** Yes, any data with frequency distribution (e.g., pixel values) can be compressed.

25. • **Q25. What is the limitation of Huffman coding?**

**A:** It doesn't adapt dynamically to data — static Huffman coding requires prior frequency knowledge.

---

## 👑 Program 2: N-Queens Problem (Backtracking)

### 📘 Code Summary:

This program solves the **N-Queens problem** using **backtracking**, with a twist — the **first queen's column position is fixed** by the user.

---

### 💬 Viva Questions & Answers (15)

1. **Q1. What is the N-Queens problem?**

**A:** Placing N queens on an NxN chessboard so that no two queens attack each other.

2. **Q2. What technique is used to solve it?**

**A:** **Backtracking**, a recursive algorithm to explore all possible placements.

3. **Q3. How is safety of a position checked?**

**A:** By ensuring the same column and both diagonals have no other queen.

4. **Q4. What are col, posDiag, and negDiag used for?**

**A:** They store occupied columns, positive diagonals ( $r+c$ ), and negative diagonals ( $r-c$ ) respectively.

5. **Q5. What happens when a valid configuration is found?**

**A:** The board state is added to the res list as a valid solution.

6. **Q6. What does backtrack(r) do?**

**A:** It attempts to place a queen in row r by testing valid columns recursively.

7. **Q7. What is the base case in recursion?**

**A:** When  $r == n$ , meaning all queens are successfully placed.

8. **Q8. What is meant by "backtracking"?**

**A:** Reverting a previous step (removing a queen) when no valid move exists further.

9. **Q9. How does fixing the first queen affect solutions?**

**A:** It reduces the total number of configurations by fixing one queen's column.

10. **Q10. Why are we using sets instead of lists?**

**A:** Sets provide  $O(1)$  lookup for checking occupied positions.

11. **Q11. What is printed as output?**

**A:** A board matrix of 1s (queen) and 0s (empty spaces).

12. **Q12. What is the time complexity?**

**A:**  $O(N!)$  in the worst case.

13. **Q13. What is the space complexity?**

**A:**  $O(N^2)$  for the board and recursive stack.

14. **Q14. Can this algorithm find all possible solutions?**

**A:** Yes, by continuing recursion even after finding one valid configuration.

15. **Q15. In real life, where is backtracking used?**

**A:** In puzzles like Sudoku, crosswords, pathfinding, and constraint satisfaction problems.

16. • **Q16. Why can't we place two queens in the same column?**

**A:** Because queens attack vertically, and each column can only have one queen.

17. • **Q17. Why do we represent diagonals as  $(r + c)$  and  $(r - c)$ ?**

**A:** It uniquely identifies each diagonal — all cells with the same  $(r + c)$  or  $(r - c)$  are on the same diagonal.

18. • **Q18. What is the advantage of using sets for column and diagonal tracking?**

**A:** They allow constant-time lookup ( $O(1)$ ), improving efficiency.

19. • **Q19. What does backtracking mean in a single sentence?**

**A:** Trying all possibilities by placing a queen and undoing if it leads to no solution.

20. • **Q20. What would happen if we remove the backtrack step (removal code)?**

**A:** The algorithm would fail because it wouldn't explore other configurations.

21. • **Q21. Why do we start recursion from row 1 instead of 0 here?**

**A:** Because the first queen's column is fixed by the user, so row 0 is already filled.

**22. • Q22. Can we solve N-Queens iteratively?**

**A:** It's possible but very complex. Recursive backtracking is more natural for this problem.

**23. • Q23. What is the output format of this program?**

**A:** It prints the chessboard as a matrix with “1” for queen positions and “0” for empty squares.

**24. • Q24. What is the minimum N for which the N-Queens problem has a solution?**

**A:** For  $N \geq 4$ , the problem has valid solutions.

**25. • Q25. What is the significance of fixing the first queen's column?**

**A:** It helps explore reduced configurations or test specific placements for optimization.

---

 **Program 3: Fibonacci Series — Recursive vs Iterative Analysis**

 **Code Summary:**

This code compares **recursive** and **iterative** Fibonacci implementations, measuring **execution time** and **memory usage** using time and tracemalloc.

---

 **Viva Questions & Answers (15)**

**1. Q1. What is the Fibonacci series?**

**A:** A sequence where each term is the sum of the two preceding ones: 0, 1, 1, 2, 3, 5, 8...

**2. Q2. How is Fibonacci implemented recursively?**

**A:** Each call computes  $F(n-1) + F(n-2)$  until  $n \leq 1$ .

**3. Q3. Why is recursion inefficient for Fibonacci?**

**A:** Because it recomputes the same values multiple times (exponential complexity).

**4. Q4. What is the iterative approach?**

**A:** Using a loop to compute Fibonacci numbers sequentially — efficient and faster.

**5. Q5. What is the time complexity of recursive Fibonacci?**

**A:**  $O(2^n)$  due to repeated computations.

6. **Q6. What is the time complexity of iterative Fibonacci?**

A:  $O(n)$  — linear time.

7. **Q7. What is the space complexity of recursive Fibonacci?**

A:  $O(n)$  due to function call stack.

8. **Q8. What is the space complexity of iterative Fibonacci?**

A:  $O(1)$  — uses only two variables.

9. **Q9. Why is tracemalloc used?**

A: To measure the current and peak memory usage during function execution.

10. **Q10. What is the purpose of gc.collect()?**

A: It forces garbage collection to clear unused memory before measurement.

11. **Q11. What is time.perf\_counter() used for?**

A: It provides high-resolution timing to measure execution duration accurately.

12. **Q12. Why do we compare both methods?**

A: To demonstrate performance trade-offs between recursion and iteration.

13. **Q13. What happens if n is large (e.g., 40 or 50)?**

A: Recursive function becomes extremely slow due to exponential growth of calls.

14. **Q14. Which approach is better for large n?**

A: The **iterative** approach — faster and memory-efficient.

15. **Q15. What real-world concept does this program teach?**

A: It shows how **algorithm design choices** (recursion vs iteration) affect **time and space efficiency**.

16. **Q16. Why is recursion slower for Fibonacci?**

A: It repeats many subproblems without storing results — leading to **exponential growth** in calls.

17. **Q17. How can we optimize recursive Fibonacci?**

A: By using **Dynamic Programming (Memoization)** to store computed results.

18. **Q18. What is memoization?**

A: A technique to cache results of previous computations to avoid redundant recursion.

19. **Q19. What is the output for n=0 and n=1?**

A: For n=0 → 0, and for n=1 → 1 (base cases).

20. **Q20. Why is iterative Fibonacci more memory efficient?**

**A:** It uses only two variables (`a`, `b`) and no function call stack.

21. **Q21. What is the use of the `tracemalloc` module?**

**A:** It tracks memory allocation to measure how much memory the function uses.

22. **Q22. Why is `time.perf_counter()` preferred over `time.time()`?**

**A:** It provides **higher precision** and includes time spent in sleep and OS calls.

23. **Q23. What's the main difference in time complexity between recursive and iterative Fibonacci?**

**A:** Recursive is  **$O(2^n)$** ; Iterative is  **$O(n)$** .

24. **Q24. What happens if we increase n to 40 or more in recursive Fibonacci?**

**A:** The program becomes extremely slow and consumes a lot of memory.

25. **Q25. Why do we compare both methods in one program?**

**A:** To understand the **impact of algorithm design** on performance — a key learning goal in Data Structures & Algorithms.

---

## Practical 4: Fractional Knapsack Problem (Greedy Algorithm)

### Concept Summary:

- You're given items with **weights and values**.
  - The goal: **maximize total value** in the knapsack with limited capacity.
  - Unlike 0/1 knapsack, you can take **fractions** of items.
  - Sorting is done by **value-to-weight ratio (v/w)**.
- 

### Top 25 Viva Questions & Answers

1. **Q1. What is the Fractional Knapsack problem?**

**A:** It's an optimization problem where items can be broken into fractions to maximize total profit within a limited capacity.

2. **Q2. What strategy is used to solve this problem?**

A: The **Greedy Strategy** — always take the item with the highest value-to-weight ratio first.

3. **Q3. What is the formula for the ratio?**

A: Value/Weight ( $v/w$ ).

4. **Q4. Why is it called “fractional”?**

A: Because we can take a portion (fraction) of an item instead of taking it whole.

5. **Q5. What's the difference between 0/1 and fractional knapsack?**

A: In 0/1, you can't split items (take or leave), while in fractional, you can take parts.

6. **Q6. What is the time complexity?**

A:  $O(n \log n)$ , due to sorting by ratio.

7. **Q7. Why do we sort the items?**

A: To ensure we pick the most valuable items first based on ratio.

8. **Q8. What data structure is best for this?**

A: A list or array, often sorted using Python's sorted().

9. **Q9. Can fractional knapsack be solved using dynamic programming?**

A: No, it doesn't require DP; greedy gives the optimal solution.

10. **Q10. What happens if two items have the same ratio?**

A: Any order gives the same result, as both contribute equally per weight.

11. **Q11. What is the base condition for capacity?**

A: Stop when total capacity becomes 0.

12. **Q12. How do you handle partial items in code?**

A: Take a fraction: `total_value += remaining_capacity * (value / weight)`.

13. **Q13. Why do we break after partial item selection?**

A: Because the knapsack is full.

14. **Q14. What is the best case scenario?**

A: When all items fit entirely within capacity.

15. **Q15. What is the worst case scenario?**

A: When only one item can be partially taken.

16. **Q16. What is greedy choice property?**

A: Making the best local choice (max ratio) leads to the global optimal solution.

**17. Q17. What is optimal substructure?**

A: The optimal solution can be built from optimal solutions of smaller subproblems.

**18. Q18. What kind of problem is it (min/max)?**

A: A maximization problem.

**19. Q19. What is the output of the algorithm?**

A: Maximum possible value for the given capacity.

**20. Q20. How do you verify if your solution is correct?**

A: By manually calculating ratios and simulating the selection process.

**21. Q21. What happens if weights are 0?**

A: It's invalid since ratio (v/w) becomes infinite.

**22. Q22. How to handle user input for items?**

A: By taking pairs of values and weights from input.

**23. Q23. What Python function helps in sorting by ratio?**

A: `sorted(items, key=lambda x: x[0]/x[1], reverse=True)`

**24. Q24. What is the final output printed in your program?**

A: "Maximum value by value/weight ratio: <value>"

**25. Q25. Real-life example?**

A: Loading a truck with goods of varying weights and profits to maximize revenue.

---



## Practical 5: 0/1 Knapsack Problem (Dynamic Programming)



### Concept Summary:

- You have items with **values** and **weights**.
  - You must **choose items** to maximize total value without exceeding capacity.
  - You can't take fractions — either include or exclude.
  - Uses **Dynamic Programming (DP)** to build a 2D table.
- 



## Top 25 Viva Questions & Answers

**1. Q1. What is the 0/1 Knapsack problem?**

**A:** A problem where we must select items to maximize total value without exceeding capacity — either include (1) or exclude (0) each item.

**2. Q2. What technique is used here?**

**A:** Dynamic Programming (DP).

**3. Q3. Why not Greedy?**

**A:** Greedy doesn't work here because items can't be fractionally divided.

**4. Q4. What is the recursive formula?**

**A:**

$dp[i][w] = \max(dp[i-1][w], dp[i-1][w - \text{weights}[i-1]] + \text{values}[i-1])$   
if weight fits.

**5. Q5. What is the base case?**

**A:** When  $i=0$  or  $w=0$ ,  $dp[i][w]=0$ .

**6. Q6. What does  $dp[i][w]$  represent?**

**A:** Maximum value attainable using first  $i$  items with capacity  $w$ .

**7. Q7. What is the size of the DP table?**

**A:**  $(n+1) \times (W+1)$ .

**8. Q8. What is the time complexity?**

**A:**  $O(n \times W)$ .

**9. Q9. What is the space complexity?**

**A:**  $O(n \times W)$  (can be reduced to  $O(W)$ ).

**10. Q10. What happens if item's weight > capacity?**

**A:** We skip the item:  $dp[i][w] = dp[i-1][w]$ .

**11. Q11. How is the table filled?**

**A:** Row by row, where each cell depends on previous computations.

**12. Q12. What does the last cell  $dp[n][W]$  represent?**

**A:** The **maximum profit** obtainable.

**13. Q13. What is the method used to find which items were selected?**

**A:** Backtracking from  $dp[n][W]$ .

**14. Q14. What condition tells us an item was included?**

**A:** If  $dp[i][w] \neq dp[i-1][w]$ , item  $i-1$  was included.

**15. Q15. How do we update capacity during backtracking?**

**A:** Subtract the included item's weight from w.

**16. Q16. Why is DP better than recursion here?**

**A:** DP avoids recomputation by storing results (overlapping subproblems).

**17. Q17. What kind of problems use DP?**

**A:** Problems with **overlapping subproblems** and **optimal substructure**.

**18. Q18. What does “optimal substructure” mean?**

**A:** The solution to a problem can be built from its subproblem solutions.

**19. Q19. What happens if all weights are larger than capacity?**

**A:** Maximum value = 0.

**20. Q20. What does selected\_items array store?**

**A:** The indices of items included in the optimal solution.

**21. Q21. Can we solve it recursively?**

**A:** Yes, but it's inefficient without memoization.

**22. Q22. What real-world applications use 0/1 knapsack?**

**A:** Budget optimization, resource allocation, cargo loading.

**23. Q23. Why do we initialize DP table with zeros?**

**A:** It represents no items or zero capacity → zero profit.

**24. Q24. Can we use floating weights or values?**

**A:** Usually integers; fractional weights fit fractional knapsack, not 0/1.

**25. Q25. What's printed in the output?**

**A:** “Maximum value: <value>” and “Selected items: <indices>”.

No.	Practical Title / Algorithm	Key Idea / Approach	Working Summary	Time Complexity	Applications / Use Cases	Key Viva Questions
1	DFS (Depth First Search)	Start from node → explore one branch fully → backtrack → explore next branch. Uses <b>stack / recursion</b> to explore as deep as possible before backtracking.	O(V + E)	Pathfinding, maze solving, topological sort	1. What data structure is used in DFS? 2. How DFS differs from BFS? 3. Applications of DFS? 4. Can DFS detect cycles? 5. Space complexity?	
2	BFS (Breadth First Search)	Start from node → explore all neighbors → then next level nodes. Uses <b>queue</b> to explore nodes level by level.	O(V + E)	Shortest path in unweighted graph, social network search	1. What data structure is used in BFS? 2. Where BFS is preferred over DFS? 3. BFS real-life example? 4. BFS vs DFS difference? 5. BFS output order example?	
3	A Algorithm (Informed Search)*	Route navigation (Google Maps), games f(n) = g(n) + h(n); chooses path with lowest f(n). Uses <b>heuristics</b> ( <b>h(n)</b> ) + cost ( <b>g(n)</b> ) to find optimal path.	O(E) (depends on heuristic)	1. What is f(n)? 2. What is heuristic? 3. A* vs Dijkstra? 4. What is admissible heuristic? 5. A* application?		
4	N-Queens Problem (Backtracking)	Try placing queen row by row → O(N!) Place N queens on N×N board so check column &	Chess puzzle, constraint satisfaction	1. What is backtracking? 2. How many diagonals checked? 3. Base		

No.	Practical Title / Algorithm	Key Idea / Approach	Working Summary	Time Complexity	Applications / Use Cases	Key Viva Questions
		no two attack each other.	diagonals → backtrack if conflict.			condition of recursion?4. Meaning of safe position?5. Time complexity?
5	<b>Fibonacci (Recursive vs Iterative)</b>	Compare recursion vs iteration in generating Fibonacci sequence.	Recursive calls → stack overhead; Recursive: $O(2^n)$ Iterative: iterative → simple loop.	Iterative: $O(n)$	Dynamic programming, financial modeling	1. Why recursion is slower?2. Which uses more memory?3. Define base case?4. Example of recursion in AI?5. Output for n=5?