# Detailed Notes — Data Structures & Algorithms in Python.

Below is a breakdown by lesson, with deeper detail, key ideas, and some additional insights.

## **Course Overview & Structure**

#### From the course page:

- The course is **beginner-friendly**, covering data structures (linked lists, stacks, queues, graphs) and algorithms (search, sorting, recursion, DP). (jovian.ai)
- You watch coding-oriented video lectures, practice in Jupyter notebooks, and solve interview-like problems. (jovian.ai)
- There are lessons, corresponding assignments, and a final project. (<u>iovian.ai</u>)
- The lessons are:
  - Binary Search, Linked Lists, Complexity Preview
  - Binary Search Trees, Traversals, Recursion
  - Sorting Algorithms & Divide & Conquer
  - Recursion and Dynamic Programming
  - Graph Algorithms (BFS, DFS, Shortest Paths)
  - o Python Interview Questions, Tips & Advice
  - o A Project where you solve a problem step-by-step (jovian.ai)
- Assignments are interleaved, e.g.

- Assignment 1: Binary Search Practice
- Assignment 2: Hash Tables & Dictionaries
- Assignment 3: Divide & Conquer Practice
- o And so on (<u>jovian.ai</u>)
- The instructor is Aakash N S background in software engineering, open source, online education. (<u>jovian.ai</u>)

# **Detailed Lesson-by-Lesson Notes & Core Concepts**

Below are deeper notes, sorted by lesson, with key ideas, sample problems, and deeper insight.

## Lesson 1: Binary Search, Linked Lists & Complexity Preview

#### **Key Topics:**

- Linear Search vs. Binary Search
- Big O notation & complexity theory
- Implementation of Linked Lists using Python classes

#### Details & insights:

#### 1. Linear Search

- Traverse list one by one, compare each element.
- Time complexity: O(N) in worst case.
- o Simple, works on unsorted data.

#### 2. Binary Search

- Works only on sorted arrays/lists.
- o Idea: compare target to middle element, then discard half of the search space.
- Time complexity: 0 (log N) (base 2)
- Must carefully handle index boundaries (start, end, mid) and off-by-one errors.

#### 3. Complexity & Big O

- Introduction to asymptotic analysis: focus on dominant term, ignore constants and lower-order terms.
- Compare best, average, worst cases.
- Space complexity (extra memory use) also matters.

#### 4. Linked Lists (using Python classes)

- Node class: value + next pointer (or reference).
- A LinkedList wrapper class might have head pointer, methods like insert, delete, search.
- Advantages: easy insertion/deletion at head, constant time for certain operations (if you have the node).
- Disadvantages: no direct indexing, so random access is O(N).

## **Lesson 2: Binary Search Trees, Traversals & Recursion**

#### **Key Topics:**

- Binary Trees and Binary Search Trees (BST)
- Tree traversal methods
- Recursion fundamentals
- Balanced BSTs and optimizations

#### **Details & insights:**

#### 1. Binary Tree & BST definitions

- A **binary tree**: each node has up to two children (left, right).
- A BST (Binary Search Tree): left subtree has smaller keys, right subtree has larger keys.
- o Common operations: insert, search, delete.

#### 2. Traversals

- **Inorder** (Left  $\rightarrow$  Node  $\rightarrow$  Right)
- **Preorder** (Node  $\rightarrow$  Left  $\rightarrow$  Right)
- **Postorder** (Left  $\rightarrow$  Right  $\rightarrow$  Node)
- These can be implemented recursively, or using a stack (iterative).

#### 3. Recursion

- Recursion is when a function calls itself with a smaller/simpler subproblem.
- Base case(s) are essential to terminate recursion.
- Many tree algorithms are naturally recursive (e.g. traversal, depth, height).

#### 4. Balanced BSTs & optimizations

- In a degenerate BST (like a linked list), operations degrade to O(N).
- Balanced BSTs (AVL, Red-Black Trees) keep depth to 0 (log N).
- In the context of the course, you may get an introduction to balancing or recognizing when tree degenerates.

## **Lesson 3: Sorting Algorithms & Divide & Conquer**

#### **Key Topics:**

- Basic sorting: bubble sort, insertion sort
- Divide & Conquer paradigm
- Merge Sort
- Quick Sort

#### **Details & insights:**

#### 1. Bubble Sort & Insertion Sort

- Bubble: repeatedly swap adjacent elements if out of order  $\rightarrow 0 (N^2)$ .
- o Insertion: build sorted portion one by one by inserting into correct spot  $\rightarrow$  also  $O(N^2)$  in worst case.

#### 2. Divide & Conquer

- A broad algorithmic strategy:
  - Divide the problem into subproblems,
  - Solve them recursively,
  - Combine results.
- Merge sort is a canonical example.

## 3. Merge Sort

- Divide the list roughly in half, sort each half, then merge two sorted halves.
- Time complexity: O(N log N) (best, average, worst).
- Space complexity: O(N) extra space (for merging).

#### 4. Quick Sort

 Pick a pivot, partition elements < pivot to one side, > pivot to other side, then recursively sort partitions.

- Average-case time complexity: 0(N log N)
- Worst-case: O(N^2) (e.g. if pivot is always worst)
- o In practice, good with randomization or choosing pivots well.
- o In-place partitioning helps reduce space overhead.

#### 5. Other Sorting Considerations

- o Stable vs. unstable sorting
- o In-place vs. not in-place
- Hybrid algorithms (e.g. switching to insertion sort for small subarrays)

## **Lesson 4: Recursion & Dynamic Programming**

#### **Key Topics:**

- Deep dive on recursion & memoization
- Subsequence problems
- 0/1 Knapsack
- Backtracking & pruning

#### **Details & insights:**

#### 1. Recursion & Memoization

- Pure recursion may recompute the same subproblem multiple times.
- o Memoization (caching intermediate results) saves time.
- o Top-down (recursive + memo) vs bottom-up (DP table) approaches.

#### 2. Subsequence Problems

- Example: Longest Common Subsequence (LCS)
- If the last characters match, you reduce both indices. Otherwise, consider skipping one character from one string.
- Use DP to avoid repeated work.

#### 3. 0/1 Knapsack

- You have items (weight, value) and a capacity.
- Recurrence: for each item, either take it (if fits) or skip it, and pick better of two
  options.
- Use DP (2D table) so that the complexity becomes 0(n \* capacity).

## 4. Backtracking & Pruning

- o Backtracking: try all possibilities (e.g. for combinatorial problems) with recursion.
- Pruning: stop exploring a branch when you know it can't yield a better solution (e.g. bounding).
- Useful for subset sum, permutations, combinations, etc.

## **Lesson 5: Graph Algorithms (BFS, DFS & Shortest Paths)**

#### **Key Topics:**

- Graph representation (adjacency list, adjacency matrix)
- BFS & DFS
- Shortest path algorithms (especially Dijkstra)
- Directed graphs & weights

#### **Details & insights:**

#### 1. Graph Representations

- Adjacency List: for each node, a list (or dict) of neighbors → efficient for sparse graphs.
- Adjacency Matrix: 2D matrix, graph[u][v] = cost or boolean → good for dense graphs or fast edge lookups.

#### 2. BFS (Breadth-First Search)

- Uses a queue
- Visits nodes in "layers" from a starting node
- Good for shortest path (in unweighted graphs)
- $\circ$  Time complexity: O(V + E) where V = nodes, E = edges

#### 3. **DFS (Depth-First Search)**

- Uses recursion or explicit stack
- Goes deep, then backtracks
- Useful for exploring connectivity, detecting cycles, topological sort
- Time complexity: also 0 (V + E)

#### 4. Shortest Paths in Weighted Graphs

- o For non-negative weights, **Dijkstra's algorithm** is classical.
- Using a priority queue (min-heap) yields a complexity like 0 ((V + E) log V).
- Key idea: maintain tentative distance to each node, repeatedly pick the undiscovered node with smallest distance, relax its edges.

#### 5. **Directed Graphs & Variations**

- o Graphs may have directed edges ( $u \rightarrow v$  but not  $v \rightarrow u$ ).
- Problems like detecting cycles, strongly connected components, topological sort may be introduced (or at least touched on).

## **Lesson 6: Python Interview Questions, Tips & Advice**

#### **Key Topics:**

- Real interview-style problems with solutions
- Coding challenge strategies
- Time management & thinking process
- Tips on writing clean, testable code

#### Details & insights:

- Emphasis likely on practicing common patterns: sliding window, two pointers, hash-based counting, recursion, DP
- Advice on debugging, writing test cases, reading the problem carefully
- Presenting your solution, articulating complexity, edge cases

## **Project & Assignments**

- After lessons and assignments, a **project** pushes you to pick a non-trivial problem, break it down, and document step-by-step solution. (jovian.ai)
- Assignments reinforce each module (e.g. implementing hash tables from scratch, divide & conquer problems, graph problems). (jovian.ai)
- You earn a certificate by completing weekly assignments. (jovian.ai)

# **Additional Depth & Extended Insights**

Here are some deeper points and expansions (beyond what might be strictly in the course) that tie into the course's topics, which may help you understand more fully.

#### On Complexity — Amortized Analysis & Constant Factors

- Sometimes operations average out over many calls (e.g. dynamic array resizing).
- Big-O ignores constant multipliers, but in real settings, constant factors and lower-order terms can matter in practice.
- Also consider worst-case vs average-case performance.

#### On Hash Tables & Collisions

- When designing a hash table from scratch, collision resolution strategies include:
  - **Chaining** (each bucket holds a list of entries)
  - Open addressing (linear probing, quadratic probing, double hashing)
- Resizing (rehashing) is important: when load factor exceeds a threshold, expand the table and rehash all entries.

#### On Balanced Trees & Alternatives

- Beyond basic BST, balanced trees like AVL, Red-Black Trees, Treaps, Splay Trees
  ensure the height remains 0(log N) under insertions & deletions.
- Sometimes **heaps** (binary heap, Fibonacci heap) are used for priority queue tasks, particularly in graph algorithms or scheduling.

## On Advanced Graph Algorithms (beyond Dijkstra)

- Bellman-Ford handles negative weights (no negative cycles).
- Floyd-Warshall computes all-pairs shortest paths.
- **Johnson's algorithm** (for sparse graphs)
- Minimum Spanning Tree (MST): Prim's and Kruskal's algorithms.
- Topological sort, strongly connected components (Kosaraju's, Tarjan's)

• Flow algorithms (e.g. Edmonds-Karp) etc.

## On More Dynamic Programming Patterns

- Common pattern types:
  - Knapsack / subsets / combinatorial optimization
  - Longest increasing subsequence (LIS)
  - Edit distance / string alignment
  - Matrix chain multiplication
  - DP on trees / graphs
- Recognizing overlapping subproblems and optimal substructure is key.

## On Space Optimizations & Trade-offs

- In DP, sometimes only a few previous rows are needed (so we can reduce 2D table to 1D).
- In recursion, careful about stack depth; tail recursion vs iteration.
- Some "optimized" algorithms trade time for space or vice versa.

## On Coding Best Practices

- Write modular functions (small pieces)
- Use assertions / unit tests
- Document time / space complexity in comments
- Consider edge cases: empty input, single element, maximum input, invalid input
- Use version control (Git) to track changes & experiment safely
- Use a test suite (like Jovian's) to automate verification

## **Suggested Additional Examples & Exercises**

To cement understanding, here are extra ideas to try (on your own):

- 1. Implement a Hash Map from scratch
  - Use open addressing or chaining
  - Support put(key, value), get(key), delete(key)
- 2. Find k-th smallest element in an unsorted array
  - Using sorting vs. using a selection algorithm (like Quickselect)
- 3. Longest Palindromic Subsequence / Substring
  - Use DP or expand-around-center techniques
- 4. **Graph problem**: shortest path on a grid with obstacles (e.g. 0/1 weights)
  - Use BFS or Dijkstra depending on weights
- 5. **Interview-style problem**: "given a string, find the minimum window substring that contains all characters of another string" typical sliding window + hash table problem.
- 6. Compare different sorting algorithms in practice
  - o Measure runtimes on random arrays of increasing size
  - Compare merge sort, quicksort, insertion sort on various data distributions (random, sorted, reversed)