

Data Structures & Algorithms

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Table of contents

Preface	3
Resources	4
1 Introduction	5
1.1 Perspective	5
1.2 Subheading	5
2 Foundations	6
2.1 Resources	6
3 Pointers	7
4 Arrays	8
5 Strings	9
6 Streams and IO	10
7 Abstract	11
8 Object-Oriented Programming	12
9 Dynamic Memory	13
10 Linked List	14
11 Iterators	15
12 Recursion	16
13 Function Objects, Functors	17
14 Error Handling & Exceptions	18
15 Stacks & Queues	19

16 Complexity Analysis	20
16.1 References	20
16.2 Runtime Overview	20
16.3 Common Time Complexities	20
16.3.1 $O(1)$ - Constant	20
16.3.2 $O(\log(N))$ - Logarithmic	20
16.3.3 $O(N)$ - Linear	21
16.3.4 $O(K \log(N))$	21
16.3.5 $O(N \log(N))$ - Log-Linear	22
16.3.6 $O(N^2)$ - Quadratic	22
16.3.7 $O(2^N)$ - Exponential	22
16.3.8 $O(N!)$ - Factorial	23
16.4 Amortized Time Complexity	23
17 STL	24
18 Heaps	25
18.1 References	25
18.2 Heap? Priority Queue?	25
18.3 Heap in C++	25
19 Trees	26
20 Searching	27
21 Sorting	28
22 Hash Maps	29
23 Graphs	30
24 Brute-Force & Greedy Algorithms	31
25 Divide and Conquer, Dynamic Programming	32
26 Backtracing, Branch and Bound Algorithms	33
27 Summary	34
28 Tips on Solving DS&A Questions	35
28.1 Problem Solving Flowchart	35
28.2 DS&A Roadmap	35
28.3 Problem Flowchart	37
28.4 ROI	37
28.5 “Academic” Algorithms	37

28.6 Keyword to Algo	37
29 Problems & Explanations	42

Preface

This is my notes on Data Structures & Algorithms in C++.

Resources

Some relevant resources:

- [EECS 280 - Programming and Intro Data Structures \(University of Michigan\)](#)
- [EECS 281 - Data Structures and Algorithms \(University of Michigan\)](#)
- [EECS 376 - Foundations of Computer Science](#)
- [Data Structures & Algorithms - Google Tech Dev Guide](#)
- [EECS 281 References](#)
- [The Algorithms](#)

Practice Resources:

- [LeetCode](#)
- [NeetCode](#)
 - [Blind 75](#)
- [Codeforces](#)

Interview Resources:

- [Leetcode Patterns](#)
- [Learning Resources - Reddit-wiki-programming](#)
- [AlgoMonster](#)
- [Tech Interview Handbook](#)

C++ Guides:

- [learncpp.com](#)

Books:

- [Algorithms - Jeff Erickson](#)
- [Cracking the Coding Interview](#)

1 Introduction

1.1 Perspective

i Note 1: Definition - Definition goes here

Definition is defined by: definition.

This notebook contains programming basics, data structures, and algorithms. The language of choice is C++, and concepts from C++ STL are also covered.

1.2 Subheading

2 Foundations

2.1 Resources

3 Pointers

4 Arrays

5 Strings

6 Streams and IO

7 Abstract

8 Object-Oriented Programming

9 Dynamic Memory

10 Linked List

11 Iterators

12 Recursion

13 Function Objects, Functors

14 Error Handling & Exceptions

15 Stacks & Queues

16 Complexity Analysis

16.1 References

- [AlgoMonster - Runtime Summary](#)

16.2 Runtime Overview

16.3 Common Time Complexities

16.3.1 $O(1)$ - Constant

Constant time complexity. Could be

- Hashmap lookup
- Array access and update
- Pushing and popping elements from a stack
- Finding and applying math formula

16.3.2 $O(\log(N))$ - Logarithmic

$\log(N)$ grows very slowly

In coding interviews, $\log(N)$ typically means:

- Binary search or variant
- Balanced binary search tree lookup
- Processing the digits of a number

Unless specified, typically $\log(N)$ refers to $\log_2(N)$

Example C++:

```
int N = 100000000;
while (N > 0) {
    // some constant operation
    N /= 2;
}
```

Many mainstream relational databases use binary trees for indexing by default, thus lookup by primary key in a relational database is $\log(N)$.

16.3.3 $O(N)$ - Linear

Linear time typically means looping through a linear data structure a constant number of times. Most commonly, this means:

- Going through array/linked list
- Two pointers
- Some types of greedy
- Tree/graph traversal
- Stack/Queue

Example C++:

```
for (int i = 1; i <= N; i++) {
    // constant time code
}

for (int i = 1; i < 5 * N + 17; i++) {
    // constant time code
}

for (int i = 1; i < N + 538238; i++) {
    // constant time code
}
```

16.3.4 $O(K \log(N))$

- Heap push/pop K times. When you encounter problems that seek the “top K elements”, you can often solve them by pushing and popping to a heap K times, resulting in an $O(K \log(N))$ runtime. e.g., K closest points, merge K sorted lists.
- Binary search K times.

Since K is constant this kind of isn't its own time complexity and can be grouped with $O(\log(N))$

16.3.5 $O(N \log(N))$ - Log-Linear

- Sorting. The default sorting algorithm's expected runtime in all mainstream languages is $N \log(N)$. For example, java uses a variant of merge sort for object sorting and a variant of Quick Sort for primitive type sorting.
- Divide and conquer with a linear time merge operation. Divide is normally $\log(N)$, and if merge is $O(N)$ then the overall runtime is $O(N \log(N))$. An example problem is smaller numbers to the right.

16.3.6 $O(N^2)$ - Quadratic

- Nested loops, e.g., visiting each matrix entry
- Many brute force solutions

```
for (int i = 1; i <= N; i++) {  
    for (int j = 1; j <= N; j++) {  
        // constant time code  
    }  
}
```

16.3.7 $O(2^N)$ - Exponential

Grows very rapidly. Often requires memoization to avoid repeated computations and reduce complexity.

- Combinatorial problems, backtracking, e.g. subsets
- Often involves recursion and is harder to analyze time complexity at first sight

E.g.: A recursive Fibonacci algorithm is $O(2^N)$

```
int Fib(int n) {  
    if (n == 0 || n == 1) {  
        return 1;  
    }  
    return Fib(n - 1) + Fib(n - 2);  
}
```


16.3.8 $O(N!)$ - Factorial

Grows very very rapidly. Only solvable by computers for small N . Often requires memoization to avoid repeated computations and reduce complexity.

- Combinatorial problems, backtracking, e.g. permutations
- Often involves recursion and is harder to analyze time complexity at first sight

16.4 Amortized Time Complexity

17 STL

18 Heaps

18.1 References

- [AlgoMonster - Heap Fundamentals](#)

18.2 Heap? Priority Queue?

Priority Queue is an **Abstract Data Type**, and Heap is the concrete data structure we use to implement a priority queue. [Source](#)

18.3 Heap in C++

19 Trees

20 Searching

21 Sorting

22 Hash Maps

23 Graphs

24 Brute-Force & Greedy Algorithms

25 Divide and Conquer, Dynamic Programming

26 Backtracing, Branch and Bound Algorithms

27 Summary

In summary...

28 Tips on Solving DS&A Questions

28.1 Problem Solving Flowchart

28.2 DS&A Roadmap

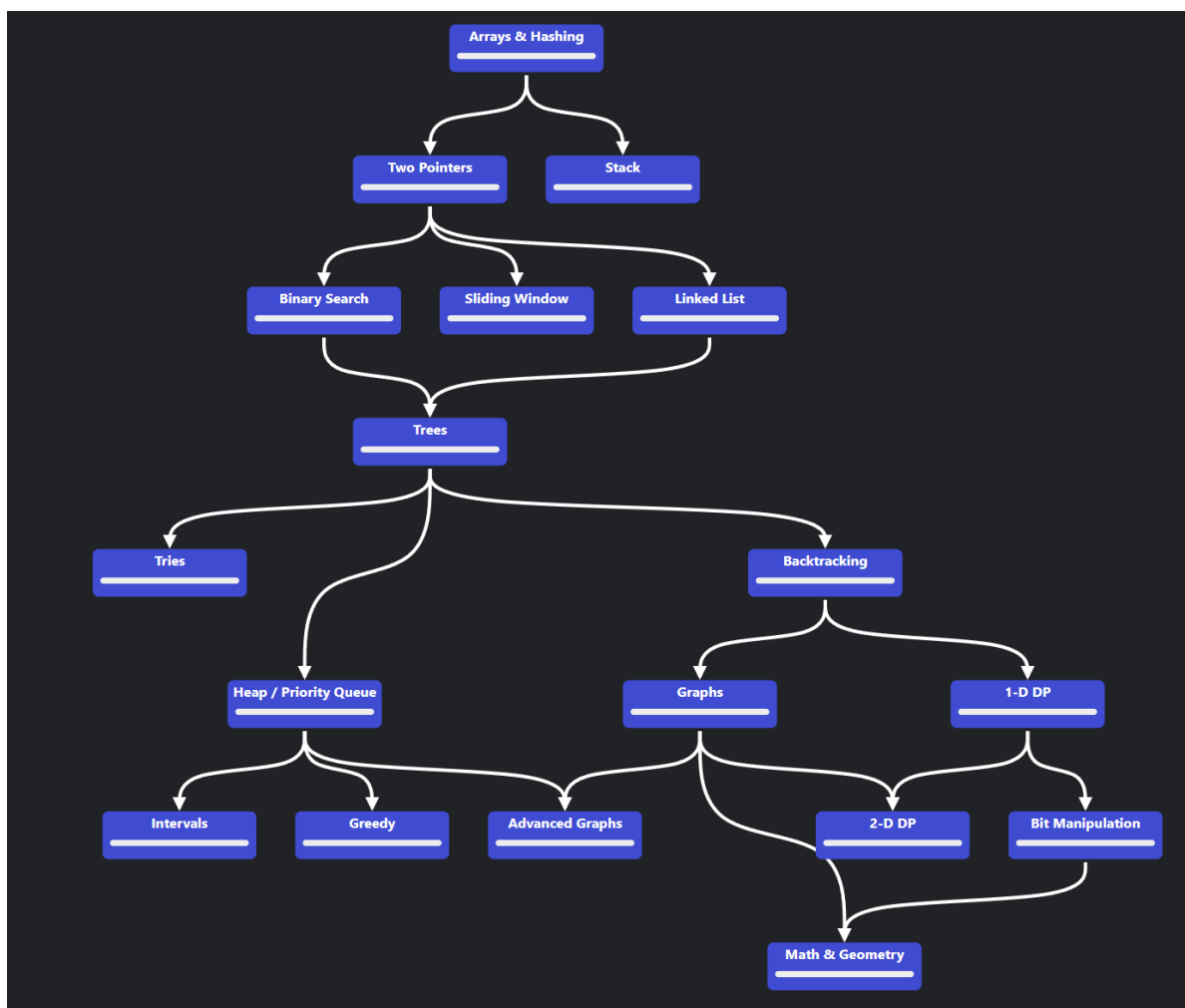


Figure 28.2: A Roadmap for studying. [Source](#)

```
If input array is sorted then
    - Binary search
    - Two pointers

If asked for all permutations/subsets then
    - Backtracking

If given a tree then
    - DFS
    - BFS

If given a graph then
    - DFS
    - BFS

If given a linked list then
    - Two pointers

If recursion is banned then
    - Stack

If must solve in-place then
    - Swap corresponding values
    - Store one or more different values in the same pointer

If asked for maximum/minimum subarray/subset/options then
    - Dynamic programming

If asked for top/least K items then
    - Heap
    - QuickSelect

If asked for common strings then
    - Map
    - Trie

Else
    - Map/Set for  $O(1)$  time &  $O(n)$  space
    - Sort input for  $O(n \log n)$  time and  $O(1)$  space
```

Figure 28.1: Tips on problem approach.[Image Source](#)

28.3 Problem Flowchart

28.4 ROI

28.5 “Academic” Algorithms

According to AlgoMonster, some **algorithms** that are very rarely/almost never asked in interviews:

- Minimal spanning tree: Kruskal’s algorithm and Prim’s algorithm
- Minimum cut: Ford-Fulkerson algorithm
- Shortest path in weight graphs: Bellman-Ford-Moore algorithm
- String search: Boyer-Moore algorithm

28.6 Keyword to Algo

[AlgoMonster](#) provides a convenient “Keyword to Algorithm” summary:

“Top k”

- Heap
 - E.g. K closest points

“How many ways..”

- DFS
 - E.g. Decode ways
- DP
 - E.g. Robot paths

“Substring”

- Sliding window
 - E.g. Longest substring without repeating characters

“Palindrome”

- two pointers: Valid Palindrome
- DFS: Palindrome Partitioning
- DP: Palindrome Partitioning II

Topic	Difficulty to Learn
Two Pointers	Easy
Sliding Window	Easy
Breadth-First Search	Easy
Depth-First Search	Medium
Backtracking	High
Heap	Medium
Binary Search	Easy
Dynamic Programming	High
Divide and Conquer	Medium
Trie	Medium
Union Find	Medium
Greedy	High

Figure 28.4: Studying to Maximizing ROI according to [AlgoMonster](#).

“Tree”

- shortest, level-order
 - BFS: Binary Tree Level-Order Traversal
- else: DFS: Max Depth

“Parentheses”

- Stack: Valid Parentheses

“Subarray”

- Sliding window: Maximum subarray sum
- Prefix sum: Subarray sum
- Hashmap: Continuous subarray sum

Max subarray

- Greedy: Kadane’s Algorithm

“X Sum”

- Two pointer: Two sum

“Max/longest sequence”

- Dynamic programming, DFS: Longest increasing subsequence
- mono deque: Sliding window maximum

“Minimum/Shortest”

- Dynamic programming, DFS: Minimal path sum
- BFS: Shortest path

“Partition/split ... array/string”

- DFS: Decode ways

“Subsequence”

- Dynamic programming, DFS: Longest increasing subsequence
- Sliding window: Longest increasing subsequence

“Matrix”

- BFS, DFS: Flood fill, Islands
- Dynamic programming: Maximal square

“Jump”

- Greedy/DP: Jump game

“Game”

- Dynamic programming: Divisor game, Stone game

“Connected component”, “Cut/remove” “Regions/groups/connections”

- Union Find: Number of connected components, Redundant connections

Transitive relationship

- If the items are related to one another and the relationship is transitive, then chances are we can build a graph and use BFS or Union Find.
 - string converting to another, BFS: Word Ladder
 - string converting to another, BFS, Union Find: Sentence Similarity
 - numbers having divisional relationship, BFS, Union Find: Evaluate Division

“Interval”

- Greedy: sort by start/end time and then go through sorted intervals Interval Pattern

29 Problems & Explanations