

DATA STRUCTURES AND ALGORITHMS MADE EASY

Data Structures and Algorithmic Puzzles

5TH
EDITION



Narasimha Karumanchi, M.Tech, IIT Bombay

Founder, CareerMonk.com

Data Structures And Algorithms Made Easy

-To All My Readers

**By
Narasimha Karumanchi**

 **Concepts**  **Problems**  **Interview Questions**

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Mother and Father, it is impossible to thank you adequately for everything you have done, from loving me unconditionally to raising me in a stable household, where your persistent efforts and traditional values taught your children to celebrate and embrace life. I could not have asked for better parents or role-models. You showed me that anything is possible with faith, hard work and determination.

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–*Narasimha Karumanchi*
M-Tech, IIT Bombay
Founder, [CareerMonk.com](https://www.careermonk.com)

Preface

Dear Reader,

Please hold on! I know many people typically do not read the Preface of a book. But I strongly recommend that you read this particular Preface.

It is not the main objective of this book to present you with the theorems and proofs on *data structures* and *algorithms*. I have followed a pattern of improving the problem solutions with different complexities (for each problem, you will find multiple solutions with different, and reduced, complexities). Basically, it's an enumeration of possible solutions. With this approach, even if you get a new question, it will show you a way to *think* about the possible solutions. You will find this book useful for interview preparation, competitive exams preparation, and campus interview preparations.

As a *job seeker*, if you read the complete book, I am sure you will be able to challenge the interviewers. If you read it as an *instructor*, it will help you to deliver lectures with an approach that is easy to follow, and as a result your students will appreciate the fact that they have opted for Computer Science / Information Technology as their degree.

This book is also useful for *Engineering degree students* and *Masters degree students* during their academic preparations. In all the chapters you will see that there is more emphasis on problems and their analysis rather than on theory. In each chapter, you will first read about the basic required theory, which is then followed by a section on problem sets. In total, there are approximately 700 algorithmic problems, all with solutions.

If you read the book as a *student* preparing for competitive exams for Computer Science / Information Technology, the content covers *all the required topics* in full detail. While writing this book, my main focus was to help students who are preparing for these exams.

In all the chapters you will see more emphasis on problems and analysis rather than on theory. In each chapter, you will first see the basic required theory followed by various problems.

For many problems, *multiple* solutions are provided with different levels of complexity. We start with the *brute force* solution and slowly move toward the *best solution* possible for that problem. For each problem, we endeavor to understand how much time the algorithm takes and how much memory the algorithm uses.

It is recommended that the reader does at least one *complete* reading of this book to gain a full understanding of all the topics that are covered. Then, in subsequent readings you can skip directly to any chapter to refer to a specific topic. Even though many readings have been done for the purpose of correcting errors, there could still be some minor typos in the book. If any are found, they will be updated at www.CareerMonk.com. You can monitor this site for any corrections and also for new problems and solutions. Also, please provide your valuable suggestions at: Info@CareerMonk.com.

I wish you all the best and I am confident that you will find this book useful.

–Narasimha Karumanchi
M-Tech, IIT Bombay
Founder, CareerMonk.com

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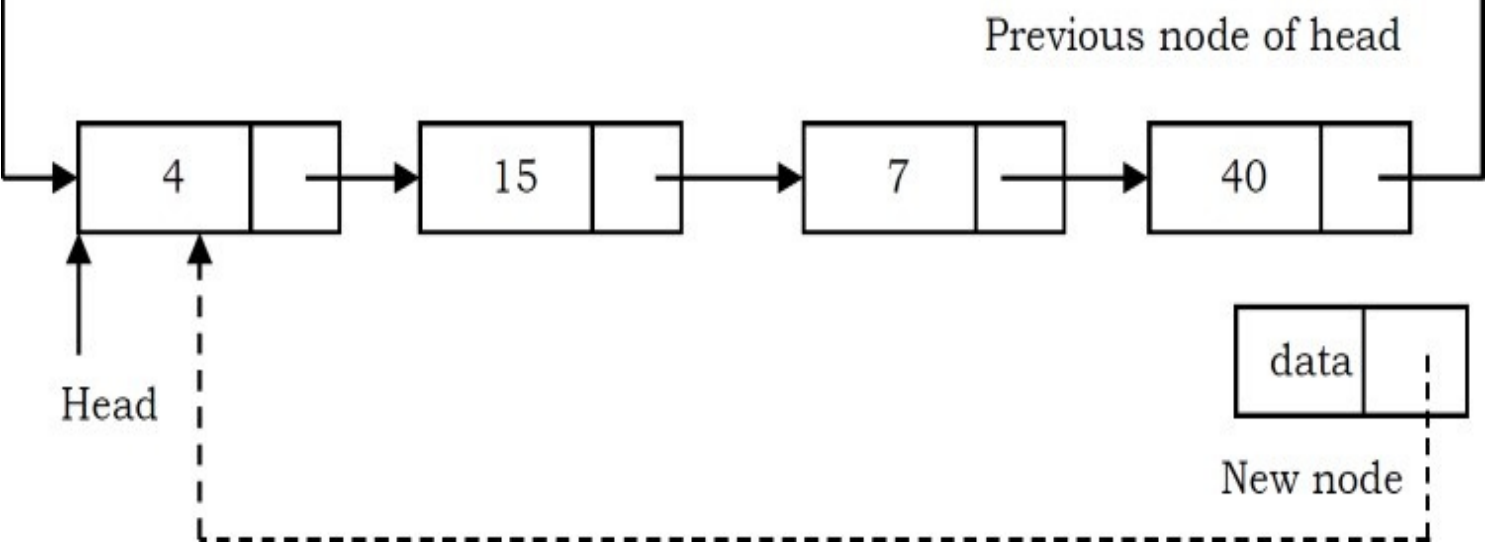
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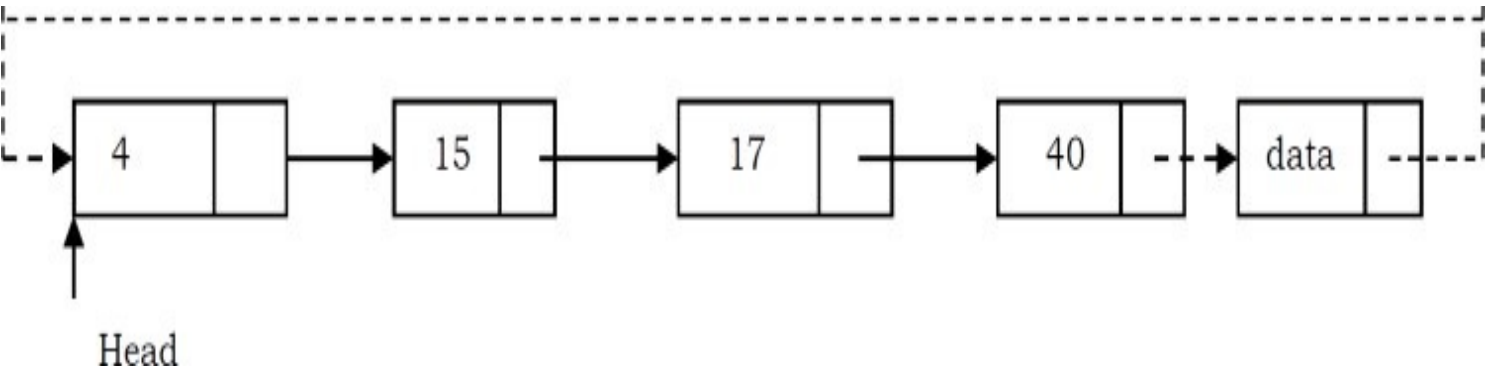
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References



- Update the next pointer of the previous node to point to the new node and we get the list as shown below.



```

struct BinaryTreeNode *DeepestNodeInBinaryTree(struct BinaryTreeNode *root){
    struct BinaryTreeNode *temp;
    struct Queue *Q;
    if(!root) return NULL;
    Q = CreateQueue();
    EnQueue(Q,root);
    while(!IsEmptyQueue(Q)) {
        temp = DeQueue(Q);
        if(temp→left)
            EnQueue(Q, temp→left);
        if(temp→right)
            EnQueue(Q, temp→right);
    }
    DeleteQueue(Q);
    return temp;
}

```

Time Complexity: $O(n)$. Space Complexity: $O(n)$.

Problem-13 Give an algorithm for deleting an element (assuming data is given) from binary tree.

Solution: The deletion of a node in binary tree can be implemented as

- Starting at root, find the node which we want to delete.
- Find the deepest node in the tree.
- Replace the deepest node's data with node to be deleted.
- Then delete the deepest node.

Problem-14 Give an algorithm for finding the number of leaves in the binary tree without using recursion.

Solution: The set of nodes whose both left and right children are NULL are called leaf nodes.

Window position	Max
[1 3 -1] -3 5 3 6 7	3
1 [3 -1 -3] 5 3 6 7	3
1 3 [-1 -3 5] 3 6 7	5
1 3 -1 [-3 5 3] 6 7	5
1 3 -1 -3 [5 3 6] 7	6
1 3 -1 -3 5 [3 6 7]	7

Input: A long array A[], and a window width w. **Output:** An array B[], B[i] is the maximum value of from A[i] to A[i+w-1]

Requirement: Find a good optimal way to get B[i]

Solution: Brute force solution is, every time the window is moved we can search for a total of w elements in the window.

Time complexity: $O(nw)$.

Problem-29 For [Problem-28](#), can we reduce the complexity?

Solution: Yes, we can use heap data structure. This reduces the time complexity to $O(n\log w)$. Insert operation takes $O(\log w)$ time, where w is the size of the heap. However, getting the maximum value is cheap; it merely takes constant time as the maximum value is always kept in the root (head) of the heap. As the window slides to the right, some elements in the heap might not be valid anymore (range is outside of the current window). How should we remove them? We would need to be somewhat careful here. Since we only remove elements that are out of the window's range, we would need to keep track of the elements' indices too.

Problem-30 For [Problem-28](#), can we further reduce the complexity?

Solution: Yes, The double-ended queue is the perfect data structure for this problem. It supports insertion/deletion from the front and back. The trick is to find a way such that the largest element in the window would always appear in the front of the queue. How would you maintain this requirement as you push and pop elements in and out of the queue?

Besides, you will notice that there are some redundant elements in the queue that we shouldn't even consider. For example, if the current queue has the elements: [10 5 3], and a new element in the window has the element 11. Now, we could have emptied the queue without considering elements 10, 5, and 3, and insert only element 11 into the queue.

- For each element of the input array, insert into the hash table. Let us say the current element is $A[X]$.
- Check whether there exists a hash entry in the table with key: $K - A[X]$.
- If such element exists then scan the element pairs of $K - A[X]$ and return all possible pairs by including $A[X]$ also.
- If no such element exists (with $K - A[X]$ as key) then go to next element.

Time Complexity: The time for storing all possible pairs in Hash table + searching = $O(n^2) + O(n^2) \approx O(n^2)$. Space Complexity: $O(n)$.

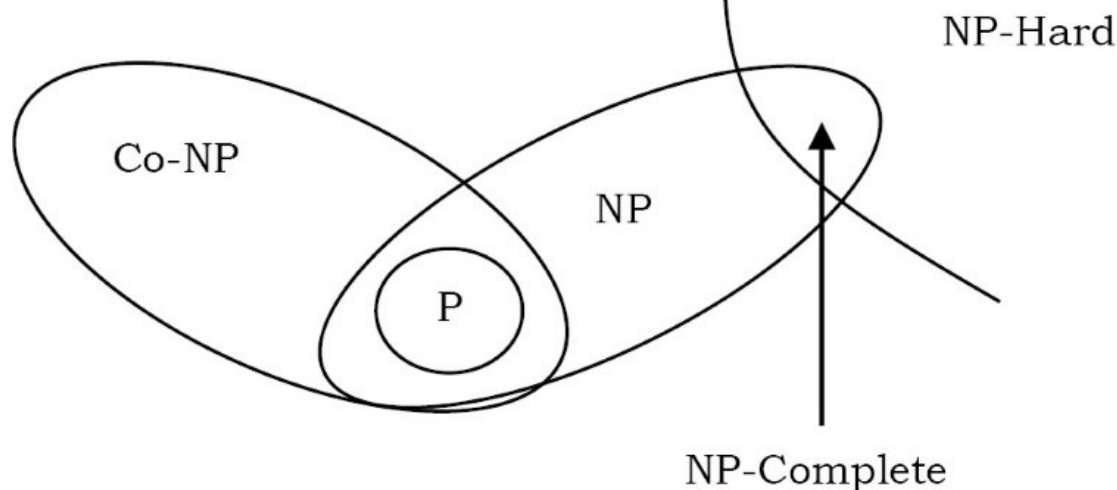
Problem-36 Given an array of n integers, the 3 – sum problem is to find three integers whose sum is closest to zero.

Solution: This is the same as that of [Problem-32](#) with K value is zero.

Problem-37 Let A be an array of n distinct integers. Suppose A has the following property: there exists an index $1 \leq k \leq n$ such that $A[1], \dots, A[k]$ is an increasing sequence and $A[k + 1], \dots, A[n]$ is a decreasing sequence. Design and analyze an efficient algorithm for finding k .

Similar question: Let us assume that the given array is sorted but starts with negative numbers and ends with positive numbers [such functions are called monotonically increasing functions]. In this array find the starting index of the positive numbers. Assume that we know the length of the input array. Design a $O(\log n)$ algorithm.

Solution: Let us use a variant of the binary search.



The set of problems that are *NP*-hard is a strict superset of the problems that are *NP*-complete. Some problems (like the halting problem) are *NP*-hard, but not in *NP*. *NP*-hard problems might be impossible to solve in general. We can tell the difference in difficulty between *NP*-hard and *NP*-complete problems because the class *NP* includes everything easier than its “toughest” problems - if a problem is not in *NP*, it is harder than all the problems in *NP*.

Does $P=NP$?

If $P = NP$, it means that every problem that can be checked quickly can be solved quickly (remember the difference between checking if an answer is right and actually solving a problem).

This is a big question (and nobody knows the answer), because right now there are lots of *NP*-complete problems that can’t be solved quickly. If $P = NP$, that means there is a way to solve them fast. Remember that “quickly” means not trial-and-error. It could take a billion years, but as long as we didn’t use trial and error, it was quick. In future, a computer will be able to change that billion years into a few minutes.

20.7 Reductions

Before discussing reductions, let us consider the following scenario. Assume that we want to solve problem *X* but feel it’s very complicated. In this case what do we do?

The first thing that comes to mind is, if we have a similar problem to that of *X* (let us say *Y*), then we try to map *X* to *Y* and use *Y*’s solution to solve *X* also. This process is called reduction.