



BY LEARNSTACK •

2026 Edition

# DSA Cheat Sheet

Complete Data Structures & Algorithms  
Reference for Placements & Interviews

Arrays

Trees

Graphs

DP

Sorting

Recursion

Hashing

*Learn. Build. Grow.*



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# About This Handbook

LearnStack DSA Complete Handbook 2026 is a premium, placement-focused reference for mastering Data Structures and Algorithms from fundamentals to advanced interview patterns. It is designed for CSE/IT students, developers refreshing DSA, self-learners preparing for product-company interviews, and competitive coding aspirants.

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INFO: What makes it different

- Every topic has working Python code.
- Interview patterns are highlighted.
- Every algorithm includes time and space complexity.
- 100+ solved problems are included for revision and practice.

TIP: How to use this handbook

- Beginners should read chapters 1-6 linearly.
- Use chapters 11-16 as focused revision blocks.
- Use chapter 17 before interviews as a solved-problem bank.
- Use chapter 19 as a final-day quick reference.

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# CHAPTER 1 - FOUNDATIONS OF DSA

## CHAPTER 1 FOUNDATIONS OF DSA Roadmap, language choice, setup, and interview mindset

### 1.1 What is DSA?

#### INFO: Definition

Data Structures are ways of organizing data. Algorithms are step-by-step procedures to solve problems. DSA matters because it controls how efficiently software stores, finds, transforms, and connects information.

- Used in databases, operating systems, compilers, search engines, APIs, and apps.
- Core of technical interviews and competitive programming.
- Improves problem-solving speed and code quality.

### 1.2 DSA Learning Roadmap

Step	Topic	Goal
1	Time & Space Complexity	Predict performance before coding
2	Arrays & Strings	Master indexing, two pointers, sliding window
3	Linked Lists	Pointers, reversal, fast/slow patterns
4	Stack & Queue	LIFO/FIFO and expression/traversal problems
5	Trees & BST	Recursive structures and search properties
6	Graphs	BFS, DFS, shortest paths, components
7	Sorting & Searching	Core algorithms and binary search patterns
8	Recursion & Backtracking	Explore solution spaces safely
9	Dynamic Programming	Optimize repeated subproblems
10	Advanced DS	Trie, DSU, segment tree, monotonic stack

### 1.3 Choosing a Language

Language	Pros	Cons	Best For
Python	Clean syntax, fast learning	Slower than C++	Beginners and interviews
Java	Strong OOP, common in campuses	Verbose	Campus placements
C++	Fast and STL-rich	Complex memory rules	Competitive coding
JavaScript	Flexible and popular	Less DS-focused	Web developers

#### TIP: Language tip

Use Python to understand DSA concepts quickly. Switch to C++ when contest speed and STL control matter.

### 1.4 Setup

## First DSA Program

INFO: What it does

Install Python, use VS Code or Jupyter, and start by writing small functions that can be tested immediately.

Python

```
# Terminal setup
# python --version
# pip install notebook

print('Hello DSA!')
```

Output: Hello DSA!

COMPLEXITY

Time:  $O(1)$  | Space:  $O(1)$

TIP

Keep each algorithm in a function so you can test it with multiple inputs.

CAUTION

Do not copy code without tracing it once.

# CHAPTER 2 - TIME & SPACE COMPLEXITY

## CHAPTER 2 TIME & SPACE COMPLEXITY Big O, space usage, and analysis habits

### INFO: Why complexity matters

Complexity tells how an algorithm grows when input size grows. In interviews, a correct but slow solution often needs optimization before it is accepted.

Notation	Name	Example	Speed
$O(1)$	Constant	Array access	Best
$O(\log n)$	Logarithmic	Binary search	Fast
$O(n)$	Linear	Single loop	OK
$O(n \log n)$	Linearithmic	Merge sort	Good for sorting
$O(n^2)$	Quadratic	Nested loops	Slow
$O(2^n)$	Exponential	Brute recursion	Avoid
$O(n!)$	Factorial	Permutations	Avoid

### $O(1)$ - Constant

#### INFO: What it does

Array access does not depend on input size.

Python

```
def first(arr):
    return arr[0] if arr else None
```

Output: Sample function returns the expected value.

#### COMPLEXITY

Time:  $O(1)$  | Space:  $O(1)$

#### TIP

Drop constants and lower-order terms in Big O.

#### CAUTION

Big O usually means worst case unless the interviewer asks otherwise.

## O(log n) - Logarithmic

INFO: What it does

Each step cuts the search space in half.

Python

```
def binary_search(arr, target):
    left, right = 0, len(arr) - 1
    while left <= right:
        mid = left + (right - left) // 2
        if arr[mid] == target:
            return mid
        if arr[mid] < target:
            left = mid + 1
        else:
            right = mid - 1
    return -1

print(binary_search([1, 3, 5, 7], 5))
```

Output: Sample function returns the expected value.

COMPLEXITY

Time: O(log n) | Space: O(1)

TIP

Drop constants and lower-order terms in Big O.

CAUTION

Big O usually means worst case unless the interviewer asks otherwise.

## O(n) - Linear

INFO: What it does

A single loop visits each element once.

Python

```
def linear_search(arr, target):
    for i, x in enumerate(arr):
        if x == target:
            return i
    return -1

print(linear_search([10, 20, 30], 20)) # 1
```

Output: Sample function returns the expected value.

COMPLEXITY

Time: O(n) | Space: O(1)

TIP

Drop constants and lower-order terms in Big O.

CAUTION

Big O usually means worst case unless the interviewer asks otherwise.

## $O(n^2)$ - Quadratic

INFO: What it does  
Nested loops compare pairs.

Python

```
def all_pairs(arr):
    ans = []
    for i in arr:
        for j in arr:
            ans.append((i, j))
    return ans
```

Output: Sample function returns the expected value.

COMPLEXITY  
Time:  $O(n^2)$  | Space:  $O(n^2)$

TIP  
Drop constants and lower-order terms in Big O.

CAUTION  
Big O usually means worst case unless the interviewer asks otherwise.

## $O(2^n)$ - Exponential

INFO: What it does  
Each recursive call branches into two calls.

Python

```
def fib_slow(n):
    if n <= 1: return n
    return fib_slow(n-1) + fib_slow(n-2)
```

Output: Sample function returns the expected value.

COMPLEXITY  
Time:  $O(2^n)$  | Space:  $O(n)$

TIP  
Drop constants and lower-order terms in Big O.

CAUTION  
Big O usually means worst case unless the interviewer asks otherwise.

## 2.5 Space Complexity

INFO: Space complexity  
Space complexity measures extra memory used by the algorithm excluding the input unless stated otherwise.  
 $O(1)$  means constant extra variables,  $O(n)$  means memory proportional to input size.

Python

```
def copy_array(arr):
    copy = arr[:] #  $O(n)$  extra space
    return copy

def in_place_increment(arr):
    for i in range(len(arr)):
        arr[i] += 1 #  $O(1)$  extra space
    return arr
```

INTERVIEW: Interview rule

Mention both time and space complexity after every solution. Explain why, not only the final notation.

## 2.6 Best, Average, Worst Case

Algorithm	Best	Average	Worst
Linear Search	$O(1)$	$O(n)$	$O(n)$
Binary Search	$O(1)$	$O(\log n)$	$O(\log n)$
Bubble Sort	$O(n)$	$O(n^2)$	$O(n^2)$
Quick Sort	$O(n \log n)$	$O(n \log n)$	$O(n^2)$
Merge Sort	$O(n \log n)$	$O(n \log n)$	$O(n \log n)$

# CHAPTER 3 - ARRAYS

## CHAPTER 3 ARRAYS

Contiguous storage, patterns, and interview problems

### INFO: Array definition

An array stores elements in indexed order. Python lists are dynamic arrays; they support  $O(1)$  average append and  $O(1)$  index access.

Index: [0] [1] [2] [3] [4]

Value: [10][20][30][40][50]

Operation	Description	Time	Space
Access	arr[i]	$O(1)$	$O(1)$
Search	Find element	$O(n)$	$O(1)$
Insert end	append	$O(1)$ avg	$O(1)$
Insert middle	Shift elements	$O(n)$	$O(1)$
Delete	Shift elements	$O(n)$	$O(1)$
Traverse	Visit all	$O(n)$	$O(1)$
Update	arr[i]=val	$O(1)$	$O(1)$

## Declaration and Traversal

### INFO: What it does

Create, access, update, and loop through a Python list.

Python

```
arr = [10, 20, 30]
arr[1] = 99
for i, value in enumerate(arr):
    print(i, value)
```

Output: Returned value depends on the input; sample calls are included where helpful.

### COMPLEXITY

Time:  $O(n)$  | Space:  $O(1)$

### TIP

Trace the algorithm on a small example before coding.

### CAUTION

Handle empty input and boundary values first.

## Linear Search

INFO: What it does  
Scan until target is found.

Python

```
def linear_search(arr, target):
    for i, x in enumerate(arr):
        if x == target:
            return i
    return -1

print(linear_search([10, 20, 30], 20)) # 1
```

Output: Returned value depends on the input; sample calls are included where helpful.

COMPLEXITY  
Time:  $O(n)$  | Space:  $O(1)$

TIP  
Trace the algorithm on a small example before coding.

CAUTION  
Handle empty input and boundary values first.

## Find Maximum and Minimum

INFO: What it does  
Track two values while traversing once.

Python

```
def max_min(arr):
    if not arr:
        return None, None
    mn = mx = arr[0]
    for x in arr[1:]:
        mn = min(mn, x)
        mx = max(mx, x)
    return mn, mx

print(max_min([4, 1, 9, 2]))
```

Output: Returned value depends on the input; sample calls are included where helpful.

COMPLEXITY  
Time:  $O(n)$  | Space:  $O(1)$

TIP  
Trace the algorithm on a small example before coding.

CAUTION  
Handle empty input and boundary values first.

## Reverse an Array

INFO: What it does

Swap two ends and move inward.

Python

```
def reverse_array(arr):
    left, right = 0, len(arr) - 1
    while left < right:
        arr[left], arr[right] = arr[right], arr[left]
        left += 1
        right -= 1
    return arr

print(reverse_array([1, 2, 3, 4]))
```

Output: Returned value depends on the input; sample calls are included where helpful.

COMPLEXITY

Time:  $O(n)$  | Space:  $O(1)$

TIP

Trace the algorithm on a small example before coding.

CAUTION

Handle empty input and boundary values first.

## Rotate Array

INFO: What it does

Normalize k and rebuild using slicing.

Python

```
def rotate_right(arr, k):
    n = len(arr)
    if n == 0: return arr
    k %= n
    return arr[-k:] + arr[:-k]

print(rotate_right([1, 2, 3, 4, 5], 2))
```

Output: Returned value depends on the input; sample calls are included where helpful.

COMPLEXITY

Time:  $O(n)$  | Space:  $O(n)$

TIP

Trace the algorithm on a small example before coding.

CAUTION

Handle empty input and boundary values first.

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