

Time & Space Complexity Cheat Sheet

BIG O NOTATION HIERARCHY (Fastest to Slowest)

$O(1) < O(\log n) < O(\sqrt{n}) < O(n) < O(n \log n) < O(n^2) < O(n^3) < O(2^n) < O(n!)$

Constant < Logarithmic < Root < Linear < Linearithmic < Quadratic < Cubic < Exponential

ACCEPTABLE TIME COMPLEXITIES FOR HACKERRANK

For $n = 10^5$ (typical Citadel constraint):

| Complexity | Max n | Example |
|---------------|-----------|-------------------------------------|
| $O(1)$ | Any | Hash lookup, array access |
| $O(\log n)$ | 10^{18} | Binary search |
| $O(\sqrt{n})$ | 10^{14} | Prime checking |
| $O(n)$ | 10^8 | Single loop |
| $O(n \log n)$ | 10^6 | Sorting, heap operations |
| $O(n^2)$ | 10^4 | Nested loops (DANGER for $n=10^5$) |
| $O(n^3)$ | 500 | Triple nested loops |
| $O(2^n)$ | 20 | Subset generation |

For 75-minute test with $n=10^5$:

- **SAFE:** $O(n)$, $O(n \log n)$
- **RISKY:** $O(n^2)$ will likely timeout
- **NO GO:** $O(n^3)$ or worse

PYTHON DATA STRUCTURES COMPLEXITY

List (Array)

| Operation | Average | Worst | Notes |
|----------------------------------|---------------|---------------|--------------------|
| <code>arr[i]</code> | $O(1)$ | $O(1)$ | Index access |
| <code>arr.append(x)</code> | $O(1)$ | $O(1)$ | Add to end |
| <code>arr.insert(i, x)</code> | $O(n)$ | $O(n)$ | Insert at position |
| <code>arr.pop()</code> | $O(1)$ | $O(1)$ | Remove last |
| <code>arr.pop(i)</code> | $O(n)$ | $O(n)$ | Remove at position |
| <code>arr.remove(x)</code> | $O(n)$ | $O(n)$ | Remove by value |
| <code>x in arr</code> | $O(n)$ | $O(n)$ | Search |
| <code>arr.sort()</code> | $O(n \log n)$ | $O(n \log n)$ | In-place sort |
| <code>sorted(arr)</code> | $O(n \log n)$ | $O(n \log n)$ | New sorted list |
| <code>arr.reverse()</code> | $O(n)$ | $O(n)$ | In-place reverse |
| <code>arr[::-1]</code> | $O(n)$ | $O(n)$ | New reversed list |
| <code>min(arr) , max(arr)</code> | $O(n)$ | $O(n)$ | Find min/max |
| Slicing <code>arr[a:b]</code> | $O(b-a)$ | $O(b-a)$ | Create sublist |

Dictionary (Hash Table)

| Operation | Average | Worst | Notes |
|-----------------------------|---------|--------|-----------------|
| <code>d[key]</code> | $O(1)$ | $O(n)$ | Access |
| <code>d[key] = value</code> | $O(1)$ | $O(n)$ | Insert/Update |
| <code>del d[key]</code> | $O(1)$ | $O(n)$ | Delete |
| <code>key in d</code> | $O(1)$ | $O(n)$ | Check existence |
| Iteration | $O(n)$ | $O(n)$ | All keys/values |

Set

| Operation | Average | Worst | Notes |
|--------------------------|---|--------|------------------|
| <code>s.add(x)</code> | $O(1)$ | $O(n)$ | Add element |
| <code>s.remove(x)</code> | $O(1)$ | $O(n)$ | Remove element |
| <code>x in s</code> | $O(1)$ | $O(n)$ | Check membership |
| <code>s1 s2</code> | $O(\text{len}(s1) + \text{len}(s2))$ | | Union |
| <code>s1 & s2</code> | $O(\min(\text{len}(s1), \text{len}(s2)))$ | | Intersection |
| <code>s1 - s2</code> | $O(\text{len}(s1))$ | | Difference |

Deque (Double-ended Queue)

| Operation | Complexity | Notes |
|----------------------------|------------|-----------------------|
| <code>append(x)</code> | $O(1)$ | Add to right |
| <code>appendleft(x)</code> | $O(1)$ | Add to left |
| <code>pop()</code> | $O(1)$ | Remove from right |
| <code>popleft()</code> | $O(1)$ | Remove from left |
| <code>d[i]</code> | $O(n)$ | Random access (slow!) |

Heap (Priority Queue)

| Operation | Complexity | Notes |
|-----------------------------|-------------|----------------------|
| <code>heappush(h, x)</code> | $O(\log n)$ | Insert |
| <code>heappop(h)</code> | $O(\log n)$ | Remove min |
| <code>h[0]</code> | $O(1)$ | Peek min |
| <code>heapify(arr)</code> | $O(n)$ | Build heap from list |

Counter

| Operation | Complexity | Notes |
|-----------|------------|-------|
|-----------|------------|-------|

| | | |
|-----------------------------|---------------|--------------------|
| Creation | $O(n)$ | Count all elements |
| <code>count[x]</code> | $O(1)$ | Get count |
| <code>most_common(k)</code> | $O(n \log k)$ | Top k elements |

COMMON ALGORITHM COMPLEXITIES

Sorting Algorithms

```
# Built-in sort -  $O(n \log n)$  time,  $O(n)$  space
arr.sort() # Timsort
sorted(arr)

# Counting sort -  $O(n + k)$  where k is range
# Only for integers in limited range
def counting_sort(arr, max_val):
    count = [0] * (max_val + 1)
    for num in arr:
        count[num] += 1

    result = []
    for num, freq in enumerate(count):
        result.extend([num] * freq)
    return result
```

Search Algorithms

```
# Linear search -  $O(n)$ 
def linear_search(arr, target):
    for i, val in enumerate(arr):
        if val == target:
            return i
    return -1

# Binary search -  $O(\log n)$ 
def binary_search(arr, target):
    left, right = 0, len(arr) - 1
    while left <= right:
        mid = (left + right) // 2
        if arr[mid] == target:
            return mid
        elif arr[mid] < target:
```

```

        left = mid + 1
    else:
        right = mid - 1
    return -1

```

Graph Algorithms

```

# BFS -  $O(V + E)$  time,  $O(V)$  space
# DFS -  $O(V + E)$  time,  $O(V)$  space
# Dijkstra -  $O((V + E) \log V)$  with heap
# Bellman-Ford -  $O(VE)$ 
# Floyd-Warshall -  $O(V^3)$ 
# Kruskal's MST -  $O(E \log E)$ 
# Prim's MST -  $O(E \log V)$  with heap
# Topological Sort -  $O(V + E)$ 

```

Tree Algorithms

```

# Tree traversal -  $O(n)$  time,  $O(h)$  space
# BST search -  $O(h)$  average,  $O(n)$  worst
# BST insert -  $O(h)$  average,  $O(n)$  worst
# Balanced tree operations -  $O(\log n)$ 

```

String Algorithms

```

# Pattern matching (naive) -  $O(nm)$ 
# KMP pattern matching -  $O(n + m)$ 
# Rabin-Karp -  $O(n + m)$  average
# String comparison -  $O(\min(n, m))$ 
# Substring search -  $O(n)$ 

```

SPACE COMPLEXITY

Common Space Patterns

```

#  $O(1)$  - Constant space
def constant_space(arr):
    result = 0
    for num in arr:
        result += num
    return result

```

```

# O(n) - Linear space
def linear_space(arr):
    return arr[:] # Copy array

# O(n) - Hash table for frequency
def frequency_count(arr):
    freq = {} # O(n) space
    for num in arr:
        freq[num] = freq.get(num, 0) + 1
    return freq

# O(h) - Recursion depth for tree
def tree_height(root):
    if not root:
        return 0
    return 1 + max(tree_height(root.left), tree_height(root.right))

# O(2^n) - All subsets
def all_subsets(arr):
    result = []

    def backtrack(index, current):
        if index == len(arr):
            result.append(current[:])
            return

        # Include current element
        current.append(arr[index])
        backtrack(index + 1, current)
        current.pop()

        # Exclude current element
        backtrack(index + 1, current)

    backtrack(0, [])
    return result

```

OPTIMIZATION TECHNIQUES

1. Use Hash Table for O(1) Lookup

```

# SLOW - O(n^2)
for num in arr1:
    if num in arr2: # O(n) search

```

```

        result.append(num)

# FAST - O(n)
set2 = set(arr2) # O(n) to build
for num in arr1: # O(n)
    if num in set2: # O(1) lookup
        result.append(num)

```

2. Avoid Repeated Calculations

```

# SLOW - O(n²)
for i in range(n):
    total = sum(arr[:i]) # Recalculates sum each time

# FAST - O(n)
prefix = [0]
for num in arr:
    prefix.append(prefix[-1] + num)

```

3. Two Pointers Instead of Nested Loops

```

# SLOW - O(n²)
for i in range(n):
    for j in range(i+1, n):
        if arr[i] + arr[j] == target:
            return [i, j]

# FAST - O(n) with sorted array
left, right = 0, n-1
while left < right:
    total = arr[left] + arr[right]
    if total == target:
        return [left, right]
    elif total < target:
        left += 1
    else:
        right -= 1

```

4. Sliding Window Instead of Recalculating

```

# SLOW - O(n²)
for i in range(n - k + 1):
    window_sum = sum(arr[i:i+k]) # O(k) each time

```

```
# FAST - O(n)
window_sum = sum(arr[:k])
for i in range(k, n):
    window_sum += arr[i] - arr[i-k] # O(1) update
```

5. Use Deque for Queue Operations

```
# SLOW - O(n2) because list.pop(0) is O(n)
queue = []
queue.append(x)
x = queue.pop(0) # O(n)

# FAST - O(1) for all operations
from collections import deque
queue = deque()
queue.append(x)
x = queue.popleft() # O(1)
```

6. Binary Search Instead of Linear Search

```
# SLOW - O(n)
for i, val in enumerate(sorted_arr):
    if val == target:
        return i

# FAST - O(log n)
left, right = 0, len(sorted_arr) - 1
while left <= right:
    mid = (left + right) // 2
    if sorted_arr[mid] == target:
        return mid
    elif sorted_arr[mid] < target:
        left = mid + 1
    else:
        right = mid - 1
```

COMMON MISTAKES TO AVOID

1. Nested Loops on Large Input

```
# TIMEOUT for n=105
```



```

for i in range(n):
    for j in range(n): # O(n²)
        process(i, j)

```

2. Sorting Inside Loop

```

# TIMEOUT - O(n² log n)
for i in range(n):
    sorted_sub = sorted(arr[:i]) # Sort every iteration

```

3. String Concatenation in Loop

```

# SLOW - O(n²) because strings are immutable
result = ""
for char in chars:
    result += char # Creates new string each time

# FAST - O(n)
result = ''.join(chars)

```

4. Using List for Frequent Membership Testing

```

# SLOW - O(n) per lookup
seen = []
for num in arr:
    if num not in seen: # O(n)
        seen.append(num)

# FAST - O(1) per lookup
seen = set()
for num in arr:
    if num not in seen: # O(1)
        seen.add(num)

```

5. Unnecessary Deep Copies

```

# SLOW - Copies entire array each time
def backtrack(arr):
    if condition:
        result.append(arr[:]) # OK - need copy here

    for i in range(len(arr)):

```

```

    new_arr = arr[:] # BAD - unnecessary copy
    new_arr[i] = x
    backtrack(new_arr)

# FAST - Modify in place
def backtrack(arr):
    if condition:
        result.append(arr[:])

    for i in range(len(arr)):
        old_val = arr[i]
        arr[i] = x # Modify
        backtrack(arr)
        arr[i] = old_val # Restore

```

QUICK COMPLEXITY CHECKS




Before implementing, ask:

1. **What's n?** (array length, string length, etc.)
2. **How many nested loops?** Each adds $O(n)$
3. **Am I sorting?** That's $O(n \log n)$
4. **Am I using hash table?** Lookups are $O(1)$
5. **Am I searching unsorted array?** That's $O(n)$
6. **Will this timeout?**
 - $n=10^5$ and $O(n^2) \rightarrow$ YES
 - $n=10^5$ and $O(n \log n) \rightarrow$ NO
 - $n=10^3$ and $O(n^2) \rightarrow$ NO

RULE OF THUMB FOR CITADEL

- **$n \leq 10$:** $O(n!)$ is acceptable (brute force permutations)
- **$n \leq 20$:** $O(2^n)$ is acceptable (subset enumeration)
- **$n \leq 500$:** $O(n^3)$ is acceptable
- **$n \leq 10^4$:** $O(n^2)$ is acceptable
- **$n \leq 10^5$:** $O(n \log n)$ or $O(n)$ required

- $n \leq 10^6$: $O(n)$ or $O(\log n)$ required

For $n=10^5$ (most Citadel problems):  $O(n)$, $O(n \log n)$  $O(n^2)$ - likely timeout 
 $O(n^3)$ or worse - guaranteed timeout