

Advanced Time and Space Complexity Cheat Sheet

Time Complexity Classes

$O(1)$: Constant Accessing array index, pushing/popping from a stack
 $O(\log n)$: Logarithmic Binary search
 $O(n)$: Linear Linear search, traversing array
 $O(n \log n)$: Linearithmic Merge sort, quick sort average
 $O(n^2)$: Quadratic Bubble sort, selection sort, nested loops
 $O(2^n)$: Exponential Recursive Fibonacci
 $O(n!)$: Factorial Solving permutations, traveling salesman brute force

Space Complexity Examples

$O(1)$: No extra space in-place sorting
 $O(n)$: Extra array, hash table, recursion stack with depth n
 $O(\log n)$: Recursive binary search call stack

Common Data Structures Time & Space

Array: Access $O(1)$, Insert/Delete $O(n)$, Space $O(n)$
Stack/Queue: Push/Pop $O(1)$, Space $O(n)$
Hash Map: Insert/Search/Delete $O(1)$ average, $O(n)$ worst, Space $O(n)$
Set: Same as hash map
Linked List: Access $O(n)$, Insert/Delete $O(1)$ if node given, Space $O(n)$
Binary Search Tree: Search/Insert/Delete $O(\log n)$ avg, $O(n)$ worst, Space $O(n)$
Heap: Insert/Delete $O(\log n)$, Access max/min $O(1)$, Space $O(n)$
Trie: Insert/Search $O(L)$, Space $O(\text{ALPHABET_SIZE} * L)$
Graph (Adj. List): Space $O(V + E)$, DFS/BFS $O(V + E)$

Tips for Analyzing Complexities

1. Loops: Linear ($O(n)$), Nested = Multiplicative ($O(n^2)$, etc.)
2. Recursive Calls: Use recurrence relations (e.g. $T(n) = 2T(n/2) + O(n)$)
3. Divide and Conquer: Often $O(n \log n)$
4. Hashing: Watch for space complexity and collisions
5. Recursion Stack: Adds implicit space usage
6. Dynamic Programming: Usually $O(n)$ or $O(n^2)$ in time and space unless optimized

Common Pitfalls

- Ignoring recursive stack space
- Assuming hash operations are always $O(1)$

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- Forgetting amortized cost (e.g., Python list append is $O(1)$ amortized)
- Misjudging input size impact n vs. n^2 vs. 2^n