

Python ICPC Cheatsheet

Comprehensive Reference for Competitive Programming

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1 Input/Output

Description: Efficient input/output is crucial in competitive programming, especially for problems with large datasets. Using `sys.stdin.readline` is significantly faster than the default `input()` function.

```
1 # Fast I/O - Essential for large inputs
2 import sys
3 input = sys.stdin.readline
4
5 # Read single integer
6 n = int(input())
7
8 # Read multiple integers on one line
9 a, b = map(int, input().split())
10
11 # Read array of integers
12 arr = list(map(int, input().split()))
13
14 # Read strings (strip to remove trailing newline)
15 s = input().strip()
16 words = input().split()
17
18 # Multiple test cases pattern
19 t = int(input())
20 for _ in range(t):
21     # process each test case
22
23 # Print without newline
24 print(x, end=' ')
25
26 # Formatted output with precision
27 print(f"{x:.6f}") # 6 decimal places
```

2 Basic Data Structures

2.1 List Operations

Description: Python lists are dynamic arrays with $O(1)$ amortized append and $O(n)$ insert/delete at arbitrary positions.

```
1 # Initialize lists
2 arr = [0] * n # n zeros
3 matrix = [[0] * m for _ in range(n)] # Correct way!
4
5 # List comprehension - concise and efficient
6 squares = [x**2 for x in range(n)]
7 evens = [x for x in arr if x % 2 == 0]
8
9 # Sorting -  $O(n \log n)$ 
10 arr.sort() # in-place, modifies arr
11 arr.sort(reverse=True) # descending
12 arr.sort(key=lambda x: (x[0], -x[1])) # custom
13 sorted_arr = sorted(arr) # returns new list
14
15 # Binary search in sorted array
16 from bisect import bisect_left, bisect_right
17 idx = bisect_left(arr, x) # leftmost position
18 idx = bisect_right(arr, x) # rightmost position
19
20 # Common operations
21 arr.append(x) #  $O(1)$  amortized
22 arr.pop() #  $O(1)$  - remove last
23 arr.pop(0) #  $O(n)$  - remove first (slow!)
24 arr.reverse() #  $O(n)$  - in-place
25 arr.count(x) #  $O(n)$  - count occurrences
26 arr.index(x) #  $O(n)$  - first occurrence
```

2.2 Deque (Double-ended Queue)

Description: Deque provides $O(1)$ append and pop from both ends, making it ideal for sliding window problems and implementing queues/stacks efficiently.

```
1 from collections import deque
2 dq = deque()
3
4 #  $O(1)$  operations on both ends
5 dq.append(x) # add to right
6 dq.appendleft(x) # add to left
7 dq.pop() # remove from right
8 dq.popleft() # remove from left
9
10 # Sliding window maximum -  $O(n)$ 
11 # Maintains decreasing order of elements
12 def sliding_max(arr, k):
13     dq = deque() # stores indices
14     result = []
15
16     for i in range(len(arr)):
17         # Remove indices outside window
18         while dq and dq[0] < i - k + 1:
19             dq.popleft()
20
21         # Remove smaller elements (not useful)
22         while dq and arr[dq[-1]] < arr[i]:
```

```

23         dq.pop()
24
25     dq.append(i)
26     if i >= k - 1:
27         result.append(arr[dq[0]])
28
29     return result

```

2.3 Heap (Priority Queue)

Description: Python's heapq implements a min-heap. For max-heap, negate values. Useful for finding k-th largest/smallest, Dijkstra's algorithm, and scheduling problems.

```

1  import heapq
2
3  # Min heap (default)
4  heap = []
5  heapq.heappush(heap, x)           # O(log n)
6  min_val = heapq.heappop(heap)     # O(log n)
7  min_val = heap[0]                 # O(1) peek
8
9  # Max heap - negate values
10 heapq.heappush(heap, -x)
11 max_val = -heapq.heappop(heap)
12
13 # Convert list to heap in-place - O(n)
14 heapq.heapify(arr)
15
16 # K largest/smallest - O(n log k)
17 k_largest = heapq.nlargest(k, arr)
18 k_smallest = heapq.nsmallest(k, arr)
19
20 # Custom comparator using tuples
21 # Compares first element, then second, etc.
22 heapq.heappush(heap, (priority, item))

```

2.4 Dictionary & Counter

Description: Hash maps with O(1) average case insert/lookup. Counter is specialized for counting occurrences.

```

1  from collections import defaultdict, Counter
2
3  # defaultdict - provides default value
4  graph = defaultdict(list) # empty list default
5  count = defaultdict(int)  # 0 default
6
7  # Counter - count elements efficiently
8  cnt = Counter(arr)
9  cnt['x'] += 1
10 most_common = cnt.most_common(k) # k most frequent
11
12 # Dictionary operations
13 d = {}
14 d.get(key, default_val)
15 d.setdefault(key, default_val)
16 for k, v in d.items():
17     pass

```

2.5 Set Operations

Description: Hash sets provide $O(1)$ membership testing and set operations.

```
1 s = set()
2 s.add(x)          #  $O(1)$ 
3 s.remove(x)       #  $O(1)$ , KeyError if not exists
4 s.discard(x)      #  $O(1)$ , no error if not exists
5
6 # Set operations - all  $O(n)$ 
7 a | b             # union
8 a & b             # intersection
9 a - b             # difference
10 a ^ b            # symmetric difference
11
12 # Ordered set workaround
13 from collections import OrderedDict
14 oset = OrderedDict.fromkeys([])
```

3 String Operations

Description: Strings in Python are immutable. For building strings, use list and join for $O(n)$ complexity instead of repeated concatenation which is $O(n^2)$.

```
1  # Common string methods
2  s.lower(), s.upper()
3  s.strip()    # remove whitespace both ends
4  s.lstrip()   # remove left whitespace
5  s.rstrip()   # remove right whitespace
6  s.split(delimiter)
7  delimiter.join(list)
8  s.replace(old, new)
9  s.startswith(prefix)
10 s.endswith(suffix)
11 s.isdigit(), s.isalpha(), s.isalnum()
12
13 # String building - EFFICIENT  $O(n)$ 
14 result = []
15 for x in data:
16     result.append(str(x))
17 s = ''.join(result)
18
19 # String concatenation - SLOW  $O(n^2)$ 
20 # s = ""
21 # for x in data:
22 #     s += str(x) # Don't do this!
23
24 # ASCII values
25 ord('a') # 97
26 chr(97)  # 'a'
27
28 # String to character array (for mutations)
29 chars = list(s)
30 chars[0] = 'x'
31 s = ''.join(chars)
```

4 Mathematics

4.1 Basic Math Operations

```
1 import math
2
3 # Common functions
4 math.ceil(x), math.floor(x)
5 math.gcd(a, b)      # Greatest common divisor
6 math.lcm(a, b)      # Python 3.9+
7 math.sqrt(x)
8 math.log(x), math.log2(x), math.log10(x)
9
10 # Powers
11 x ** y
12 pow(x, y, mod)     # (x^y) % mod - efficient modular exp
13
14 # Infinity
15 float('inf'), float('-inf')
16
17 # Custom GCD using Euclidean algorithm - O(log min(a,b))
18 def gcd(a, b):
19     while b:
20         a, b = b, a % b
21     return a
22
23 def lcm(a, b):
24     return a * b // gcd(a, b)
```

4.2 Combinatorics

Description: Compute combinations and permutations. For modular arithmetic, compute factorial arrays and use modular inverse.

```
1 from math import factorial, comb, perm
2
3 # nCr (combinations) - "n choose r"
4 comb(n, r) # Built-in Python 3.8+
5
6 # nPr (permutations)
7 perm(n, r) # Built-in Python 3.8+
8
9 # Manual nCr implementation
10 def ncr(n, r):
11     if r > n: return 0
12     r = min(r, n - r) # Optimization: C(n,r) = C(n,n-r)
13     num = den = 1
14     for i in range(r):
15         num *= (n - i)
16         den *= (i + 1)
17     return num // den
18
19 # Precompute factorials with modulo
20 MOD = 10**9 + 7
21 def modfact(n):
22     fact = [1] * (n + 1)
23     for i in range(1, n + 1):
24         fact[i] = fact[i-1] * i % MOD
25     return fact
26
27 # Modular combination using precomputed factorials
```



```
28 def modcomb(n, r, fact, modinv_fact):  
29     if r > n: return 0  
30     return fact[n] * modinv_fact[r] % MOD * modinv_fact[n-r] % MOD
```

5 Number Theory

Description: Essential algorithms for problems involving primes, modular arithmetic, and divisibility.

5.1 Modular Arithmetic

```
1 # Modular inverse using Fermat's Little Theorem
2 # Only works when mod is prime
3 #  $a^{-1} = a^{(mod-2)} \pmod{p}$ 
4 def modinv(a, mod):
5     return pow(a, mod - 2, mod)
6
7 # Extended Euclidean Algorithm
8 # Returns (gcd, x, y) where  $ax + by = \gcd(a, b)$ 
9 # Can find modular inverse for any coprime a, mod
10 def extgcd(a, b):
11     if b == 0:
12         return a, 1, 0
13     g, x1, y1 = extgcd(b, a % b)
14     x = y1
15     y = x1 - (a // b) * y1
16     return g, x, y
```

5.2 Sieve of Eratosthenes

Description: Find all primes up to n in $O(n \log \log n)$ time. Memory: $O(n)$.

```
1 def sieve(n):
2     is_prime = [True] * (n + 1)
3     is_prime[0] = is_prime[1] = False
4
5     for i in range(2, int(n**0.5) + 1):
6         if is_prime[i]:
7             # Mark multiples as composite
8             for j in range(i*i, n + 1, i):
9                 is_prime[j] = False
10
11     return is_prime
12
13 # Get list of primes
14 primes = [i for i in range(n+1) if is_prime[i]]
```

5.3 Prime Factorization

Description: Decompose n into prime factors in $O(\sqrt{n})$ time.

```
1 def factorize(n):
2     factors = []
3     d = 2
4
5     # Check divisors up to  $\sqrt{n}$ 
6     while d * d <= n:
7         while n % d == 0:
8             factors.append(d)
9             n //= d
10        d += 1
11
12    # If  $n > 1$ , it's a prime factor
13    if n > 1:
14        factors.append(n)
15
```

```
16     return factors
17
18 # Get prime factors with counts
19 from collections import Counter
20 def prime_factor_counts(n):
21     return Counter(factorize(n))
```

6 Graph Algorithms

6.1 Graph Representation

Description: Adjacency list is most common for sparse graphs. Use defaultdict for convenience.

```
1 from collections import defaultdict, deque
2
3 # Unweighted graph
4 graph = defaultdict(list)
5 for _ in range(m):
6     u, v = map(int, input().split())
7     graph[u].append(v)
8     graph[v].append(u) # for undirected
9
10 # Weighted graph - store (neighbor, weight) tuples
11 graph[u].append((v, weight))
```

6.2 BFS (Breadth-First Search)

Description: Explores graph level by level. Finds shortest path in unweighted graphs. Time: $O(V+E)$, Space: $O(V)$.

```
1 def bfs(graph, start):
2     visited = set([start])
3     queue = deque([start])
4     dist = {start: 0}
5
6     while queue:
7         node = queue.popleft()
8
9         for neighbor in graph[node]:
10             if neighbor not in visited:
11                 visited.add(neighbor)
12                 queue.append(neighbor)
13                 dist[neighbor] = dist[node] + 1
14
15     return dist
16
17 # Grid BFS - common in maze/path problems
18 def grid_bfs(grid, start):
19     n, m = len(grid), len(grid[0])
20     visited = [[False] * m for _ in range(n)]
21     queue = deque([start])
22     visited[start[0]][start[1]] = True
23
24     # 4 directions: right, down, left, up
25     dirs = [(0,1), (1,0), (0,-1), (-1,0)]
26
27     while queue:
28         x, y = queue.popleft()
29
30         for dx, dy in dirs:
31             nx, ny = x + dx, y + dy
32
33             # Check bounds and validity
34             if (0 <= nx < n and 0 <= ny < m
35                 and not visited[nx][ny]
36                 and grid[nx][ny] != '#'):
37
38                 visited[nx][ny] = True
```

6.3 DFS (Depth-First Search)

Description: Explores as far as possible along each branch. Used for connectivity, cycles, topological sort. Time: $O(V+E)$, Space: $O(V)$.

```

1  # Recursive DFS
2  def dfs(graph, node, visited):
3      visited.add(node)
4
5      for neighbor in graph[node]:
6          if neighbor not in visited:
7              dfs(graph, neighbor, visited)
8
9  # Iterative DFS using stack
10 def dfs_iterative(graph, start):
11     visited = set()
12     stack = [start]
13
14     while stack:
15         node = stack.pop()
16
17         if node not in visited:
18             visited.add(node)
19
20             for neighbor in graph[node]:
21                 if neighbor not in visited:
22                     stack.append(neighbor)
23
24 # Cycle detection in undirected graph
25 def has_cycle(graph, n):
26     visited = [False] * n
27
28     def dfs(node, parent):
29         visited[node] = True
30
31         for neighbor in graph[node]:
32             if not visited[neighbor]:
33                 if dfs(neighbor, node):
34                     return True
35             # Back edge to non-parent = cycle
36             elif neighbor != parent:
37                 return True
38
39     return False
40
41 # Check all components
42 for i in range(n):
43     if not visited[i]:
44         if dfs(i, -1):
45             return True
46
47 return False

```

7 Shortest Path Algorithms

7.1 Dijkstra's Algorithm

Description: Finds shortest paths from a source to all vertices in weighted graphs with non-negative edges. Time: $O((V+E) \log V)$ with heap.

```
1 import heapq
2
3 def dijkstra(graph, start, n):
4     # Initialize distances to infinity
5     dist = [float('inf')] * n
6     dist[start] = 0
7
8     # Min heap: (distance, node)
9     heap = [(0, start)]
10
11     while heap:
12         d, node = heapq.heappop(heap)
13
14         # Skip if already processed with better distance
15         if d > dist[node]:
16             continue
17
18         # Relax edges
19         for neighbor, weight in graph[node]:
20             new_dist = dist[node] + weight
21
22             if new_dist < dist[neighbor]:
23                 dist[neighbor] = new_dist
24                 heapq.heappush(heap, (new_dist, neighbor))
25
26     return dist
27
28 # Path reconstruction
29 def dijkstra_with_path(graph, start, n):
30     dist = [float('inf')] * n
31     parent = [-1] * n
32     dist[start] = 0
33     heap = [(0, start)]
34
35     while heap:
36         d, node = heapq.heappop(heap)
37         if d > dist[node]:
38             continue
39
40         for neighbor, weight in graph[node]:
41             new_dist = dist[node] + weight
42             if new_dist < dist[neighbor]:
43                 dist[neighbor] = new_dist
44                 parent[neighbor] = node
45                 heapq.heappush(heap, (new_dist, neighbor))
46
47     return dist, parent
48
49 def reconstruct_path(parent, target):
50     path = []
51     while target != -1:
52         path.append(target)
53         target = parent[target]
54     return path[::-1]
```

8 Topological Sort

Description: Linear ordering of vertices in a DAG (Directed Acyclic Graph) such that for every edge $u \rightarrow v$, u comes before v . Used for task scheduling, course prerequisites, build systems. Time: $O(V+E)$.

8.1 Kahn's Algorithm (BFS-based)

Advantages: Detects cycles, can process nodes level by level.

```
1 from collections import deque
2
3 def topo_sort(graph, n):
4     # Count incoming edges for each node
5     indegree = [0] * n
6     for u in range(n):
7         for v in graph[u]:
8             indegree[v] += 1
9
10    # Start with nodes having no dependencies
11    queue = deque([i for i in range(n)
12                  if indegree[i] == 0])
13    result = []
14
15    while queue:
16        node = queue.popleft()
17        result.append(node)
18
19        # Remove this node from graph
20        for neighbor in graph[node]:
21            indegree[neighbor] -= 1
22
23        # If neighbor has no more dependencies
24        if indegree[neighbor] == 0:
25            queue.append(neighbor)
26
27    # If not all nodes processed, cycle exists
28    return result if len(result) == n else []
```

8.2 DFS-based Topological Sort

Advantages: Simpler code, uses less space.

```
1 def topo_dfs(graph, n):
2     visited = [False] * n
3     stack = []
4
5     def dfs(node):
6         visited[node] = True
7
8         # Visit all neighbors first
9         for neighbor in graph[node]:
10             if not visited[neighbor]:
11                 dfs(neighbor)
12
13        # Add to stack after visiting all descendants
14        stack.append(node)
15
16    # Process all components
17    for i in range(n):
18        if not visited[i]:
19            dfs(i)
```

```
20  
21     # Reverse stack gives topological order  
22     return stack[::-1]
```


9 Union-Find (Disjoint Set Union)

Description: Efficiently tracks disjoint sets and supports union and find operations. Used for Kruskal's MST, connected components, cycle detection. Time: $O(\alpha(n)) \approx O(1)$ per operation with path compression and union by rank.

Applications:

- Kruskal's minimum spanning tree
- Detecting cycles in undirected graphs
- Finding connected components
- Network connectivity problems

```
1 class UnionFind:
2     def __init__(self, n):
3         # Each node is its own parent initially
4         self.parent = list(range(n))
5         # Rank for union by rank optimization
6         self.rank = [0] * n
7
8     def find(self, x):
9         # Path compression: point directly to root
10        if self.parent[x] != x:
11            self.parent[x] = self.find(self.parent[x])
12        return self.parent[x]
13
14    def union(self, x, y):
15        # Find roots
16        px, py = self.find(x), self.find(y)
17
18        # Already in same set
19        if px == py:
20            return False
21
22        # Union by rank: attach smaller tree under larger
23        if self.rank[px] < self.rank[py]:
24            px, py = py, px
25
26        self.parent[py] = px
27
28        # Increase rank if trees had equal rank
29        if self.rank[px] == self.rank[py]:
30            self.rank[px] += 1
31
32        return True
33
34    def connected(self, x, y):
35        return self.find(x) == self.find(y)
36
37    # Count number of disjoint sets
38    def count_sets(self):
39        return len(set(self.find(i)
40                        for i in range(len(self.parent))))
41
42    # Example: Detect cycle in undirected graph
43    def has_cycle_uf(edges, n):
44        uf = UnionFind(n)
45        for u, v in edges:
```

```
46     if uf.connected(u, v):  
47         return True # Cycle found  
48     uf.union(u, v)  
49 return False
```

10 Binary Search

Description: Search in $O(\log n)$ time. Works on monotonic functions (sorted arrays, or functions where condition transitions from false to true exactly once).

10.1 Template for Finding First/Last Position

```
1  # Find FIRST position where check(mid) is True
2  def binary_search_first(left, right, check):
3      while left < right:
4          mid = (left + right) // 2
5
6          if check(mid):
7              right = mid # Could be answer, search left
8          else:
9              left = mid + 1 # Not answer, search right
10
11     return left
12
13 # Find LAST position where check(mid) is True
14 def binary_search_last(left, right, check):
15     while left < right:
16         mid = (left + right + 1) // 2 # Round up!
17
18         if check(mid):
19             left = mid # Could be answer, search right
20         else:
21             right = mid - 1 # Not answer, search left
22
23     return left
24
25 # Example: Integer square root
26 def sqrt_binary(n):
27     left, right = 0, n
28
29     while left < right:
30         mid = (left + right + 1) // 2
31
32         if mid * mid <= n:
33             left = mid # mid might be answer
34         else:
35             right = mid - 1
36
37     return left
38
39 # Binary search on answer - common pattern
40 def min_days_to_make_bouquets(bloomDay, m, k):
41     # Can we make m bouquets in 'days' days?
42     def can_make(days):
43         bouquets = consecutive = 0
44         for bloom in bloomDay:
45             if bloom <= days:
46                 consecutive += 1
47                 if consecutive == k:
48                     bouquets += 1
49                     consecutive = 0
50             else:
51                 consecutive = 0
52         return bouquets >= m
53
54     if len(bloomDay) < m * k:
```

```
55     return -1
56
57     # Binary search on number of days
58     return binary_search_first(
59         min(bloomDay), max(bloomDay), can_make)
```

11 Dynamic Programming

Description: Solve problems by breaking them into overlapping subproblems. Store results to avoid recomputation.

11.1 Longest Increasing Subsequence

Description: Find length of longest strictly increasing subsequence. Time: $O(n \log n)$ using binary search.

```
1 def lis(arr):
2     from bisect import bisect_left
3
4     # dp[i] = smallest ending value of LIS of length i+1
5     dp = []
6
7     for x in arr:
8         # Find position to place x
9         idx = bisect_left(dp, x)
10
11        if idx == len(dp):
12            dp.append(x) # Extend LIS
13        else:
14            dp[idx] = x # Better ending for this length
15
16    return len(dp)
17
18 # LIS with actual sequence
19 def lis_with_sequence(arr):
20     from bisect import bisect_left
21
22     n = len(arr)
23     dp = []
24     parent = [-1] * n
25     dp_idx = [] # indices in dp
26
27     for i, x in enumerate(arr):
28         idx = bisect_left(dp, x)
29
30         if idx == len(dp):
31             dp.append(x)
32             dp_idx.append(i)
33         else:
34             dp[idx] = x
35             dp_idx[idx] = i
36
37         if idx > 0:
38             parent[i] = dp_idx[idx - 1]
39
40     # Reconstruct sequence
41     result = []
42     idx = dp_idx[-1]
43     while idx != -1:
44         result.append(arr[idx])
45         idx = parent[idx]
46
47     return result[::-1]
```

11.2 0/1 Knapsack

Description: Maximum value with weight capacity. Each item can be taken 0 or 1 time. Time: $O(n \times \text{capacity})$, Space: $O(n \times \text{capacity})$.

```
1 def knapsack(weights, values, capacity):
2     n = len(weights)
3     # dp[i][w] = max value using first i items,
4     #           weight <= w
5     dp = [[0] * (capacity + 1) for _ in range(n + 1)]
6
7     for i in range(1, n + 1):
8         for w in range(capacity + 1):
9             # Don't take item i-1
10            dp[i][w] = dp[i-1][w]
11
12            # Take item i-1 if it fits
13            if weights[i-1] <= w:
14                dp[i][w] = max(
15                    dp[i][w],
16                    dp[i-1][w - weights[i-1]] + values[i-1]
17                )
18
19     return dp[n][capacity]
20
21 # Space-optimized O(capacity)
22 def knapsack_optimized(weights, values, capacity):
23     dp = [0] * (capacity + 1)
24
25     for i in range(len(weights)):
26         # Iterate backwards to avoid using updated values
27         for w in range(capacity, weights[i] - 1, -1):
28             dp[w] = max(dp[w],
29                         dp[w - weights[i]] + values[i])
30
31     return dp[capacity]
```

11.3 Edit Distance (Levenshtein Distance)

Description: Minimum operations (insert, delete, replace) to transform s1 to s2. Time: $O(m \times n)$, Space: $O(m \times n)$.

```
1 def edit_dist(s1, s2):
2     m, n = len(s1), len(s2)
3     # dp[i][j] = edit distance of s1[:i] and s2[:j]
4     dp = [[0] * (n + 1) for _ in range(m + 1)]
5
6     # Base cases: empty string transformations
7     for i in range(m + 1):
8         dp[i][0] = i # Delete all
9     for j in range(n + 1):
10        dp[0][j] = j # Insert all
11
12    for i in range(1, m + 1):
13        for j in range(1, n + 1):
14            if s1[i-1] == s2[j-1]:
15                # Characters match, no operation needed
16                dp[i][j] = dp[i-1][j-1]
17            else:
18                dp[i][j] = 1 + min(
19                    dp[i-1][j], # Delete from s1
```

```
20         dp[i][j-1],      # Insert into s1
21         dp[i-1][j-1]      # Replace in s1
22     )
23
24     return dp[m][n]
```

12 Array Techniques

12.1 Prefix Sum

Description: Precompute cumulative sums for $O(1)$ range queries. Time: $O(n)$ preprocessing, $O(1)$ query.

```
1 # 1D prefix sum
2 prefix = [0] * (n + 1)
3 for i in range(n):
4     prefix[i + 1] = prefix[i] + arr[i]
5
6 # Range sum query [l, r] inclusive
7 range_sum = prefix[r + 1] - prefix[l]
8
9 # 2D prefix sum - for rectangle sum queries
10 def build_2d_prefix(matrix):
11     n, m = len(matrix), len(matrix[0])
12     prefix = [[0] * (m + 1) for _ in range(n + 1)]
13
14     for i in range(1, n + 1):
15         for j in range(1, m + 1):
16             prefix[i][j] = (matrix[i-1][j-1] +
17                             prefix[i-1][j] +
18                             prefix[i][j-1] -
19                             prefix[i-1][j-1])
20
21     return prefix
22
23 # Rectangle sum from (x1,y1) to (x2,y2) inclusive
24 def rect_sum(prefix, x1, y1, x2, y2):
25     return (prefix[x2+1][y2+1] -
26             prefix[x1][y2+1] -
27             prefix[x2+1][y1] +
28             prefix[x1][y1])
```

12.2 Difference Array

Description: Efficiently perform range updates. $O(1)$ per update, $O(n)$ to reconstruct.

```
1 # Initialize difference array
2 diff = [0] * (n + 1)
3
4 # Add 'val' to range [l, r]
5 def range_update(diff, l, r, val):
6     diff[l] += val
7     diff[r + 1] -= val
8
9 # After all updates, reconstruct array
10 def reconstruct(diff):
11     result = []
12     current = 0
13     for i in range(len(diff) - 1):
14         current += diff[i]
15         result.append(current)
16     return result
17
18 # Example: Multiple range updates
19 diff = [0] * (n + 1)
20 for l, r, val in updates:
21     range_update(diff, l, r, val)
```



```
22 final_array = reconstruct(diff)
```

12.3 Sliding Window

Description: Maintain a window of elements while traversing. Time: $O(n)$.

```
1  # Fixed size window
2  def max_sum_window(arr, k):
3      window_sum = sum(arr[:k])
4      max_sum = window_sum
5
6      # Slide window: add right, remove left
7      for i in range(k, len(arr)):
8          window_sum += arr[i] - arr[i - k]
9          max_sum = max(max_sum, window_sum)
10
11     return max_sum
12
13 # Variable size window - two pointers
14 def min_subarray_sum_geq_target(arr, target):
15     left = 0
16     current_sum = 0
17     min_len = float('inf')
18
19     for right in range(len(arr)):
20         current_sum += arr[right]
21
22         # Shrink window while condition holds
23         while current_sum >= target:
24             min_len = min(min_len, right - left + 1)
25             current_sum -= arr[left]
26             left += 1
27
28     return min_len if min_len != float('inf') else 0
29
30 # Longest substring with at most k distinct chars
31 def longest_k_distinct(s, k):
32     from collections import defaultdict
33
34     left = 0
35     char_count = defaultdict(int)
36     max_len = 0
37
38     for right in range(len(s)):
39         char_count[s[right]] += 1
40
41         # Shrink if too many distinct
42         while len(char_count) > k:
43             char_count[s[left]] -= 1
44             if char_count[s[left]] == 0:
45                 del char_count[s[left]]
46             left += 1
47
48         max_len = max(max_len, right - left + 1)
49
50     return max_len
```

13 Advanced Data Structures

13.1 Segment Tree

Description: Supports range queries and point updates in $O(\log n)$. Can be modified for range updates with lazy propagation.

```
1 class SegmentTree:
2     def __init__(self, arr):
3         self.n = len(arr)
4         # Tree size: 4n is safe upper bound
5         self.tree = [0] * (4 * self.n)
6         self.build(arr, 0, 0, self.n - 1)
7
8     def build(self, arr, node, start, end):
9         if start == end:
10             # Leaf node
11             self.tree[node] = arr[start]
12         else:
13             mid = (start + end) // 2
14             # Build left and right subtrees
15             self.build(arr, 2*node+1, start, mid)
16             self.build(arr, 2*node+2, mid+1, end)
17             # Combine results (sum in this case)
18             self.tree[node] = (self.tree[2*node+1] +
19                               self.tree[2*node+2])
20
21     def update(self, node, start, end, idx, val):
22         if start == end:
23             # Leaf node - update value
24             self.tree[node] = val
25         else:
26             mid = (start + end) // 2
27             if idx <= mid:
28                 # Update left subtree
29                 self.update(2*node+1, start, mid, idx, val)
30             else:
31                 # Update right subtree
32                 self.update(2*node+2, mid+1, end, idx, val)
33             # Recompute parent
34             self.tree[node] = (self.tree[2*node+1] +
35                               self.tree[2*node+2])
36
37     def query(self, node, start, end, l, r):
38         # No overlap
39         if r < start or end < l:
40             return 0
41
42         # Complete overlap
43         if l <= start and end <= r:
44             return self.tree[node]
45
46         # Partial overlap
47         mid = (start + end) // 2
48         left_sum = self.query(2*node+1, start, mid, l, r)
49         right_sum = self.query(2*node+2, mid+1, end, l, r)
50         return left_sum + right_sum
51
52     # Public interface
53     def update_val(self, idx, val):
54         self.update(0, 0, self.n-1, idx, val)
```

```

55
56     def range_sum(self, l, r):
57         return self.query(0, 0, self.n-1, l, r)

```

13.2 Fenwick Tree (Binary Indexed Tree)

Description: Simpler than segment tree, supports prefix sum and point updates in $O(\log n)$. More space efficient.

```

1  class FenwickTree:
2      def __init__(self, n):
3          self.n = n
4          # 1-indexed for easier implementation
5          self.tree = [0] * (n + 1)
6
7      def update(self, i, delta):
8          # Add delta to position i (1-indexed)
9          while i <= self.n:
10             self.tree[i] += delta
11             # Move to next node: add LSB
12             i += i & (-i)
13
14      def query(self, i):
15          # Get prefix sum up to i (1-indexed)
16          s = 0
17          while i > 0:
18             s += self.tree[i]
19             # Move to parent: remove LSB
20             i -= i & (-i)
21          return s
22
23      def range_query(self, l, r):
24          # Sum from l to r (1-indexed)
25          return self.query(r) - self.query(l - 1)
26
27  # Usage example
28  bit = FenwickTree(n)
29  for i, val in enumerate(arr, 1):
30      bit.update(i, val)
31
32  # Range sum [l, r] (1-indexed)
33  result = bit.range_query(l, r)

```

13.3 Trie (Prefix Tree)

Description: Tree for storing strings, enables fast prefix searches. Time: $O(m)$ for operations where m is string length.

```

1  class TrieNode:
2      def __init__(self):
3          self.children = {} # char -> TrieNode
4          self.is_end = False # End of word marker
5
6  class Trie:
7      def __init__(self):
8          self.root = TrieNode()
9
10     def insert(self, word):
11         # Insert word - O(len(word))
12         node = self.root

```

```

13     for char in word:
14         if char not in node.children:
15             node.children[char] = TrieNode()
16             node = node.children[char]
17     node.is_end = True
18
19 def search(self, word):
20     # Exact word search - O(len(word))
21     node = self.root
22     for char in word:
23         if char not in node.children:
24             return False
25         node = node.children[char]
26     return node.is_end
27
28 def starts_with(self, prefix):
29     # Prefix search - O(len(prefix))
30     node = self.root
31     for char in prefix:
32         if char not in node.children:
33             return False
34         node = node.children[char]
35     return True
36
37 # Find all words with given prefix
38 def words_with_prefix(self, prefix):
39     node = self.root
40     for char in prefix:
41         if char not in node.children:
42             return []
43         node = node.children[char]
44
45     # DFS to collect all words
46     words = []
47     def dfs(n, path):
48         if n.is_end:
49             words.append(prefix + path)
50         for char, child in n.children.items():
51             dfs(child, path + char)
52
53     dfs(node, "")
54     return words

```

14 Bit Manipulation

Description: Efficient operations using bitwise operators. Useful for sets, flags, and optimization.

```
1  # Check if i-th bit (0-indexed) is set
2  is_set = (n >> i) & 1
3
4  # Set i-th bit to 1
5  n |= (1 << i)
6
7  # Clear i-th bit (set to 0)
8  n &= ~(1 << i)
9
10 # Toggle i-th bit
11 n ^= (1 << i)
12
13 # Count set bits (popcount)
14 count = bin(n).count('1')
15 count = n.bit_count() # Python 3.10+
16
17 # Get lowest set bit
18 lsb = n & -n # Also n & (~n + 1)
19
20 # Remove lowest set bit
21 n &= (n - 1)
22
23 # Check if power of 2
24 is_pow2 = n > 0 and (n & (n - 1)) == 0
25
26 # Check if power of 4
27 is_pow4 = n > 0 and (n & (n-1)) == 0 and (n & 0x55555555) != 0
28
29 # Iterate over all subsets of set represented by mask
30 mask = (1 << n) - 1 # All bits set
31 submask = mask
32 while submask > 0:
33     # Process submask
34     submask = (submask - 1) & mask
35
36 # Iterate through all k-bit masks
37 def iterate_k_bits(n, k):
38     mask = (1 << k) - 1
39     while mask < (1 << n):
40         # Process mask
41         yield mask
42         # Gosper's hack
43         c = mask & -mask
44         r = mask + c
45         mask = (((r ^ mask) >> 2) // c) | r
46
47 # XOR properties
48 # a ^ a = 0 (number XOR itself is 0)
49 # a ^ 0 = a (number XOR 0 is itself)
50 # XOR is commutative and associative
51 # Find unique element when all others appear twice:
52 def find_unique(arr):
53     result = 0
54     for x in arr:
55         result ^= x
56     return result
57
```

```

58 # Subset enumeration
59 n = 5 # Number of elements
60 for mask in range(1 << n):
61     subset = [i for i in range(n) if mask & (1 << i)]
62     # Process subset
63
64 # Check parity (odd/even number of 1s)
65 def parity(n):
66     count = 0
67     while n:
68         count ^= 1
69         n &= n - 1
70     return count # 1 if odd, 0 if even
71
72 # Swap two numbers without temp variable
73 a, b = 5, 10
74 a ^= b
75 b ^= a
76 a ^= b
77 # Now a=10, b=5

```

15 Matrix Operations

Description: Matrix operations for DP optimization, graph algorithms, and recurrence relations.

15.1 Matrix Multiplication

```
1  # Standard matrix multiplication -  $O(n^3)$ 
2  def matmul(A, B):
3      n, m, p = len(A), len(A[0]), len(B[0])
4      C = [[0] * p for _ in range(n)]
5
6      for i in range(n):
7          for j in range(p):
8              for k in range(m):
9                  C[i][j] += A[i][k] * B[k][j]
10
11     return C
12
13 # With modulo
14 def matmul_mod(A, B, mod):
15     n = len(A)
16     C = [[0] * n for _ in range(n)]
17
18     for i in range(n):
19         for j in range(n):
20             for k in range(n):
21                 C[i][j] = (C[i][j] +
22                             A[i][k] * B[k][j]) % mod
23
24     return C
```

15.2 Matrix Exponentiation

Description: Compute M^n in $O(k^3 \log n)$ where k is matrix dimension. Used for solving linear recurrences efficiently.

```
1  def matpow(M, n, mod):
2      size = len(M)
3
4      # Identity matrix
5      result = [[1 if i==j else 0
6                  for j in range(size)]
7                 for i in range(size)]
8
9      # Binary exponentiation
10     while n > 0:
11         if n & 1:
12             result = matmul_mod(result, M, mod)
13             M = matmul_mod(M, M, mod)
14             n >>= 1
15
16     return result
17
18 # Example: Fibonacci using matrix exponentiation
19 #  $F(n) = \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix}^n$ 
20 def fibonacci(n, mod):
21     if n == 0: return 0
22     if n == 1: return 1
23
24     M = [[1, 1], [1, 0]]
25     result = matpow(M, n - 1, mod)
```

```

26     return result[0][0]
27
28 # Linear recurrence:  $a(n) = c_1 a(n-1) + c_2 a(n-2) + \dots$ 
29 # Build transition matrix and use matrix exponentiation
30 def linear_recurrence(coeffs, init, n, mod):
31     k = len(coeffs)
32
33     # Transition matrix
34     #  $[a(n), a(n-1), \dots, a(n-k+1)]$ 
35     M = [[0] * k for _ in range(k)]
36     M[0] = coeffs # First row
37     for i in range(1, k):
38         M[i][i-1] = 1 # Identity for shifting
39
40     # Initial state vector
41     state = init[::-1] # Reverse order
42
43     if n < k:
44         return init[n]
45
46     #  $M^{(n-k+1)}$ 
47     result_matrix = matpow(M, n - k + 1, mod)
48
49     # Multiply with initial state
50     result = 0
51     for i in range(k):
52         result = (result + result_matrix[0][i] * state[i]) % mod
53
54     return result

```


16 Miscellaneous Tips

16.1 Python-Specific Optimizations

```
1 # Fast input for large datasets
2 import sys
3 input = sys.stdin.readline
4
5 # Increase recursion limit for deep DFS/DP
6 sys.setrecursionlimit(10**6)
7
8 # Deep copy (be careful with performance)
9 from copy import deepcopy
10 new_list = deepcopy(old_list)
```

16.2 Useful Libraries

```
1 # Iterator tools - powerful combinations
2 from itertools import *
3
4 # permutations(iterable, r) - all r-length permutations
5 perms = list(permutations([1,2,3], 2))
6 # [(1,2), (1,3), (2,1), (2,3), (3,1), (3,2)]
7
8 # combinations(iterable, r) - r-length combinations
9 combs = list(combinations([1,2,3], 2))
10 # [(1,2), (1,3), (2,3)]
11
12 # product - cartesian product
13 prod = list(product([1,2], ['a', 'b']))
14 # [(1, 'a'), (1, 'b'), (2, 'a'), (2, 'b')]
15
16 # accumulate - running totals
17 acc = list(accumulate([1,2,3,4]))
18 # [1, 3, 6, 10]
19
20 # chain - flatten iterables
21 chained = list(chain([1,2], [3,4]))
22 # [1, 2, 3, 4]
```

16.3 Common Patterns

```
1 # Lambda sorting with multiple keys
2 arr.sort(key=lambda x: (-x[0], x[1]))
3 # Sort by first desc, then second asc
4
5 # All/Any - short-circuit evaluation
6 all(x > 0 for x in arr) # True if all positive
7 any(x > 0 for x in arr) # True if any positive
8
9 # Zip - parallel iteration
10 for a, b in zip(list1, list2):
11     pass
12
13 # Enumerate - index and value
14 for i, val in enumerate(arr):
15     print(f"arr[{i}] = {val}")
16
17 # Custom comparison function
18 from functools import cmp_to_key
```

```

19
20 def compare(a, b):
21     # Return -1 if a < b, 0 if equal, 1 if a > b
22     if a + b > b + a:
23         return -1
24     return 1
25
26 arr.sort(key=cmp_to_key(compare))
27
28 # defaultdict with lambda
29 from collections import defaultdict
30 d = defaultdict(lambda: float('inf'))
31
32 # Multiple assignment
33 a, b = b, a # Swap
34 a, *rest, b = [1,2,3,4,5] # a=1, rest=[2,3,4], b=5

```

16.4 Common Pitfalls

```

1 # Integer division - floors toward negative infinity
2 print(7 // 3) # 2
3 print(-7 // 3) # -3 (not -2!)
4
5 # For ceiling division toward zero:
6 def div_ceil(a, b):
7     return -(-a // b)
8
9 # Modulo with negative numbers
10 print((-5) % 3) # 1 (not -2!)
11 print(5 % -3) # -1
12
13 # List multiplication creates references!
14 matrix = [[0] * m] * n # WRONG! All rows same object
15 matrix[0][0] = 1 # Changes all rows!
16
17 # Correct way
18 matrix = [[0] * m for _ in range(n)]
19
20 # Float comparison - don't use ==
21 a, b = 0.1 + 0.2, 0.3
22 print(a == b) # False!
23
24 # Use epsilon comparison
25 eps = 1e-9
26 print(abs(a - b) < eps) # True
27
28 # String immutability
29 s = "abc"
30 # s[0] = 'd' # ERROR!
31 s = 'd' + s[1:] # OK
32
33 # For many string mutations, use list
34 chars = list(s)
35 chars[0] = 'd'
36 s = ''.join(chars)
37
38 # Mutable default arguments - dangerous!
39 def func(arr=[]): # WRONG!
40     arr.append(1)
41     return arr

```

```
42
43 # Each call modifies same list
44 print(func()) # [1]
45 print(func()) # [1, 1]
46
47 # Correct way
48 def func(arr=None):
49     if arr is None:
50         arr = []
51     arr.append(1)
52     return arr
```