

# 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 (pronounced "deck") provides  $O(1)$  append and pop operations from both ends, unlike lists which have  $O(n)$  for operations at the front. Essential for BFS, sliding window problems, implementing efficient queues/stacks, and maintaining monotonic queues. Use when you need fast insertions/deletions at both ends.

```

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 module implements a min-heap (smallest element always at index 0). Provides  $O(\log n)$  insert and extract-min operations,  $O(n)$  heapify, and  $O(1)$  peek. For max-heap, negate values before insertion. Critical for Dijkstra's algorithm, Prim's MST, k-th largest/smallest problems, merge k sorted lists, and any problem requiring repeated access to minimum/maximum elements. More efficient than sorting when you only need partial ordering.

```

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  $O(1)$  average-case membership testing, insertion, and deletion. Unlike lists, sets store only unique elements (no duplicates) and are unordered. Essential for removing duplicates, fast membership queries, and mathematical set operations (union, intersection, difference). Use when element uniqueness matters or you need fast lookups without caring about order. For sorted sets, consider using sorted containers or maintaining a sorted list separately.

```

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([])
15

```

```

16 \subsection{Skip List}
17 \textbf{Description:} Skip lists are probabilistic balanced data
    structures that support  $O(\log n)$  expected time for search,
    insert, and delete. They are an alternative to balanced BSTs
    and are simple to implement. Good when you want ordered set
    /map operations with expected logarithmic time.
18
19 \textbf{How it works:} A skip list is composed of multiple
    levels of sorted linked lists; level 0 contains all elements
    and higher levels act as “express lanes” with a subset of
    elements (each element appears in a random number of levels
    , forming a “tower”). To search, start at the head on the
    highest level, move right while the next key is less than
    the target, and drop down one level when you can no longer
    move right; repeat until level 0. New node levels are chosen
    randomly with probability  $p$  (common choice  $p=0.5$ ). For
    contests, use  $p=0.5$  and  $L_{\max}$   $\approx \lceil \log_2 N \rceil + 2$  for safety.
20
21 \begin{lstlisting}
22 import random
23
24 class SkipNode:
25     def __init__(self, key, value=None, level=0):
26         self.key = key
27         self.value = value
28         # forward pointers for levels 0..level
29         self.forward = [None] * (level + 1)
30
31 class SkipList:
32     def __init__(self, max_level=16, p=0.5):
33         self.max_level = max_level
34         self.p = p
35         self.level = 0
36         self.head = SkipNode(None, level=max_level)
37
38     def random_level(self):
39         lvl = 0
40         while random.random() < self.p and lvl < self.max_level:
41             lvl += 1
42         return lvl
43
44     def search(self, key):
45         node = self.head
46         # start from highest level
47         for i in range(self.level, -1, -1):
48             while node.forward[i] and node.forward[i].key < key:
49                 node = node.forward[i]
50         node = node.forward[0]
51         if node and node.key == key:
52             return node.value
53         return None
54
55     def insert(self, key, value=None):
56         update = [None] * (self.max_level + 1)
57         node = self.head
58         for i in range(self.level, -1, -1):
59             while node.forward[i] and node.forward[i].key < key:
60                 node = node.forward[i]
61             update[i] = node
62         node = node.forward[0]
63
64         # If key exists, update value
65         if node and node.key == key:
66             node.value = value
67             return
68
69         lvl = self.random_level()
70         if lvl > self.level:
71             for i in range(self.level + 1, lvl + 1):
72                 update[i] = self.head
73             self.level = lvl
74
75         new_node = SkipNode(key, value, lvl)
76         for i in range(lvl + 1):
77             new_node.forward[i] = update[i].forward[i]
78             update[i].forward[i] = new_node
79
80     def delete(self, key):
81         update = [None] * (self.max_level + 1)
82         node = self.head
83         for i in range(self.level, -1, -1):
84             while node.forward[i] and node.forward[i].key < key:
85                 node = node.forward[i]
86             update[i] = node

```

```

87         node = node.forward[0]
88
89         if not node or node.key != key:
90             return False # not found
91
92         for i in range(self.level + 1):
93             if update[i].forward[i] != node:
94                 break
95             update[i].forward[i] = node.forward[i]
96
97         # adjust current level
98         while self.level > 0 and self.head.forward[self.level]
99         is None:
100             self.level -= 1
101         return True
102
103 # Example usage:
104 # s = SkipList()
105 # s.insert(10, 'ten')
106 # print(s.search(10))
107 # s.delete(10)

```

## 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)

```

### 3.1 KMP Pattern Matching

**Description:** Find all occurrences of pattern in text. Time:  $O(n+m)$ .

```

1 def kmp_search(text, pattern):
2     # Build LPS (Longest Proper Prefix which is Suffix)
3     def build_lps(pattern):
4         m = len(pattern)
5         lps = [0] * m
6         length = 0 # Length of previous longest prefix
7         i = 1
8
9         while i < m:
10             if pattern[i] == pattern[length]:
11                 length += 1
12                 lps[i] = length
13                 i += 1
14             else:
15                 if length != 0:
16                     length = lps[length - 1]
17                 else:
18                     lps[i] = 0
19                     i += 1

```

```

20         return lps
21
22
23     n, m = len(text), len(pattern)
24     lps = build_lps(pattern)
25
26     matches = []
27     i = j = 0 # Indices for text and pattern
28
29     while i < n:
30         if text[i] == pattern[j]:
31             i += 1
32             j += 1
33
34         if j == m:
35             matches.append(i - j)
36             j = lps[j - 1]
37         elif i < n and text[i] != pattern[j]:
38             if j != 0:
39                 j = lps[j - 1]
40             else:
41                 i += 1
42
43     return matches

```

```

19     text_hash = 0
20     matches = []
21
22     for i in range(n):
23         # Add new character
24         text_hash = (text_hash * BASE +
25                     ord(text[i])) % MOD
26
27         # Remove old character if window full
28         if i >= m:
29             text_hash = (text_hash -
30                         ord(text[i - m]) * power) % MOD
31             text_hash = (text_hash + MOD) % MOD
32
33         # Check match
34         if i >= m - 1 and text_hash == pattern_hash:
35             # Verify actual match (avoid hash collision)
36             if text[i - m + 1:i + 1] == pattern:
37                 matches.append(i - m + 1)
38
39     return matches

```

## 3.2 Z-Algorithm

**Description:** Compute Z-array where  $Z[i]$  = length of longest substring starting from  $i$  that matches prefix. Time:  $O(n)$ .

```

1 def z_algorithm(s):
2     n = len(s)
3     z = [0] * n
4     l, r = 0, 0
5
6     for i in range(1, n):
7         if i <= r:
8             z[i] = min(r - i + 1, z[i - l])
9
10        while i + z[i] < n and s[z[i]] == s[i + z[i]]:
11            z[i] += 1
12
13        if i + z[i] - 1 > r:
14            l, r = i, i + z[i] - 1
15
16    return z
17
18 # Pattern matching using Z-algorithm
19 def z_search(text, pattern):
20     # Concatenate pattern + $ + text
21     s = pattern + '$' + text
22     z = z_algorithm(s)
23
24     matches = []
25     m = len(pattern)
26
27     for i in range(m + 1, len(s)):
28         if z[i] == m:
29             matches.append(i - m - 1)
30
31     return matches

```

## 3.3 Rabin-Karp (Rolling Hash)

**Description:** Fast pattern matching using hashing. Average:  $O(n+m)$ , Worst:  $O(nm)$ .

```

1 def rabin_karp(text, pattern):
2     MOD = 10**9 + 7
3     BASE = 31 # Prime base for hashing
4
5     n, m = len(text), len(pattern)
6     if m > n:
7         return []
8
9     # Compute hash of pattern
10    pattern_hash = 0
11    power = 1
12    for i in range(m):
13        pattern_hash = (pattern_hash * BASE +
14                        ord(pattern[i])) % MOD
15    if i < m - 1:
16        power = (power * BASE) % MOD
17
18    # Rolling hash

```

# 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 # First precompute inverse factorials
29 def compute_inv_factorials(n, mod):
30     fact = modfact(n)
31     inv_fact = [1] * (n + 1)
32     inv_fact[n] = pow(fact[n], mod - 2, mod)
33     for i in range(n - 1, -1, -1):
34         inv_fact[i] = inv_fact[i + 1] * (i + 1) % mod
35     return fact, inv_fact
36
37 def modcomb(n, r, fact, inv_fact, mod):
38     if r > n or r < 0: return 0
39     return fact[n] * inv_fact[r] % mod * inv_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) (mod 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))
22
23 # Count divisors
24 def count_divisors(n):

```

```

25     count = 0
26     i = 1
27     while i * i <= n:
28         if n % i == 0:
29             count += 1 if i * i == n else 2
30             i += 1
31     return count
32
33 # Sum of divisors
34 def sum_divisors(n):
35     total = 0
36     i = 1
37     while i * i <= n:
38         if n % i == 0:
39             total += i
40             if i != n // i:
41                 total += n // i
42             i += 1
43     return total

```

### 5.4 Chinese Remainder Theorem

**Description:** Solve system of congruences  $x \equiv a_1 \pmod{m_1}$ ,  $x \equiv a_2 \pmod{m_2}$ , ... Time:  $O(n \log M)$  where M is product of moduli.

```

1 def chinese_remainder(remainders, moduli):
2     # Solve x = remainders[i] (mod moduli[i])
3     # Assumes moduli are pairwise coprime
4
5     def extgcd(a, b):
6         if b == 0:
7             return a, 1, 0
8         g, x1, y1 = extgcd(b, a % b)
9         return g, y1, x1 - (a // b) * y1
10
11    total = 0
12    prod = 1
13    for m in moduli:
14        prod *= m
15
16    for r, m in zip(remainders, moduli):
17        p = prod // m
18        g, inv, _ = extgcd(p, m)
19        # inv may be negative, normalize it
20        inv = (inv % m + m) % m
21        total += r * inv * p
22
23    return total % prod

```

### 5.5 Euler's Totient Function

**Description:**  $\phi(n)$  = count of numbers  $\leq n$  coprime to n. Time:  $O(\sqrt{n})$ .

```

1 def euler_phi(n):
2     result = n
3     p = 2
4
5     while p * p <= n:
6         if n % p == 0:
7             # Remove factor p
8             while n % p == 0:
9                 n //= p
10            # Multiply by (1 - 1/p)
11            result -= result // p
12            p += 1
13
14    if n > 1:
15        result -= result // n
16
17    return result
18
19 # Phi for range [1, n] using sieve
20 def phi_sieve(n):
21     phi = list(range(n + 1)) # phi[i] = i initially
22
23    for i in range(2, n + 1):
24        if phi[i] == i: # i is prime
25            for j in range(i, n + 1, i):
26                phi[j] = phi[j] // i * (i - 1)
27
28    return phi

```



## 5.6 Fast Exponentiation with Matrix

**Description:** Already covered in matrix section, but useful pattern.

```
1 # Modular exponentiation
2 def mod_exp(base, exp, mod):
3     result = 1
4     base %= mod
5
6     while exp > 0:
7         if exp & 1:
8             result = (result * base) % mod
9         base = (base * base) % mod
10        exp >>= 1
11
12    return result
```

# 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
39                 queue.append((nx, ny))
```

## 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
48
49 # Cycle detection in directed graph
50 def has_cycle_directed(graph, n):
51     WHITE, GRAY, BLACK = 0, 1, 2
52     color = [WHITE] * n
53
54     def dfs(node):
55         color[node] = GRAY
56
57         for neighbor in graph[node]:
58             if color[neighbor] == GRAY:
59                 return True # Back edge = cycle
60             if color[neighbor] == WHITE:
61                 if dfs(neighbor):
62                     return True
63
64         color[node] = BLACK
65         return False
66
67     for i in range(n):
68         if color[i] == WHITE:
69             if dfs(i):
70                 return True
71     return False
72
73 # Connected components count
74 def count_components(graph, n):
75     visited = [False] * n
76     count = 0
77
78     def dfs(node):
79         visited[node] = True
```

```

80     for neighbor in graph[node]:
81         if not visited[neighbor]:
82             dfs(neighbor)
83
84     for i in range(n):
85         if not visited[i]:
86             dfs(i)
87             count += 1
88
89     return count
90
91 # Bipartite check (2-coloring)
92 def is_bipartite(graph, n):
93     color = [-1] * n
94
95     def bfs(start):
96         from collections import deque
97         queue = deque([start])
98         color[start] = 0
99
100        while queue:
101            node = queue.popleft()
102
103            for neighbor in graph[node]:
104                if color[neighbor] == -1:
105                    color[neighbor] = 1 - color[node]
106                    queue.append(neighbor)
107                elif color[neighbor] == color[node]:
108                    return False
109
110        return True
111
112     for i in range(n):
113         if color[i] == -1:
114             if not bfs(i):
115                 return False
116
117     return True

```

## 6.4 Strongly Connected Components (SCC)

**Description:** Find all SCCs in directed graph using Tarjan's algorithm. Time:  $O(V+E)$ .

```

1 def tarjan_scc(graph, n):
2     index_counter = [0]
3     stack = []
4     lowlink = [0] * n
5     index = [0] * n
6     on_stack = [False] * n
7     index_initialized = [False] * n
8     sccs = []
9
10    def strongconnect(v):
11        index[v] = index_counter[0]
12        lowlink[v] = index_counter[0]
13        index_counter[0] += 1
14        index_initialized[v] = True
15        stack.append(v)
16        on_stack[v] = True
17
18        for w in graph[v]:
19            if not index_initialized[w]:
20                strongconnect(w)
21                lowlink[v] = min(lowlink[v], lowlink[w])
22            elif on_stack[w]:
23                lowlink[v] = min(lowlink[v], index[w])
24
25        if lowlink[v] == index[v]:
26            scc = []
27            while True:
28                w = stack.pop()
29                on_stack[w] = False
30                scc.append(w)
31                if w == v:
32                    break
33            sccs.append(scc)
34
35    for v in range(n):
36        if not index_initialized[v]:
37            strongconnect(v)
38
39    return sccs

```

## 6.5 Bridges and Articulation Points

**Description:** Find critical edges (bridges) and vertices (articulation points). Time:  $O(V+E)$ .

```

1 def find_bridges(graph, n):
2     visited = [False] * n
3     disc = [0] * n
4     low = [0] * n
5     parent = [-1] * n
6     time = [0]
7     bridges = []
8
9     def dfs(u):
10        visited[u] = True
11        disc[u] = low[u] = time[0]
12        time[0] += 1
13
14        for v in graph[u]:
15            if not visited[v]:
16                parent[v] = u
17                dfs(v)
18                low[u] = min(low[u], low[v])
19
20                # Bridge condition
21                if low[v] > disc[u]:
22                    bridges.append((u, v))
23            elif v != parent[u]:
24                low[u] = min(low[u], disc[v])
25
26    for i in range(n):
27        if not visited[i]:
28            dfs(i)
29
30    return bridges
31
32 def find_articulation_points(graph, n):
33     visited = [False] * n
34     disc = [0] * n
35     low = [0] * n
36     parent = [-1] * n
37     time = [0]
38     ap = set()
39
40    def dfs(u):
41        children = 0
42        visited[u] = True
43        disc[u] = low[u] = time[0]
44        time[0] += 1
45
46        for v in graph[u]:
47            if not visited[v]:
48                children += 1
49                parent[v] = u
50                dfs(v)
51                low[u] = min(low[u], low[v])
52
53                # Articulation point conditions
54                if parent[u] == -1 and children > 1:
55                    ap.add(u)
56                if parent[u] != -1 and low[v] >= disc[u]:
57                    ap.add(u)
58            elif v != parent[u]:
59                low[u] = min(low[u], disc[v])
60
61    for i in range(n):
62        if not visited[i]:
63            dfs(i)
64
65    return list(ap)

```

## 6.6 Lowest Common Ancestor (LCA)

**Description:** Find LCA of two nodes in a tree. Binary lifting preprocessing:  $O(n \log n)$ , Query:  $O(\log n)$ .

```

1 class LCA:
2     def __init__(self, graph, root, n):
3         self.n = n
4         self.LOG = 20 # log2(n) + 1
5         self.parent = [[-1] * self.LOG for _ in range(n)]
6         self.depth = [0] * n
7
8         # DFS to set parent and depth
9         visited = [False] * n

```

```

11 def dfs(node, par, d):
12     visited[node] = True
13     self.parent[node][0] = par
14     self.depth[node] = d
15
16     for neighbor in graph[node]:
17         if not visited[neighbor]:
18             dfs(neighbor, node, d + 1)
19
20 dfs(root, -1, 0)
21
22 # Binary lifting preprocessing
23 for j in range(1, self.LOG):
24     for i in range(n):
25         if self.parent[i][j-1] != -1:
26             self.parent[i][j] = self.parent[
27                 self.parent[i][j-1]][j-1]
28
29 def lca(self, u, v):
30     # Make u deeper
31     if self.depth[u] < self.depth[v]:
32         u, v = v, u
33
34     # Bring u to same level as v
35     diff = self.depth[u] - self.depth[v]
36     for i in range(self.LOG):
37         if (diff >> i) & 1:
38             u = self.parent[u][i]
39
40     if u == v:
41         return u
42
43     # Binary search for LCA
44     for i in range(self.LOG - 1, -1, -1):
45         if self.parent[u][i] != self.parent[v][i]:
46             u = self.parent[u][i]
47             v = self.parent[v][i]
48
49     return self.parent[u][0]
50
51 def dist(self, u, v):
52     # Distance between two nodes
53     l = self.lca(u, v)
54     return self.depth[u] + self.depth[v] - 2 * self.depth[l]

```

```

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]

```

## 7.2 Bellman-Ford Algorithm

**Description:** Finds shortest paths with negative edges. Detects negative cycles. Time:  $O(VE)$ .

```

1 def bellman_ford(edges, n, start):
2     # edges = [(u, v, weight), ...]
3     dist = [float('inf')] * n
4     dist[start] = 0
5
6     # Relax edges n-1 times
7     for _ in range(n - 1):
8         for u, v, w in edges:
9             if dist[u] != float('inf') and \
10                dist[u] + w < dist[v]:
11                 dist[v] = dist[u] + w
12
13     # Check for negative cycles
14     for u, v, w in edges:
15         if dist[u] != float('inf') and \
16            dist[u] + w < dist[v]:
17             return None # Negative cycle exists
18
19     return dist

```

# 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

```

## 7.3 Floyd-Warshall Algorithm

**Description:** All-pairs shortest paths. Works with negative edges (no negative cycles). Time:  $O(V^3)$ .

```

1 def floyd_warshall(n, edges):
2     # Initialize distance matrix
3     dist = [[float('inf')] * n for _ in range(n)]
4
5     for i in range(n):
6         dist[i][i] = 0
7
8     for u, v, w in edges:
9         dist[u][v] = min(dist[u][v], w)
10
11     # Dynamic programming
12     for k in range(n): # Intermediate vertex
13         for i in range(n):
14             for j in range(n):
15                 dist[i][j] = min(dist[i][j],
16                                dist[i][k] + dist[k][j])
17
18     return dist
19
20 # Check for negative cycle
21 def has_negative_cycle(dist, n):
22     for i in range(n):
23         if dist[i][i] < 0:
24             return True
25     return False

```



## 7.4 Minimum Spanning Tree

### 7.4.1 Kruskal's Algorithm

**Description:** MST using Union-Find. Sort edges by weight. Time:  $O(E \log E)$ .

```
1 def kruskal(n, edges):
2     # edges = [(weight, u, v), ...]
3     edges.sort() # Sort by weight
4
5     uf = UnionFind(n)
6     mst_weight = 0
7     mst_edges = []
8
9     for weight, u, v in edges:
10        if uf.union(u, v):
11            mst_weight += weight
12            mst_edges.append((u, v, weight))
13
14    return mst_weight, mst_edges
15
16 class UnionFind:
17     def __init__(self, n):
18         self.parent = list(range(n))
19         self.rank = [0] * n
20
21     def find(self, x):
22         if self.parent[x] != x:
23             self.parent[x] = self.find(self.parent[x])
24         return self.parent[x]
25
26     def union(self, x, y):
27         px, py = self.find(x), self.find(y)
28         if px == py:
29             return False
30         if self.rank[px] < self.rank[py]:
31             px, py = py, px
32         self.parent[py] = px
33         if self.rank[px] == self.rank[py]:
34             self.rank[px] += 1
35         return True
```

### 7.4.2 Prim's Algorithm

**Description:** MST using heap. Good for dense graphs. Time:  $O(E \log V)$ .

```
1 import heapq
2
3 def prim(graph, n):
4     # graph[u] = [(v, weight), ...]
5     visited = [False] * n
6     min_heap = [(0, 0)] # (weight, node)
7     mst_weight = 0
8
9     while min_heap:
10        weight, u = heapq.heappop(min_heap)
11
12        if visited[u]:
13            continue
14
15        visited[u] = True
16        mst_weight += weight
17
18        for v, w in graph[u]:
19            if not visited[v]:
20                heapq.heappush(min_heap, (w, v))
21
22    return mst_weight
```

## 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
```

```
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
14    result = []
15
16    while queue:
17        node = queue.popleft()
18        result.append(node)
19
20        # Remove this node from graph
21        for neighbor in graph[node]:
22            indegree[neighbor] -= 1
23
24        # If neighbor has no more dependencies
25        if indegree[neighbor] == 0:
26            queue.append(neighbor)
27
28    # If not all nodes processed, cycle exists
29    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)
```

```

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]

```

### 11.4 Longest Common Subsequence (LCS)

**Description:** Longest subsequence common to two sequences. Time:  $O(m \times n)$ .

```

1 def lcs(s1, s2):
2     m, n = len(s1), len(s2)
3     dp = [[0] * (n + 1) for _ in range(m + 1)]
4
5     for i in range(1, m + 1):
6         for j in range(1, n + 1):
7             if s1[i-1] == s2[j-1]:
8                 dp[i][j] = dp[i-1][j-1] + 1
9             else:
10                dp[i][j] = max(dp[i-1][j], dp[i][j-1])
11
12    return dp[m][n]
13
14 # Reconstruct LCS
15 def lcs_string(s1, s2):
16     m, n = len(s1), len(s2)
17     dp = [[0] * (n + 1) for _ in range(m + 1)]
18
19     for i in range(1, m + 1):
20         for j in range(1, n + 1):
21             if s1[i-1] == s2[j-1]:

```

```

22         dp[i][j] = dp[i-1][j-1] + 1
23     else:
24         dp[i][j] = max(dp[i-1][j], dp[i][j-1])
25
26 # Backtrack
27 result = []
28 i, j = m, n
29 while i > 0 and j > 0:
30     if s1[i-1] == s2[j-1]:
31         result.append(s1[i-1])
32         i -= 1
33         j -= 1
34     elif dp[i-1][j] > dp[i][j-1]:
35         i -= 1
36     else:
37         j -= 1
38
39     return ''.join(reversed(result))

```

### 11.5 Coin Change

**Description:** Minimum coins to make amount, or count ways. Time:  $O(n \times \text{amount})$ .

```

1 # Minimum coins
2 def coin_change_min(coins, amount):
3     dp = [float('inf')] * (amount + 1)
4     dp[0] = 0
5
6     for coin in coins:
7         for i in range(coin, amount + 1):
8             dp[i] = min(dp[i], dp[i - coin] + 1)
9
10    return dp[amount] if dp[amount] != float('inf') else -1
11
12 # Count ways
13 def coin_change_ways(coins, amount):
14     dp = [0] * (amount + 1)
15     dp[0] = 1
16
17     for coin in coins:
18         for i in range(coin, amount + 1):
19             dp[i] += dp[i - coin]
20
21    return dp[amount]

```

### 11.6 Palindrome Partitioning

**Description:** Minimum cuts to partition string into palindromes. Time:  $O(n^2)$ .

```

1 def min_palindrome_partition(s):
2     n = len(s)
3
4     # is_pal[i][j] = True if s[i:j+1] is palindrome
5     is_pal = [[False] * n for _ in range(n)]
6
7     # Every single character is palindrome
8     for i in range(n):
9         is_pal[i][i] = True
10
11    # Check all substrings
12    for length in range(2, n + 1):
13        for i in range(n - length + 1):
14            j = i + length - 1
15            if s[i] == s[j]:
16                is_pal[i][j] = (length == 2 or
17                                is_pal[i+1][j-1])
18
19    # dp[i] = min cuts for s[0:i+1]
20    dp = [float('inf')] * n
21
22    for i in range(n):
23        if is_pal[0][i]:
24            dp[i] = 0
25        else:
26            for j in range(i):
27                if is_pal[j+1][i]:
28                    dp[i] = min(dp[i], dp[j] + 1)
29
30    return dp[n-1]

```

### 11.7 Subset Sum

**Description:** Check if subset sums to target. Time:  $O(n \times \text{sum})$ .

```

1 def subset_sum(arr, target):
2     n = len(arr)
3     dp = [[False] * (target + 1) for _ in range(n + 1)]
4
5     # Base case: sum 0 is always achievable
6     for i in range(n + 1):
7         dp[i][0] = True
8
9     for i in range(1, n + 1):
10        for s in range(target + 1):
11            # Don't take arr[i-1]
12            dp[i][s] = dp[i-1][s]
13
14            # Take arr[i-1] if possible
15            if s >= arr[i-1]:
16                dp[i][s] = dp[i][s] or dp[i-1][s - arr[i-1]]
17
18    return dp[n][target]
19
20 # Space optimized
21 def subset_sum_optimized(arr, target):
22     dp = [False] * (target + 1)
23     dp[0] = True
24
25     for num in arr:
26         for s in range(target, num - 1, -1):
27             dp[s] = dp[s] or dp[s - num]
28
29    return dp[target]

```

## 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

```

## 13.4 Treap (Randomized Balanced BST)

**Description:** Ordered set/map with expected  $O(\log n)$  insert, erase, search, k-th, and rank. Combines a BST by key and a heap by random priority. Stores unique keys; for multiset, store (key, uid) or maintain a count.

```

1 import random
2
3 class TreapNode:
4     __slots__ = ("key", "prio", "left", "right", "size")
5     def __init__(self, key):
6         self.key = key
7         self.prio = random.randint(1, 1 << 30)
8         self.left = None
9         self.right = None
10        self.size = 1
11
12    def _sz(t):
13        return t.size if t else 0
14
15    def _upd(t):
16        if t:
17            t.size = 1 + _sz(t.left) + _sz(t.right)
18
19

```



```

19 def _merge(a, b):
20     # assumes all keys in a < all keys in b
21     if not a or not b:
22         return a or b
23     if a.prio > b.prio:
24         a.right = _merge(a.right, b)
25         _upd(a)
26         return a
27     else:
28         b.left = _merge(a, b.left)
29         _upd(b)
30         return b
31
32 def _split(t, key):
33     # returns (l, r): l has keys < key, r has keys >= key
34     if not t:
35         return (None, None)
36     if key <= t.key:
37         l, t.left = _split(t.left, key)
38         _upd(t)
39         return (l, t)
40     else:
41         t.right, r = _split(t.right, key)
42         _upd(t)
43         return (t, r)
44
45 def _erase(t, key):
46     if not t:
47         return None
48     if key == t.key:
49         return _merge(t.left, t.right)
50     if key < t.key:
51         t.left = _erase(t.left, key)
52     else:
53         t.right = _erase(t.right, key)
54     _upd(t)
55     return t
56
57 class Treap:
58     def __init__(self):
59         self.root = None
60
61     def __len__(self):
62         return _sz(self.root)
63
64     def contains(self, key):
65         t = self.root
66         while t:
67             if key == t.key:
68                 return True
69             t = t.left if key < t.key else t.right
70         return False
71
72     def insert(self, key):
73         if self.contains(key):
74             return
75         node = TreapNode(key)
76         l, r = _split(self.root, key)
77         self.root = _merge(_merge(l, node), r)
78
79     def remove(self, key):
80         self.root = _erase(self.root, key)
81
82     def kth_smallest(self, k):
83         # 0-indexed k
84         t = self.root
85         while t:
86             ls = _sz(t.left)
87             if k < ls:
88                 t = t.left
89             elif k == ls:
90                 return t.key
91             else:
92                 k -= ls + 1
93                 t = t.right
94         return None # k out of range
95
96     def count_less_than(self, key):
97         # number of keys < key
98         t, cnt = self.root, 0
99         while t:
100             if key <= t.key:
101                 t = t.left
102             else:
103                 cnt += 1 + _sz(t.left)

```

```

104         t = t.right
105         return cnt
106
107     def lower_bound(self, key):
108         # smallest key >= key; returns None if none
109         t, ans = self.root, None
110         while t:
111             if t.key >= key:
112                 ans = t.key
113                 t = t.left
114             else:
115                 t = t.right
116         return ans
117
118 # Usage example
119 T = Treap()
120 for x in [5, 1, 7, 3]:
121     T.insert(x)
122 T.contains(3) # True
123 T.kth_smallest(1) # 3 (0-indexed)
124 T.count_less_than(6) # 3 (1,3,5)
125 T.remove(5)
126 len(T) # 3

```

## 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) = [[1,1],[1,0]]^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) = c1*a(n-1) + c2*a(n-2) + ...
29 # Build transition matrix and use matrix exponentiation
30 def linear_recurrence(coeffs, init, n, mod):
31     k = len(coeffs)
32
33     if n < k:
34         return init[n]
35
36     # Transition matrix
37     # [a(n), a(n-1), ..., a(n-k+1)]
38     M = [[0] * k for _ in range(k)]
39     M[0] = coeffs # First row
40     for i in range(1, k):
41         M[i][i-1] = 1 # Identity for shifting
42
43     # Initial state vector [a(k-1), a(k-2), ..., a(0)]
44     state = init[k-1::-1]
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
55
56 # Example: Tribonacci T(n) = T(n-1) + T(n-2) + T(n-3)
57 def tribonacci(n, mod):
58     if n == 0: return 0
59     if n == 1 or n == 2: return 1
60
61     coeffs = [1, 1, 1]
62     init = [0, 1, 1]
63     return linear_recurrence(coeffs, init, n, mod)

```

## 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 # Threading for higher stack limit (CAUTION: use carefully)
9 import threading
10 threading.stack_size(2**26) # 64MB
11 sys.setrecursionlimit(2**20)
12
13 # Deep copy (be careful with performance)
14 from copy import deepcopy
15 new_list = deepcopy(old_list)
16
17 # Fast output (for printing large results)
18 import sys
19 print = sys.stdout.write # Only use for string output

```

### 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 # chain - flatten iterables
20 chained = list(chain([1,2], [3,4]))
21 # [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 # Each call modifies same list
43
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
53
54 # Generator expressions save memory
55 sum(x*x for x in range(10**6)) # Memory efficient
56 # vs
57 sum([x*x for x in range(10**6)]) # Creates full list
58
59 # Ternary operator
60 x = a if condition else b
61
62 # Dictionary get with default
63 count = d.get(key, 0) + 1
64
65 # Matrix rotation 90 degrees clockwise
66 def rotate_90(matrix):
67     return [list(row) for row in zip(*matrix[::-1])]
68
69 # Matrix transpose
70 def transpose(matrix):
71     return [list(row) for row in zip(*matrix)]

```

## 16.5 Time Complexity Reference

Common time complexities (Python, rough guides for 1–2s limits):

- $O(1)$ ,  $O(\log n)$ : instant
- $O(n)$ : usually fine up to  $\sim 10^7$  operations ( $\sim 1$ s)
- $O(n \log n)$ : OK for  $n$  up to several  $10^5$  depending on constants
- $O(n\sqrt{n})$ : risky in Python (may be OK for  $n$  up to a few  $10^4$  with low constants)
- $O(n^2)$ : often TLE for  $n > 10^4$
- $O(2^n)$ : TLE for  $n > 20$  (unless heavy pruning/memoization)
- $O(n!)$ : TLE for  $n > 11$

Input size guidelines (Python-focused):

- $n \leq 12$ :  $O(n!)$  (brute-force permutations)
- $n \leq 20$ :  $O(2^n)$  (subset DP / bitmask DP)
- $n \leq 500$ :  $O(n^3)$  may sometimes pass for small constants
- $n \leq 5000$ :  $O(n^2)$  borderline; optimize heavily
- $n \leq 10^6$ :  $O(n \log n)$  common;  $O(n)$  preferred when possible
- $n \leq 10^7$ :  $O(n)$  may be OK for tight loops
- $n > 10^7$ : aim for  $O(n)$  with very low constants, or  $O(\log n)/O(1)$

### Complexity examples (Python implementations)

- $O(1)$ : array access, dictionary lookup, push/pop from list end.
- $O(\log n)$ : binary search (bisect), heap push/pop (heapq), operations in sortedcontainers.
- $O(n)$ : single-pass scans, two-pointers, prefix sums, counting frequencies (Counter).
- $O(n \log n)$ : sorting (Timsort via sorted()/list.sort()), heap construction, divide-and-conquer merges.
- $O(n\sqrt{n})$ : sqrt-decomposition queries, some Mo's algorithm variants (constant-sensitive).
- $O(n^2)$ : nested loops for pairwise checks, naive DP on pairs (be cautious for  $n > 10,000$ ).
- $O(n^3)$ : triple loops (Floyd–Warshall), usually too slow unless  $n \leq 200$ .
- $O(2^n)$ : bitmask DP, subset enumerations, recursion over subsets (recommended for  $n \leq 20$ ).
- $O(n!)$ : full permutations, exhaustive search over orderings (recommended for  $n \leq 10$ ; occasionally up to 11).

**How to use:** This quick reference maps input size  $n$  (left) to typical feasible time complexities (right) for contest time limits (1–2s) targeting Python implementations. Use it to pick algorithmic ap-

proaches and to decide when to optimize or change strategy.

### Notes on filling the table:

- Start by checking the problem's time limit and target language. These guidelines are Python-focused (assume roughly  $\approx 10^7$  simple operations/s; actual throughput depends on implementation details and input shapes).
- Convert algorithm cost to operation count: roughly cost =  $c \cdot f(n)$ . If cost > time.limit  $\times$  ops.per.sec, it will TLE.
- When in doubt, aim one complexity class lower (e.g. prefer  $O(n \log n)$  over  $O(n^2)$  for  $n$  around  $10^5$ ).
- Consider memory limits—some faster algorithms use more memory (e.g. segment trees vs. Fenwick tree).
- For multivariate inputs, replace  $n$  with the product/dominant parameter (e.g.  $n \cdot m$ ) and apply the same rules.
- If an algorithm theoretically fits but is close to the limit, try to reduce constant factors: use local variables, avoid heavy Python objects in inner loops, use built-in functions, or move hot code to PyPy/Cython if allowed.

## 17 Computational Geometry

### 17.1 Basic Geometry

**Description:** Fundamental geometric operations for 2D points.

```
1 import math
2
3 # Point operations
4 def dist(p1, p2):
5     # Euclidean distance
6     return math.sqrt((p1[0] - p2[0])**2 + (p1[1] - p2[1])**2)
7
8 def cross_product(O, A, B):
9     # Cross product of vectors OA and OB
10    # Positive: counter-clockwise
11    # Negative: clockwise
12    # Zero: collinear
13    return (A[0] - O[0]) * (B[1] - O[1]) - \
14           (A[1] - O[1]) * (B[0] - O[0])
15
16 def dot_product(A, B, C, D):
17     # Dot product of vectors AB and CD
18     return (B[0] - A[0]) * (D[0] - C[0]) + \
19           (B[1] - A[1]) * (D[1] - C[1])
20
21 # Check if point is on segment
22 def on_segment(p, q, r):
23     # Check if q lies on segment pr
24     return (q[0] <= max(p[0], r[0]) and
25            q[0] >= min(p[0], r[0]) and
26            q[1] <= max(p[1], r[1]) and
27            q[1] >= min(p[1], r[1]))
28
29 # Segment intersection
30 def segments_intersect(p1, q1, p2, q2):
31     o1 = cross_product(p1, q1, p2)
32     o2 = cross_product(p1, q1, q2)
33     o3 = cross_product(p2, q2, p1)
34     o4 = cross_product(p2, q2, q1)
35
36     # General case
37     if o1 * o2 < 0 and o3 * o4 < 0:
38         return True
39
40     # Special cases (collinear)
41     if o1 == 0 and on_segment(p1, p2, q1):
42         return True
43     if o2 == 0 and on_segment(p1, q2, q1):
44         return True
45     if o3 == 0 and on_segment(p2, p1, q2):
46         return True
47     if o4 == 0 and on_segment(p2, q1, q2):
48         return True
49
50     return False
```

### 17.2 Convex Hull

**Description:** Find convex hull using Graham's scan. Time:  $O(n \log n)$ .

```
1 def convex_hull(points):
2     # Graham's scan algorithm
3     points = sorted(points) # Sort by x, then y
4
5     if len(points) <= 2:
6         return points
7
8     # Build lower hull
9     lower = []
10    for p in points:
11        while (len(lower) >= 2 and
12              cross_product(lower[-2], lower[-1], p) <= 0):
13            lower.pop()
14        lower.append(p)
15
16    # Build upper hull
17    upper = []
18    for p in reversed(points):
19        while (len(upper) >= 2 and
20              cross_product(upper[-2], upper[-1], p) <= 0):
21            upper.pop()
22        upper.append(p)
23
24    # Remove last point (duplicate of first)
25    return lower[:-1] + upper[:-1]
26
27 # Convex hull area
28 def polygon_area(points):
29     # Shoelace formula
30     n = len(points)
31     area = 0
32
33     for i in range(n):
34         j = (i + 1) % n
35         area += points[i][0] * points[j][1]
36         area -= points[j][0] * points[i][1]
37
38     return abs(area) / 2
```

### 17.3 Point in Polygon

**Description:** Check if point is inside polygon. Time:  $O(n)$ .

```
1 def point_in_polygon(point, polygon):
2     # Ray casting algorithm
3     x, y = point
4     n = len(polygon)
5     inside = False
6
7     p1x, p1y = polygon[0]
8     for i in range(1, n + 1):
9         p2x, p2y = polygon[i % n]
10
11        if y > min(p1y, p2y):
12            if y <= max(p1y, p2y):
13                if x <= max(p1x, p2x):
14                    if p1y != p2y:
15                        xinters = (y - p1y) * (p2x - p1x) / \
16                                (p2y - p1y) + p1x
17
18                    if p1x == p2x or x <= xinters:
19                        inside = not inside
20
21        p1x, p1y = p2x, p2y
22
23    return inside
```

### 17.4 Closest Pair of Points

**Description:** Find closest pair using divide and conquer. Time:  $O(n \log n)$ .

```
1 def closest_pair(points):
2     points_sorted_x = sorted(points, key=lambda p: p[0])
3     points_sorted_y = sorted(points, key=lambda p: p[1])
4
5     def closest_recursive(px, py):
6         n = len(px)
7
8         # Base case: brute force
9         if n <= 3:
```

```

10 min_dist = float('inf')
11 for i in range(n):
12     for j in range(i + 1, n):
13         min_dist = min(min_dist, dist(px[i], px[j]))
14     return min_dist
15
16 # Divide
17 mid = n // 2
18 midpoint = px[mid]
19
20 pyl = [p for p in py if p[0] <= midpoint[0]]
21 pyr = [p for p in py if p[0] > midpoint[0]]
22
23 # Conquer
24 dl = closest_recursive(px[:mid], pyl)
25 dr = closest_recursive(px[mid:], pyr)
26 d = min(dl, dr)
27
28 # Combine: check strip
29 strip = [p for p in py if abs(p[0] - midpoint[0]) < d]
30
31 for i in range(len(strip)):
32     j = i + 1
33     while j < len(strip) and strip[j][1] - strip[i][1] <
34         d = min(d, dist(strip[i], strip[j]))
35         j += 1
36
37 return d
38
39 return closest_recursive(points_sorted_x, points_sorted_y)

```

```

45 flow = min(residual[u][v] for u, v in path)
46
47 # Update residual graph
48 for u, v in path:
49     residual[u][v] -= flow
50     residual[v][u] += flow
51
52 max_flow_value += flow
53
54 return max_flow_value
55
56 # Example usage
57 # graph[u][v] = capacity
58 graph = defaultdict(lambda: defaultdict(int))
59 graph[0][1] = 10
60 graph[0][2] = 10
61 graph[1][3] = 4
62 graph[1][4] = 8
63 graph[2][4] = 9
64 graph[3][5] = 10
65 graph[4][3] = 6
66 graph[4][5] = 10
67
68 n = 6 # Number of nodes
69 result = max_flow(graph, 0, 5, n)

```

## 18 Network Flow

### 18.1 Maximum Flow - Edmonds-Karp (BFS-based Ford-Fulkerson)

**Description:** Find maximum flow from source to sink. Time:  $O(VE^2)$ .

```

1 from collections import deque, defaultdict
2
3 def max_flow(graph, source, sink, n):
4     # graph[u][v] = capacity from u to v
5     # Build residual graph
6     residual = defaultdict(lambda: defaultdict(int))
7     for u in graph:
8         for v in graph[u]:
9             residual[u][v] = graph[u][v]
10
11 def bfs_path():
12     # Find augmenting path using BFS
13     parent = {source: None}
14     visited = {source}
15     queue = deque([source])
16
17 while queue:
18     u = queue.popleft()
19
20     if u == sink:
21         # Reconstruct path
22         path = []
23         while parent[u] is not None:
24             path.append((parent[u], u))
25             u = parent[u]
26         return path[::-1]
27
28 for v in range(n):
29     if v not in visited and residual[u][v] > 0:
30         visited.add(v)
31         parent[v] = u
32         queue.append(v)
33
34 return None
35
36 max_flow_value = 0
37
38 # Find augmenting paths
39 while True:
40     path = bfs_path()
41     if path is None:
42         break
43
44 # Find minimum capacity along path

```

### 18.2 Dinic's Algorithm (Faster)

**Description:** Faster max flow using level graph and blocking flow. Time:  $O(V^2E)$ .

```

1 from collections import deque, defaultdict
2
3 class Dinic:
4     def __init__(self, n):
5         self.n = n
6         self.graph = defaultdict(lambda: defaultdict(int))
7
8     def add_edge(self, u, v, cap):
9         self.graph[u][v] += cap
10
11     def bfs(self, source, sink):
12         # Build level graph
13         level = [-1] * self.n
14         level[source] = 0
15         queue = deque([source])
16
17 while queue:
18     u = queue.popleft()
19
20     for v in range(self.n):
21         if level[v] == -1 and self.graph[u][v] > 0:
22             level[v] = level[u] + 1
23             queue.append(v)
24
25 return level if level[sink] != -1 else None
26
27 def dfs(self, u, sink, pushed, level, start):
28     if u == sink:
29         return pushed
30
31 while start[u] < self.n:
32     v = start[u]
33
34     if (level[v] == level[u] + 1 and
35         self.graph[u][v] > 0):
36
37         flow = self.dfs(v, sink,
38                         min(pushed, self.graph[u][v]),
39                         level, start)
40
41         if flow > 0:
42             self.graph[u][v] -= flow
43             self.graph[v][u] += flow
44             return flow
45
46     start[u] += 1
47
48 return 0
49
50 def max_flow(self, source, sink):
51     flow = 0
52
53 while True:
54     level = self.bfs(source, sink)

```



```

55         if level is None:
56             break
57
58         start = [0] * self.n
59
60         while True:
61             pushed = self.dfs(source, sink, float('inf'),
62                             level, start)
63             if pushed == 0:
64                 break
65             flow += pushed
66
67         return flow

```

## 18.3 Min Cut

**Description:** Find minimum cut after computing max flow.

```

1 def min_cut(graph, source, n, residual):
2     # After running max_flow, residual graph is available
3     # Min cut = set of reachable nodes from source
4     visited = [False] * n
5     queue = deque([source])
6     visited[source] = True
7
8     while queue:
9         u = queue.popleft()
10        for v in range(n):
11            if not visited[v] and residual[u][v] > 0:
12                visited[v] = True
13                queue.append(v)
14
15    # Cut edges
16    cut_edges = []
17    for u in range(n):

```

```

18        if visited[u]:
19            for v in range(n):
20                if not visited[v] and graph[u][v] > 0:
21                    cut_edges.append((u, v))
22
23    return cut_edges

```

## 18.4 Bipartite Matching

**Description:** Maximum matching in bipartite graph using flow.

```

1 def max_bipartite_matching(left_size, right_size, edges):
2     # edges = [(left_node, right_node), ...]
3     # Add source (0) and sink (left_size + right_size + 1)
4
5     n = left_size + right_size + 2
6     source = 0
7     sink = n - 1
8
9     graph = defaultdict(lambda: defaultdict(int))
10
11    # Source to left nodes
12    for i in range(1, left_size + 1):
13        graph[source][i] = 1
14
15    # Left to right edges
16    for l, r in edges:
17        graph[l + 1][left_size + r + 1] = 1
18
19    # Right nodes to sink
20    for i in range(1, right_size + 1):
21        graph[left_size + i][sink] = 1
22
23    return max_flow(graph, source, sink, n)

```