

# Python ICPC Cheatsheet

Comprehensive Reference for Competitive Programming

## Contents

<b>1</b>	<b>Input/Output</b>	<b>2</b>	7.4.1	Kruskal's Algo- rithm . . . . .	11
			7.4.2	Prim's Algorithm .	11
<b>2</b>	<b>Basic Data Structures</b>	<b>2</b>	<b>8</b>	<b>Topological Sort</b>	<b>11</b>
2.1	List Operations . . . . .	2	8.1	Kahn's Algorithm (BFS- based) . . . . .	11
2.2	Deque (Double-ended Queue) . . . . .	2	8.2	DFS-based Topological Sort . . . . .	11
2.3	Heap (Priority Queue) . .	3	<b>9</b>	<b>Union-Find (Disjoint Set Union)</b>	<b>12</b>
2.4	Dictionary & Counter . .	3	<b>10</b>	<b>Binary Search</b>	<b>12</b>
2.5	Set Operations . . . . .	3	10.1	Template for Finding First/Last Position . . . .	12
<b>3</b>	<b>String Operations</b>	<b>3</b>	<b>11</b>	<b>Dynamic Programming</b>	<b>13</b>
3.1	KMP Pattern Matching .	4	11.1	Longest Increasing Sub- sequence . . . . .	13
3.2	Z-Algorithm . . . . .	4	11.2	0/1 Knapsack . . . . .	13
3.3	Rabin-Karp (Rolling Hash)	4	11.3	Edit Distance (Leven- shtein Distance) . . . . .	14
<b>4</b>	<b>Mathematics</b>	<b>5</b>	11.4	Longest Common Subse- quence (LCS) . . . . .	14
4.1	Basic Math Operations . .	5	11.5	Coin Change . . . . .	14
4.2	Combinatorics . . . . .	5	11.6	Palindrome Partitioning .	14
<b>5</b>	<b>Number Theory</b>	<b>5</b>	11.7	Subset Sum . . . . .	15
5.1	Modular Arithmetic . . .	5	<b>12</b>	<b>Array Techniques</b>	<b>15</b>
5.2	Sieve of Eratosthenes . . .	6	12.1	Prefix Sum . . . . .	15
5.3	Prime Factorization . . .	6	12.2	Difference Array . . . . .	15
5.4	Chinese Remainder The- orem . . . . .	6	12.3	Sliding Window . . . . .	16
5.5	Euler's Totient Function .	6	<b>13</b>	<b>Advanced Data Structures</b>	<b>16</b>
5.6	Fast Exponentiation with Matrix . . . . .	7	13.1	Segment Tree . . . . .	16
<b>6</b>	<b>Graph Algorithms</b>	<b>7</b>	13.2	Fenwick Tree (Binary In- dexed Tree) . . . . .	17
6.1	Graph Representation . .	7	13.3	Trie (Prefix Tree) . . . . .	17
6.2	BFS (Breadth-First Search) . . . . .	7	13.4	Treap (Randomized Bal- anced BST) . . . . .	17
6.3	DFS (Depth-First Search)	7	<b>14</b>	<b>Bit Manipulation</b>	<b>18</b>
6.4	Strongly Connected Components (SCC) . . . .	8	<b>15</b>	<b>Matrix Operations</b>	<b>19</b>
6.5	Bridges and Articulation Points . . . . .	9	15.1	Matrix Multiplication . .	19
6.6	Lowest Common Ances- tor (LCA) . . . . .	9	15.2	Matrix Exponentiation . .	19
<b>7</b>	<b>Shortest Path Algorithms</b>	<b>10</b>	<b>16</b>	<b>Miscellaneous Tips</b>	<b>20</b>
7.1	Dijkstra's Algorithm . . .	10			
7.2	Bellman-Ford Algorithm .	10			
7.3	Floyd-Warshall Algorithm	10			
7.4	Minimum Spanning Tree .	11			

16.1 Python-Specific Optimizations . . . . .	20	17.3 Point in Polygon . . . . .	23
16.2 Useful Libraries . . . . .	20	17.4 Closest Pair of Points . . . . .	23
16.3 Common Patterns . . . . .	20		
16.4 Common Pitfalls . . . . .	20	<b>18 Network Flow</b>	<b>23</b>
16.5 Time Complexity Reference . . . . .	21	18.1 Maximum Flow - Edmonds-Karp (BFS-based Ford-Fulkerson) . . . . .	23
<b>17 Computational Geometry</b>	<b>22</b>	18.2 Dinic's Algorithm (Faster)	24
17.1 Basic Geometry . . . . .	22	18.3 Min Cut . . . . .	25
17.2 Convex Hull . . . . .	22	18.4 Bipartite Matching . . . . .	25

## 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
15 #      newline)
16 s = input().strip()
17 words = input().split()
18
19 # Multiple test cases pattern
20 t = int(input())
21 for _ in range(t):
22     # process each test case
23
24 # Print without newline
25 print(x, end=' ')
26
27 # Formatted output with precision
28 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)] #
4  #      Correct way!
5  # List comprehension - concise and

```

```

efficient
squares = [x**2 for x in range(n)]
evens = [x for x in arr if x % 2 == 0]

# Sorting - O(n log n)
arr.sort() # in-place, modifies arr
arr.sort(reverse=True) # descending
arr.sort(key=lambda x: (x[0], -x[1])) #
custom
sorted_arr = sorted(arr) # returns new
list

# Binary search in sorted array
from bisect import bisect_left,
bisect_right
idx = bisect_left(arr, x) # leftmost
position
idx = bisect_right(arr, x) # rightmost
position

# Common operations
arr.append(x) # O(1) amortized
arr.pop() # O(1) - remove last
arr.pop(0) # O(n) - remove first
(slow!)
arr.reverse() # O(n) - in-place
arr.count(x) # O(n) - count
occurrences
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
22         # useful)
23         while dq and arr[dq[-1]] < arr[i]:
24             dq.pop()
25
26         dq.append(i)
27         if i >= k - 1:
28             result.append(arr[dq[0]])
29
30     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,
22 # etc.
23 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
5 # default
6 count = defaultdict(int) # 0 default
7
8 # Counter - count elements efficiently
9 cnt = Counter(arr)
10 cnt['x'] += 1
11 most_common = cnt.most_common(k) # k
12 # most frequent
13
14 # Dictionary operations
15 d = {}
16 d.get(key, default_val)
17 d.setdefault(key, default_val)
18 for k, v in d.items():
19     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
4 # exists
5 s.discard(x) # O(1), no error if not
6 # exists
7
8 # Set operations - all O(n)
9 a | b # union
10 a & b # intersection
11 a - b # difference
12 a ^ b # symmetric difference
13
14 # Ordered set workaround
15 from collections import OrderedDict
16 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
4     ends
5 s.lstrip() # remove left whitespace
6 s.rstrip() # remove right whitespace
7 s.split(delimiter)
8 delimiter.join(list)
9 s.replace(old, new)
10 s.startswith(prefix)
11 s.endswith(suffix)
12 s.isdigit(), s.isalpha(), s.isalnum()
13
14 # String building - EFFICIENT O(n)
15 result = []
16 for x in data:
17     result.append(str(x))
18 s = ''.join(result)
19
20 # String concatenation - SLOW O(n^2)
21 # s = ""
22 # for x in data:
23 #     s += str(x) # Don't do this!
24
25 # ASCII values
26 ord('a') # 97
27 chr(97) # 'a'
28
29 # String to character array (for
30     mutations)
31 chars = list(s)
32 chars[0] = 'x'
33 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
3     #   which is Suffix)
4     def build_lps(pattern):
5         m = len(pattern)
6         lps = [0] * m
7         length = 0 # Length of previous
8             longest prefix
9         i = 1
10
11         while i < m:
12             if pattern[i] == pattern[
13                 length]:
14                 length += 1
15                 lps[i] = length
16                 i += 1
17             else:
18                 if length != 0:
19                     length = lps[length
20                         - 1]
21                 else:
22                     lps[i] = 0
23                     i += 1
24
25         return lps
26
27 n, m = len(text), len(pattern)
28 lps = build_lps(pattern)
29
30 matches = []
31 i = j = 0 # Indices for text and
32     pattern
33
34 while i < n:
35     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] !=
38             pattern[j]:
39             if j != 0:
40                 j = lps[j - 1]
41             else:
42                 i += 1
43
44 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 -
9                 1])
10
11         while i + z[i] < n and s[z[i]]
12             == s[i + z[i]]:
13             z[i] += 1
14
15         if i + z[i] - 1 > r:
16             l, r = i, i + z[i] - 1
17
18 return z
19
20 # Pattern matching using Z-algorithm
21 def z_search(text, pattern):
22     # Concatenate pattern + '$' + text
23     s = pattern + '$' + text
24     z = z_algorithm(s)
25
26     matches = []
27     m = len(pattern)
28
29     for i in range(m + 1, len(s)):
30         if z[i] == m:
31             matches.append(i - m - 1)
32
33 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 *
14             BASE +
15             ord(pattern[i]))
16         % MOD
17         if i < m - 1:
18             power = (power * BASE) % MOD
19
20     # Rolling hash
21     text_hash = 0
22     matches = []
23     for i in range(n):
24         # Add new character
25         text_hash = (text_hash * BASE +
26             ord(text[i])) % MOD
27
28         # Remove old character if window
29         # full
30         if i >= m:
31             text_hash = (text_hash -
32                 ord(text[i - m])
33                 * power) % MOD
34             text_hash = (text_hash + MOD
35                 ) % MOD
36
37         # Check match
38         if i >= m - 1 and text_hash ==
39             pattern_hash:
40             # Verify actual match (avoid
41             # hash collision)
42             if text[i - m + 1:i + 1] ==
43                 pattern:
44                 matches.append(i - m +
45                     1)
46
47     return matches

```

## 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
6  divisor
7  math.lcm(a, b) # Python 3.9+
8  math.sqrt(x)
9  math.log(x), math.log2(x), math.log10(x)
10
11 # Powers
12 x ** y
13 pow(x, y, mod) # (x^y) % mod -
14 # efficient modular exp
15
16 # Infinity
17 float('inf'), float('-inf')
18
19 # Custom GCD using Euclidean algorithm -
20 # O(log min(a,b))
21 def gcd(a, b):
22     while b:
23         a, b = b, a % b
24     return a
25
26 def lcm(a, b):
27     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
13     (n, r) = C(n, n-r)
14     num = den = 1
15     for i in range(r):
16         num *= (n - i)
17         den *= (i + 1)
18     return num // den
19
20 # Precompute factorials with modulo
21 MOD = 10**9 + 7
22 def modfact(n):
23     fact = [1] * (n + 1)
24     for i in range(1, n + 1):
25         fact[i] = fact[i-1] * i % MOD
26     return fact
27
28 # Modular combination using precomputed
29 # factorials
30 # First precompute inverse factorials
31 def compute_inv_factorials(n, mod):
32     fact = modfact(n)
33     inv_fact = [1] * (n + 1)
34     inv_fact[n] = pow(fact[n], mod - 2,
35         mod)
36     for i in range(n - 1, -1, -1):
37         inv_fact[i] = inv_fact[i + 1] *
38             (i + 1) % mod
39     return fact, inv_fact
40
41 def modcomb(n, r, fact, inv_fact, mod):
42     if r > n or r < 0: return 0
43     return fact[n] * inv_fact[r] % mod *
44         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
2  # Theorem
3  # Only works when mod is prime
4  # a^(-1) = a^(mod-2) (mod p)
5  def modinv(a, mod):
6      return pow(a, mod - 2, mod)
7
8  # Extended Euclidean Algorithm
9  # Returns (gcd, x, y) where ax + by =
10 # gcd(a,b)
11 # Can find modular inverse for any
12 # 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

```

```

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

```

## 5.2 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                 ):
10                 is_prime[j] = False
11
12     return is_prime
13
14 # Get list of primes
15 primes = [i for i in range(n+1) if
16     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
30             i += 1
31     return count
32
33 # Sum of divisors
34 def sum_divisors(n):
35     total = 0
36     i = 1

```

## 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     ):
3     # Solve x = remainders[i] (mod
4     # moduli[i])
5     # Assumes moduli are pairwise
6     # coprime
7
8     def extgcd(a, b):
9         if b == 0:
10             return a, 1, 0
11         g, x1, y1 = extgcd(b, a % b)
12         return g, y1, x1 - (a // b) * y1
13
14     total = 0
15     prod = 1
16     for m in moduli:
17         prod *= m
18
19     for r, m in zip(remainders, moduli):
20         p = prod // m
21         g, inv, _ = extgcd(p, m)
22         # inv may be negative, normalize
23         it
24         inv = (inv % m + m) % m
25         total += r * inv * p
26
27     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] =
22     i initially

```

```

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 * (
27                     i - 1)
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) %
9              mod
10             base = (base * base) % mod
11             exp >>= 1
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,
2     deque
3
4  # Unweighted graph
5  graph = defaultdict(list)
6  for _ in range(m):
7      u, v = map(int, input().split())
8      graph[u].append(v)
9      graph[v].append(u) # for undirected
10
11 # Weighted graph - store (neighbor,
12    weight) tuples
13 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[

```

```

14         node] + 1
15     return dist
16
17 # Grid BFS - common in maze/path
18    problems
19 def grid_bfs(grid, start):
20     n, m = len(grid), len(grid[0])
21     visited = [[False] * m for _ in
22         range(n)]
23     queue = deque([start])
24     visited[start[0]][start[1]] = True
25
26     # 4 directions: right, down, left,
27     up
28     dirs = [(0,1), (1,0), (0,-1), (-1,0)
29         ]
30
31     while queue:
32         x, y = queue.popleft()
33
34         for dx, dy in dirs:
35             nx, ny = x + dx, y + dy
36
37             # Check bounds and validity
38             if (0 <= nx < n and 0 <= ny
39                 < m
40                 and not visited[nx][ny]
41                 and grid[nx][ny] != '#'):
42                 visited[nx][ny] = True
43                 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
22                 visited:
23                     stack.append(
24                     neighbor)
25
26 # Cycle detection in undirected graph
27 def has_cycle(graph, n):
28     visited = [False] * n
29
30     def dfs(node, parent):
31         visited[node] = True

```

```

30         for neighbor in graph[node]:
31             if not visited[neighbor]:
32                 if dfs(neighbor, node):
33                     return True
34             # Back edge to non-parent =
35             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
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         for neighbor in graph[node]:
103             if color[neighbor] ==
104             -1:
105                 color[neighbor] = 1
106             - color[node]
107             queue.append(
108             neighbor)
109             elif color[neighbor] ==
110             color[node]:
111                 return False
112
113         return True
114
115     for i in range(n):
116         if color[i] == -1:
117             if not bfs(i):
118                 return False
119
120     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
22            [v], lowlink[w])
23            elif on_stack[w]:
24                lowlink[v] = min(lowlink
25                [v], index[w])
26
27        if lowlink[v] == index[v]:
28            scc = []
29            while True:
30                w = stack.pop()
31                on_stack[w] = False
32                scc.append(w)
33                if w == v:
34                    break
35            sccs.append(scc)
36
37    for v in range(n):
38        if not index_initialized[v]:
39            strongconnect(v)
40
41    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
24             elif v != parent[u]:
25                 low[u] = min(low[u], disc[v])
26
27     for i in range(n):
28         if not visited[i]:
29             dfs(i)
30
31     return bridges
32
33 def find_articulation_points(graph, n):
34     visited = [False] * n
35     disc = [0] * n
36     low = [0] * n
37     parent = [-1] * n
38     time = [0]
39     ap = set()
40
41     def dfs(u):
42         children = 0
43         visited[u] = True
44         disc[u] = low[u] = time[0]
45         time[0] += 1
46
47         for v in graph[u]:
48             if not visited[v]:
49                 children += 1
50                 parent[v] = u
51                 dfs(v)
52                 low[u] = min(low[u], low[v])
53
54             # Articulation point conditions
55             if parent[u] == -1 and children > 1:
56                 ap.add(u)
57             if parent[u] != -1 and low[v] >= disc[u]:
58                 ap.add(u)
59             elif v != parent[u]:
60                 low[u] = min(low[u], disc[v])
61
62         return ap

```

```

60 for i in range(n):
61     if not visited[i]:
62         dfs(i)
63
64 return list(ap)
65

```

## 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
10
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[self.parent[i][j-1]][j-1]
27
28     def lca(self, u, v):
29         # Make u deeper
30         if self.depth[u] < self.depth[v]:
31             u, v = v, u
32
33         # Bring u to same level as v
34         diff = self.depth[u] - self.depth[v]
35         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 def dist(self, u, v):
51     # Distance between two nodes
52     l = self.lca(u, v)
53     return self.depth[u] + self.
54     depth[v] - 2 * self.depth[l]

```

```

46 return dist, parent
47
48 def reconstruct_path(parent, target):
49     path = []
50     while target != -1:
51         path.append(target)
52         target = parent[target]
53     return path[::-1]
54

```

## 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
15         # better distance
16         if d > dist[node]:
17             continue
18
19         # Relax edges
20         for neighbor, weight in graph[
21             node]:
22             new_dist = dist[node] +
23             weight
24             if new_dist < dist[neighbor
25             ]:
26                 dist[neighbor] =
27                 new_dist
28                 heapq.heappush(heap, (
29                 new_dist, neighbor))
30
31     return dist
32
33 # Path reconstruction
34 def dijkstra_with_path(graph, start, n):
35     dist = [float('inf')] * n
36     parent = [-1] * n
37     dist[start] = 0
38     heap = [(0, start)]
39
40     while heap:
41         d, node = heapq.heappop(heap)
42         if d > dist[node]:
43             continue
44
45         for neighbor, weight in graph[
46             node]:
47             new_dist = dist[node] +
48             weight
49             if new_dist < dist[neighbor
50             ]:
51                 dist[neighbor] =
52                 new_dist
53                 parent[neighbor] = node
54                 heapq.heappush(heap, (
55                 new_dist, neighbor))

```

### 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')
10             and \
11                 dist[u] + w < dist[v]:
12                 dist[v] = dist[u] + w
13
14     # Check for negative cycles
15     for u, v, w in edges:
16         if dist[u] != float('inf') and \
17             dist[u] + w < dist[v]:
18             return None # Negative
19             cycle exists

```

### 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
4             range(n)]
5
6     for i in range(n):
7         dist[i][i] = 0
8
9     for u, v, w in edges:
10         dist[u][v] = min(dist[u][v], w)
11
12     # Dynamic programming
13     for k in range(n): # Intermediate
14         vertex
15         for i in range(n):
16             for j in range(n):
17                 dist[i][j] = min(dist[i
18                 ][j],
19                 dist[i][k] + dist[k][j])
20
21     return dist
22
23 # Check for negative cycle
24 def has_negative_cycle(dist, n):
25     for i in range(n):
26         if dist[i][i] < 0:
27             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,
13                             weight))
14
15     return mst_weight, mst_edges
16
17 class UnionFind:
18     def __init__(self, n):
19         self.parent = list(range(n))
20         self.rank = [0] * n
21
22     def find(self, x):
23         if self.parent[x] != x:
24             self.parent[x] = self.find(
25                 self.parent[x])
26         return self.parent[x]
27
28     def union(self, x, y):
29         px, py = self.find(x), self.find(
30             y)
31         if px == py:
32             return False
33         if self.rank[px] < self.rank[py]:
34             px, py = py, px
35         self.parent[py] = px
36         if self.rank[px] == self.rank[py]:
37             self.rank[px] += 1
38         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
8     mst_weight = 0
9
10    while min_heap:
11        weight, u = heapq.heappop(
12            min_heap)
13
14        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,
21                             (w, v))
22
23     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
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
11    # dependencies
12    queue = deque([i for i in range(n)
13                    if indegree[i] == 0])
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
25        # dependencies
26        if indegree[neighbor] == 0:
27            queue.append(neighbor)
28
29    # If not all nodes processed, cycle
30    # exists
31    return result if len(result) == n
32    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
14     # all descendants
15     stack.append(node)
16
17     # Process all components
18     for i in range(n):
19         if not visited[i]:
20             dfs(i)
21
22     # Reverse stack gives topological
23     # order
24     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
4          # initially
5          self.parent = list(range(n))
6          # Rank for union by rank
7          # optimization
8          self.rank = [0] * n
9
10     def find(self, x):
11         # Path compression: point
12         # directly to root
13         if self.parent[x] != x:
14             self.parent[x] = self.find(
15                 self.parent[x])
16         return self.parent[x]
17
18     def union(self, x, y):
19         # Find roots
20         px, py = self.find(x), self.find(
21             y)
22
23         # Already in same set
24         if px == py:
25             return False
26
27         # Union by rank: attach smaller
28         # tree under larger
29         if self.rank[px] < self.rank[py]:
30             px, py = py, px

```

```

26     self.parent[py] = px
27
28     # Increase rank if trees had
29     # equal rank
30     if self.rank[px] == self.rank[py]:
31         self.rank[px] += 1
32
33     return True
34
35     def connected(self, x, y):
36         return self.find(x) == self.find(
37             y)
38
39     # Count number of disjoint sets
40     def count_sets(self):
41         return len(set(self.find(i)
42             for i in range(len(
43                 self.parent))))
44
45     # Example: Detect cycle in undirected
46     # graph
47     def has_cycle_uf(edges, n):
48         uf = UnionFind(n)
49         for u, v in edges:
50             if uf.connected(u, v):
51                 return True # Cycle found
52             uf.union(u, v)
53         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)
2  # is True
3  def binary_search_first(left, right,
4  check):
5      while left < right:
6          mid = (left + right) // 2
7
8          if check(mid):
9              right = mid # Could be
10             answer, search left
11         else:
12             left = mid + 1 # Not answer
13             , search right
14
15     return left
16
17 # Find LAST position where check(mid) is
18 # True
19 def binary_search_last(left, right,
20 check):
21     while left < right:
22         mid = (left + right + 1) // 2 #
23         Round up!
24
25         if check(mid):
26             left = mid # Could be
27             answer, search right
28         else:
29             right = mid - 1 # Not
30             answer, search left

```

```

22     return left
23
24 # Example: Integer square root
25 def sqrt_binary(n):
26     left, right = 0, n
27
28     while left < right:
29         mid = (left + right + 1) // 2
30
31         if mid * mid <= n:
32             left = mid # mid might be
33             answer
34         else:
35             right = mid - 1
36
37     return left
38
39 # Binary search on answer - common
40 # pattern
41 def min_days_to_make_bouquets(bloomDay,
42                               m, k):
43     # Can we make m bouquets in 'days'
44     # days?
45     def can_make(days):
46         bouquets = consecutive = 0
47         for bloom in bloomDay:
48             if bloom <= days:
49                 consecutive += 1
50                 if consecutive == k:
51                     bouquets += 1
52                     consecutive = 0
53             else:
54                 consecutive = 0
55         return bouquets >= m
56
57     if len(bloomDay) < m * k:
58         return -1
59
60 # Binary search on number of days
61 return binary_search_first(
62     min(bloomDay), max(bloomDay),
63     can_make)

```

```

13     else:
14         dp[idx] = x # Better
15         ending for this length
16
17     return len(dp)
18
19 # LIS with actual sequence
20 def lis_with_sequence(arr):
21     from bisect import bisect_left
22
23     n = len(arr)
24     dp = []
25     parent = [-1] * n
26     dp_idx = [] # indices in dp
27
28     for i, x in enumerate(arr):
29         idx = bisect_left(dp, x)
30
31         if idx == len(dp):
32             dp.append(x)
33             dp_idx.append(i)
34         else:
35             dp[idx] = x
36             dp_idx[idx] = i
37
38         if idx > 0:
39             parent[i] = dp_idx[idx - 1]
40
41 # Reconstruct sequence
42 result = []
43 idx = dp_idx[-1]
44 while idx != -1:
45     result.append(arr[idx])
46     idx = parent[idx]
47
48 return result[::-1]

```

## 11 Dynamic Programming

**Description:** Solve problems by breaking them into overlapping sub-problems. 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
5     # LIS of length i+1
6     dp = []
7
8     for x in arr:
9         # Find position to place x
10        idx = bisect_left(dp, x)
11
12        if idx == len(dp):
13            dp.append(x) # Extend LIS

```

## 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
4     # items, weight <= w
5     dp = [[0] * (capacity + 1) for _ in
6           range(n + 1)]
7
8     for i in range(1, n + 1):
9         for w in range(capacity + 1):
10            # Don't take item i-1
11            dp[i][w] = dp[i-1][w]
12
13            # Take item i-1 if it fits
14            if weights[i-1] <= w:
15                dp[i][w] = max(
16                    dp[i][w],
17                    dp[i-1][w - weights[
18                        i-1]] + values[i-1]
19                )
20
21     return dp[n][capacity]
22
23 # Space-optimized O(capacity)
24 def knapsack_optimized(weights, values,
25                         capacity):
26     dp = [0] * (capacity + 1)
27
28     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     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 # Reconstruct LCS
14
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              :
9              dp[i] = min(dp[i], dp[i -
    coin] + 1)
10
11     return dp[amount] if dp[amount] !=
    float('inf') else -1
12
13 # Count ways
14 def coin_change_ways(coins, amount):
15     dp = [0] * (amount + 1)
16     dp[0] = 1
17
18     for coin in coins:
19         for i in range(coin, amount + 1):
20             :
21             dp[i] += dp[i - coin]
22
23     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
17                             == 2 or
18                             is_pal[i
19                             +1][j-1])
20
21 # dp[i] = min cuts for s[0:i+1]
22 dp = [float('inf')] * n
23
24 for i in range(n):
25     if is_pal[0][i]:
26         dp[i] = 0
27     else:
28         for j in range(i):
29             if is_pal[j+1][i]:
30                 dp[i] = min(dp[i],
31                             dp[j] + 1)
32
33 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 _
4         in range(n + 1)]
5
6     # Base case: sum 0 is always
7     # achievable
8     for i in range(n + 1):
9         dp[i][0] = True
10
11     for i in range(1, n + 1):
12         for s in range(target + 1):
13             # Don't take arr[i-1]
14             dp[i][s] = dp[i-1][s]
15
16             # Take arr[i-1] if possible
17             if s >= arr[i-1]:
18                 dp[i][s] = dp[i][s] or
19                 dp[i-1][s - arr[i-1]]
20
21     return dp[n][target]
22
23 # Space optimized
24 def subset_sum_optimized(arr, target):
25     dp = [False] * (target + 1)
26     dp[0] = True
27
28     for num in arr:
29         for s in range(target, num - 1,
30             -1):
31             dp[s] = dp[s] or dp[s - num]
32
33     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
13         range(n + 1)]
14
15     for i in range(1, n + 1):
16         for j in range(1, m + 1):
17             prefix[i][j] = (matrix[i-1][
18                 j-1] +
19                             prefix[i-1][j
20                             ] +
21                             prefix[i][j
22                             -1] -
23                             prefix[i-1][j
24                             -1])
25
26     return prefix
27
28 # Rectangle sum from (x1,y1) to (x2,y2)
  inclusive
29 def rect_sum(prefix, x1, y1, x2, y2):
30     return (prefix[x2+1][y2+1] -
31             prefix[x1][y2+1] -
32             prefix[x2+1][y1] +
33             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,
```



```

55         val)
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             :
16                 node.children[char] =
           TrieNode()
17                 node = node.children[char]
18                 node.is_end = True
19
20     def search(self, word):
21         # Exact word search - O(len(word))
22         node = self.root
23         for char in word:
24             if char not in node.children
25             :
26                 return False
27                 node = node.children[char]
28                 return node.is_end
29
30     def starts_with(self, prefix):
31         # Prefix search - O(len(prefix))
32         node = self.root
33         for char in prefix:
34             if char not in node.children
35             :
36                 return False
37                 node = node.children[char]
38                 return True
39
40     # Find all words with given prefix
41     def words_with_prefix(self, prefix):
42         node = self.root
43         for char in prefix:
44             if char not in node.children
45             :
46                 return []
47                 node = node.children[char]
48
49         # DFS to collect all words
50         words = []
51         def dfs(n, path):
52             if n.is_end:
53                 words.append(prefix +
           path)
54             for char, child in n.
           children.items():
55                 dfs(child, path + char)
56
57         dfs(node, "")
58         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    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,
78
node), r)
79
80    def remove(self, key):
81        self.root = _erase(self.root, key)
82
83    def kth_smallest(self, k):
84        # 0-indexed k
85        t = self.root
86        while t:
87            ls = _sz(t.left)
88            if k < ls:
89                t = t.left
90            elif k == ls:
91                return t.key
92            else:
93                k -= ls + 1
94                t = t.right
95        return None # k out of range
96
97    def count_less_than(self, key):
98        # number of keys < key
99        t, cnt = self.root, 0
100        while t:
101            if key <= t.key:
102                t = t.left
103            else:
104                cnt += 1 + _sz(t.left)
105                t = t.right
106        return cnt
107
108    def lower_bound(self, key):
109        # smallest key >= key; returns None if none
110        t, ans = self.root, None
111        while t:
112            if t.key >= key:
113                ans = t.key
114                t = t.left
115            else:
116                t = t.right
117        return ans
118
119    # Usage example
120    T = Treap()
121    for x in [5, 1, 7, 3]:
122        T.insert(x)
123    T.contains(3) # True
124    T.kth_smallest(1) # 3 (0-indexed)
125    T.count_less_than(6) # 3 (1,3,5)
126    T.remove(5)
127    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,
13                                M, mod)
14            M = matmul_mod(M, M, mod)
15            n >>= 1
16
17    return result
18
19 # Example: Fibonacci using matrix
    exponentiation
20 # F(n) = [[1,1],[1,0]]^n
21 def fibonacci(n, mod):
22     if n == 0: return 0
23     if n == 1: return 1
24
25     M = [[1, 1], [1, 0]]
26     result = matpow(M, n - 1, mod)
27     return result[0][0]
28
29 # Linear recurrence: a(n) = c1*a(n-1) +
    c2*a(n-2) + ...
30 # Build transition matrix and use matrix
    exponentiation
def linear_recurrence(coeffs, init, n,

```

```

31     mod):
32     k = len(coeffs)
33
34     if n < k:
35         return init[n]
36
37     # Transition matrix
38     # [a(n), a(n-1), ..., a(n-k+1)]
39     M = [[0] * k for _ in range(k)]
40     M[0] = coeffs # First row
41     for i in range(1, k):
42         M[i][i-1] = 1 # Identity for
43         # shifting
44
45     # Initial state vector [a(k-1), a(k-2), ..., a(0)]
46     state = init[k-1::-1]
47
48     # M^(n-k+1)
49     result_matrix = matpow(M, n - k + 1,
50                             mod)
51
52     # Multiply with initial state
53     result = 0
54     for i in range(k):
55         result = (result + result_matrix
56                  [0][i] * state[i]) % mod
57
58     return result
59
60 # Example: Tribonacci T(n) = T(n-1) + T(
61 # n-2) + T(n-3)
62 def tribonacci(n, mod):
63     if n == 0: return 0
64     if n == 1 or n == 2: return 1
65
66     coeffs = [1, 1, 1]
67     init = [0, 1, 1]
68     return linear_recurrence(coeffs,
69                               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/
6 # DP
7 sys.setrecursionlimit(10**6)
8
9 # Threading for higher stack limit (
10 # CAUTION: use carefully)
11 import threading
12 threading.stack_size(2**26) # 64MB
13 sys.setrecursionlimit(2**20)
14
15 # Deep copy (be careful with performance
16 # )
17 from copy import deepcopy
18 new_list = deepcopy(old_list)
19
20 # Fast output (for printing large
21 # results)
22 import sys
23 print = sys.stdout.write # Only use for
24 # string output

```

## 16.2 Useful Libraries

```

1 # Iterator tools - powerful combinations
2 from itertools import *
3
4 # permutations(iterable, r) - all r-
5 # length permutations
6 perms = list(permutations([1,2,3], 2))
7 # [(1,2), (1,3), (2,1), (2,3), (3,1),
8 # (3,2)]
9
10 # combinations(iterable, r) - r-length
11 # combinations
12 combs = list(combinations([1,2,3], 2))
13 # [(1,2), (1,3), (2,3)]
14
15 # product - cartesian product
16 prod = list(product([1,2], ['a','b']))
17 # [(1,'a'), (1,'b'), (2,'a'), (2,'b')]
18
19 # accumulate - running totals
20 acc = list(accumulate([1,2,3,4]))
21 # [1, 3, 6, 10]
22
23 # chain - flatten iterables
24 chained = list(chain([1,2], [3,4]))
25 # [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
7 # positive
8 any(x > 0 for x in arr) # True if any
9 # positive
10
11 # Zip - parallel iteration
12 for a, b in zip(list1, list2):
13     pass
14
15 # Enumerate - index and value
16 for i, val in enumerate(arr):
17     print(f"arr[{i}] = {val}")
18
19 # Custom comparison function
20 from functools import cmp_to_key
21
22 def compare(a, b):
23     # Return -1 if a < b, 0 if equal, 1
24     # if a > b
25     if a + b > b + a:
26         return -1
27     return 1
28
29 arr.sort(key=cmp_to_key(compare))
30
31 # DefaultDict with lambda
32 from collections import defaultdict
33 d = defaultdict(lambda: float('inf'))
34
35 # Multiple assignment
36 a, b = b, a # Swap
37 a, *rest, b = [1,2,3,4,5] # a=1, rest
38 # = [2,3,4], b=5

```

## 16.4 Common Pitfalls

```

1 # Integer division - floors toward
2 # 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
15 # rows same object
16 matrix[0][0] = 1      # Changes all
17 # rows!
18
19 # Correct way
20 matrix = [[0] * m for _ in range(n)]
21
22 # Float comparison - don't use ==
23 a, b = 0.1 + 0.2, 0.3
24 print(a == b)        # False!
25
26 # Use epsilon comparison
27 eps = 1e-9
28 print(abs(a - b) < eps) # True
29
30 # String immutability
31 s = "abc"
32 # s[0] = 'd' # ERROR!
33 s = 'd' + s[1:] # OK
34
35 # For many string mutations, use list
36 chars = list(s)
37 chars[0] = 'd'
38 s = ''.join(chars)
39
40 # Mutable default arguments - dangerous!
41 def func(arr=[]): # WRONG!
42     arr.append(1)
43     return arr
44
45 # Each call modifies same list
46 print(func()) # [1]
47 print(func()) # [1, 1]
48
49 # Correct way
50 def func(arr=None):
51     if arr is None:
52         arr = []
53     arr.append(1)
54     return arr
55
56 # Generator expressions save memory
57 sum(x*x for x in range(10**6)) # Memory
58 # efficient
59
60 # vs
61 sum([x*x for x in range(10**6)]) #
62 # Creates full list
63
64 # Ternary operator
65 x = a if condition else b
66
67 # Dictionary get with default
68 count = d.get(key, 0) + 1
69
70 # Matrix rotation 90 degrees clockwise
71 def rotate_90(matrix):
72     return [list(row) for row in zip(*
73         matrix[::-1])]
74
75 # Matrix transpose

```

```

70 def transpose(matrix):
71     return [list(row) for row in zip(*
72         matrix)]

```

## 16.5 Time Complexity Reference

```

1 # Common time complexities (Python,
2 # rough guides for 1--2s limits):
3 # O(1), O(log n): instant
4 # O(n): usually fine up to ~10^7
5 # operations (~1s)
6 # O(n log n): OK for n up to several
7 # 10^5 depending on constants
8 # O(n sqrt(n)): risky in Python (may be
9 # OK for n up to a few 10^4 with low
10 # constants)
11 # O(n^2): often TLE for n > 10^4
12 # O(2^n): TLE for n > 20 (unless heavy
13 # pruning/memoization)
14 # O(n!): TLE for n > 11
15
16 # Input size guidelines (Python-focused)
17 :
18 # n <= 12: O(n!) (brute-force
19 # permutations)
20 # n <= 20: O(2^n) (subset DP / bitmask
21 # DP with n up to ~20)
22 # n <= 500: O(n^3) may sometimes pass
23 # for small constants
24 # n <= 5000: O(n^2) borderline; optimize
25 # heavily
26 # n <= 10^6: O(n log n) common; O(n)
27 # preferred when possible
28 # n <= 10^7: O(n) may be ok for tight
29 # loops
30 # n > 10^7: aim for O(n) with very low
31 # 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 approaches 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  $\text{cost} = c \cdot f(n)$ . If  $\text{cost} > \text{time\_limit} \times \text{ops\_per\_sec}$ , 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
7                     + (p1[1] - p2[1])**2)
8
9 def cross_product(O, A, B):
10     # Cross product of vectors OA and OB
11     # Positive: counter-clockwise
12     # Negative: clockwise
```

```
# Zero: collinear
return (A[0] - O[0]) * (B[1] - O[1])
- \
    (A[1] - O[1]) * (B[0] - O[0])

def dot_product(A, B, C, D):
    # Dot product of vectors AB and CD
    return (B[0] - A[0]) * (D[0] - C[0])
    + \
        (B[1] - A[1]) * (D[1] - C[1])

# Check if point is on segment
def on_segment(p, q, r):
    # Check if q lies on segment pr
    return (q[0] <= max(p[0], r[0]) and
            q[0] >= min(p[0], r[0]) and
            q[1] <= max(p[1], r[1]) and
            q[1] >= min(p[1], r[1]))

# Segment intersection
def segments_intersect(p1, q1, p2, q2):
    o1 = cross_product(p1, q1, p2)
    o2 = cross_product(p1, q1, q2)
    o3 = cross_product(p2, q2, p1)
    o4 = cross_product(p2, q2, q1)

    # General case
    if o1 * o2 < 0 and o3 * o4 < 0:
        return True

    # Special cases (collinear)
    if o1 == 0 and on_segment(p1, p2, q1):
        return True
    if o2 == 0 and on_segment(p1, q2, q1):
        return True
    if o3 == 0 and on_segment(p2, p1, q2):
        return True
    if o4 == 0 and on_segment(p2, q1, q2):
        return True

    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
4     # then y
5
6     if len(points) <= 2:
7         return points
8
9     # Build lower hull
10    lower = []
11    for p in points:
12        while (len(lower) >= 2 and
13              cross_product(lower[-2],
14                            lower[-1], p) <= 0):
15            lower.pop()
16            lower.append(p)
17
18    # Build upper hull
19    upper = []
20    for p in reversed(points):
21        while (len(upper) >= 2 and
22              cross_product(upper[-2],
23                            upper[-1], p) <= 0):
```

```

21         upper.pop()
22         upper.append(p)
23
24         # Remove last point (duplicate of
25         # first)
26         return lower[:-1] + upper[:-1]
27
28     # Convex hull area
29     def polygon_area(points):
30         # Shoelace formula
31         n = len(points)
32         area = 0
33
34         for i in range(n):
35             j = (i + 1) % n
36             area += points[i][0] * points[j][1]
37             area -= points[j][0] * points[i][1]
38
39     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 -
16                                     p1y) * (p2x - p1x) / \
17                                     (p2y -
18                                     p1y) + p1x
19
20                     if p1x == p2x or x
21                     <= xinters:
22                         inside = not
23                         inside
24
25         p1x, p1y = p2x, p2y
26
27     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
3      =lambda p: p[0])
4      points_sorted_y = sorted(points, key
5      =lambda p: p[1])
6
7      def closest_recursive(px, py):
8          n = len(px)
9
10         # Base case: brute force
11         if n <= 3:
12             min_dist = float('inf')
13             for i in range(n):

```

```

12         for j in range(i + 1, n)
13         :
14             min_dist = min(
15             min_dist, dist(px[i], px[j]))
16         return min_dist
17
18         # Divide
19         mid = n // 2
20         midpoint = px[mid]
21
22         pyl = [p for p in py if p[0] <=
23         midpoint[0]]
24         pyr = [p for p in py if p[0] >
25         midpoint[0]]
26
27         # Conquer
28         dl = closest_recursive(px[:mid],
29         pyl)
30         dr = closest_recursive(px[mid:],
31         pyr)
32         d = min(dl, dr)
33
34         # Combine: check strip
35         strip = [p for p in py if abs(p
36         [0] - midpoint[0]) < d]
37
38         for i in range(len(strip)):
39             j = i + 1
40             while j < len(strip) and
41             strip[j][1] - strip[i][1] < d:
42                 d = min(d, dist(strip[i]
43                 ], strip[j]))
44             j += 1
45
46         return d
47
48     return closest_recursive(
49         points_sorted_x, points_sorted_y)

```

## 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,
2  defaultdict
3
4  def max_flow(graph, source, sink, n):
5      # graph[u][v] = capacity from u to v
6      # Build residual graph
7      residual = defaultdict(lambda:
8      defaultdict(int))
9
10     for u in graph:
11         for v in graph[u]:
12             residual[u][v] = graph[u][v]
13
14     def bfs_path():
15         # Find augmenting path using BFS
16         parent = {source: None}
17         visited = {source}
18         queue = deque([source])
19
20         while queue:
21             u = queue.popleft()
22
23             if u == sink:
24                 # Reconstruct path
25                 path = []

```

```

23         while parent[u] is not
24             None:
25                 path.append((parent[
26                     u], u))
27                 u = parent[u]
28                 return path[::-1]
29
30         for v in range(n):
31             if v not in visited and
32                 residual[u][v] > 0:
33                 visited.add(v)
34                 parent[v] = u
35                 queue.append(v)
36
37         return None
38
39         max_flow_value = 0
40
41         # Find augmenting paths
42         while True:
43             path = bfs_path()
44             if path is None:
45                 break
46
47             # Find minimum capacity along
48             # path
49             flow = min(residual[u][v] for u,
50                         v in path)
51
52             # Update residual graph
53             for u, v in path:
54                 residual[u][v] -= flow
55                 residual[v][u] += flow
56
57             max_flow_value += flow
58
59         return max_flow_value
60
61         # Example usage
62         # graph[u][v] = capacity
63         graph = defaultdict(lambda: defaultdict(
64             int))
65         graph[0][1] = 10
66         graph[0][2] = 10
67         graph[1][3] = 4
68         graph[1][4] = 8
69         graph[2][4] = 9
70         graph[3][5] = 10
71         graph[4][3] = 6
72         graph[4][5] = 10
73
74         n = 6 # Number of nodes
75         result = max_flow(graph, 0, 5, n)

```

## 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,
2     defaultdict
3
4 class Dinic:
5     def __init__(self, n):
6         self.n = n
7         self.graph = defaultdict(lambda:
8             defaultdict(int))
9
10    def add_edge(self, u, v, cap):
11        self.graph[u][v] += cap

```

```

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
22                 self.graph[u][v] > 0:
23                 level[v] = level[u]
24                 + 1
25                 queue.append(v)
26
27         return level if level[sink] !=
28             -1 else None
29
30 def dfs(self, u, sink, pushed, level
31         , start):
32     if u == sink:
33         return pushed
34
35     while start[u] < self.n:
36         v = start[u]
37
38         if (level[v] == level[u] + 1
39             and
40                 self.graph[u][v] > 0):
41             flow = self.dfs(v, sink,
42                             min(
43                                 pushed, self.graph[u][v]),
44                                 level,
45                                 start)
46
47             if flow > 0:
48                 self.graph[u][v] -=
49                     flow
50                 self.graph[v][u] +=
51                     flow
52                 return flow
53
54         start[u] += 1
55
56     return 0
57
58 def max_flow(self, source, sink):
59     flow = 0
60
61     while True:
62         level = self.bfs(source,
63                             sink)
64         if level is None:
65             break
66
67         start = [0] * self.n
68
69         while True:
70             pushed = self.dfs(source
71                                 , sink, float('inf'),
72                                 level,
73                                 start)
74             if pushed == 0:
75                 break
76             flow += pushed
77
78     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
3     # graph is available
4     # Min cut = set of reachable nodes
5     # from source
6     visited = [False] * n
7     queue = deque([source])
8     visited[source] = True
9
10    while queue:
11        u = queue.popleft()
12        for v in range(n):
13            if not visited[v] and
14            residual[u][v] > 0:
15                visited[v] = True
16                queue.append(v)
17
18    # Cut edges
19    cut_edges = []
20    for u in range(n):
21        if visited[u]:
22            for v in range(n):
23                if not visited[v] and
24                graph[u][v] > 0:
25                    cut_edges.append((u,
26                    v))
27    return cut_edges
```

### 18.4 Bipartite Matching

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

```
1 def max_bipartite_matching(left_size,
2                             right_size, edges):
3     # edges = [(left_node, right_node),
4     # ...]
5     # Add source (0) and sink (left_size
6     # + right_size + 1)
7
8     n = left_size + right_size + 2
9     source = 0
10    sink = n - 1
11
12    graph = defaultdict(lambda:
13    defaultdict(int))
14
15    # Source to left nodes
16    for i in range(1, left_size + 1):
17        graph[source][i] = 1
18
19    # Left to right edges
20    for l, r in edges:
21        graph[l + 1][left_size + r + 1]
22        = 1
23
24    # Right nodes to sink
25    for i in range(1, right_size + 1):
26        graph[left_size + i][sink] = 1
27
28    return max_flow(graph, source, sink,
29    n)
```