

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
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)] # 
        # Correct way!
4
5 # List comprehension - concise and

```

```

6 efficient
7 squares = [x**2 for x in range(n)]
8 evens = [x for x in arr if x % 2 == 0]
9
10 # Sorting - O(n log n)
11 arr.sort() # in-place, modifies arr
12 arr.sort(reverse=True) # descending
13 arr.sort(key=lambda x: (x[0], -x[1])) #
        # custom
14 sorted_arr = sorted(arr) # returns new
        # list
15
16 # Binary search in sorted array
17 from bisect import bisect_left,
        bisect_right
18 idx = bisect_left(arr, x) # leftmost
        position
19 idx = bisect_right(arr, x) # rightmost
        position
20
21 # Common operations
22 arr.append(x) # O(1) amortized
23 arr.pop() # O(1) - remove last
24 arr.pop(0) # O(n) - remove first
        (slow!)
25 arr.reverse() # O(n) - in-place
26 arr.count(x) # O(n) - count
        occurrences
27 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)
8 peek
9
10 # Max heap - negate values
11 heapq.heappush(heap, -x)
12 max_val = -heapq.heappop(heap)
13
14 # Convert list to heap in-place - O(n)
15 heapq.heapify(arr)
16
17 # K largest/smallest - O(n log k)
18 k_largest = heapq.nlargest(k, arr)
19 k_smallest = heapq.nsmallest(k, arr)
20
21 # Custom comparator using tuples
22 # Compares first element, then second,
23 # etc.
24 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,
2 Counter
3
4 # defaultdict - provides default value
5 graph = defaultdict(list) # empty list
6 default
7 count = defaultdict(int) # 0 default
8
9 # Counter - count elements efficiently
10 cnt = Counter(arr)
11 cnt['x'] += 1
12 most_common = cnt.most_common(k) # k
13 most frequent
14
15 # Dictionary operations
16 d = {}
17 d.get(key, default_val)
18 d.setdefault(key, default_val)
19 for k, v in d.items():
20     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
# String building - EFFICIENT O(n)
14 result = []
15 for x in data:
16     result.append(str(x))
17 s = ''.join(result)
18
# String concatenation - SLOW O(n^2)
19 # s = ""
20 # for x in data:
21 #     s += str(x) # Don't do this!
22
# ASCII values
23 ord('a') # 97
24 chr(97) # 'a'
25
26 # String to character array (for
27 # mutations)
28 chars = list(s)
29 chars[0] = 'x'
30 s = ''.join(chars)
31

```

```

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

```

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 - 1])
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]))
15         % MOD
16         if i < m - 1:
17             power = (power * BASE) % MOD
18
19     # Rolling hash
20     text_hash = 0
21     matches = []
22
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 def compute_inv_factorials(n, mod):
31     fact = modfact(n)
32     inv_fact = [1] * (n + 1)
33     inv_fact[n] = pow(fact[n], mod - 2,
34             mod)
35     for i in range(n - 1, -1, -1):
36         inv_fact[i] = inv_fact[i + 1] *
37             (i + 1) % mod
38     return fact, inv_fact
39
40 def modcomb(n, r, fact, inv_fact, mod):
41     if r > n or r < 0: return 0
42     return fact[n] * inv_fact[r] % mod *
43             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
8              composite
9              for j in range(i*i, n + 1, i):
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         else:
31             i += 1
32     return count
33
34 # Sum of divisors
35 def sum_divisors(n):
36     total = 0
            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      # Solve  $x = remainders[i] \pmod{moduli[i]}$ 
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] =
22                   i initially

```

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: $O(V)$.

```
def bfs(graph, start):
    visited = set([start])
    queue = deque([start])
    dist = {start: 0}

    while queue:
        node = queue.popleft()

        for neighbor in graph[node]:
            if neighbor not in visited:
                visited.add(neighbor)
                queue.append(neighbor)
                dist[neighbor] = dist[
```

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: $O(V)$.

```

# Recursive DFS
def dfs(graph, node, visited):
    visited.add(node)

    for neighbor in graph[node]:
        if neighbor not in visited:
            dfs(graph, neighbor, visited
)

# Iterative DFS using stack
def dfs_iterative(graph, start):
    visited = set()
    stack = [start]

    while stack:
        node = stack.pop()

        if node not in visited:
            visited.add(node)

            for neighbor in graph[node]:
                if neighbor not in
visited:
                    stack.append(
neighbor)

# Cycle detection in undirected graph
def has_cycle(graph, n):
    visited = [False] * n

    def dfs(node, parent):
        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 =
36             cycle
37             elif neighbor != parent:
38                 return True
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
72     return False
73
74 # Connected components count
75 def count_components(graph, n):
76     visited = [False] * n
77     count = 0
78
79     def dfs(node):
80         visited[node] = True
81         for neighbor in graph[node]:
82             if not visited[neighbor]:
83                 dfs(neighbor)
84
85         for i in range(n):
86             if not visited[i]:
87                 dfs(i)
88                 count += 1
89
90     return count
91
92 # Bipartite check (2-coloring)
93 def is_bipartite(graph, n):
94     color = [-1] * n
95
96     def bfs(start):
97         from collections import deque
98         queue = deque([start])
99         color[start] = 0
100
101         while queue:
102             node = queue.popleft()
103
104             for neighbor in graph[node]:
105                 if color[neighbor] ==
106                     -1:
107                         color[neighbor] = 1
108                         queue.append(
109                             neighbor)
110                         elif color[neighbor] ==
111                             color[node]:
112                             return False
113
114     for i in range(n):
115         if color[i] == -1:
116             if not bfs(i):
117                 return False
118
119     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
19 [v])
20
21                 # Bridge condition
22                 if low[v] > disc[u]:
23                     bridges.append((u, v
24 ))
25
26             elif v != parent[u]:
27                 low[u] = min(low[u],
28 disc[v])
29
30     for i in range(n):
31         if not visited[i]:
32             dfs(i)
33
34     return bridges
35
36     def find_articulation_points(graph, n):
37         visited = [False] * n
38         disc = [0] * n
39         low = [0] * n
40         parent = [-1] * n
41         time = [0]
42         ap = set()
43
44         def dfs(u):
45             children = 0
46             visited[u] = True
47             disc[u] = low[u] = time[0]
48             time[0] += 1
49
50             for v in graph[u]:
51                 if not visited[v]:
52                     children += 1
53                     parent[v] = u
54                     dfs(v)
55                     low[u] = min(low[u], low
56 [v])
57
58                     # Articulation point
59                     conditions
59                     if parent[u] == -1 and
60                     children > 1:
61                         ap.add(u)
62                     if parent[u] != -1 and
63                     low[v] >= disc[u]:
64                         ap.add(u)
65                     elif v != parent[u]:
66                         low[u] = min(low[u],
67 disc[v])
68
69     return ap

```

```

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
6         for _ in range(n)]
7         self.depth = [0] * n
8
9         # DFS to set parent and depth
10        visited = [False] * n
11
12        def dfs(node, par, d):
13            visited[node] = True
14            self.parent[node][0] = par
15            self.depth[node] = d
16
17            for neighbor in graph[node]:
18                if not visited[neighbor]:
19                    dfs(neighbor, node,
20 d + 1)
21
22        dfs(root, -1, 0)
23
24        # Binary lifting preprocessing
25        for j in range(1, self.LOG):
26            for i in range(n):
27                if self.parent[i][j-1]
28                != -1:
29                    self.parent[i][j] =
30                    self.parent[
31                        self.parent[i][j-1]
32                    ][j-1]
33
34        def lca(self, u, v):
35            # Make u deeper
36            if self.depth[u] < self.depth[v
37 ]:
38                u, v = v, u
39
40                # Bring u to same level as v
41                diff = self.depth[u] - self.
42                depth[v]
43                for i in range(self.LOG):
44                    if (diff >> i) & 1:
45                        u = self.parent[u][i]
46
47                if u == v:
48                    return u
49
50                    # Binary search for LCA
51                    for i in range(self.LOG - 1, -1,
52 -1):
53                        if self.parent[u][i] != self.
54 parent[v][i]:
55                            u = self.parent[u][i]
56                            v = self.parent[v][i]
57
58                    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]
55
56
57     def reconstruct_path(self, target):
58         path = []
59         while target != -1:
60             path.append(target)
61             target = self.parent[target]
62
63         return path[::-1]

```

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
25            if new_dist < dist[neighbor]:
26                dist[neighbor] =
27                new_dist
28                heapq.heappush(heap, (
29                    new_dist, neighbor))
30
31    return dist
32
33
34 # Path reconstruction
35 def dijkstra_with_path(graph, start, n):
36     dist = [float('inf')] * n
37     parent = [-1] * n
38     dist[start] = 0
39     heap = [(0, start)]
40
41     while heap:
42         d, node = heapq.heappop(heap)
43         if d > dist[node]:
44             continue
45
46         for neighbor, weight in graph[
47             node]:
48             new_dist = dist[node] +
49             weight
50             if new_dist < dist[neighbor]:
51                 dist[neighbor] =
52                 new_dist
53                 parent[neighbor] = node
54                 heapq.heappush(heap, (
55                     new_dist, neighbor))
56
57    return dist, parent

```

7.2 Bellman-Ford Algorithm

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

```

def bellman_ford(edges, n, start):
    # edges = [(u, v, weight), ...]
    dist = [float('inf')] * n
    dist[start] = 0

    # Relax edges n-1 times
    for _ in range(n - 1):
        for u, v, w in edges:
            if dist[u] != float('inf'):
                if dist[u] + w < dist[v]:
                    dist[v] = dist[u] + w
                    and \
                dist[u] + w < dist[v]:
                    dist[v] = dist[u] + w

    # Check for negative cycles
    for u, v, w in edges:
        if dist[u] != float('inf') and
            dist[u] + w < dist[v]:
                return None # Negative
    cycle exists

return dist

```

7.3 Floyd-Warshall Algorithm

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

```

def floyd_marshall(n, edges):
    # Initialize distance matrix
    dist = [[float('inf')]] * n for _ in range(n)

    for i in range(n):
        dist[i][i] = 0

    for u, v, w in edges:
        dist[u][v] = min(dist[u][v], w)

    # Dynamic programming
    for k in range(n): # Intermediate vertex
        for i in range(n):
            for j in range(n):
                dist[i][j] = min(dist[i][j],
                                   dist[i][k],
                                   dist[k][j],
                                   dist[i][k] + dist[k][j])

    return dist

# Check for negative cycle
def has_negative_cycle(dist, n):
    for i in range(n):
        if dist[i][i] < 0:
            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: $O(E \log E)$.

```
def kruskal(n, edges):
    # edges = [(weight, u, v), ...]
    edges.sort()    # Sort by weight

    uf = UnionFind(n)
    mst_weight = 0
    mst_edges = []

    for weight, u, v in edges:
        if uf.union(u, v):
            mst_weight += weight
            mst_edges.append((u, v, weight))

    return mst_weight, mst_edges

class UnionFind:
    def __init__(self, n):
        self.parent = list(range(n))
        self.rank = [0] * n

    def find(self, x):
        if self.parent[x] != x:
            self.parent[x] = self.find(
                self.parent[x])
        return self.parent[x]

    def union(self, x, y):
        px, py = self.find(x), self.find(y)
        if px == py:
            return False
        if self.rank[px] < self.rank[py]:
            px, py = py, px
            self.parent[py] = px
            if self.rank[px] == self.rank[py]:
                self.rank[px] += 1
            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]:
```

```

        continue

visited[u] = True
mst_weight += weight

for v, w in graph[u]:
    if not visited[v]:
        heapq.heappush(min_heap,
(w, v))

urn 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.

```

from collections import deque

def topo_sort(graph, n):
    # Count incoming edges for each node
    indegree = [0] * n
    for u in range(n):
        for v in graph[u]:
            indegree[v] += 1

    # Start with nodes having no
    # dependencies
    queue = deque([i for i in range(n)
                  if indegree[i] == 0])
    result = []

    while queue:
        node = queue.popleft()
        result.append(node)

        # Remove this node from graph
        for neighbor in graph[node]:
            indegree[neighbor] -= 1

            # If neighbor has no more
            # dependencies
            if indegree[neighbor] == 0:
                queue.append(neighbor)

    # If not all nodes processed, cycle
    # exists
    return result if len(result) == n
    else []

```

8.2 DFS-based Topological Sort

Advantages: Simpler code, uses less space.

```
def topo_dfs(graph, n):
    visited = [False] * n
    stack = []
```

```

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]
25
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(y)
37
38    # Count number of disjoint sets
39    def count_sets(self):
40        return len(set(self.find(i)
41                     for i in range(len(
42                     self.parent))))
43
44    # Example: Detect cycle in undirected
45    # graph
46    def has_cycle_uf(edges, n):
47        uf = UnionFind(n)
48        for u, v in edges:
49            if uf.connected(u, v):
50                return True # Cycle found
51            uf.union(u, v)
52        return False

```

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

```

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 # Round up!
23
24         if check(mid):
25             left = mid # Could be
26             answer, search right
27         else:
28             right = mid - 1 # Not
29             answer, search left

```

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.

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]
4     # and s2[:j]
5     dp = [[0] * (n + 1) for _ in range(m + 1)]
6
7     # Base cases: empty string
8     # transformations
9     for i in range(m + 1):
10         dp[i][0] = i # Delete all
11     for j in range(n + 1):
12         dp[0][j] = j # Insert all
13
14     for i in range(1, m + 1):
15         for j in range(1, n + 1):
16             if s1[i-1] == s2[j-1]:
17                 # Characters match, no
18                 # operation needed
19                 dp[i][j] = dp[i-1][j-1]
20             else:
21                 dp[i][j] = 1 + min(
22                     dp[i-1][j], # Delete from s1
23                     dp[i][j-1], # Insert into s1
24                     dp[i-1][j-1]) # Replace in s1
25
26     return dp[m][n]

```

11.4 Longest Common Subsequence (LCS)

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

```

# Reconstruct LCS
def lcs_string(s1, s2):
    m, n = len(s1), len(s2)
    dp = [[0] * (n + 1) for _ in range(m + 1)]

    for i in range(1, m + 1):
        for j in range(1, n + 1):
            if s1[i-1] == s2[j-1]:
                dp[i][j] = dp[i-1][j-1]
            else:
                dp[i][j] = max(dp[i-1][j],
                                dp[i][j-1])

    # Backtrack
    result = []
    i, j = m, n
    while i > 0 and j > 0:
        if s1[i-1] == s2[j-1]:
            result.append(s1[i-1])
            i -= 1
            j -= 1
        elif dp[i-1][j] > dp[i][j-1]:
            i -= 1
        else:
            j -= 1

    return ''.join(reversed(result))

```

11.5 Coin Change

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

```

# Minimum coins
def coin_change_min(coins, amount):
    dp = [float('inf')] * (amount + 1)
    dp[0] = 0

    for coin in coins:
        for i in range(coin, amount + 1):
            :
                dp[i] = min(dp[i], dp[i - coin] + 1)

    return dp[amount] if dp[amount] != float('inf') else -1

# Count ways
def coin_change_ways(coins, amount):
    dp = [0] * (amount + 1)
    dp[0] = 1

    for coin in coins:
        for i in range(coin, amount + 1):
            :
                dp[i] += dp[i - coin]

    return dp[amount]

```

11.6 Palindrome Partitioning

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

```
def min_palindrome_partition(s):  
    n = len(s)
```

```

4      # is_pal[i][j] = True if s[i:j+1] is
5      # palindrome
6      is_pal = [[False] * n for _ in range
7          (n)]
8
9      # Every single character is
10     # palindrome
11     for i in range(n):
12         is_pal[i][i] = True
13
14     # Check all substrings
15     for length in range(2, n + 1):
16         for i in range(n - length + 1):
17             j = i + length - 1
18             if s[i] == s[j]:
19                 is_pal[i][j] = (length
20                  == 2 or
21                  is_pal[i
22                      +1][j-1])
23
24             # dp[i] = min cuts for s[0:i+1]
25             dp = [float('inf')] * n
26
27             for i in range(n):
28                 if is_pal[0][i]:
29                     dp[i] = 0
30                 else:
31                     for j in range(i):
32                         if is_pal[j+1][i]:
33                             dp[i] = min(dp[i],
34                             dp[j] + 1)
35
36             return dp[n-1]

```

11.7 Subset Sum

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

```

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,
-1):
30                 dp[s] = dp[s] or dp[s - num]
31
32     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
10    queries
11    def build_2d_prefix(matrix):
12        n, m = len(matrix), len(matrix[0])
13        prefix = [[0] * (m + 1) for _ in
14                  range(n + 1)]
15
16        for i in range(1, n + 1):
17            for j in range(1, m + 1):
18                prefix[i][j] = (matrix[i-1][
19                    j-1] +
20                        prefix[i-1][j]
21                    ] +
22                        prefix[i][j-1] -
23                        prefix[i-1][j-1])
24
25        return prefix
26
27    # Rectangle sum from (x1, y1) to (x2, y2)
28    # 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,
10             window_sum)
11
12      return max_sum
13
14  # Variable size window - two pointers
15  def min_subarray_sum_geq_target(arr,
16      target):
17      left = 0
18      current_sum = 0
19      min_len = float('inf')
20
21      for right in range(len(arr)):
22          current_sum += arr[right]
23
24          # Shrink window while condition holds
25          while current_sum >= target:
26              min_len = min(min_len, right -
27                  left + 1)
28              current_sum -= arr[left]
29              left += 1
30
31      return min_len if min_len != float('
32          inf') else 0
33
34  # Longest substring with at most k
35  # distinct chars
36  def longest_k_distinct(s, k):
37      from collections import defaultdict
38
39      left = 0
40      char_count = defaultdict(int)
41      max_len = 0
42
43      for right in range(len(s)):
44          char_count[s[right]] += 1
45
46          # Shrink if too many distinct
47          while len(char_count) > k:
48              char_count[s[left]] -= 1
49              if char_count[s[left]] == 0:
50                  del char_count[s[left]]
51              left += 1
52
53              max_len = max(max_len, right -
54                  left + 1)
55
56      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
5          bound
6          self.tree = [0] * (4 * self.n)
7          self.build(arr, 0, 0, self.n - 1)
8
9      def build(self, arr, node, start,
10         end):
11          if start == end:
12              # Leaf node
13              self.tree[node] = arr[start]
14          else:
15              mid = (start + end) // 2
16              # Build left and right
17              subtrees
18              self.build(arr, 2*node+1,
19                  start, mid)
20              self.build(arr, 2*node+2,
21                  mid+1, end)
22              # Combine results (sum in
23              # this case)
24              self.tree[node] = (self.tree
25                  [2*node+1] +
26                                  self.tree
27                  [2*node+2])
28
29      def update(self, node, start, end,
30         idx, val):
31          if start == end:
32              # Leaf node - update value
33              self.tree[node] = val
34          else:
35              mid = (start + end) // 2
36              if idx <= mid:
37                  # Update left subtree
38                  self.update(2*node+1,
39                  start, mid, idx, val)
40              else:
41                  # Update right subtree
42                  self.update(2*node+2,
43                  mid+1, end, idx, val)
44                  # Recompute parent
45                  self.tree[node] = (self.tree
46                  [2*node+1] +
47                                  self.tree
48                  [2*node+2])
49
50      def query(self, node, start, end, l,
51         r):
52          # No overlap
53          if r < start or end < l:
54              return 0
55
56          # Complete overlap
57          if l <= start and end <= r:
58              return self.tree[node]
59
60          # Partial overlap
61          mid = (start + end) // 2
62          left_sum = self.query(2*node+1,
63                  start, mid, l, r)
64          right_sum = self.query(2*node+2,
65                  mid+1, end, l, r)
66          return left_sum + right_sum
67
68  # Public interface
69  def update_val(self, idx, val):
70      self.update(0, 0, self.n-1, idx,
71

```

```

55     val)
56
57     def range_sum(self, l, r):
58         return self.query(0, 0, self.n
59             -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
5          # implementation
6          self.tree = [0] * (n + 1)
7
8      def update(self, i, delta):
9          # Add delta to position i (1-
10         indexed)
11         while i <= self.n:
12             self.tree[i] += delta
13             # Move to next node: add LSB
14             i += i & (-i)
15
16     def query(self, i):
17         # Get prefix sum up to i (1-
18         indexed)
19         s = 0
20         while i > 0:
21             s += self.tree[i]
22             # Move to parent: remove LSB
23             i -= i & (-i)
24
25     def range_query(self, l, r):
26         # Sum from l to r (1-indexed)
27         return self.query(r) - self.
28         query(l - 1)
29
30     # Usage example
31     bit = FenwickTree(n)
32     for i, val in enumerate(arr, 1):
33         bit.update(i, val)
34
35     # Range sum [l, r] (1-indexed)
36     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 ->
4          TrieNode
5          self.is_end = False # End of
6          word marker
7
8  class Trie:
9      def __init__(self):
10         self.root = TrieNode()
11
12     def insert(self, word):
13         # 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] =
16             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                 return False
26             node = node.children[char]
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
36     # Find all words with given prefix
37     def words_with_prefix(self, prefix):
38         node = self.root
39         for char in prefix:
40             if char not in node.children:
41                 return []
42             node = node.children[char]
43
44         # DFS to collect all words
45         words = []
46         def dfs(n, path):
47             if n.is_end:
48                 words.append(prefix +
49                             path)
50                 for char, child in n.
51                 children.items():
52                     dfs(child, path + char)
53
54         dfs(node, "")
55
56     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",
5                  "right", "size")
6
7  def __init__(self, key):
8      self.key = key
9      self.prio = random.randint(1, 1
10                                << 30)
11      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
70            else t.right
71            return False
72
73        def insert(self, key):
74            if self.contains(key):
75                return
76            node = TreapNode(key)
77            l, r = _split(self.root, key)

```

```

78            self.root = _merge(_merge(l, 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
110        # None if none
111        t, ans = self.root, None
112        while t:
113            if t.key >= key:
114                ans = t.key
115                t = t.left
116            else:
117                t = t.right
118        return ans
119
120    # Usage example
121    T = Treap()
122    for x in [5, 1, 7, 3]:
123        T.insert(x)
124    T.contains(3)          # True
125    T.kth_smallest(1)     # 3 (0-indexed)
126    T.count_less_than(6)  # 3 (1,3,5)
127    T.remove(5)
128    len(T)                # 3

```

```

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')

```

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
28     (n & 0x55555555) != 0
29
30 # Iterate over all subsets of set
31 # represented by mask
32 mask = (1 << n) - 1 # All bits set
33 submask = mask
34 while submask > 0:
35     # Process submask
36     submask = (submask - 1) & mask
37
38 # Iterate through all k-bit masks
39 def iterate_k_bits(n, k):
40     mask = (1 << k) - 1
41     while mask < (1 << n):
42         # Process mask
43         yield mask
44         # Gosper's hack
45         c = mask & ~mask
46         r = mask + c
47         mask = (((r ^ mask) >> 2) // c)
48         | r
49
50 # XOR properties
51 # a ^ a = 0 (number XOR itself is 0)
52 # a ^ 0 = a (number XOR 0 is itself)
53 # XOR is commutative and associative
54 # Find unique element when all others
55 # appear twice:
56 def find_unique(arr):
57     result = 0
58     for x in arr:
59         result ^= x
60     return result
61
62 # Subset enumeration
63 n = 5 # Number of elements
64 for mask in range(1 << n):
65     subset = [i for i in range(n) if
66             mask & (1 << i)]
67     # Process subset
68
69 # Check parity (odd/even number of 1s)
70 def parity(n):
71     count = 0
72     while n:
73         count ^= 1
74         n &= n - 1
75     return count # 1 if odd, 0 if even
76
77 # Swap two numbers without temp variable
78 a, b = 5, 10
79 a ^= b
80 b ^= a
81 a ^= b
82 # 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]
10            ]][j]
11
12    return C
13
14 # With modulo
15 def matmul_mod(A, B, mod):
16     n = len(A)
17     C = [[0] * n for _ in range(n)]
18
19     for i in range(n):
20         for j in range(n):
21             for k in range(n):
22                 C[i][j] = (C[i][j] +
23                             A[i][k] * B[k]
24                         ) % mod
25
26    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
20 # exponentiation
21 # F(n) = [[1,1],[1,0]]^n
22 def fibonacci(n, mod):
23     if n == 0: return 0
24     if n == 1: return 1
25
26     M = [[1, 1], [1, 0]]
27     result = matpow(M, n - 1, mod)
28     return result[0][0]
29
30 # Linear recurrence: a(n) = c1*a(n-1) +
31 # c2*a(n-2) + ...
32 # Build transition matrix and use matrix
33 # exponentiation
34 def linear_recurrence(coeffs, init, n,
35

```

```

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-
46     # -2), ..., a(0)]
47     state = init[k-1::-1]
48
49     #  $M^{n-k+1}$ 
50     result_matrix = matpow(M, n - k + 1,
51         mod)
52
53     # Multiply with initial state
54     result = 0
55     for i in range(k):
56         result = (result + result_matrix
57             [0][i] * state[i]) % mod
58
59     return result
60
61 # Example: Tribonacci T(n) = T(n-1) + T(
62 # n-2) + T(n-3)
63 def tribonacci(n, mod):
64     if n == 0: return 0
65     if n == 1 or n == 2: return 1
66
67     coeffs = [1, 1, 1]
68     init = [0, 1, 1]
69     return linear_recurrence(coeffs,
70         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-
4 length permutations
5 perms = list(permutations([1,2,3], 2))
6 # [(1,2), (1,3), (2,1), (2,3), (3,1),
6     (3,2)]
7
8 # combinations(iterable, r) - r-length
8 combinations
9 combs = list(combinations([1,2,3], 2))
9 # [(1,2), (1,3), (2,3)]
10
11 # product - cartesian product
12 prod = list(product([1,2], ['a','b']))
13 # [(1,'a'), (1,'b'), (2,'a'), (2,'b')]
14
15 # accumulate - running totals
16 acc = list(accumulate([1,2,3,4]))
16 # [1, 3, 6, 10]
17
18 # chain - flatten iterables
19 chained = list(chain([1,2], [3,4]))
19 # [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
6 positive
7 any(x > 0 for x in arr) # True if any
7 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
21     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
34 = [2,3,4], b=5

```

16.4 Common Pitfalls

```

1 # Integer division - floors toward
1 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)]) # Creates full list
62
63  # Ternary operator
64  x = a if condition else b
65
66  # Dictionary get with default
67  count = d.get(key, 0) + 1
68
69  # Matrix rotation 90 degrees clockwise
70  def rotate_90(matrix):
71      return [list(row) for row in zip(*
72          matrix[::-1])]

# Matrix transpose
70  def transpose(matrix):
71      return [list(row) for row in zip(*
72          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 ($\sim 1\text{s}$)
- $O(n \log n)$: 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 (heappq), 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 (as-sume 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 # Point operations
3

```

```

def dist(p1, p2):
    # Euclidean distance
    return math.sqrt((p1[0] - p2[0])**2
                    + (p1[1] - p2[1])**2)

def cross_product(O, A, B):
    # Cross product of vectors OA and OB
    # Positive: counter-clockwise
    # 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: $O(n \log n)$.

```

def convex_hull(points):
    # Graham's scan algorithm
    points = sorted(points) # Sort by x
    , then y

    if len(points) <= 2:
        return points

    # Build lower hull
    lower = []
    for p in points:
        while (len(lower) >= 2 and
               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
39

```

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 \
19                            <= xinters:
20                                inside = not
21
22                                p1x, p1y = p2x, p2y
23
24    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

```

```

4             =lambda p: p[1])
5
6     def closest_recursive(px, py):
7         n = len(px)
8
9         # Base case: brute force
10        if n <= 3:
11            min_dist = float('inf')
12            for i in range(n):
13                for j in range(i + 1, n):
14                    min_dist = min(
15                        min_dist, dist(px[i], px[j]))
16
17        return min_dist
18
19        # Divide
20        mid = n // 2
21        midpoint = px[mid]
22
23        pyl = [p for p in py if p[0] <=
24               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:], pyr)
31        d = min(dl, dr)
32
33        # Combine: check strip
34        strip = [p for p in py if abs(p
35                           [0] - midpoint[0]) < d]
36
37        for i in range(len(strip)):
38            j = i + 1
39            while j < len(strip) and
40                  strip[j][1] - strip[i][1] < d:
41                      d = min(d, dist(strip[i]
42                                      [1], strip[j]))
43                      j += 1
44
45        return d
46
47    return closest_recursive(
48        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(VE²).

```

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         for u in graph:
10            for v in graph[u]:
11                residual[u][v] = graph[u][v]
12
13    def bfs_path():
14        # Find augmenting path using BFS
15        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
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)

```

```

3 class Dinic:
4     def __init__(self, n):
5         self.n = n
6         self.graph = defaultdict(lambda:
7             defaultdict(int))
8
9     def add_edge(self, u, v, cap):
10        self.graph[u][v] += cap
11
12    def bfs(self, source, sink):
13        # Build level graph
14        level = [-1] * self.n
15        level[source] = 0
16        queue = deque([source])
17
18        while queue:
19            u = queue.popleft()
20
21            for v in range(self.n):
22                if level[v] == -1 and
23                    self.graph[u][v] > 0:
24                    level[v] = level[u]
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
42                flow = self.dfs(v, sink,
43                                 min(
44                                 pushed, self.graph[u][v]),
45                                 level,
46                                 start)
47
48                if flow > 0:
49                    self.graph[u][v] -=
50                    flow
51                    self.graph[v][u] +=
52                    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
```

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

```

```

64           break
65           flow += pushed
66
67       return flow
23   return cut_edges

```

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))

```

18.4 Bipartite Matching

Description: Maximum matching in bipartite graph using flow.

```

def max_bipartite_matching(left_size,
                           right_size, edges):
    # edges = [(left_node, right_node),
    # ...]
    # Add source (0) and sink (left_size
    # + right_size + 1)

    n = left_size + right_size + 2
    source = 0
    sink = n - 1

    graph = defaultdict(lambda:
                       defaultdict(int))

    # Source to left nodes
    for i in range(1, left_size + 1):
        graph[source][i] = 1

    # Left to right edges
    for l, r in edges:
        graph[l + 1][left_size + r + 1]
        = 1

    # Right nodes to sink
    for i in range(1, right_size + 1):
        graph[left_size + i][sink] = 1

    return max_flow(graph, source, sink,
                  n)

```