

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
# efficient
6 squares = [x**2 for x in range(n)]
7 evens = [x for x in arr if x % 2 == 0]
8
9 # Sorting - O(n log n)

```

```

10 arr.sort() # in-place, modifies arr
11 arr.sort(reverse=True) # descending
12 arr.sort(key=lambda x: (x[0], -x[1])) #
        # custom
13 sorted_arr = sorted(arr) # returns new
        # list
14
15 # Binary search in sorted array
16 from bisect import bisect_left,
        bisect_right
17 idx = bisect_left(arr, x) # leftmost
        # position
18 idx = bisect_right(arr, x) # rightmost
        # position
19
20 # Common operations
21 arr.append(x) # O(1) amortized
22 arr.pop() # O(1) - remove last
23 arr.pop(0) # O(n) - remove first
        # (slow!)
24 arr.reverse() # O(n) - in-place
25 arr.count(x) # O(n) - count
        # occurrences
26 arr.index(x) # O(n) - first
        # occurrence

```

2.2 Deque (Double-ended Queue)

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

```

1 from collections import deque
2 dq = deque()
3
4 # O(1) operations on both ends
5 dq.append(x) # add to right
6 dq.appendleft(x) # add to left
7 dq.pop() # remove from right
8 dq.popleft() # remove from left
9
10 # Sliding window maximum - O(n)
11 # Maintains decreasing order of elements
12 def sliding_max(arr, k):
13     dq = deque() # stores indices
14     result = []
15
16     for i in range(len(arr)):
17         # Remove indices outside window

```

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
7 min_val = heapq.heappop(heap)    # O(log n)
8
9 min_val = heap[0]                # O(1)
10 peek
11
12 # Max heap - negate values
13 heapq.heappush(heap, -x)
14 max_val = -heapq.heappop(heap)
15
16 # Convert list to heap in-place - O(n)
17 heapq.heapify(arr)
18
19 # K largest/smallest - O(n log k)
20 k_largest = heapq.nlargest(k, arr)
21 k_smallest = heapq.nsmallest(k, arr)
22
23 # Custom comparator using tuples
24 # Compares first element, then second,
25 # etc.
26 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
```

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()           # O(1)
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([()])
17
```

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 $O(n^2)$.

```
# Common string methods
s.lower(), s.upper()
s.strip()      # remove whitespace both
               ends
s.lstrip()     # remove left whitespace
s.rstrip()    # remove right whitespace
s.split(delimiter)
delimiter.join(list)
s.replace(old, new)
s.startswith(prefix)
s.endswith(suffix)
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
19 # String concatenation - SLOW O(n^2)
20 # s = ""
21 # for x in data:
22 #     s += str(x) # Don't do this!
23
24 # ASCII values
25 ord('a') # 97
26 chr(97) # 'a'
27
28 # String to character array (for
# mutations)
29 chars = list(s)
30 chars[0] = 'x'
31 s = ''.join(chars)

```

3.1 KMP Pattern Matching

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

```

1 def kmp_search(text, pattern):
2     # Build LPS (Longest Proper Prefix
# which is Suffix)
3     def build_lps(pattern):
4         m = len(pattern)
5         lps = [0] * m
6         length = 0 # Length of previous
# longest prefix
7         i = 1
8
9         while i < m:
10             if pattern[i] == pattern[length]:
11                 length += 1
12                 lps[i] = length
13                 i += 1
14             else:
15                 if length != 0:
16                     length = lps[length - 1]
17                 else:
18                     lps[i] = 0
19                     i += 1
20
21         return lps
22
23 n, m = len(text), len(pattern)
24 lps = build_lps(pattern)
25
26 matches = []
27 i = j = 0 # Indices for text and
# pattern
28
29 while i < n:
30     if text[i] == pattern[j]:
31         i += 1
32         j += 1
33
34     if j == m:
35         matches.append(i - j)
36         j = lps[j - 1]
37     elif i < n and text[i] != pattern[j]:
38         if j != 0:
39             j = lps[j - 1]
40         else:
41             i += 1
42

```

```

43     return matches

```

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 +
# ord(pattern[i])) % MOD
14        if i < m - 1:
15            power = (power * BASE) % MOD
16
17    # Rolling hash
18    text_hash = 0
19    matches = []
20
21    for i in range(n):
22

```

```

23     # Add new character
24     text_hash = (text_hash * BASE +
25                 ord(text[i])) % MOD
26
27     # Remove old character if window
28     full
29     if i >= m:
30         text_hash = (text_hash -
31                     ord(text[i - m])) *
32             power) % MOD
33         text_hash = (text_hash + MOD
34 ) % MOD
35
36     # Check match
37     if i >= m - 1 and text_hash ==
38 pattern_hash:
39         # Verify actual match (avoid
40         hash collision)
41         if text[i - m + 1:i + 1] ==
42 pattern:
43             matches.append(i - m +
44 1)
45
46 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) = (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
13 def extgcd(a, b):
14     if b == 0:
15         return a, 1, 0
16     g, x1, y1 = extgcd(b, a % b)
17     x = y1
18     y = x1 - (a // b) * y1
19     return g, x, y

```

5.2 Sieve of Eratosthenes

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

```

1 def sieve(n):
2     is_prime = [True] * (n + 1)
3     is_prime[0] = is_prime[1] = False

```

```

4
5     for i in range(2, int(n**0.5) + 1):
6         if is_prime[i]:
7             # Mark multiples as
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 2
31             i += 1
32     return count
33
34 # Sum of divisors
35 def sum_divisors(n):
36     total = 0
37     i = 1
38     while i * i <= n:
39         if n % i == 0:
40             total += i
41             if i != n // i:
42                 total += n // i
43             i += 1
44     return total

```

```

1 def chinese_remainder(remainders, moduli):
2     """
3         Solve  $x \equiv remainders[i] \pmod{moduli[i]}$ 
4         Assumes moduli are pairwise coprime
5     """
6
7     def extgcd(a, b):
8         if b == 0:
9             return a, 1, 0
10            g, x1, y1 = extgcd(b, a % b)
11            return g, y1, x1 - (a // b) * y1
12
13    total = 0
14    prod = 1
15    for m in moduli:
16        prod *= m
17
18    for r, m in zip(remainders, moduli):
19        p = prod // m
20        g, inv, _ = extgcd(p, m)
21        # inv may be negative, normalize it
22        inv = (inv % m + m) % m
23        total += r * inv * p
24
25    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
24     for i in range(2, n + 1):
25         if phi[i] == i: # i is prime
26             for j in range(i, n + 1, i):
27                 phi[j] = phi[j] // i * (
28                     i - 1)
29
30     return phi

```

5.4 Chinese Remainder Theo- rem

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.

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

```

6 Graph Algorithms

6.1 Graph Representation

Description: Adjacency list is most common for sparse graphs. Use default dict 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
16    return dist
17
18 # Grid BFS - common in maze/path
19 # problems
20 def grid_bfs(grid, start):
21     n, m = len(grid), len(grid[0])
22     visited = [[False] * m for _ in
23               range(n)]
24     queue = deque([start])
25     visited[start[0]][start[1]] = True
26
27     # 4 directions: right, down, left,
28     # up
29     dirs = [(0,1), (1,0), (0,-1), (-1,0)]
30
31     while queue:
32         node = queue.popleft()
33
34         for dir in dirs:
35             x, y = node[0] + dir[0], node[1] + dir[1]
36
37             if 0 <= x < n and 0 <= y < m and
38               not visited[x][y]:
39                 visited[x][y] = True
40                 queue.append((x,y))
41
42    return visited

```

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
32
33         for neighbor in graph[node]:
34             if not visited[neighbor]:
35                 if dfs(neighbor, node):
36                     return True
37
38             # Back edge to non-parent =
39             cycle
40                 elif neighbor != parent:
41                     return True
42
43             return False
44
45 # Check all components
46 for i in range(n):
47     if not visited[i]:
48         if dfs(i, -1):
49             return True
50
51 return False
```

6.4 Strongly Connected Components (SCC)

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

```

f tarjan_scc(graph, n):
    index_counter = [0]
    stack = []
    lowlink = [0] * n
    index = [0] * n
    on_stack = [False] * n
    index_initialized = [False] * n
    sccs = []

    def strongconnect(v):
        index[v] = index_counter[0]
        lowlink[v] = index_counter[0]
        index_counter[0] += 1
        index_initialized[v] = True
        stack.append(v)
        on_stack[v] = True

        for w in graph[v]:
            if not index_initialized[w]:
                strongconnect(w)
                lowlink[v] = min(lowlink[v],
                                   lowlink[w])
            elif on_stack[w]:
                lowlink[v] = min(lowlink[v],
                                  index[w])

        if lowlink[v] == index[v]:
            scc = []
            while True:
                w = stack.pop()
                on_stack[w] = False
                scc.append(w)
                if w == v:
                    break
            sccs.append(scc)

    for v in range(n):
        if not index_initialized[v]:
            strongconnect(v)

    return sccs

```

6.5 Bridges and Articulation Points

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

```
f find_bridges(graph, n):
    visited = [False] * n
    disc = [0] * n
    low = [0] * n
    parent = [-1] * n
    time = [0]
    bridges = []

    def dfs(u):
        visited[u] = True
        disc[u] = low[u] = time[0]
        time[0] += 1

        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         # 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],
26             disc[v])
27     for i in range(n):
28         if not visited[i]:
29             dfs(i)
30     return bridges
31 def find_articulation_points(graph, n):
32     visited = [False] * n
33     disc = [0] * n
34     low = [0] * n
35     parent = [-1] * n
36     time = [0]
37     ap = set()
38
39     def dfs(u):
40         children = 0
41         visited[u] = True
42         disc[u] = low[u] = time[0]
43         time[0] += 1
44
45         for v in graph[u]:
46             if not visited[v]:
47                 children += 1
48                 parent[v] = u
49                 dfs(v)
50                 low[u] = min(low[u], low
51                     [v])
52
53             # Articulation point
54             conditions
55             if parent[u] == -1 and
56             children > 1:
57                 ap.add(u)
58             if parent[u] != -1 and
59             low[v] >= disc[u]:
60                 ap.add(u)
61             elif v != parent[u]:
62                 low[u] = min(low[u],
63                 disc[v])
64
65     for i in range(n):
66         if not visited[i]:
67             dfs(i)
68     return list(ap)
69
70
71     self.depth = [0] * n
72
73     # DFS to set parent and depth
74     visited = [False] * n
75
76     def dfs(node, par, d):
77         visited[node] = True
78         self.parent[node][0] = par
79         self.depth[node] = d
80
81         for neighbor in graph[node]:
82             if not visited[neighbor
83 ]:
84                 dfs(neighbor, node,
85 d + 1)
86
87     dfs(root, -1, 0)
88
89     # Binary lifting preprocessing
90     for j in range(1, self.LOG):
91         for i in range(n):
92             if self.parent[i][j-1]
93             != -1:
94                 self.parent[i][j] =
95                     self.parent[
96                         self.parent[i][j-1]
97 ]
98
99     def lca(self, u, v):
100        # Make u deeper
101        if self.depth[u] < self.depth[v
102 ]:
103            u, v = v, u
104
105            # Bring u to same level as v
106            diff = self.depth[u] - self.
107            depth[v]
108            for i in range(self.LOG):
109                if (diff >> i) & 1:
110                    u = self.parent[u][i]
111
112            if u == v:
113                return u
114
115            # Binary search for LCA
116            for i in range(self.LOG - 1, -1,
117 -1):
118                if self.parent[u][i] != self.
119                parent[v][i]:
120                    u = self.parent[u][i]
121                    v = self.parent[v][i]
122
123        return self.parent[u][0]
124
125    def dist(self, u, v):
126        # Distance between two nodes
127        l = self.lca(u, v)
128        return self.depth[u] + self.
129        depth[v] - 2 * self.depth[l]
130
131

```

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)]

```

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 # 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                 dist[neighbor] =
51                 new_dist
52                 parent[neighbor] = node
53                 heapq.heappush(heap, (
54                     new_dist, neighbor))
55
56     return dist, parent
57
58 def reconstruct_path(parent, target):
59     path = []
60     while target != -1:
61         path.append(target)
62         target = parent[target]
63     return path[::-1]

```

7.2 Bellman-Ford Algorithm

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

```

1 def bellman_ford(edges, n, start):
2     # edges = [(u, v, weight), ...]
3     dist = [float('inf')] * n
4     dist[start] = 0
5
6     # Relax edges n-1 times

```

```

7     for _ in range(n - 1):
8         for u, v, w in edges:
9             if dist[u] != float('inf'):
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
20
21     return dist

```

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], dist[i][k] + dist[k][j])
19
20     return dist
21
22 def has_negative_cycle(dist, n):
23     for i in range(n):
24         if dist[i][i] < 0:
25             return True
26     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     return mst_weight, mst_edges
15
16 class UnionFind:
17     def __init__(self, n):
18         self.parent = list(range(n))
19         self.rank = [0] * n
20
21     def find(self, x):
22         if self.parent[x] != x:
23             self.parent[x] = self.find(
24                 self.parent[x])
25         return self.parent[x]
26
27     def union(self, x, y):
28         px, py = self.find(x), self.find(y)
29         if px == py:
30             return False
31         if self.rank[px] < self.rank[py]:
32             px, py = py, px
33             if self.rank[px] == self.rank[py]:
34                 self.rank[px] += 1
35             return True

```

7.4.2 Prim's Algorithm

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

```

1 import heapq
2
3 def prim(graph, n):
4     # graph[u] = [(v, weight), ...]
5     visited = [False] * n
6     min_heap = [(0, 0)] # (weight, node)
7     mst_weight = 0
8
9     while min_heap:
10        weight, u = heapq.heappop(
11            min_heap)
12
13        if visited[u]:
14            continue
15
16        visited[u] = True
17        mst_weight += weight
18
19        for v, w in graph[u]:
20            if not visited[v]:
21                heapq.heappush(min_heap,
22                               (w, v))
23
24    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-
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
31
32             self.parent[py] = px
33
34             # Increase rank if trees had
35             equal rank
36             if self.rank[px] == self.rank[py]:
37                 self.rank[px] += 1
38
39             return True
40
41     def connected(self, x, y):
42         return self.find(x) == self.find(
43             y)
44
45     # Count number of disjoint sets
46     def count_sets(self):
47         return len(set(self.find(i)
48                     for i in range(len(
49                         self.parent))))))
50
51     # Example: Detect cycle in undirected
52     # graph
53

```

```
def has_cycle_uf(edges, n):
    uf = UnionFind(n)
    for u, v in edges:
        if uf.connected(u, v):
            return True # Cycle found
        uf.union(u, v)
    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

```

# Find FIRST position where check(mid)
# is True
def binary_search_first(left, right,
check):
    while left < right:
        mid = (left + right) // 2

        if check(mid):
            right = mid # Could be
            answer, search left
        else:
            left = mid + 1 # Not answer
            , search right

    return left

# Find LAST position where check(mid) is
# True
def binary_search_last(left, right,
check):
    while left < right:
        mid = (left + right + 1) // 2 # Round up!

        if check(mid):
            left = mid # Could be
            answer, search right
        else:
            right = mid - 1 # Not
            answer, search left

    return left

# Example: Integer square root
def sqrt_binary(n):
    left, right = 0, n

    while left < right:
        mid = (left + right + 1) // 2

        if mid * mid <= n:
            left = mid # mid might be
            answer
        else:
            right = mid - 1

    return left

# Binary search on answer - common
# pattern
def min_days_to_make_bouquets(bloomDay,
m, k):

```

```

41     # Can we make m bouquets in 'days'
42     days?
43     def can_make(days):
44         bouquets = consecutive = 0
45         for bloom in bloomDay:
46             if bloom <= days:
47                 consecutive += 1
48                 if consecutive == k:
49                     bouquets += 1
50                     consecutive = 0
51             else:
52                 consecutive = 0
53         return bouquets >= m
54
55     if len(bloomDay) < m * k:
56         return -1
57
58     # Binary search on number of days
59     return binary_search_first(
60         min(bloomDay), max(bloomDay),
61         can_make)

```

```

36         if idx > 0:
37             parent[i] = dp_idx[idx - 1]
38
39         # Reconstruct sequence
40         result = []
41         idx = dp_idx[-1]
42         while idx != -1:
43             result.append(arr[idx])
44             idx = parent[idx]
45
46     return result[::-1]

```

11 Dynamic Programming

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

11.1 Longest Increasing Subsequence

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

```

1  def lis(arr):
2      from bisect import bisect_left
3
4      # dp[i] = smallest ending value of
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
14         else:
15             dp[idx] = x # Better
16             ending for this length
17
18     return len(dp)
19
20 # LIS with actual sequence
21 def lis_with_sequence(arr):
22     from bisect import bisect_left
23
24     n = len(arr)
25     dp = []
26     parent = [-1] * n
27     dp_idx = [] # indices in dp
28
29     for i, x in enumerate(arr):
30         idx = bisect_left(dp, x)
31
32         if idx == len(dp):
33             dp.append(x)
34             dp_idx.append(i)
35         else:
36             dp[idx] = x
37             dp_idx[idx] = i

```

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,
5      #           weight <= w
6      dp = [[0] * (capacity + 1) for _ in
7            range(n + 1)]
8
9      for i in range(1, n + 1):
10         for w in range(capacity + 1):
11             # Don't take item i-1
12             dp[i][w] = dp[i-1][w]
13
14             # Take item i-1 if it fits
15             if weights[i-1] <= w:
16                 dp[i][w] = max(
17                     dp[i][w],
18                     dp[i-1][w - weights[
19                         i-1]] + values[i-1]
20                 )
21
22     return dp[n][capacity]
23
24 # Space-optimized O(capacity)
25 def knapsack_optimized(weights, values,
26                         capacity):
27     dp = [0] * (capacity + 1)
28
29     for i in range(len(weights)):
30         # Iterate backwards to avoid
31         # using updated values
32         for w in range(capacity, weights[
33             i] - 1, -1):
34             dp[w] = max(dp[w],
35                         dp[w - weights[i]]
36                         + values[i])
37
38     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]
4      #           and s2[:j]
5      dp = [[0] * (n + 1) for _ in range(m +
6          1)]

```

```

5      # Base cases: empty string
6      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
16                 # operation needed
17                 dp[i][j] = dp[i-1][j-1]
18             else:
19                 dp[i][j] = 1 + min(
20                     dp[i-1][j],           #
21                     Delete from s1
22                     dp[i][j-1],           #
23                     Insert into s1
24                     dp[i-1][j-1]           #
25                     Replace in s1
26                     )
27
28     return dp[m][n]

```

11.4 Longest Common Subsequence (LCS)

Description: Longest subsequence common to two sequences. Time: O(m×n).

```

1 def lcs(s1, s2):
2     m, n = len(s1), len(s2)
3     dp = [[0] * (n + 1) for _ in range(m
4         + 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]
9             else:
10                dp[i][j] = max(dp[i-1][j]
11                               , dp[i][j-1])
12
13     return dp[m][n]
14
15 # Reconstruct LCS
16 def lcs_string(s1, s2):
17     m, n = len(s1), len(s2)
18     dp = [[0] * (n + 1) for _ in range(m
19         + 1)]
20
21     for i in range(1, m + 1):
22         for j in range(1, n + 1):
23             if s1[i-1] == s2[j-1]:
24                 dp[i][j] = dp[i-1][j-1]
25             else:
26                 dp[i][j] = max(dp[i-1][j]
27                               , dp[i][j-1])
28
29     # Backtrack
30     result = []
31     i, j = m, n
32     while i > 0 and j > 0:
33         if s1[i-1] == s2[j-1]:
34             result.append(s1[i-1])
35             i -= 1
36             j -= 1
37         elif dp[i-1][j] > dp[i][j-1]:
38             i -= 1
39

```

```

        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×amount).

```

1 # Minimum coins
2 def coin_change_min(coins, amount):
3     dp = [float('inf')] * (amount + 1)
4     dp[0] = 0
5
6     for coin in coins:
7         for i in range(coin, amount + 1):
8             dp[i] = min(dp[i], dp[i - coin] + 1)
9
10    return dp[amount] if dp[amount] !=
11        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             dp[i] += dp[i - coin]
21
22     return dp[amount]

```

11.6 Palindrome Partitioning

Description: Minimum cuts to partition string into palindromes. Time: O(n²).

```

1 def min_palindrome_partition(s):
2     n = len(s)
3
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 + 1][j - 1])
22
23    # dp[i] = min cuts for s[0:i+1]
24    dp = [float('inf')] * n
25
26    for i in range(n):
27        if is_pal[0][i]:
28            dp[i] = 0
29
30    for i in range(1, n):
31        for j in range(i + 1, n + 1):
32            if is_pal[i][j]:
33                dp[i] = min(dp[i], dp[i - 1] + 1)
34
35    return dp[n - 1]

```

```

25     else:
26         for j in range(i):
27             if is_pal[j+1][i]:
28                 dp[i] = min(dp[i],
29                             dp[j] + 1)
30

```

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 _ in range(n + 1)]
4
5     # Base case: sum 0 is always achievable
6     for i in range(n + 1):
7         dp[i][0] = True
8
9     for i in range(1, n + 1):
10        for s in range(target + 1):
11            # Don't take arr[i-1]
12            dp[i][s] = dp[i-1][s]
13
14            # Take arr[i-1] if possible
15            if s >= arr[i-1]:
16                dp[i][s] = dp[i][s] or
17                dp[i-1][s - arr[i-1]]
18
19    return dp[n][target]
20
21 # Space optimized
22 def subset_sum_optimized(arr, target):
23     dp = [False] * (target + 1)
24     dp[0] = True
25
26     for num in arr:
27         for s in range(target, num - 1, -1):
28             dp[s] = dp[s] or dp[s - num]
29
30

```

```

17     j-1) +
18     ] + prefix[i-1][j]
19     -1) -
20     prefix[i-1][j]
21
22     return prefix
23
24 # Rectangle sum from (x1,y1) to (x2,y2)
25 # inclusive
26 def rect_sum(prefix, x1, y1, x2, y2):
27     return (prefix[x2+1][y2+1] -
28             prefix[x1][y2+1] -
29             prefix[x2+1][y1] +
30             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 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][

```

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         while current_sum >= target:
25             min_len = min(min_len, right - left + 1)
26             current_sum -= arr[left]
27             left += 1
28
29     return min_len

```

```

18     for right in range(len(arr)):
19         current_sum += arr[right]
20
21         # Shrink window while condition
22         holds
23             while current_sum >= target:
24                 min_len = min(min_len, right
25 - left + 1)
26                 current_sum -= arr[left]
27                 left += 1
28
29             return min_len if min_len != float('
30             inf') else 0
31
32 # Longest substring with at most k
33 # distinct chars
34 def longest_k_distinct(s, k):
35     from collections import defaultdict
36
37     left = 0
38     char_count = defaultdict(int)
39     max_len = 0
40
41     for right in range(len(s)):
42         char_count[s[right]] += 1
43
44         # Shrink if too many distinct
45         while len(char_count) > k:
46             char_count[s[left]] -= 1
47             if char_count[s[left]] == 0:
48                 del char_count[s[left]]
49             left += 1
50
51             max_len = max(max_len, right -
52             left + 1)
53
54     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 -
8             1)
9
10    def build(self, arr, node, start,
11             end):
12        if start == end:
13            # Leaf node
14            self.tree[node] = arr[start]
15        else:
16            mid = (start + end) // 2
17            # Build left and right
18            subtrees
19            self.build(arr, 2*node+1,
20                      start, mid)
21            self.build(arr, 2*node+2,
22                      mid+1, end)
23            # Combine results (sum in
24            this case)
25            self.tree[node] = (self.tree
26 [2*node+1] +
27                         self.tree
28 [2*node+2])
29
30    def update(self, node, start, end,
31              idx, val):
32        if start == end:
33            # Leaf node - update value
34            self.tree[node] = val
35        else:
36            mid = (start + end) // 2
37            if idx <= mid:
38                # Update left subtree
39                self.update(2*node+1,
40                           start, mid, idx, val)
41            else:
42                # Update right subtree
43                self.update(2*node+2,
44                           mid+1, end, idx, val)
45            # Recompute parent
46            self.tree[node] = (self.tree
47 [2*node+1] +
48                         self.tree
49 [2*node+2])
50
51    def query(self, node, start, end, l,
52              r):
53        # No overlap
54        if r < start or end < l:
55            return 0
56
57        # Complete overlap
58        if l <= start and end <= r:
59            return self.tree[node]
60
61        # Partial overlap
62        mid = (start + end) // 2
63        left_sum = self.query(2*node+1,
64                           start, mid, l, r)
65        right_sum = self.query(2*node+2,
66                           mid+1, end, l, r)
67        return left_sum + right_sum
68
69    # Public interface
70    def update_val(self, idx, val):
71        self.update(0, 0, self.n-1, idx,
72                   val)
73
74    def range_sum(self, l, r):
75        return self.query(0, 0, self.n
76 -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             i += i & -i

```

```

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-
16         # indexed)
17         s = 0
18         while i > 0:
19             s += self.tree[i]
20             # Move to parent: remove LSB
21             i -= i & (-i)
22         return s
23
24     def range_query(self, l, r):
25         # Sum from l to r (1-indexed)
26         return self.query(r) - self.
27         query(l - 1)
28
29     # Usage example
30     bit = FenwickTree(n)
31     for i, val in enumerate(arr, 1):
32         bit.update(i, val)
33
34     # Range sum [l, r] (1-indexed)
35     result = bit.range_query(l, r)
36
37     # Find all words with given prefix
38     def words_with_prefix(self, prefix):
39         node = self.root
40         for char in prefix:
41             if char not in node.children:
42                 return []
43             node = node.children[char]
44
45         # DFS to collect all words
46         words = []
47         def dfs(n, path):
48             if n.is_end:
49                 words.append(prefix +
50                     path)
51             for char, child in n.
52                 children.items():
53                     dfs(child, path + char)
54
55         dfs(node, "")
56
57     return words

```

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))
14        node = self.root
15        for char in word:
16            if char not in node.children:
17                node.children[char] =
18                    TrieNode()
19                node = node.children[char]
20                node.is_end = True
21
22    def search(self, word):
23        # Exact word search - O(len(word))
24        node = self.root
25        for char in word:
26            if char not in node.children:
27                return False
28            node = node.children[char]
29        return node.is_end
30
31    def starts_with(self, prefix):
32        # Prefix search - O(len(prefix))
33        node = self.root
34        for char in prefix:
35            if char not in node.children:
36                return False
37            node = node.children[char]
38        return True

```

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     def __init__(self, key):
7         self.key = key
8         self.prio = random.randint(1, 1
9             << 30)
10        self.left = None
11        self.right = None
12        self.size = 1
13
14    def _sz(t):
15        return t.size if t else 0
16
17    def _upd(t):
18        if t:
19            t.size = 1 + _sz(t.left) +
20            _sz(t.right)
21
22    def _merge(a, b):
23        # assumes all keys in a < all keys
24        # in b
25        if not a or not b:
26            return a or b
27        if a.prio > b.prio:
28            a.right = _merge(a.right, b)
29            _upd(a)
30            return a
31        else:
32            b.left = _merge(a, b.left)
33            _upd(b)
34            return b
35
36    def _split(t, key):
37        # returns (l, r): l has keys < key,
38        # r has keys >= key
39        if not t:
40            return (None, None)
41        if t.key < key:
42            return (_split(t.right, key),
43                    t)
44        else:
45            return (t,
46                    _split(t.left, key))

```

```

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
107    def lower_bound(self, key):
108        # smallest key >= key; returns
109        # 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
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

```

```

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

```

```
    A[i][k] * B[k  
    ][j]) % mod  
  
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.

15 Matrix Operations

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

15.1 Matrix Multiplication

```

1 # Standard matrix multiplication - O(n
2     ^3)
3
4 def matmul(A, B):
5     n, m, p = len(A), len(A[0]), len(B
6         [0])
7     C = [[0] * p for _ in range(n)]
8
9     for i in range(n):
10        for j in range(p):
11            for k in range(m):
12                C[i][j] += A[i][k] * B[k
13                    ][j]
14
15     return C
16
17 # With modulo
18 def matmul_mod(A, B, mod):
19     n = len(A)
20     C = [[0] * n for _ in range(n)]
21
22     for i in range(n):
23         for j in range(n):
24             for k in range(n):
25                 C[i][j] = (C[i][j] +
26                             A[i][k] * B[k][j]) % mod
27
28     return C
29
30 # Matrix exponentiation - O(k * log(n))
31 def matpow(M, n):
32     if n < k:
33         return init[n]
34
35
36     # Transition matrix
37     # [a(n), a(n-1), ..., a(n-k+1)]
38     M = [[0] * k for _ in range(k)]
39     M[0] = coeffs # First row
40
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
54     # Multiply with initial state
55     result = 0
56     for i in range(k):
57         result = (result + result_matrix
58             [0][i] * state[i]) % mod
59
60     return result

```

```

55 # Example: Tribonacci T(n) = T(n-1) + T(n-2) + T(n-3)
56 def tribonacci(n, mod):
57     if n == 0: return 0
58     if n == 1 or n == 2: return 1
59
60     coeffs = [1, 1, 1]
61     init = [0, 1, 1]
62     return linear_recurrence(coeffs,
63                             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:
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
3 print(7 // 3) # 2
4 print(-7 // 3) # -3 (not -2!)
5
6 # For ceiling division toward zero:
7 def div_ceil(a, b):
8     return -(-a // b)
9
10 # Modulo with negative numbers
11 print((-5) % 3) # 1 (not -2!)
12 print(5 % -3) # -1
13
14 # List multiplication creates references
15 matrix = [[0] * m] * n # WRONG! All
16 # rows same object
17 matrix[0][0] = 1 # Changes all
18 # rows!
19
20 # Correct way
21 matrix = [[0] * m for _ in range(n)]
22
23 # Float comparison - don't use ==
24 a, b = 0.1 + 0.2, 0.3
25 print(a == b) # False!
26
27 # Use epsilon comparison
28 eps = 1e-9
29 print(abs(a - b) < eps) # True
30
31 # String immutability
32 s = "abc"

```

```

30 # s[0] = 'd' # ERROR!
31 s = 'd' + s[1:] # OK
32
33 # For many string mutations, use list
34 chars = list(s)
35 chars[0] = 'd'
36 s = ''.join(chars)
37
38 # Mutable default arguments - dangerous!
39 def func(arr=[]): # WRONG!
40     arr.append(1)
41     return arr
42
43 # Each call modifies same list
44 print(func()) # [1]
45 print(func()) # [1, 1]
46
47 # Correct way
48 def func(arr=None):
49     if arr is None:
50         arr = []
51     arr.append(1)
52     return arr
53
54 # Generator expressions save memory
55 sum(x*x for x in range(10**6)) # Memory
56 # efficient
57 sum([x*x for x in range(10**6)]) # Creates full list
58
59 # Ternary operator
60 x = a if condition else b
61
62 # Dictionary get with default
63 count = d.get(key, 0) + 1
64
65 # Matrix rotation 90 degrees clockwise
66 def rotate_90(matrix):
67     return [list(row) for row in zip(*matrix[::-1])]
68
69 # Matrix transpose
70 def transpose(matrix):
71     return [list(row) for row in zip(*matrix)]

```

16.5 Time Complexity Reference

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

- $O(1)$, $O(\log n)$: instant
 - $O(n)$: usually fine up to $\sim 10^7$ operations ($\sim 1\text{s}$)
 - $O(n \log n)$: 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 (assume roughly $\approx 10^7$ simple operations/s; actual throughput depends on implementation details and input shapes).

- Convert algorithm cost to operation count: roughly cost = $c \cdot f(n)$. If cost > 39
time_limit \times ops_per_sec, it will TLE.
40
41
 - When in doubt, aim one complexity class lower (e.g. prefer $O(n \log n)$ over $O(n^2)$ for n around 10^5).
42
43
 - Consider memory limits—some faster algorithms use more memory (e.g. segment trees vs. Fenwick tree).
44
45
46
 - For multivariate inputs, replace n with the product/dominant parameter (e.g. $n \cdot m$) and apply the same rules.
47
48
49
 - 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.
50
51

```

    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: $O(n \log n)$.

17 Computational Geometry

17.1 Basic Geometry

Description: Fundamental geometric operations for 2D points.

```

import math
# Point operations
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:

```

```

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):
            lower.pop()
        lower.append(p)

    # Build upper hull
    upper = []
    for p in reversed(points):
        while (len(upper) >= 2 and
               cross_product(upper[-2], upper[-1], p) <= 0):
            upper.pop()
        upper.append(p)

    # Remove last point (duplicate of first)
    return lower[:-1] + upper[:-1]

# Convex hull area
def polygon_area(points):
    # Shoelace formula
    n = len(points)
    area = 0

    for i in range(n):
        j = (i + 1) % n
        area += points[i][0] * points[j][1]
    area -= points[j][0] * points[i][1]

    return abs(area) / 2

```

17.3 Point in Polygon

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

```
def point_in_polygon(point, polygon):
```

```

# Ray casting algorithm
x, y = point
n = len(polygon)
inside = False

pix, p1y = polygon[0]
for i in range(1, n + 1):
    p2x, p2y = polygon[i % n]

    if y > min(p1y, p2y):
        if y <= max(p1y, p2y):
            if x <= max(pix, p2x):
                if p1y != p2y:
                    xinters = (y - p1y) * (p2x - pix) / \
                                         (p2y - p1y) + pix

    if pix == p2x or x <= xinters:
        inside = not inside

pix, p1y = p2x, p2y

return inside

```

```
    strip[j][1] - strip[i][1] < d:  
        d = min(d, dist(strip[i], strip[j]))  
        j += 1  
  
    return d  
  
return closest_recursive(  
    points_sorted_x, points_sorted_y)
```

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

```

8 Network Flow

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

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

```

from collections import deque,
defaultdict

def max_flow(graph, source, sink, n):
    # graph[u][v] = capacity from u to v
    # Build residual graph
    residual = defaultdict(lambda:
        defaultdict(int))
    for u in graph:
        for v in graph[u]:
            residual[u][v] = graph[u][v]

    def bfs_path():
        # Find augmenting path using BFS
        parent = {source: None}
        visited = {source}
        queue = deque([source])

        while queue:
            u = queue.popleft()

            if u == sink:
                # Reconstruct path
                path = []
                while parent[u] is not
None:
                    path.append((parent[
u], u))
                    u = parent[u]
                return path[::-1]

            for v in range(n):
                if v not in visited and
residual[u][v] > 0:
                    visited.add(v)
                    parent[v] = u
                    queue.append(v)

        return None

    max_flow_value = 0

    # Find augmenting paths
    while True:
        path = bfs_path()
        if path is None:
            break

        # Find minimum capacity along
        # path
        flow = min(residual[u][v] for u,
v in path)

        # Update residual graph

```

```

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

```

```

36             flow = self.dfs(v, sink,
37                         min(
38                             pushed, self.graph[u][v]),
39                             level,
40                             start)
41
42             if flow > 0:
43                 self.graph[u][v] -=
44                 flow
45                 self.graph[v][u] +=
46                 flow
47
48             return flow
49
50             start[u] += 1
51
52         return 0
53
54 def max_flow(self, source, sink):
55     flow = 0
56
57     while True:
58         level = self.bfs(source,
59                             sink)
60
61         if level is None:
62             break
63
64         start = [0] * self.n
65
66         while True:
67             pushed = self.dfs(source,
68                                 sink, float('inf'),
69                                 level,
70                                 start)
71
72             if pushed == 0:
73                 break
74                 flow += pushed
75
76         return flow

```

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
12
13    def bfs(self, source, sink):
14        # Build level graph
15        level = [-1] * self.n
16        level[source] = 0
17        queue = deque([source])
18
19        while queue:
20            u = queue.popleft()
21
22            for v in range(self.n):
23                if level[v] == -1 and
24                    self.graph[u][v] > 0:
25                    level[v] = level[u]
26                    queue.append(v)
27
28            return level if level[sink] !=
29 -1 else None
30
31    def dfs(self, u, sink, pushed, level
32           , start):
33        if u == sink:
34            return pushed
35
36        while start[u] < self.n:
37            v = start[u]
38
39            if (level[v] == level[u] + 1
40                and
41                self.graph[u][v] > 0):

```

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
28    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)
```