

# COMP4128 Notes

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## 1 C++ features

- *cin* is slower than *scanf()*, revert to *scanf* if necessary
- To insist on non-scientific notation, use `std::fixed` from `<iostream>`. To change number of decimal places use *setprecision()* from `<iomanip>`
- Use the string stream type to take in a string and extract integers from it. In other cases, *stoi()* might suffice.
- *pair* `<T,T>`
- The `<algorithm>` header provides useful functions like *fill()*, *search()*, *count()*, *max()*, *min()*, *swap(x,y)*, *sort()*, etc.
- *next\_permutation()* rearranges elements in a range into the lexicographically next permutation. This helps us run an exhaustive search (brute force).

- The following snippet iterates through all bit sequences of length *n* in which exactly *k* bits are set.

```
int selected[n];
fill(selected, selected+n-k, 0);
fill(selected+n-k, selected+n, 1);

do {
    // ...
} while (next_permutation(selected, selected+n));
```

Figure 1: Next permutation

## 2 Debugging

Debugging can be tricky in a time environment. The key is to not panic and analyse your code carefully.

## 2.1 TIME-LIMIT-EXCEEDED (TLE)

Check for the following in the order they are listed.

- Check your complexity analysis. Computers can run 200 million computations per second. If your  $n > 10,000$  then no amount of optimization will allow it to run quick enough. As a general rule, substitute your given max  $n$  into your complexity analysis and check to see if it runs quicker than 100 million. If I have  $n$  up to  $10^5$  and my complexity analysis is  $O(n \log n)$ , then this is  $10^5 \times \log(10^5) < 200,000,000$  and so our algorithm is quick enough.
- Check for infinite loops.
- If there are a lot of inputs via `Iostream`, you can speed this up by adding

---

```
int main() {  
    cin.tie(nullptr);  
    cin.sync_with_stdio(false);  
}
```

---

- Replace vectors with arrays
- replace stacks and queues with arrays
- replace small sets with bitsets
- Look for computations that are repeated
- Sort your data to begin with.
- Remove debugging print statements.
- Change order of indices in a 2D array to increase caching.  $grid[col][row] \Rightarrow grid[row][col]$ .
- Consider early exiting.

## 2.2 RUN-ERROR/RUNTIME-ERROR

- Check for accessing out of array bounds.
- Check that your variables are initialized before using them.
- Check for accessing into or deleting from empty data structures.
- Compile with all flags using `-Wall`. `-Wextra` and `-Wshadow` may also help.

## 2.3 WRONG-ANSWER

- Write test cases. Think of possible edge cases.
- Take your time and read what you're typing. Open your code in a different editor.
- You can try run a slow brute force algorithm and compare it against yours. The below script compares your solution with a brute force solution. *soln* is an your solution executable, *brute\_force* is the brute force executable and *gen* is a random data generator.

---

```
./gen > random.in; while diff <(<./soln < random.in)
<(<./brute_force < random.in); do echo "all good"; ./gen
> random.in; done
```

---

- Don't rely on sample cases, these are poor for actually testing a solution.
- Are there cases such that:
  1. a loop will never run
  2. a branch of an if-statement will not run
  3. the start is also the end? Where  $\text{left} = \text{right}$ ? Usually for situations where the size of some data structure or the assignment of a variable is 0.
  4. the answer is 0 or negative?
  5. floating numbers aren't printed with enough precision?
  6. you've misinterpreted the bounds? can  $N = 0$  or  $N = 1$ ? Is it true that  $\text{INT\_MIN} < \text{value} < \text{INT\_MAX}$ ?

## 3 Greedy Algorithms

Reduce the number of states explored to just a single best choice per stage. Never consider alternatives. Prove to yourself that your algorithm achieves the best solution at every stage. If there is a counter example to algorithm, then you may need to consider another programming technique.

### 3.1 Linear Sweep

Having an order is better than not having an order. Handle states one after the other rather than all at once. Sorting a data set then linear sweeping is never a bad first approach. Recall stabbing end points from activity selection, this is not uncommon in competitive programming.

## 3.2 Coordinate Compression

For most algorithms, the actual values of the coordinates is irrelevant. We only care about the relative order. If coordinates are up to 1 billion but there are only  $n \leq 100,000$  points, then there are only  $O(n)$  interesting coordinates.

A solution is to replace each coordinate by its rank among all coordinates so we can preserve the relative order of values while making the maximum coordinate  $O(n)$ . A simple sort or range tree can solve this issue. We can use *pair* or *tuple* to associate compressed and uncompressed coordinates.

---

```
#include <bits/stdc++.h>
using namespace std;
// coordinates -> ( compressed coordinates ).
map <int , int > coordMap;
void compress(vector <int >& values) {
    for (int v : values) coordMap[v] = 0;
    int cId = 0;
    for (auto it = coordMap.begin (); it != coordMap.end (); ++it) {
        it ->second = cId ++;
    }
    for (int &v : values) v = coordMap[v];
}
```

---

### 3.3 Binary Search

A monotonic relationship for our purposes is the ability to remove  $1/2$  our search space for each check of a value. Say our search space is increasing (sorted) for an array  $A$  with values  $1, 2, \dots, n$  and we're searching for some value  $x$ . If  $A[1/2] > x$ , then we can eliminate all  $A[1/2]..A[n]$ ; the right side of our search space.

Conversely, if  $A[1/2] < x$  then we can eliminate all  $1..A[1/2]$ ; the left side of our search space. This can be extended further to solving a key problem, given a monotone function, find the largest/smallest  $x$  such that  $f(x)$  is less than/greater than/equal to  $y$ .

---

```
#include <bits/stdc++.h>
using namespace std;
// Find the smallest X such that f(X) is true;
int binarysearch (function <bool(int)> f) {
    int lo = 0;
    int hi = 100000;
    int bestSoFar = -1;
    // Range [lo , hi];
    while (lo <= hi) {
        int mid = (lo + hi) / 2;
        if (f(mid)) {
            bestSoFar = mid;
            hi = mid - 1;
        } else {
            lo = mid + 1;
        }
    }
    return bestSoFar ;
}
```

---

### 3.4 Binary Search alternatives

We can avoid implementing a binary search all together by using the `< algorithm >` header. Given a sorted array you can use these functions `binary_search()`, `lower_bound()`, `upper_bound()` and `equal_range()`.

---

```
#include <algorithm>
#include <iostream>
using namespace std;
const int N = 100100;
int a[N];
int main () {
    int n;
    cin >> n;
    assert(n <= N);
    for (int i = 0; i < n; i++) cin >> a[i];
    assert( is_sorted (a, a+n));
    int x;
    cin >> x;
    bool found = binary_search (a, a+n, x);
    cout << (found ? "found " : "did not find ") << x;
    int y;
    cin >> y;
    int i = lower_bound (a, a+n, y) - a;
    if (i < n)
        cout << "a[" << i << "] = " << a[i] << " is the first entry
            to compare >= " << y;
    else
        cout << "all elements of a[] compare < " << y;
    int z;
    cin >> z;
    int j = upper_bound (a,a+n,z) - a;
    if (j < n)
        cout << "a[" << j << "] = " << a[j] << " is the first entry
            to compare > " << z;
    else
        cout << "all elements of a[] compare <= " << z;
}
```

---

### 3.5 Discrete Binary Search

A Discrete Binary search is particularly useful for converting problems from optimization problems to decision problems. "What is the smallest  $x$  for which you can..." => "Given the value  $x$ , can you...". Let  $f(x)$  be the outcome of the decision problem for a given  $x$ , so  $f$  is an integer values function with range  $\{0, 1\}$ .

It is sometimes the case in such problems that increasing  $x$  does not make

it any harder for the condition to hold (if it is true for  $x$  then it is also true for  $x + 1$ ). This  $f$  is all zeros up to the first 1, after which it is all ones. This is a monotonic function we can binary search on but particularly, we're trying to find the smallest  $x$  such that  $f(x) = 1$ .

## 4 Data structures

### 4.0.1 Vectors

- $O(1)$  access time
- contiguous block of memory
- `push_back()` adds to end of vector

### 4.1 Stacks

- `push()` and `pop()` operations in  $O(1)$
- Last in, first out
- `< stack >` header

### 4.2 Queues

- `push()` and `pop()` operations in  $O(1)$
- First in, first out
- `< queue >` header

### 4.3 Sets

- `< set >` header with  $O(\log n)$  random access
- Implemented as red-black trees
- Since sets are ordered, you can run `lower_bound(x)` and `upper_bound(x)` which returns the next element not less than  $x$
- Multiset in `< multiset >`
- Unordered multiset in `< unordered_multiset >`

### 4.4 Maps

- `< map >` header
- Acts a dictionary with  $O(\log n)$  random access
- Implemented as a red/black tree of key, value pairs
- Can use `lower_bound(x)` and `upper_bound(x)` for the same reason

## 4.5 Implementation of sets & maps

---

```
include <iostream >
#include <map >
#include <set >
using namespace std;
set <int > s;
map <int , char > m;
int main () {
    s.insert (2); s.insert (4); s.insert (1);
    m = {{1,'a'}, {4,'c'}, {2,'b'}};
    // Check membership :
    cout << (s.find (2) != s.end ()) << ' ' << (s.find (3) != s.end
        ()) << '\n'; // 1 0
    // NOT binary_search (s.begin (), s.end (), 2), which takes
        linear time
    // Access map:
    cout << m[1] << '\n'; // 'a'
    // WARNING : Access to non - existent data just silently adds it
        , avoid this.
    // cout << m[3] << '\n ' ; // null character
    // Lower and upper bounds :
    cout << *s. lower_bound (2) << '\n'; // 2
    // NOT * lower_bound (s.begin (), s.end (), 2), which takes
        linear time
    cout << *s. upper_bound (2) << '\n'; // 4
    auto it = m. lower_bound (2);
    cout << it ->first << ' ' << it ->second << '\n'; // 2 b
    // Move around with prev/next or increment / decrement
    cout << prev(it)->first << '\n'; // 1
    cout << (++it)->first << '\n'; //
}
```

---

## 4.6 Order Statistic Trees

The main issue with sets and maps is that they do not track index information. You can not query what the  $k$ -th number is or how many numbers are  $< x$ . We can support a similar operation with order statistic trees.

- `find_by_order( $x$ )` finds the  $x$ th element 0-indexed
- `order_of_key( $x$ )` outputs the number of elements that are  $< x$ .
- Both of these are still  $O(\log n)$ .



## 4.7 OST Implementation

---

```
#include <bits/stdc++.h>
#include <ext/pb_ds/assoc_container.hpp>
#include <ext/pb_ds/tree_policy.hpp>

using namespace __gnu_pbds;
using namespace std;

typedef tree<int, null_type, less<int>, rb_tree_tag,
            tree_order_statistics_node_update> ordered_set;

ordered_set myset;
int main() {
    myset.insert(2);
    myset.insert(4);
    myset.insert(1);
    printf("%d\n", *(myset.find_by_order(0))); // 1
    printf("%d\n", (int)myset.order_of_key(3)); // 2
    printf("%d\n", (int)myset.order_of_key(4)); // 2
    printf("%d\n", (int)myset.size()); // 3
}
```

---

## 4.8 Heaps

- push() and pop() in  $O(\log)$
- top() for the top of the heap
- < queue > header
- Heap proper: the value stored in every node is greater than its children
- Shape property: the tree is as close in shape to a complete binary tree as possible

## 4.9 Union-Find

Used to represent disjoint set of items. For some set of elements, union find supports:

- union( $x, y$ ): union the disjoint sets that contain  $x$  and  $y$ .
- find( $x$ ): return a canonical representative for the set that  $x$  is in.
- both operations take  $O(\log n)$  time with improved size heuristic implementation (below)

## 4.10 Union-Find Implementation

---

```
int parent[N];
int subtree_size[N];

void init(int n) {
    for (int i = 0; i < n; i++) {
        parent[i] = i;
        subtree_size[i] = 1;
    }
}

int root(int x) {
    // only roots are their own parents
    return parent[x] == x ? x : root(parent[x]);
}

void join(int x, int y) {
    // join roots
    x = root(x); y = root(y);
    // test whether already connected
    if (x == y) return;
    // size heuristic
    // hang smaller subtree under root of larger subtree
    if (subtree_size[x] < subtree_size[y]) {
        parent[x] = y;
        subtree_size[y] += subtree_size[x];
    } else {
        parent[y] = x;
        subtree_size[x] += subtree_size[y];
    }
}
```

---

## 4.11 Range trees

Data structures that support operations on a range of values. Can support any operator that is associative (sum, min, max, gcd, lcm, merge). This works by decomposing  $[0, n)$  into ranges such that:

- each item belongs to much fewer than  $n$  ranges
- any subarray can be decomposed into much fewer than  $n$  ranges.

Updates and queries over a range tree is  $O(\log n)$  with  $O(n)$  amount of nodes in the tree

## 4.12 Range Tree Query

---

```
const int N = 100100;
// the number of additional nodes created can be as high as the next
// power of two up from N ( $2^{17} = 131,072$ )
int tree[1<<18];

int n; // actual length of underlying array

// query the sum over [qL, qR) (0-based)
// i is the index in the tree, rooted at 1 so children are 2i and 2i+1
// instead of explicitly storing each node's range of responsibility
// [cL,cR), we calculate it on the way down
// the root node is responsible for [0, n)
int query(int qL, int qR, int i = 1, int cL = 0, int cR = n) {
    // the query range exactly matches this node's range of responsibility
    if (cL == qL && cR == qR)
        return tree[i];
    // we might need to query one or both of the children
    int mid = (cL + cR) / 2;
    int ans = 0;
    // query the part within the left child [cL, mid), if any
    if (qL < mid) ans += query(qL, min(qR, mid), i * 2, cL, mid);
    // query the part within the right child [mid, cR), if any
    if (qR > mid) ans += query(max(qL, mid), qR, i * 2 + 1, mid, cR);
    return ans;
}
```

---

## 4.13 Range Tree Update

---

```
// p is the index in the array (0-based)
// v is the value that the p-th element will be updated to
// i is the index in the tree, rooted at 1 so children are 2i and 2i+1
// instead of explicitly storing each node's range of responsibility
// [cL,cR), we calculate it on the way down
// the root node is responsible for [0, n)
void update(int p, int v, int i = 1, int cL = 0, int cR = n) {
    if (cR - cL == 1) {
        // this node is a leaf, so apply the update
        tree[i] = v;
        return;
    }
    // figure out which child is responsible for the index (p) being
    // updated
    int mid = (cL + cR) / 2;
    if (p < mid)
```

```

    update(p, v, i * 2, cL, mid);
else
    update(p, v, i * 2 + 1, mid, cR);
// once we have updated the correct child, recalculate our stored
// value.
tree[i] = tree[i*2] + tree[i*2+1];
}

```

---

#### 4.14 Range Tree Debugg

---

```

// print the entire tree to stderr
// instead of explicitly storing each node's range of responsibility
// [cL,cR), we calculate it on the way down
// the root node is responsible for [0, n)
void debug(int i = 1, int cL = 0, int cR = n) {
    // print current node's range of responsibility and value
    cerr << "tree[" << cL << "," << cR << "] = " << tree[i];

    if (cR - cL > 1) { // not a leaf
        // recurse within each child
        int mid = (cL + cR) / 2;
        debug(i * 2, cL, mid);
        debug(i * 2 + 1, mid, cR);
    }
}

```

---

## 5 Dynamic Programming

For recursive solutions, time complexity bound is  $O(t \times n^r)$  where  $t$  is the time spent in the recursive function,  $n$  is the number of recursive branches and  $r$  is the recursion depth.

For DP, it comes down to carefully determining the number of subproblems and the time taken to solve each subproblem. The idea of DP is to breakdown a large problem into a smaller subset of many subproblems that relate in some way. That is, I should be able to notice the transition of one subproblem to another. Rather than recurse into each subproblem like DNC, we pick and choose which subproblems to calculate and store it for future subproblems to utilise.

## 5.1 Top Down vs Bottom Up DP

Top down DP takes a mathematical recurrence and translates it directly to code. Subproblems are cached using a caching function; memoisation.

---

```
int f(int n) {
    // base cases
    if (n == 0 || n == 1) return 1;
    // return the answer from the cache if we already have one
    if (cache[n]) return cache[n];
    // calculate the answer and store it in the cache
    return cache[n] = f(n-1) + f(n-2);
}
```

---

Bottom-up DP starts at base cases and builds up answers one-by-one.

---

```
f[0] = 1, f[1] = 1;
for (int i = 2; i <= n; i++) f[i] = f[i-1] + f[i-2];
```

---

## 5.2 How to DP

1. Choose some order to do the problem in. How will you build up your solution?
2. Pick a tentative state/subproblem containing the parameters you think are necessary to determine the end result.
3. Test your state/subproblem is sufficient enough by trying to make a recurrence.

If you notice your DP is too slow consider the following:

1. Is everything in the state/subproblem necessary?
2. Is there a better order to solve states/subproblems in?
3. Can a data structure be used to speed up your recurrence?

## 5.3 2D DP - Integer Knapsack

---

```
int dp[N+2][S+1];

for (int i = N; i >= 1; --i) {
    // everything from larger i will be available here
    for (int r = 0; r <= S; ++r) {
        // we have declared the array larger, so if i == N, dp[i+1][...]
        // will be zero.
    }
}
```

```

    int m = dp[i+1][r];
    // bounds check so we don't go to a negative array index
    if (r - s[i] >= 0) m = max(m, dp[i+1][r-s[i]] + v[i]);
    dp[i][r] = m;
}
}

```

---

## 5.4 Exponential DP

Sometimes our DP states are very large;  $2^n$ . We thankfully have a work around this by representing the state as a bitset.

## 5.5 Bitset

A bitset is an integer which represents a set. The  $i$ th least significant bit is 1 if the  $i$ th element is in the set, and 0 otherwise. For example 01101101 represents the set 0, 1, 2, 3, 4, 6. We can now perform extremely quick operations to manipulate the set:

- Singleton set:  $1 \ll i$ .
- Set complement:  $\sim x$ .
- Set intersection:  $x \& y$ .
- Set union:  $x | y$ .
- Symmetric difference:  $x \wedge y$ .
- Membership test:  $x \& (1 \ll y)$ .
- Size of set (C++20): `< bit > header using popcount( $x$ )`.
- Size of set (before C++20): `_builtin_popcount( $x$ )`. Two underscores followed by the text is intended.
- Least significant bit:  $x \& (-x)$ .
- Iterate over all sets and subsets:

---

```

// for all sets
for (int set = 0; set < (1<<n); set++) {
    // for all subsets of that set
    for (int subset = set; subset; subset = (subset-1) & set) {
        // do something with the subset
    }
}

```

---

## 5.6 Roof Tiling

You want to tile your roof with  $n$  tiles in a straight line, each of which is either black or white. Due to regulations, for every  $m$  consecutive tiles on your roof, at least  $k$  of them must be black. Given  $n$ ,  $m$  and  $k$  ( $1 \leq n \leq 60, 1 \leq k \leq m \leq 15, m \leq n$ ), how many valid tiling are there?

Subproblem/State: Let  $f(i, S)$  be the number of ways to have tiled the first  $i$  tiles, such that out of the last  $m$  tiles, the ones that are black are exactly the ones in our set  $S$ .

Recurrence:  $f(i, S) = f(i-1, S \gg 1) + f(i-1, (S \gg 1) | (1 \ll (m-1)))$ , where  $f(i-1, (S \gg 1) | (1 \ll (m-1)))$  is taking the union of the set with an extra black tile (adding that tile) with the last  $m-1$ th tile (we care if this was black or white).

Time Complexity:  $O(n2^m)$ .

---

```
// base case
for (int set = 0; set < (1<<m); set++)
    dp[m%2][set] = (popcount(set) >= k);

for (int i = m+1; i <= n; i++) {
    fill(dp[i%2], dp[i%2] + (1<<M), 0);
    for (int set = 0; set < (1<<m); set++)
        if (popcount(set) >= k)
            dp[i%2][set] = dp[(i+1)%2][set>>1]
                + dp[(i+1)%2][(set>>1) | (1<<(m-1))];
}
// answer is sum over all sets of dp[n%2][set]
```

---

## 5.7 DP with Data Structures

Sometimes you have the correct state space, but the cost of recursion is too high. We can try to speed up our recursive step using data structures. If your states are doing some associative query taking  $O(n)$  time, then we can speed this using a range tree for  $O(\log n)$ .

## 5.8 Segment Cover

Consider the  $n + 1$  points on the real line  $0, 1, \dots, n$ . You are given  $m$  segments, each with a range  $[s_i, e_i]$  and a cost  $c_i$ . Output the minimum cost necessary to obtain a subset of the segments which covers all  $n + 1$  points. First line,  $n, m$ .  $1 \leq n, m \leq 100,000$ . The following  $m$  lines describe a segment as a triple  $(s_i, e_i, c_i)$ .

---

```
#include <bits/stdc++.h> // algorithm, iostream, utility, vector
using namespace std;
typedef long long ll;
const int N = 100100;
const ll INF = (1LL << 61);
int n, m;
vector<pair<int, ll>> segments[N]; // (start, cost)
ll dp[N];
int tree[1<<18]; // range min tree with point update
void update(int p, ll v, int i = 1, int cL = 0, int cR = n);
ll query(int qL, int qR, int i = 1, int cL = 0, int cR = n); // [qL, qR)

int main() {
    cin >> n >> m;
    for (int i = 0; i < m; i++) {
        int s, e, c;
        cin >> s >> e >> c;
        segments[e].emplace_back(s, c); // preprocess: collate by end
        point
    }

    for (int i = 0; i <= n; i++)
        update(i, INF);

    for (int i = 0; i <= n; i++) {
        dp[i] = INF;
        for (auto seg : segments[i]) {
            ll prevcost = seg.first == 0 ? 0 : query(seg.first-1, i);
            dp[i] = min(dp[i], prevcost + seg.second);
        }
        update(i, dp[i]);
    }
    cout << dp[n] << '\n';
}
```

---



## 6 Graph Algorithms

Represented with an adjacency list and occasionally an adjacency matrix.

### 6.1 Adjacency List

---

```
#include <iostream>
#include <vector>

int N = 1e6 + 5; // number of vertices in graph
vector<int> edges[N]; // each vertex has a list of connected vertices

void add(int u, int v) {
    edges[u].push_back(v);
    // Warning: If the graph has self-loops, you need to check this.
    if (v != u) {
        edges[v].push_back(u);
    }
}

// iterate over edges from u (since C++11)
for (int v : edges[u]) cout << v << '\n';

// iterate over edges from u (before C++11)
vector<int>::iterator it = edges[u].begin();
for (; it != edges[u].end(); ++it) {
    int v = *it;
    cout << v << '\n';
}

// or just a regular for loop will work too
for (unsigned int i = 0; i < edges[u].size(); i++) {
    cout << edges[u][i] << '\n';
}
```

---

## 6.2 BFS

---

```
vector<int> edges[N];
// dist from start. -1 if unreachable.
int dist[N];
// previous node on a shortest path to the start
// Useful for reconstructing shortest paths
int prev[N];

void bfs(int start) {
    fill(dist, dist+N, -1);
    dist[start] = 0;
    prev[start] = -1;

    queue<int> q;
    q.push(start);
    while (!q.empty()) {
        int c = q.front();
        q.pop();
        for (int nxt : edges[c]) {
            // Push if we have not seen it already.
            if (dist[nxt] == -1) {
                dist[nxt] = dist[c] + 1;
                prev[nxt] = c;
                q.push(nxt);
            }
        }
    }
}
```

---

## 6.3 DFS

---

```
// global arrays are initialised to zero for you
bool seen[N];

void dfs(int u) {
    if (seen[u]) return;
    seen[u] = true;
    for (int v : edges[u]) dfs(v);
}
```

---

## 6.4 DFS Trees

DFS alone is usually our traversal of choice to solve many problems. These problems can include: undirected cycle detection, connectivity, flood fill, etc. The main power of DFS is a specific invariant.

**Main Invariant:** By the time we return from a vertex  $v$  in our DFS, we have visited every vertex  $v$  can reach that does not require passing through an already visited vertex.

In undirected graphs, if we only consider edges that visit a vertex for the first time, these edges form a tree. All other edges are called "back-edges". **These back edges always go directly upwards to an ancestor in the DFS tree.** It often helps to bring structure to graphs by creating a DFS tree to solve a problem.

## 6.5 Bridge finding

Consider  $G$  an undirected, simple, connected graph. A bridge of  $G$  is an edge whose removal disconnects  $G$ . Output all bridges of  $G$ .

**Input:** First line, 2 integers  $V, E$  the number of vertices and edges respectively. Next  $E$  lines, each a pair,  $u_i v_i$ . Guaranteed  $u_i \neq v_i$  and no unordered pair appears twice.  $1 \leq V, E, \leq 100,000$ .

We claim that **back edges can not be bridges** and **a tree edge is a bridge iff there is no back edge going "past it"**. More formally, it is enough to know within each subtree of the DFS tree, what the highest node a back edge in this subtree can reach. We can compute this recursively using a DFS tree.

---

```
int preorder[N]; // order of pushing to the DFS stack initialise to -1
int T = 0; // counter for preorder sequence
int reach[N]; // reach[u] is the smallest preorder index of any vertex
               reachable from u
vector<pair<int,int>> bridges;

void dfs(int u, int from = -1) {
    preorder[u] = T++;
    reach[u] = preorder[u];

    for (int v : edges[u]) if (v != from) {
        if (preorder[v] == -1) { // u--v is a tree edge
            // v hasn't been seen before, so recurse
            dfs(v, u);
            // if v can't reach anything earlier than itself
            // then u--v is a bridge
            if (reach[v] == preorder[v]) bridges.emplace_back(u,v);
            // anything reachable from v is reachable from u
            reach[u] = min(reach[u], reach[v]);
        }
    }
}
```

---

## 6.6 Directed Cycle Detection

We can use the concept of back edges to detect directed cycles. If there is a cycle  $C$  and  $u \in C$  is the first vertex our DFS visits in the cycle then all vertices in the cycle will be in the subtree of  $u$  in the DFS tree. Hence this subtree must have some backedge to  $u$ .

We can rephrase this algorithm as checking if any edge visits a vertex we are still recursing from. This means we reach a vertex  $v$  that we are still trying to build the subtree for. So  $v$  is an ancestor. We simply just mark the vertex as active in the preorder step and unmark it in the post order step.

---

```
// the vertices that are still marked active when this returns are the
// ones in the cycle we detected
bool has_cycle(int u) {
    if (seen[u]) return false;
    seen[u] = true;
    active[u] = true;
    for (int v : edges[u]) {
        if (active[v] || has_cycle(v)) return true;
    }
    active[u] = false;
    return false;
}
```

---

## 6.7 Tree representation

Trees are undirected, connected, have unique simple paths between any two vertices,  $E = V - 1$ , no cycles and where any edge removal disconnects the graph.

---

```
const int N = 1e6 + 5;
// We need the list of edges to construct our representation. But we
// don't use it afterwards.
vector<int> edges[N];
int par[N]; // Parent. -1 for the root.
vector<int> children[N]; // Your children in the tree.
int size[N]; // As an example: size of each subtree.

void constructTree(int c, int cPar = -1) {
    par[c] = cPar;
    size[c] = 1;
    for (int nxt : edges[c]) {
        if (nxt == par[c]) continue;
        constructTree(nxt, c); children[c].push_back(nxt);
        size[c] += size[nxt];
    }
}
```

---

With the tree representation of a graph, every node has a parent (except the root) and a list of children. We can do a lot by recursing through the children array. A linear sweep on the array is just a DFS, a DP on the array is to DP on the tree and we can perform range tree queries for paths over a tree.

## 6.8 Shortest Distance on a tree - LCA

Given a labeled rooted tree  $T$ , and  $Q$  queries of form, "What is the vertex furthest away from the root in the tree that is an ancestor of vertices labeled  $u$  and  $v$ ?".

**Input:** A rooted tree  $T(1 \leq |T| \leq 1,000,000)$ , as well as  $Q(1 \leq Q \leq 1,000,000)$  pairs of integers  $u$  and  $v$ .

**Algorithm** Instead of walking up single edges, in a tree representation we can use our precomputed  $\text{parent}[u][k]$  to greedily continue moving up by the largest power of 2 possible until we're at the desired vertex. We're trying to find the **Lowest common ancestor (LCA)** in this step.

Fist, move both  $u$  and  $v$  to the same depth then binary search for the maximum amount you can jump up without reaching the same vertex. The parent of that vertex is the LCA. The implementation sets a search range that is jumping up in decreasing powers of 2 order and rejecting any jumps that results in  $u$  and  $v$  being at the same vertex.

## 6.9 LCA - Preprocess

---

```
// parent[u][k] is the 2^k-th parent of u
void preprocess() {
    for (int i = 0; i < n; i++) {
        // assume parent[i][0] (the parent of i) is already filled in
        for (int j = 1; (1<<j) < n; j++) parent[i][j] = -1;
    }

    // fill in the parent for each power of two up to n
    for (int j = 1; (1<<j) < n; j++) {
        for (int i = 0; i < n; i++) {
            if (parent[i][j-1] != -1) {
                // the 2^j-th parent is the 2^(j-1)-th parent of the 2^(j-1)-th
                // parent
                parent[i][j] = parent[parent[i][j-1]][j-1];
            }
        }
    }
}
```

---

## 6.10 LCA - Querying

---

```
int lca (int u, int v) {
    // make sure u is deeper than v
    if (depth[u] < depth[v]) swap(u,v);

    // log2s[i] holds the largest k such that 2^k <= i
    // precompute by DP: log2s[i] = log2s[i/2] + 1
    for (int i = log2s[depth[u]]; i >= 0; i--) {
        // repeatedly raise u by the largest possible power of two until it
        // is the same depth as v
        if (depth[u] - (1<<i) >= depth[v]) u = parent[u][i];
    }

    if (u == v) return u;

    for (i = log2s[depth[u]]; i >= 0; i--)
        if (parent[u][i] != -1 && parent[u][i] != parent[v][i]) {
            // raise u and v as much as possible without having them coincide
            // this is important because we're looking for the lowest common
            // ancestor, not just any
            u = parent[u][i];
            v = parent[v][i];
        }

    // u and v are now distinct but have the same parent, and that parent
    // is the LCA
    return parent[u][0];
}
```

---

We know what the paths between  $u$  and  $v$  look like, it's  $u \rightarrow \text{lca}$  followed by  $\text{lca} \rightarrow v$ . We also need the lengths of certain ranges, similar to a range tree. Sum has an inverse and so we can use a cumulative sum data structure instead.

Store  $\text{dist}(\text{root}, u)$  for all  $u$  then  $\text{dist}(u, v) = \text{dist}(\text{root}, u) + \text{dist}(\text{root}, v) - 2 \times \text{dist}(\text{root}, \text{lca})$ .

## 6.11 LCA Algorithm

---

```
#include <bits/stdc++.h>
using namespace std;

const int MAXN = 100000, LOGN = 18;
struct edge { int nd; long long d; };
int parent[MAXN+5][LOGN];
long long distToRoot[MAXN+5];
vector<edge> children[MAXN+5];
// Code to set up LCA and tree representation
void construct_tree(int c, int cPar = -1);
int lca(int a, int b);

void calc_dists_to_root(int c) {
    for (auto edg : children[c]) {
        distToRoot[edg.nd] = distToRoot[c] + edg.d;
        calc_dists_to_root(edg.nd);
    }
}

long long find_tree_dist(int a, int b) {
    int cLca = lca(a, b);
    return distToRoot[a] + distToRoot[b] - 2 * distToRoot[cLca];
}
```

---

The time complexity of this algorithm is  $O(n \log n)$  time preprocessing and  $O(\log n)$  time per query. You must do jumps to the power of 2.

## 6.12 Directed Acyclic Graphs (DAG)

1. Every DAG has a minimal vertex, one with no incoming edges.
2. Every DAG can be linearly ordered. There must be some ordered of vertices such that edges only go from  $v_i \rightarrow v_j$  where  $i < j$ . This is also known as a topological sort.

## 6.13 Topological sort

An ordering of vertices that has the property that if some vertex  $u$  has a directed edge pointing to another vertex  $v$ , then  $v$  comes after  $u$  in the ordering. If the graph has a cycle, then there does not exist a valid topological sort.

Due to the DFS invariant, for an acyclic graph every vertex after  $v$  in the sort is returned from before  $v$ . A vertex is only added after it's children have been visited (and thus added).

---

```

// if the edges are in ASCENDING order of node number,
// this produces the lexicographically GREATEST ordering

void dfs(int u, vector<int>& postorder) {
    if (seen[u]) return;
    seen[u] = true;
    for (int v : edges[u]) dfs(v, postorder);
    postorder.push_back(u);
}

vector<int> topsort() {
    vector<int> res;
    for (int i = 0; i < n; i++) dfs(i, res);
    reverse(res.begin(), res.end()); // #include <algorithm>
    return res;
}

```

---

## 6.14 Strongly Connected Components (SCC)

An SCC is a maximal subset of vertices of a directed graph such that every vertex in the subset can reach every other vertex in that component. Condensing every SCC to a single vertex (condensation graph) results in a directed acyclic graph. We have two SCC algorithms.

## 6.15 Targan's Algorithm

---

```

// we will number the vertices in the order we see them in the DFS
int dfs_index[MAX_VERTICES];
// for each vertex, store the smallest number of any vertex we see
// in its DFS subtree
int lowlink[MAX_VERTICES];

// explicit stack
stack<int> s; // #include <stack>
bool in_stack[MAX_VERTICES];

// arbitrarily number the SCCs and remember which one things are in
int scc_counter;
int which_scc[MAX_VERTICES];

void connect(int v) {
    // a static variable doesn't get reset between function calls
    static int i = 1;
    // set the number for this vertex
    // the smallest numbered thing it can see so far is itself
    lowlink[v] = dfs_index[v] = i++;
}

```



```

s.push(v);
in_stack[v] = true;

for (auto w : edges[v]) { // for each edge v -> w
    if (!dfs_index[w]) { // w hasn't been visited yet
        connect(w);
        // if w can see something, v can too
        lowlink[v] = min(lowlink[v], lowlink[w]);
    }
    else if (in_stack[w]) {
        // w is already in the stack, but we can see it
        // this means v and w are in the same SCC
        lowlink[v] = min(lowlink[v], dfs_index[w]);
    }
}
// v is the root of an SCC
if (lowlink[v] == dfs_index[v]) {
    ++scc_counter;
    int w;
    do {
        w = s.top(); s.pop();
        in_stack[w] = false;
        which_scc[w] = scc_counter;
    } while (w != v);
}
}

```

---

## 6.16 Kosaraju's Algorithm

---

```

void dfs(int u) {
    if (seen[u])
        return;
    seen[u] = true;
    for (int v : edges[u])
        dfs(v);
    postorder[p++] = u;
}

void dfs_r(int u, int mark) {
    if (seen_r[u])
        return;
    seen_r[u] = true;
    scc[u] = mark;
    for (int v : edges_r[u])
        dfs_r(v, mark);
}

int compute_sccs() {

```

```

int sccs = 0;
for (int i = 1; i <= n; i++)
    if (!seen[i])
        dfs(i);
for (int i = p - 1; i >= 0; i--) {
    int u = postorder[i];
    if (!seen_r[u] // ignore visited vertices
        dfs_r(u, sccs++);
}
return sccs;
}

```

---

## 6.17 Minimum Spanning Trees (MST)

A spanning tree for some graph  $G$  is a subgraph of  $G$  that is a tree, and also connects all of the vertices of  $G$ . A MST is a spanning tree with minimum sum of edge weights.

To construct a minimum spanning tree of some graph  $G$ , we maintain a set of spanning forests, initially composed of just the vertices of the graph and no edges, and we keep adding edges until we have a spanning tree. If we add  $|V| - 1$  edges and avoid constructing cycles, we'll have a spanning tree.

We restrict ourselves to only edges that cross components that we haven't connected yet. If  $e$  has a minimum weight of edges that connects components we haven't connected yet, then there is a spanning tree containing  $e$ . We have Kruskal's Algorithm that does this in  $O(|E| \log |E|)$  time.

---

```

struct edge {
    int u, v, w;
};
bool operator< (const edge& a, const edge& b) {
    return a.w < b.w;
}

edge edges[N];
int p[N];
int root (int u); // union-find root with path compression
void join (int u, int v); // union-find join with size heuristic

int mst() {
    sort(edges, edges+m); // sort by increasing weight
    int total_weight = 0;
    for (int i = 0; i < m; i++) {
        edge& e = edges[i];
        // if the endpoints are in different trees, join them
        if (root(e.u) != root(e.v)) {

```

```
        total_weight += e.w;
        join(e.u, e.v);
    }
}
return total_weight;
}
```

---

## 7 Shortest Paths

Given a weighted directed graph  $G$  with two specific vertices  $s$  and  $t$ , we want to find the shortest path that goes between  $s$  and  $t$  on the graph. Generally, algorithms which solve the shortest path problem also solve the single source shortest path problem, which computes shortest paths from a single source vertex to every other vertex.

### 7.1 BFS

BFS can solve shortest paths in unweighted graphs or graphs with edge weight of 1.

### 7.2 Single Source Shortest Paths

Most single source shortest paths algorithms rely on the basic idea of building shortest paths iteratively. At any point, we keep what we think is the shortest path to each vertex and we update this by "relaxing" edges.

**Relax( $u, v$ ):** if the currently found shortest path from our source  $s$  to a vertex  $v$  could be improved by using the edge  $(u, v)$ , update it. For non-negative edge weights, we can get away with only relaxing vertices for which we know the optimal distance. But this becomes trickier with negative edge weights.

If we keep track for each  $v$  of its most recently relaxed incoming edge, we can find the actual path from the source to  $v$ . This is because, for each  $v$  we know the vertex we would've come from if we followed the shortest path from the source. We can work backwards from  $v$  to the source to find the shortest path from the source to  $v$ . So keep relaxing edges until we can't anymore, then we'll have a shortest path.

### 7.3 Dijkstra's Algorithm

For non-negative edge weights, visit each vertex  $u$  starting from the source  $s$ . Whenever we visit the vertex  $u$ , we relax all of the edges coming out of  $u$ . We always process the next unprocessed vertex with the smallest distance from the source. Dijkstra's uses a priority queue keyed on each vertex's current known shortest distance from the source. This runs in  $O(E \log V)$  using a priority queue.

## 7.4 Dijkstra's Implementation

---

```
#include <queue>
using namespace std;

typedef pair<int, int> edge; // (distance, vertex)
const int N = 100100;
vector<edge> edges[N];
int dist[N];
bool seen[N];
priority_queue<edge, vector<edge>, greater<edge>> pq;

void dijkstra (int s) {
    fill(seen, seen+N, false);
    pq.push(edge(0, s)); // distance to s itself is zero
    while (!pq.empty()) {
        // choose (d, v) so that d is minimal
        // i.e. the closest unvisited vertex
        edge cur = pq.top();
        pq.pop();
        int v = cur.second, d = cur.first;
        if (seen[v]) continue;

        dist[v] = d;
        seen[v] = true;

        // relax all edges from v
        for (edge nxt : edges[v]) {
            int u = nxt.second, weight = nxt.first;
            if (!seen[u])
                pq.push(edge(d + weight, u));
        }
    }
}
```

---

## 7.5 Bellman-Ford Algorithm

If a graph has no negative cycles, then all shortest paths from  $u$  have length  $\leq |V| - 1$ . Conversely, a negative cycle implies there is a shortest path of length  $|V|$  better than any path of length  $|V| - 1$ . So we should instead build up shortest paths by number of edges, not just from  $u$  outwards.

Bellman-Ford involves trying to relax every edge of the graph  $|V| - 1$  times and update our tentative shortest paths each time. Because every shortest path has at most  $|V| - 1$  edges, if after all these global relaxations, relaxations can still be made, then there exists a negative cycle.

**Subproblem:**  $f(v, k)$ , the shortest  $s \rightarrow v$  path using up to  $k$  edges.

**Recurrence:**  $f(v, k) = \min\{f(v, k-1), \min_{u: e=(u,v) \in E} (f(u, k-1) + w_e)\}$

**Base cases:**  $f(v, 0) = 0$  for  $v = s$  or  $\infty$  otherwise.

**Time complexity:**  $O(VE)$ .

## 7.6 Bellman-Ford implementation

---

```
struct edge {
    int u, v, w; // u -> v of weight w
    edge(int _u, int _v, int _w) : u(_u), v(_v), w(_w) {}
};
vector<int> dist(n);
vector<edge> edges;

// global relaxation: try to relax every edge in the graph
// returns whether any distance was updated
bool relax() {
    bool relaxed = false;
    for (edge e : edges) {
        // we don't want to relax an edge from an unreachable vertex
        if (dist[e.u] != INF && dist[e.v] > dist[e.u] + e.w) {
            relaxed = true;
            dist[e.v] = dist[e.u] + e.w;
        }
    }
    return relaxed;
}

// Puts distances from source vertex 0 in dist
// Returns true if there is a negative cycle, false otherwise.
// NOTE: You can't trust the dist array if this function returns True.
bool bellman_ford() {
    fill(dist.begin(), dist.end(), INF);
    dist[0] = 0;
    // V-1 global relaxations
    // if no updates are made in an entire round, we can early exit
    // SPFA optimises this further
    for (int i = 0; i < n - 1; i++) if (!relax()) break;
    // there is a negative cycle iff any edge can be relaxed further
    // therefore try a Vth global relaxation
    // true if any changes made, false otherwise
    // can micro-optimize by early exiting when the first change is made
    return relax();
}
```

---

## 7.7 Floyd-Warsall Algorithm

Finds all pairs of shortest paths between all vertices in a graph using dynamic programming in  $O(V^3)$  time.

**Subproblem:** Let  $f(u, v, i)$  be the length of the shortest path between  $u$  and  $v$  using only the first  $i$  vertices as intermediate vertices. The key is to build this up for increasing values of  $i$ .

**Base case:** Then  $f(u, u, 0) = 0$  for all vertices  $u$ ,  $f(u, v, 0) = w_e$  if there is an edge from  $u$  to  $v$  and  $\infty$  otherwise.

**Recurrence:**  $f(u, v, i) = \min(f(u, v, i-1), f(u, i, i-1) + f(i, v, i-1))$ .

## 7.8 Floyd-Warshall Implementation

---

```
// the distance between everything is infinity
for (int u = 0; u < n; ++u)
    for (int v = 0; v < n; ++v)
        dist[u][v] = INF;

// update the distances for every directed edge
for (edge e : edges)
    // each edge u -> v with weight w
    dist[e.u][e.v] = e.w;

// every vertex can reach itself
for (int u = 0; u < n; ++u)
    dist[u][u] = 0;

for (int i = 0; i < n; i++)
    for (int u = 0; u < n; u++)
        for (int v = 0; v < n; v++)
            // dist[u][v] is the length of the shortest path from u to v using
            // only 0 to i as intermediate vertices
            // now that we're allowed to also use i, the only new path that
            // could be shorter is u -> i -> v
            dist[u][v] = min(dist[u][v], dist[u][i] + dist[i][v]);
```

---

## 7.9 Reconstructing Paths

As well as keeping track of a distance table, any time the improved path (via  $i$ ) is used, note that the next thing on the path from  $u$  to  $v$  is going to be the next thing on the path from  $u$  to  $i$ . We should already know this since we're keeping track of it. When initializing the table with the edges in the graph, don't forget to set  $v$  as next on the path from  $u$  to  $v$  for each edge  $u \rightarrow v$ .