

* hash function = translation + compression

* Two ways of compressing the output of the hash

- function $h(t)$ into range $[0..m]$. m : table size
 - $|h(t)| \bmod m$ (m should be a prime)
 - $(a \cdot h(t) + b) \bmod m$ (a and b are prime numbers → used if you can't control m)

* separate chaining: "searching for a key is $\Theta(\frac{N}{m})$ " (erasing) → "removing a key is $\Theta(1)$ " (key → $\frac{N}{m}$)

* inserting a key is $\Theta(1)$ if duplicate keys allowed $\Theta(\frac{N}{m})$ if duplicates are not allowed

* open addressing: linear, quadratic probing + double hashing

* linear probing disadvantage: primary clustering

* evenly distribute $\Theta(1)$ cluster (half, $\frac{1}{3} \dots$) $\Theta(n)$

* quadratic hashing disadvantage: secondary clustering.

* quadratic residues $\equiv (e+j^2) \bmod m$ only map to some collision number

* double hashing: $t(key) + j \cdot t'(key)$ second hash that's used if there's collision

* if linear probing is used and the hash function distributes keys evenly, the expected number of probes to successfully search for an existing element is $\frac{1}{2}(1 + \frac{1}{1-\alpha})$ $\alpha = \frac{d}{m}$ → load factor $\frac{m}{n}$ → size of table

* complexity of std::unordered_map <= worst case $\Theta(n)$ average case $\Theta(1)$ + worst case $\Theta(n)$

* Adjacency Matrix

A $|V| \times |V|$ matrix, uses $\Theta(V^2)$ space to represent dense graphs

$O(|E|)$ check edge exists between $O(|V|^2)$ iterate over all edges

$O(|E|)$ add edge to graph

$O(|E|)$ remove edge from graph

* DFS and BFS adjacency list $\Theta(|V+E|)$, adjacency matrix $\Theta(|V|^2)$ Building "forest"

* Prim's Algorithm: Matrix (linear search) $\Theta(|V|^2)$ dense works on negative edges fails on directed graphs list (binary heap), $\Theta(E \log V)$ sparse $\Theta(|E| \log E) = \Theta(|E|)$

* Kurskal's Algorithm: $\Theta(|E| \log |E|)$ / $\Theta(|E| \log |V|)$ sparse 0-1 Knapsack capacity fractional knapsack greedy $\Theta(n \log n)$

* Brute Force $\Theta(n^2)$ DP $\Theta(nC)$ # items

$F(i, c) = \begin{cases} 0, & \text{if } i = 0 \\ \max(F(i-1, c), F(i-1, c - w_i) + v_i), & \text{if } i > 0, c > 0 \end{cases}$

* Dijkstra's Algorithm: recording v, k_v, d_v, p_v of children, can be stored as a binary tree

Adjacency matrix (linear search) = $\Theta(|V|^2)$ dense
sparse

Adjacency list (binary heap) = $\Theta(|E| \log |V|)$

* Branch and Bound: our current best estimate of partial seq

- minimization: keep upper bound, change lower bound optimistic underestimate
- maximization: keep lower bound, change upper bound optimistic overestimate

* Backtracking → solving N-Queens (constraint satisfaction: satisfying constraint is not necessarily optimal)

* For separate chaining, collisions only occur between elements that have the same hash value. But that's not true for open addressing

* array-based binary tree

- insert (Best Case) $\Theta(1)$ assuming no duplicate elements
- insert (Worst Case) $\Theta(n)$
- remove key (Worst Case) $\Theta(n)$
- Find Parent $\Theta(1)$ *
- Find child $\Theta(1)$

* pointer-based binary

- insert (Best Case) $\Theta(1)$
- insert (Worst Case) $\Theta(n)$
- remove (Worst Case) $\Theta(n)$
- Find Parent $\Theta(n)$
- Find child $\Theta(1)$

* Space Worst Case $\Theta(2^n)$ Space Worst Case $\Theta(n)$ Insertion into AVL tree = $\Theta(1)$ or $\Theta(n)$ Deletion from AVL = $\Theta(\log n)$ *

* AVL:

AVL:	Best-Case	Average	Worst-Case
Finding a value	$\Theta(1)$	$\Theta(\log n)$	$\Theta(\log n)$
Inserting a value	$\Theta(\log n)$	$\Theta(\log n)$	$\Theta(\log n)$
Deleting a value	$\Theta(\log n)$	$\Theta(\log n)$	$\Theta(\log n)$

* std::map<>, for search, insert, delete = $\Theta(\log n)$ average and worst-case time

* A "unique" shortest path must be included in the MST Depth = n , space = $2^0 + 2^1 + 2^2 + \dots + 2^{n-2} + 2^{n-1} = 2^n - 1 = \Theta(2^n)$

A tree with at most k children: $1 \left(\frac{1-k^n}{1-k} \right) = \Theta(k^n)$

* As d increases, the performance of separate chaining does not deteriorate as quickly as the performances of double the size of the hash table if the load factor > 0.5 open addressing methods

* Bottom-up DP can be used any time top-down is used

* DP can not be used to solve a problem that can also be solved using divide and conquer

* A proper or full binary tree where every node either has 0 or 2 children. (all non-leaf nodes in a proper binary tree have 2 children)

* A complete binary tree is a tree in which every depth, except possibly the last, is completely filled. If the last depth (the bottom row) is not completely filled, then all nodes that do exist at that depth must be filled from left to right with no gaps

* General trees, where each node can have any number of children, can be stored as a binary tree

* Shortest path

- ① DFS: single-source shortest path problem in $\Theta(|V+E|)$ or $\Theta(|V|^2)$ time if acyclic graph
- ② BFS: unweighted or same-weight graphs
- ③ Dijkstra's: non-negative weighted/unweighted graphs (can also be directed)

* Only connected graphs have a MST

* Given a connected graph with n vertices, $n-1$ edges are in the MST of the graph

* 2D vector: $\text{vector} < \text{vector} < T >$ $\text{vec_2d}(m, \text{vector} < T > (\text{on}, \text{data}))$

class PersonComparator {

public:

```
bool operator() (const Person & p1, const Person & p2) const {
    return p1.get_age() < p2.get_age();
}
```

Σ

priority_queue<Person, vector<Person>, PersonComparator> myPQ;

$$\text{balance factor} = \text{left height} - \text{right height}$$

