#### 1 Java

#### 1.1 Access Modifier

C: Class; P: Package; SC: Subclass; W: World

Modifier	С	Р	SC	W
public	Y	Y	Y	Y
protected	Y	Y	Y	N
none	Y	Y	N	N
private	Y	N	N	N

### 1.2 Comparable

Implement Comparable<T>:

int compareTo(T o)

#### 1.3 hashCode

- If two objects are equal, hashCode must return the same result
- hashCode must return the same result hen invoked on the same object more than once

int hashCode() {}
boolean equals(Object o) {}

### 2 Algorithmic Analysis

Amortized Cost Algo has amortized cost T(n) if  $\forall$  k  $\in$  Z, cost of k operations is  $\leq kT(n)$ 

#### 2.1 Master's Theorem

 $T(n) = aT(\frac{n}{b}) + f(n)$  where  $a \ge 1$ , b > 1

- $f(n) \in O(n^c)$ ,  $c < \log_b a$ then  $T(n) \in \Theta(n^{\log_b a})$
- $f(n) \in O(n^c \log^k n), c = \log_b a$ then  $T(n) \in \Theta(n^c \log^{k+1} n)$
- $f(n) \in O(n^c), c > \log_b a$  and  $\exists k$ st  $af(\frac{n}{b}) \le kf(n)$  then  $T(n) \in \Theta(n^{\log_b a})$

#### 2.1.1 Common Ones

- $T(n) = T(n/2) + \Theta(1) = O(\log n)$  (Binary Search)
- $T(n) = 2T(n/2) + \Theta(1) = O(n)$  (Binary Tree Traversal)
- $T(n) = 2T(n/2) + \Theta(\log n) = O(n)$  (Optimal Sorted Matrix Search)

```
• T(n) = 2T(n/2) + O(n) = O(n \log n)
(Merge Sort)
```

## 3 Abstract Data Types

**Bag** Insert(i); Draw()

Stack (LIFO) empty(); peek(); pop() push(i)

Queue (FIFO) add(); offer(i); peek(); poll(); remove()

 ${\bf Dequeue} \ \ {\bf double\text{-}ended} \ \ {\bf queue}$ 

### 4 Searching

• Binary Search  $O(\log n)$ 

```
int binarySearch(int[] arr, int key) {
    int start = 0, end = arr.length - 1;
    int found = -1;
    while(start <= end) {</pre>
        int mid = start + (end - start)/2;
        if(arr[mid] < key) {</pre>
            start = mid + 1:
        } else if(arr[mid] > key) {
            end = mid - 1:
        } else {
            found = mid;
            // if we want first instance
            end = mid - 1;
            // if we want last instance
            // start = mid + 1:
    return found;
```

One sided Binary Search Suppose one side is bounded, eg  $[1, \infty)$ . Use the sequence  $[1,2,4,8,16..., 2^k...]$  If it works for  $2^k$ , then search on  $[2^{k-1}, 2^k]$ 

**Peak Finding** A[j] in array A is peak if (i) A[j] > A[j-1] (ii) A[j] > A[j+1]. If only one item in array, vacously true

**1D Peak Finding**  $O(\log n)$  D&C

 $\label{eq:analytic_problem} \begin{array}{l} \text{if } a[n/2] < a[n/2-1] \text{ look at } 1..n/2-1 \\ \text{else if } a[n/2] < a[n/2+1] \text{ look at } n/2+1..n \\ \text{else return } a[n/2] \end{array}$ 

**2D Peak Finding** O(m+n) D&C

find max in border + cross O(m+n)
if max is peak return
else go into quadrant with higher number

### 5 Sorting

Bubble Sort Stable, In-place, W&A  $O(n^2)$ , B O(n), S O(1); Invariant: At iteration i , the sub-array A[1 .. i] is sorted and any element in A[i + 1 .. A . size] is greater or equal to any element in A[1 .. i]

Selection Sort In-Place, Unstable; find minimum element and swap. W,A,B  $O(n^2)$ , S O(n/1); Invariant: a[0...i-1] is sorted all entries in a[i..n-1] are larger than or equal to the entries in a[0...i-1]

Insertion Sort In-place, Stable; W  $O(n^2)$ , B  $O(n \log n)$ , S O(n); Invariant: The subarray a[i] consists of the original elements in sorted order.

Merge Sort Stable, In Place; W/B  $O(n \log n)$ , S O(n)

Quick Sort In-place, Unstable; W  $O(n^2)$ A/B  $O(n \log n)$  S  $O(\log n)$ 

### 6 Geometric Algorithms

### **6.1** Jarvis March O(hn)

- 1. Find somewhere to start, e.g. y-min coordinate
- 2. Add point with maximum angle from horizon O(n)
- 3. Keep adding points with maximum angle from previous

# **6.2** Line Intersection Algorithm $O(n \log n)$

- 1. Divide into two equal size sets (along vertical line)
- 2. Recursively find convex hulls (base case 3 points)
- 3. Merge convex hulls
  - (a) Find upper tangent lines
    - i. while (u, v, w) clockwise, decrement v

- ii. while (v, w, z) clockwise, increment w
- (b) Find lower tangent lines
  - i. while (w, v, u) clockwise, increment v
  - ii. while (z, v, u) clockwise, decrement w

### **6.3** Quick Hull $O(n \log n)$

- 1. Choose pivot, construct two subproblems, delete interior points
- 2. recurse on subproblems

### 7 Trees

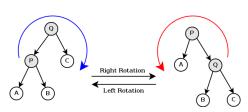
### 7.1 Binary Trees (height h)

 $h(v) = max\left(h(v.left), h(v.right)\right) + 1$ 

- BST: left ST < key < right ST
- traversal O(n) IN:LSR, PRE:SLR, POST:LRS
- insert, search, findMax, findMin: O(h)
- successor O(h):
  - if hasRightChild, smallest node in right sub-tree
  - else, first parent node that is left child
- delete O(h): switch numChild
  - 0: remove v
  - 1: remove v, connect child(v) to parent(v)
  - 2: swap with successor(v), remove(v)

### 7.2 AVL Trees (height $h = \log n$ )

- Property: Every node is height-balanced
- |v.left.height v.right.height| < 1



- insert  $O(\log n)$ :
  - insert key in BST

- walk up, perform max 2 rotations if out- 7.3.3 Orthogonal Range Searching of-balance
- delete(v): (log n rotations)
  - If v has 2 children, swap with successor
  - delete v, and reconnect children
  - for every ancestor of deleted node \* rotate if out-of-balance
- Splay Trees: Rotate nodes that are accessed to root. consider using where operations are non-random.

### 7.3 Augmented Trees

### 7.3.1 Rank Tree (Order Statistics)

- store weight of tree in each node:
- w(v) = w(v.left) + w(v.right) + 1
- select(k)  $O(\log n)$ : finds node with rank k

```
rank = left.weight + 1;
 if (k == rank)
   return v
 else if (k < rank)
   return left.select(k)
 else return right.select(k-rank)
```

• rank(v)  $O(\log n)$ : computes rank of node

```
rank = v.left.weight + 1
 while (v != null) do
   if v is left child do nothing
   if v is right child,
       rank += v.parent.left.weight + 1
   v = v.parent
```

#### 7.3.2 Interval Trees

- Each node is an interval (m, n), m < n
- Sort by m, augment node with maximum n of children in each node
- search(x)  $O(\log n)$ :

```
if x in c
 return c
else if c has no left child
 search in right subtree
else if x > max endpoint in c.left
 search in right subtree
else search in left subtree
```

• findAll(x)  $O(k \log n)$  for k overlapping intervals

```
search(x)
store it somewhere else
remove interval
repeat until no intervals found
```

#### 1. 1D

- (a) use a binary tree search tree
- (b) store all points in the leaves of the tree, internal nodes store only copies
- (c) each internal node v stores the max of any leaf in the left subtree
- (d) Query Time:  $O(k + \log n)$
- (e) Building Tree:  $O(n \log n)$

#### 2. k-dim Tree

- (a) each node in the x-tree has a set of points in its subtree
- (b) store the y-tree at each x-node containing all points
- (c) Query Time:  $O(k + \log^d n)$
- (d) Building Tree:  $O(n \log^{d-1} n)$
- (e) Space:  $O(n \log^{d-1} n)$

#### Hash Tables

- n: #items, m: #buckets
- Simple Uniform Hashing: Keys are equally likely to map to every bucket, and are mapped independently
  - $-load(ht) = \frac{n}{m}$
- $-E_{\text{search}} = 1 + \frac{n}{m}$
- Assume  $m = \tilde{\Omega}(n)$ ,  $E_{\text{search}} = O(1)$

### 8.1 Hash Functions

#### 8.1.1 Division

•  $h(k) = k \mod m$ , choose m prime

### 8.1.2 Multiplication

- fix table size:  $m=2^r$ , for some r
- fix word size: w, size of key in bits
- fix odd constant A,  $A > 2^{w-1}$
- $h(k) = (Ak) \mod 2^w >> (w r)$

### 8.1.3 Rolling Hash

• When key changes by single character

### 8.2 Chaining

- bucket stores linked list, containing (object, value)
- Worst insert O(1 + cost(h))
- Expected search =  $1 + \frac{n}{m} = O(1)$
- Worst search O(n)

### 8.3 Open Addressing

- One item per slot, probe sequence of buckets until find only one
- $h(key, i): U \mapsto 1..m$ , i is no. of collisions
- search: keep probing until empty bucket, or exhausted entire table
- delete: set key to tombstone value, so probe sequence still works
- insert: on deleted cell, overwrite, else find next available slot
- good hash function:
  - 1. h(key, i) enumerates all possible buckets
  - 2. Simple Uniform Hashing
- Linear: h(k, i) = h(k) + i, Clustering
- Double:  $h(k, i) = f(k) + i \cdot g(k) \mod m$
- Insert, Search:  $\frac{1}{1-\alpha}$  where  $\alpha = \frac{n}{m} \le 1$
- good: saves space, rare mem alloc, better cache perf
- bad: sensitive to hash, load

#### 8.4 Cuckoo Hashing

- Resolving hash collisions with worst-case constant lookup time
- Lookup: inspection of just two locations in the hash table
- Insertion: Insert in one of two cells if empty; else kick out other keys to second location.
- If infinite loop, hash function is rebuilt in place

### 8.5 Table resizing

- Scan old table  $O(m_1)$ , create new table  $O(m_2)$ , insert each element O(1), total  $O(m_1 + m_2 + n)$
- O(n) amor: if n == m, m = 2m, if  $n < \frac{m}{4}$

### 8.6 Fingerprint Hash Table (FHT)

- Vector of 0/1 bits
- no false negatives, but has false positives.  $P_{\text{no FP}} = \left(\frac{1}{e}\right)^{n/m}$

#### 8.7 Bloom Filter

• use n hash functions. More space per item. but require n collisions for false positive.

- $P_{\text{coll}} = (1 e^{-kn/m})^k$
- Two hash functions, h(k) and t(k), two tables  $T_1$  and  $T_2$
- insert:  $T_1[h(k)] = 1$ ,  $T_2[h(k)] = 1$
- search: if  $T_1[h(k)]$  and  $T_2[h(k)]$  both 1 return true

### Graphs

Type	Space	v,w	any	all
List Mat	$O(V+E)$ $O(V^2)$	slow fast	fast slow	

#### 9.1 Simple search

• BFS/DFS do not explore all paths

### **9.1.1 BFS** O(V + E)

bfs(root)

Q.enqueue(root)

```
while Q is not empty:
  current = Q.dequeue()
  visit(current)
  for each node n adj to current
    if n not visited
     n.parent = current
      Q.enqueue(n)
```

### **9.1.2 DFS** O(V + E)

• Same as BFS, but use stack instead of queue

### 9.1.3 Topological Sort (DAG)

- Post-order DFS
- Kahn's Algorithm (first append all nodes with no incoming edges to result set, remove edges connected to these nodes and repeat, also O(V+E))

#### 9.2 SSSP

### 9.2.1 Bellman-Ford O(EV)

do V number of times for (Edge e : graph) relax(e)

- can terminate early if no improvement
- can detect negative cycle: perform V times, then perform once more, if have changes it has negative cycle
- if all weights are the same, use BFS

#### 9.2.2 Dijkstra $O(E \log V)$

- Doesn't work with negative edge weights
- can terminate once end is found

```
add start to PQ
dist[i] = INF for all i
dist[start] = 0
while PQ not empty
 w = pq.dequeue()
 for each edge e connected to w
   if edge is improvement
     update pq[w] O(logn)
     update dist[w]
```

#### 9.2.3 DAG

• Toposort, relax in order

### 9.3 Heap

- implements priority queue, is a complete binary tree
- priority of parent > priority of child
- insert: create new leaf, bubbleUp
- decreaseKev: update priority, bubbleDown
- delete: swap with leaf, delete, and then bubble
- store in array:
  - -left(x) = 2x + 1
  - right(x) = 2x + 2
  - parent(x) = |(x-1)/2|

### 9.3.1 Heap Sort

- 1. Heapify (insert n items) O(n log n)
- 2. Extract from heap n times  $(O(n \log n))$
- 3. Improvement: recursively join 2 heaps and bubble root down (base case single node) O(n)
- 4. O(n log n) worst case, deterministic, in-place

### 9.3.2 UFDS (weighted)

- union(p,q)  $O(\log n)$ 
  - find parent of p and q
  - make root of smaller tree root of larger tree
- find(k)  $O(\log n)$ 
  - search up the tree, return the root
  - (PC): update all traversed nodes parent to root
- WU with PC, union and find  $O(\alpha(m,n))$

#### 9.4 MST

• acyclic subset of edges that connects all nodes, and has minimum weight

### 9.4.1 Properties

- 1. Cutting edge in MST results in 2 MSTs
- 2. Cycle Poperty:  $\forall$  cycle, max weight edge is not in MST
- 3. Cut Property:  $\forall$  partitions, min weight edge across cut is in MST

#### 9.4.2 Prim's $O(E \log V)$

```
T = \{start\}
enqueue start's edges in PQ
while PQ not empty
  e = PQ.dequeue()
  if (vertex v linked with e not in T)
    T = T U \{v, e\}
  else
    ignore edge
MST = T
```

### 9.4.3 Kruskal's $O(E \log V)$

- Use union find
- connect two nodes if they are in the same blue tree

```
Sort E edges by increasing weight
T = \{\}
for (i = 0; i < edgeList.length; i++)</pre>
  if adding e = edgelist[i] does
  not form a cycle
    add e to T
    else ignore e
MST = T
```

### 9.4.4 Boruvka's $O(E \log V)$

```
T = { one-vertex trees }
While T has more than one component:
 For each component C of T:
   Begin with an empty set of edges S
   For each vertex v in C:
     Find the cheapest edge from v
     to a vertex outside of C, and
     add it to S
   Add the cheapest edge in S to T
 Combine trees connected by edges
```

#### 9.4.5 Variants

MST = T

- 1. Same weight: BFS/DFS O(E)
- 2. Edges have weight 1..k:
  - Kruskal's

- Bucket sort Edges O(E)
- Union/check  $O(\alpha(V))$
- Total cost:  $O(\alpha(V)E)$
- Prim's
  - Use array of size k as PQ, each slot holds linked list of nodes
- insert/remove nodes O(V)
- decreaseKey O(E)
- 3. Directed MST
  - $\forall$  node except root, add minimum incoming edge O(E)
- 4. MaxST
  - negate all weights, run MST algo

#### 9.4.6 MST Problems

- 1. How do I add an edge (A.B) of weight k into graph G and find MST quickly?
  - Use cycle property; max edge in any cycle is not in MST
  - only add (A,B) if k is not the max weight edge
  - $\bullet$  O(V + E) time to find max edge along  $A \to B$  with DFS
- 2. Given an undirected graph with Kpower plants, find the minimum cost to connect all other sites.
  - run Prim's, use super source
  - weight of new edges are zero
  - this is a single MST
- 3. How do I make Kruskal run faster when sorting?
  - Store edges in separate linked lists
  - To process edges in increasing weight, process all edges in one linked list then the next
  - Time: O(E) or  $O(E\alpha(m,n))$
  - Space: O(E), need to store all E edges
- 4. Minimum Bottleneck Spanning Tree (MBST)
  - General idea: If I use some edge e that is not in the MST to replace some edge e' in the MST, then my max. edge is max (max edge on original MST, e).

- Intuitively, my MST would then fulfill the condition of MBST.
- Note: Every MST is an MBST, but not every MBST is an MST
- 5. Find maximum distance between 2 vertices in MST
  - Bruteforce: perform DFS starting from every single location since there is only one path from any node to another
  - DFS: O(V + E), doing it V times,  $O(V(V+E)) = O(V^2)$  since E =V-1
  - Space: O(V), need to store all the edges in MST

### 9.5 Floyd-Warshall (APSP)

- Shortest paths have optimal substructure
- Shortest paths have overlapping subprob-
- Idea: gradually allow usage of intermediate vertices
- Invariant: At step k, shortest path via nodes 0 to k are correct.

```
// precondition: A[i][j] contains weight
// of edge (i,j) or inf if no edge
int[][] APSP(A) {
 // len = # vertices
 // clone A into S
 for(int k = 0; k < len; k++)
     for(int i = 0; i < len; i++)
          for(int j = 0; j < len; j++)
             S[i][i] =
                  Math.min(S[i][j],
                           S[i][k] + S[k][j]);
 return S;
```

### 9.6 Network Flow

**k-edge connected** Source and target are k-edge connected if there are k edge disjoint paths(don't share edges) from source to target.

Max flow st-cut property with minimum capacity(outgoing from s, ignore incoming to s)

Min cut Let S be the nodes reachable from the source in the residual graph. T=all other nodes,  $S \to T$  is minimum cut **Augmenting Path** path in residual graph from s to t that has no 0 weight edges

#### 9.6.1 Ford-Fulkerson

- 1. Start with 0 flow
- 2. While there exists augmenting path:
  - find an augmenting path
  - compute bottleneck (min edge)
  - increase flow on the path by bottleneck capacity

Time Complexity:

- DFS: O(|F|E)
- BFS(Edmonds-Karp, shortest augmenting path):  $O(VE^2)$
- Dinitz:  $O(V^2E)$

# 9.7 Graph Algorithms on Trees9.7.1 Min Vertex Cover

- Idea: transform tree into DAG, run DP
- only two possiblities for each vertex; taken or not

### 9.7.2 SSSP

- On a weighted tree, any graph traversal algorithm (eg. DFS, BFS) can obtain the shortest path to any vertice in O(V)
- Weight of shortest path between two vertices is the sum of the weights of edges on the unique path

#### 9.7.3 ASSP

• Run SSSP on V vertices in total  $O(V^2)$ , compared to  $O(V^3)$  FW algorithm

#### 9.7.4 Diameter

- Originally, run FW in  $O(V^3)$  and do an  $O(V^2)$  all-pairs check, to total  $O(V^2)$ .
- Now, only need 2 O(V) traversals: DF-S/BFS from any vertex s to find the furthest vertex x. Then do a DFS/BFS one more time from vertex x to find furthest vertex y. Length of unique path along x to y is the diameter of the tree.

### 10 Parallel Algorithms

#### 10.1 Parallel Fibonacci

```
parallelFib(n) {
   if(n < 2) then
   return n;
   x = spawn parallelFib(n - 1);
   y = spawn parallelFib(n - 2);
   sync;
   return x + y;
}</pre>
```

- Critical Path:  $T_{\infty}$ , Parallelism =  $T_1/T_{\infty}$
- $T_{\infty}(n) = max(T_{\infty}(n-1), T_{\infty}(n-2)) + O(1) = O(n)$
- $T_p > T_1/p$
- $T_p > T_{\infty}$ , cannot run slower on more processors
- Goal:  $T_p = (T_1/p) + T_{\infty}$ ,  $T_1/p$  is the parallel part,  $T_{\infty}$  is the sequential part

#### 10.2 Matrix Addition

Before: • Work analysis:  $T_1(n) = O(n^2)$ 

• critical path analysis:  $T_{\infty}(n) = O(n^2)$  After:

```
pMatAdd(A,B,C,i,j,n)
  if(n == 1)
    C[i,j] = A[i,j] + B[i,j];
  else:
    spawn pMatAdd(A,B,C,i,j,n/2);
    spawn pMatAdd(A,B,C,i,j + n/2,n/2);
    spawn pMatAdd(A,B,C,i + n/2,j,n/2);
    spawn pMatAdd(A,B,C,i + n/2,j + n/2,n/2);
    spawn pMatAdd(A,B,C,i + n/2,j + n/2,n/2);
    sync;
```

• Work Analysis:  $T_1(n) = 4T_1(n/2) + O(1) = O(n^2)$ 

```
• Critical Path Analysis:
  T_{\infty}(n/2) + O(1) = O(\log n)
10.3 Parallelized Merge Sort
pMerge(A[1..k], B[1..m], C[1..n])
  if (m > k) then pMerge(B, A, C);
  else if (n==1) then C[1] = A[1];
  else if (k==1) and (m==1) then
    if (A[1] \le B[1]) then
      C[1] = A[1]; C[2] = B[1];
      C[1] = B[1]; C[2] = A[1];
  else
    // binary search for j where
    // B[j] \le A[k/2] \le B[j+1]
    spawn pMerge(A[1..k/2],
                 B[1..i],
                 C[1..k/2+j])
    spawn pMerge(A[k/2+1..1],
                 B[j+1..m],
                 C[k/2+j+1..n])
    synch;
pMergeSort(A, n)
  if (n=1) then return;
    X = \text{spawn pMergeSort}(A[1..n/2], n/2)
    Y = \text{spawn pMergeSort}(A[n/2+1, n], n/2)
    A = \text{spawn pMerge}(X, Y);
   Critical Path Analysis:
• T_{\infty}(n): critical path of parallel merge for
  A.B of length n
• k > n/2 elements in A, n - k elements in
  B, k/2 + (n-k) < 3n/4
```

- $T_{\infty}(n)T_{\infty}(3n/4) + O(\log n)O(\log^2 n)$
- $T_1(n) = T_1(\alpha n) + T_1((1-\alpha)n) + O(\log n) \approx 2T_1(n/2) + O(\log n) = O(n)$
- Parallel Sorting:  $T_{\infty}(n) = T_{\infty}(n/2) + O(\log^2 n) = O(\log^3 n)$