

-) Amortized Cost = $\frac{\text{Total cost of operations}}{\# \text{operations}}$
-) Pointer: a variable that stores the memory address of an object
-) Hard removal: The data removed is completely removed from the backing structure
 - Set a position to null is a hard removal method
-) Soft removal: The data is still inside the backing data structure, but cannot be accessed by user

DATA STRUCTURE

-) List: An ordered, 0-aligned, contiguous collection of data
 - start from 0
 - no gaps
- 2 standard implementation of list : ArrayList, LinkedList
-) ArrayList : CS 1332 lecture note
-) LinkedList : 3 different implementations : Singly-Linked List, Doubly-Linked List, Circularly Linked List
-) Singly-Linked List : CS 1332 lecture
- Methods that require going to the end of the list takes $O(n)$
 - AddBack() \rightarrow tackled using tail pointer
 - RemoveBack() \rightarrow a problem where it's impossible to assign "tail" to 2nd-to-last node
 - \rightarrow cannot go backward &

- Note on `removeLast()`: cannot directly set the data of "tail" to null because the node itself still there, meaning the 2nd-to-last node is not pointing to null
 - Instead, has to traverse the list to the 2nd-to-last node, then set its pointer to point to null

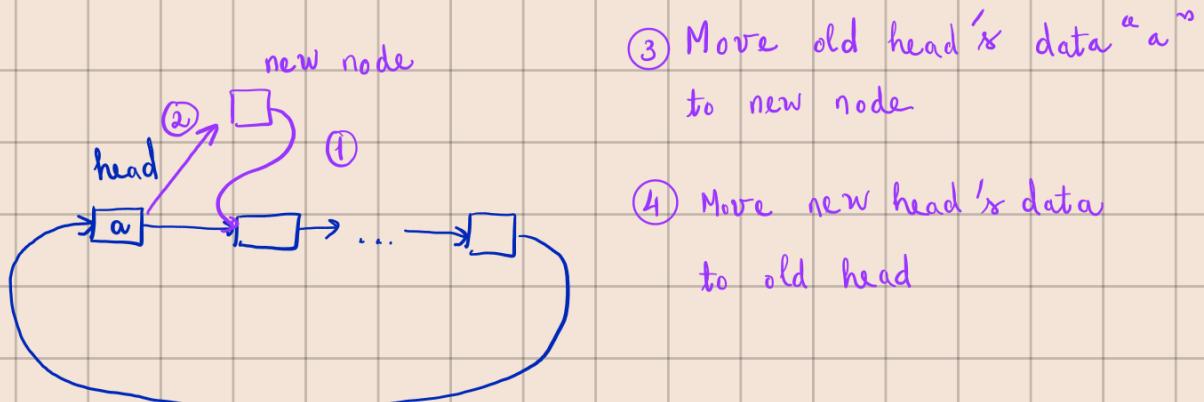
) Doubly Linked List : Can do all methods in $O(1)$ (except

`addAtIndex ($\frac{n}{2}$)`), but cost memory and more complex

) Circular Singly Linked List : like a singly linked list, but the last node points to the 1st node

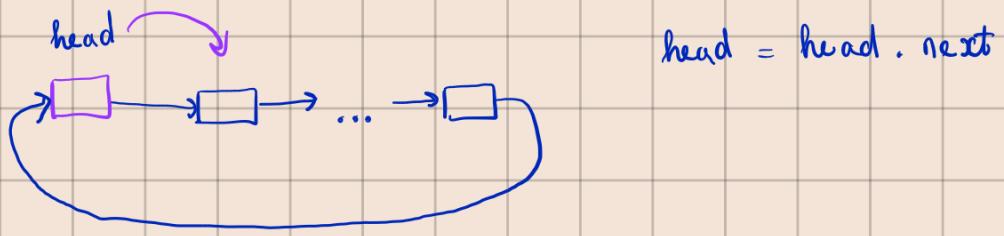
- "tail" is not needed

• Add Front () :

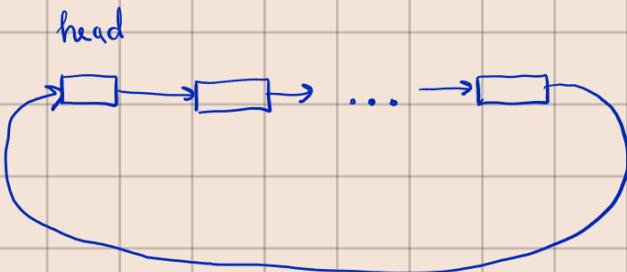


• Add Back () :

- ① `AddFront (data)`
- ② `MoveFrontToBack ()`



• Remove Front () :

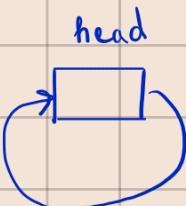


- Remove current head is hard (because of the "tail" pointer)
- Remove and node is easy \rightarrow move the data of and node to head then remove and node

$head.data = head.next.data$

$head.next = head.next.next$

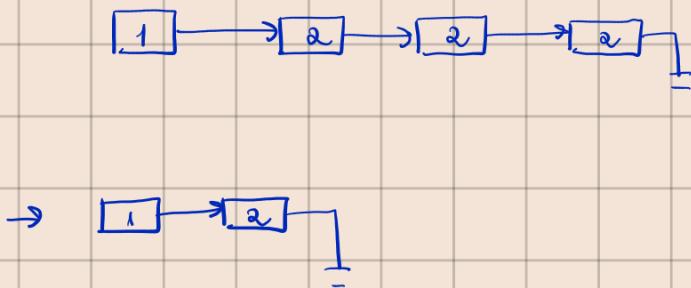
✳ Not worked if CSLL only has 1 node



• Remove Back () : iterate to the 2nd-to-last node

(Recursion)

- Pointer Reinforcement: each node determine whether it remains in the list
Ex: remove duplicates in a linked list

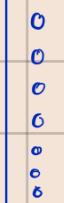


```
1 public void removeDuplicates() {  
2     head = rRemove(head);  
3 }  
4  
5 private Node<T> rRemove(Node<T> curr) {  
6     if (curr == null) {  
7         return null;  
8     }  
9     curr.next = rRemove(curr.next);  
10    if (curr.next != null  
11        && curr.data.equals(curr.next.data)) {  
12        return curr.next;  
13    }  
14    return curr;  
15 }
```

the node that replaces curr in the list
Base case
return the node that replaces curr.next
in the list (recursive step)
condition for curr to be in the list

only remove / add from top

- Stack:



- Data at the bottom stay in the stack the longest (Last In First Out - LIFO)
- push () := addTop ()
- pop () := removeTop ()
- peek () := getTop
- Problem: don't have addIndex () and removeIndex ()

top



) Array Stack :

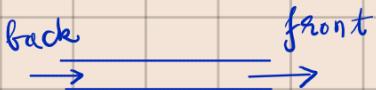
- top : backing ArrayList's back
- push () := add Back ()
- pop () := remove Back ()
- peek () := get Back ()

top



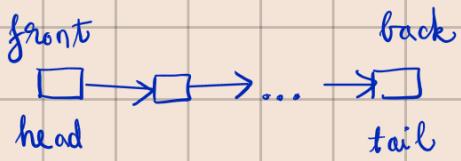
) LinkedStack :

- top : backing LinkedList's head
- push () := add Front ()
- pop () := rem Front ()
- peek () := get Head ()



) Queue :

- Features : $\underbrace{\text{add To Back}(), \text{rem From Front}()}_{\text{enqueue}(), \text{dequeue}()}$

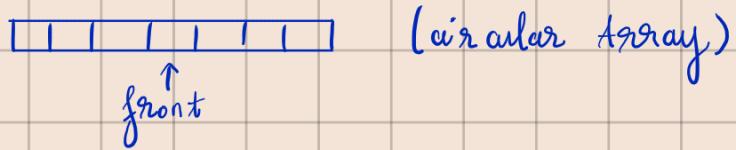


) Linked Queue :

- back : backing LinkedList's tail
- front : backing LinkedList's head

$$\Rightarrow \left\{ \begin{array}{l} \text{enqueue} () := \text{add Back With Tail} () \\ \text{dequeue} () := \text{rem Front} () \end{array} \right.$$

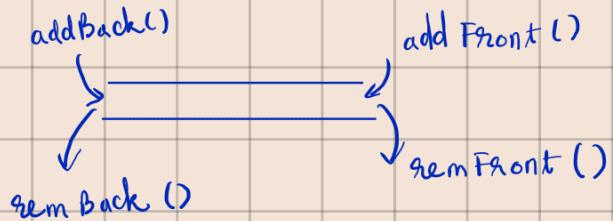
2) Array Queue :



(circular array)

- Backing DS is circular Array, with front index
 - $\text{back} = (\text{front} + \text{size} - 1) \% \text{BackingArr.length}$
- When resize backing Arr, set front back to 0
- enqueue () := addIndex ($(\text{front} + \text{size} - 1) \% \text{BackingArr.length}$)
dequeue () :
 - Set front's data to null
 - $\text{front} := (\text{front} + 1) \% \text{BackingArr.length}$

3) Deque :

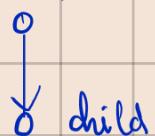


- 4) Linked Deque : Have to deal with $O(n)$ method (removeBack()) if using SLL
→ Use DLL as the backing DS

- 5) Array Deque : Use circular array as the backing DS

- 6) Trees : G is a tree if G is loop-free and has no cycles

- Parent & Child :



- A node can have any # of children
- Exactly 1 node w/ 0 parent

.) leaf : Node that has 0 children

(BT)

.) Binary Tree : Each node has at most 2 children : "left" and "right"

• Full BT : each node has exactly 0 or 2 children

• Complete BT : satisfy 2 conditions :

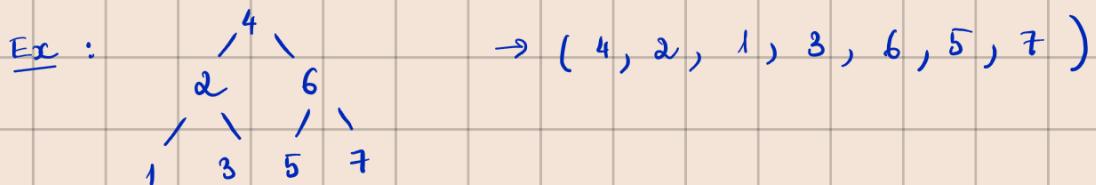
- ① Each depth of the tree except the lowest depth must have maximum # nodes
- ② Lowest depth must be filled from left \rightarrow right

\Rightarrow For each n , there is exactly one shape a complete tree has

• Balanced BT : A node is balanced if its children have height that differ by 0 or 1. A tree is balanced if every node is balanced

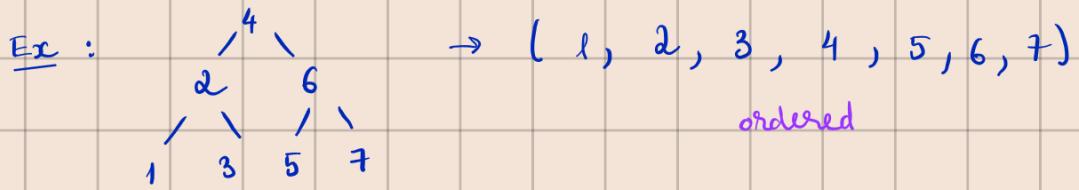
.) Binary Search Tree (BST) : A binary tree where for a node with data x , all the data in the left subtree $< x$ and all the data in the right subtree $> x$.

.) Pre-order traversal : parent \rightarrow left sub-tree \rightarrow right sub-tree



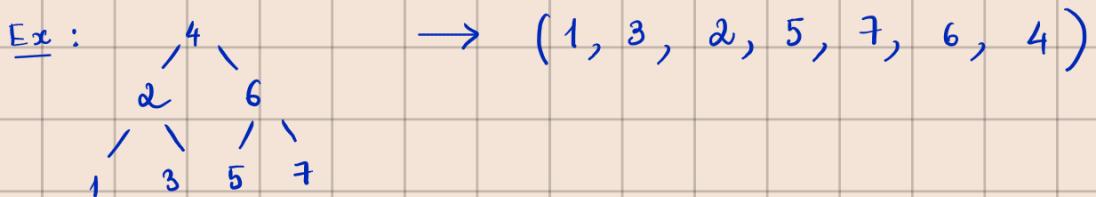
• Pre-order traversal is unique

.) In-order traversal : left sub-tree \rightarrow parent \rightarrow right sub-tree



\rightarrow This outputs ordered list of data

- Post-order traversal : left sub-tree \rightarrow right sub-tree \rightarrow parent



- Tree Operations :

- contains () :

- If $\text{data} > \text{currNode}. \text{data}$: search right

- If $\text{data} < \text{currNode}. \text{data}$: search left

Worst case :



Generally runtime is $O(\log n)$

- add () : Any data can be added to leaf position &

- If $\text{data} > \text{currNode}.data$: go right
- If $\text{data} < \text{currNode}.data$: go left

```

public void Add(T data)
    root = addH(root, data)

private Node addH(Node curr, T data):
    if curr == null:
        Node newNode = new Node(data)
        return newNode
    if curr.data == data:
        //do nothing, data already in tree
    if curr.data < data:
        //recurse right
        curr.right = addH(curr.right, data)
    if curr.data > data:
        //recurse left
        curr.left = addH(curr.left, data)

    return curr

```

*(return node) replaces curr
ptr reinforcement*

replace null w/ new Node

- remove () : There are 3 cases :

- Case 1 : The removed data has no children :

- ptr reinforcement :

if currNode.data == data : return null

- Case 2 : The _____ 1 child :

- ptr reinforcement :

if currNode.data == data : return currNode.nextNode

- Case 3 : The _____ 2 children :

- Find the node containing the data using ptr reinforcement

- Delete the node's data

- Replace with its successor / predecessor

smallest

largest

data > curr.data

data < curr.data

- Remove the S / PS (1 or 0 child this time)

fixin node

replaces curr

```

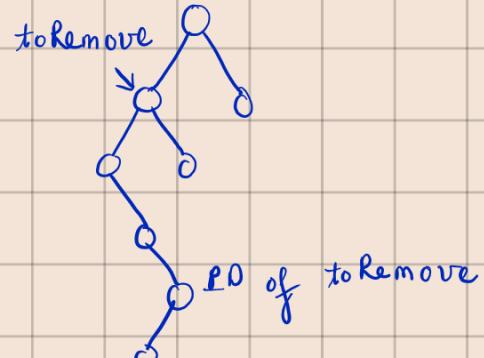
public T remove(T data):
    Node dummy = new Node(null)
    root = removeH(root, data, dummy)
    return dummy.data

private Node removeH(Node curr, T data, Node dummy):
    if curr == null:
        //data not in tree
        throw exception
    if curr.data < data:
        //recurse right
    if curr.data > data:
        //recurse left
    if curr.data == data:
        dummy.data = curr.data
        if curr has 0 kids:
            return null
        if curr has 1 kid:
            return curr.kid
        if curr has 2 kids:
            //get predecessor
            Node dummy2 = new Node(null)
            curr.left = getPredecessor(curr.left, dummy2)
            curr.data = dummy2.data
    return curr

```

as R as possible

1 Left



return the node replacing curr

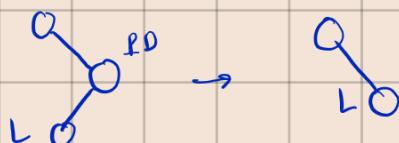
```

private Node getPredecessor(Node curr, Node dummy):
    // find the rightmost descendant
    if curr.right == null: → curr is LD
    //found predecessor
    dummy.data = curr.data
    return curr.left
else:
    curr.right = getPredecessor(curr.right, dummy)
    return curr

```

ptr reinforcement
as well

clever trick to return more
than one value



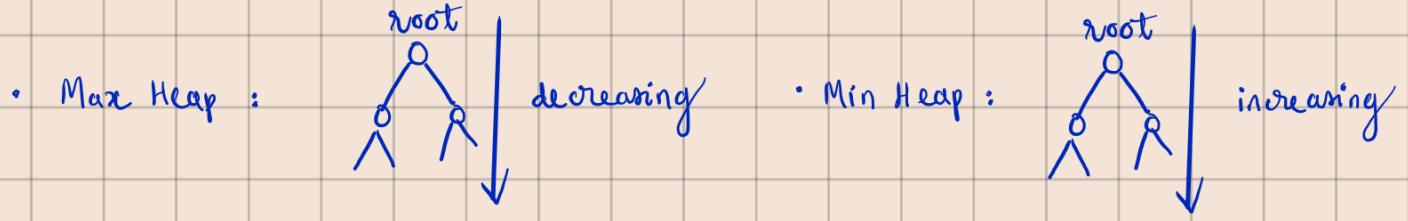
- Base case of getPredecessor(): we want to remove the LD as well,

therefore the parent of the LD will point to the left (L) node instead

- getPredecessor() will reconstruct the path it travels through to get to the LD of toremove.

- Reading recursion: checking the base case $n=0$, then check the case $n=0+1, \dots$, until you get the concept

• Heap : A tree data structure where for every node, the value of its children is greater or smaller than its own value



- Heap is a complete tree
- Order : for a max heap , node with greater value has higher order
min heap , _____ smaller
- parent has higher order than children
- Backing DS : array (because there is no gaps between data)
 - Storing data of a heap into array : use level-order traversal
 - The 0-index element is null

let A be the backing array

- add () :
 - ① A.addBack() → now heap may be out-of-order
 - ② Upheap :
while newNode has higher order than parent
swap (newNode, parent)
if newNode is root :
break
- remove () : remove root (highest order)
 - ① remove A[1]
 - ② move A [size] to index 1 → newNode

③ Down heap :

while newNode has lower order than its children

if only 1 bad child :

swap (newNode, bad child)

if 2 bad children :

swap (newNode, bad child with higher order)

if newNode is leaf (have 0 children) :

break

• buildHeap () :

✓

starting from right \rightarrow left

for each internal node v in reverse order level :

downHeap (v)

Runtime : $O(n)$

• HashMap : A data structure that stores (key, value) pairs

let M be a hashmap

• Rules : keys must be unique and immutable

• put(key, value) : put (key, value) pair into M, if key already existed, replace old value with new input value

• get(key) : return associated value

We want put() and get() to be $O(1)$ \Rightarrow hashCode() and backing array
hash value

• hashCode() : map data of arbitrary size to a fixed-size value

• Two objects are equal \Rightarrow they have the same hash value

• Preference : 2 diff objects have diff hash code

• $index = |\text{object} \cdot \text{hashCode}() \% \text{A.length}|$

- Collisions : put 2 diff keys into the same spot of backing array
 - good hashCode() can reduce this
 - increase size of backing array
- Load factor : a value indicate when to resize, i.e. when

$$\frac{\# \text{ items}}{\text{A.len}} \geq \text{load factor}$$
- Resize : when doing resize, len changes so indices of keys must be recalculated
- Open addressing : 1 item per index \rightarrow Linear / Quadratic Probing
- Closed addressing : ≥ 2 items per index \rightarrow External chaining
 - \downarrow no back ptr
- External chaining : one index hold a linked list
 - put() $\rightarrow O(n)$ (adding to back)
 - resize cost : $O(n)$
- Linear / quadratic probing : keep track of deleted items (DEL mark)
 - put() : stop when seeing size non-removed entry or meeting removed entry with key-value
 - (1) compute id
 - (2) At id :
 - if key already existed : update value
 - if diff keys : probe right ($id + 1$)
 - if DEL : remember if it the 1st one
 - if null :
 - if 1st DEL = null : put key here
 - else : put key into 1st DEL

Runtime : Worst case $O(n)$

- get() : (*)

① Compute id

② At id :

if keys are same : return key's value

if diff keys : probe right

if DEL : probe right

if null : stop, no found

Runtime : Worst case $O(n)$

- resize : everything can collide after resize $\rightarrow O(n^2)$

- Quadratic probing can help reduce runtime by creating gaps, but this causes searching for empty space to be inefficient

- What do we do when we cannot find an empty space? resize (not once)

- Why the max # attempts is $\frac{arr.\text{len}}{2}$? After that the indices will start repeating
resize once may not be enough

⇒ SkipList : 20 linked list → where a node has 4 ptrs : next, previous, above, below

- Rules :

① All data must be in lowest level

② If in level, must be lower levels (no data can fall between 2 empty levels)

- contains() :

if data > cur's data : go right

else if data < _____ : go down

if hit null : not in list

- add() : most left word item is $-\infty$, most rightful item is ∞

① Flip coin

② if head : # layers $\leftarrow 1$ (add on top)

if tail : start adding

③ Adding : if data > curr's data : go right
if data < curr's data :

add before curr

go up

continue until fall out of the list

- remove() : similar to contains(), but after found and remove the element at the current level, go down and repeat until fall out of list

Runtime :

- for add() : random

- for contains() :

- Best case is similar to BST $\rightarrow O(\log n)$

- Worst case : all data is in a single linked list

horizontal (all HEADS)

vertical (all TAILS)

$\Rightarrow O(n)$

- for removes() : $O(n)$

) AVL : self-balancing binary search tree

- Balance : A node is balanced if its children have heights diff by 0 or 1. A tree containing all balanced nodes is balanced

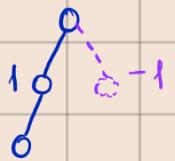
- null node has height -1

BF

- Balance factor := node.left.height - node.right.height

$$\cdot \text{BF} \in \{-2, -1, 0, 1, 2\}$$

- Ex :



root.BF = 2 → lean too much to the left

- Fixing unbalanced subtree : Let A, B, C form a sub-tree, we have 4 cases :

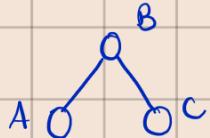
- ① Rotate left :



$$A.BF = -2$$

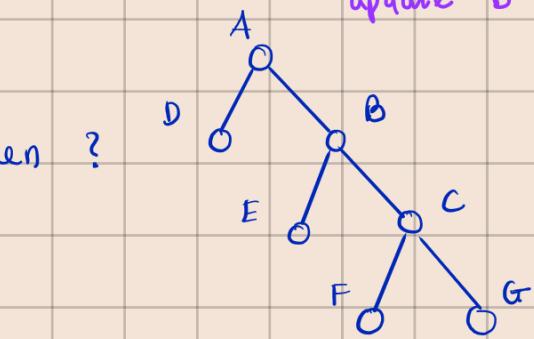
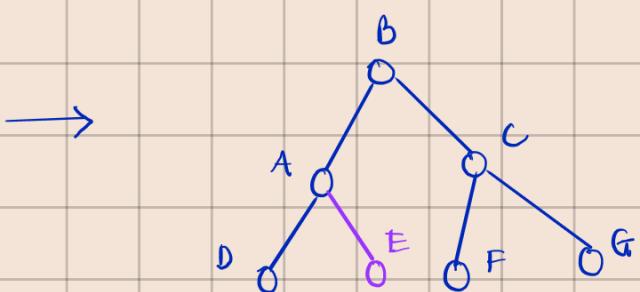
$$A.\text{right}.BF = -1$$

- Rotate left :



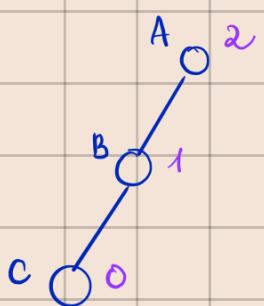
Notice that height and BF of C doesn't change → only need to update B and A

What if B already had 2 children ?



- attach left child to old root or right child
- tip when coding : create temp node point to B to keep track of what we want to return
- what node need to update ? A and B

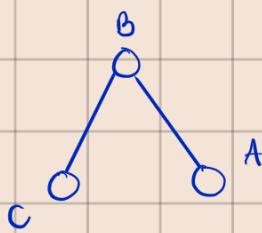
- ② Rotate right :



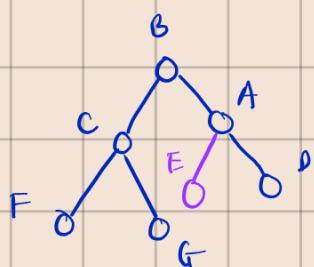
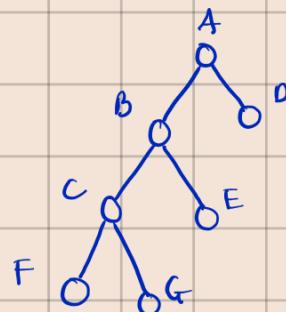
$$A.BF = 2$$

$$A.\text{left}.BF = 1$$

Rotate right :

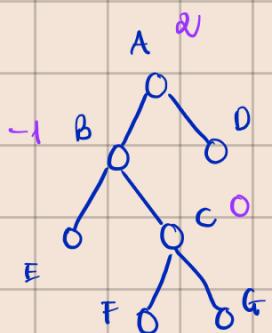


When B has 2 children :



: attach right child to old root
as left child

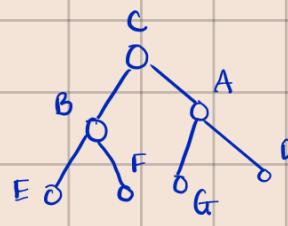
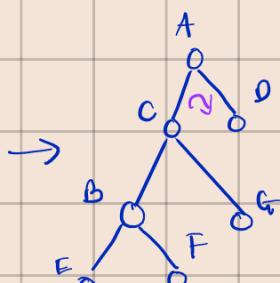
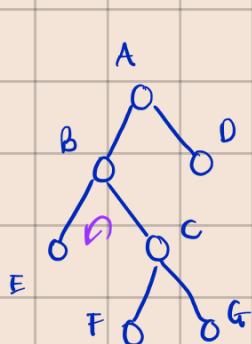
③ Rotate left-right



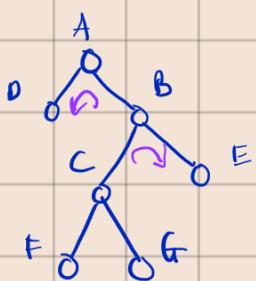
$$A \cdot BF = 2$$

$$A \cdot \text{left} \cdot BF = -1$$

Rotate left-right : rotate left for sub-tree rooted with B, then
rotate right for sub-tree rooted at A



④ Rotate right-left :



Add() :

- ① BST. Add()
- ② Update height (from newly add node \rightarrow root)
- ③ Update BF
- ④ Fix unbalanced subtree

• Since the method to fix unbalance subtree is used together with ptr reinforcement, it is more convenient to design the method to return the rebalanced root Node rebalance (Node parent)

• Where to put it ? When we want to rebalance a subtree rooted at parent where BF of parent = ± 2
 → It makes sure that the children and grandchildren node are balanced
 → Makes sure that parent's BF is always ± 2

Remove() :

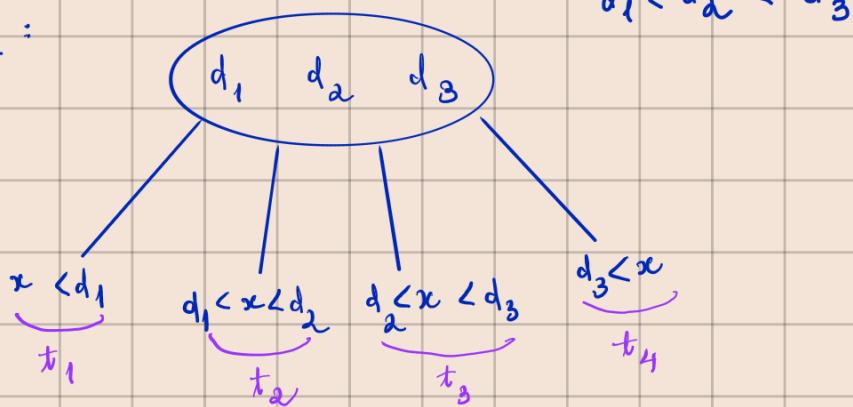
- ① BST. Remove() \rightarrow need to remove the successor / predecessor, so also need to update height / BF here
- ② Update height (from newly add node \rightarrow root)
- ③ Update BF
- ④ Fix unbalanced node

Runtime : all $O(\log n)$

\Rightarrow 2-4 Tree: A tree where each node has 1, 2, or 3 data

- # kids = # data + 1
- Shape property: all leaves are at the same depth

- Order property:



- Contains():

if $data < d_1$: search t_1
else if $d_1 < data < d_2$: search t_2
else if $d_2 < data < d_3$: search t_3
else if $d_3 < data$: search t_4

- Add():

- ① Use contain() to find leaf node
- ② Add to existing node

Case: leaf node is already full, adding to it causes "overflow"
→ still add, then push 2nd / 3rd data to parent → promotion
→ Split leaf node into 2 new nodes

- What if parent is overflow? continuing do promotion for parent node

- Remove():

① Use contain() to find node containing the data

Case 1 : data in leaf node with ≥ 2 data

→ Just remove

Case 2 : data in leaf node with only 1 data

→ look into sibling nodes to see if they have ≥ 2 data

- If yes, then after removing the data in leaf node, move parent node to the now-empty leaf node, and push sibling data to the place left over by parent (technically rotation)

- If no, then we encounter "underflow"

→ pull data from parent down to now-empty node, then merge with sibling → fusion

What if parent is empty? continuing to fusion in parent node

Case 3 : data in internal node and successor / predecessor of data have ≥ 2 data

→ Replace data with successor / predecessor

→ Remove the succ / pred (Guarantee to be in leaf node)

→ The rest is case 1

Case 4 : data in internal node and succ / pred of data is alone in its node

→ Replace data w/ succ / pred

→ Remove the succ / pred the same way in case 2

• Runtime:

• Add() : contain() : $O(\log n)$

overflows : at most $\log n$ overflow $\rightarrow O(\log n)$

$\Rightarrow O(\log n)$

- Remove() : contain() : $O(\log n)$

transfer : at most once

underflow : at most $\log n$ underflows $\rightarrow O(\log n)$

- Contain() : $O(\log n)$
-

ALGORITHM

→) Sorting : there are 3 criteria for sorting algs:

- In-place : Doesn't need additional data structure to sort
 - Adaptability : If data is partial sorted, then runtime improves
 - Stability : Items with the same value will stay in the same relative order after sorting
-
- In-place and Adaptability are easy to detect
 - For Stability : if swap only adjacent elements → probably stable
if _____ further apart → probably not stable

→) Bubble Sort: For each pair of adjacent items, if out of order, swap

- After i-th iteration, the item in index size - i will be correct
- Optimization :
 - keep track of the index of the last item that got swapped → k
→ For each following iteration, stop at k (indices from k → end are correct)
 - If no swaps made during an iteration → finish
- In-place, Adapt, and Stable
- Runtime : $O(n^2)$

→) Insertion Sort: In i-th iteration, put i-th element into the sorted portion of the list

... ? ? ? ? | i-th
... ↓ ↓ ↓ ↓

- At i-th iteration, indices from 0 → i are sorted
- Runtime : $O(n^2)$
- Stable, Adaptive, In-place

- .) Selection Sort : In i -th iteration , put the i -th largest element to correct position
- Actually the one that has the fewest # swaps
 - Not stable, Not adaptive , In-place
- .) Cocktail Shaker Sort : Run bubble sort from left \rightarrow right then right to left
- The next iteration of BB starts at the element right after the last swap position of current BB
 - If no swap in one iteration \rightarrow Finish
 - Stable, Adaptive, In-place
 - Runtime : $O(n^2)$
- .) Best iterative sorts : Insertion Sort (fewest compares)
- .) HeapSort : Add all data to heap , then remove item by item
- Note : Use `BuildHeap()` to add data
 - Not Stable, not adaptive, not In-place
 - Runtime : $O(n \log n)$
- .) MergeSort : Recursively sort left half , and then right - half , and then merge left and right halves
- Note :
 - $\text{left} = \text{data}[0 : \text{size}/2]$ (right-exclusive)
 - $\text{right} = \text{data}[\text{size}/2 : \text{size}]$
 - For merge , create a new temp array to hold the sorted elements , then copy them into the original position
 - Stable , not adaptive , not In-place

) Fried Sort: break data into lists of size \sqrt{n} , then run insertion sort on each list, then merge them into a single sorted list

- Stable, not Adaptive, not In-place
 - Runtime:
 - Insertion sorts: $O(n)$
 - Add 1 element to temp after need to compare with \sqrt{n} others
 $\rightarrow O(n\sqrt{n})$
- $\Rightarrow O(n\sqrt{n})$

) QuickSort: recursively partition the left subarray and right subarray according to a pivot chosen at random

< pivot | pivot | > pivot

• Partition ():

① Swap (pivot, A[start])

② let $i = 1$, $j = \text{size} - 1$

• Find i s.t $A[i] > \text{pivot}$ from index 1 \rightarrow end

• Find j s.t $A[j] < \text{pivot}$ from index end $\rightarrow 1$

• Swap ($A[i]$, $A[j]$)

• Repeat until $j < i$ (j moves to the region containing elements $< \text{pivot}$)

③ Swap (pivot, $A[j]$)

• Not stable, not adaptive, in-place

• Runtime: average $O(n \log n)$

worst $O(n^2)$ (keep choosing pivot which is the min / max element)

) Quick Select: similar to QuickSort, but only need to work on one side
 $\rightarrow O(n)$ runtime

) Radix Sort : Sort integers by 10's places (each place can only hold digit from 0 → 9, except the largest place which may include negative digits, so digit from -9 → 9)

- # iterations : # digits of the longest int = k
- Represent each digit by a Bucket (linked list) containing the integers that has that digit at the current 10's place.
- Runtime : $O(kn)$
- Stable, not adaptive, not in-place

) Pattern matching : given a string text and string pattern, find all occurrences of pattern in text

) Boyer Moore :

- Last Occurrence Table : mapping each char in the pattern to the last index it appears in the pattern, and all other chars got mapped to -1 (Use hashmap)
- Searching algorithm :
 - ① Starting from the end of the pattern, compare char-by-char with the end portion of the text
 - ② If every char matches : add the start index of the end portion of the text to result, then move start forward 1 position
 - ③ Else : Let failtable be the failure table

let k be the index of the char of text that fails

lastId

If $\text{failtable}[\text{getKey}(\text{text}[k])] < k :$
 $\text{start} += \text{pattern.length} - \text{lastId}$

Else : $\text{start} += 1$

) KMP : relies on prefix & suffix of the pattern to determine shifting

- Failure Table : record the length of the longest prefix that is also the suffix of the pattern $[0 \dots i]$ (right inclusive)
- Algorithm for creating FT :

$$\text{FT}[0] = 0$$

let $i :=$ the starting index of prefix

$j :=$ _____ suffix

$i = 0, j = 1$ (the substring is $c_1 c_2$)
i j (at both heads)
↓ ↓

while $j < \text{pat.length}$: length of prefix

If $\text{pat}[i] == \text{pat}[j]$: write $\overbrace{i+1}^{\text{length of prefix}}$ to j
 $i++$, $j++$

Else :

If $i \neq 0$: $i = \text{FT}[i-1]$

If $i = 0$: $\text{FT}[j] = 0$, $j++$

(i, j now at both heads again and
 $\text{pat}[i] \neq \text{pat}[j]$)

i) Searching algorithm: Acts like normal Brute Force, except:

- Comparing go from left \rightarrow right of pattern
- When a mismatch occurs at index k of pattern

$$\text{start} += k - FT[k-1]$$

For next comparing iteration, start at id $FT[k-1]$ of pattern

- If all chars match, treat the next char in text at mismatch and shiftx like normal
- If mismatch right at start: start ++ (shift 1 pos)

ii) Rabin-Karp: Compare the whole pattern with the curr portion of text using hash

• Searching algorithm:

If $\text{hash}(\text{pat}) = \text{hash}(\text{curr})$: (equal hashes doesn't mean equal objects)

\rightarrow compare pat and curr char by char

Else: shift pat by 1, hash of new portion computed using rolling hash

$$\text{pattern} = \overline{c_n \dots c_0}$$

• Coding Tip: $\text{hash}(\text{pat}) = \sum_{i=0}^n c_i \cdot \text{base}^i$

① Have to compute hash of $\overline{c_0 \dots c_l}$ where $l = \text{pat.length}$ manually (can't use rolling hash)

• $\text{hashFirst}()$ return hash of $\overline{c_0 \dots c_l}$ and base^n (don't need to use $\text{Math.exp}()$)

(2) Rolling hash : let $h = \text{hash}(c_i \dots c_j)$
 $\text{hash}(\underline{c_{i+1} \dots c_{j+1}}) = (\text{hash}(\underline{c_i \dots c_j}) - c_i * \text{base}^n) * \text{base} + c_{j+1}$

(3) Stop when $i = \text{end} - l + 1$

) Galil Rule : A rule that can optimize Boyer Moore

Ex : text = aaaa aaa ... a (length n) ($m < n$)
pattern = aaa ... a (length m)
 \Rightarrow Runtime = $O(mn)$ (just like Brute Force)

- We can make pattern become $g g g \dots g$ where $g = a \dots a$ where $g.\text{length} = \text{pat.length} - FT[\text{end}]$
- When we get a complete match between pat and text, treat this like the way KMP does : shift by period ($= \text{pat.length} - FT[\text{end}]$, next char in text seen as mismatch) and start to check at id $\text{pat.length} - \text{period}$ ($= FT[\text{end}]$)

a b a b a b

Ex FT: 0 1 2 3 4 5 period = ab
 0 0 1 2 3 4 period.length = 2

) Depth - First Search (DFS) : An algorithm to traverse the graph, where the visited vertices are marked to avoid later. The order of visits is similar to the process of build up and tear down a stack

- Runtime : $O(|V| + |E|)$

- Space-complex : $O(\log n)$

) Breadth - First Search (BFS) : Traverse the graph layer by layer, where visited are marked to avoid later. The order of visits is similar to the process of build up and tear down a queue. Stop when the queue is empty.

- Runtime : $O(|V| + |E|)$

- Space complex: $O(n/2)$

) Dijkstra : Find the shortest path from a node to all other nodes in a weight graph. Use priority queue to keep track of the shortest path. Vertices returned by priority queue's `remove()` has value of min distance to the starting node and thus marked as visited. Stop when the queue is empty or all vertices are visited.

- Runtime : $O((V+E) \log V)$

) Prim : Find the MST by using a set of visited vertices S containing vertices reached by the chosen edges. Use priority queue to keep track of the shortest edge that incident to the vertices of S (to be added). Stop when all edges visited or the queue is empty

- Runtime: $O(|E| \log |E|)$

) Kruskal : Find the MST by picking the shortest edges possible throughout the graph that don't form a cycle. Use union-find data structure to keep track of cycles.

- Runtime: $\text{union}() / \text{find}() = O(\alpha(V)) \approx O(1)$

inverse Ackermann func
↓

$\Rightarrow O(|E| \log |E|)$

