

Arrays

An array is a set of elements put next to each other in RAM. Each element has an address and we can go to any element in the array in $O(1)$ time if we know the index or address of element. Thanks to that each element is next to each other in memory we can easily go to the next element by adding size of the element to current's element address.

Stacks

pop - remove last added element from stack

push - add element to stack at the end

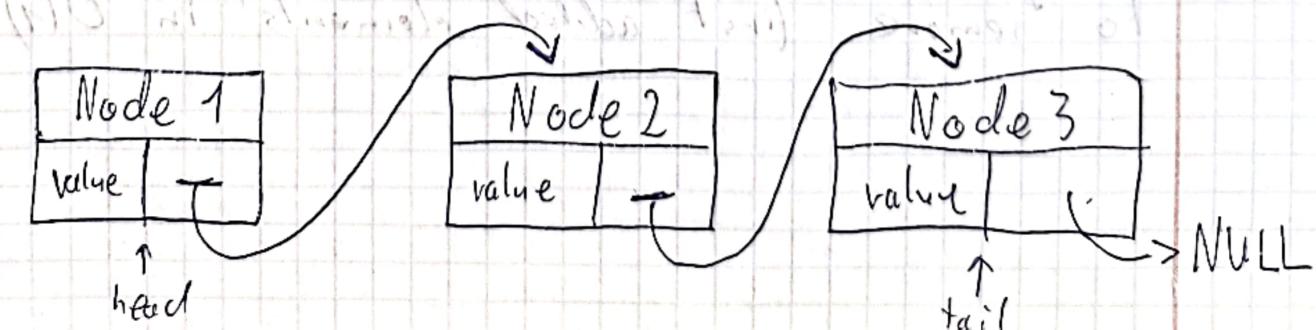
top - return last ^{added} element on the stack

stack is LIFO - Last in, First out data structure

Linked Lists

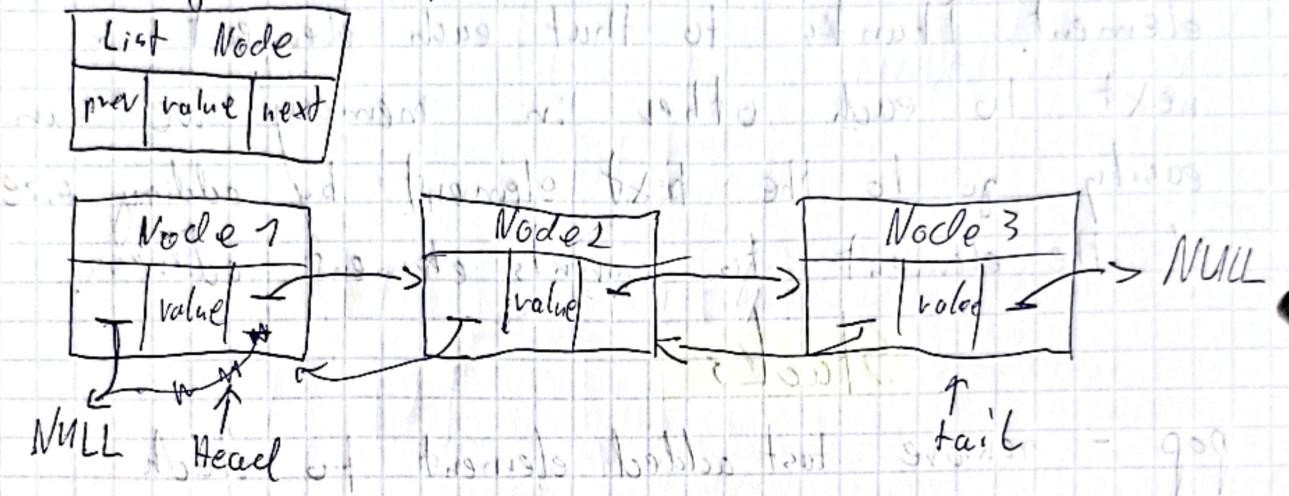
Linked lists consists of list nodes, each node has a value and address to the next element in the list. Last element in the list points to a NULL. First element

is called a head and last is called a tail.
Elements of linked lists don't have to be next to each other in memory.



Doubly linked lists

Doubly linked lists are very similar to singly linked lists, but here each node has 2 pointers, pointing to next and previous elements. Thanks to that accessing elements is easier and traveling both forwards and backwards is possible.



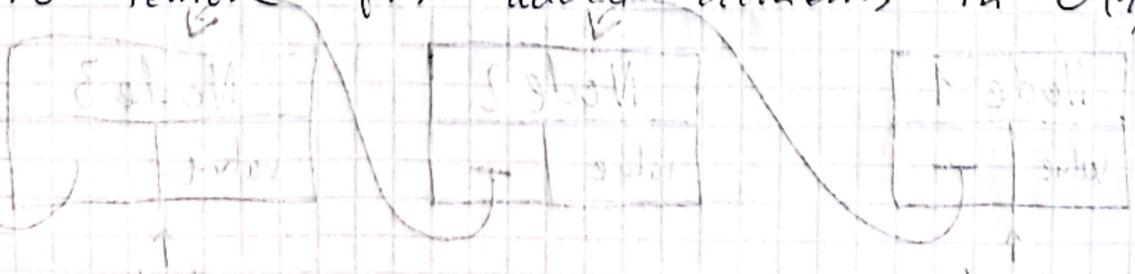
Array Vs Linked List

Operation	Big-O time Array	Big-O time Linked list
Access i-th element	$O(1)$	$O(n)$
Insert/remove end	$O(1)$	$O(1)$
Insert/remove middle	$O(n)$	$O(1)$

Queues

Queues follow **FIFO** - First In, First Out.

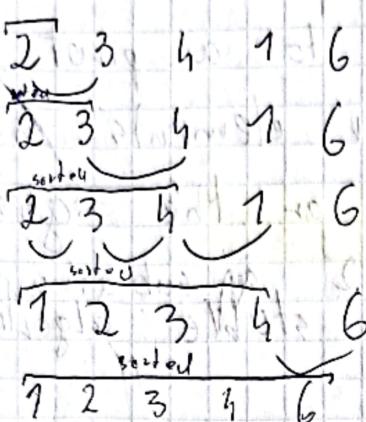
Queues have 2 operations: enqueue - add element, and dequeue - remove element. They are usually implemented using linked lists because they allow to remove first added elements in $O(1)$ time.



Sorting

Insertion Sort $O(n^2)$

sorted



Insertion sort starts with

sorted 1 element array and

go through sort the rest

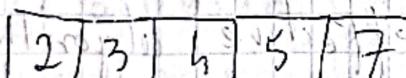
of elements putting them

in right places in sorted

part, by comparing from left.

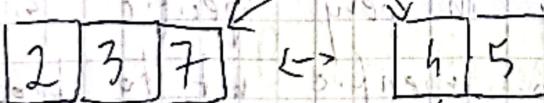
- If there is a "tie" while sorting algorithm which preserve starting order of equal elements is called stable and that one which doesn't guarantee that is called unstable

MergeSort $O(n \log n)$

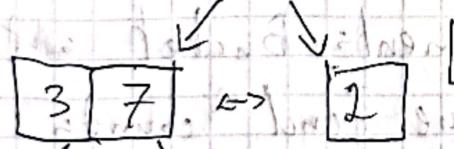


MergeSort uses ~~divide~~ technique called

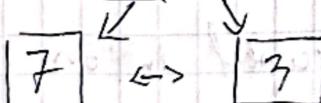
divide and conquer.



We divide our



array to approximately



halves until we have single

elements, then we compare them

and merge. Merge sort is

a recursive algorithm.

Quick Sort

$O(n^2)$, $\Theta(n \log n)$

Quick sort is another recursive algorithm, it also splits the array to 2 parts like mergesort but it does it according to a pivot (usually last element). It puts smaller elements on the left of pivot and bigger on the right and then calls itself on the 2 subarrays. Quick sort is usually not a stable algorithm.

6	2	4	1	(3)
---	---	---	---	-----

2	(1)	3	6	6
---	-----	---	---	---

1	2	4	6
---	---	---	---

After division pivot is always in a right place so there is no need to sort it further.

Bucket Sort

$O(n)$

Bucket Sort is a very effective algorithm but it can be used only in very specific situations where we have given specific not too broad range of values of the elements. Bucket sort creates "buckets" for each value and counts how many times each value occurred in an array and at the end simply overwrites the array with values from the bucket. It is not a stable algorithm.

2	1	2	0	1	0	2
---	---	---	---	---	---	---



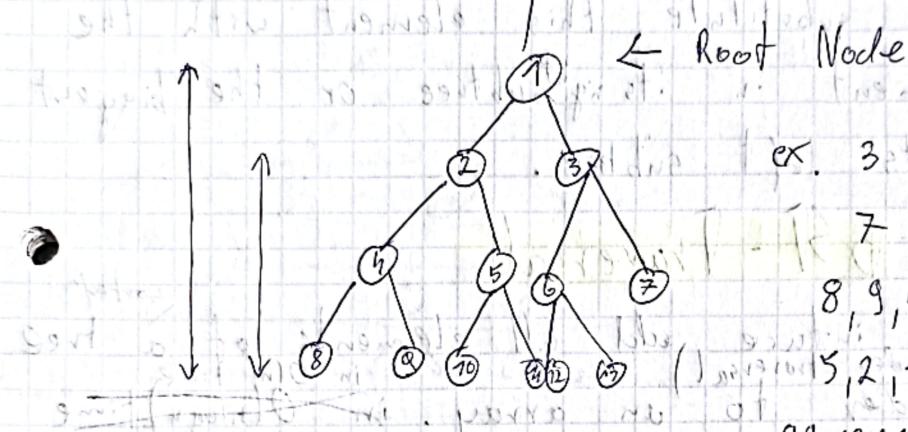
0	0	1	2	2	2
---	---	---	---	---	---

0	1	2
a	a	3

Binary Search $O(\log n)$

- It is a search algorithm which with each iteration narrows our search by half.
To run it the array must be already sorted.

Binary Tree



ex. 3 is a parent of 7 and 6

8, 9, 6 are descendants of 2
5, 2, 1 are ancestors of 10 and 11

8, 9, 10, 11, 12, 13 are leaf nodes

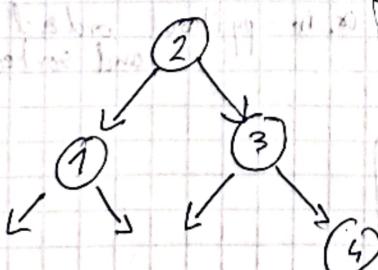
height is numbers of nodes from a node to a leaves

ex. height of 2 is 3, height of 8 is 4

Depth of a node is height calculated from it to a root node.

Binary Search Trees (BST)

They are similar to binary trees but have a sorted property. All smaller elements are on the left of a node and bigger on the right.



Thanks to that searching time is like in sorted array $O(\log n)$ which is actually a height of the tree.

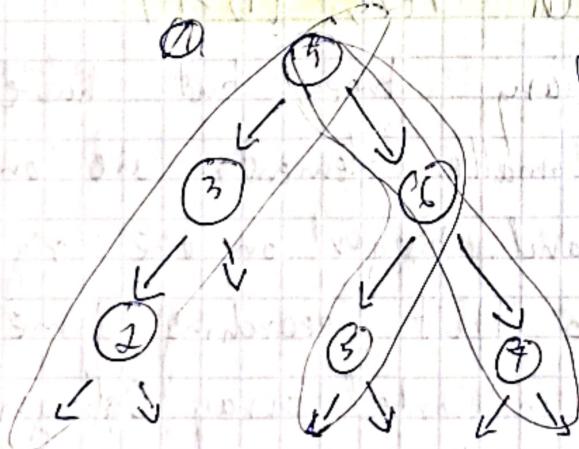
The advantage against the array is the fact that adding and removing elements from BST takes also $O(\log n)$ time compared

We can add elements to BST as a leaves or in the middle. Adding as leaves is easier and more preferable but it usually result in less balanced tree. While removing an element we have 2 cases: ~~AN~~ easy one where removed element doesn't have child nodes or have only 1. The hard one where it has 2 child nodes, here we have to substitute this element with the smallest element in its right subtree or the biggest element in its left subtree.

BST - Traversal

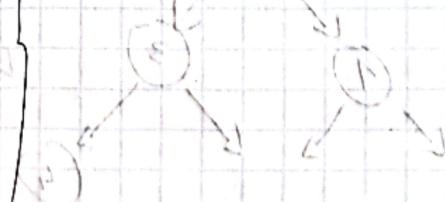
We can (for instance add all elements of a tree in sorted order to an array. in $O(n)$ time) ~~in $O(n \log n)$ time~~

To do it we have to visit all left descendents of the node before visiting its right node. This is an example of Depth-First Search (DFS) where we first go as deep as we can in each node.



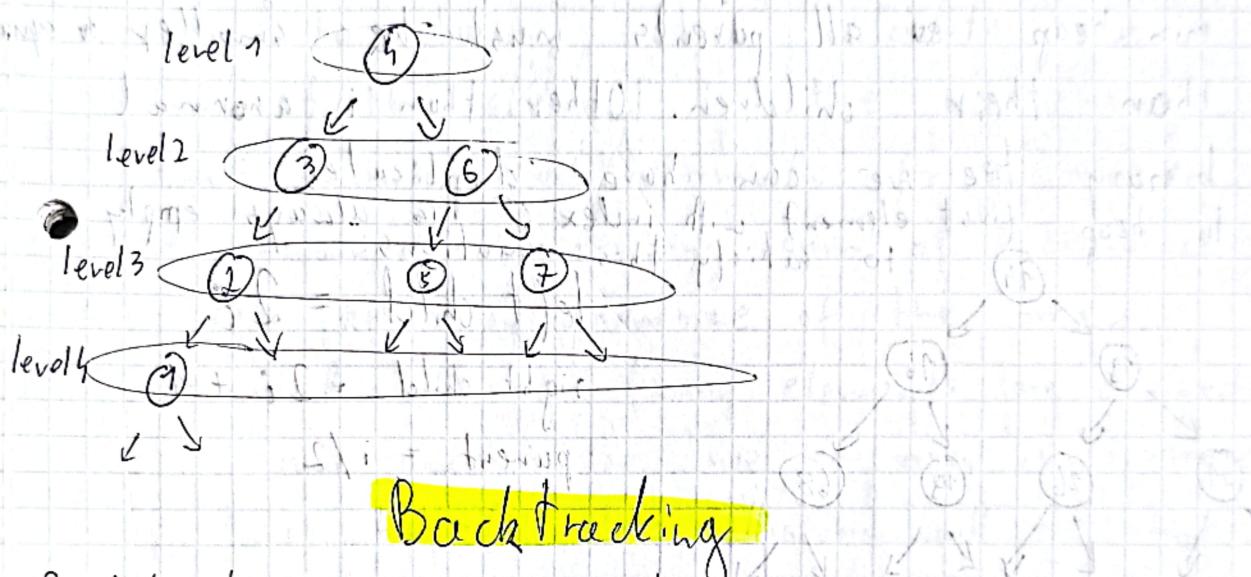
DFS

Binary trees are one of the ways of implementing other DSs like sets and ex. in python ordered-set and sorted-dict



Breadth-First Search BFS

- In BFS in contrary to DFS we firstly visit each level of the tree, BFS uses queues to append each child nodes of a currently process node and go to the next node in queue, because queues are First-in, First-out datastructure.



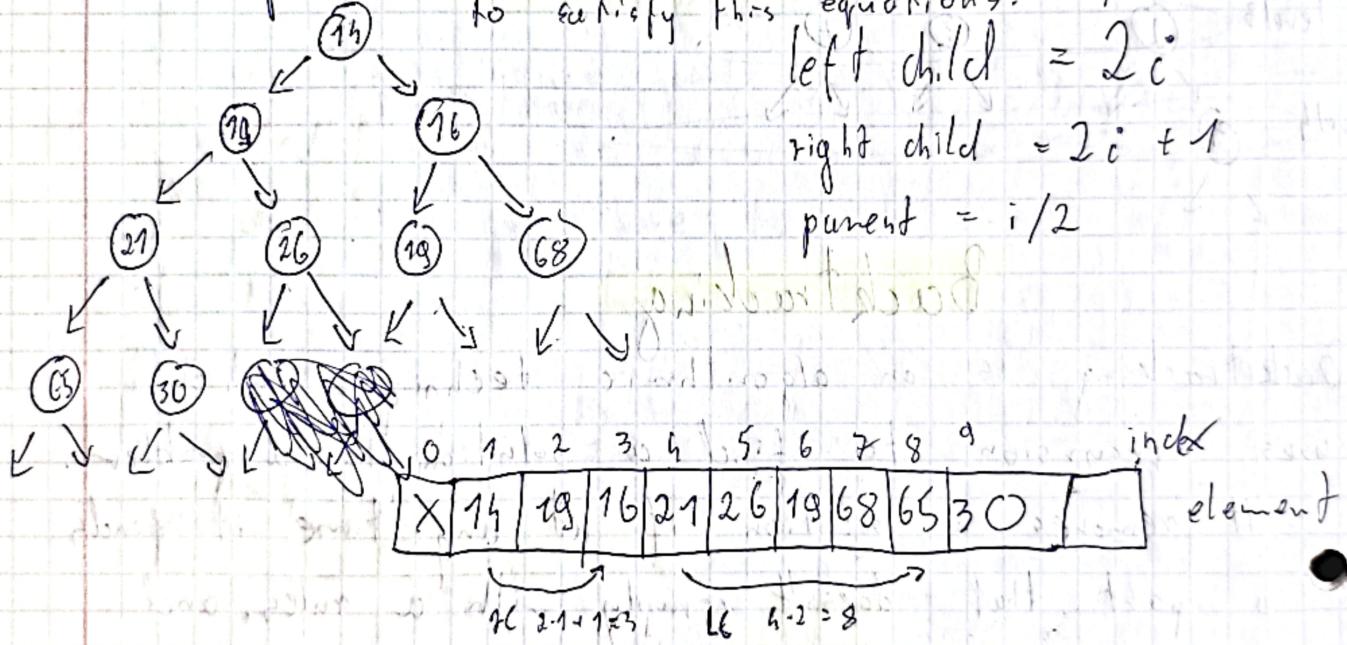
Backtracking

- Backtracking is an algorithmic technique which uses recursion to find a solution to a problem.
- It removes a solution if at any time it finds a part that doesn't comply with a rules, and go back to last moment where it has a choice ~~which hasn't been tried yet~~.

Heap / Priority Queue

Heap is a data structure which is implemented using complete binary tree (doesn't have missing nodes except at leaves) which is really an array underneath. This is called a structure property, there is also an order property. It can be a max or min heap, which are analogous so for min heap all parents must be smaller or equal than their children. Other than in a normal binary tree we can have duplicates.

In heap, first element with index 0 is always empty to satisfy this equations:



To add an element to heap we add it at the end and swap with parents until we have an order property satisfied.

To remove root element we remove root and put last element in its place and then swaps it with its child nodes until order property is satisfied. Time complexity of adding/removing is equal to height of a tree which is $O(\log n)$.

Heapify allows us to create a heap from unsorted array in $O(n)$ time

Hash sets/maps

Hash maps are a very useful datastructures which allow us to insert, delete and search values in a $O(1)$ time. It is so fast thanks to hashing which calculate the index of the element in the list using some mathematical formula for ex. ~~sum~~ the remainder of devision sum of ASCII values of the key by ~~size~~ size of the array.

To avoid collisions (many elements are assigned the same index) the size of array is increased if it is half-full. The collisions will still happen and then we can use arrays in arrays or use open addressing where we assign other index to a key if that index is occupied for eg. the next empty, ~~in~~ the square of it etc.

Graphs

Graphs are like similar to linked lists and trees, but great graphs allow loops. There can be different implementations of graphs; connections of nodes are called edges

Matrix: graph is represented

as 2D array. Uses 0 to represent a free space and 1 a blocked one.

Free spaces (0) represent nodes

and edges are implicit, they are represented by possibility of moving in given direction

1	0	0	1	1
0	1	0	1	0
0	1	0	0	0
0	0	0	1	1
1	1	0	0	1

Adjacency Matrix: usually represented as

square matrix. Here nodes are represented by vertices

Here each element represents

the wage or possibility of moving from node i.e. 3 to node 4

matrix [3][4]

1	2	3	4
0	0	0	0
1	1	0	0
0	0	0	1
0	1	0	0

Adjacency List: here we

store information about neighbours

in a list (of Node type). It is the most popular way to represent a graph because it uses the least amount of space (it don't have representation of nodes that doesn't exist)

```
class Node:  
    def __init__(self, val):  
        self.val = val  
        self.neighbors = []
```

Matrix-DFS

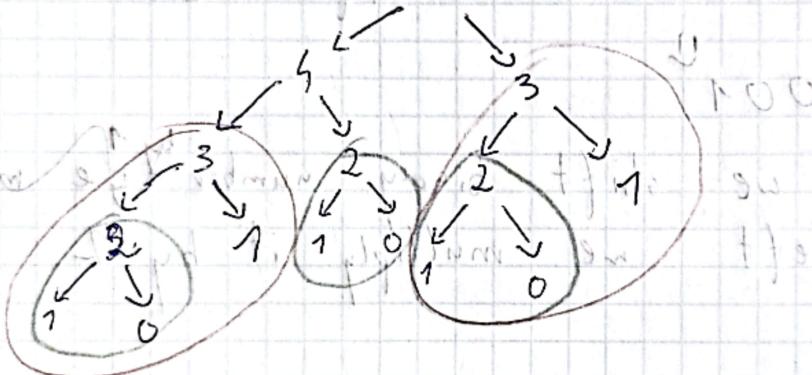
Dynamic Programming

Dynamic programming is about solving complex problems by solving some simpler problems first and then using them to solve starting problem.

Memoization (Top Down DP) we store, in cache (usually hash map) all currently solved problems so we don't have to solve them again.

i.e. calculating Fibonacci sequence

Input of fib(5) = 5 - <010>



Bottom up DP (True dynamic programming) here we start with the base case and work our way up. i.e. Fibonacci sequence

0	1	2	3	4	5
0	1	1	2	3	5

+

But here we really need last 2 elements to calculate the next element so we can save space

0	1	1	2	3	5
2	3	3	5		

Bit Manipulation

AND	
0 & 0	0
0 & 1	0
1 & 0	0
1 & 1	1

OR	
0 0	0
0 1	1
1 0	1
1 1	1

XOR	
0 ^ 0	0
0 ^ 1	1
1 ^ 0	1
1 ^ 1	0

Bit shifting

001 << 1 - shift 1 bit to left

010

010 >> 1 - shift 1 bit to right

001

Basically when we shift binary number by 1 to left we multiply it by 2.