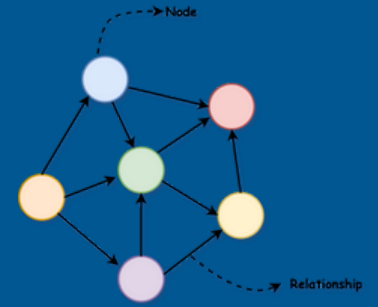


ALGORITHMS JS



DATA STRUCTURES



Eimard Sobrino Zurera
Software Engineer



www.eimardsobrinouzera.com



[linkedin.com/in/eimardsobrinouzera](https://www.linkedin.com/in/eimardsobrinouzera)



<https://gitlab.com/esobrinouz>

TS

Array List

Advantages

- ArrayList optimized for **retrieving items**.
- Simple **creation and usage**
- Foundational **building block** for complex **data structures**

Drawbacks

- **Expensive to manipulate** (insert/delete or resequence values)
- **Inefficient to sort**
- **Fixed size**

Applications

- Basic spreadsheets
- Within complex structures such as hash tables

Time Complexity

Average

Access	Search	Insertion	Deletion
$O(1)$	$O(n)$	$O(n)$	$O(n)$

Worst

Access	Search	Insertion	Deletion
$O(1)$	$O(n)$	$O(n)$	$O(n)$

Space Complexity

Worst $O(n)$

TS

Linked List

Advantages

- LinkedList **optimized for manipulation**
- Inserts and deletes are just changes of the pointer in nodes, **no collapsing or expanding needed**
- **Less complex than restructuring an array**

Drawbacks

- **Expensive to retrieve data**
- **Uses more memory than arrays**
- **Inefficient to traverse the list backward**

Applications

- Best used when data must be added and removed in quick succession from unknown locations
- You can add or remove items from the beginning of the list in constant time. For specific applications, this can be useful

Time Complexity

Average

Access	Search	Insertion	Deletion
0 (n)	0 (n)	0 (1)	0 (1)

Worst

Access	Search	Insertion	Deletion
0 (n)	0 (n)	0 (1)	0 (1)

Space Complexity

Worst 0 (n)

TS

Binary Search Tree

A binary tree is a tree in which each node has up to two children. Not all trees are binary trees. A node is called 'leaf' node if it has no children.

A binary search tree is a binary tree in which every node fits a specific ordering property: ***all left descendents $\leq n < \text{all right descendents}$*** . This must be true for each node n .

The definition can vary slightly with respect to equality.

Considerations

- **Fast for search** | Length is only $O(\text{height})$
- Notion of order
- **Optimized for adding and finding items** because of notion of order ($O(\log n)$)
- **Not optimized for sorted lists** (become $O(n)$)
- Binary search trees **are not as fast as the more complicated hash table**

Applications

- Storing hierarchical data such as a file location
- Binary search trees are excellent for tasks needing searching or ordering of data

Time Complexity

Average

Access	Search	Insertion	Deletion
$O(\log(n))$	$O(\log(n))$	$O(\log(n))$	$O(\log(n))$

Worst

Access	Search	Insertion	Deletion
$O(n)$	$O(n)$	$O(n)$	$O(n)$

Space Complexity

Worst $O(n)$

TS

Binary Heap

A binary heap is very **similar to a binary tree with special rules around implementation**. A binary heap will be as **compact** as possible which means that **all the children of each node are as full as they can be and left children are filled out first**.

We can implement Binary Heap in two forms: **Max Binary Heap** and **Min Binary Heap**.

Considerations *(Max Binary Heap)*

- Every node have at most two children
- Each **parent node's value will be always greater** than its children and there is no guarantee in its children's order.
- **Root** node value is **always** greater than **all** the nodes
- Useful **only for very specific scenarios**
- You **cannot extrac a node other than the maximum**

Applications

- A Binary Heap data structure is a **specialized structure used to save the information** and has very **specific** use cases like **sorting** and **priority queue**.
- **Emergency ward system** with high critical tasks to carry out
- Task scheduling based on priority

Time Complexity

Average

Access	Search	Insertion	Deletion
0 (log(n))	0 (log(n))	0 (log(n))	0 (log(n))

Worst

Access	Search	Insertion	Deletion
0 (log(n))	0 (log(n))	0 (log(n))	0 (log(n))

Space Complexity

Worst 0 (n)



Time complexity to find minimum or maximum of all elements is always constant:

0 (1)

TS

Quicksort | Mergesort | Heapsort

Quicksort

In quicksort we **pick a random element and partition the array**, such that all numbers that are less than the partitioning element come before all elements that are greater than it

Best	Average	Worst	Memory	inplace?	is stable?	remarks
$O(n \log(n))$	$O(2n \log(n))$	$O(n^2 / 2)$	$O(\log(n))$	YES	NO	<ul style="list-style-type: none">$N \log N$ probabilistic guarantee fastest in practice.39% more compares than mergesortFaster than mergesort in practice because of less data movementRandom shuffle guarantee against worst case

Mergesort

Merge sort **divides the array in half, sorts each of those halves**, and then merges them back together. Each of those halves has the same algorithm applied to it.

Best	Average	Worst	Memory	inplace?	is stable?	remarks
$O(n \log(n))$	$O(n \log(n))$	$O(n \log(n))$	$O(n)$	NO	YES	<ul style="list-style-type: none">$N \log N$ guarantee, stableNo in-place sorting, linear extra space

Heapsort

Heapsort **divides its input into a sorted and an unsorted region**, and it iteratively shrinks the unsorted region by **extracting the largest** element from it and **inserting it into the sorted region**.

Best	Average	Worst	Memory	inplace?	is stable?	remarks
$O(n \log(n))$	$O(2n \log(n))$	$O(2n \log(n))$	$O(1)$	YES	NO	<ul style="list-style-type: none">Optimal for both time and space, butInner loop longer than quicksortMakes poor use of cache memoryNot stable

Disjktra

Disjktra algorithm is a way to **find out the shortest path between two points in a weighted directed graph** (which might have cycles). All edges must have positive values.

Disjktra's algorithm finds the minimum weight path from a start node s to every node on the graph.

Implementation and complexity

The **runtime** of this algorithm **depends** heavily **on** the **implementation** of the priority queue. Assume you have v vertices and e nodes:

- Implement priority queue **with array**: Each operation would take $O(V)$. Additionally you would update the values of the paths weight per each edge, so that's $O(V)$. Therefore, the total runtime is $O(V^2)$.
- If you implement the priority queue **with min heap**, then remove calls per each vertex will take $O(v \log(v))$. Additionally on each edge you might call one update/insert, so that's $O(e \log(v))$. Therefore the total runtime is $O((v+e) \log v)$.

Which is better?

If the **graphs has a lot of edges**, then v^2 will be close to e . In this case you might be **better off with the array** implementation, as $O(V^2)$ is better than $O(V + V^2)$.

However, if the **graph is sparse**, then e is much less than v^2 . In this case, the **min heap** implementation **may be better**.

Applications

- **Representing shortest path between two locations**, as GPS system does.
- **Represent cities** and edges representing travel time
- **Task scheduling** based on priority

TS

BFS && DFS | Graph Search

The two most common ways to search a graph are deep-first search and breadth-first search

Breadth-first Search

We start at the root (or arbitrary node) and explore each neighbour before going to any of their children. **We go deep before we go wide.**

- If we want to **find the shortest path** between nodes, **BFS** is generally a **better solution**.
- BFS is **not recursive**. It **uses a queue**.
- Often **used as building block** in other algorithms
- Some applications: **Social Networking** websites, **Peer to Peer** Networks, **Crawlers** in search engines
- Complexity **$O(V + E)$**

Deep-first Search

We start at the root (or arbitrary node) and explore each branch completely before moving on to the next branch. **We go deep first we go deep.**

- BFS tends to be used in different scenarios. **DFS is often preferred if we want to visit every node in the graph.**
- BFS is **not recursive**. It **uses a queue**.
- By itself, DFS is not too **useful** but it is **for** specific tasks such as **count connected components**, determine **connectivity**, or find the bridges/**articulation points**, detects **cycles**, even generate **mazes**.
- Complexity **$O(V + E)$**

TS

RB Tree | AVL Tree

These are two types of self balanced binary search trees.

RB Tree

Red-black trees (a type of self-balancing binary search tree) do **not ensure quite as strict balancing**, but the balancing is still good enough to ensure $O(\log(n))$ insertions, deletions and retrievals.

- It requires a bit less memory
- Can rebalance faster
- Used in situation where the tree will be modified frequently

Properties

- Every node is either **black** or **red**
- Root is **black**
- The leaves, which are NULL nodes, are considered **black**
- Every node must have two **black children**. That is, **red node** cannot have red children (although a **black node** can have **black children**)
- Every path from a node to its leaves must have the same number of **black children**

Time Complexity (For RB Trees and AVL Trees)

Average

Access

$O(\log(n))$

Search

$O(\log(n))$

Insertion

$O(\log(n))$

Deletion

$O(\log(n))$

Worst

Access

$O(\log(n))$

Search

$O(\log(n))$

Insertion

$O(\log(n))$

Deletion

$O(\log(n))$

Space Complexity

Worst

$O(n)$

AVL Tree

An AVL tree is one of the two common ways to implement tree balancing.

- **AVL trees are also Binary Search Trees** (lesser items to left, greater items to the right)
- **used to keep tree as flat as possible**
- **Rigidly balance tree** => **Provide faster look-ups**
- **Insertion/Deletion is not that fast**
- **Only uses Rotations for balancing**

Properties

An AVL tree stores in each node the height of the subtrees rooted at this node. Then, for any node, we can check if it is height balanced: that the height of the left subtree and the height of the right subtree differ by no more than one. This prevents situations where the tree gets too lopsided.

$$\text{balance}(n) = n.\text{left.height} - n.\text{right.height} \\ -1 \leq \text{balance}(n) \leq 1$$