# C++ Data Structures and Algorithms Cheat Sheet

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# 1.0 Data Structures

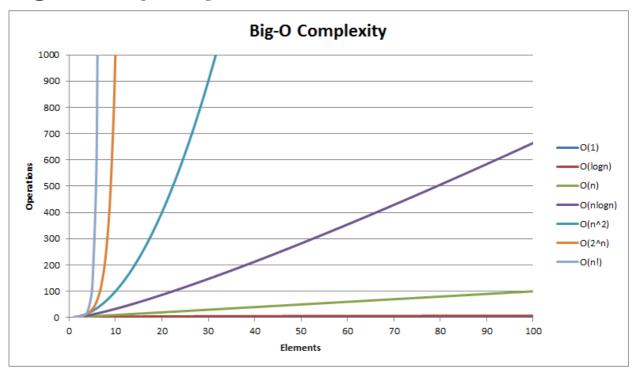
# 1.1 Overview

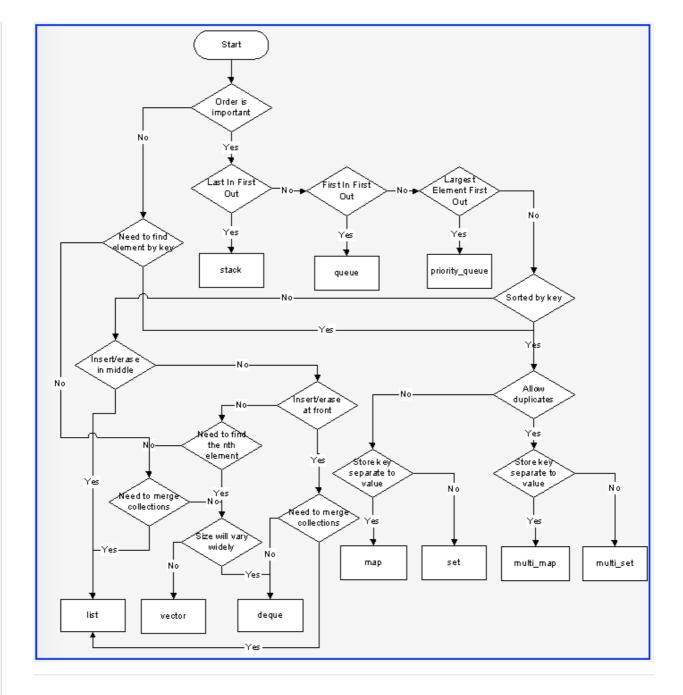


#### **Data Structures**

Data Structure	Time Complexity					Space Complexity			
	Average			Worst			Worst		
	Indexing	Search	Insertion	Deletion	Indexing	Search	Insertion	Deletion	
Basic Array	0(1)	0(n)	-	-	0(1)	0(n)	-	-	0(n)
Dynamic Array	0(1)	0(n)	0(n)	0(n)	0(1)	0(n)	0(n)	0(n)	0(n)
Singly-Linked List	0(n)	0(n)	0(1)	0(1)	0(n)	0(n)	0(1)	0(1)	0(n)
Doubly-Linked List	0(n)	0(n)	0(1)	0(1)	0(n)	0(n)	0(1)	0(1)	0(n)
Skip List	O(log(n))	0(log(n))	0(log(n))	0(log(n))	0(n)	0(n)	0(n)	0(n)	O(n log(n))
Hash Table	-	0(1)	0(1)	0(1)	-	0(n)	0(n)	0(n)	0(n)
Binary Search Tree	O(log(n))	O(log(n))	O(log(n))	O(log(n))	0(n)	0(n)	0(n)	0(n)	0(n)
Cartresian Tree	-	0(log(n))	0(log(n))	0(log(n))	-	0(n)	0(n)	0(n)	0(n)
B-Tree	O(log(n))	0(log(n))	0(log(n))	0(log(n))	O(log(n))	O(log(n))	O(log(n))	O(log(n))	0(n)
Red-Black Tree	O(log(n))	O(log(n))	O(log(n))	O(log(n))	O(log(n))	O(log(n))	0(log(n))	O(log(n))	0(n)
Splay Tree	-	O(log(n))	O(log(n))	O(log(n))	-	0(log(n))	0(log(n))	O(log(n))	0(n)
AVL Tree	0(log(n))	0(log(n))	0(log(n))	0(log(n))	0(log(n))	O(log(n))	O(log(n))	0(log(n))	0(n)

# **Big-O Complexity Chart**





# 1.2 Vector std::vector

#### Use for

- Simple storage
- Adding but not deleting
- Serialization
- Quick lookups by index
- Easy conversion to C-style arrays
- Efficient traversal (contiguous CPU caching)

#### Do not use for

- Insertion/deletion in the middle of the list
- Dynamically changing storage

Non-integer indexing

#### **Time Complexity**

Operation	Time Complexity
Insert Head	0(n)
Insert Index	0(n)
Insert Tail	0(1)
Remove Head	0(n)
Remove Index	0(n)
Remove Tail	0(1)
Find Index	0(1)
Find Object	0(n)

```
std::vector<int> v;
//-----
// General Operations
//-----
// Insert head, index, tail
v.insert(v.begin(), value);  // head
v.insert(v.begin() + index, value);  // index
v.push_back(value);
                                        // tail
// Access head, index, tail
int head = v.front();  // head
int value = v.at(index);  // index
int tail = v.back();  // tail
// Size
unsigned int size = v.size();
// Iterate
for(std::vector<int>::iterator it = v.begin(); it != v.end(); it++) {
    std::cout << *it << std::endl;</pre>
}
// Remove head, index, tail
                               // head
v.erase(v.begin());
v.erase(v.begin() + index);  // index
v.pop_back();
                                // tail
```

```
// Clear
v.clear();
```

# 1.3 Deque std::deque

#### Use for

- Similar purpose of std::vector
- Basically std::vector with efficient push\_front and pop\_front

#### Do not use for

• C-style contiguous storage (not guaranteed)

#### **Notes**

- Pronounced 'deck'
- Stands for Double Ended Queue

```
std::deque<int> d;
//-----
// General Operations
//-----
// Insert head, index, tail
d.push_front(value);
                                    // head
d.insert(d.begin() + index, value);  // index
d.push_back(value);
                                      // tail
// Access head, index, tail
int head = d.front();  // head
int value = d.at(index);  // index
int tail = d.back();  // tail
// Size
unsigned int size = d.size();
// Iterate
for(std::deque<int>::iterator it = d.begin(); it != d.end(); it++) {
   std::cout << *it << std::endl;</pre>
}
// Remove head, index, tail
d.pop_front();
                             // head
d.erase(d.begin() + index);
                             // index
d.pop_back();
                              // tail
```

```
// Clear
d.clear();
```

# 1.4 List std::list and std::forward\_list

#### Use for

- Insertion into the middle/beginning of the list
- Efficient sorting (pointer swap vs. copying)

#### Do not use for

Direct access

## **Time Complexity**

Operation	Time Complexity
Insert Head	0(1)
Insert Index	0(n)
Insert Tail	0(1)
Remove Head	0(1)
Remove Index	O(n)
Remove Tail	0(1)
Find Index	0(n)
Find Object	0(n)

```
std::list<int> 1;

//-----
// General Operations
//-----
// Insert head, index, tail
1.push_front(value);  // head
1.insert(1.begin() + index, value);  // index
1.push_back(value);  // tail
// Access head, index, tail
```

```
// head
int head = 1.front();
                                          // index
int value = std::next(l.begin(), index);
int tail = 1.back();
                                                            // tail
// Size
unsigned int size = 1.size();
// Iterate
for(std::list<int>::iterator it = 1.begin(); it != 1.end(); it++) {
   std::cout << *it << std::endl;</pre>
}
// Remove head, index, tail
1.pop_front();
                             // head
                            // index
1.erase(l.begin() + index);
1.pop_back();
                             // tail
// Clear
1.clear();
//-----
// Container-Specific Operations
//-----
// Splice: Transfer elements from list to list
// splice(iterator pos, list &x)
       splice(iterator pos, list &x, iterator i)
//
       splice(iterator pos, list &x, iterator first, iterator last)
1.splice(1.begin() + index, list2);
// Remove: Remove an element by value
1.remove(value);
// Unique: Remove duplicates
1.unique();
// Merge: Merge two sorted lists
1.merge(list2);
// Sort: Sort the list
1.sort();
// Reverse: Reverse the list order
1.reverse();
```

# 1.5 Map std::map and std::unordered\_map

Use for

- Key-value pairs
- Constant lookups by key

- Searching if key/value exists
- Removing duplicates
- std::map
  - Ordered map
- std::unordered\_map
  - Hash table

#### Do not use for

Sorting

#### **Notes**

- Typically ordered maps ( std::map ) are slower than unordered maps ( std::unordered\_map )
- Maps are typically implemented as binary search trees

# **Time Complexity**

std::map

Operation	Time Complexity
Insert	0(log(n))
Access by Key	O(log(n))
Remove by Key	O(log(n))
Find/Remove Value	O(log(n))

#### std::unordered\_map

Operation	Time Complexity
Insert	0(1)
Access by Key	0(1)
Remove by Key	0(1)
Find/Remove Value	

```
std::map<std::string, std::string> m;
//-----
// General Operations
```

```
//-----
// Insert
m.insert(std::pair<std::string, std::string>("key", "value"));
// Access by key
std::string value = m.at("key");
// Size
unsigned int size = m.size();
// Iterate
for(std::map<std::string, std::string>::iterator it = m.begin(); it != m.end(); i
   std::cout << *it << std::endl;</pre>
}
// Remove by key
m.erase("key");
// Clear
m.clear();
//-----
// Container-Specific Operations
//-----
// Find if an element exists by key
bool exists = (m.find("key") != m.end());
// Count the number of elements with a certain key
unsigned int count = m.count("key");
```

#### 1.6 Set std::set

#### Use for

- Removing duplicates
- Ordered dynamic storage

#### Do not use for

- Simple storage
- Direct access by index

#### **Notes**

• Sets are often implemented with binary search trees

#### **Time Complexity**

Operation	Time Complexity		
Insert	0(log(n))		
Remove	O(log(n))		
Find	O(log(n))		

#### **Example Code**

```
std::set<int> s;
//-----
// General Operations
//-----
// Insert
s.insert(20);
// Size
unsigned int size = s.size();
// Iterate
for(std::set<int>::iterator it = s.begin(); it != s.end(); it++) {
   std::cout << *it << std::endl;</pre>
}
// Remove
s.erase(20);
// Clear
s.clear();
// Container-Specific Operations
//-----
// Find if an element exists
bool exists = (s.find(20) != s.end());
// Count the number of elements with a certain value
unsigned int count = s.count(20);
```

# 1.7 Stack std::stack

#### Use for

- First-In Last-Out operations
- Reversal of elements

## **Time Complexity**

Operation	Time Complexity
Push	0(1)
Рор	0(1)
Тор	0(1)

## **Example Code**

# 1.8 Queue std::queue

#### Use for

- First-In First-Out operations
- Ex: Simple online ordering system (first come first served)
- Ex: Semaphore queue handling
- Ex: CPU scheduling (FCFS)

#### **Notes**

• Often implemented as a std::deque

```
std::queue<int> q;
```

```
//----
// General Operations
//-----
// Insert
q.push(value);

// Access head, tail
int head = q.front();  // head
int tail = q.back();  // tail

// Size
unsigned int size = q.size();

// Remove
q.pop();
```

# 1.9 Priority Queue std::priority\_queue

#### Use for

- First-In First-Out operations where **priority** overrides arrival time
- Ex: CPU scheduling (smallest job first, system/user priority)
- Ex: Medical emergencies (gunshot wound vs. broken arm)

#### **Notes**

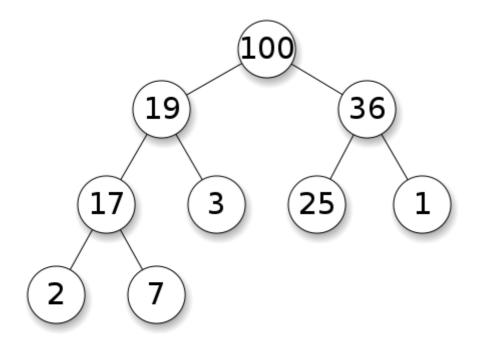
• Often implemented as a std::vector

# 1.10 Heap std::priority\_queue

#### **Notes**

- A heap is essentially an instance of a priority queue
- A min heap is structured with the root node as the smallest and each child subsequently larger than its parent
- A max heap is structured with the root node as the largest and each child subsequently smaller than its parent
- A min heap could be used for Smallest Job First CPU Scheduling
- A max heap could be used for *Priority* CPU Scheduling

#### Max Heap Example (using a binary tree)

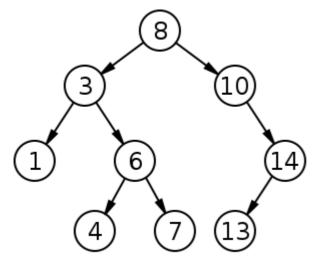


# 2.0 Trees

# 2.1 Binary Tree

- A binary tree is a tree with at most two (2) child nodes per parent
- Binary trees are commonly used for implementing  $O(\log(n))$  operations for ordered maps, sets, heaps, and binary search trees
- Binary trees are **sorted** in that nodes with values greater than their parents are inserted to the **right**, while nodes with values less than their parents are inserted to the **left**

# **Binary Search Tree**



# 2.2 Balanced Trees

- Balanced trees are a special type of tree which maintains its balance to ensure  $O(\log(n))$  operations
- When trees are not balanced the benefit of log(n) operations is lost due to the highly vertical structure
- Examples of balanced trees:
  - AVL Trees
  - Red-Black Trees

# 2.3 Binary Search

#### Idea:

- 1. If current element, return
- 2. If less than current element, look left
- 3. If more than current element, look right
- 4. Repeat

#### **Data Structures:**

- Tree
- Sorted array

#### Space:

0(1)

#### **Best Case:**

0(1)

#### Worst Case:

• 0(log n)

# Average:

• 0(log n)

#### Visualization:



# 2.4 Depth-First Search

#### Idea:

- 1. Start at root node
- 2. Recursively search all adjacent nodes and mark them as searched
- 3. Repeat

#### **Data Structures:**

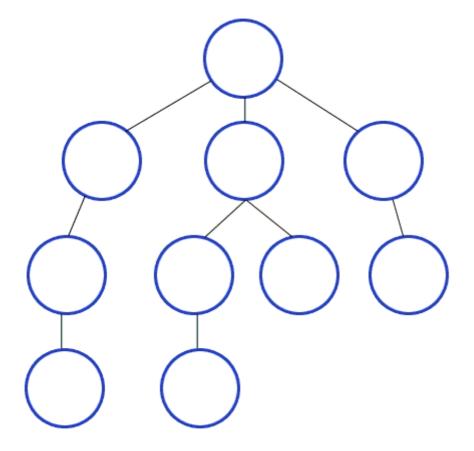
- Tree
- Graph

## Space:

• O(V), V = number of verticies

#### Performance:

• O(E), E = number of edges



# 2.5 Breadth-First Search

#### Idea:

- 1. Start at root node
- 2. Search neighboring nodes first before moving on to next level

#### **Data Structures:**

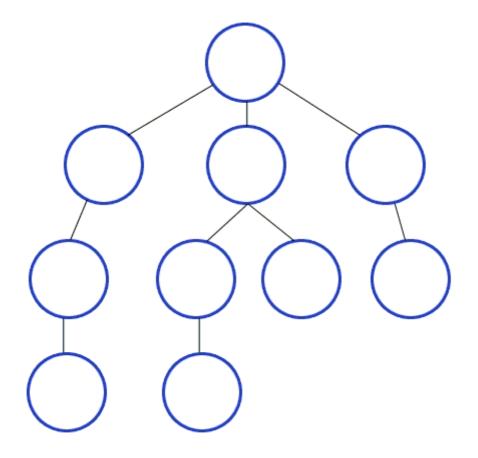
- Tree
- Graph

# Space:

• O(V), V = number of verticies

## Performance:

• O(E), E = number of edges



# 3.0 NP Complete Problems

# 3.1 NP Complete

- NP Complete means that a problem is unable to be solved in polynomial time
- NP Complete problems can be *verified* in polynomial time, but not *solved*

# 3.2 Traveling Salesman Problem

# 3.3 Knapsack Problem

Implementation

# 4.0 Algorithms

# 4.1 Insertion Sort

#### Idea

- 1. Iterate over all elements
- 2. For each element:
  - o Check if element is larger than largest value in sorted array
- 3. If larger: Move on
- 4. If smaller: Move item to correct position in sorted array

#### **Details**

- Data structure: Array
- Space: 0(1)
- Best Case: Already sorted, O(n)
- Worst Case: Reverse sorted, 0(n^2)
- **Average**: 0(n^2)

#### **Advantages**

- Easy to code
- Intuitive
- Better than selection sort and bubble sort for small data sets
- Can sort in-place

#### Disadvantages

• Very inefficient for large datasets

#### Visualization

6 5 3 1 8 7 2 4

#### 4.2 Selection Sort

#### Idea

- 1. Iterate over all elements
- 2. For each element:

o If smallest element of unsorted sublist, swap with left-most unsorted element

#### **Details**

• Data structure: Array

• Space: 0(1)

• Best Case: Already sorted, 0(n^2)

• Worst Case: Reverse sorted, 0(n^2)

• **Average:** 0(n^2)

# **Advantages**

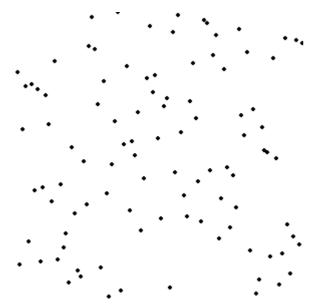
• Simple

• Can sort in-place

• Low memory usage for small datasets

# Disadvantages

• Very inefficient for large datasets



8
5
2
6
9
3
1
4
0
7

## 4.3 Bubble Sort

#### Idea

- 1. Iterate over all elements
- 2. For each element:
  - o Swap with next element if out of order
- 3. Repeat until no swaps needed

#### **Details**

- Data structure: Array
- Space: 0(1)
- Best Case: Already sorted O(n)
- Worst Case: Reverse sorted, 0(n^2)
- **Average**: 0(n^2)

## **Advantages**

• Easy to detect if list is sorted

## Disadvantages

- Very inefficient for large datasets
- Much worse than even insertion sort

# 4.4 Merge Sort

#### Idea

- 1. Divide list into smallest unit (1 element)
- 2. Compare each element with the adjacent list
- 3. Merge the two adjacent lists
- 4. Repeat

#### **Details**

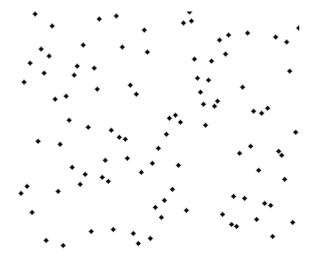
- Data structure: Array
- **Space**: O(n) auxiliary
- Best Case: O(nlog(n))
- Worst Case: Reverse sorted, O(nlog(n))
- Average: O(nlog(n))

#### **Advantages**

- High efficiency on large datasets
- Nearly always O(nlog(n))
- Can be parallelized
- Better space complexity than standard Quicksort

#### Disadvantages

- Still requires O(n) extra space
- Slightly worse than Quicksort in some instances



6 5 3 1 8 7 2 4

# 4.5 Quicksort

#### Idea

- 1. Choose a **pivot** from the array
- 2. Partition: Reorder the array so that all elements with values *less* than the pivot come before the pivot, and all values *greater* than the pivot come after
- 3. Recursively apply the above steps to the sub-arrays

#### **Details**

• Data structure: Array

• **Space**: 0(n)

• Best Case: O(nlog(n))

• Worst Case: All elements equal, 0(n^2)

• Average: O(nlog(n))

#### **Advantages**

- Can be modified to use O(log(n)) space
- Very quick and efficient with large datasets
- Can be parallelized
- Divide and conquer algorithm

## Disadvantages

- Not stable (could swap equal elements)
- Worst case is worse than Merge Sort

# **Optimizations**

- Choice of pivot:
  - o Choose median of the first, middle, and last elements as pivot
  - $\circ \ \ \ \text{Counters worst-case complexity for already-sorted and reverse-sorted}$

