

Abstract Data Types (ADT) List Stack Queue Data Abstraction: Separation of a data type's logical properties from its implementation. -Logical Properties --What are the possible values? What operations will be needed? -Implementation --How can this be done in Java, C++, or any other programming language? ADT is a set of objects together with a set of operations. A data type that does not describe or belong to any specific data, yet allows the specification of organization and manipulation of data.	Stack Implementations Array LinkedList Operations take constant time. Size can grow/shrink easily. Overflow: When element count in an array exceeds array size. Underflow: Pop from an empty stack. Evaluate infix expressions , 2 stacks algorithm (Dijkstra): -- <i>Value</i> : Push onto the value stack. -- <i>Operator</i> : Push onto the operator stack. -- <i>Left parenthesis</i> : Ignore. -- <i>Right parenthesis</i> : Pop operator and two values; push the result of applying that operator to those values onto the operand (value) stack.	Trees (cont) Subtrees Remaining nodes are partitioned into trees themselves, called subtrees . Each subtree is connected by a directed edge from the root . Degree Number of subtrees of a node. Leaf / Terminal Node Node with degree 0 . Parent Child Ancestors Path from node ₁ to node _k Depth Length of the unique path from root to node. Height Length of the longest downward path from the node to a leaf . Height of a Tree Height of the root . For its implementation, a node can hold: -Its first child. -Its next sibling. Thus siblings would be held as a linked list . Without parent/previous sibling information, each node holds only 2 references.	Binary Tree Each node can have at most 2 children. Full BT When each node has 2 or 0 children. Perfect BT It's full and each leaf has the same depth . Tree Traversal Preorder Parent first. Visit root. Traverse left subtree. Traverse right subtree. Postorder Parent last. Traverse left subtree. Traverse right subtree. Visit root. Inorder Left-Parent-Right Traverse left subtree. Visit root. Traverse right subtree
List Operations Find (First occurrence) Insert Remove FindKth MakeEmpty PrintList	Queue - First In Last Out List Operations Enqueue Dequeue MakeEmpty		
List Implementations Simple Array Simple (Singly) Linked List	Queue Implementations Circular Array (Circular Queue) Linked List <i>Examples:</i> -Calls to a call center -Jobs in the printer -Network operations on routers -CPU usage queues		
Stack - LIFO (Last In First Out) List Operations Push Pop Top (Peek) MakeEmpty	Trees Tree Collection of nodes such that: Root Unless empty, trees have a root .		Search Tree ADT - Binary Search Tree Provides inorder traversal. Average case: Depth of all nodes on average log(N) Balanced BST maintains all operations at $h=O(\log N)$ time

AVL (Adelson-Velskii and Landis) Tree	
It's a BST.	
Height of the left subtree and the right subtree differ by at most 1 .	
Empty tree has height -1.	
Balancing After Insertion	
Left-Left	Single Right Rotation
Right--	Single Right Rotation
Right	
Left-	Double Left-Right
Right	Rotation
Right-	Double Right-Left
Left	Rotation
-	
-	
-	
-	
-	
-	
-	
-	
-	
-	

Algorithm Analysis	
<i>Problem Solving: Life Cycle</i>	
Problem Definition	
Functional Requirements	Calculate the mean of n numbers etc. What should the program do?
Nonfunctional Requirements	Performance Requirements: How fast should it run? etc. How should the program do? Can be considered as Quality Attributes
Algorithm Design	

Algorithm Analysis (cont)	
Algorithm	A clearly specified set of instructions for the program to follow.
Knuth's Characterization (5 properties as requirements for an algorithm)	
~Input	0 or more, externally produced quantities
~Output	1 or more quantities
~Definiteness	Clarity, precision of each instruction
~Finiteness	The algorithm has to stop after a finite amount of steps
~Effectiveness	Each instruction has to be basic enough and feasible

Algorithm Analysis	
Given an algorithm, will it satisfy the requirements?	
Given a number of algorithms to perform the same computation, which one is "best"?	
The analysis required to estimate the "- resource use" of an algorithm	Space and Time Complexity
Implementation	

Algorithm Analysis (cont)	
Testing	
Maintenance	Bug fixes, version management, new features etc.
Space Complexity	
Space Complexity	The amount of memory required by an algorithm to run to completion
Fixed Part	The size required to store certain data/variables, that is independent of the size of the problem, eg. name of the input/-output files,
Variable Part	Space needed by variables, whose size is dependent on the size of the problem.
$S(P)=c+S_p$	c = constant, S_p = instance characteristics which depends on a particular instance

Pseudocode	
Control Flow	
if... then... [else...]	
while... do...	
repeat... until...	
for... do...	
Indentation instead of braces	

Pseudocode (cont)	
Method Declaration	
Algorithm Method (arg [, arg...])	
Input...	
Output	
Method Call	
var.method(arg [, arg...])	
MethodReturn Value	
return expression	
MethodExpressions	
←	Assignment (= in code)
=	Equality Check (== in code)
Superscripts etc. mathematical formatting allowed	
Experimental Approach	Can't always use
Low Level Algorithm Analysis	Make an addition = 1 operation
Using Primitive Operations	Calling a method or returning from a method = 1 operation
	Index in an array = 1 operation.
	Comparison = 1 operation, etc.
Method: Count the primitive operations to find $O(f(n))$	
Growth rate of the running time $T(n)$ is an intrinsic property of an algorithm	Not dependent on hardware.

Pseudocode (cont)	Dictionary Operations	Hash Table (cont)	Hash Table (cont)
Asymptotic Notation Big-Oh, Big Omega, Big Theta, Little-Oh	Find	insert, find, remove take $O(1+\lambda)$ on average	Eliminates primary clustering
Characterizing Algorithms As A Function Of Input Size	Insert	Closed Hashing (Probing Hash Tables)	Unless if TableSize prime and $\lambda < 1/2$, cannot guarantee finding empty cell
Solving Recursive Equations	Remove	$h_i(x) = (\text{hash}(x) + f(i)) \bmod \text{TableSize}$, $f(0)=0$	Secondary Clustering Elements that hash to the same position probe to the same alternative cells, clustering there
by Repeated	Note: No operations that require ordering information	f = Collision Resolution Strategy	
Substitution	Dictionary Implementations	--Linear Probing	
	Lists	--Quadratic Probing	
	Binary Search Trees	--Double Hashing	
	Hash Tables		
by Telescoping	Hash Table	Linear Probing	Double Hashing
	Collision Resolving	f(i)=i (linear function of i)	f(i)=i-hash₂(x) (includes another hash function)
	Separate Chaining (Open Hashing)	Primary Clustering	Example: $\text{hash}_2(x)=R - (x \bmod R)$, where R is a prime < TableSize
	Open Addressing (Closed Hashing - Probing Hash Tables)		Rehashing When λ too big, make bigger TableSize and rehash everything. Takes $O(n)$ but happens rarely..
	--Open Hashing	Insertion time can get long due to blocks of occupied cells are formed	
	--Closed Hashing	Primary Clustering : Any key that hashes into the cluster - even if the keys map to different values- will require several attempts to resolve collision and then it will be added to the cluster.	ALL Closed Hashing cannot work with $\lambda = 1$
			-- Quadratic probing can fail if $\lambda > 0.5$
	Separate Chaining		-- Linear probing and Double hashing are slow if $\lambda > 0.5$
	Each cell in the hash table is the head of a linked list	Worst Case : find, insert	Open Hashing becomes slow once $\lambda > 2$
	Elements are stored in the hash-specified linked list	Deletion requires Lazy Deletion to not mess up the table	Quadratic Probing Proof:
	Records in the linked list can be ordered by:		-
	order of insertion, key value, frequency of access	Quadratic Probing	-
	λ = Load Factor	f(i)=i² (quadratic function of i)	-
	$\lambda \approx n/\text{TableSize}$		-
Dictionary ADT			-
A collection of (key, value) pairs such that each key appears at most once			-
Idea: Use the key as the index information to reach the key	Key-Value Mapping		-

Cuckoo Hashing

2 hash tables

Only insert at the 1st table
Move the value in the 1st table if collision

May cause cycles but if $\lambda < 0.5$, cycle probability low. Still possible, so specify maximum iteration count after which you rehash.

Time Complexity

$O(f(N)) = T(N) \leq cf(N)$ when $N \geq n_0$. **Upper-bound**

$O(g(N)) = T(N) \geq cf(N)$ when $N \geq n_0$. **Lower-bound**

$\Theta(h(N)) = T(N) = O(h(N))$ and $T(N) = \Omega(h(N))$
Tight-bound (Exact)

$o(p(N)) = T(N) < cp(N)$ **Strict Upper-bound**

$f(N)$ is $o(g(N))$ if it's $O(g(N))$ but not $\Theta(g(N))$

$O(1)$ constant

$O(\log N)$ logarithmic

$O(\log^2 N)$ log-squared

$O(N)$ linear

$O(N^2)$ quadratic

$O(N^3)$ cubic

$O(2^N)$ exponential

$f(n) \leq O(g(n))$ is the wrong usage. You need to say $f(n)$ **is** $(=)$ $O(g(n))$

Priority Queues (Heaps)

For applications that require a sorted (but not fully sorted) order of procession of keys.

Priority Queues (Heaps) (cont)

Jobs sent to a printer, Simulation Environments (Discrete Event Simulators)

Priority Queue Operations

insert

deleteMin

Priority Queue Implementations

Simple (Singly) Insert $O(1)$, deleteMin $O(n)$

Linked

List

Sorted Insert $O(n)$, deleteMin $O(1)$

Linked

List

Binary Search Tree (BST) $\Theta(\log n)$ average for insert and deleteMin

Binary Heap Can implement as a **single array**, doesn't require **links**, $O(\log n)$ **worst-case** for insert and deleteMin

d-Heaps Parents can have **d** children

Leftist Heaps

Skew Heaps

Binomial Queues

Binary Heap

Classic method for **priority queues**

Simply called **dynamic-memory allocation** (Different from heap in **Heap**)

Structure Property

Binary Heap (cont)

Heap is a **complete binary tree** is a BT filled completely, except maybe for the **bottom row**, which is filled from **left-to-right**

Array implementation:

For any element in position i :

--**Left child** is in position $2i$

--**Right child** is in position $2i+1$ (after left child)

--**Parent** in position $\lfloor i/2 \rfloor$

Order Property

Every **parent** is **smaller than or equal to** its children, so **findMin** is $O(1)$

A **Max Heap** is the reverse, allowing constant access to the max element

Binary Heap Methods

insert Insert element in **position 0**. Find its possible position, make an empty node, then **percolate up** until the **new element** can be put into the empty position.

Worst case runtime is $O(\log n)$ Average runtime is **constant**

Binary Heap Methods (cont)

deleteMin Delete/make root empty, put the **last element** into array **position 0**, **percolate down** until the **last element** can be put into the empty position.

Worst case runtime is $O(\log n)$ Average runtime is also $O(\log n)$ since an element at the bottom is likely to still go down to the bottom.

Building a Heap **Iterating insertion:** $O(N \log N)$ worst case. $O(N)$ on average.

buildHeap First fill the leafs. ~Half of the tree is filled already. Then, as you place the next elements, the subtrees will all be valid heaps - do **percolate down**. Thus $O(\log \text{Height})$ operations for each node.

Binary Heap Methods (cont)

There are **more high depth nodes** than **high height nodes** in a heap thus buildHeap is a faster method, **O(N) worst case**

Sum of all **heights**:

$$S = \log N + 2(\log N - 1) + \dots + 2^k(\log N - k), k = \log N$$

$$2S = 2\log N + \dots + 2^{k+1}(\log N - k)$$

$$2S - S = 2 + 2^2 + \dots + 2^k - \log N \text{ since } k = \log N \text{ thus } 2^{k+1}(\log N - k) = 0$$

$$S = 2^{k+1} - 2 - \log N$$

$$S = 2N - \log N - 2 \text{ thus } O(N)$$

Sum of all **depths**:

$$S = 0 + 2 \cdot 1 + 2^2 \cdot 2 + \dots + 2^{\log N - 1} \cdot (\log N - 1)$$

$$S = N \log N - 2N + 2 \text{ thus } O(N \log N)$$

Priority Queue Applications (cont)

If $k=N$, we get a sorted list. This is heapSort which is $O(N \log N)$

Discrete Event Simulation
Instead of experimenting, put all events to happen in a queue. Advance clock to the next event each **tick**. Events are stored in a heap to find the next one easily.

--Tick A quantum unit

Priority Queue Applications

Operating System Design

Some Graph Algorithms

Selection and Sorting Problems
Given a list of N elements, and an integer k , the selection problem is to find the k th smallest element.

Take N elements, apply buildHeap, do deleteMin k times.

