

<h3>Abstract Data Types (ADT)</h3> <p>List</p> <p>Stack</p> <p>Queue</p> <hr/> <p>Data Abstraction: Separation of a data type's logical properties from its implementation.</p> <p>-Logical Properties</p> <p>--What are the possible values? What operations will be needed?</p> <p>-Implementation</p> <p>--How can this be done in Java, C++, or any other programming language?</p> <p>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.</p>	<h3>Stack Implementations</h3> <p>Array</p> <p>LinkedList Operations take constant time. Size can grow/shrink easily.</p> <hr/> <p>Overflow: When element count in an array exceeds array size.</p> <p>Underflow: Pop from an empty stack.</p> <p>Evaluate infix expressions, 2 stacks algorithm (Dijkstra):</p> <p>--<i>Value:</i> Push onto the value stack.</p> <p>--<i>Operator:</i> Push onto the operator stack.</p> <p>--<i>Left parenthesis:</i> Ignore.</p> <p>--<i>Right parenthesis:</i> Pop operator and two values; push the result of applying that operator to those values onto the operand (value) stack.</p>	<h3>Trees (cont)</h3> <p>Subtrees Remaining nodes are partitioned into trees themselves, called subtrees. Each subtree is connected by a directed edge from the root.</p> <p>Degree Number of subtrees of a node.</p> <p>Leaf / Terminal Node Node with degree 0.</p> <p>Parent</p> <p>Child</p> <p>Ancestors</p> <p>Path from node₁ to node_k</p> <p>Depth Length of the unique path from root to node.</p> <p>Height Length of the longest downward path from the node to a leaf.</p> <p>Height of a Tree Height of the root.</p> <p>For its implementation, a node can hold:</p> <ul style="list-style-type: none"> -Its first child. -Its next sibling. <p>Thus siblings would be held as a linked list.</p> <p>Without parent/previous sibling information, each node holds only 2 references.</p>	<h3>Binary Tree</h3> <p>Each node can have at most 2 children.</p> <p>Full BT When each node has 2 or 0 children.</p> <p>Perfect BT It's full and each leaf has the same depth.</p> <hr/> <h3>Tree Traversal</h3> <p>Preorder Parent first.</p> <p>Visit root. Traverse left subtree. Traverse right subtree.</p> <p>Postorder Parent last.</p> <p>Traverse left subtree. Traverse right subtree. Visit root.</p> <p>Inorder Left-Parent-Right</p> <p>Traverse left subtree. Visit root. Traverse right subtree</p>
<h3>List Operations</h3> <p>Find (First occurrence)</p> <p>Insert</p> <p>Remove</p> <p>FindKth</p> <p>MakeEmpty</p> <p>PrintList</p>	<h3>Queue - First In Last Out List Operations</h3> <p>Enqueue</p> <p>Dequeue</p> <p>MakeEmpty</p>	<p>Height of a Tree Height of the root.</p> <p>For its implementation, a node can hold:</p> <ul style="list-style-type: none"> -Its first child. -Its next sibling. <p>Thus siblings would be held as a linked list.</p> <p>Without parent/previous sibling information, each node holds only 2 references.</p>	<h3>Search Tree ADT - Binary Search Tree</h3> <p>Provides inorder traversal.</p> <p>Average case: Depth of all nodes on average $\log(N)$</p> <p>Balanced BST maintains all operations at $h=O(\log N)$ time</p>
<h3>List Implementations</h3> <p>Simple Array</p> <p>Simple (Singly) Linked List</p>	<h3>Queue Implementations</h3> <p>Circular Array (Circular Queue)</p> <p>Linked List</p> <p><i>Examples:</i></p> <ul style="list-style-type: none"> -Calls to a call center -Jobs in the printer -Network operations on routers -CPU usage queues 	<h3>Trees</h3> <p>Tree Collection of nodes such that:</p> <p>Root Unless empty, trees have a root.</p>	



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-Right Single Left Rotation

Left-Right Double Left-Right Rotation

Right-Left Double Right-Left Rotation

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Algorithm Analysis

Problem Solving: Life Cycle

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 Space and Time Complexity
required to estimate the "- resource use" of an algorithm

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 Not dependent on hardware.

running time T(n)

is an **intrinsic property** of an algorithm

Pseudocode (cont)	Dictionary Operations	Hash Table (cont)	Hash Table (cont)
Asymptotic Big-Oh, Big	Find	insert, find, remove take $O(1+\lambda)$	Eliminates primary clustering
Notation Omega, Big Theta, Little-Oh	Insert	on average	
Characterizing Algorithms As A Function Of Input Size	Remove	Closed Hashing (Probing Hash Tables)	Unless if TableSize prime and $\lambda < 1/2$, cannot guarantee finding empty cell
Solving Recursive Equations	Note: No operations that require ordering information	$h_i(x) = (\text{hash}(x) + f(i)) \bmod$	
by $T(N) = T(N/2) + c = T(N/4) + c + c = \dots = T(N/2^k) + kc$, choose $k = \log N$, $T(N) = T(1) + c \log N = \Theta(\log N)$	Dictionary Implementations	TableSize, $f(0)=0$	
Repeated	Lists	f = Collision Resolution Strategy	Secondary Clustering Elements that hash to the same position probe to the same alternative cells, clustering there
Substitution	Binary Search Trees	--Linear Probing	
by Telescoping $T(N) = T(N/2) + c$	Hash Tables	--Quadratic Probing	
$T(N/2) = T(N/4) + c$	Hash Table	--Double Hashing	Double Hashing $f(i) = i - \text{hash}_2(x)$ (includes another hash function)
...	Collision Resolving	Linear Probing $f(i) = i$ (linear function of i)	Example: $\text{hash}_2(x) = R - (x \bmod R)$, where R is a prime $< \text{TableSize}$
$+ \dots - \dots$ (canceling opposite terms)	Separate Chaining (Open Hashing)	Primary Clustering $n < \text{TableSize}$ guarantees finding a free cell	
$T(N) = T(1) + c \log N = \Theta(\log N)$	Open Addressing (Closed Hashing - Probing Hash Tables)	Insertion time can get long due to blocks of occupied cells are formed	
-	--Open Hashing Collisions are stored outside of the table	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.	
-	--Closed Hashing Collisions are stored at another slot in the table	Worst Case $O(n)$: find, insert	ALL Closed Hashing cannot work with $\lambda = 1$
-	Separate Chaining	Deletion requires Lazy Deletion to not mess up the table	-- Quadratic probing can fail if $\lambda > 0.5$
-	Each cell in the hash table is the head of a linked list	Quadratic Probing $f(i) = i^2$ (quadratic function of i)	-- Linear probing and Double hashing are slow if $\lambda > 0.5$
-	Elements are stored in the hash-specified linked list		Open Hashing becomes slow once $\lambda > 2$
-	Records in the linked list can be ordered by: order of insertion, key value, frequency of access		Quadratic Probing Proof: - - - - - - -
-	$\lambda = \text{Load Factor}$		
-	$\lambda \approx n / \text{TableSize}$		
-	Dictionary ADT		
A collection of (key, value) pairs such that each key appears at most once	A collection of (key, value) pairs such that each key appears at most once		
<i>Idea:</i> Use the key as the index information to reach the key	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 $\Theta(\log n)$ average for insert and deleteMin

Binary Search Tree (BST)

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 **Heap** (Different from heap in **dynamic memory allocation**)

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 leaves. ~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.

