# CSE 332 Autumn 2023 Lecture 11: B Trees and Hashing

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# B Trees (aka B+ Trees)

- Two types of nodes:
  - Internal Nodes
    - Sorted array of M 1 keys
    - Has *M* children
    - No other data!
  - Leaf Nodes
    - Sorted array of *L* key-value pairs
- Subtree between values a and b must contain only keys that are  $\geq a$  and < b

-1

- If a is missing use  $-\infty$
- If b is missing use  $\infty$



### Find

- Start at the root node
- Binary search to identify correct subtree
- Repeat until you reach a leaf node
- Binary search the leaf to get the value



# B Tree Structure Requirements

- Root:
  - If the tree has  $\leq L$  items then root is a leaf node
  - Otherwise it is an internal node
- Internal Nodes:
  - Must have at least  $\left[\frac{M}{2}\right]$  children (at least half full)
- Leaf Nodes:
  - Must have at least Must have at least  $\left[\frac{L}{2}\right]$  items (at least half full)
  - All leaves are at the same depth

#### Insertion Summary

- Binary search to find which leaf should contain the new item
- If there's room, add it to the leaf array (maintaining sorted order)
- If there's not room, **split** 
  - Make a new leaf node, move the larger  $\left|\frac{L+1}{2}\right|$  items to it
  - If there's room in the parent internal node, add new leaf to it (with new key bound value)
  - If there's not room in the parent internal node, **split** that!
    - Make a new internal node and have it point to the larger  $\left|\frac{M+1}{2}\right|$
    - If there's room in the parent internal node, add this internal node to it
    - If there's not room, repeat this process until there is!

#### Insertion TLDR

- Find where the item goes by repeated binary search
- If there's room, just add it
- If there's not room, split things until there is

















		3		5			
-							
	1		3		5		
	2		4		6		







#### Let's do it together!

• M = 3, L = 3





# Running Time of Find

- Maximum number of leaves:
  - $\frac{2n}{L}$ •  $\Theta\left(\frac{n}{L}\right)$
- Maximum height of the tree:
  - $2 \log_M \frac{2n}{L}$ •  $\Theta\left(\log_M \frac{n}{L}\right)$
- Find:
  - One binary search per level of the tree
    - $\Theta(\log_2 M)$  per search
  - One binary search in the leaf
    - $\Theta(\log_2 L)$

Overall:  $\Theta\left(\log_2 M \cdot \log_M \frac{n}{L} + \log_2 L\right)$ Usually simplified to:  $\Theta(\log_2 M \cdot \log_M n)$ 

# Running Time of Insert



















# Delete Summary

- Find the item
- Remove the item from the leaf
  - If that causes the leaf to be underfull, adopt from a neighbor
  - If that would cause the neighbor to be underfull, merge those two leaves
  - Update the parent
    - If that causes the parent to be underfull, adopt from a neighbor
    - If that causes the neighbor to be underfull, merge
    - Update the parent

• ...

# Delete TLDR

- Find and remove from leaf
- Keep doing this until everything is "full enough":
  - If the node is now too small, adopt from a neighbor
  - If the neighbor is too small then merge

# Aside: Implementation

- What an internal node class might look like:
  - int M
  - int[] keys
  - Node[] children
  - int num\_children
- What a leaf node class might look like:
  - int L
  - E[] data
  - int num\_items

# Next topic: Hash Tables

Data Structure	Time to insert	Time to find	Time to delete		
Unsorted Array	$\Theta(n)$	$\Theta(n)$	$\Theta(n)$		
Unsorted Linked List	$\Theta(n)$	$\Theta(n)$	$\Theta(n)$		
Sorted Array	$\Theta(n)$	$\Theta(\log n)$	$\Theta(n)$		
Sorted Linked List	$\Theta(n)$	$\Theta(n)$	$\Theta(n)$		
Binary Search Tree	$\Theta(n)$	$\Theta(n)$	$\Theta(n)$		
AVL Tree	$\Theta(\log n)$	$\Theta(\log n)$	$\Theta(\log n)$		
Hash Table (Worst case)	$\Theta(n)$	$\Theta(n)$	$\Theta(n)$		
Hash Table (Average)	Θ(1)	Θ(1)	Θ(1)		

# Two Different ideas of "Average"

- Expected Time
  - The expected number of operations a randomly-chosen input uses
  - Assumed randomness from somewhere
    - Most simply: from the input
    - Preferably: from the algorithm/data structure itself
  - f(n) = sum of the running times for each input of size n divided by the number of inputs of size n
- Amortized Time
  - The long-term average per-execution cost (in the worst case)
  - Rather than look at the worst case of one execution, look at the total worst case of a sequential chain of many executions
    - Why? The worst case may be guaranteed to be rare
  - f(n) = the sum of the running times from a sequence of n sequential calls to the function divided by n

### Amortized Example

- ArrayList Insert:
  - Worst case:  $\Theta(n)$

0	1	2	3	4	5	6	7	8				

# Amortized Example

• ArrayList Insert:

• ...

- First 8 inserts: 1 operation each
- 9<sup>th</sup> insert: 9 operations
- Next 7 inserts: 1 operation each
- 17<sup>th</sup> insert: 17 operations
- Next 15 inserts: 1 operation each

Do x operations with cost 1 Do 1 operation with cost x Do x operations with cost 1 Do 1 operation with cost 2x Do 2x operations with cost 1 Do 1 operation with cost 4x Do 4x operations with cost 1 Do 1 operation with cost 8x

• • •

Amortized: each operation cost 2 operations  $\Theta(1)$ 

0 1 2 3 4 5 6 7 8

0	1	2	3	4	5	6	7	8							
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# Hash Tables

- Motivation:
  - Why not just have a gigantic array?

# Hash Tables

- Idea:
  - Have a small array to store information
  - Use a hash function to convert the key into an index
    - Hash function should "scatter" the keys, behave as if it randomly assigned keys to indices
  - Store key at the index given by the hash function
  - Do something if two keys map to the same place (should be very rare)
    - Collision resolution



# Example



- Key: Phone Number
- Value: People
- Table size: 10
- h(phone) = number as an integer % 10
- h(8675309) = 9

# What Influences Running time?