



CSE373: Data Structures & Algorithms Lecture 5: Dictionary ADTs; Binary Trees

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Today's Outline

Announcements

- Homework 1 due TODAY at 10:59pm ©
- Homework 2 out
 - Due online next Friday 10:59 pm

Today's Topics

- Finish Asymptotic Analysis
- Dictionary ADT (a.k.a. Map): associate keys with values
 - Extremely common
- Binary Trees

Summary of Asymptotic Analysis

Analysis can be about:

- The problem or the algorithm (usually algorithm)
- Time or space (usually time)
- Best-, worst-, or average-case (usually worst)
- Upper-, lower-, or tight-bound (usually upper)

• The most common thing we will do is give an O upper bound to the worst-case running time of an algorithm.

Addendum: Timing vs. Big-O Summary

- Big-O
 - Examine the algorithm itself, not the implementation
 - Reason about performance as a function of *n*
 - For *small n*, an algorithm with worse asymptotic complexity might be faster
- Timing
 - Compare implementations
 - Focus on data sets other than worst case
 - Determine what the constants actually are

Let's take a breath

- So far we've covered
 - Simple ADTs: stacks, queues, lists
 - Some math (proof by induction)
 - Algorithm analysis
 - Asymptotic notation (Big-Oh)
- Coming up....
 - Many more ADTs!
 - Starting with dictionaries

The Dictionary (a.k.a. Map) ADT



A Modest Few Uses

Used to store information with some key and retrieve it efficiently

- Lots of programs do that!
- Search: phone directories
- Networks: router tables
- Operating systems: page tables
- Compilers: symbol tables
- Databases: dictionaries with other nice properties
- Biology: genome maps
- ...

Possibly the most widely used ADT

Simple implementations

For dictionary with *n* key/value pairs

•	Unsorted linked-list	insert $O(1)^*$	find O(n)	delete O(n)
•	Unsorted array	<i>O</i> (1)*	<i>O</i> (n)	<i>O</i> (n)
•	Sorted linked list	<i>O</i> (n)	<i>O</i> (n)	<i>O</i> (n)
•	Sorted array	<i>O</i> (n)	O(log n)	<i>O</i> (n)

* Unless we need to check for duplicates

We'll see a Binary Search Tree (BST) probably does better but not in the worst case (unless we keep it balanced)

Lazy Deletion

10	12	24	30	41	42	44	45	50
\checkmark	×	\checkmark	\checkmark	\checkmark	\checkmark	×	\checkmark	\checkmark

A general technique for making delete as fast as find:

- Instead of actually removing the item just mark it deleted

Pros:

- Simpler
- Can do removals later in batches
- If re-added soon thereafter, just unmark the deletion

Cons:

- Extra *space* for the "is-it-deleted" flag
- Data structure full of deleted nodes wastes *space*
- May complicate other operations

Better dictionary data structures

There are many good data structures for (large) dictionaries

- 1. Binary trees
- 2. AVL trees
 - Binary search trees with *guaranteed balancing*
- 3. B-Trees
 - Also always balanced, but different and shallower
 - B-Trees are not the same as Binary Trees
 - B-Trees generally have large branching factor
- 4. Hashtables
 - Not tree-like at all

Skipping: Other balanced trees (e.g., red-black, splay)

Tree terms

Root (tree) Leaves (tree) Children (node) Parent (node) Siblings (node) Ancestors (node) Descendents (node) Subtree (node)

Depth (node) Height (tree) Degree (node) Branching factor (tree)



More tree terms

- There are many kinds of trees
 - Binary trees, linked lists, etc...
- There are many kinds of binary trees
 - binary search tree, binary heaps
- A tree can be balanced or not
 - A balanced tree with *n* nodes has a height of $O(\log n)$
 - Use different "balance conditions" to achieve this

Kinds of trees

Certain terms define trees with specific structure

- Binary tree: Each node has at most 2 children (branching factor 2)
- *n*-ary tree: Each node has at most *n* children (branching factor *n*)
- Perfect tree: Each row completely full
- Complete tree: Each row completely full except maybe the bottom row, which is filled from left to right



What is the height of a perfect binary tree with n nodes? A complete 14-ary tree?

Binary Trees

- Binary tree: Each node has at most 2 children (branching factor 2)
- Binary tree is
 - A root (with data)
 - A left subtree (may be empty)
 - A right subtree (may be empty)
- Representation:



• For a dictionary, data will include a key and a value



Binary Tree Representation



Binary Trees: Some Numbers

Recall: height of a tree = longest path from root to leaf (count edges)

For binary tree of height *h*: - max # of nodes: 2^{h+1} -1 - max # of leaves: 2^h - min # of leaves: – min # of nodes: h + 1For n nodes, we cannot do better than $O(\log n)$ height and we want to avoid O(n) height 16

Calculating height

What is the height of a tree with root **root**?

```
int treeHeight(Node root) {
          ???
}
```

Calculating height

What is the height of a tree with root **root**?

Running time for tree with *n* nodes:

O(n) – single pass over tree

Note: non-recursive is painful – need your own stack of pending nodes; much easier to use recursion's call stack

Tree Traversals

A *traversal* is an order for visiting all the nodes of a tree

- Pre-order: root, left subtree, right subtree
- *In-order*: left subtree, root, right subtree
- *Post-order*: left subtree, right subtree, root



(an expression tree)

```
void inOrderTraversal(Node t) {
    if(t != null) {
        inOrderTraversal(t.left);
        process(t.element);
        inOrderTraversal(t.right);
    }
}
```





= completed node ✓ = element has been processed

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 + * 2 4 5
- In-order: left subtree, root, right subtree
 2*4+5
- Post-order: left subtree, right subtree, root
 2 4 * 5 +



(an expression tree)