

CSCI 136
Data Structures &
Advanced Programming

Lecture 22

Fall 2017

Instructor: Bills

Announcement

Power outage (3-5am)

We'll be shutting down systems at 10pm tonight

Rebooting at 9am tomorrow

Last Time

- Wrap up Binary Tree Iterators
- Breadth-First and Depth-First Search
- Array Representations of (Binary) Trees
- Application: Huffman Encoding

Today

Improving Huffman's Algorithm

- Priority Queues & Heaps
 - A “somewhat-ordered” data structure
 - Conceptual structure
 - Efficient implementations

Huffman Codes

- Example
 - AN_ANTARCTIC_PENGUIN
 - Compute letter frequencies

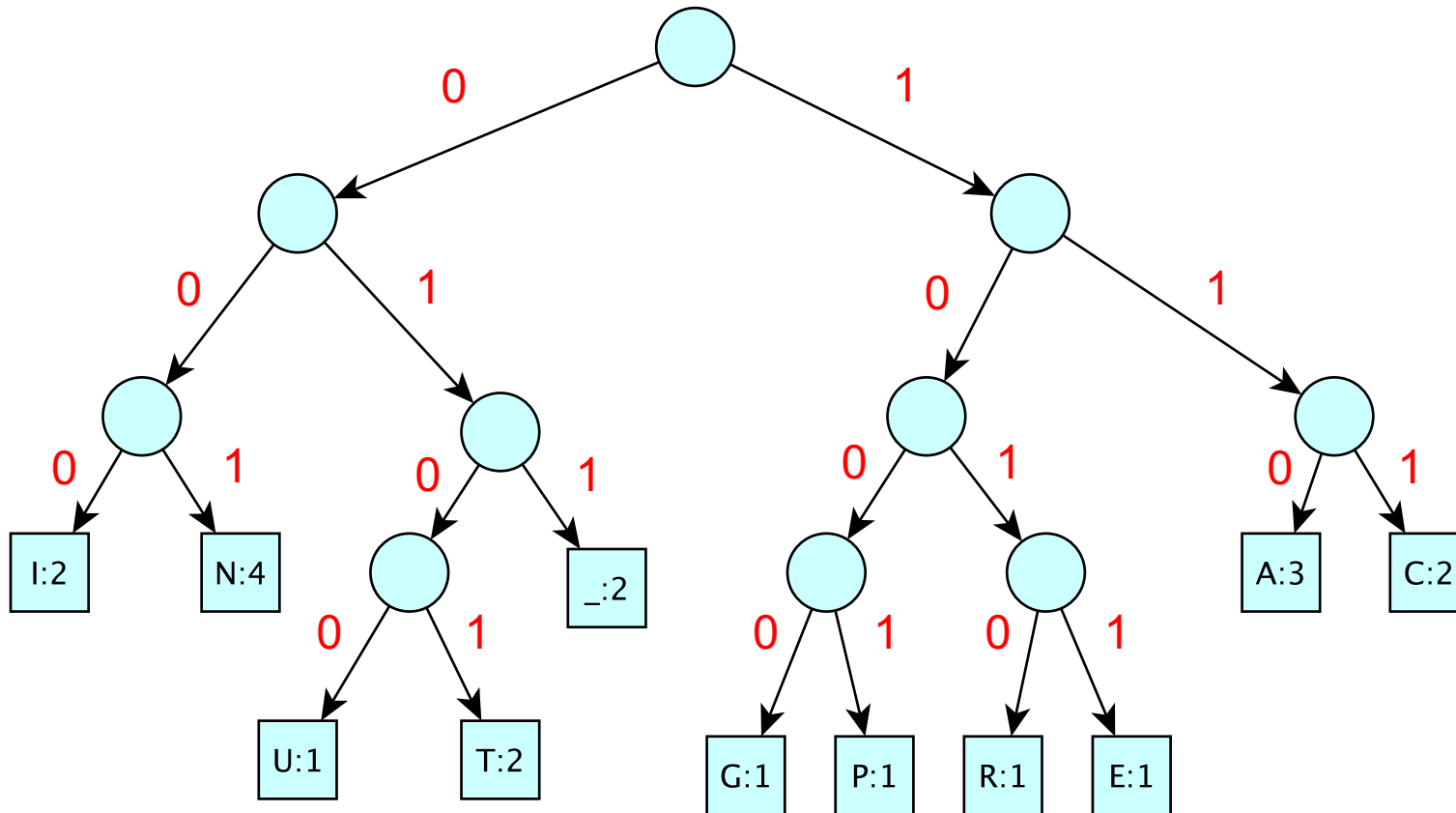
A	C	E	G	I	N	P	R	T	U	_
3	2	1	1	2	4	1	1	2	1	2

- **Key Idea:** Use fewer bits for most common letters

A	C	E	G	I	N	P	R	T	U	_
3	2	1	1	2	4	1	1	2	1	2
110	111	1011	1000	000	001	1001	1010	0101	0100	011

- Uses 67 bits to encode entire string

The Encoding Tree



Left = 0; Right = 1

Huffman Encoding Algorithm

Input: symbols of alphabet with frequencies

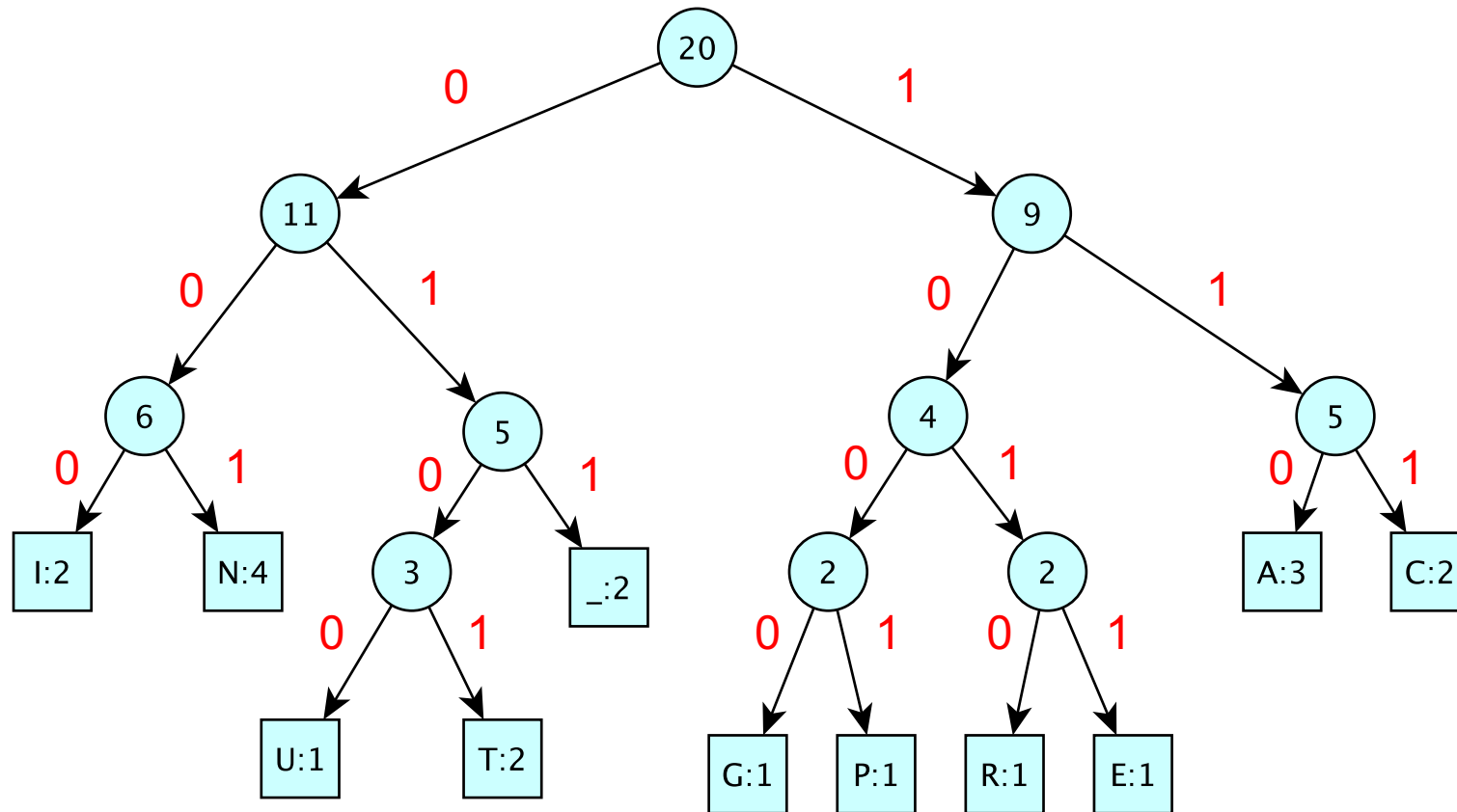
- Huffman encode as follows
 - Create a single-node tree for each symbol: key is frequency; weight is letter
 - while there is more than one tree
 - Find two trees T1 and T2 with lowest weights
 - Merge them into new tree T with:
$$\text{weight} = T1.\text{weight} + T2.\text{weight}$$
- Theorem: The tree computed by Huffman is an optimal encoding for given frequencies

Demo

- To run the Huffman code demo found on course webpage:

```
java -jar huffman.jar
```


The Encoding Tree (With Weights)



Left = 0; Right = 1

*Each node's value is the sum of the frequencies of all its children

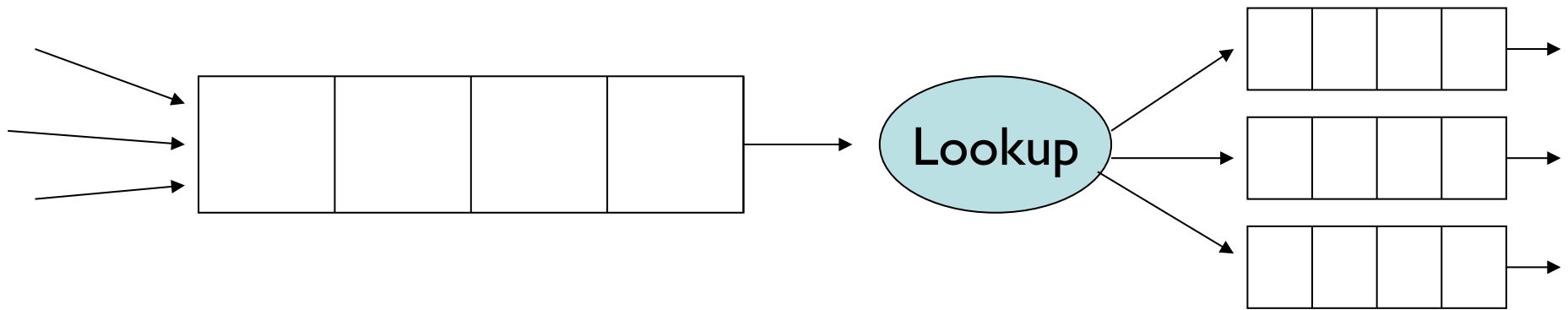
Implementing the Algorithm

- Keep a Vector of Binary Trees
- Sort them by decreasing frequency
 - Removing two smallest frequency trees is fast
- Insert merged tree into correct sorted location in Vector
- Running Time:
 - $O(n \log n)$ for initial sorting
 - $O(n^2)$ for while loop
- Can we do better...?

What Huffman Encoder Needs

- A structure S to hold items with *priorities*
- S should support operations
 - `add(E item); // add an item`
 - `E removeMin(); // remove min priority item`
- S should be designed to make these two operations fast
- If, say, they both ran in $O(\log n)$ time, the Huffman while loop would take $O(n \log n)$ time instead of $O(n^2)$!
- We've seen this situation before....

Priority Queues



Packet Sources May Be Ordered by Sender

sysnet.cs.williams.edu

bull.cs.williams.edu

yahoo.com

spammer.com

priority = 1 (best)

2

10

100 (worst)

Priority Queues

- Priority queues are also used for:
 - Scheduling processes in an operating system
 - Priority is function of time lost + process priority
 - Order services on server
 - Backup is low priority, so don't do when high priority tasks need to happen
 - Scheduling future events in a simulation (lab next week!)
 - Medical waiting room
 - Huffman codes - order by tree size/weight
 - A variety of graph/network algorithms
 - To roughly rank choices that are generated out of order

Priority Queues

- Name is misleading: They are **not FIFO**
- Always dequeue object with **highest priority** (smallest rank) regardless of when it was enqueued
- Data can be received/inserted in any order, but it is always returned/removed according to priority
- Like ordered structures (i.e., OrderedVectors and OrderedLists), PQs require comparisons of values

An Apology

- On behalf of computer scientists everywhere, I'd like to apologize for the confusion that inevitably results from the fact that

Higher Priority \leftrightarrow Lower Rank

- The PQ removes the *lowest ranked* value in an ordering: that is, the *highest priority* value!

We're sorry!

PQ Interface

```
public interface PriorityQueue<E extends Comparable<E>> {  
    public E getFirst(); // peeks at minimum element  
    public E remove(); // removes minimum element  
    public void add(E value); // adds an element  
    public boolean isEmpty();  
    public int size();  
    public void clear();  
}
```


Notes on PQ Interface

- Unlike previous structures, we do not extend any other interfaces
 - Many reasons: For example, it's not clear that there's an obvious iteration order
- PriorityQueue uses Comparables: methods *consume* Comparable parameters and *return* Comparable values
 - Could be made to use Comparators instead...

Implementing PQs

- Queue?
 - Wouldn't work so well because we can't insert and remove in the "right" way (i.e., keeping things ordered)
- OrderedVector?
 - Keep ordered vector of objects
 - $O(n)$ to add/remove from vector
 - Details in book...
 - Can we do better than $O(n)$?
- Heap!
 - Partially ordered binary tree

Heap

- A heap is a special type of tree
 - Root holds smallest (highest priority) value
 - Subtrees are also heaps (this is important!)
- Values increase in priority (decrease in rank) from leaves to root (from descendant to ancestor)
- *Invariant for nodes:* For each child of each node
 - `node.value() <= child.value()` // if child exists
- Several valid heaps for same data set (no unique representation)

Inserting into a PQ

- Add new value as a leaf
- “Percolate” it up the tree
 - while (value < parent’s value) swap with parent
- This operation preserves the heap property since new value was the only one violating heap property
- Efficiency depends upon speed of
 - Finding a place to add new node
 - Finding parent
 - Tree height

Removing From a PQ

- Get value from root node (highest priority)
- Find a leaf, delete it, put its *data* in the root
- “Push” *data* down through the tree
 - while (*data.value* > value of (at least) one child)
 - Swap *data* with data of **smaller** child
- This operation preserves the heap property
- Efficiency depends upon speed of
 - Finding a leaf
 - Finding locations of children
 - Height of tree

Implementing Heaps

- VectorHeap
 - Use conceptual array representation of BT (ArrayTree)
 - But use extensible Vector instead of array (makes adding elements easier)
 - Note:
 - Root of tree is location 0 of Vector
 - Children of node in location i are in locations $2i+1$ (left) and $2i+2$ (right)
 - Parent of node i is in location $(i-1)/2$
 - Remember: dividing Integers truncates the result

Implementing Heaps

- Strategy: tree modifications that always preserve tree *completeness*, but may violate heap property. Then fix.
 - Add/remove never add gaps to array
 - We always add in next available array slot (left-most available spot in binary tree)
 - We always remove using “final” leaf
 - *Heap Invariant* becomes
 - $\text{data}[i] \leq \text{data}[2i+1]; \text{data}[i] \leq \text{data}[2i+2]$ (or kids might be null)
 - When elements are added and removed, do small amount of work to “re-heapify”
 - How small? Note: finding a node’s child or parent takes constant time, as does finding “final” leaf or next slot for adding
 - Since this heap corresponds to a full binary tree, the depth of the tree is $O(\log n)$, so percolate/pushDown takes $O(\log n)$ time!

VectorHeap Summary

- Let's look at VectorHeap code....
- Add/remove are both $O(\log n)$
- Data is not completely sorted
 - “Partial” order is maintained: all root-to-leaf paths
- Note: `VectorHeap(Vector<E> v)`
 - Takes an unordered Vector and uses it to construct a heap
 - How?

Heapifying A Vector (or array)

Problem: You are given a Vector V that is not a valid heap, and you want to “heapify” V

- Method I: Top-Down
 - Assume $V[0..k]$ satisfies the heap property
 - Now call `percolateUp` on item in location $k+1$
 - Then $V[0..k+1]$ satisfies the heap property

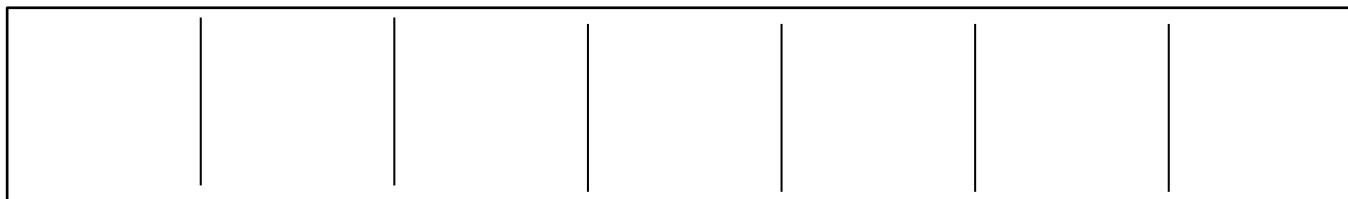


----->
Grow heap one element at a time

Heapifying A Vector (or array)

Problem: You are given a Vector V that is not a valid heap, and you want to “heapify” V

- Method II: Bottom-up
 - Assume $V[k..n]$ satisfies the heap property
 - Now call pushDown on item in location $k-1$
 - Then $V[k-1..n]$ satisfies heap property



← - - - - -
Grow heap one element at a time

Top-Down vs Bottom-Up

- Top-down heapify: elements at depth d may be swapped d times: Total # of swaps is

$$\sum_{d=1}^h d2^d = (h - 1)2^{h+1} = (\log n - 1)2n + 2$$

(recall: $h = \log n$)

- This is $O(n \log n)$
- Some intuition: most of the elements are in the lowest levels of the tree, so each of them might have to move to root: $O(\log n)$ swaps per element

Top-Down vs Bottom-Up

- Bottom-up heapify: elements at depth d may be swapped $h-d$ times: Total # of swaps is

$$\sum_{d=1}^h (h-d)2^d = 2^{h+1} - h - 2 = 2n - \log n + 2$$

- This is $O(n)$ --- beats top-down!
- Some intuition: most of the elements are in the lowest levels of the tree, so each of them will only be pushed down (swapped) a small number of times **SO COOL!!!**

Some Sums (for your toolbox)

$$\sum_{d=0}^{d=k} 2^d = 2^{k+1} - 1$$

All of these can be proven by (weak) induction.

$$\sum_{d=0}^{d=k} r^d = (r^{k+1} - 1) / (r - 1)$$

Try these to hone your skills

$$\sum_{d=1}^{d=k} d * 2^d = (k - 1) * 2^{k+1} + 2$$

The second sum is called a geometric series. It works for any $r \neq 0$

$$\sum_{d=1}^{d=k} (k - d) * 2^d = 2^{k+1} - k - 2$$