Discussion 7(11/16)

ECE 17

Statistical Analysis/Timing of code

- Always think before you implement code.
- Why?
 - Blindly implemented code and its subsequent debugging might take longer than thinking before coding
 - You might dig yourself into a hole, trying to debug dysfunctional code
 - Bad/inefficient usage of algorithms

Statistical Analysis/Timing of code

- This is for HW5 Part 1 tips, for word count and dictionary populating
 - Words with capitalized characters are the same word
 - "Hello" and "hello" are the same word
 - Sample solution => set all the words lowercase or read/parse everything as lowercase
 - Punctuation
 - "Isn't" is a word
 - So only checking for letters here is not sufficient:
 - isalpha() or if(char >= 'a' && char <= 'z') by themselves will fail here

Timing

- Remember every nested loop you have equals to exponential growth in time
- For example:

Timing

Example: Given a **color** and **shape** and arrays possible versions of each(**shape**[] and **color**[]). Find out of the given **color** and **shape** combination is possible. Let's say each array has 20 elements

- The bad version will run 20² times = 4000 iterations!!!
- If we have 100 possibilities of each, that will be 10,000 iterations!!!!

Lets try a better version

Timing

- Boolean shapePossible = false;
- Boolean colorPossible = false;
- for(int i = 0; i < shape.size; i++)</pre>
 - shapePossible= (shape==shape[i])?true:false;
- for(int i = 0; i < color.size; i++)</pre>
 - colorPossible= (color==color[i])?true:false;
- if(colorPossible && shapePossible)
 - Cout << "Yes";

This will run 20+20 times = 40 iterations, a huge improvement.

If we have 100 possibilities of each, it will be 200 iterations, instead of 10,000

Data Compression

- Basic idea: what if we can represent longer repetitive/common with a shorter place holder
- Used in zip files
- We would want lossless compression, aka compression that has no data loss
 - Non lossless or lossy compression will be compression with data loss.
 - Ex: turning a 4k photo into a 2k photo. You can never get the lost data back

Data Compression

Fun example:

Spider-man was **happy** to see **happy** and now they're both **happy**

If we set the word "happy" to "1", we now get:

Spider-man was $\underline{\mathbf{1}}$ to see $\underline{\mathbf{1}}$ and now they're both $\underline{\mathbf{1}}$

In replacing one word, we shaved down 12 characters

Dictionary Compression

We can expand our compression example, what if we swap every word out. We know almost every english sentence/paragraph/written work as repetitive words.

Ex:

Jingle bells, jingle bells, jingle all the way

We use the dictionary:

12121345

The encoded version uses so much less data

| Dictionary | | |
|------------|----------|--|
| word | encoding | |
| jingle | 1 | |
| bells, | 2 | |
| all | 3 | |
| the | 4 | |
| way | 5 | |

Huffman Encoding

**NOTE, this is a very very watered down version, please do your own research for this

- We want a even better version of compression, so we will use Huffman Encoding.
 - used for JPEG and MPEG-2
- Huffman Encoding takes in account for the frequency of the input and creates encoding where the more common a entry is, the shorter its encoding is.
 - Ex: if 'a' shows up 100 times and 'b' shows up 15 times, the length of encoding of 'a' will be shorter than length of encoding 'b'
- Goal: set the most common entry(word or character) with the shortest encoded value to yield lowest overall data use

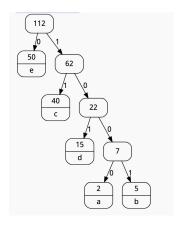
Huffman encoding

- General idea:
 - We first collect the frequency of each entry(word or character) from the input
 - Then we generate a tree with the most common entry as root
 - To generate the encoding, we set each path in the tree as '0' or '1', and we "walk" from the root to that node

Huffman encoding

- Example: Build a huffman encoding for this freq table:
 - The table means: for entry 'a', it showed up 2 times. 'b' showed up 5 times, 'c' showed up 40 times, and so on.
- 1) We build a frequency tree
 - Each node on the main path denotes the sum of characters below it.
 - b) Each node representing a character has its frequency above it

| Frequency Table | | |
|-----------------|------|--|
| Entry | Freq | |
| a | 2 | |
| b | 5 | |
| С | 40 | |
| d | 15 | |
| е | 50 | |



Huffman Encoding

3. We build the Encoding Table based on our tree

This is achieved by "walking" from the root to that node

Notice how the highest frequency = shortest encoding

And opposite for lowest frequency

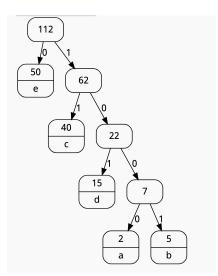
4. Now, used this encoding to replace your entries(chars, words) in your data input

**Again, this is a VERY VERY watered down

Version, so please do you own research for

Huffman encoding

| Frequency Table | | | |
|-----------------|------|----------|--|
| Entry | Freq | Encoding | |
| е | 50 | 0 | |
| С | 40 | 11 | |
| d | 15 | 101 | |
| b | 5 | 1001 | |
| а | 2 | 1000 | |



Any questions?

