ECE 264 – Spring 2013 Huffman Coding

(Derived from an assignment by Professor Vijay Raghunathan)

Huffman coding is a widely used compression algorithm used in JPEG compression as well as in MP3 audio compression. This document explains the technique. Please see the respective SPEC files for PA10 and PA11 for concrete details on what to do.

1 ASCII Coding

Many programming languages use ASCII (which stands for American Standard Code for Information Interchange) encoding to represent characters. In ASCII encoding, every character is encoded (represented) with the same number of bits (8-bits) per character. Since there are 256 different values that can be represented with 8-bits, there are potentially 256 different characters in the ASCII character set, as shown in the ASCII character table available at http://www.asciitable.com/.

Let us now look at a simple example of ASCII encoding of characters. Using ASCII encoding (8 bits per character) the 13-character string "go go gophers" requires 13 * 8 = 104 bits. The table below shows how the coding works.

Character	ASCII Code	8-bit binary value
Space	32	00100000
e	101	01100101
g	103	01100111
h	104	01101000
o	111	01101111
p	112	01110000
r	114	01110010
S	115	01110011

The given string would be written as the following stream of bits (the spaces would not be written, just the 0's and 1's)

2 From ASCII Coding Towards Huffman Coding

Next, let us see how we might use fewer bits using a simpler coding scheme. Since there are only 8 different characters in "go go gophers", it is possible to use only 3-bits to encode the 8 different characters. We might, for example, use the coding shown in the table below (keep in mind that other 3-bit encodings are also possible).

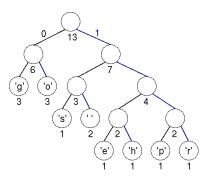
Character	Code Value	3-bit binary value
g	0	000
O	1	001
p	2	010
h	3	011
e	4	100
r	5	101
S	6	110
Space	7	111

Now the string "go go gophers" would be encoded as: 000 001 111 000 001 111 000 001 111 000 011 110 010 101 110. As you can see, by using three bits per character instead of eight bits per character that ASCII uses, the string "go go gophers" uses a total of 39 bits instead of 104 bits.

However, even in this improved coding scheme, we used the same number of bits to represent each character, irrespective of how often the character appears in our string. Even more bits can be saved if we use fewer than three bits to encode characters like g, o, and space that occur frequently and more than three bits to encode characters like e, h, p, r, and s that occur less frequently in "go go gophers". This is the basic idea behind Huffman coding: to use fewer bits for characters that occur more frequently. We will see how this is done using a tree data structure that stores the characters as its leaf nodes, and whose root-to-leaf paths provide the bit sequence used to encode the characters.

2.1 Towards a Coding Tree

Using a binary tree for coding, all characters are stored at the leaves of a tree. A left-edge is numbered 0 and a right-edge is numbered 1. The code for any character/leaf node is obtained by following the root-to-leaf path and concatenating the 0's and 1's. The specific structure of the tree determines the coding of any leaf node using the 0/1 edge convention described. As an example, the tree below yields the coding table following it.



Character	Binary code
, ,	101
'e'	1100
'g'	00
'h'	1101
o'	01
'p'	1110
'r'	1111
's'	100

Using this coding, "go go gophers" is encoded (again, spaces would not appear in the bit-stream) as: 00 01 101 00 01 101 00 01 1110 1101 1100 1111 100. This is a total of 37 bits, two bits fewer than the improved encoding in which each of the 8 characters has a 3-bit encoding! The bits are saved by coding frequently occurring characters like 'g' and 'o' with fewer bits (here two bits) than characters that occur less frequently like 'p', 'h', 'e', and 'r'.

by right-right-left-right to the letter 'h'. Continuing thus yields a decoded string "sphere."

2.2 Prefix codes

When all characters are stored in leaves, and every interior (non-leaf) node has two children, the coding induced by the 0/1 convention outlined above satisfies a very important property called the *prefix property* which states that no bit-sequence encoding of a character is the prefix of the bit-sequence encoding of any other character. This makes it possible to decode a bitstream using the coding tree by following root-to-leaf paths. The tree shown above for "go go gophers" satisfies this prefix property and is an optimal tree. There are other trees that use 37 bits; for example you can simply swap any sibling nodes and get a different encoding that uses the same number of bits. Next, we look at an algorithm for constructing such an optimal tree. This algorithm is called Huffman coding, and was invented by David A. Huffman in 1952 when he was a Ph.D. student at MIT.

3 Huffman Coding

In the previous section we saw examples of how a stream of bits can be generated from an encoding. We also saw how the tree can be used to decode a stream of bits. We will discuss how to construct the tree here using Huffman's algorithm.

We will assume that associated with each character is a weight that is equal to the number of times the character occurs in a file. For example, in the string "go go gophers", the characters 'g' and 'o' have weight 3, the space has weight 2, and the other characters have weight 1. When compressing a file, we will need to first read the file and calculate these weights. Assume that all the character weights have been calculated. Huffman's algorithm assumes that we are building a single tree from a group (or forest) of trees. Initially, all the trees have a single node containing a character and the character's weight. Iteratively, a new tree is formed by picking two trees and making a new tree whose child nodes are the roots of the two trees. The weight of the new tree is the sum of the weights of the two sub-trees. This decreases the number of trees by one in each iteration. The process iterates until there is only one tree left. The algorithm is as follows:

1. Begin with a forest of trees. All trees have just one node, with the weight of the tree equal to the weight of the character in the node. Characters that

occur most frequently have the highest weights. Characters that occur least frequently have the smallest weights.

2. Repeat this step until there is only one tree:

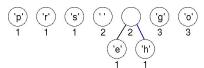
Choose two trees with the smallest weights; call these trees T1 and T2. Create a new tree whose root has a weight equal to the sum of the weights T1 + T2 and whose left sub-tree is T1 and whose right sub-tree is T2.

3. The single tree left after the previous step is an optimal encoding tree.

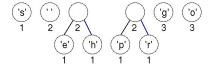
We shall use the string "go go gophers" as an example. Initially we have the forest shown below. The nodes are shown with a weight that represents the number of times the node's character occurs in the given input string/file.



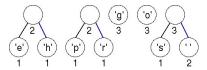
We pick two minimal nodes. There are five nodes with the minimal weight of 1. In this implementation, we maintain a priority queue with items arranged according to their weights. When two items have the same weight, a leaf node (i.e., a node associated with an ASCII character) is always ordered first. If both nodes are leaf nodes, they are ordered according to their ASCII coding. If both nodes are non-leaf nodes, they are ordered according to the creation times of the nodes. We always pick the first two items in the priority queue, namely, nodes for characters 'e' and 'h'. We create a new tree whose root is weighted by the sum of the weights chosen. The order of the nodes in the priority queue also determines the left and right child nodes of the new root. We now have a forest of seven trees as shown here. Although the newly created node has the same weight as Space, it is ordered after Space in the priority queue because Space is an ASCII character.



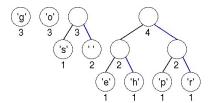
Choosing the first two (minimal) nodes in the priority queue yields another tree with weight 2 as shown below. There are now six trees in the forest of trees that will eventually build an encoding tree.



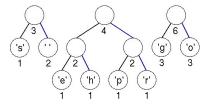
Again we must choose the first two (minimal) nodes in the priority queue. The lowest weight is the 'e'-node/tree with weight equal to 1. There are three trees with weight 2; the one chosen corresponds to an ASCII character because of the way we order the nodes in the priority queue. The new tree has a weight of 3, which will be placed last in the priority queue according to our ordering strategy.



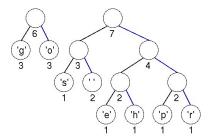
Now there are two trees with weight equal to 2. These are joined into a new tree whose weight is 4. There are four trees left, one whose weight is 4 and three with a weight of 3.



The first two minimal (weight 3) trees in the priority queue are joined into a tree whose weight is 6. There are three trees left.

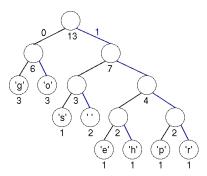


The minimal trees have weights of 3 and 4; these are joined into a tree with weight 7.



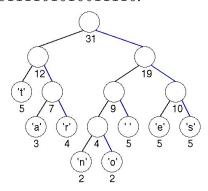
Finally, the last two trees are joined into a final tree whose weight is 13, the sum of the two weights 6 and 7. This tree is the tree we used to illustrate Huffman coding

above. Note that you can easily come up with an alternative optimal tree by using a different ordering strategy to order trees of the same weights. In that case, the bit patterns for each character are different, but the total number of bits used to encode "go go gophers" is the same.



We now show another tree to compress the string "streets are stone stars are not" optimally. To encode "streets are" we would have the following bits:

1110001111011000111101010011110.



Character	Binary code
, ,	101
'a'	010
'e'	110
'n'	1000
'o'	1001
'r'	011
's'	111
<u>'t'</u>	00

It is important to note that you cannot use the tree built for the string "go go gophers" to decode the bitstreams obtained from the encoding of "streets are stone

stars are not" as the encoding is performed using a different tree.

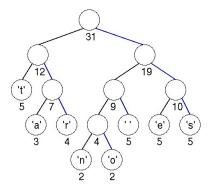
4 Reading Huffman Headers

You must store some initial information in the compressed file that will be used by the decompression/unhuffing program. Basically, you must store the tree that is used to compress the original file. This is because the decompression program needs this exact same tree in order to decode the data. The header information contains:

- The topology of the Huffman coding tree. To store the tree at the beginning of the file, we use a post-order traversal, writing each node visited. When you encounter a leaf node, you write a 1 followed by the ASCII character of the leaf node. When you encounter a non-leaf node, you write a 0. To indicate the end of the Huffman coding tree, we write another 0.
- The total number of characters in the input file, followed by a newline character.

Consider the string "go go gophers", the header information is $"1g1o01s1 \ 01e1h01p1r0000013\n"$, where "\n" is the newline character. The post-order traversal of the Huffman coding tree gives us " $1g1o01s1 \ 01e1h01p1r0000$ ". Another "0" separates the topology from "13", which is the number of characters in the input file.

For the string "streets are stone stars are not", the header information is "1t1a1r001n1o01 $01e1s000031\n$ ". The resulting tree looks like this:



The numbers below each leaf node corresponds to the number of instances each characters appears in the file. For assignment PA10, you do not need to worry

about this.

In these two examples, we use characters 0 and 1 to distinguish between non-leaf and leaf nodes (and 0 to indicate the end of a topology). As there are eight leaf nodes in each of the two examples, there are eight 1's, seven 0's for non-leaf nodes, and another one 0 to indicate that we have reached the end of a topology. This approach used a total of 24 bytes.

To construct a Huffman coding tree from the header information, we make use of a stack. When a 1 is read, we read the corresponding ASCII character and push a node containing the character onto the stack. When a 0 is read, if the stack contains only one element, we have constructed the entire Huffman coding tree. Otherwise, there must be more than one element in the stack. We create a new node, and pop the top two elements off the stack. We make the first element off the stack the right child of the new node, and the second element off the stack the left child of the new node. After that, we push the newly created node onto the stack.