

# Prefix Codes: Equiprobable Words, Unequal Letter Costs

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## Abstract

We consider the following variant of Huffman coding in which the costs of the letters, rather than the probabilities of the words, are non-uniform: Given an alphabet of *unequal-length* letters, find a minimum-average-length prefix-free set of  $n$  codewords over the alphabet. We show new structural properties of such codes, leading to an  $O(n \log^2 r)$  time algorithm for finding them. This new algorithm is simpler and faster than the previously best known  $O(nr \min\{\log n, r\})$  one due to Perl, Garey, and Even [5].

Keywords: Algorithms, Huffman Codes, Prefix Codes, Trees.

## 1 Introduction

The well-known Huffman coding problem [2] is the following: given a sequence of probabilities  $\langle p_1, p_2, \dots, p_n \rangle$ , construct a binary prefix code  $\langle w_1, w_2, \dots, w_n \rangle$  minimizing the expected length  $\sum_i p_i \text{length}(w_i)$ . (A binary prefix code is a set of binary strings, none of which is a prefix of another.)

A natural generalization of the problem is to allow the codewords to be strings over an arbitrary alphabet of  $r \geq 2$  letters. Further, the letters are allowed to have arbitrary non-negative lengths  $\langle c_1 \leq c_2 \leq \dots \leq c_r \rangle$ . The length of a codeword is then the sum of the lengths of its letters. For instance, the “dots and dashes” of Morse code are a variable-length alphabet with length corresponding to transmission time. This generalization of Huffman coding to a variable-length alphabet has been considered by many authors, including Altenkamp and Melhorn [1], and Karp [3]. Apparently no polynomial-time algorithm for it is known, nor is it known to be NP-hard.

In this paper we consider the special case of the general problem in which the codewords are sent with equal probability, i.e., each  $p_i$  equals  $1/n$ . This is a variant of Huffman coding in which the lengths of the letters, rather than the codeword probabilities, are non-uniform. This problem is equivalent to one of finding a tree of a particular type that has minimal external path length among all trees of that

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type with  $n$  leaves. These two equivalent problems were previously considered by Perl, Garey, and Even [5], who gave an  $O(rn \min\{r, \log n\})$ -time algorithm. In what follows we describe a simpler,  $O(n \log^2 r)$ -time algorithm based on new insights into the structure of optimal codes.

In section 2 we define *shallow* trees and their properties and prove that there is a small set of shallow trees that, among themselves, must contain a tree with minimal external path length. In Section 3 we use the properties of shallow trees to develop an algorithm that constructs all of them quickly. The shallow tree with minimal cost will be the one that describes an optimal encoding.

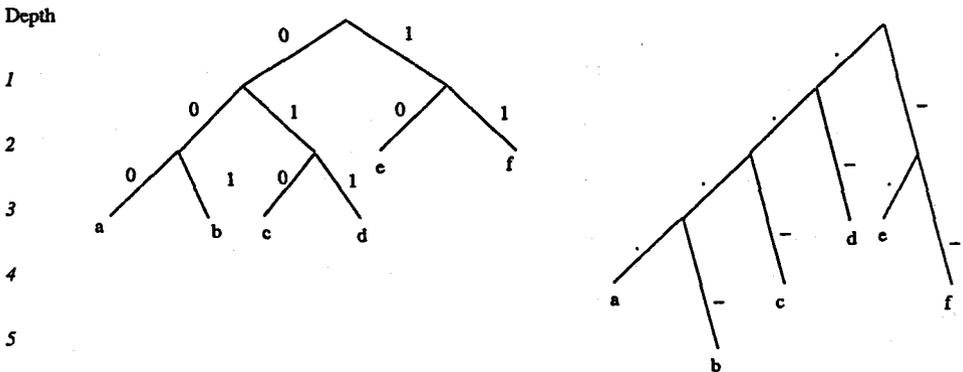


Figure 1: Two Huffman trees for the 6 symbols a,b,c,d,e,f, which all occur with probability  $1/6$ . The tree on the left is the optimal tree that uses the alphabet  $\{0, 1\}$ ,  $length(0) = length(1) = 1$  while the tree on the right is for the alphabet  $\{., -\}$  with  $length(.) = 1$  and  $length(-) = 2$ . The corresponding sets of codewords are

$$a = 000, \quad b = 001, \quad c = 011, \quad d = 011, \quad e = 10, \quad f = 11$$

and

$$a = \dots, \quad b = \dots, \quad c = \dots, \quad d = \dots, \quad e = \dots, \quad f = \dots$$

## 2 Shallow Trees

Fix an instance of the problem, given by the lengths  $\langle c_1 \leq c_2 \leq \dots \leq c_r \rangle$  of the letters and the number  $n$  of (equiprobable and prefix-free) codewords required.

We assume the standard tree representation of prefix codes. The finite words over the alphabet of  $r$  letters correspond to the nodes of the infinite, rooted, ordered  $r$ -ary tree. If an edge in the tree goes from a node to its  $i$ th child, the edge has length  $c_i$  and is labeled with the  $i$ th letter in the alphabet. The labels along the path from the root to a node spell the corresponding word and the length of the path is the length of this word. A prefix code corresponds to a set of nodes none of which is a descendant of another.

In the remainder of the text, the term "tree" refers to any subtree  $T$  containing the root. In any such tree,  $n$  of the leaves will be identified as *terminals*; their

corresponding words form a prefix code. The term “node” refers to any node of the infinite tree, while the term “non-terminal” refers to any node in the subtree  $T$  that is not a terminal. The notation  $\text{child}_i(u)$  denotes the  $i$ th child of node  $u$ .

The cost  $c(T)$  of such a tree is the sum of the depths of the terminals. This is also called the *external weighted path length* of the tree. The goal is to find an optimal (minimum-cost) tree. A *proper* tree is a tree in which every non-terminal has degree at least two. It is easy to see that some optimal tree is proper so we may restrict ourselves to finding an optimal proper tree.

Our basic tool for understanding the structure of optimal trees is a standard swapping argument. For example, in any proper optimal tree, no non-terminal is deeper than any terminal. Otherwise, the terminal and the subtree rooted at the non-terminal could be swapped, decreasing the average depth of the terminals.

Intuitively, this suggests that the optimal, proper trees have the following form for some  $m$ . The non-terminals are some  $m$  shallowest (i.e., least-depth) nodes of the infinite tree, while the terminals are some  $n$  shallowest children of these nodes in the infinite tree. (In general, when we refer to the children of a set of nodes we exclude the nodes in the set itself.) Note that the “ $m$  shallowest” nodes are not necessarily unique. Our algorithm constructs a sequence of such trees, one for each possible number of non-terminals, and returns the best one. Note too that, in the definition of a shallow tree, a node may be non-terminal but still have no children. It is for this reason that we talk of *terminal* and *non-terminal* nodes in place of the more common *internal* nodes and *leaves*.

Formally, a tree  $T$  is *shallow* provided that (i) for any non-terminal  $u$  of  $T$  and any node  $w$  that is not a non-terminal of  $T$ ,  $\text{depth}(u) \leq \text{depth}(w)$  and (ii) for any terminal  $u$  of  $T$  and any node  $w$  of the infinite tree that is not in  $T$  but is a child of a non-terminal of  $T$ ,  $\text{depth}(u) \leq \text{depth}(w)$ .

Shallow trees have the nice property that they are optimal among all trees that share the same number of non-terminals.

**Lemma 1** *Any shallow tree  $T$  satisfies  $c(T) \leq c(T')$  for every proper tree  $T'$  with the same number of non-terminals.*

**Proof.** Fix a shallow tree  $T$ . If there are no proper trees  $T'$  with the same number of non-terminals, the lemma is trivially true. Otherwise, among such trees, consider those that minimize  $c(T')$ . Among these let  $T^*$  be one that maximizes the number of shared non-terminals with  $T$  (where  $T^*$  and  $T$  are considered as finite subsets of the infinite tree).

Suppose for contradiction that the set of non-terminals of  $T$  differs from that of  $T^*$ . Among all non-terminals of  $T$  that are not non-terminals of  $T^*$  let  $u$  be one whose parent is a non-terminal of  $T^*$ . Let  $w$  be any non-terminal of  $T^*$  that is not a non-terminal of  $T$ . Since  $T$  is shallow,  $\text{depth}(u) \leq \text{depth}(w)$ .

In  $T^*$ , node  $u$ , if present, is a terminal. Node  $w$ , on the other hand, has at least two terminal descendants, because  $T^*$  is proper. In  $T^*$ , consider swapping  $u$  and  $w$ 's subtrees. (More specifically, make  $u$  a non-terminal. If  $u$  was a terminal in  $T^*$ , make  $w$  a terminal, otherwise delete  $w$ . For each previous descendant  $x$  of  $w$ , delete  $x$  and add the corresponding descendant  $y$  of  $u$  (as a terminal if  $x$  was a terminal).) The swap doesn't increase  $c(T^*)$  yet increases the number of non-terminals shared with  $T$ . By the choice of  $T^*$ , this is a contradiction.

Thus,  $T$  and  $T^*$  have the same set of non-terminals. Since  $T$  is shallow, clearly  $c(T) \leq c(T^*)$ .  $\square$

As an aside, a similar argument proves something like the converse: if a proper tree is optimal among all trees with the same number of non-terminals, then it is shallow.

Lemma 1 implies that it suffices to consider shallow, proper trees:

**Lemma 2** Let  $m_{\min} = \lceil (n-1)/(r-1) \rceil$ . Let  $\langle T_{m_{\min}}, T_{m_{\min}+1}, T_{m_{\min}+2}, \dots \rangle$  be any sequence of shallow trees such that for each  $m$ ,  $T_m$  has  $m$  non-terminals. Then one of the  $T_m$  is proper and optimal.

**Proof.** Let  $m$  be the minimum number of non-terminals of any optimal tree. Since the optimal tree has degree bounded by  $r$ ,  $m \geq m_{\min}$ . By Lemma 1,  $T_m$  is optimal. Further,  $T_m$  must be proper; otherwise, it would be easy to construct an optimal tree with fewer non-terminals.  $\square$

It is this lemma which is at the core of our algorithm for finding an optimal tree; the algorithm generates such a sequence of shallow trees and returns the one which has minimal cost. The lemma guarantees that this tree will be optimal. The rest of the paper is devoted to examining the properties of shallow trees which enable the identification of a minimal cost shallow tree in  $O(n \log^2 r)$  time.

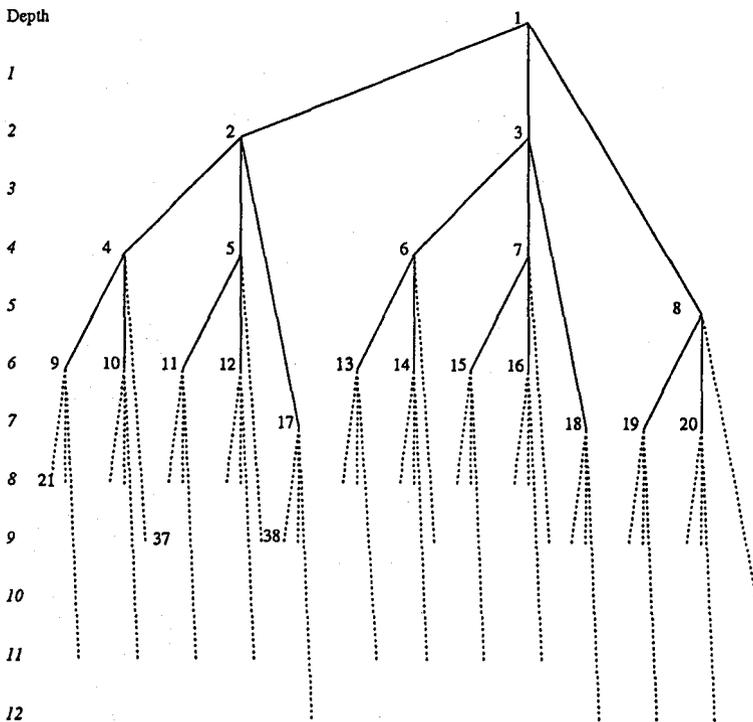


Figure 2: The top of a labelled infinite tree with  $r = 3$ ,  $c_1 = 2$ ,  $c_2 = 2$ , and  $c_3 = 5$ .

## 2.1 Defining the Trees

To determine a unique sequence of trees, order the nodes of the infinite tree as  $1, 2, 3, \dots$ , in order of increasing depth. Break ties arbitrarily, except that if two nodes  $u$  and  $w$  are of equal depth and both are  $i$ th children for some  $i$ , then  $u < w$  iff  $\text{parent}(u) < \text{parent}(w)$ . For the sake of notation, identify each node with its rank in this ordering, so that 1 is the root, 2 is a minimum-depth child of the root, etc. Figure 2 illustrates the top section of such a labelling for  $r = 3$ ,  $c_1 = 2$ ,  $c_2 = 2$ , and  $c_3 = 5$ . These values of  $r$  and  $c_j$  will be the ones we assume in all later examples as well.

For each  $m \geq m_{\min}$ , let  $T_m$  denote the "shallowest" tree with  $m$  non-terminals with respect to the ordering of the nodes. That is, the non-terminal nodes of  $T_m$  are the nodes  $\{1, \dots, m\}$ ; the terminals are the minimum  $n$  nodes among the children of  $\{1, \dots, m\}$  in the infinite tree. Since the ordering of the nodes respects depth, each  $T_m$  is shallow. Figure 3 presents  $T_5$ ,  $T_6$ ,  $T_7$ , and  $T_8$  for  $n = 10$  using the labelling of Figure 2.

By Lemma 2, to find an optimal tree it suffices to consider the set of trees  $\{T_m : T_m \text{ is proper}\}$ .

## 2.2 Relation of Successive Trees

Next we turn our attention to the relation of  $T_{m+1}$  to  $T_m$ .

**Lemma 3** *For  $m \geq m_{\min}$ , the new non-terminal (node  $m + 1$ ) in  $T_{m+1}$  is the minimum terminal of  $T_m$ .*

**Proof.** The parent of  $m+1$  is in  $\{1, \dots, m\}$ , so  $m+1$  is one of the children of  $\{1, \dots, m\}$  in the infinite tree. Among these children,  $m + 1$  is necessarily the minimum. The result follows from the definition of  $T_m$ .  $\square$

**Lemma 4** *For  $m \geq m_{\min}$ , provided the new non-terminal (node  $m + 1$ ) has degree at least one in  $T_{m+1}$ , each terminal of  $T_{m+1}$  is either a child of  $m + 1$  or a terminal of  $T_m$ .*

**Proof.** Let node  $m + 1$  have degree  $d$  in  $T_{m+1}$ . Let the set of children of nodes  $\{1, \dots, m\}$  in the infinite tree be  $\mathcal{C}$ . The terminals of tree  $T_{m+1}$  consist of the minimum  $d$  children of node  $m + 1$  together with the minimum  $n - d$  nodes in  $\mathcal{C} - \{m + 1\}$ . These  $n - d$  nodes, together with node  $m + 1$  (the minimum node in  $\mathcal{C}$ ), are the the  $n - d + 1$  minimum nodes in  $\mathcal{C}$ . If  $d \geq 1$ , then by the definition of  $T_m$ , each such node is a terminal in  $T_m$ .  $\square$

The main significance of Lemmas 3 and 4 is that they will allow an efficient construction of  $T_{m+1}$ . Moreover, they also imply that if  $T_m$  is not proper, neither is any subsequent tree.

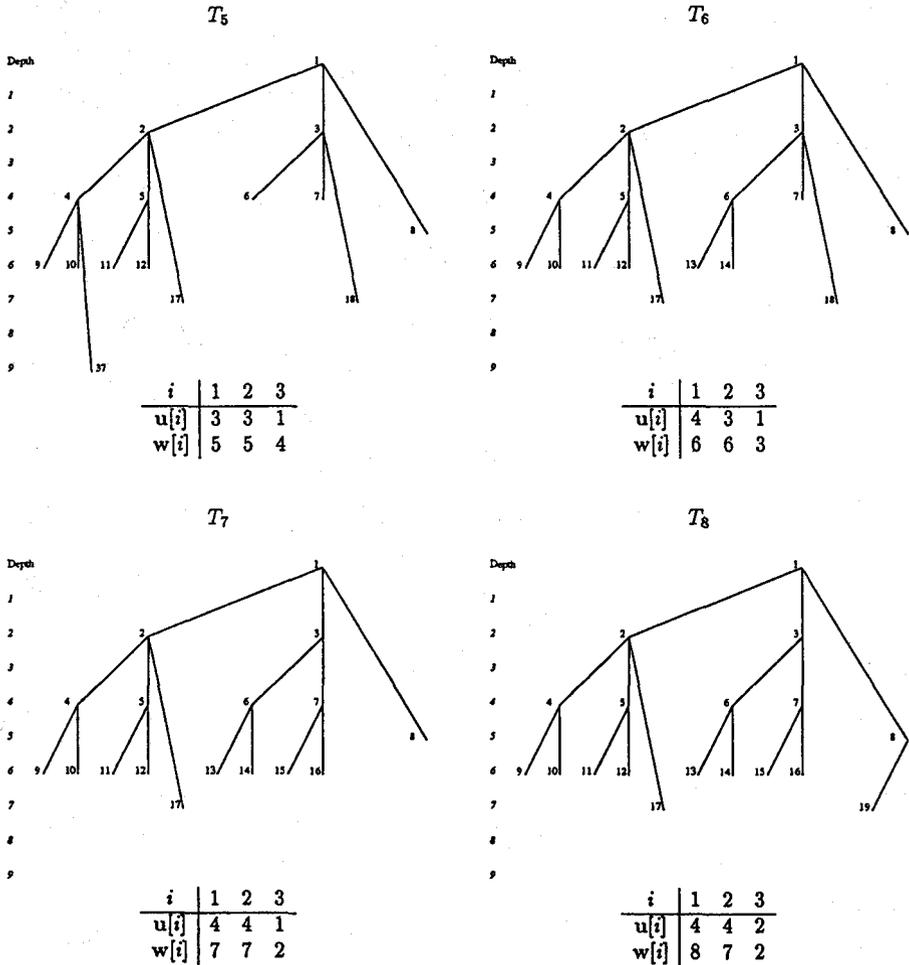


Figure 3: The trees  $T_5$ ,  $T_6$ ,  $T_7$ , and  $T_8$  for  $r = 3$ ,  $c_1 = 2$ ,  $c_2 = 2$ , and  $c_3 = 5$ . The node numbering is that of the previous figure. calculating the external path lengths we find that  $c(T_5) = 60$ ,  $c(T_6) = 59$ ,  $c(T_7) = 60$ , and  $c(T_8) = 62$ .

**Lemma 5** *One of the trees  $\langle T_{m_{\min}}, T_{m_{\min}+1}, \dots, T_{m_{\max}} \rangle$  is optimal and proper, where  $m_{\max} = \min\{m : T_{m+1} \text{ is improper}\} - 1$ .*

**Proof.** By lemma 4, if  $T_m$  is improper, then so is  $T_{m+1}$  — either node  $m + 1$  has degree zero in  $T_{m+1}$  or the non-terminal in  $T_m$  that had degree less than two also has degree less than two in  $T_{m+1}$ . Hence, for  $m \geq m_{\max}$ , tree  $T_m$  is improper. Thus lemma 2 implies that one of the trees  $\langle T_{m_{\min}}, T_{m_{\min}+1}, \dots, T_{m_{\max}} \rangle$  is proper and optimal.  $\square$

For  $n = 10$ ,  $m_{\min} = \lceil \frac{10-1}{3-1} \rceil = 5$  and referring back to Figure 3 shows that  $T_8$  is improper. The lemma then implies that one of  $T_5$ ,  $T_6$ , or  $T_7$  must have minimal external path length. Straight calculation shows that  $T_6$  with  $c(T_6) = 59$  is the optimal one.

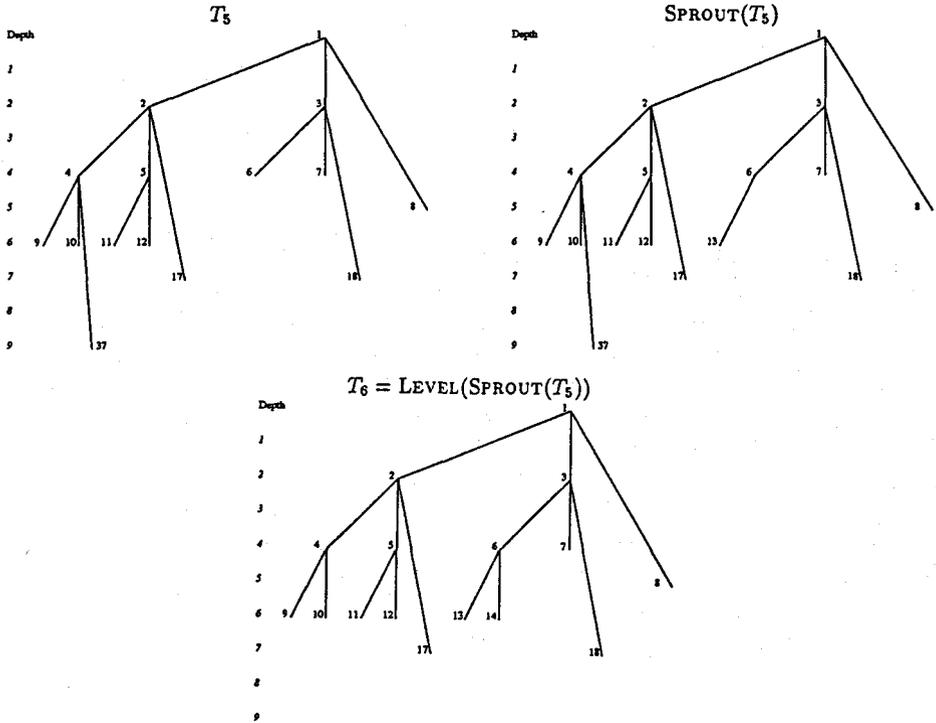


Figure 4: **SPROUTING** and **LEVELING**  $T_5$  yields  $T_6$ .

As an aside, note that a proper tree can have at most  $n - 1$  non-terminals corresponding to every non-terminal having exactly two children. This implies that  $m_{\max} \leq n - 1$ , a fact which will later be needed in the proof of Lemma 9.

### 3 Computing the Trees

Two basic operations are used to compute the trees.

To **SPROUT** a tree is to make its minimum terminal a non-terminal and add the minimum child of this non-terminal as a terminal.

To **LEVEL** a tree is to add  $c$  children of the maximum non-terminal to the tree as terminals and to remove the  $c$  largest terminals in the tree. The  $c$  children are the minimum  $c$  children not yet in the tree, where  $c$  is maximum such that all children added are less than all terminals deleted.

The algorithm computes the initial tree  $T_{m_{\min}}$  then repeatedly **SPROUTS** and **LEVELS** to obtain successive trees until the tree so obtained is not proper. Lemmas 3 and 4 imply that, as long as node  $m + 1$  has degree at least one in  $T_{m+1}$  (it will if  $T_{m+1}$  is proper), **SPROUTING** and **LEVELING**  $T_m$  yields  $T_{m+1}$ . Figure 4 illustrates this operation.

**Observation 6** Let  $m = m_{\max}$ . If node  $m + 1$  has degree one in  $T_{m+1}$  then **SPROUTING** and **LEVELING**  $T_m$  yields tree  $T_{m+1}$ . If node  $m + 1$  has degree zero in  $T_{m+1}$ , then the maximum terminal in  $T_m$  is less than the minimum child of node  $m + 1$  and **SPROUTING** and **LEVELING**  $T_m$  yields a tree in which non-terminal  $m + 1$  has degree one. Hence, the algorithm always correctly identifies  $T_{m_{\max}}$  and terminates correctly, having considered all relevant trees.

To **SPROUT** requires identification and conversion of the minimum terminal of the current tree, whereas to **LEVEL** requires identification and replacement of (no more than  $r$ ) maximum terminals by children of the new non-terminal. One could identify the maximum and minimum terminals in  $O(\log n)$  time by storing all terminals in two standard priority queues (one to detect the minimum, the other to detect the maximum). At most  $r$  terminals are replaced in computing each tree and because  $m_{\max} \leq n - 1$ , only  $O(n)$  trees are computed. This approach yields an  $O(rn \log n)$ -time algorithm.

By a more careful use of the structure of the trees, we improve upon this analysis in two ways. First, we give an amortized analysis showing that in total, only  $O(n \log r)$ , rather than  $O(rn)$ , terminals are replaced. Second, we show how to reduce the number of non-terminals in each priority queue to at most  $r$ . This yields an  $O(n \log^2 r)$ -time algorithm.

Both reductions will be seen to follow from the observation that  $T_m$  must have the following simple structure.

**Lemma 7** In any  $T_m$ , if  $u$  and  $w$  are non-terminals with  $u < w$ , and the  $i$ th child of  $w$  is in the tree, then so is the  $i$ th child of  $u$ . If the  $i$ th child of  $w$  is a non-terminal, then so is the  $i$ th child of  $u$ .

**Proof.** Straightforward from the definition of  $T_m$  and the condition on breaking ties in ordering the nodes.  $\square$

**Corollary 8** Node  $m$  has minimal degree among all non-terminals in  $T_m$ .

### 3.1 Only $O(n \log r)$ Replacements Total

The number of terminals replaced while obtaining  $T_m$  from  $T_{m-1}$  is at most the degree of non-terminal  $m$  in  $T_m$ . Although this degree might be  $r$  for many  $m$ , the sum of these degrees is  $O(n \log r)$ :

**Lemma 9** Let  $d_m$  be the degree of non-terminal  $m$  in tree  $T_m$ . Then  $\sum_m d_m$  is  $O(n \log r)$ .

**Proof.** By Lemma 7, within  $T_m$ , node  $m$  is the lowest-degree non-terminal. The sum of the  $m$  non-terminals' degrees is  $(m + n - 1)$ . Thus,  $d_m$  is at most the average  $(m + n - 1)/m = 1 + (n - 1)(1/m)$ .

$$\sum_{m=m_{\min}}^{m_{\max}} d_m \leq (m_{\max} - m_{\min} + 1) + (n - 1) \sum_{m=m_{\min}}^{m_{\max}} 1/m = O(m_{\max} - m_{\min} + n \log(m_{\max}/m_{\min})).$$

The result follows from  $m_{\min} = \lceil \frac{n-1}{r-1} \rceil$  and  $m_{\max} \leq n - 1$ .  $\square$

### 3.2 Limiting the Relevant Terminals

To reduce the number of terminals that must be considered in finding the minimum and maximum terminals, consider, for each  $i$ , the terminals which are  $i$ th children.

**Lemma 10** *In any  $T_m$ , for any  $i$ , the set of non-terminals whose  $i$ th children are terminals is of the form  $\{u_i, u_i+1, \dots, w_i\}$  for some  $u_i$  and  $w_i$ . The minimum among terminals that are  $i$ th children is  $\text{child}_i(u_i)$  (the  $i$ th child of  $u_i$ ). The maximum among these terminals is  $\text{child}_i(w_i)$ .*

**Proof.** A straightforward consequence of Lemma 7. □

Figure 3 presents  $u_i$  and  $w_i$  for the trees  $T_5, T_6, T_7$ , and  $T_8$  when  $n = 10$ .

This lemma implies that the minimum terminal in  $T_m$  is the minimum among  $\{\text{child}_i(u_i), i = 1, \dots, r\}$ . The minimum terminal in  $T$  can be found by storing only these  $r$  particular children in a priority queue in place of storing all  $n$  terminals in a priority queue. This reduces the cost of finding the minimum from  $O(\log n)$  to  $O(\log r)$ . Similarly the maximum terminal can be found in  $O(\log r)$  time by storing  $\{\text{child}_i(w_i), i = 1, \dots, r\}$ , in a priority queue.

### 3.3 The Algorithm in Detail

The full algorithm has two distinct steps. The first constructs the base tree  $T_{m_{\min}}$ . The second starts with  $T_{m_{\min}}$  and, by **SPROUTING** and **LEVELING**, iteratively constructs the sequence of shallow trees

$$\langle T_{m_{\min}}, T_{m_{\min}+1}, T_{m_{\min}+2}, \dots, T_{m_{\max}} \rangle$$

and returns one which has smallest external path length.  $T_{m_{\max}}$  is the last proper tree in the sequence. Lemma 5 guarantees that this sequence contains an optimal tree so the tree that the algorithm returns is an optimal tree. We now describe how to implement the first part of the algorithm in  $O(n \log r)$  time and the second in  $O(n \log^2 r)$  time; the full algorithm will therefore run in  $O(n \log^2 r)$  time.

The skeleton of the final algorithm is shown in Figure 5. Procedure **CREATE- $T_{m_{\min}}$**  creates tree  $T_{m_{\min}}$ , the variable **C** contains the external path length of current tree  $T_m$  and **mDeg** contains the degree of node  $m$  in tree  $T_m$ . As presented, the algorithm computes only the cost of an optimal tree. It is easily modified to compute the actual tree. Note that to check that the current tree  $T_m$  is proper, by Observation 6 and Corollary 8, it suffices to check that non-terminal  $m$  has degree at least two.

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COMPUTE - TREES( $(c_1, c_2, \dots, c_r), n$ )
1  CREATE- $T_{m_{\min}}$ ;
2  WHILE (mDeg  $\geq$  2) DO
    — Compute  $T_{m+1}$  from  $T_m$  —
3      SPROUT( $T$ )
4      LEVEL( $T$ )
5       $C_{\min} \leftarrow \min\{C, C_{\min}\}$ 
6  RETURN  $C_{\min}$ 

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Figure 5: Algorithm to Find An Optimal Variable-Length Prefix Code

Recall that the nodes of the infinite tree are labelled in order of increasing depth with ties broken arbitrarily except for the requirement that if  $u$  and  $v$  are both of equal depth and both are  $i$ th children of their respective parents, then  $u < v$  iff  $\text{parent}(u) < \text{parent}(v)$ . Depending upon  $c_1, c_2, \dots, c_r$ , there may be many such labellings. The algorithm to be presented breaks ties using the specific rule that if  $u$  and  $v$  have the same depth,  $u = \text{child}_i(u')$ ,  $v = \text{child}_j(v')$ , and  $u' < v'$  then  $u < v$ . If  $u' = v'$  then  $u < v$  iff  $i < j$  (this can only occur if  $c_i = c_j$ ). Figure 2 illustrates this labelling for  $r = 3$ ,  $c_1 = 2$ ,  $c_2 = 2$ , and  $c_3 = 5$ . Fixing the labelling also fixes the shallow trees. Figure 3 illustrates the shallow trees with 10 non-terminals for these  $r$  and  $c$  values.

A tree  $T_m$  can be fully represented by the following data structures:

$N$  — The number of terminals.

$m$  — The number of non-terminals. Also the rank of the maximum non-terminal.

$C$  — The sum of the depths of the terminals.

$m\text{Deg}$  — The degree of non-terminal  $m$ .

$D[u]$  — The depth of each non-terminal  $u$ .

$u[i]$  — The rank of the minimum non-terminal (if any) whose  $i$ th child is a terminal ( $1 \leq i \leq r$ ).

$w[i]$  — The rank of the maximum non-terminal (if any) whose  $i$ th child is a terminal ( $1 \leq i \leq r$ ).

**low-queue** — A priority queue for finding the minimum terminal.

Contains  $\{\text{child}_i(u[i]) : \text{appropriate } i\}$ .

**high-queue** — A priority queue for finding the maximum terminal.

Contains  $\{\text{child}_i(w[i]) : \text{appropriate } i\}$ .

For an example refer back to figure 3. Tree  $T_6$  has

$$N = 10, \quad C = 59, \quad m\text{Deg} = 2,$$

$$D[1] = 0, \quad D[2] = 2, \quad D[3] = 3, \quad D[4] = 4, \quad D[5] = 4, \quad D[6] = 4,$$

$$u[1] = 4, \quad u[2] = 3, \quad u[3] = 1, \quad w[1] = 6, \quad w[2] = 6, \quad w[3] = 3$$

$$\text{low-queue} = \{\text{child}_1(4), \text{child}_2(3), \text{child}_3(1)\},$$

$$\text{high-queue} = \{\text{child}_1(6), \text{child}_2(6), \text{child}_3(3)\}.$$

Generally, the algorithm knows the ranks of the non-terminals in the current tree, but not the terminals. The relevant terminals are referenced via their parents, generally one of the non-terminals  $m$ ,  $u[i]$ , or  $w[i]$ . The order of any two terminals,  $\text{child}_i(u)$  and  $\text{child}_j(w)$ , is determined by considering their respective depths  $D[u] + c_i$  and  $D[w] + c_j$ . If one terminal has lesser depth, that terminal has the smaller label. Otherwise, ties are broken in the manner described at the beginning of this

subsection. If  $u < w$  then  $\text{child}_i(u) < \text{child}_j(w)$ . Otherwise  $u = w$  and  $\text{child}_i(u) < \text{child}_j(w)$  for  $i < j$ . This is how the  $r$  terminals are ordered within each of the priority queues. Inserting and deleting an item from each of these queues can be done in  $O(\log r)$  time. Since the minimum terminal in low-queue is the minimum terminal in  $T_m$  this gives a method of finding the minimum terminal in  $T_m$  in  $O(\log r)$  time. Similarly, using high-queue permits finding the largest terminal in  $T_m$  in  $O(\log r)$  time.

The reason for using this indirect method of comparing terminals in place of explicitly calculating and comparing the terminals' labels is that the terminals might have rank higher than  $\Theta(n \log^2 r)$ . Explicitly calculating the labels could therefore cost more than all the rest of the algorithm, effectively destroying the  $O(n \log^2 r)$  running time of the algorithm.

In the definitions of the priority queues, "appropriate" values of  $i$  are those for which some non-terminal has an  $i$ th child in the current tree. Note that Lemmas 4 and 7 imply that if, for some  $i$  and  $T_m$ , no non-terminal has an  $i$ th child in  $T_m$ , then no non-terminal has an  $i$ th child in  $T_{m+1}$ . Subsequently,  $i$ th children of non-terminals will always be "inappropriate" and need not be considered.

Corresponding to each variable  $u[i]$  (resp.  $w[i]$ ) is a terminal in low-queue (resp. high-queue). When such a variable is changed, the priority queues are updated by the following routine:

UPDATE-QS( $T, i$ )

- 1 IF ( $u[i] \leq w[i]$ ) THEN
- 2     Update  $\text{child}_i(u[i])$  in low-queue and  $\text{child}_i(w[i])$  in high-queue  
to maintain the queues' invariants.
- 3 ELSE Delete both nodes from their respective queues.

Line 2 replaces the old value of  $\text{child}_i(u[i])$  in low-queue ( $\text{child}_i(w[i])$  in high-queue) by its new value. Line 3 will only be executed if  $\text{child}_i(u[i]) > \text{child}_i(w[i])$  which will only happen if the tree no longer contains any  $i$ th child as a terminal.

The routines SPROUT and LEVEL are shown in Figure 6.

SPROUT( $T$ )

—Make the minimum terminal a non-terminal—

- 1  $m \leftarrow m + 1$ ;
  - 2 Let  $\text{child}_i(u[i])$  be the minimum terminal in low-queue.
  - 3  $D[m] \leftarrow D[u[i]] + c_i$ ;  $u[i] \leftarrow u[i] + 1$ ; UPDATE-QS( $T, i$ )
  - 4  $C \leftarrow C - D[m]$ ;  $m\text{Deg} \leftarrow 0$ ;
- Add smallest child as a terminal—
- 5 ADD-TERMINAL( $T$ )

LEVEL( $T$ )

- 1 WHILE ( $m\text{Deg} < r$  and  $\text{child}_{m\text{Deg}+1}(m)$  is less than the max. terminal  $\text{child}_i(w[i])$  in high-queue) DO
  - 2 ADD-TERMINAL( $T$ )
- Delete the maximum terminal—
- 3  $C \leftarrow C - (D[w[i]] + c_i)$
  - 4  $w[i] \leftarrow w[i] - 1$ ; UPDATE-QS( $T, i$ )

ADD-TERMINAL( $T$ )

- 1  $m\text{Deg} \leftarrow m\text{Deg} + 1$ ;  $C \leftarrow C + D[m] + c_{m\text{Deg}}$ ;
- 2  $w[m\text{Deg}] \leftarrow m$ ; UPDATE-QS( $T, m\text{Deg}$ )

Figure 6: The Operations SPROUT and LEVEL.

**Construction of the First Trees.** Tree  $T_{m_{\min}}$  has a simple structure. Its non-terminals are the nodes  $\{1, 2, \dots, m_{\min}\}$ . Its terminals are the  $n$  shallowest children of nodes  $\{1, 2, \dots, m_{\min}\}$ .

To construct  $T_{m_{\min}}$  we assume that  $n > r$ , otherwise  $T_{m_{\min}}$  is simply the root and its first  $n$  children. For  $1 \leq m < m_{\min}$ , define  $T_m$  to be the tree with non-terminals  $\{1, \dots, m\}$  and all of the  $(r-1)m + 1$  children of  $\{1, \dots, m\}$  as terminals. The proof of Lemma 3 generalizes easily to these trees; node  $m + 1$  is the minimum terminal of  $T_m$ .

The tree  $T_1$  is easy to construct. It is the tree with 1 root and  $r$  children. Inductively construct the tree  $T_m$  from the tree  $T_{m-1}$ ,  $m < m_{\min} - 1$  as follows: find the minimum terminal in  $T_m$  by taking the minimum terminal out of low-queue. Label this node  $m$ , make it a non-terminal, and add all of its children to  $T_m$  as terminals.

Finally, construct  $T_{m_{\min}}$  from  $T_{m_{\min}-1}$  by making the lowest terminal of  $T_{m_{\min}-1}$  into node  $m_{\min}$ . Add the  $n - (r-1)(m_{\min} - 1)$  minimum children of node  $m_{\min}$  as terminals bringing the total number of terminals in the current tree to  $n$ . Level the resulting tree.

Since only  $O(n/r)$  trees are constructed while computing  $T_{m_{\min}}$ , and each tree can be constructed from the previous tree in  $O(r \log r)$  time, the time required to compute  $T_{m_{\min}}$  is  $O(n \log r)$ . (If desired, the time for each tree  $T_m$  with  $m < m_{\min}$  can be reduced to  $O(\log r)$ , because maximum terminals are not replaced in constructing such a tree.)

**Construction of the Remaining Trees.** The algorithm constructs the sequence of trees

$$\langle T_{m_{\min}}, T_{m_{\min}+1}, T_{m_{\min}+2}, \dots, T_{m_{\max}} \rangle$$

as described previously. Tree  $T_m$  is found by SPROUTING and then LEVELING its predecessor  $T_{m-1}$ . The cost is  $O(d_m \log r)$  time, where  $d_m$  is the degree of the new non-terminal  $m$  in  $T_m$ . By Lemma 9 this part of the algorithm runs in  $O((\sum_m d_m) \log r) = O(n \log^2 r)$  time.

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