A Work-Efficient Algorithm for Constructing Huffman Codes

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Abstract

We present an algorithm for parallel construction of Huffman codes in \( O(\frac{1}{p} \log p) \) time with \( p \) processors, improving the previous result of Levcopoulos and Przytycka.

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1 Introduction

A Huffman code for an alphabet $a_1, a_2, \ldots, a_n$ with weights $w_1, w_2, \ldots, w_n$ is a prefix code that minimizes weighted codeword length, defined as $\sum_{i=1}^{n} w_i l_i$, where $l_i$ is the length of the $i$th codeword.

The best known parallel algorithm [LP95] for this problem that uses $n$ processors is due to Larmore and Przytycka and runs in $O(\sqrt{n \log n})$ time. Their algorithm is based on reducing the problem of constructing a Huffman code to the concave least weight subsequence problem. Levcopoulos and Przytycka [LevPrz95] have presented an algorithm for the efficient construction of Huffman trees with the sublinear number of processors. Their algorithm runs in $O(\frac{n}{\sqrt{p \log(p+1)}} \log^2 p + \log(n - \sqrt{p \log p}))$ with $p$ processors. However, we observe that the algorithm of [LevPrz95] contains a technical flaw since construction of left-justified trees is not combined with the greedy Huffman algorithm in an appropriate way.

In this paper we further improve the result of Levcopoulos and Przytycka and describe an algorithm that works in $O(\frac{n}{\sqrt{p} \log p})$ time with $p$ processors. Our algorithm uses a combination of ideas inspired by several prior approaches to the problem. In particular, we combine the methods used for the construction of almost-optimal Huffman codes (see [KP96], [BKN02]), the greedy paradigm of Huffman’s algorithm [H51], and the CLMS approach from [LP95].

2 Algorithm Description

The technique for the construction of approximate Huffman codes in parallel given in [KP96, BKN02] is to divide elements $a_j$ into classes $W_j$, $j = m, \ldots, 1$ such that $W_j = \{ w_i | 1/2^j \leq w_i < 1/2^{j-1} \}$. This ensures that the sum of the weights of any two elements of a given class is greater than the weight of any other element of that same class. Basically, Huffman’s algorithm can be reformulated as follows. Let $W_k$ be the class which contains the smallest element, and initialize $W_k$ to be empty. For each $j$, in descending order, consecutive pairs of elements of the merged list $W_j + W_j$ are combined and stored as a list $W_{j-1}$, then $W_{j-1}$ and $W_{j-1}$ are merged. It follows that elements from the same class can be processed in parallel. See [KP96] for a detailed description of this algorithm that also considers the case of an odd number of elements in $W_j + W_j$. A limitation of this approach is that the number of parallel steps is proportional to the number of different classes $W_j$.

In this paper we will show how we can reduce the number of steps for
the classes of small size. A binary tree $T$ is called left-justified if the height of every leaf of the left subtree of $T$ is greater than or equal to the height of the right subtree of $T$, and if every proper subtree of $T$ is left-justified. The following theorem was proven in [LP95].

**Theorem 1** A left-justified Huffman tree $T$ for elements $a_1, a_2, \ldots, a_p$ can be constructed in $O(\sqrt{p \log p})$ time with $p$ processors.

In general, a tree constructed by Huffman’s original algorithm is not left-justified. We call a tree constructed by Huffman’s algorithm a greedy Huffman tree.

To combine the algorithm of [LP95] with the approach of [KP96] and [BKN02] we need the following theorem.

**Theorem 2** A greedy Huffman tree $T$ for elements $a_1, a_2, \ldots, a_p$ can be constructed in $O(\sqrt{p \log p})$ time with $p$ processors.

We begin by describing Huffman’s algorithm by formulating of a lemma.

Let $b_1, \ldots, b_{n-1}$ be the internal nodes of the greedy Huffman tree, and let $v_i$ be the weight of $b_i$, defined to be the sum of the weights of the leaves of the subtree rooted at $b_i$. We index the internal nodes in the order in which Huffman’s algorithm produces them; thus $v_1 \leq v_2 \leq \ldots v_{n-1}$.

Let $F_r$ be the greedy Huffman forest of $r$ roots, i.e., the forest obtained after $n-r$ steps of Huffman’s algorithm. Note that $F_n$ is the forest consisting of $n$ singleton nodes, while $F_1$ is greedy Huffman tree. The leaves of each $F_r$ are the original items. Huffman’s algorithm constructs $F_{n-r}$ by combining the two least weight trees of $F_{n-r-1}$ into a single tree, creating the new root $b_j$.

We will use the following tie-breaking rule. We assign a t-value to every node in a Huffman tree. The t-value of the leaf $a_i$ will be $2^i$, and the t-value of an internal node will be the sum of t-values of its children. Thus all nodes have distinct t-values. In case two nodes have the same weight, the one with the smaller t-value will be taken to be the smaller one.

This method reduces the problem with ties to the problem with no ties, by simply changing the weight of the $a_i$ to $w_i + \epsilon 2^i$, where $\epsilon$ is some very small positive constant.

This tie-breaking scheme can be implemented at a cost of $O(1)$ per comparison, as follows. For each node, keep track of the index of the largest leaf in the subtree rooted at that node, and call that the dominant t-value of that node. If it becomes necessary to compare the t-values of two nodes, simply compare their dominant t-values. Since the subtrees are disjoint, they will have distinct dominant t-values.
Lemma 1

1. \( F_n \) consists of the singleton trees \( \{a_i\} \).

2. For \( j > 1 \), one of the roots of \( F_{n-j} \) is \( b_j \). Furthermore, \( F_{n-j} - b_j = F_{n-j+1} \), and the children of \( b_j \) are the roots of the two smallest weight trees in \( F_{n-j+1} \).

3. If \( k < \ell < j \) and \( b_k \) is a root of \( F_{n-j} \), then \( b_\ell \) is a root of \( F_{n-j} \).

4. \( F_{n-j} \) has leaves \( a_1, \ldots, a_n \), internal nodes \( b_1, \ldots, b_j \), and roots \( b_{i+1}, \ldots, b_j \), \( a_{j-i+1}, \ldots, a_n \), for some \( i < j \).

5. Let \( B_{n-j} \) be the set of children of non-leaf roots of \( F_{n-j} \). Then the sum of the weights of any two elements of \( B_{n-j} \) is greater than the weight of any other element of \( B_{n-j} \).

Proof:

Part 1 and Part 2 are the well-known observations which justify Huffman’s algorithm. Part 3 is proved by contradiction, as follows. If \( k < \ell < j \), and \( b_k \) is a root of \( F_{n-j} \) and \( b_\ell \) is not, then \( F_{n-j} \) could be improved by exchanging \( b_k \) and \( b_\ell \), since \( v_k < v_\ell \). Part 4 follows from Part 1, Part 2, Part 3, and a simple computation. Part 5 is proved by contradiction in the same manner as Part 3; if it were false, the smallest non-leaf root of \( F_{n-j} \) could be exchanged with the largest member of \( B_{n-j} \), improving \( F_{n-j} \).

We now summarize the reduction given in [LP95]. For \( 0 < m < n \), let \( S_m = \sum_{i=1}^{m} v_i \), the sum of the weights of the first \( m \) symbols. Let \( G \) be the weighted directed acyclic graph whose nodes are the integers \( 0, 1, \ldots, n - 1 \), and whose edges are the pairs \( (i, j) \) such that \( i < j \) and \( 2j - i \leq n \), where the edge \( (i, j) \) has weight \( S_{2j-i} \). Define \( b_{00} = 0 \), and for any \( 0 < j < n \), let \( b_{0j} \) be defined to be the next-to-the-last node in the minimum weight path from 0 to \( j \). Let \( f_j \) be the total weight of that minimum weight path. More formally, \( f_0 = 0 \) and, for \( j > 0 \), \( f_j = \min_{i < j} \{ f_i + S_{2j-i} : i < j \} \), and \( b_{0j} \) is that choice of \( i \) for which the minimum value of \( f_j \) is achieved.

We will use the following lemmas from [LP95].

Lemma 2 The graph \( G \) has the concave Monge property.

Lemma 3 We can find minimum weight paths from 0 to \( j \) for \( 1 \leq j \leq n \) in \( O(\sqrt{n}) \) time with \( n \) processors.
Algorithm CLWS described in the paper [LP] finds minimum weight paths for all j. According to theorem 2.7 of [LP] it works in $O(n \log n/m + n^2/m + nm \log n/p)$ time with p processors for any $0 < m \leq n$. By setting $p = n$ and $m = \sqrt{n}$ we get the result of the Lemma.

Lemma 4

1. The weighted path length of $F_{n-j}$ is equal to $f_j$.
2. For $0 < j < n$, $f_j - f_{j-1} = v_j$.
3. If $back_j = i$, then the roots of $F_{n-j}$ are $b_{i+1}, \ldots, b_j, a_{2j-i+1}, \ldots, a_n$.
4. For $0 < j < n$, $back_{j-1} \leq back_j \leq back_{j-1} + 2$. Let $i = back_{j-1}$. Then
   
   (a) If $back_j = i$ then the two children of $b_j$ are $a_{2j-i+1}$ and $a_{2j-i}$.
   (b) If $back_j = i + 1$ then the two children of $b_j$ are $b_{i+1}$ and $a_{2j-i-1}$.
   (c) If $back_j = i + 2$ then the two children of $b_j$ are $b_{i+1}$ and $b_{i+2}$.

Proof:

We first prove Part 1 by strong induction. For $j = 0$, it is trivial, so assume $j > 0$. Part 3 of Lemma 1 allows us to choose $i < j$ such that $b_k$ is a root of $F_{n-j}$ if and only if $i < k \leq j$. $F_{n-j}$ is then obtained from $F_{n-j}$ by deleting the roots $b_k$ for $i < k \leq j$. Since $F_{n-j}$ must have $n - j$ roots, of which $j - i$ are the $b_k$, it must have $n - 2j + i$ roots which are original items. It follows from Part 3 of Lemma 1 that removal of $b_k$ for $i < k \leq j$ causes all leaves which are not roots of $F_{n-i}$ to move up one level in the tree. Hence the weighted path length is decreased by the sum of the weights of those leaves, which is $S_{2j-i}$, which is also the weight of the edge $(i, j)$ in $G$.

By the inductive hypothesis, the weighted path length of $F_i$ is $f_i$, thus the weighted path length of $F_j$ is $f_i + S_{2j-i} \geq f_j$.

To prove that $f_j$ is also an upper bound for the weighted path length of $F_j$, let $i = back_j$. We consider two cases. If $i = 0$, then $f_j = S_{2j-i}$, which is the weighted path length of the forest obtained by combining $a_1, \ldots, a_{2j}$ in pairs, resulting in a forest with $n - j$ roots and weighted path length $f_j$.

Now suppose $i > 0$. By the inductive hypothesis, the weighted path length of $F_i$ is $f_i$. Let $m = back_i$. We know that $2i - m \leq 2j - i$, since otherwise a smaller weight path from 0 to $j$ could be obtained by replacing the edges $(m, i)$ and $(i, j)$ by the edges $(m, i-1)$ and $(i-1, j)$. The forest obtained from $F_i$ by combining the roots $b_m + 1, \ldots, b_i, a_{2i-m+1}, \ldots, a_{2j-i}$ in pairs then has $n - j$ roots and weighted path length $f_j$. 
Part 2 follows from Part 1 and Part 2 of Lemma 1, since the difference between the weighted path lengths of \( F_{n-j} \) and \( F_{n-j+1} \) is the weight of the subtree rooted at \( b_j \).

Part 3 follows from the above discussion and from Part 4 of Lemma 1.

We now prove Part 4. The two children of \( b_j \) are the two roots of \( F_{j-1} \) of smallest weight, which must be in the set \( \{ b_{i+1}, b_{i+2}, a_{2j-i-1}, a_{2j-i} \} \). There are three possibilities, giving us the three cases.

This completes the proof of Lemma 4. □

Theorem 2 follows from Lemma 3 and Part 4 of Lemma 4.

Now we finish the description of our algorithm. Suppose that classes \( W_0, W_1, \ldots, W_t \) are already processed. If the number of elements in \( W_j \) exceeds \( p \), we construct \( W_{j-1} \) as described above. If \( |W_j| < p \) then classes \( W_0, W_1, \ldots, W_j \) such that \( |W_0| + |W_1| + \ldots + |W_j| \) ≤ \( p \) and \( |W_0| + |W_1| + \ldots + |W_j| > p \) are considered. Using theorem 2 we can construct a greedy Huffman tree \( T' \) for elements of classes \( W_0, W_1, \ldots, W_j \).

We consider the sets \( N \) of all nodes \( S \) in \( T' \), such that the weight of at least one son of \( S \) is in \( \left[ 2^{j-1}, 2^j \right) \). It is easy to see that the set \( N \) can be used instead of the set \( W_{i-1} \) from the previous algorithm.

There are at most \( n/p \) large classes with more than \( p \) elements. Therefore there are at most \( n/p \) groups of small classes. The total time to process all groups of small classes can therefore be limited by \( O((n/p) \log n) \). The time necessary to process a large class is \( O(\log \log p \cdot |W_i|/p) \). Hence the total time for processing all large classes is less than \( O(n/p \log \log p) \).

Therefore the running time of the modified algorithm is \( O((n/\sqrt{p}) \log p) \).

3 Open Problems

It is an open problem whether there exists an \( O((n/\sqrt{p})) \) \( p \)-processor algorithm for Huffman coding. Such an algorithm can perhaps be constructed with a faster CLWS algorithm for a Monge graph with limited edge lengths.

Another open problem is the construction of faster parallel algorithms based on the CLWS approach, that would use the special properties of Monge graphs corresponding to Huffman codes (for instance, the fact that there are only a linear number of distinct weights on the edges.)

References


