



Experiment No.8
Implementation Huffman encoding(Tree) using Linked List
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Experiment No. 8: Huffman encoding (Tree) using Linked list

Aim: Implementation Huffman encoding (Tree) using Linked list

Objective:

The objective of this experiment is to implement and evaluate the Huffman coding algorithm using a linked list data structure to assess its efficiency in data compression.

Theory:

Huffman Coding is a widely used data compression algorithm that assigns variable-length codes to symbols in a dataset. It is a lossless compression technique that is based on the frequency of symbols in the data. The key idea behind Huffman Coding is to assign shorter codes to more frequent symbols and longer codes to less frequent symbols, thus optimizing the compression ratio.

Traditional Huffman Coding:



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1. **Frequency Calculation:** In traditional Huffman Coding, the first step is to calculate the frequency of each symbol in the dataset.
2. **Huffman Tree Construction:** Next, a binary tree called the Huffman tree is constructed. This tree is built using a greedy algorithm, where the two least frequent symbols are merged into a new node, and this process continues until a single tree is formed.
3. **Code Assignment:** The codes are assigned to symbols by traversing the Huffman tree. Left branches are assigned the binary digit "0," and right branches are assigned the binary digit "1." The codes are assigned such that no code is a prefix of another, ensuring unambiguous decoding.
4. **Compression:** The original data is then encoded using the Huffman codes, resulting in a compressed representation of the data.

Adaptive Huffman Coding:

1. **Initial Tree:** In Adaptive Huffman Coding, an initial tree is created with a predefined structure that includes a special symbol for escape. The escape symbol is used to signal that a new symbol is being introduced.
2. **Tree Update :**As symbols are encountered in the data, the tree is updated dynamically. When a symbol is encountered for the first time, it is added as a leaf node to the tree, and the escape symbol is used to navigate to its parent. This process ensures that new symbols can be encoded even if they were not present when the tree was initially created.
3. **Code Assignment:** The codes are assigned dynamically as the tree changes. More frequent symbols have shorter codes, and less frequent symbols have longer codes.



4. Compression: The data is encoded using the dynamically updated Huffman tree, resulting in compressed data. The tree is updated as the data is processed.

Algorithm

Adaptive Huffman Coding algorithm using the FGK (Faller-Gallager-Knuth) variant:

Step1:- Initialization:

- Create an initial tree with the escape symbol and any predefined symbols.
- Set a pointer to the escape symbol.

Step 2:- Data Processing:

- Start processing symbols from the input data stream.
- If a symbol is encountered for the first time, add it as a new leaf node to the tree.

Update the tree structure as needed to maintain the prefix property.

- Use the escape symbol as a way to navigate to the parent node of the new symbol.
- After each symbol is processed, the tree is adjusted to maintain the prefix property and optimal code lengths.

Step 3:- Code Assignment:

- Traverse the tree to assign variable-length codes to the symbols dynamically.
- More frequent symbols have shorter codes, and less frequent symbols have longer codes.

Step 4:- Compression:

- Encode the input data using the dynamically updated Huffman tree.
- The compressed data consists of the variable-length codes for each symbol.

Adaptive Huffman Coding adapts to the data as it is processed, allowing for efficient encoding of symbols even if their frequencies change over time. This adaptability



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makes it a suitable choice for scenarios where the data distribution is not known in advance or may change dynamically.

Code:

```
#include <stdio.h>
#include <stdlib.h>

// Define the structure for a Huffman Tree Node
struct HuffmanNode {
    char data;
    unsigned frequency;
    struct HuffmanNode* left;
    struct HuffmanNode* right;
};

// Define a structure for a Min Heap Node
struct MinHeapNode {
    struct HuffmanNode* node;
    struct MinHeapNode* next;
};

// Define a structure for a Min Heap
struct MinHeap {
    struct MinHeapNode* head;
};

// Function to create a new Min Heap Node
struct MinHeapNode* createMinHeapNode(struct HuffmanNode* node) {
    struct MinHeapNode* newNode = (struct MinHeapNode*)malloc(sizeof(struct MinHeapNode));
    newNode->node = node;
    newNode->next = NULL;
```



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```
        return newNode;
    }

// Function to create a new Min Heap
struct MinHeap* createMinHeap() {
    struct MinHeap* minHeap = (struct MinHeap*)malloc(sizeof(struct MinHeap));
    minHeap->head = NULL;
    return minHeap;
}

// Function to insert a Min Heap Node
void insertMinHeap(struct MinHeap* minHeap, struct MinHeapNode* node) {
    if (minHeap->head == NULL) {
        minHeap->head = node;
    } else {
        if (node->frequency < minHeap->head->frequency) {
            node->next = minHeap->head;
            minHeap->head = node;
        } else {
            struct MinHeapNode* current = minHeap->head;
            while (current->next != NULL && current->next->frequency < node->frequency) {
                current = current->next;
            }
            node->next = current->next;
            current->next = node;
        }
    }
}

// Function to extract the minimum node from the Min Heap
```



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```
struct MinHeapNode* extractMin(struct MinHeap* minHeap) {
    struct MinHeapNode* temp = minHeap->head;
    minHeap->head = minHeap->head->next;
    return temp;
}

// Function to build the Huffman Tree
struct HuffmanNode* buildHuffmanTree(char data[], int frequency[], int n) {
    struct HuffmanNode *left, *right, *top;

    // Create a Min Heap and insert all characters into it
    struct MinHeap* minHeap = createMinHeap();
    for (int i = 0; i < n; ++i) {
        struct HuffmanNode* node = (struct HuffmanNode*)malloc(sizeof(struct
HuffmanNode));
        node->data = data[i];
        node->frequency = frequency[i];
        node->left = node->right = NULL;
        insertMinHeap(minHeap, createMinHeapNode(node));
    }

    // Build the Huffman Tree
    while (minHeap->head != NULL) {
        left = extractMin(minHeap)->node;
        right = extractMin(minHeap)->node;

        top = (struct HuffmanNode*)malloc(sizeof(struct HuffmanNode));
        top->data = '\0';
        top->frequency = left->frequency + right->frequency;
        top->left = left;
        top->right = right;
    }
}
```



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```
        insertMinHeap(minHeap, createMinHeapNode(top));
    }
    return extractMin(minHeap)->node;
}

// Function to print the Huffman codes for each character
void printHuffmanCodes(struct HuffmanNode* root, int arr[], int top) {
    if (root->left) {
        arr[top] = 0;
        printHuffmanCodes(root->left, arr, top + 1);
    }

    if (root->right) {
        arr[top] = 1;
        printHuffmanCodes(root->right, arr, top + 1);
    }

    if (root->data) {
        printf("%c: ", root->data);
        for (int i = 0; i < top; i++) {
            printf("%d", arr[i]);
        }
        printf("\n");
    }
}

int main() {
    char data[] = {'a', 'b', 'c', 'd', 'e', 'f'};
    int frequency[] = {5, 9, 12, 13, 16, 45};
    int n = sizeof(data) / sizeof(data[0]);
```



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```
struct HuffmanNode* root = buildHuffmanTree(data, frequency, n);

int arr[100], top = 0;
printf("Huffman Codes:\n");
printHuffmanCodes(root, arr, top);

return 0;
}
```

Output:

Huffman Codes:

a: 1100

c: 1101

b: 111

f: 0

e: 10

d: 111

Conclusion:

1) What are some real-world applications of Huffman coding, and why it is preferred in those applications?

❖ Data Compression:

Huffman coding is extensively used in data compression algorithms like ZIP, GZIP, and DEFLATE. It's efficient for compressing text, images, and other data types.

❖ Image and Video Compression:

JPEG (Joint Photographic Experts Group) uses Huffman coding to compress images. MPEG (Moving Picture Experts Group) standards also employ Huffman coding for video compression.



❖ **Text Compression:**

Huffman coding is used in text editors, search engines, and document storage systems to compress and store text data efficiently.

❖ **Data Encryption:**

- Huffman codes are employed in various encryption and data encoding schemes.
- They assist in reducing the size of encrypted data while preserving data security.

❖ **Data Transmission:** Huffman coding can be used in data transmission protocols, especially when data needs to be sent over networks with limited bandwidth. By reducing the size of the data, it can improve the speed and efficiency of data transmission.

❖ **Data storage:** In applications where storage space is at a premium, such as in embedded systems, Huffman coding can be used to optimize data storage. This is crucial in scenarios where memory resources are constrained.

2) What are the Limitations and potential drawbacks of using Huffman coding in practical data compression scenarios?

1. **Not Suitable for All Data Types:** Huffman coding is most effective for data with non-uniform symbol frequencies, where some symbols are more common than others. For data with a relatively uniform distribution, other compression methods like Run-Length Encoding (RLE) may be more efficient.
2. **Variable-Length Codes:** Huffman coding generates variable-length codes, which can complicate storage and transmission. Decoding variable-length codes efficiently requires additional bookkeeping, making it less suitable for certain real-time applications.
3. **Limited Compression for Small Data Sets:** For very small data sets, Huffman coding might not provide significant compression benefits due to the overhead of maintaining the codebook and encoding/decoding data.



4. **Complexity and Overhead:** Creating and maintaining the Huffman tree, especially in real-time applications, can add complexity and computational overhead. It may not be the best choice for scenarios where speed is of the essence.
5. **Lossless Compression Only:** Huffman coding is primarily used for lossless compression, which preserves data integrity. It is not suitable for applications where lossy compression is acceptable or preferred, such as image or audio compression.
6. **Lack of Security:** Huffman coding does not provide any security or encryption of the data. It is not designed to protect data from unauthorized access, and additional encryption methods may be required when security is a concern.
7. **Encoding and decoding complexity:** Huffman coding requires constructing and maintaining a binary tree during both encoding and decoding. While the algorithm itself is relatively simple, constructing the Huffman tree and decoding variable-length codes can be computationally intensive, especially for large datasets. This complexity may not be suitable for real-time applications or environments with limited computational resources.
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