



<b>Experiment No.8</b>
Implementation Huffman encoding(Tree) using Linked List
Name: Swarup Satish Kakade
Roll No: 19
Date of Performance:
Date of Submission:
Marks:
Sign:

**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:



# **Vidyavardhini's College of Engineering and Technology**

## **Department of Artificial Intelligence & Data Science**

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



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;
```



# Vidyavardhini's College of Engineering and Technology

## Department of Artificial Intelligence & Data Science

```
    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
```



```
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;
    }
}
```



```
        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]);
```



# Vidyavardhini's College of Engineering and Technology

## Department of Artificial Intelligence & Data Science

```
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?

Huffman coding plays a significant role in a variety of technological applications:

#### 1) Data Compression:

Huffman coding finds wide application in data compression techniques such as ZIP, GZIP, and DEFLATE. These algorithms efficiently reduce the size of data, making them suitable for compressing text, images, and various other data formats.





### **2)Image and Video Compression:**

Within the realm of image compression, JPEG (Joint Photographic Experts Group) utilizes Huffman coding to achieve efficient compression. In video compression, standards like MPEG (Moving Picture Experts Group) also leverage Huffman coding for effective data size reduction.

### **3)Text Compression:**

Huffman coding is commonly used in text editors, search engines, and document storage systems. It aids in compactly storing text data, making it a valuable tool for efficient data management.

### **4)Data Encryption:**

Beyond compression, Huffman codes are instrumental in various encryption and data encoding schemes. They contribute to the reduction of encrypted data size while maintaining data security, enhancing the overall encryption process.

In summary, Huffman coding is a versatile and fundamental technique that plays a vital role in various domains, enabling efficient data compression, storage, and security in the digital world.

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

1. **Data Type Compatibility:** Huffman coding is exceptionally efficient when dealing with data characterized by uneven symbol frequencies, where some symbols occur more frequently than others. However, for data that exhibits a relatively uniform distribution of symbols, alternative compression techniques like Run-Length Encoding (RLE) may prove to be more effective.
2. **Variable-Length Codes:** One distinctive feature of Huffman coding is its generation of variable-length codes. While this is advantageous for compression, it can introduce complexities in storage and transmission.



# Vidyavardhini's College of Engineering and Technology

## Department of Artificial Intelligence & Data Science

Efficiently decoding variable-length codes demands additional housekeeping, rendering it less ideal for real-time applications.

3. **Limited Impact on Small Data Sets:** In the case of very compact data sets, the advantages of Huffman coding may be outweighed by the overhead of managing the codebook and encoding/decoding processes, resulting in relatively modest compression gains.
4. **Complexity and Overhead:** Constructing and maintaining the Huffman tree, particularly in real-time contexts, can introduce complexity and computational overhead. Therefore, it may not be the preferred choice in situations where rapid processing is paramount.
5. **Exclusive Use in Lossless Compression:** Huffman coding is primarily employed for lossless compression, preserving data integrity throughout the process. It is not suitable for scenarios where some degree of data loss is acceptable or preferred, as seen in image and audio compression.
6. **Absence of Security Features:** It's important to note that Huffman coding does not incorporate any security or data encryption capabilities. It is not designed to safeguard data from unauthorized access. In scenarios where data security is a concern, supplementary encryption methods are necessary.