

# Image Processing and Computer Vision



# RUN LENGTH CODING

- ❖ Run-length coding is a very widely used and simple compression technique which does not assume a memoryless source
  - We replace runs of symbols (possibly of length one) with pairs of (*run-length*, *symbol*)
  - For images, the maximum run-length is the size of a row
- ❖ Run length coding (RLC) is effective when long sequences of the same symbol occur.
- ❖ Run length coding exploits the spatial redundancy by coding the number of symbols in a run.

# RUN LENGTH CODING

- ❖ The term run is used to indicate the repetition of a Symbol, while the term run length is used to represent the number of repeated symbols.
- ❖ Run length coding maps a sequence of numbers into a sequence of symbol pairs (run value).
- ❖ Images with large areas of constant shade are good candidates for this kind of compression.
- ❖ It is used in windows bitmap file format.
- ❖ Run length coding can be classified into
  - I) 1-D run length coding.
  - II) 2-D run length coding.

# RUN LENGTH CODING

I) I-D run length coding-

In 1-D run length coding, each scan line is encoded independently.

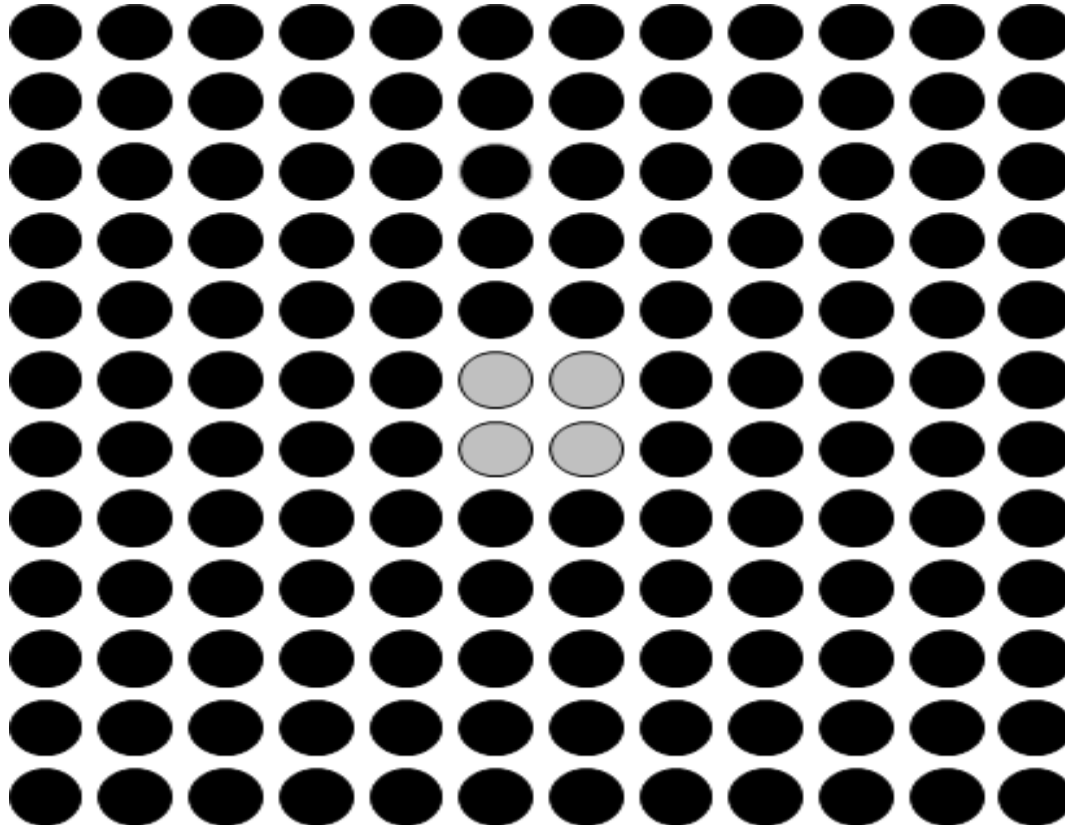
1-D RLC utilizes only the horizontal correlation between pixels on the same scan line.

2-D RLC utilizes both horizontal and vertical correlation between pixels.

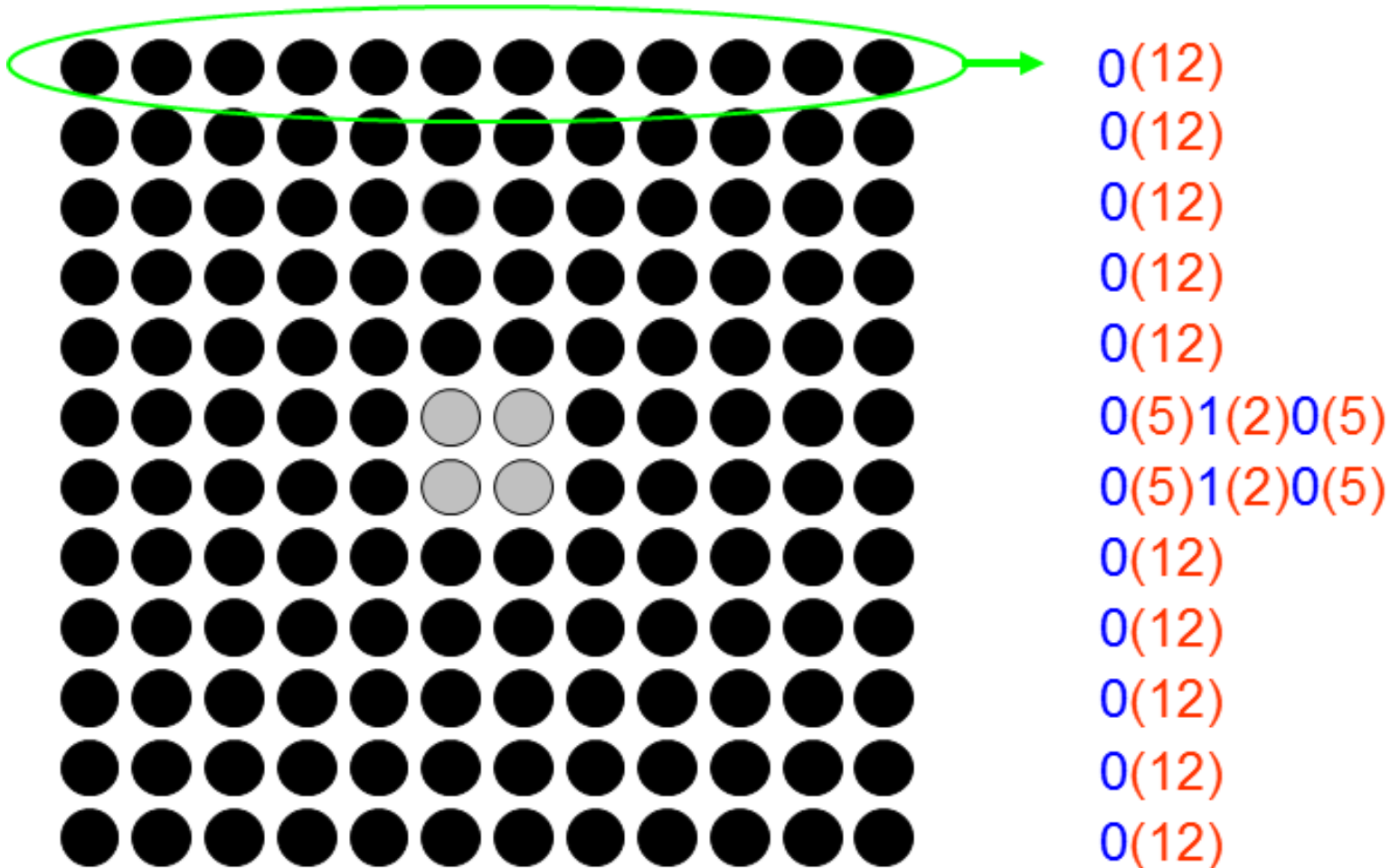
# RUN LENGTH CODING

- ❑ 2-D run length coding
  - ❖ The 1-D run length coding utilizes the correlation between pixels within a scanline.
  - ❖ In order to utilize correlation between pixels in neighbouring scan lines to achieve higher coding efficiency, 2-D run length coding was developed.
  - ❖ In RLC, two values are transmitted- the first value indicates the number of times a particular symbol has occurred. The second value indicates the actual symbol.

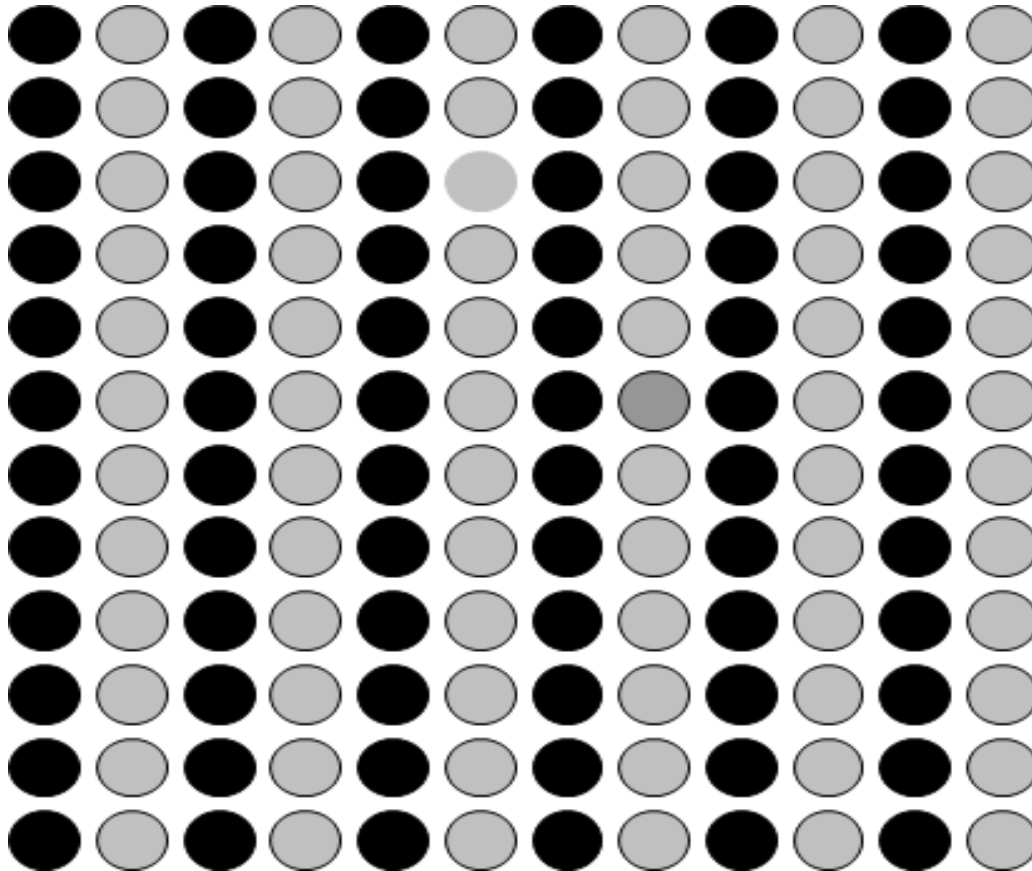
# RUN LENGTH CODING



# RUN LENGTH CODING



# RUN LENGTH LIMITATIONS





# ENTROPY

- The average bit rate of a coding scheme is given by:

$$\bar{b} = \sum_{i=1}^L p_i b_i$$

- $p_i$  = probability of occurrence of the  $i^{\text{th}}$  reconstruction level
- $b_i$  = number of bits assigned to the  $i^{\text{th}}$  message

# ENTROPY

- ❖ In information theory, **entropy** is a measure of the uncertainty associated with a random variable.
- ❖ This term usually refers to the **Shannon entropy**, which quantifies the expected value of the information contained in a message.
- ❖ Entropy is typically measured in bits.
- ❖ Entropy, in an information sense, is a measure of unpredictability.
- ❖ Compressed message is more unpredictable, because there is no redundancy; each bit of the message is communicating a unique bit of information.
- ❖ Entropy is a measure of how much information the message contains.

# ENTROPY

$$E = \text{entropy}(I)$$

E=a scalar value representing the entropy of grayscale image I. Entropy is a statistical measure of randomness that can be used to characterize the texture of the input image. Entropy is defined as

$$H(X) = \sum_{i=1}^n p(x_i) I(x_i) = \sum_{i=1}^n p(x_i) \log_b \frac{1}{p(x_i)} = - \sum_{i=1}^n p(x_i) \log_b p(x_i),$$

where  $p$  contains the histogram counts returned from `imhist`.

where  $b$  is the base of the logarithm used. Common values of  $b$  are 2,10.

# A SIMPLE EXAMPLE

- ❖ Suppose we have a message consisting of 5 symbols, e.g.  
[▶ ♣♣♠ 😊 ▶ ♣☀ ▶ 😊 ]
- ❖ How can we code this message using 0/1 so the coded message will have minimum length (for transmission or saving!)

- ❖ 5 symbols → at least 3 bits
- ❖ For a simple encoding,  
length of code is  $10 \times 3 = 30$  bits

▶	000
♣	001
😊	010
♠	011
☀	100

# A SIMPLE EXAMPLE – CONT.

❖ Intuition: Those symbols that are more frequent should have smaller codes, yet since their length is not the same, there must be a way of distinguishing each code

❖ For Huffman code,  
length of encoded message  
will be ▶♣♣♠😊 ▶♣☀▶😊  
 $= 3*2 + 3*2 + 2*2 + 3 + 3 = 24\text{bits}$

Symbol	Freq.	Code
▶	3	00
♣	3	01
😊	2	10
♠	1	110
☀	1	111

# ENTROPY CODING

**Example 7.3** Calculate the entropy for the symbols shown in Table 7.2.

**Table 7.2** Symbols and their distribution

Symbol	1	2	3	4	5	6
Probability	0.4	0.2	0.2	0.1	0.05	0.05

*Solution* Entropy =  $-\sum p_i \times \log_2 p_i$ ; as  $\log_2 x = \log_{10} x / \log_{10} 2$

$$\begin{aligned} &= -[0.4 \times (\log_{10}(0.4)/\log_{10}(2)) + 0.2 \times (\log_{10}(0.2)/\log_{10}(2)) + 0.2 \times \\ &\quad (\log_{10}(0.2)/\log_{10}(2)) + 0.1 \times (\log_{10}(0.1)/\log_{10}(2)) + 0.05 \times (\log_{10}(0.05)/\log_{10}(2)) \\ &\quad + 0.05 \times (\log_{10}(0.05)/\log_{10}(2))] \\ &= -[-0.5288 - 0.4644 - 0.4644 - 0.3322 - 0.2161 - 0.2161] \\ &= 2.22 \end{aligned}$$

# HUFFMAN CODE

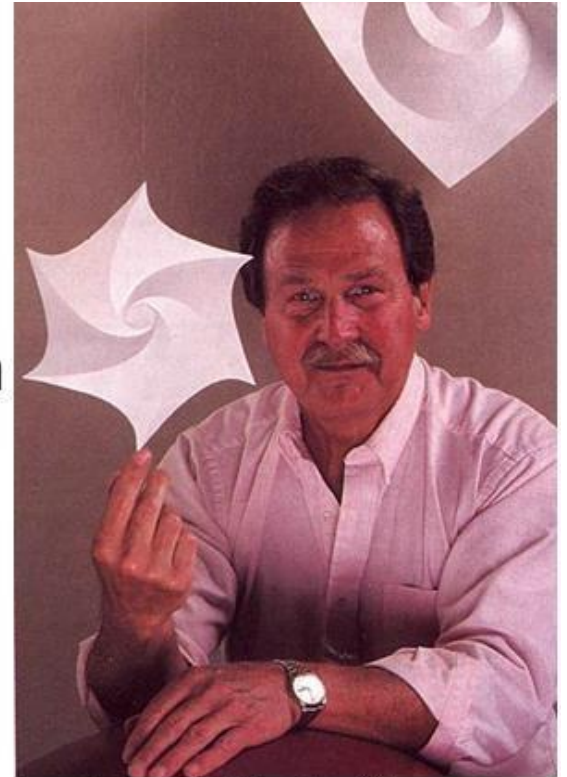
- Developed by David A. Huffman in 1951
- Widely used in Computers, HDTV, modems etc
- Robert M. Fano assigned a term paper on efficient representation of numbers



**Robert M. Fano**



**Claude Shannon**



**David A. Huffman**



# HUFFMAN CODING

- ❖ Huffman codes are optimal codes that map one symbol with one code word.
- ❖ In Huffman coding, it is assumed that each pixel intensity has associated with it a certain probability of occurrence, and this probability is spatially invariant.
- ❖ Huffman coding assigns a binary code to each intensity value, with shorter codes going to intensities with higher probability.
- ❖ If the probabilities can be estimated then the table of Huffman codes can be fixed in both the encoder and the decoder.



# HUFFMAN CODING

❑ The Parameters involved in Huffman coding are as follows.

❖ **Entropy**

❖ **Average Length**

❖ **Efficiency**

❖ **Variance.**

❑ **Prefix Code**-A code is a prefix code if no code word is the prefix of another code word. The main advantage of a prefix code is that it is uniquely decodable.

Example-Huffman code.

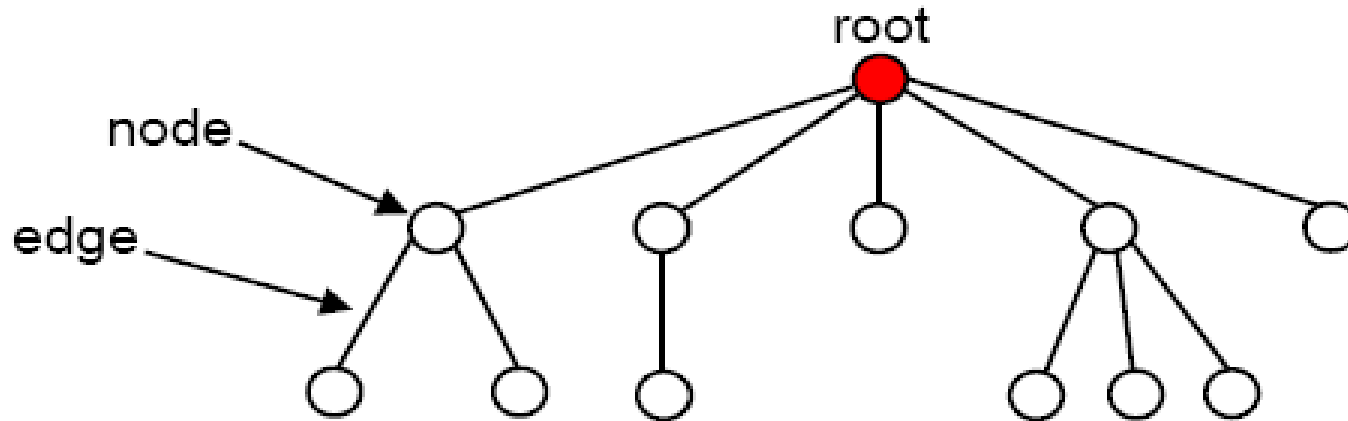
# HUFFMAN CODING ALGORITHM

- ❖ Initialization: Put all symbols on a list sorted according to their frequency counts.
- From the list pick two symbols with the lowest frequency counts. Form a Huffman subtree that has these two symbols as child nodes and create a parent node.
- Assign the sum of the children's frequency counts to the parent and insert it into the list such that the order is maintained.
- Delete the children from the list.
- Repeat until the list has only one symbol left
- ❖ Assign a codeword for each leaf based on the path from the root.

# PROPERTIES OF HUFFMAN CODES

- ❖ No Huffman code is the prefix of any other Huffman codes so decoding is unambiguous
- ❖ The Huffman coding technique is optimal (but we must know the probabilities of each symbol for this to be true)
- ❖ Symbols that occur more frequently have shorter Huffman codes

# WHAT IS A TREE?

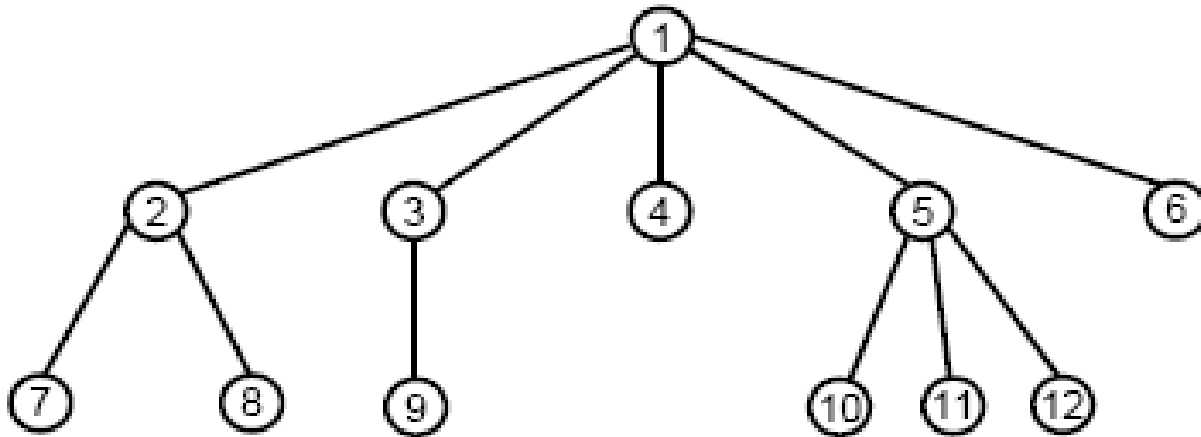


A tree consists of:

- a set of *nodes*
- a set of *edges*, each of which connects a pair of nodes.

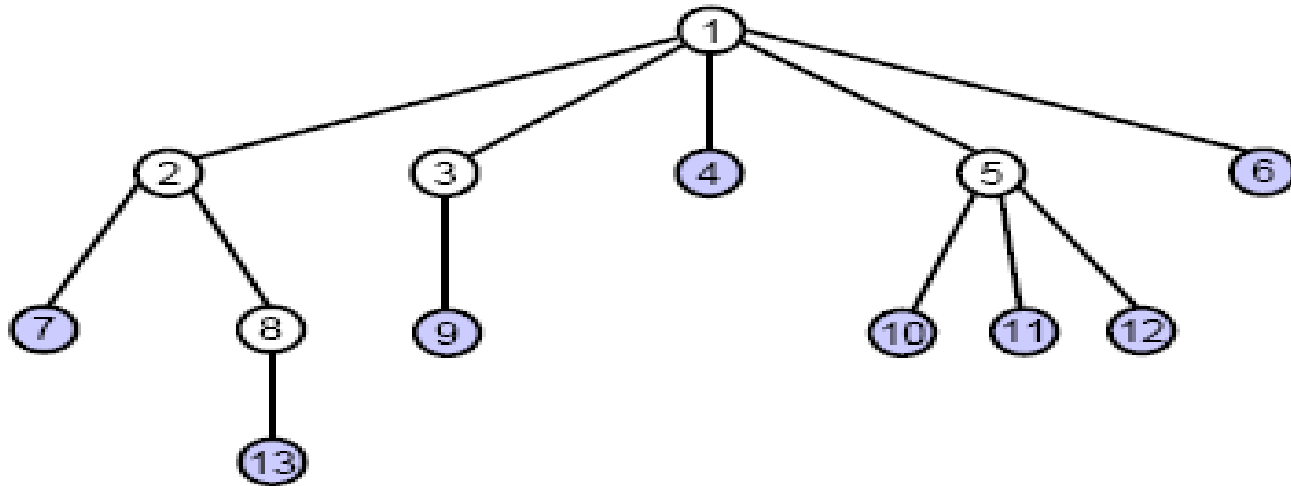
The node at the “top” of the tree is called the *root* of the tree.

# RELATIONSHIPS BETWEEN NODES



- ❖ If a node  $N$  is connected to other nodes that are directly below it in the tree,  $N$  is referred to as their *parent* and they are referred to as its *children*.
- ❖ example: node 5 is the parent of nodes 10, 11, and 12
- ❖ Each node is the child of *at most one* parent.

# TYPES OF NODES



- ❖ A *leaf node* is a node without children.
- ❖ An *interior node* is a node with one or more children.

# EXAMPLE

**Example 9.2** Obtain the Huffman code for the word '**COMMITTEE**'

**Solution** Total number of symbols in the word '**COMMITTEE**' is 9.

$$\text{Probability of a symbol} = \frac{\text{Total number of occurrence of symbol in a message}}{\text{Total number of symbol in the message}}$$

$$\text{Probability of the symbol C} = p(C) = 1/9$$

$$\text{Probability of the symbol O} = p(O) = 1/9$$

$$\text{Probability of the symbol M} = p(M) = 2/9$$

$$\text{Probability of the symbol I} = p(I) = 1/9$$

$$\text{Probability of the symbol T} = p(T) = 2/9$$

$$\text{Probability of the symbol E} = p(E) = 2/9$$



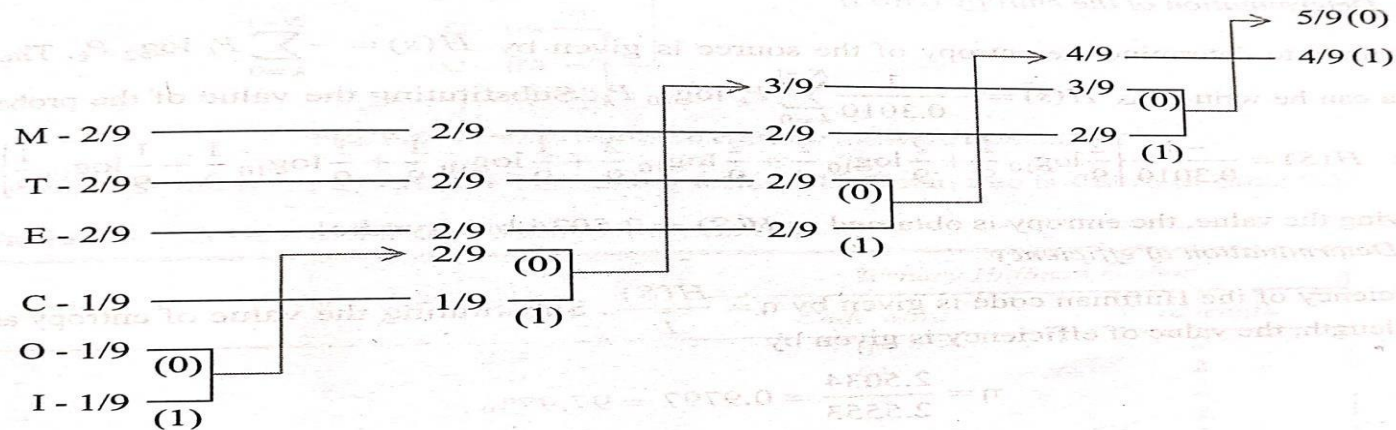
**Step 1** Arrange the symbols into descending order according to the probability. This is illustrated in Table 9.1.

**Table 9.1** Arrangement of symbols in descending order of probability

Symbol	Probability
M	2/9
T	2/9
E	2/9
C	1/9
O	1/9
I	1/9

**Step 2** Construction of Huffman tree

The Huffman tree corresponding to the term COMMITTEE is shown below.



**Step 3** Code word from the Huffman Tree

The code word for each symbol and the corresponding word length is shown in Table 9.2.

**Table 9.2** Code word from Huffman tree

Symbol	Probability	Binary Huffman method	
		Codeword	Word length
M	2/9	01	2
T	2/9	10	2
E	2/9	11	3
C	1/9	001	4
O	1/9	0000	4
I	1/9	0001	



$$H(S) = \frac{-1}{0.3010} \left\{ \frac{2}{9} \log_{10} \frac{2}{9} + \frac{2}{9} \log_{10} \frac{2}{9} + \frac{2}{9} \log_{10} \frac{2}{9} + \frac{1}{9} \log_{10} \frac{1}{9} + \frac{1}{9} \log_{10} \frac{1}{9} + \frac{1}{9} \log_{10} \frac{1}{9} \right\}$$

$$H(S) = 2.5034 \text{ bits/symbol}$$

Finding the average length  $\bar{L}$  (Ternary Huffman Coding)

$$\bar{L} = \sum_{k=0}^{N-1} P_k l_k = \frac{2}{9} \times 2 + \frac{2}{9} \times 2 + \frac{2}{9} \times 2 + \frac{1}{9} \times 2 + \frac{1}{9} \times 2 + \frac{1}{9} \times 2$$

$$= 0.4444 + 0.4444 + 0.4444 + 0.2222 + 0.2222 + 0.2222$$

$$\bar{L} = 1.9998 \text{ bits/symbol}$$

**Efficiency of the compression in ternary Huffman Coding ( $\eta$ )**

The efficiency of the Huffman code is given by

$$\eta = \frac{H(S)}{\bar{L}}$$

By substituting the value of  $H(S)$ ,  $\bar{L}$  we get

$$\eta = \frac{2.5034}{1.9998} = 125.18\%$$

$$\eta = 1.2518$$

# HUFFMAN CODING EXAMPLE

## Input Image

1	2	5	7
2	3	7	5
7	2	1	3
6	4	7	1

## STEP: 1

### Probability of occurrence of gray level

Probability of '1' =  $P(1) = 3/16$

Probability of '2' =  $P(2) = 3/16$

Probability of '3' =  $P(3) = 2/16$

Probability of '4' =  $P(4) = 1/16$

Probability of '5' =  $P(5) = 2/16$

Probability of '6' =  $P(6) = 1/16$

Probability of '7' =  $P(7) = 4/16$

# HUFFMAN CODING EXAMPLE

Pixel Value	1	2	3	4	5	6	7
Probability of occurrence	3/16	3/16	2/16	1/16	2/16	1/16	4/16

**STEP 2:** Arrange the probability in descending order

Code	Symbol	Probability	Step 1	Step 2	Step 3	Step 4	Step 5
	7	4/16					
							12

# HUFFMAN CODING EXAMPLE

Pixel Value	1	2	3	4	5	6	7
Probability of occurrence	3/16	3/16	2/16	1/16	2/16	1/16	4/16

**STEP 2:** Arrange the probability in descending order

Code	Symbol	Probability	Step 1	Step 2	Step 3	Step 4	Step 5
	7	4/16					
	1	3/16					
	2	3/16					
	3	2/16					
	5	2/16					
	4	1/16					
	6	1/16					12

# HUFFMAN CODING EXAMPLE

Pixel Value	1	2	3	4	5	6	7
Probability of occurrence	3/16	3/16	2/16	1/16	2/16	1/16	4/16

## STEP 2: Arrange the probability in descending order

Code	Symbol	Probability	Step 1	Step 2	Step 3	Step 4	Step 5
(01)	7	4/16	4/16	4/16	5/16	7/16	9/16 (0)
(11)	1	3/16	3/16	4/16	4/16	5/16 (0)	7/16 (1)
(000)	2	3/16	3/16	3/16	4/16 (0)	4/16 (1)	
(001)	3	2/16	2/16	3/16 (0)	3/16 (1)		
(100)	5	2/16	2/16 (0)	2/16 (1)			
(1010)	4	1/16	2/16				
(1011)	6	1/16					12

# HUFFMAN CODING EXAMPLE

Pixel Value (or) Symbol	1	2	3	4	5	6	7
Probability, $P_k$	3/16	3/16	2/16	1/16	2/16	1/16	4/16
Code	11	000	001	1010	100	1011	01

# HUFFMAN CODING EXAMPLE

## STEP: 3 To find the **Average length**

$$\begin{aligned}\bar{L} &= P_1 l_1 + P_2 l_2 + P_3 l_3 + P_4 l_4 + P_5 l_5 + P_6 l_6 + P_7 l_7 \\ &= \frac{3}{16} \times 2 + \frac{3}{16} \times 3 + \frac{2}{16} \times 3 + \frac{1}{16} \times 4 + \frac{2}{16} \times 3 + \frac{1}{16} \times 4 + \frac{4}{16} \times 2 \\ &= \frac{1}{16} [6 + 9 + 6 + 4 + 6 + 4 + 8] \\ &= \frac{43}{16} = 2.6875\end{aligned}$$

Where  $P_k$  is Probability of the symbol or pixel value

$l_k$  is length of the code for the corresponding symbol

$\bar{L}$  is Average length

## STEP: 4 To find the Entropy, H(s)

$$H(s) = -\sum_k P_k \log_2(P_k)$$

$$\begin{aligned} H(s) &= -\left\{ \frac{3}{16} \times \log_2\left(\frac{3}{16}\right) + \frac{3}{16} \times \log_2\left(\frac{3}{16}\right) + \frac{2}{16} \times \log_2\left(\frac{2}{16}\right) + \frac{1}{16} \times \log_2\left(\frac{1}{16}\right) + \frac{2}{16} \times \log_2\left(\frac{2}{16}\right) + \frac{1}{16} \times \log_2\left(\frac{1}{16}\right) + \frac{4}{16} \times \log_2\left(\frac{4}{16}\right) \right\} \\ &= -\left\{ \frac{3}{16} \times (-2.4153) + \frac{3}{16} \times (-2.4153) + \frac{2}{16} \times (-3.0003) + \frac{1}{16} \times (-4.0004) + \frac{2}{16} \times (-3.0003) + \frac{1}{16} \times (-4.0004) + \frac{4}{16} \times (-2) \right\} \\ &= \frac{1}{16} \{ 7.2459 + 7.2459 + 6.0006 + 4.0004 + 6.0006 + 4.0004 + 8 \} \\ &= \frac{1}{16} \times 42.4938 = 2.6559 \end{aligned}$$

## STEP: 5 To find the Efficiency $\eta = \frac{H(s)}{\bar{L}}$

$$\begin{aligned} \eta &= \frac{2.6559}{2.6875} \\ &= 0.9882 \end{aligned}$$

$$\text{Efficiency in Percentage} = 0.9882 \times 100 = \mathbf{98.82\%}$$