



COMSATS Institute of  
Information Technology

ECI750 Multimedia Data Compression

# Lecture 4

## *Huffman Coding – I*

Dr. Shadan Khattak

Department of Electrical Engineering

COMSATS Institute of Information Technology - Abbottabad



# Huffman Coding

- In this lecture, we will study
  - How can we generate Huffman codes when the probability model of the source is known?
  - Different types of Huffman codes, including
    - Minimum Variance Huffman Codes,
    - Extended Huffman Codes, and
    - Non-binary Huffman Codes

# Huffman Coding

- The Huffman Coding Algorithm
  - Developed by David Huffman as part of a class assignment
  - Huffman codes are prefix codes and are optimum for a given model (set of probabilities)
  - It is based on two ideas:
    - In an optimum code, symbols that occur more frequently (have a higher probability of occurrence) will have shorter codewords than symbols that occur less frequently.
    - In an optimum code, the two symbols that occur least frequently will have the same length.

# Huffman Coding

- Huffman Coding Principle:
  - Starting with two least probable symbols,  $\gamma$  and  $\delta$ , of an alphabet  $A$ , if the codeword for  $\gamma$  is  $[m]0$ , the codeword for  $\delta$  should be  $[m]1$ , where  $m$  is a string of 1s and 0s.
  - Now the two symbols can be combined into a group, which represents a new symbol  $\psi$  in the alphabet set.
    - The symbol  $\psi$  has the probability  $P(\gamma) + P(\delta)$ .
  - Recursively determine the bit pattern  $[m]$  using the new alphabet set.

# Huffman Coding

- Example:

$$A = \{a_1, a_2, a_3, a_4, a_5\}$$

$$P(a_1) = P(a_3) = 0.2, P(a_2) = 0.4, P(a_4) = P(a_5) = 0.1$$

$$H = 2.122 \text{ bits/symbols}$$

Symbol	Step 1	Step 2	Step 3	Step 4	Codeword
$a_2$	0.4	0.4	0.4	0.6 0	1
$a_1$	0.2	0.2	0.4 } 0	0.4 1	01
$a_3$	0.2	0.2 } 0	0.2 } 1		000
$a_4$	0.1 } 0	0.2 } 1			0010
$a_5$	0.1 } 1				0011



# Huffman Coding

- Example:

- Average codeword length:

$$l = 0.4 * 1 + 0.2 * 2 + 0.2 * 3 + 0.1 * 4 + 0.1 * 4 = 2.2 \text{ bits/symbol}$$

- *Redundancy = Average Codeword Length – Entropy*  
$$= 2.2 - 2.122$$
$$= 0.078 \text{ bits/symbol}$$

- For Huffman codes, the redundancy is zero when the probabilities are negative powers of 2.

# Huffman Coding

- **Minimum Variance Huffman Code**

- When more than two symbols in a Huffman tree have the same probability, different merge orders produce different Huffman codes.

Symbol	Step 1	Step 2	Step 3	Step 4	Codeword
$a_2$	0.4	0.4	0.4	0.6 0	00
$a_1$	0.2	0.2	0.4 } 0	0.4 1	10
$a_3$	0.2	0.2 } 0	0.2 } 1		11
$a_4$	0.1 } 0	0.2 } 1			010
$a_5$	0.1 } 1				011

The average codeword length is still 2.2 bits/symbol. But variances are different!

- We prefer a code with smaller length-variance.
- To create a minimum variance Huffman code, put the combined letter as high in the list as possible!

# Huffman Coding

- **Minimum Variance Huffman Code**

- Why are we interested in minimum variance codes?
  - The greater the variance, the more difficult is the buffer design problem
- Consider the following:
- For both the codes considered in Slide 5 and Slide 7, the average codeword length is 2.2 bits/symbol.
- If we have to transmit 10,000 symbols/sec, we need a transmission capacity of  $10,000 \times 2.2 = 22,000$  bits/sec.
- If we are transmitting only symbols  $a_4$  and  $a_5$  for a few seconds, then,
  - For the code on Slide 5:  $10,000 \times 4 = 40,000$  bits/sec  $\Rightarrow$  18,000 bits/sec buffering
  - For the code on Slide 7:  $10,000 \times 3 = 30,000$  bits/sec  $\Rightarrow$  8,000 bits/sec buffering



# Huffman Coding

- **Length of the Huffman Code**

- The Huffman coding procedure generates an optimum code.
- The average codeword length  $\hat{l}$  of an optimum code (and thus the Huffman code) is bounded below by the entropy of the source ( $S$ ) and bounded above by the entropy of the source plus 1 bit i.e.,

$$H(S) \leq \hat{l} < H(S) + 1$$

- Given a sequence of positive integers  $\{l_1, l_2, \dots, l_k\}$  satisfies

$$\sum_{i=1}^k 2^{-l_i} \leq 1$$

There exists a uniquely decodable code whose codeword lengths are given by  $\{l_1, l_2, \dots, l_k\}$

# Huffman Coding

- **Extended Huffman Codes**

- As  $P_{max}$  increases, the efficiency of Huffman coding decreases.

# Huffman Coding

- **Extended Huffman Codes**

- **Example 1:**

- Consider a source that puts out *iid* letters from the alphabet  $A = \{a_1, a_2, a_3\}$  with the probability model:  $P\{a_1\} = 0.8, P\{a_2\} = 0.02, P\{a_3\} = 0.18$
- Entropy = 0.816 bits/symbol
- Huffman Code:

Letter	Codeword
$a_1$	0
$a_2$	11
$a_3$	10

- Average codeword length: 1.2 bits/symbol
- Redundancy: 0.384 bits/symbol (47% of the entropy)
- $\Rightarrow$  to code this sequence, we would need 47% more bits than the minimum required.

# Huffman Coding

- Extended Huffman Codes

- Example 2:

- Average codeword length: 1.7228 bits/symbol
- Average codeword length in terms of the original alphabet:  $1.7228/2 = 0.8614$  bits/symbol
- Redundancy: 0.045 bits/symbol (5.5% of the entropy)
- $\Rightarrow$  to code this sequence, we would need 5.5% more bits than the minimum required.

Letter	Probability	Code
$a_1a_1$	0.64	0
$a_1a_2$	0.016	10101
$a_1a_3$	0.144	11
$a_2a_1$	0.016	101000
$a_2a_2$	0.0004	10100101
$a_2a_3$	0.0036	1010011
$a_3a_1$	0.1440	100
$a_3a_2$	0.0036	10100100
$a_3a_3$	0.0324	1011

# Huffman Coding

- **Extended Huffman Codes**

- By coding blocks of symbols together, we can reduce the redundancy of Huffman codes.
- Blocking two symbols together, the alphabet size grows from  $m$  to  $m^2$
- As we block more and more symbols together, the size of the alphabet grows exponentially, and the Huffman coding scheme becomes more impractical.
- Under these conditions, a more practical coding technique is *arithmetic coding* which we are going to see in the next week.

# Huffman Coding

- **Non-Binary Huffman Codes**

- Huffman codes can be applied to n-ary code space. For example, codewords composed of  $\{0,1,2\}$  called ternary Huffman code.

Letter	Probability	Codeword
$a_1$	0.20	2
$a_2$	0.05	021
$a_3$	0.20	00
$a_4$	0.20	01
$a_5$	0.25	1
$a_6$	0.10	020