

Image Processing and Computer Vision



IMAGE COMPRESSION

- **Definition:** Compression means storing data in a format that requires less space than usual.
- Data compression is particularly useful in communications because it enables devices to transmit the same amount of data in fewer bits.
- The bandwidth of a digital communication link can be effectively increased by compressing data at the sending end and decompressing data at the receiving end.
- There are a variety of data compression techniques, but only a few have been standardized.

IMAGE COMPRESSION

Image processing is the application of 2D signal processing methods to images

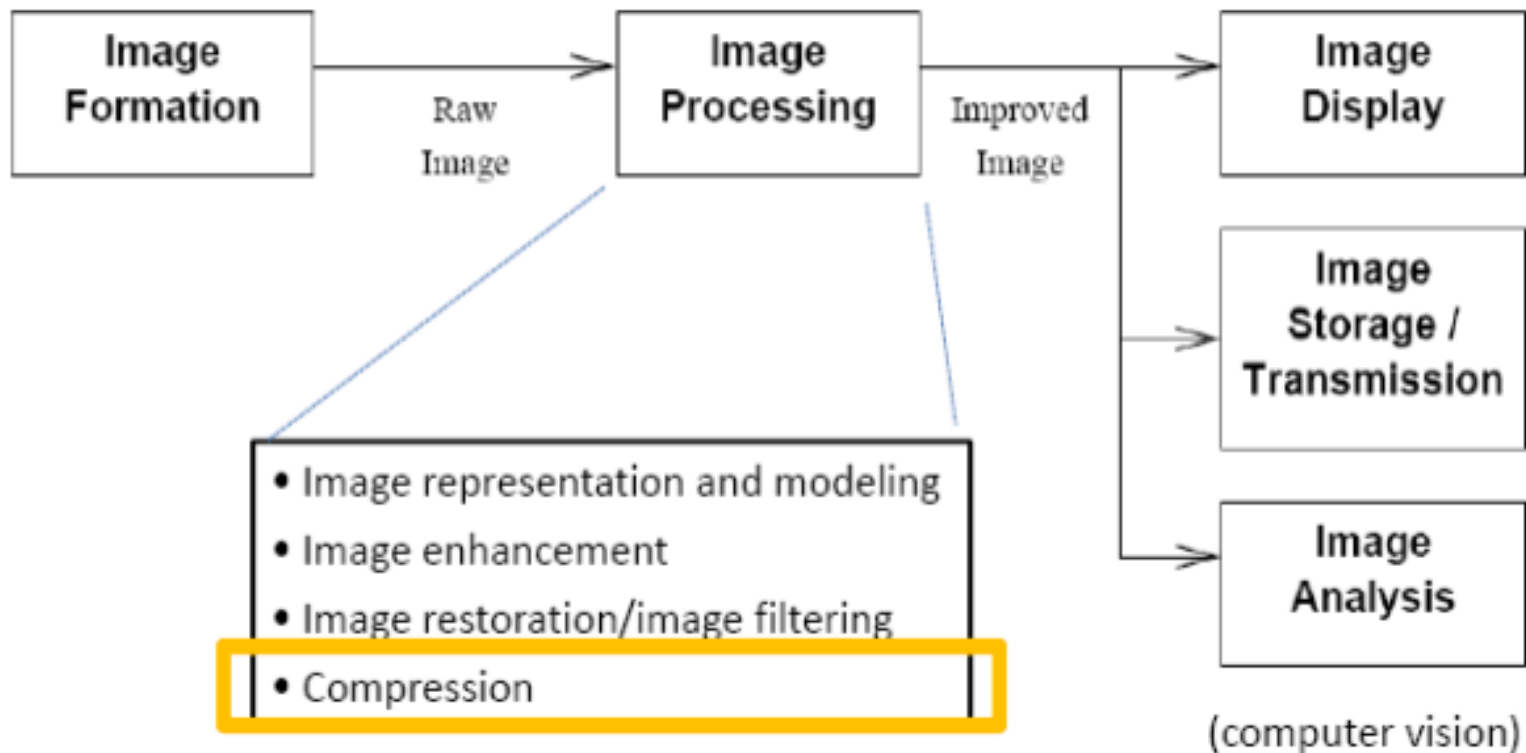


IMAGE COMPRESSION

- Image and video data compression refers to a process in which the amount of data used to represent image and video is reduced to meet a bit rate requirement (below or at most equal to the maximum available bit rate).
 - The quality of the reconstructed image or video satisfies a requirement for a certain application.
 - The complexity of computation involved is affordable for the application.
- **The basis of compression process is removal of redundant data.**

WHAT IS THE NEED FOR COMPRESSION

- In terms of storage, the capacity of a storage can be effectively increased.
- In terms of communications, the bandwidth of a digital communication link can be effectively increased .
- At any given time, the ability of the internet to data is fixed. Thus, if data can effectively be compressed wherever possible, significant improvement of data throughput can be achieved.
- Many files can be combined into one compressed documents making sending easier.

NEED FOR COMPRESSION

- To minimize the Storage Space
- To enable higher rate of Data transfer
- There is still a need to develop newer and faster algorithms adapted to image data

Types of Data Compression

- There are two main types of data compression : Lossy and Lossless.
- In Lossy data compression the message can never be recovered exactly as it was before it was compressed.
- In a Lossless data compression file the original message can be exactly decoded.
- Lossless compression is ideal for text.
- Huffman coding is type of lossless data compression.

TYPES OF COMPRESSION

- There are two types of compression
 - ❖ **Lossless compression-** It can recover the exact original data after compression.
 - It is used mainly for compressing database records, spreadsheet or word processing files, medical imaging, where exact replication of the original is essential.
 - ❖ **Lossy compression-** It will result in a certain loss of accuracy for a substantial increase in compression.
 - Lossy compression is more effective when used to compress graphic images and digitized voice where losses outside visual or aural perception can be tolerated.

TYPES OF COMPRESSION: LOSSLESS

Basic Idea:

- ✦ Convert each gray-value (symbol) into a binary sequence (codeword)
- ✦ Codewords may have different lengths for different gray-values.

Exploit statistical redundancy

- ✦ Assign shorter binary sequences to more common gray levels, and longer codewords to less common ones.
- ✦ How do we make this assignment quantitatively?
Use the notion of **Entropy**.

TYPES OF COMPRESSION: LOSSLESS

- ❖ Perfectly reproduce the image at the decoder($MSE=0$, $PSNR=Inf$)
- ❖ Reduction in bit rate is usually limited.
- ❖ Can not guarantee a fixed rate.
- ❖ Applications: –Fax Machines–Medical Images

LOSSLESS COMPRESSION METHODS

- ❖ Run length Encoding (RLE)
- ❖ Huffman coding
- ❖ Arithmetic coding
- ❖ LZ and LWZ coding
- ❖ Predictive coding
- ❖ Variable length coding etc

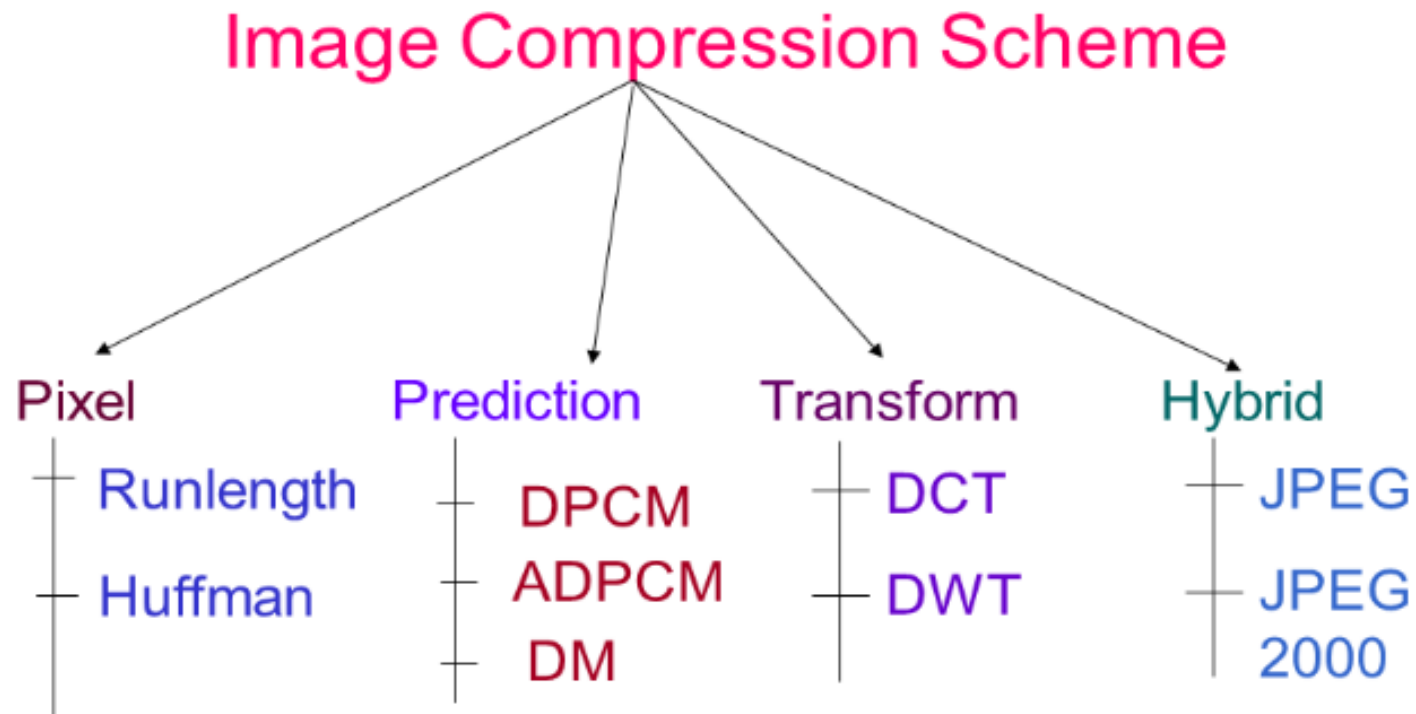
TYPES OF COMPRESSION: LOSSY

- ❑ Idea: some deviation of decompressed image from original is often acceptable.
- ❑ Human visual system might not perceive loss or tolerate it. Thus, omit irrelevant details that humans cannot perceive.
- ❑ No need to represent more than the visible resolution in:
 - ❖ Space
 - ❖ Time
 - ❖ Brightness
 - ❖ color

HOW TO ACHIEVE COMPRESSION

- Minimizing the Redundancy in the Image.
- Types of Redundancy-Mainly two types
 - ❖ Statistical Redundancy- Interpixel Redundancy & Coding
- Redundancy.
 - Interpixel Redundancy-Spatial Redundancy & Temporal Redundancy.
 - ❖ Psychovisual Redundancy.

IMAGE COMPRESSION TECHNIQUES



Compression Algorithms

- Huffman Coding
- Run Length Encoding
- Shift Codes
- Arithmetic Codes
- Block Truncation Codes
- Transform codes
- Vector Quantization

Huffman Coding

- Huffman coding is a popular compression technique that assigns variable length codes (VLC) to symbols, so that the most frequently occurring symbols have the shortest codes.
- On decompression the symbols are reassigned their original fixed length codes.
- The idea is to use short bit strings to represent the most frequently used characters
- and to use longer bit strings to represent less frequently used characters.
- That is, the most common characters, usually **space**, **e**, and **t** are assigned the shortest codes.
- In this way the total number of bits required to transmit the data can be considerably less than the number required if the fixed length ASCII representation is used.
- A Huffman code is a binary tree with branches assigned the value 0 or 1.
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Huffman Algorithm

- To each character, associate a binary tree consisting of just one node.
- To each tree, assign the character's frequency, which is called the tree's weight.
- Look for the two lightest-weight trees. If there are more than two, choose among them randomly.
- Merge the two into a single tree with a new root node whose left and right sub trees are the two we chose.
- Assign the sum of weights of the merged trees as the weight of the new tree.
- Repeat the previous step until just one tree is left.

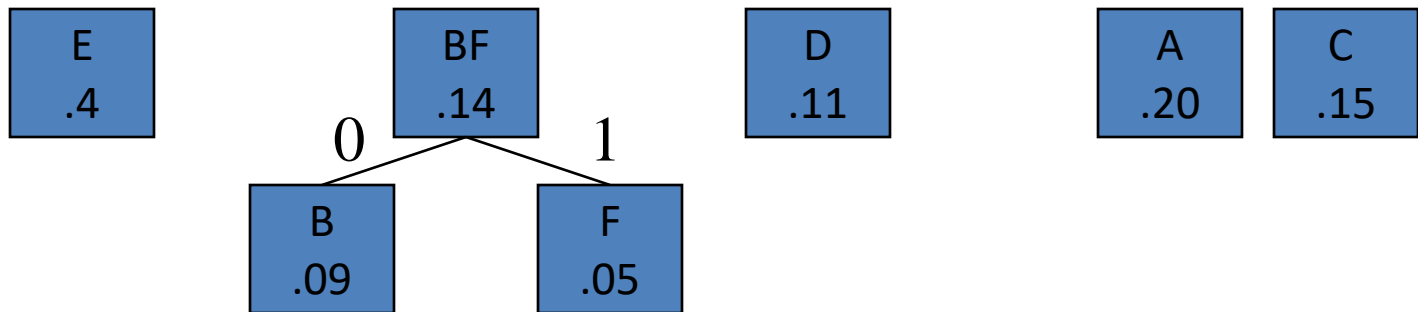
Advantages & Disadvantages

- The problem with Huffman coding is that it uses an integral number of bits in each code.
- If the entropy of a given character is 2.5 bits, the Huffman code for that character must be either 2 or 3 bits, not 2.5.
- Though Huffman coding is inefficient due to using an integral number of bits per code, it is relatively easy to implement and very efficient for coding and decoding.
- It provides the best approximation for coding symbols when using fixed width codes.

Huffman Coding Example

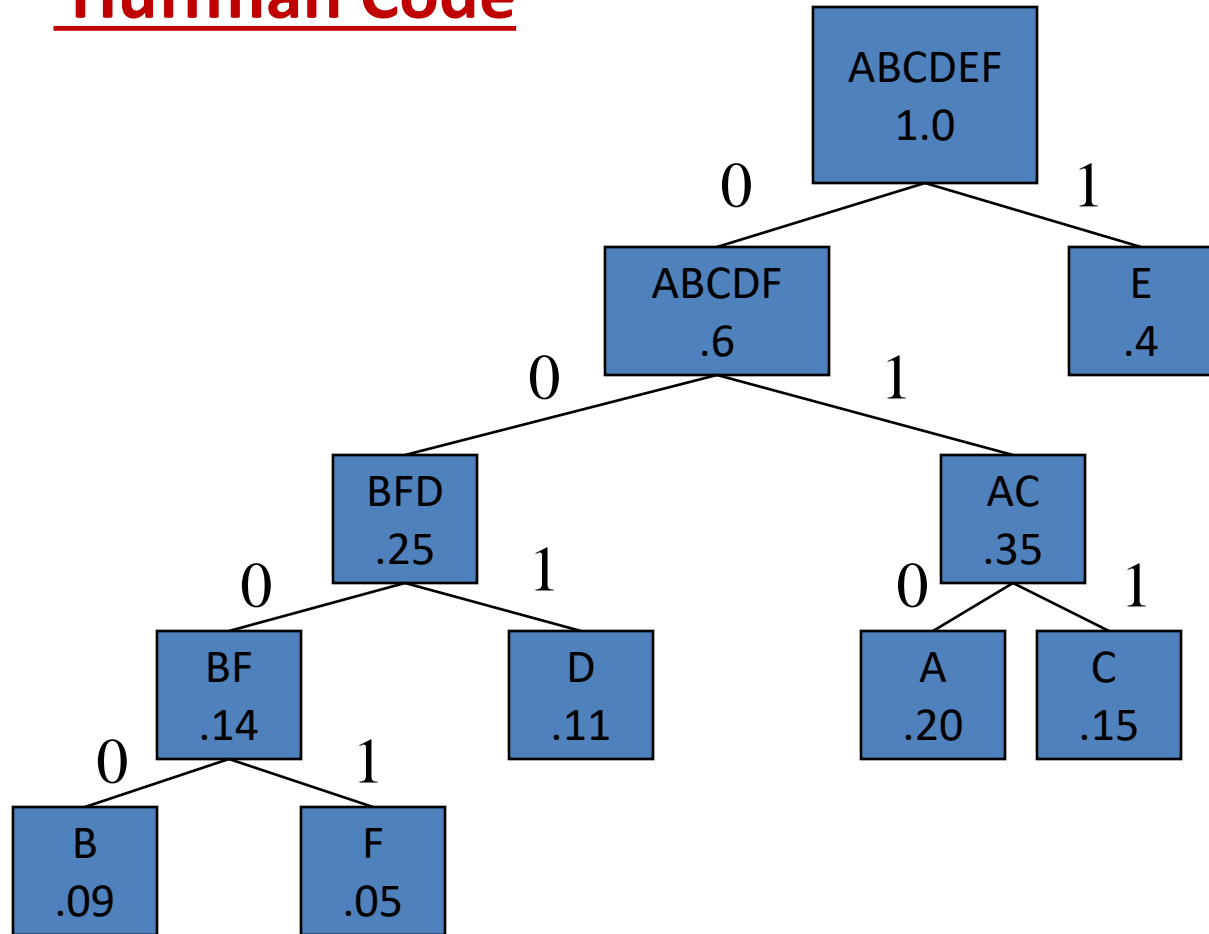
- Character frequencies
 - A: 20% (.20)
 - B: 9% (.09)
 - C: 15%
 - D: 11%
 - E: 40%
 - F: 5%
- No other characters in the document

Huffman Code



Huffman Code

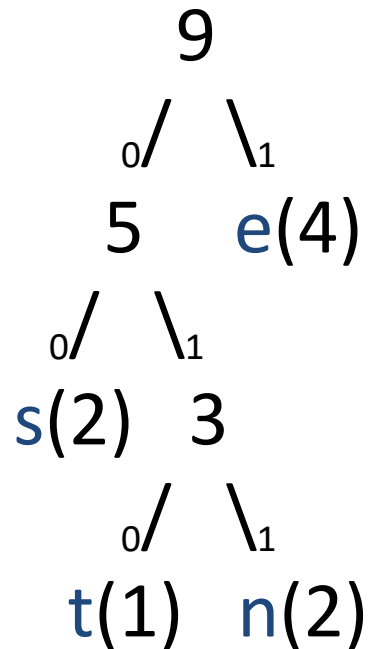
- Codes
 - A: 010
 - B: 0000
 - C: 011
 - D: 001
 - E: 1
 - F: 0001



- Note
 - None are prefixes of another

Huffman Coding

- TENNESSEE



- ENCODING

- E : 1
- S : 00
- T : 010
- N : 011

Average code length = $(1*4 + 2*2 + 3*2 + 3*1) / 9 = 1.89$

Average Code Length

Average code length =

$$\sum_{i=0,n} (\text{length} * \text{frequency}) / \sum_{i=0,n} \text{frequency}$$

$$= \{ 1(4) + 2(2) + 3(2) + 3(1) \} / (4+2+2+1)$$

$$= 17 / 9 = 1.89$$

ENTROPY

Entropy is a measure of information content: the more probable the message, the lower its information content, the lower its entropy

$$\text{Entropy} = -\sum_{i=1,n} (p_i \log_2 p_i)$$

(p - probability of the symbol)

$$\begin{aligned} &= - (0.44 * \log_2 0.44 + 0.22 * \log_2 0.22 \\ &\quad + 0.22 * \log_2 0.22 + 0.11 * \log_2 0.11) \\ &= - (0.44 * \log 0.44 + 2(0.22 * \log 0.22 + 0.11 * \log 0.11) \\ &\quad / \log 2 \\ &= 1.8367 \end{aligned}$$