

Basic concepts of image Compression technique: JPEG



Introduction

- To understand the need for compact image representation, consider the amount of data required to represent a 2 hour Standard Definition (SD) using $720 \times 480 \times 24$ bit pixel arrays.
- A video is a sequence of video frames where each frame is a full color still image.
- Because video player must display the frames sequentially at rates near 30fps, SD video data must be accessed at

$$30\text{fps} \times (720 \times 480)\text{ppf} \times 3\text{bpp} = 31,104,000 \text{ bps}$$

fps – frames per second,

ppf – pixels per frame,

bpp – bytes per pixel & bps – bytes per second



Introduction

Thus a 2 hour movie consists of

$31,104,000 \text{ bps} \times (60^2) \text{ sph} \times 2 \text{ hrs} \approx 2.24 \times 10^{11} \text{ bytes.}$

OR

224GB of data

sph = second per hour

- ❑ Twenty seven 8.5GB dual layer DVDs are needed to store it.
- ❑ To put a 2hr movie on a single DVD, each frame must be compressed by a factor of around 26.3.
- ❑ The compression must be even higher for HD, where image resolution reach $1920 \times 1080 \times 24 \text{ bits}/\text{image.}$



Data and Information

- *Data is not the same thing as information.*
- Data is the means with which information is expressed. The amount of data can be much larger than the amount of information.
- Data that provide no relevant information = *redundant data or redundancy.*
Image coding or compression has a goal to reduce the amount of data by reducing the amount of redundancy



Data Redundancy

Redundant data :

Representation that contain irrelevant or repeated information.

- n_1 = data.
- n_2 = data – redundancy (i.e., data after compression).
- **Compression ratio** = $CR = n_1/n_2$

Relative redundancy = $RD = 1 - 1/CR$



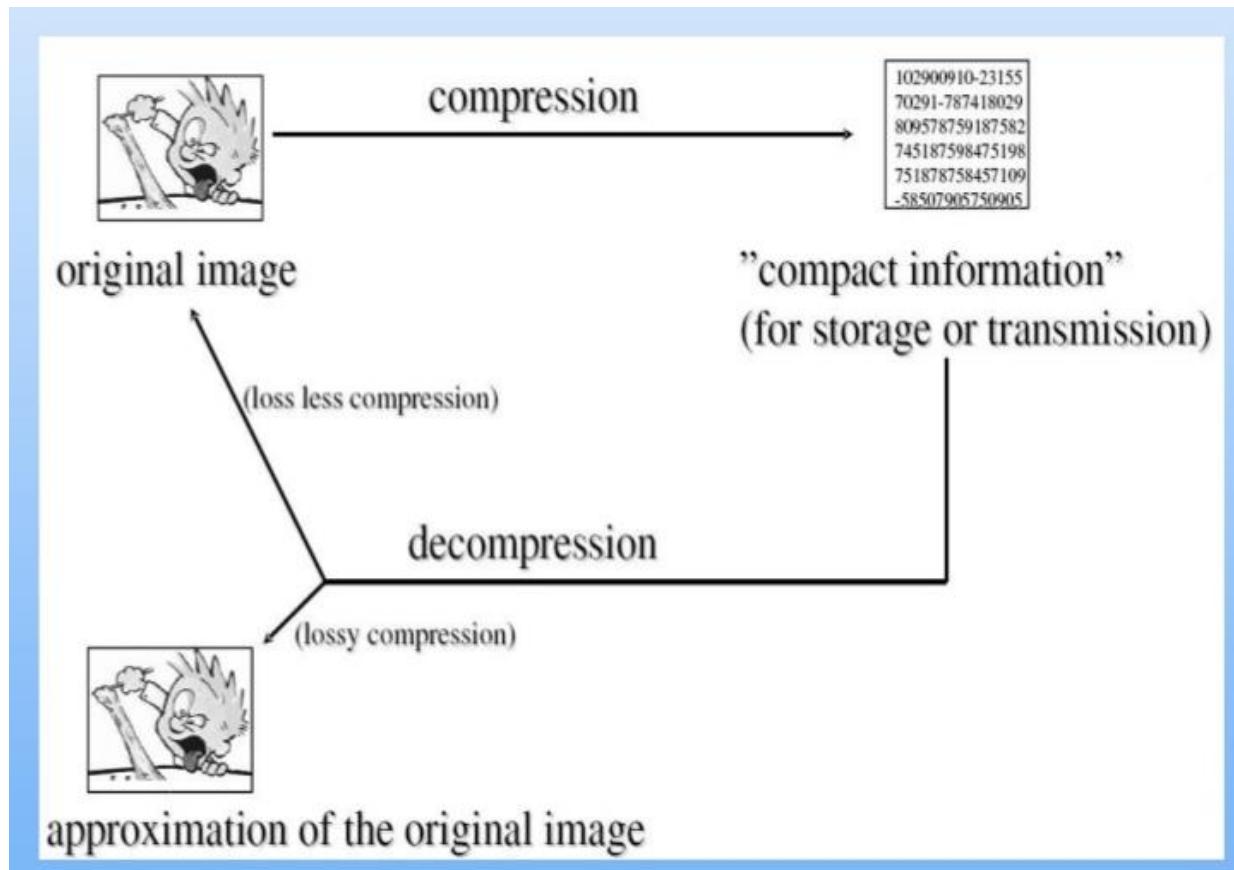
What is image compression?

- Image compression refers to the process of redundancy amount of data required **to represent the given quantity of information for digital image**. The basis of reduction process is **removal of redundant data**.



Why image compression?

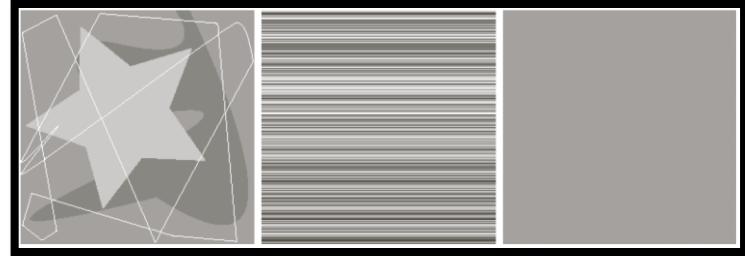
- Reducing the amount of data
- Reducing transmission time



Different types of Redundancy

→ Three redundancies in 2-D arrays:

- Coding redundancy
- Inter-pixel redundancy
- Irrelevant information



CR Coding Redundancy.

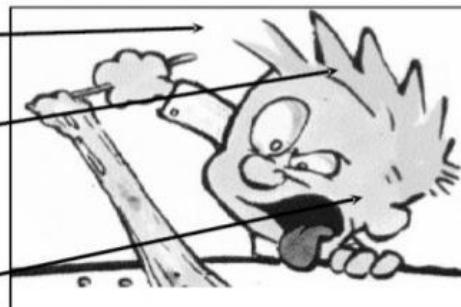
IR Interpixel Redundancy.

PVR Psycho-Visual Redundancy

CR: some graylevels are more common than others

IR: the same graylevel covers large areas

PVR: the eye can only resolve 32 graylevels locally



Data Redundancy

TYPES OF DATA REDUNDANCY

- ❑ Three principal types of data redundancies that can be identified and exploited in digital images
 - 1. Coding redundancy
 - 2. Spatial or temporal (interpixel) redundancy
 - 3. Psychovisual redundancy (irrelevant information)
- ❑ Data compression attempts to reduce one or more of these redundancy types.



Coding Redundancy

TYPES OF DATA REDUNDANCY

- ❑ **Coding redundancy**

The 8-bit codes that are used to represent the intensities in most 2-D intensity arrays contain more bits than are needed to represent the intensities.

- ❑ **Spatial or temporal (interpixel) redundancy**

Interpixel redundancy implies that pixel values are correlated (i.e., A pixel value can be reasonably predicted by its neighbors).

- ❑ **Psychovisual redundancy (irrelevant information)**

Most images contain information that is ignored by the human visual system and/or irrelevant to the intended use of the image. It is redundant in the sense that it is not used.



Coding Redundancy



r_k	$p_r(r_k)$	Code 1	$I_I(r_k)$	Code 2	$I_2(r_k)$
$r_{87} = 87$	0.25	01010111	8	01	2
$r_{128} = 128$	0.47	10000000	8	1	1
$r_{186} = 186$	0.25	11000100	8	000	3
$r_{255} = 255$	0.03	11111111	8	001	3
r_k for $k \neq 87, 128, 186, 255$	0	—	8	—	0

Inter-pixel Redundancy

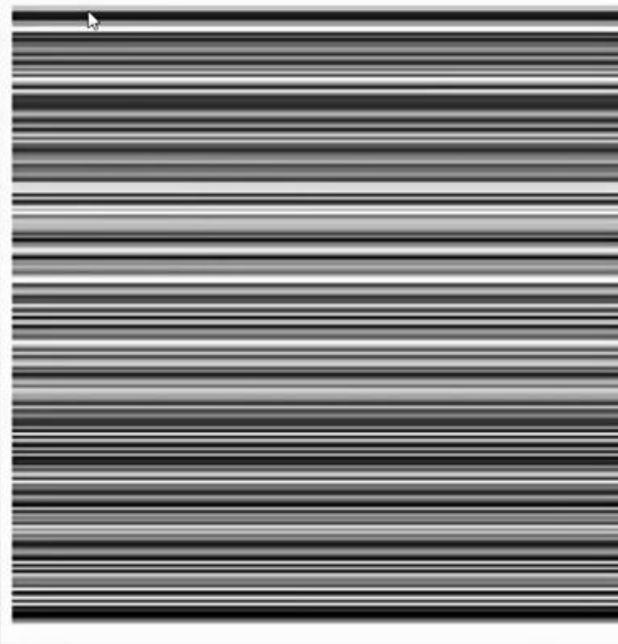
SPATIAL AND TEMPORAL REDUNDANCY

CONSIDER THE FOLLOWING IMAGE OF SIZE 256×256 .
IN THE CORRESPONDING 2-D IMAGE:

256, 255

256, 0

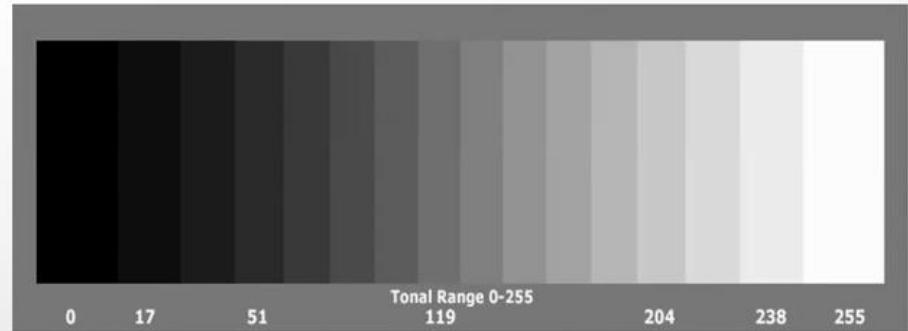
256, 1



Irrelevant information

- *Information that not used by the human visual systems*

PSYCHOVISUAL REDUNDANCY



Coding Redundancy

CODING - DEFINITIONS

- **Code:** a list of symbols (letters, numbers, bits etc.)
- **Code word:** a sequence of symbols used to represent some information (e.g., Gray levels).
- **Code word length:** number of symbols in a code word.

Example: (binary code, symbols: 0,1, length: 3)

CODE	CODE WORD
0:	000
1:	001
2:	010
3:	011
4:	100
5:	101
6:	110
7:	111

CODING REDUNDANCY

- $r_k \rightarrow$ Input Intensity Value e.g. 0 – 255 for grayscale image
- $l(r_k) \rightarrow$ No. of bits used to represent r_k
- Then average no of bits required to represent each pixel is

$$L_{avg} = \sum_{k=0}^{L-1} l(r_k) p_r(r_k)$$

Say, uniform width str. (all pixels are using 8 bits to represent intensity)

→ equal length code (fixed length code)

→ m-bit fixed length code (8 bit fixed length code in our case)

r_k	$p_r(r_k)$	Code 1	$I_I(r_k)$	Code 2	$I_2(r_k)$
$r_{87} = 87$	0.25	01010111	8	01	2
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r_k for $k \neq 87, 128, 186, 255$	0	—	8	—	0



CODING REDUNDANCY

- In this case, $l(r_k) = 8$ [each rq. 8 bits in fixed length]

$$L_{avg} = 8 \sum_{k=0}^{L-1} p_r(r_k) = 8 * 1 = 8$$

- Coding redundancy tries to reduce L_{avg}
- Thus, total no. of bits required to represent an $M \times N$ image is MNL_{avg}
- The code can be an equal length code, or variable length code.

r_k	$p_r(r_k)$	Code 1	$I_1(r_k)$	Code 2	$I_2(r_k)$
$r_{87} = 87$	0.25	01010111	8	01	2
$r_{128} = 128$	0.47	10000000	8	1	1
$r_{186} = 186$	0.25	11000100	8	000	3
$r_{255} = 255$	0.03	11111111	8	001	3
r_k for $k \neq 87, 128, 186, 255$	0	—	8	—	0

Cont.

CODING REDUNDANCY



Code 1 → equal length code
Code 2 → variable length code
(obt by Huffman Coding)
Min bits → higher probability

r_k	$p_r(r_k)$	Code 1	$I_1(r_k)$	Code 2	$I_2(r_k)$
$r_{87} = 87$	0.25	01010111	8	01	2
$r_{128} = 128$	0.47	10000000	8	1	1
$r_{186} = 186$	0.25	11000100	8	000	3
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r_k for $k \neq 87, 128, 186, 255$	0	—	8	—	0

CODING REDUNDANCY

- For code 1, $L_{avg} = 8$
- On the other hand, using code 2, the average length of the encoded pixels is

$$L_{avg} = 0.25(2) + 0.47(1) + 0.25(3) + 0.03(3) = 1.81 \text{ bits}$$

- The resulting compression and corresponding relative redundancy are

Thus, 77.4% of the data in the original 8-bit 2-D intensity array is redundant.

$$C = \frac{256 \times 256 \times 8}{256 \times 256 \times 1.81} \approx 4.42$$

$$R = 1 - \frac{1}{4.42} = 0.774$$



Compression Method

HUFFMAN CODING

- A measure to reduce coding redundancy
- Most popular coding redundancy technique
- Variable length code
- Min length code is assigned to one with highest probability



Cont.

HUFFMAN CODING

Say, Image size: 10 x 10 (5 bit image)

Frequency:

$$a_2 = 40 \quad a_6 = 30$$

$$a_1 = 10$$

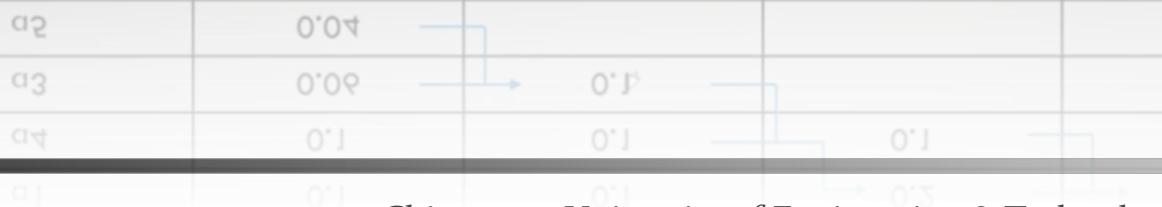
$$a_4 = 10$$

$$a_3 = 6$$

$$a_5 = 4$$

$$P(a_2) = 40/100 = 0.4$$

Symbols (like intensity levels)	Probabilities (sorted)	Source Reduction (do till two values are left) (Maintain in sorted order here as well)			
		1	2	3	4
a_2	0.4	0.4	0.4	0.4	0.6
a_6	0.3	0.3	0.3	0.3	0.4
a_1	0.1	0.1	0.2	0.3	
a_4	0.1	0.1	0.1		
a_3	0.06	0.1			
a_5	0.04				



Cont.

Symbols (like intensity levels)	Probabilities (sorted)	Source Reduction (do till two values are left) (Maintain in sorted order here as well)			
		1	2	3	4
a2	0.4 1	0.4	0.4	0.4	0.6 0
a6	0.3 00	0.3	0.3	0.3 00	0.4 1
a1	0.1 011	0.1	0.2 010	0.3 01	
a4	0.1 0100	0.1 0100	0.1 011		
a3	0.06 01010	0.1 0101			
a5	0.04 01011				

Encoded String: 010100111100

Decoding : a3 a1a2 a2 a6

Parameters:

- Average length of code

$$L_{avg} = 0.4 * 1 + 0.3 * 2 + 0.1 * 3 + 0.1 * 4 + 0.06 * 5 + 0.04 * 5 = 2.2 \text{ bits/symbol}$$

- Total no. of bits to be transmitted

$$10 * 10 * 2.2 = 220 \text{ bits}$$

- Entropy = 2.1396

- How much you saved = $\frac{10 * 10 * 5 - 10 * 10 * 2.2}{10 * 10 * 5} = 0.56 = 56\%$



Reference

- Digital Image Processing, Rafael C.Gonzalez, Richard E.Woods, 3rd Edition.
- Digital Image Processing: Part II, Huiyu, Jiahua Wu, Jianguo Zhang.
- Jpeg Image Compression Using Discrete Cosine Transform - A Survey,A.M.Raid, W.M.Khedr, M. A. El-dosuky and Wesam Ahmed, International Journal of Computer Science & Engineering Survey (IJCSES) Vol.5, No.2, April 2014.
- Tarek Ouni and Mohamed Abid , International Journal of Signal Processing, Image Processing and Pattern Recognition Vol. 5, No. 3, September, 2012 , Scan Methods and Their Application in Image Compression .

