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# 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 x 480 x 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  $1920 \times 1080 \times 24 \text{ bits/image}$ .



# 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.

- ▣  $n1 = \text{data}$ .
- ▣  $n2 = \text{data} - \text{redundancy}$  (i.e., data after compression).
- ▣ **Compression ratio** =  $CR = n1/n2$

$$\text{Relative redundancy} = RD = 1 - 1/CR$$



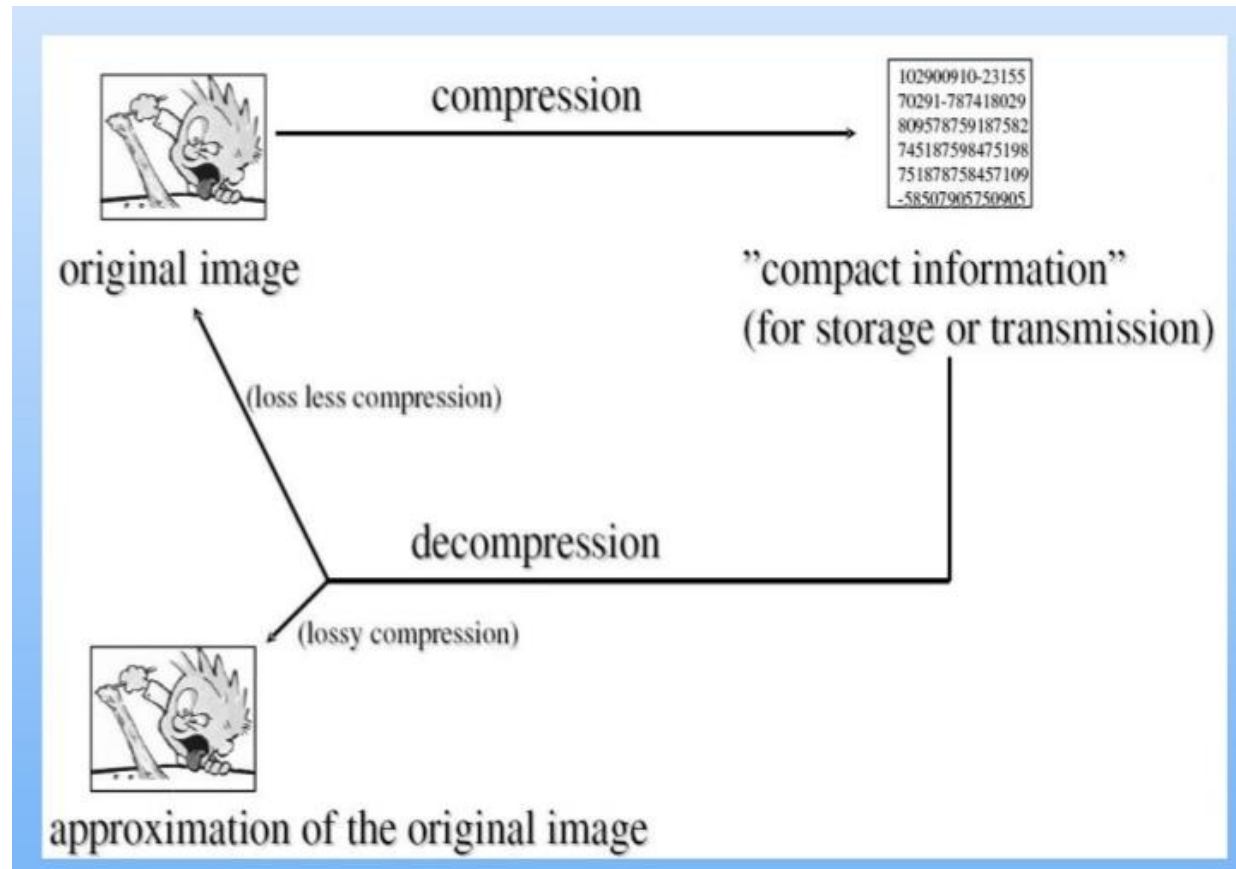
# What is image compression?

- 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:

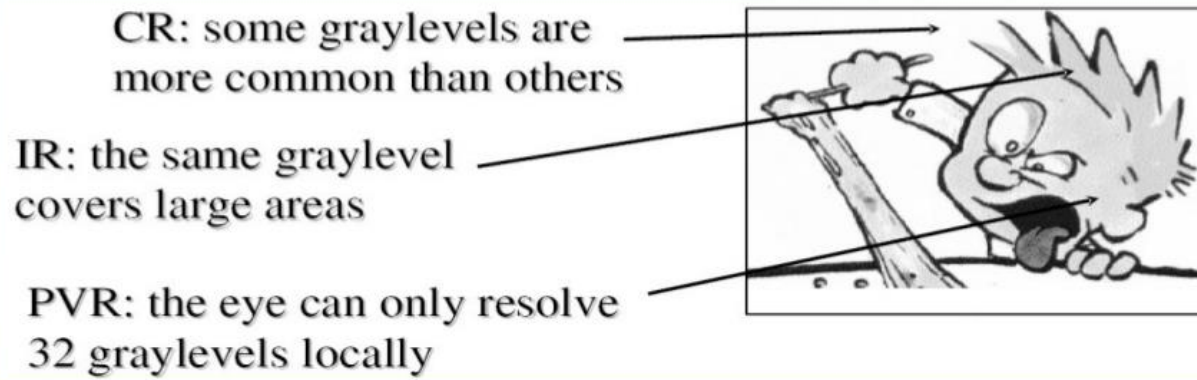
- Coding redundancy
- Inter-pixel redundancy
- Irrelevant information



**CR** Coding Redundancy.

**IR** Interpixel Redundancy.

**PVR** Psycho-Visual Redundancy





# 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	$l_1(r_k)$	Code 2	$l_2(r_k)$
$r_{87} = 87$	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

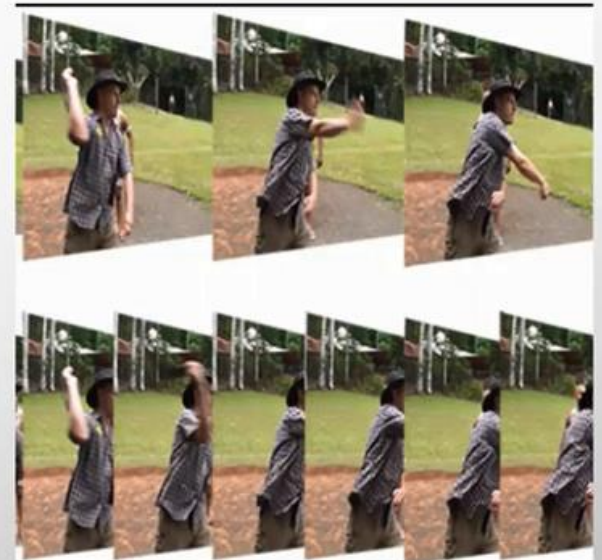
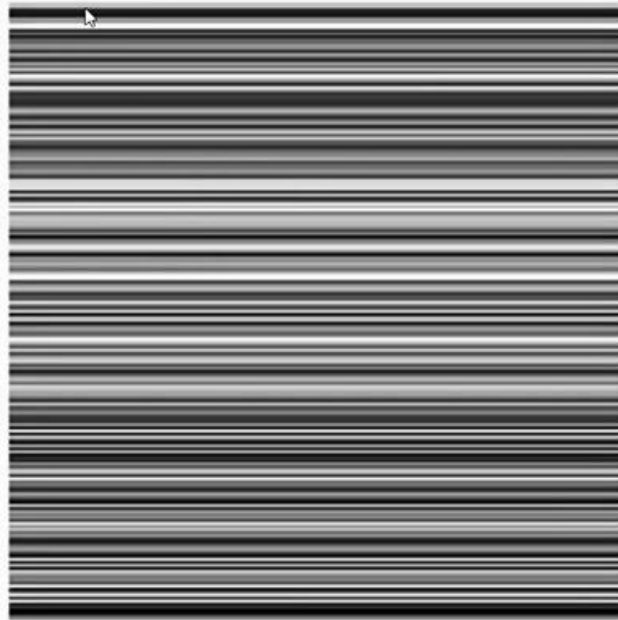
## SPATIAL AND TEMPORAL REDUNDANCY

CONSIDER THE FOLLOWING IMAGE OF SIZE  $256 \times 256$ .  
IN THE CORRESPONDING 2-D IMAGE:

256, 255

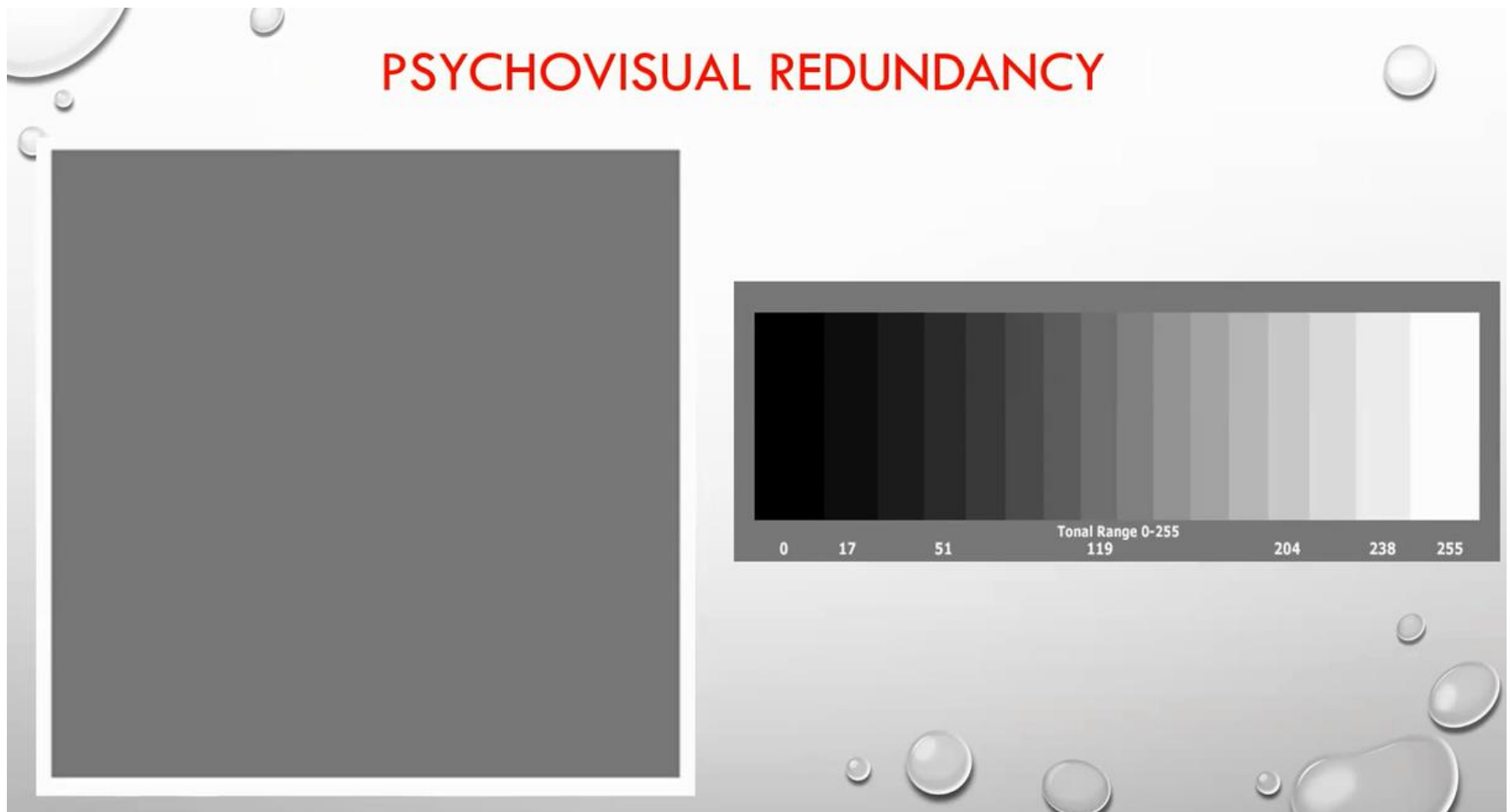
256, 0

256, 1



# Irrelevant information

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





# 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	4: 100
1: 001	5: 101
2: 010	6: 110
3: 011	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)  
 $\rightarrow$  equal length code (fixed length code)  
 $\rightarrow$  m-bit fixed length code (8 bit fixed length code in our case)

$r_k$	$p_r(r_k)$	Code 1	$l_1(r_k)$	Code 2	$l_2(r_k)$
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# 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	$l_1(r_k)$	Code 2	$l_2(r_k)$
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$r_{128} = 128$	0.47	10000000	8	1	1
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# 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	$l_1(r_k)$	Code 2	$l_2(r_k)$
$r_{87} = 87$	0.25	01010111	8	01	2
$r_{128} = 128$	0.47	10000000	8	1	1
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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:

$\alpha_2 = 40$        $\alpha_6 = 30$        $\alpha_1 = 10$        $\alpha_4 = 10$        $\alpha_3 = 6$        $\alpha_5 = 4$

$P(\alpha_2) = 40/100 = 0.4$

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
$\alpha_2$	0.4	0.4	0.4	0.4	0.6
$\alpha_6$	0.3	0.3	0.3	0.3	0.4
$\alpha_1$	0.1	0.1	0.2	0.3	
$\alpha_4$	0.1	0.1	0.1		
$\alpha_3$	0.06	0.1			
$\alpha_5$	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 a1 a2 a2 a6

## Parameters:

1. Average length of code

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

2. Total no. of bits to be transmitted

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

3. Entropy = 2.1396

4. 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, 3<sup>rd</sup> Edition.
- Digital Image Processing: Part II, Huiyu, Jiahua Wu, Jianguo Zhang.
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- 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 .

