# EXERCISE 3: IMAGE PROCESSING PART 1

#### Multimedia Database SS 23

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#### 1a) Image Formats

#### Vector Image

- Makes use of sequential commands or mathematical statements which place lines or shapes in a 2-D or 3-D environment.
- Each control point has a definite (x,y)
  position and determines a graphic
  primitive, which may be shapes,
  curves, splines
- Occupies less space,
- Suitable for printing, since vector images scale well

#### Raster Image

- Raster Images are pixel based
- Use bitmaps to store information
- Consumes more memory
- Scaling up leads to degradation.





#### 1b) Image Formats: GIF

- Color model: RGB
- Color depth: 8 Bits per pixel range of 24 Bits color space (256/16 M)
- Lempel-Ziv-Welch (LZW) compression lossless
- Small file size
- Can represent animations, logos, etc...
- but not suitable for photos due to color limitation.





#### 1b) Image Formats: PNG

- Color model: RGB/RGBA
- Color depth: variable grey scale (1-16 Bpp) Indexed (1-8 Bpp 24 Bits Color palette), TrueColor (24/48 Bpp)
- Compression "Deflate": Combination of LZSS and Huffman Coding (lossless)
- Doesn't support animation





#### 1b) Image Formats: JPEG

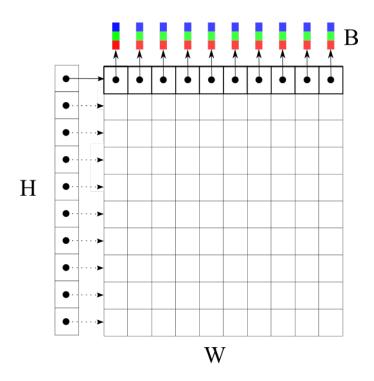
- Color model: RGB
- Color depth: 24 Bits per pixel (most common)
- DCT-based compression, lossy
- Doesn't support animation, transparency

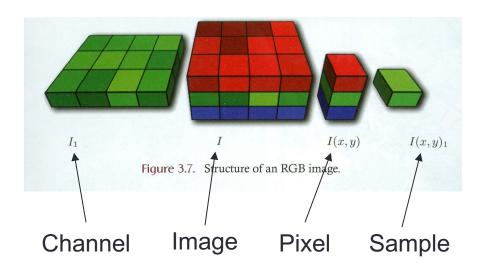




### 2a) Image Processing

- Naive internal representation: 3-D Array
- int[][] raster









### 2) Image Processing

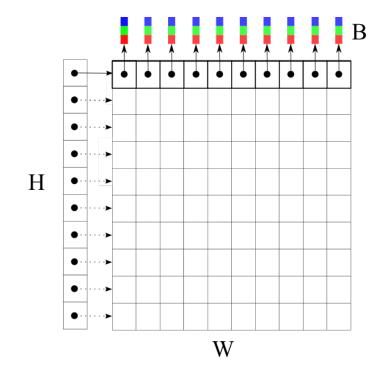
- 24bpp RGB image: 1 Byte = 8 bits per sample
  - Size(raster) = w x h x b samples (sample = channel value)
  - → Efficient implementation would require at most w x h x b Bytes of memory
- Practically -> more
  - Java: multi-dimensional arrays = array of an array





### 2a) Image Processing

- Memory Consumption using 3-D raster array
- References (Pointers):
  - H+WH (4 Bytes for each)
- Object Header (Array length attribute) :
  - 1+H +WH (4 Bytes each)
- •Total amount of unnecessary memory consumed by this 3-D representation:
  - 4(H+WH) + 4(WH + H +1)
  - ~ 8 WH
- Image data
  - 3WH
- Total (W=H=12)
  - -4(H+WH) + 4(WH + H + 1) + 4(3WH) = 2980 Bytes







### 2b) Image Processing

- Till now, a sample was stored in an integer value (4-Bytes per sample).
- However, most image processing applications require a color depth of at most 24 bits per pixel (1-Byte per sample).
- Solution: pack the entire pixel into a single int value.

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#### 2b) Image Processing

- Memory Consumption using 2-D array
- References (Pointers) :
  - H (4 Bytes for each)
- Array length attribute:
  - 1+H (4 Bytes each)
- Total amount of unnecessary memory consumed
  - 2H+1
  - ~ 2H
- Image data:
  - WH (4 Bytes each)
- Total (W=H=12)
  - 4 \* (2H + 1 + WH) = 676 bytes
  - Improvement by a factor 4.4





### 2c) Unpacking

#### Operations to pack and unpack a pixel

- Right-shift operator (A>>B); it shifts each bit of A to the right by B bits, and fills the least significant (rightmost) by zero.
- ▶ Left-shift operator (A<<B); it shifts each bit of A to the left by B bits, and fills the least significant (leftmost) by zero.</p>

#### Bitwise logical operators:

• AND: &

• OR:

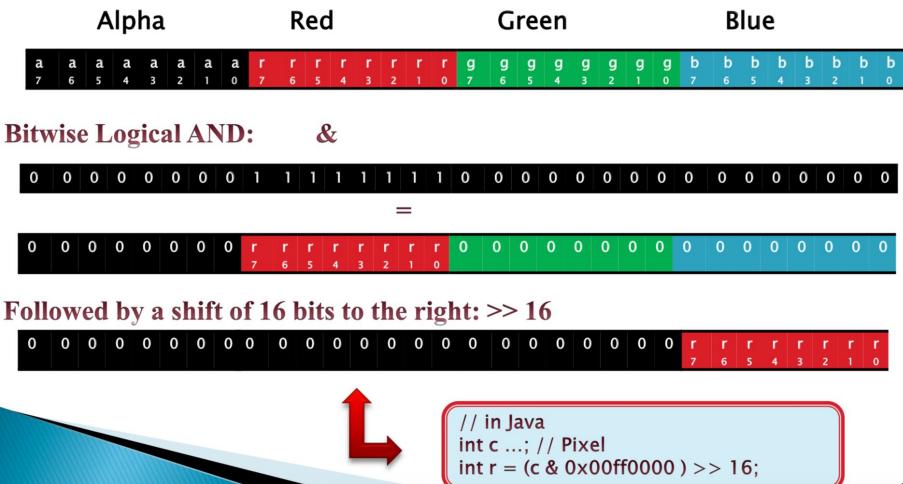
XOR: ^

Negation: ~





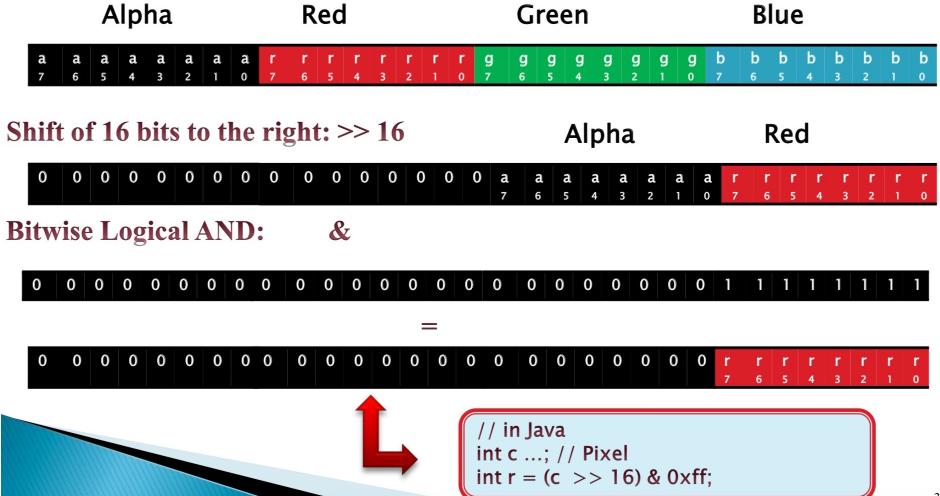
### 2c)







## 2c)







#### 2d) Image Resolution

- Screen Size (Diagonal) = 6.1 inches
- Aspect Ratio = 19.5 : 9 → H = (19.5/9) x W
- Resolution = 828 x 1792 pixels

According to Pythagora's Theorem:

 $W^2 + \{ (19.5/9) \times W \}^{2=} (6.1)^2$ 

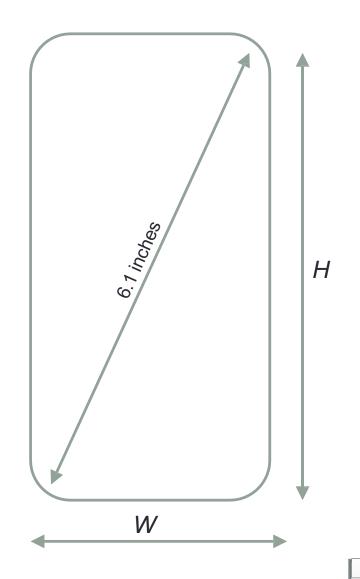
Solving for W:

W = 2.5561 inches

Therefore,

2.5561 inches = 828 pixels

1 inch  $\cong$  324 pixels





#### 3a) Uniform Quantization

- Quantization: For an Image I, containing m different colors  $C = \{C_1, C_2, ..., C_m\}$ , where  $m \le 2^{24}$  (For a 3 x 8 bit image), we have to replace the original colors with a set  $C' = \{C_1, C_2, ..., C_n\}$ , with  $n \le m$  in such a way that we have minimal perceptible degradation in the resulting image.
- Uniform Quantization.
  - Each axis of the color space is divided into equal sized segments. (The number of these segments depends on storage requirements. For e.g. 8 for red, 8 for green and 4 for blue for 8 bit storage)
  - Each of the original colors is then mapped to the region which it falls in (256 for 8 bits).
  - Representative colors for each region is then the average of all the colors mapped to that region

#### 3a) Median Cut Quanization

Source: An Overview of Color Quantization Techniques by Steven Segenchuk https://web.cs.wpi.edu/~matt/courses/cs563/talks/color\_quant/CQindex.html

- Median Cut Algorithm
  - Find the smallest box which contains all the colors in the image.
  - Sort the enclosed colors along the longest axis of the box.
  - Split the box into 2 regions at median of the sorted list.
  - Repeat the above process until the original color space has been divided into 256 regions.
  - The representative colors are found by averaging the colors in each box.





### 3b) (Random) Noise Dithering

•Reduce effects of quantization(such as color banding) by adding uniformly distributed white noise (dither signal) to the input image prior to quantization.

Quantization (Without Dithering):

$$P(x,y) = Q(I(x,y)) = floor(\frac{I(x,y)}{256} 2^{b})$$



Quantization (With Dithering):

$$P(x,y) = Q(I(x,y) + noise(x,y))$$



