#### Histograms

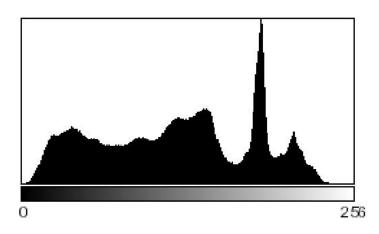
CMPUT 206

Instructor: Nilanjan Ray

Source: http://www.imagingbook.com/

# What is an image histogram





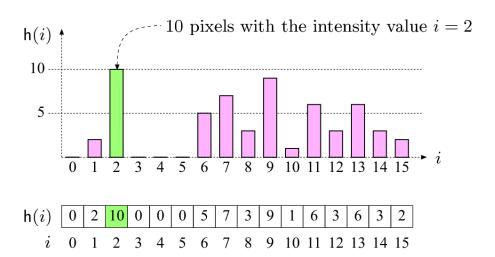
Count: 1920000 Min: 0 Mean: 118.848 Max: 251

StdDev: 59.179 Mode: 184 (30513)

### How to define a histogram

h(i) = the number of pixels in I with the intensity value i

Or, formally, 
$$h(i) = \operatorname{card} \{(u,v) \mid I(u,v) = i \}$$



### Example: Histogram

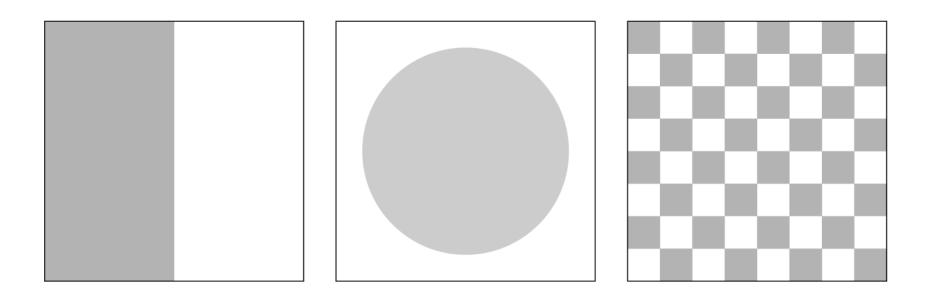
Consider a 3-by-4 image with 8 gray levels

| 3 | 0 | 1 | 2 |
|---|---|---|---|
| 4 | 3 | 6 | 7 |
| 3 | 2 | 1 | 4 |

Its histogram h is as follows:

| Gray level i   | 0 | 1 | 2 | 3 | 4 | 5 | 6 | 7 |
|----------------|---|---|---|---|---|---|---|---|
| Histogram h[i] | 1 | 2 | 2 | 3 | 2 | 0 | 1 | 1 |

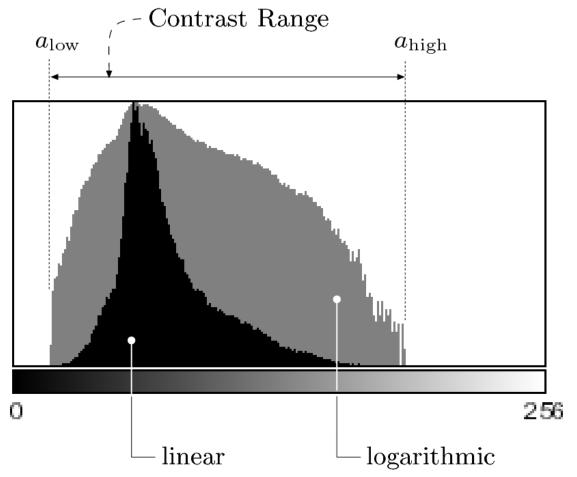
# Are histograms unique?



Three images with same histogram

Going from an image to its histogram, what information is lost?

# Histogram interpretation

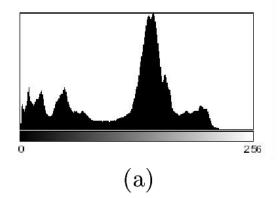


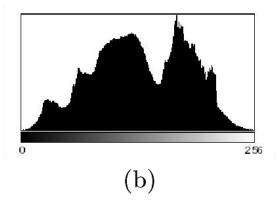
# Exposure

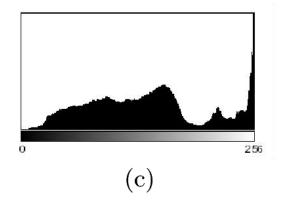










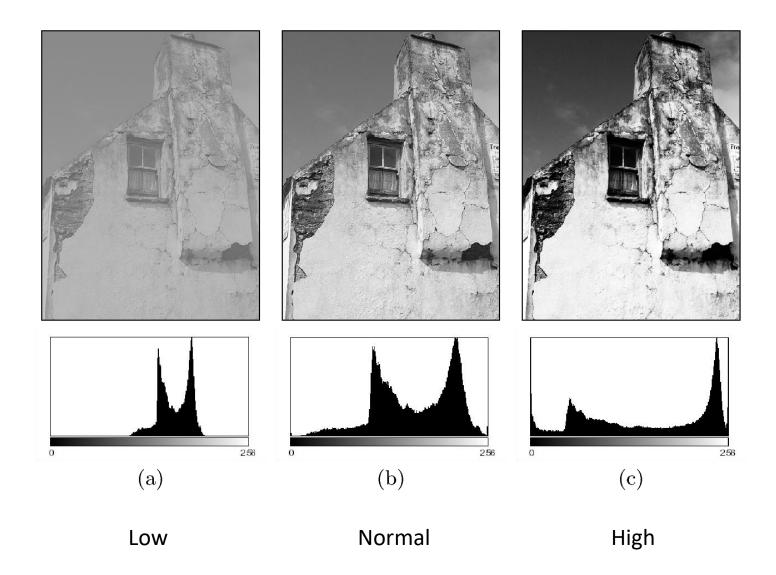


Under exposed

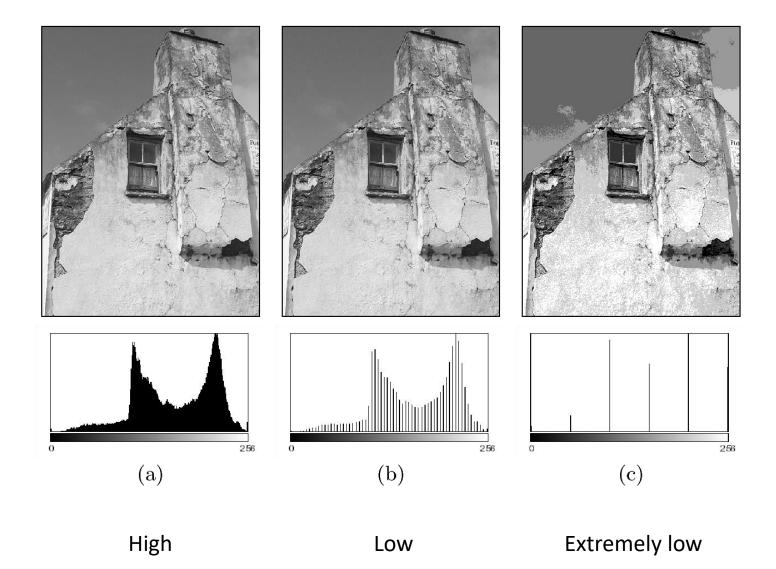
Normal exposure

Over exposed

#### Contrast



# Dynamic range



#### How to compute a histogram

- Let's compute histogram for a L gray level image
- Initialize histogram array: h[n] = 0, n=0,...,L-1
- Let height and width be the number of row and columns of image I, respectively
- For i from 0 to height-1
  - For j from 0 to width-1
    - h[ l[i, j] ] += 1

### Cumulative histogram

$$H(i) = \sum_{j=0}^{i} h(j) \quad \text{for } 0 \le i < K$$

$$\mathsf{H}(i) = \begin{cases} \mathsf{h}(0) & \text{for } i = 0 \\ \mathsf{H}(i-1) + \mathsf{h}(i) & \text{for } 0 < i < K \end{cases}$$

$$H(K-1) = \sum_{j=0}^{K-1} h(j) = M \cdot N$$

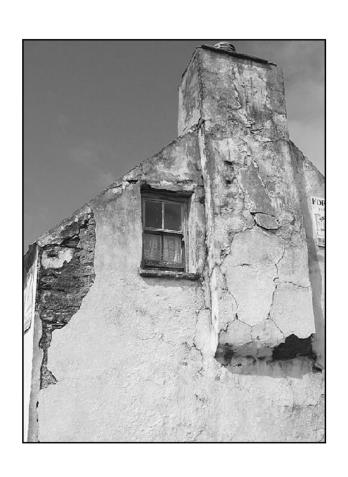
# Example: Cumulative histogram

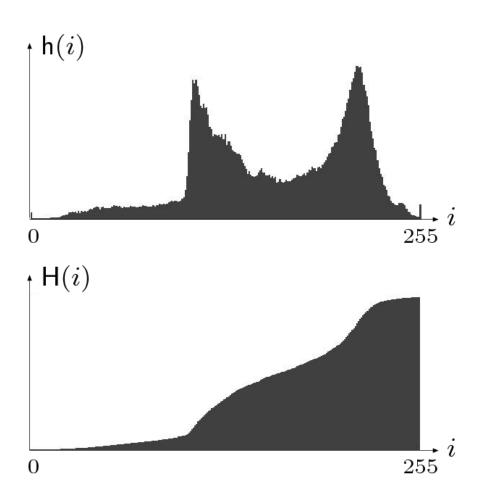
#### 8 gray level Image

| 3 | 0 | 1 | 2 |
|---|---|---|---|
| 4 | 3 | 6 | 7 |
| 3 | 2 | 1 | 4 |

| Gray level i                 | 0 | 1 | 2 | 3 | 4  | 5  | 6  | 7  |
|------------------------------|---|---|---|---|----|----|----|----|
| Histogram h[i]               | 1 | 2 | 2 | 3 | 2  | 0  | 1  | 1  |
| Cumulative<br>histogram H[i] | 1 | 3 | 5 | 8 | 10 | 10 | 11 | 12 |

# Example of cumulative histogram





### Normalized histogram

- Suppose h(i) is a histogram
- It's normalized version is defined as

$$p(i) = h(i) / \sum_{i=0}^{K-1} h(i) = h(i) / (MN)$$

• One can think of the normalized histogram as a probability mass function: p(i) means the probability of a pixel value to be i

#### Normalized cumulative histogram

- Suppose H(i) is a cumulative histogram
- It's normalized version is defined as: H(i)/(MN)

# Example: Normalized histogram and normalized cumulative histogram

8 gray level Image

| 3 | 0 | 1 | 2 |
|---|---|---|---|
| 4 | 3 | 6 | 7 |
| 3 | 2 | 1 | 4 |

| Gray level i                    | 0    | 1    | 2    | 3    | 4     | 5     | 6     | 7    |
|---------------------------------|------|------|------|------|-------|-------|-------|------|
| Histogram h[i]                  | 1    | 2    | 2    | 3    | 2     | 0     | 1     | 1    |
| Cumulative<br>histogram H[i]    | 1    | 3    | 5    | 8    | 10    | 10    | 11    | 12   |
| Normalized<br>histogram         | 1/12 | 2/12 | 2/12 | 3/12 | 2/12  | 0     | 1/12  | 1/12 |
| Normalized cumulative histogram | 1/12 | 3/12 | 5/12 | 8/12 | 10/12 | 10/12 | 11/12 | 1    |

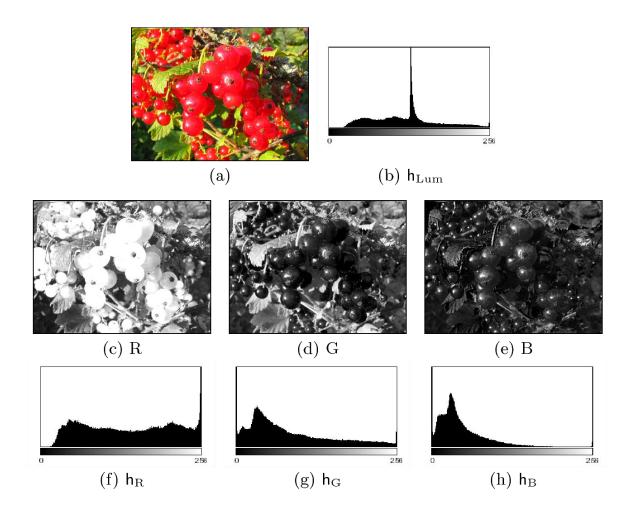
### Binning for histogram

Sometimes, we need to create bins for histograms. Suppose, we have a 14 bit image. So, the range of values is 0 to 16384. If we want to create a histogram of 256 bins, then it would look like:

$$\begin{array}{llll} \mathsf{h}(0) & \leftarrow & 0 \leq I(u,v) < & 64 \\ \mathsf{h}(1) & \leftarrow & 64 \leq I(u,v) < & 128 \\ \mathsf{h}(2) & \leftarrow & 128 \leq I(u,v) < & 192 \\ \vdots & \vdots & \vdots & \vdots \\ \mathsf{h}(j) & \leftarrow & a_j \leq I(u,v) < & a_{j+1} \\ \vdots & \vdots & \vdots & \vdots \\ \mathsf{h}(255) & \leftarrow & 16320 \leq I(u,v) < 16384 \end{array}$$

Notice that the bin widths are same.

# Color image histogram



# Matching histograms and some applications

- Histogram matching has many applications image processing
  - Image segmentation
  - Tracking
  - Content-based image retrieval
  - **—** ...
- A quick way for content-based image retrieval can be based on histogram matching
  - In a database of images, find out those that have histograms similar to the histogram of a query image

#### Histogram match metrics

- Before matching two histograms, they are converted into normalized histograms
- Several metrics exist for matching normalized histograms
  - Bhattacharya coefficient
  - Kullback-Liebler divergence
  - Diffusion distance (<a href="http://www.ist.temple.edu/~hbling/publication/Ling-2006">http://www.ist.temple.edu/~hbling/publication/Ling-2006</a> &Okada06cvpr.pdf

**—** ....

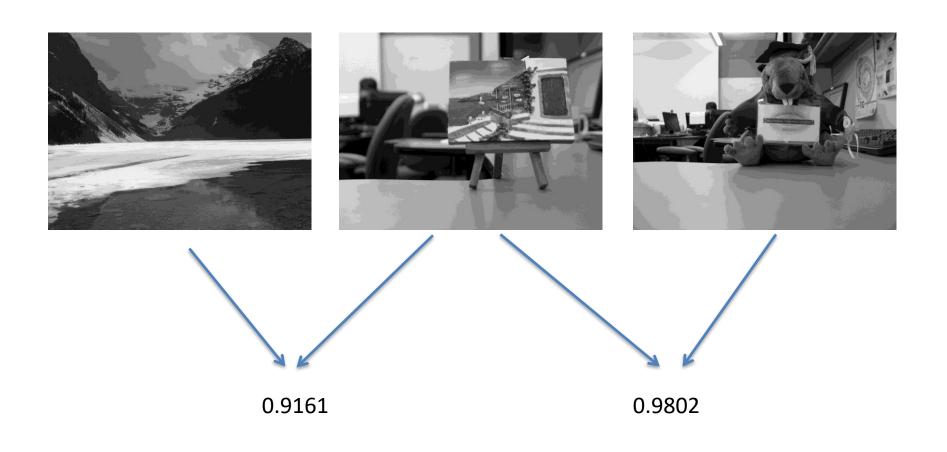
# Matching two normalized histograms with Bhattacharya coefficient

- Suppose two p(i) and q(i) are two normalized histograms
- Bhattacharya coefficient is defined as

$$BC(p,q) = \sum_{i=0}^{K-1} \sqrt{p(i)q(i)}$$

- For a perfect match BC is 1, for a complete mismatch BC is 0
- A higher BC value implies a better match

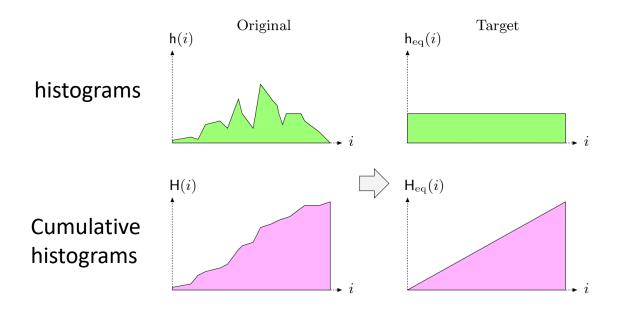
#### Bhattacharya coefficient: Examples



Notice that BC value is higher for the similar pair of images

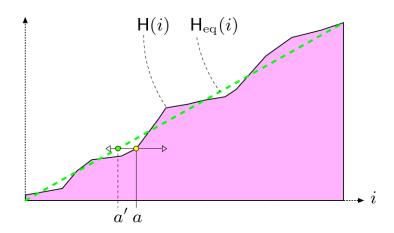
# Histogram equalization

Also known as "histogram flattening"



### Histogram equalization...

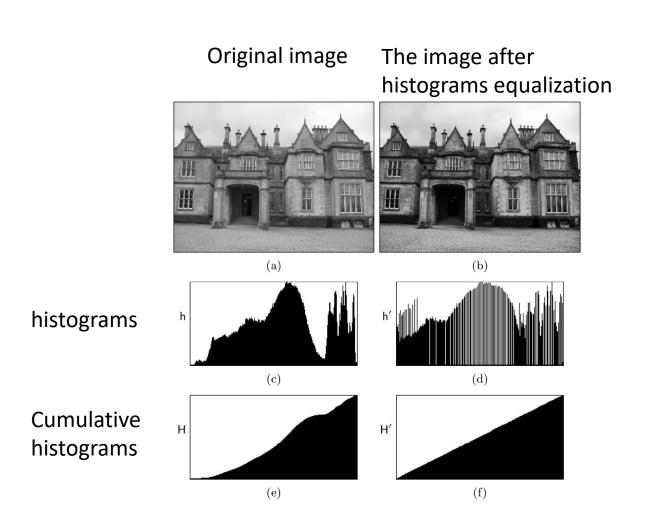
Cumulative histograms



$$a' = floor[\frac{K-1}{MN}H(a) + 0.5]$$

Can we derive this formula?

### Histogram equalization: example



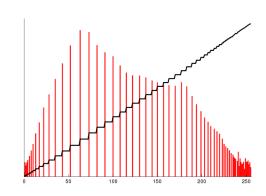
# A histogram equalization example from Wikipedia



0 50 100 150 200 250

Original image





Histograms (red) and cumulative histograms (black)

After histogram equalization

#### Histogram specification

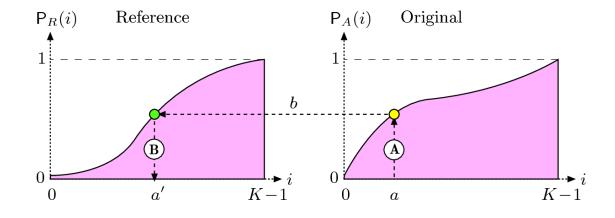
Prob[output image pixel value  $\leq a'$ ] = Prob[input image pixel value  $\leq a$ ]

$$\Rightarrow P_{R}(a') = P_{A}(a)$$

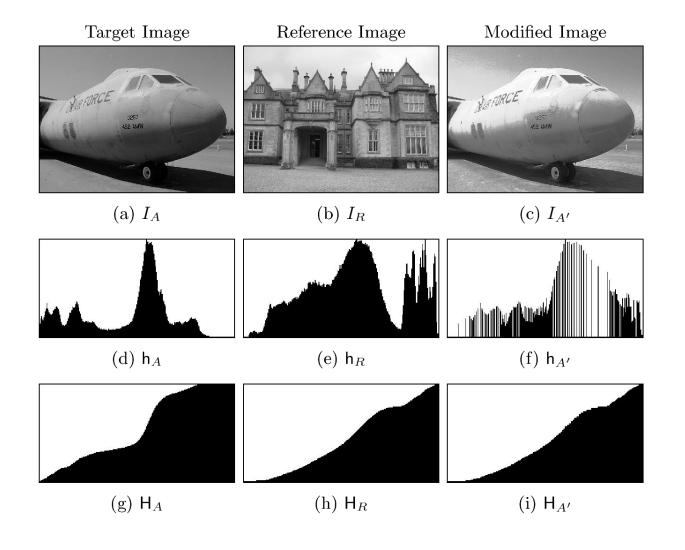
$$\Rightarrow a' = P_R^{-1}(P_A(a))$$

P<sub>A</sub> and P<sub>R</sub> are normalized cumulative histograms





# Histogram specification: example



# Local (Adaptive) histogram equalization

- Histogram equalization does not work well when the distribution of pixel values varies a lot over local windows in an image.
- So, doing histogram equalization within a sliding window is often more useful than the global histogram equalization.
- By the way, what is a <u>sliding window</u>?

#### Local histogram equalization: example

http://scikit-image.org/docs/dev/auto\_examples/color\_exposure/plot\_local\_equalize.html

