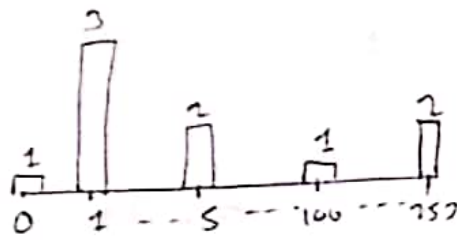


4a

## Histogram and Point Operators

### \* Histogram

1	5	255
1	100	255
5	1	0



Gray Scale  
Image

Histogram

⇒ A histogram is a frequently used tool for analyzing images.

$h(g) \Rightarrow$  Number of pixel that take value  $g$ .



⇒ Probability that a random pixel has value  $g$ .

$$P(g) = \frac{h(g)}{N} \quad \{ N = I \times J \}$$

⇒ Histogram can be calculated with linear complexity.  $O(N)$

### \* Cumulative Histogram

$$H(g) = \sum_{x=0}^g h(x) = \text{Number of pixel with } g(i,j) \leq g$$

$$P(g) = \frac{h(g)}{N}$$

$$F(g) = \frac{H(g)}{N}$$

## \* Mean, Variance, Median of Histogram

⇒ All three values are a characteristic of the image.

⇒ Mean describes the brightness.

⇒ Variance describes contrast.

⇒ Median is a robust description of the brightness.

$$\mu_g = \frac{1}{N} \sum_i \sum_j g(i,j) \rightarrow O(N)$$

$$\mu_g = \frac{1}{N} \sum_g g h(g) \rightarrow O(1)$$

$$\sigma_g^2 = \frac{1}{N-1} \sum_i \sum_j (g(i,j) - \mu_g)^2 \rightarrow O(N)$$

$$\sigma_g^2 = \frac{1}{N-1} \sum_g (g - \mu_g)^2 h(g) \rightarrow O(1)$$

$$\boxed{\text{med}(g) = F^{-1}(0.5)}$$

## \* Types of Operator

① Global Operator

② Local operator

③ Point operator

## \* Point Operator

→ The point operator maps the pixel value to a new value only based on the value and the location of the input pixel in the image.

$$b(i,j) = f(a(i,j), p)$$

↓                      ↓                      ↓                      ↘  
Output            operator            Input            Parameter

## \* Linear Transformation

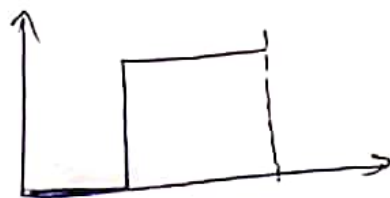
$$\boxed{b(i,j) = k + m a(i,j)} \quad \left\{ p = [k, m]^T \right\}$$

$$\boxed{\mu_b = k + m \mu_a} \quad \boxed{\sigma_b = |m| \sigma_a}$$

## \* Nonlinear Transformations

⇒ Example: thresholding

$$b(a) = \begin{cases} b_0 & \text{if } a < T \\ b_1 & \text{otherwise} \end{cases}$$



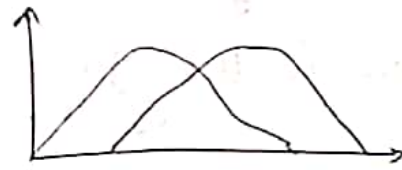
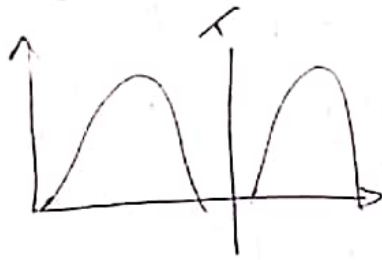
{ Binary Image }

## Application

⇒ Foreground/background Separation.



## \* Foreground and Background Separation

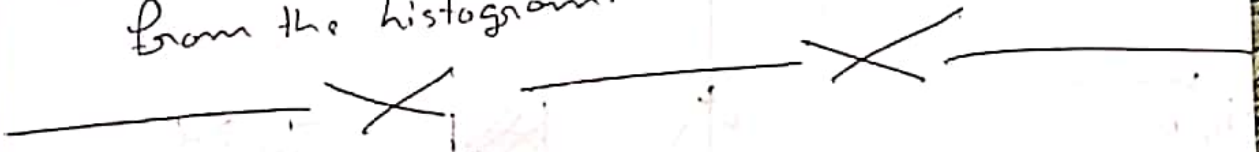


⇒ Often, the ratio is unknown but we might assume a bi-modal distribution of gray values.

⇒ We can use a sum of two Gaussians to represent the bi-modal distribution.

$$P(g) = \alpha G(g; \mu_1, \sigma_1^2) + (1-\alpha) G(g; \mu_2, \sigma_2^2)$$

⇒ The parameters  $\alpha, \mu_1, \sigma_1, \mu_2, \sigma_2$  are estimated from the histogram.



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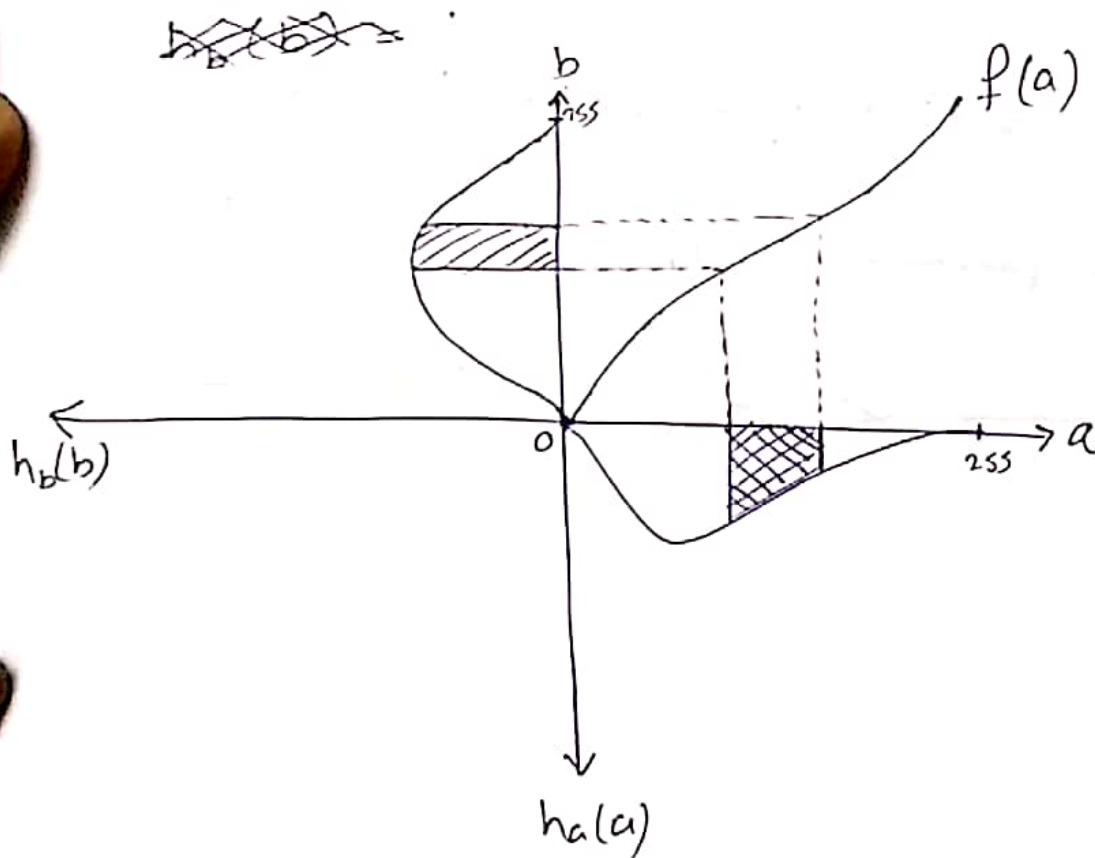
\* How do the operators Affect the Histogram of the Image?



→ Monotonous function  $b = f(a)$

→ Histogram of image  $a(i,j)$  is  $h_a(a)$

→ Compute the histogram  $h_b(b)$  of image  $b$ .

↑ Goal



$\Rightarrow$   and  should have same areas, as function is monotonic.

→ If slope of  $f(a)$  is  $> 1$ , histogram of  $b$  is being stretched and if slope of  $f(a) < 1$ , then histogram of  $b$  is being compressed.

$$h_a(a) da = h_b(b) db$$

$$h_b(b) = \frac{h_a(a)}{\left| \frac{db}{da} \right|} = \frac{h_a(a)}{|f'(a)|}$$

$$h_b(b) = \frac{h_a(f^{-1}(b))}{|f'(f^{-1}(b))|}$$

### \* Histogram Equalization

→ For this to happen,

$$h_b(b) = \text{const} = \frac{h_a(a)}{\left| \frac{db}{da} \right|}$$

$$\text{let } \frac{1}{K} = \frac{h_a(a)}{\left| \frac{db}{da} \right|}$$

$$\Rightarrow db = K h_a(a) da$$

$$\Rightarrow b = K \int_{\alpha=0}^{\alpha=a} h_a(\alpha) d\alpha + C$$

$\downarrow$   
 $H(a)$  {Cumulative  
 histogram}

$$\Rightarrow b(a) = K H_a(a) + C$$

$$b(0) = 0$$

$$b(255) = 255$$

$$K = \frac{255}{N - H(0)}$$

$$C = -H(0) \frac{255}{N - H(0)}$$



$$b(a) = \frac{255}{N - H(0)} (H(a) - H(0))$$

### \* Noise Variance Equalization

→ Goal: adjust the variance of the intensities to a fixed value.

→ Useful for statistical analysis of images

→ Realized by a monotonic point operator.

$$\sigma_a^2 = ma$$

Goal:  $\sigma_b = \sigma_0$  for all intensities

⇒ Variance propagation yields

$$\sigma_b = \left| \frac{db}{da} \right| \sigma_a = \sigma_0$$

$$\Rightarrow \frac{db}{da} \sigma_a = \sigma_0 \quad \left\{ \text{Monotonic function} \right\}$$

$$db = \frac{\sigma_0}{\sigma_a} da = \frac{\sigma_0}{\sqrt{m}} a^{-1/2} da \quad \left\{ \text{as } \sigma_a^2 = ma \right\}$$

$$\Rightarrow b = \frac{2\sigma_0}{\sqrt{m}} \sqrt{a} + C$$

$$\begin{aligned} \Rightarrow a=0 &\rightarrow b=0 \\ a=255 &\rightarrow b=255 \end{aligned}$$



$$C=0$$

$$\sigma_0 = \frac{255}{\sqrt{255}} \frac{\sqrt{m}}{2} = \frac{1}{2} \sqrt{255m}$$

## \* Point Operator on Multiple Image

$$\Rightarrow b = f(a_1, \dots, a_m)$$

• After two input images

$$b = f(a_1, a_2)$$

## \* Masking

$$b(i, j) = \begin{cases} a_2(i, j) & \text{if } a_1(i, j) = 1 \\ 0 & \text{otherwise} \end{cases}$$

## \* Montage of Image

$$b(i, j) = \begin{cases} a_2(i, j) & \text{if } a_1(i, j) = 1 \\ a_3(i, j) & \text{otherwise} \end{cases}$$

## \* Point Operators

$\Rightarrow$  Point operators can be efficiently computed using look-up table.

