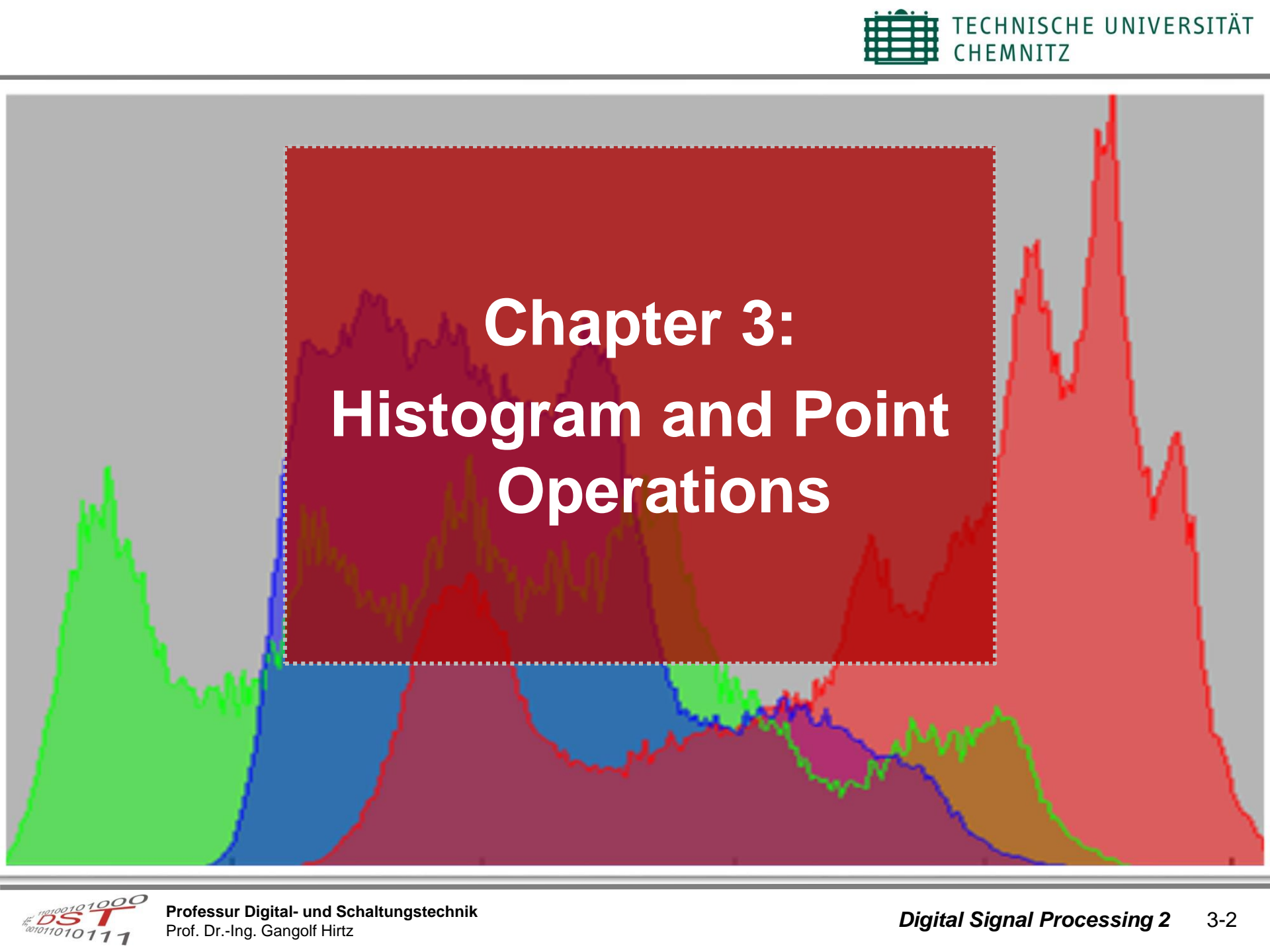


# Solution Exercise 2: Thresholding



# Chapter 3: Histogram and Point Operations

## Histogram calculation

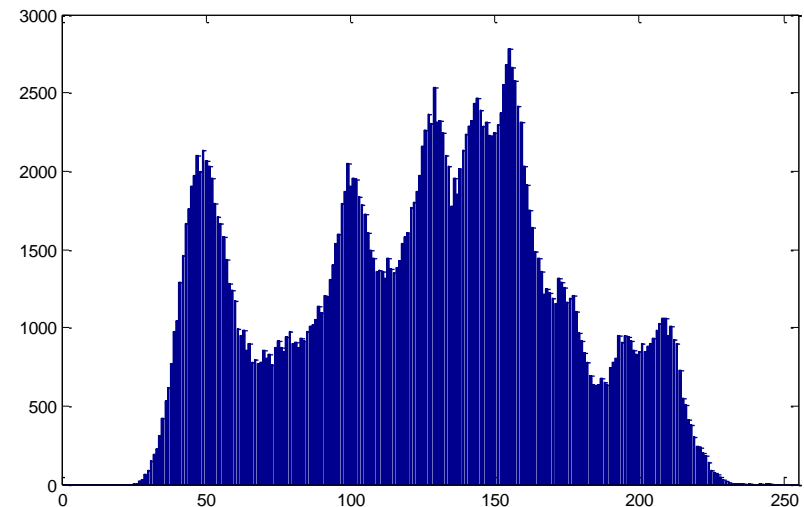
- Each pixel represents a quantized measurement of the intensity
- The Histogram is a vector, which contains one element of each quantization level
  - N-bit resolution  $\rightarrow 2^N$  quantization levels  $\rightarrow K=2^N$  histogram elements
  - E.g. 8-bit  $\rightarrow K=2^8=256$  entries
- Each element contains the number of pixels whose value corresponds to the index of the element
- Example: calculation of a histogram in MATLAB:

```
% read image
I = imread('image.bmp');
[rows cols] = size(I);

% initialize histogram with zeros
h = zeros(256,1);

% iterate over all pixels
for n = 1:rows
    for m = 1:cols
        % increment histogram value at index I(n,m)
        h(I(n,m)+1) = h(I(n,m)+1) + 1;
    end
end

% show histogram
bar(h);
```



Histogram distribution of Lena

## Mean value (expectation value)

- Provides information about the brightness of the image (mean value  $\mu$ )

Calculation from the histogram: 
$$\mu = \sum_{i_{\min}}^{i_{\max}} i \cdot h_n(i)$$

Calculation from the image: 
$$\mu = \frac{1}{M \cdot N} \sum_{m=0}^{M-1} \sum_{n=0}^{N-1} I(n, m)$$

## Variance

- Provides information about the dispersion of the values

Calculation from the histogram: 
$$\sigma^2 = \sum_{i_{\min}}^{i_{\max}} (i - \mu)^2 \cdot h_n(i)$$

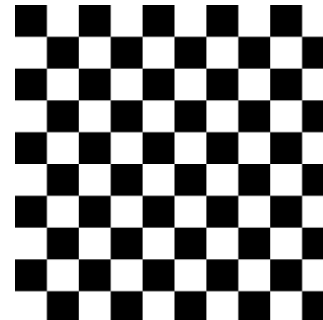
Calculation from the image: 
$$\sigma^2 = \frac{1}{M \cdot N} \sum_{m=0}^{M-1} \sum_{n=0}^{N-1} I^2(n, m) - \mu^2$$

Examples:



$$\mu = 127$$

$$\sigma^2 = 0$$

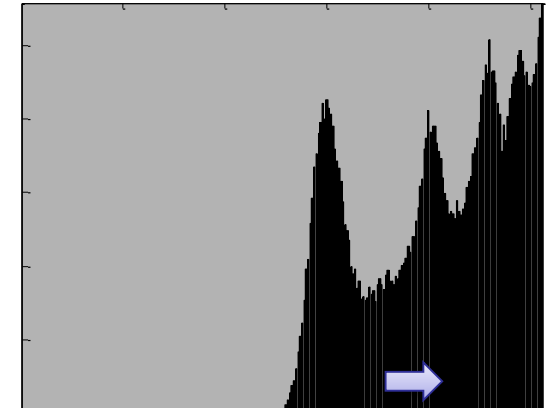
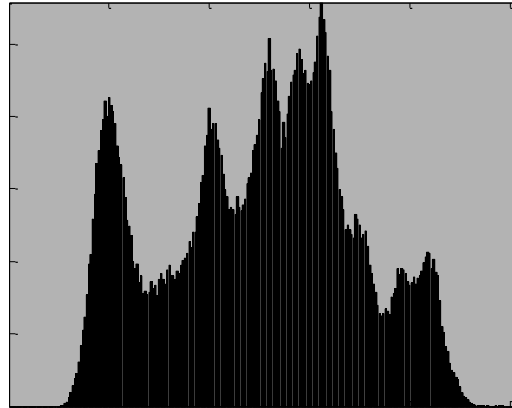
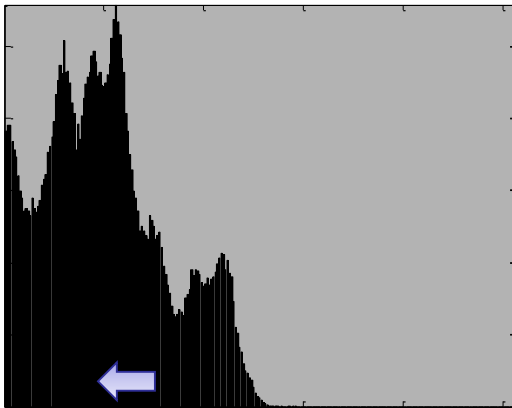


$$\mu = 127,5$$

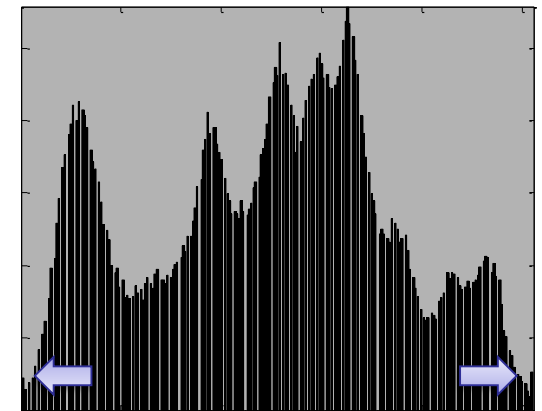
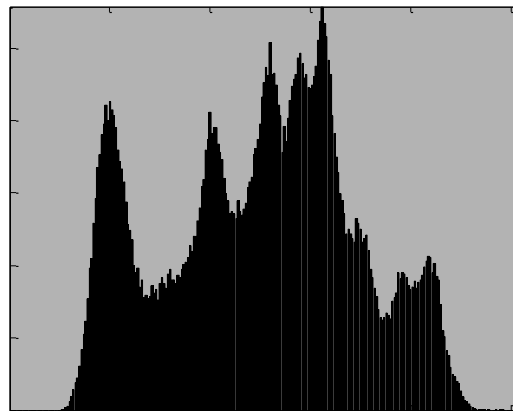
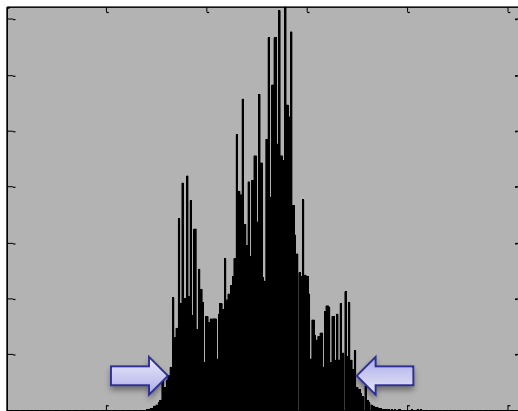
$$\sigma^2 = 127,5^2 = 16256,25$$

## Examples

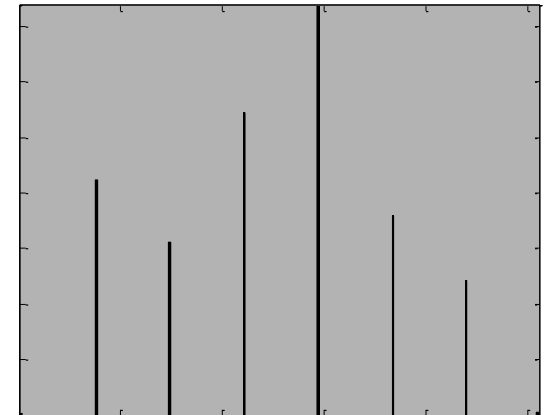
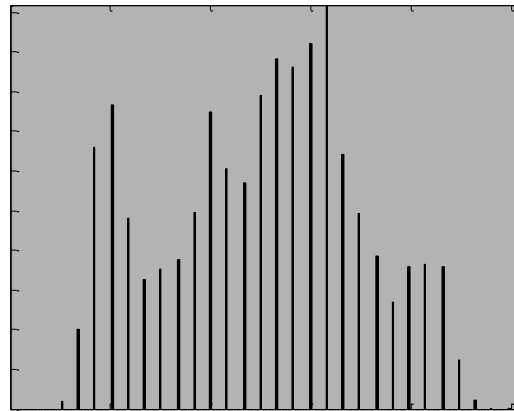
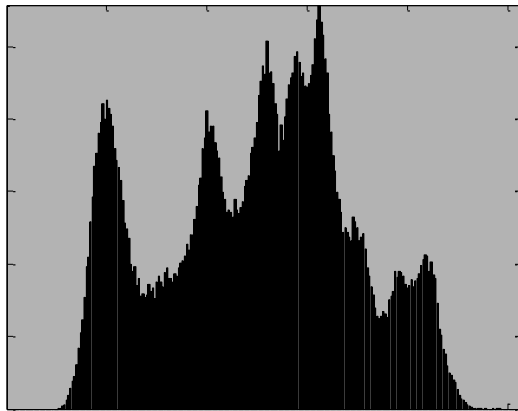
### Brightness



## Contrast



## Quantization noise



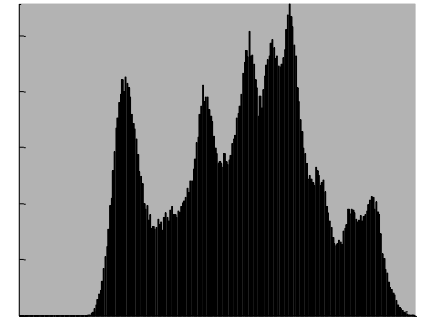
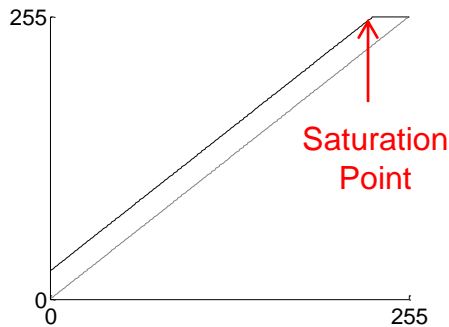


## Brightness Adjustment

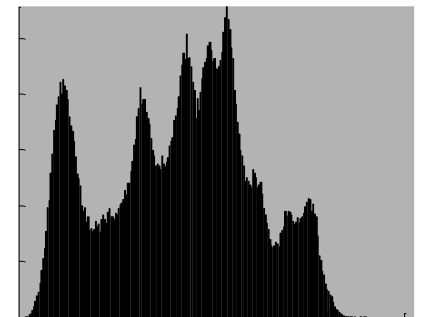
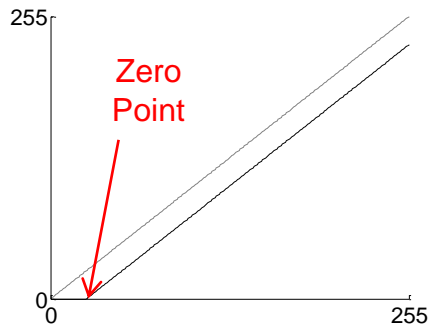
$$f(m, n) = I(m, n) + a \quad a \in [0, 255]$$

$$J(m, n) = \begin{cases} f(m, n), & \text{if } 0 \leq f(m, n) \leq 255 \\ 0, & \text{if } f(m, n) < 0 \\ 255, & \text{if } f(m, n) > 255 \end{cases}$$

$a > 0$ :



$a < 0$ :





## Contrast Adjustment

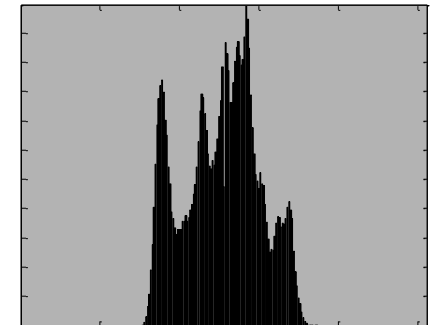
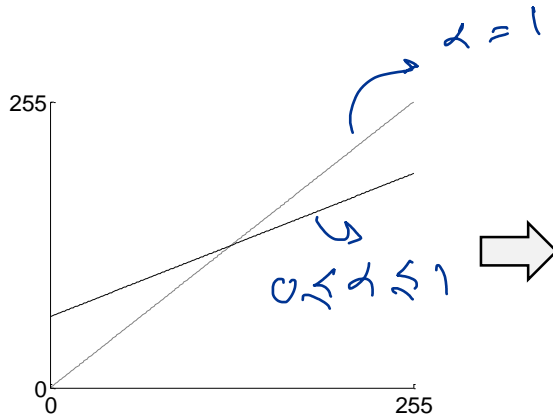
$\alpha \rightarrow$  Contrast factor.

$$f(m,n) = a(I(m,n) - s) + s \quad a, s \in [0,255]$$

$$J(m,n) = \begin{cases} f(m,n), & \text{if } 0 \leq f(m,n) \leq 255 \\ 0, & \text{if } f(m,n) < 0 \\ 255, & \text{if } f(m,n) > 255 \end{cases}$$

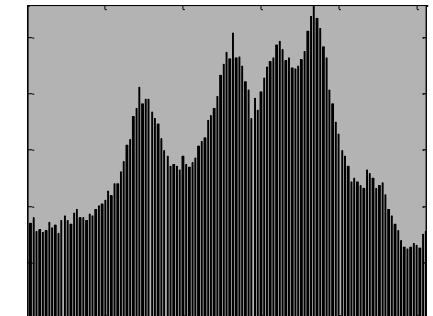
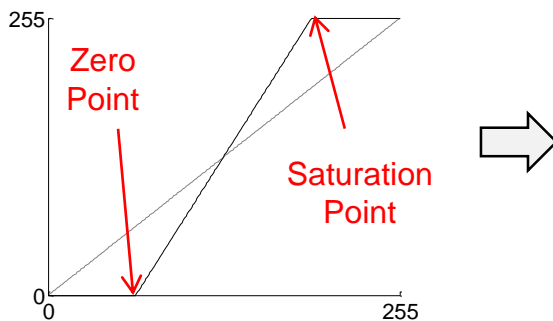
$s$  is the center of the  
contrast function,  
e.g.  $s = 127$

$0 \leq a < 1$ :



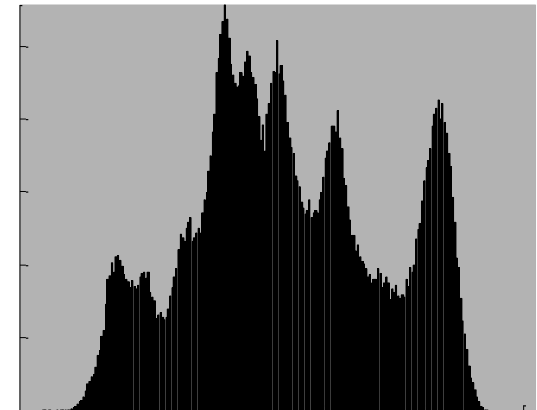
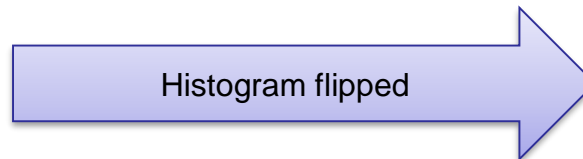
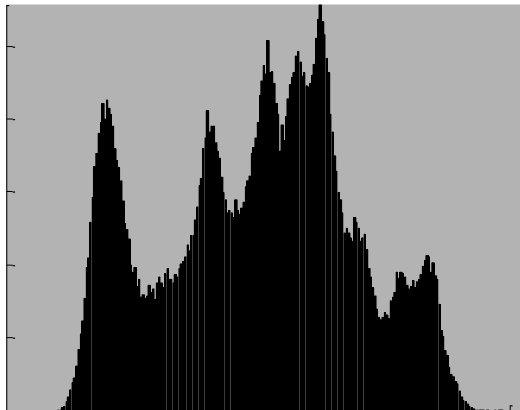
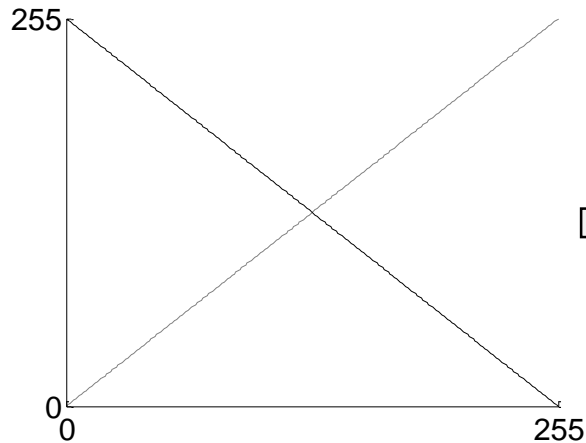
$s = 127$

$a > 1$ :



## Inversion

$$J(m,n) = 255 - I(m,n)$$



## Binning (quantization)

- Images with a high bit depth would generate huge histograms
  - Example: 14-bit  $\rightarrow K=2^{14}=16384$  histogram entries
- To reduce the size of the histogram, binning can be done
- The intensity values are counted in  $B$  intervals (bins)  $[i_j, i_{j+1})$
- For equally sized bins, the interval size is  $k_B=K/B$
- The bin index can be calculated with  $j = \left\lfloor I(m, n) \frac{B}{K} \right\rfloor$

**Example:** 14-bit image, histogram with 256 bins

$$\begin{aligned}h(0) &\leftarrow 0 && \leq I(n, m) < 64 \\h(1) &\leftarrow 64 && \leq I(n, m) < 128 \\M &\leftarrow M && \leq I(n, m) < M \\h(j) &\leftarrow i_j && \leq I(n, m) < i_{j+1} \\M &\leftarrow M && \leq I(n, m) < M \\h(255) &\leftarrow 16320 && \leq I(n, m) < 16384\end{aligned}$$

## Third exercise

- Implement some histogram operations

## Expected Output (3 of 6)

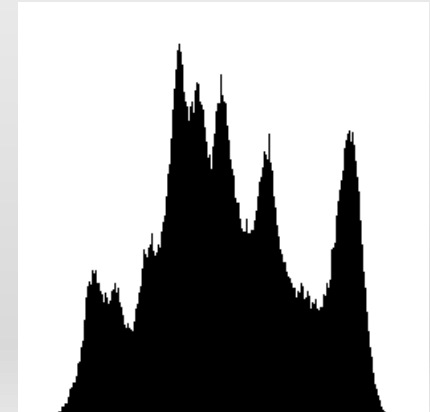
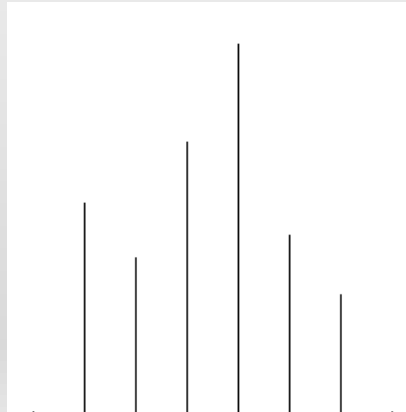
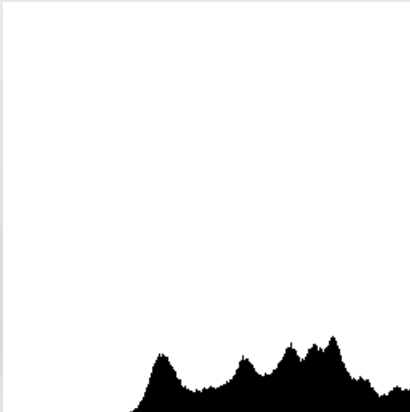
Adjusted Brightness



Quantized



Inverted



**That is all for today.**