

# Image Enhancement

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## References

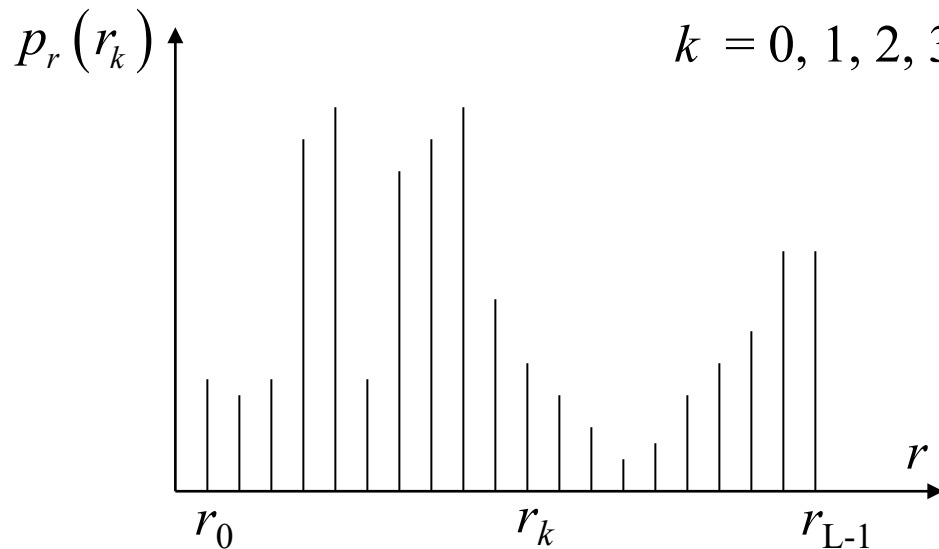
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- Rafael C. Gonzalez and Richard E. Woods, Digital Image Processing, Third Edition, Pearson Education, 2008:
  - Histogram Equalization: Chapter 3.3.1
- [https://en.wikipedia.org/wiki/Histogram\\_equalization](https://en.wikipedia.org/wiki/Histogram_equalization)

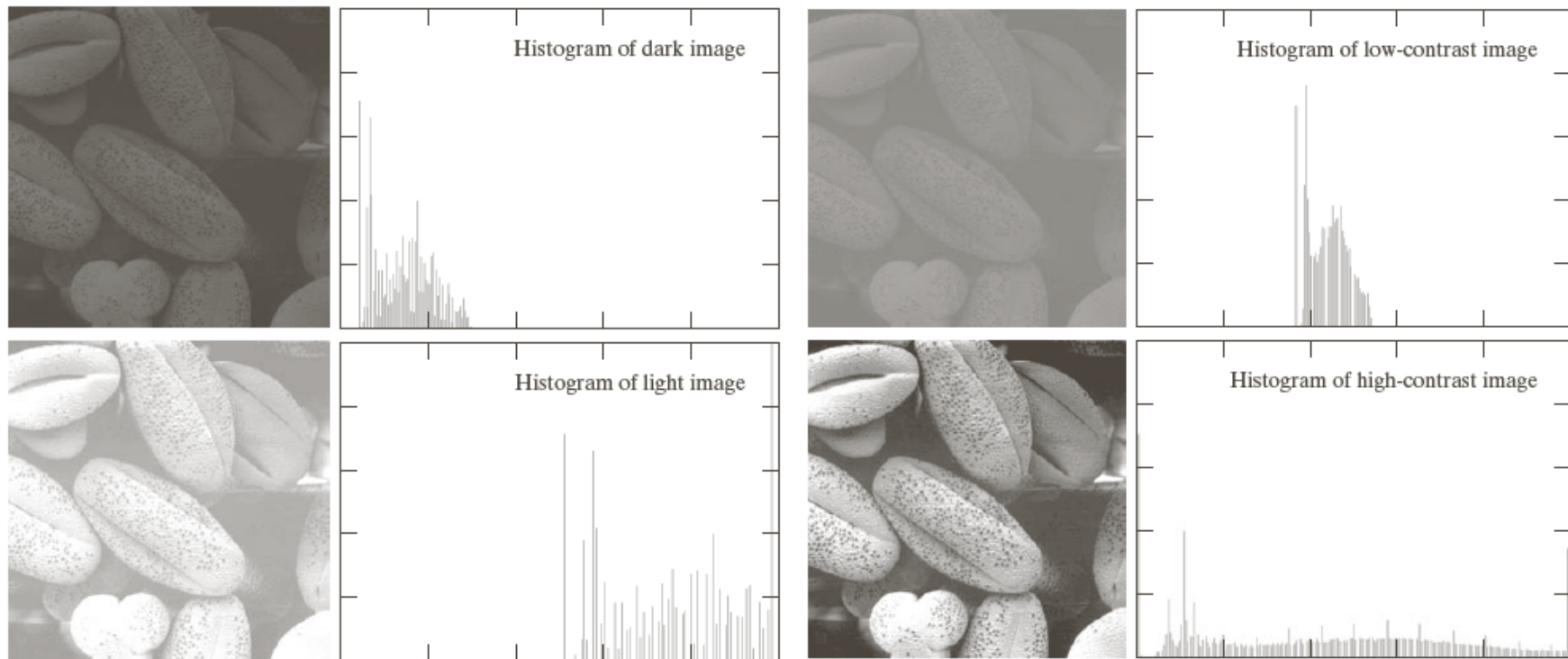
- A digital image with  $L$  gray levels  $r_k$ . The probability of occurrence of gray level  $r_k$  is given by

$$p_r(r_k) = \frac{n_k}{N} \quad \text{where} \quad \begin{aligned} n_k &= \text{number of pixels with gray level } r_k \\ N &= \text{total number of pixels in an image} \end{aligned}$$

$$k = 0, 1, 2, 3, \dots, L-1$$



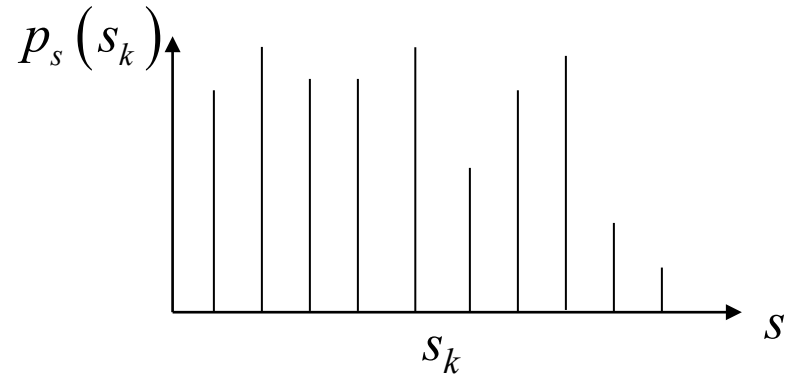
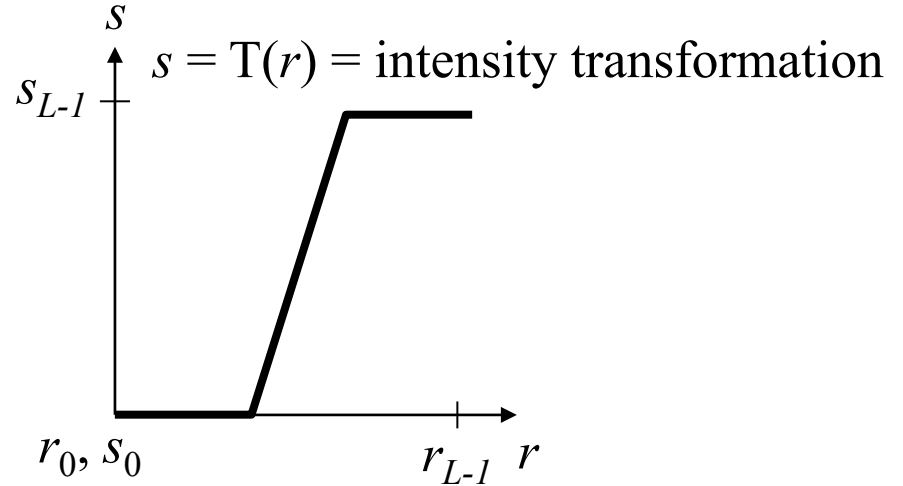
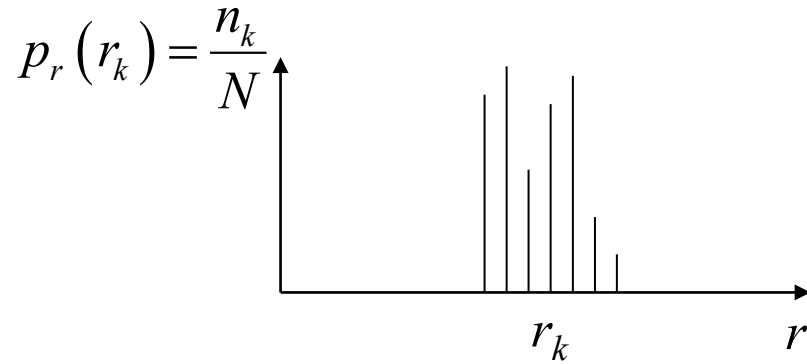
## Images with different histograms



**FIGURE 3.16** Four basic image types: dark, light, low contrast, high contrast, and their corresponding histograms.

## Contrast stretching

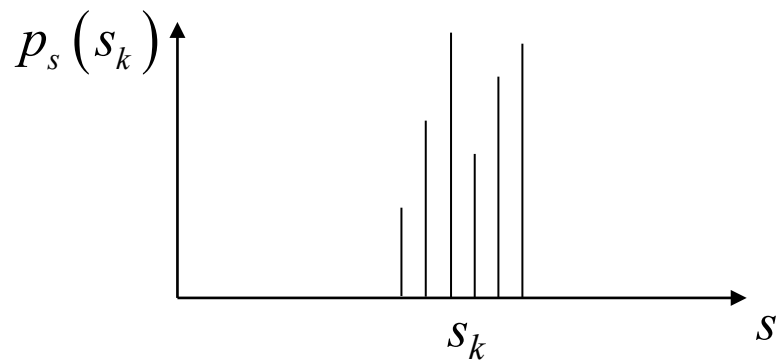
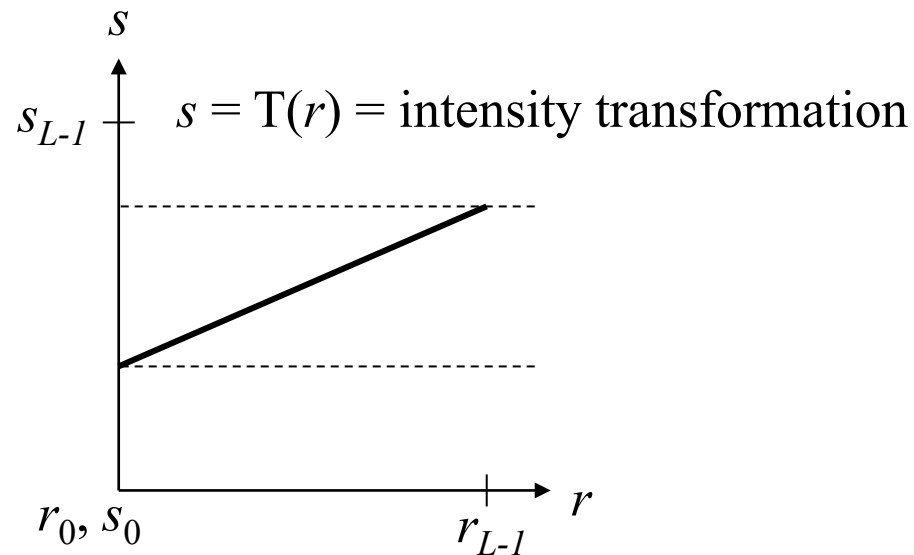
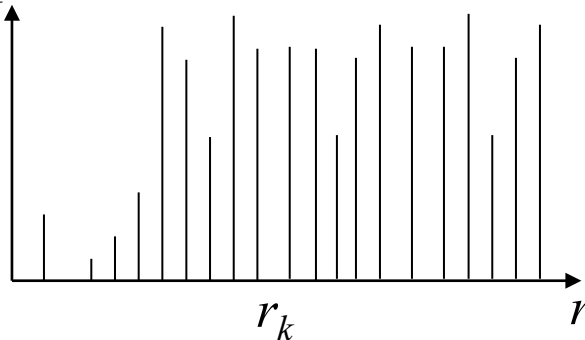
Histogram of a low contrast image



## Contrast compressing

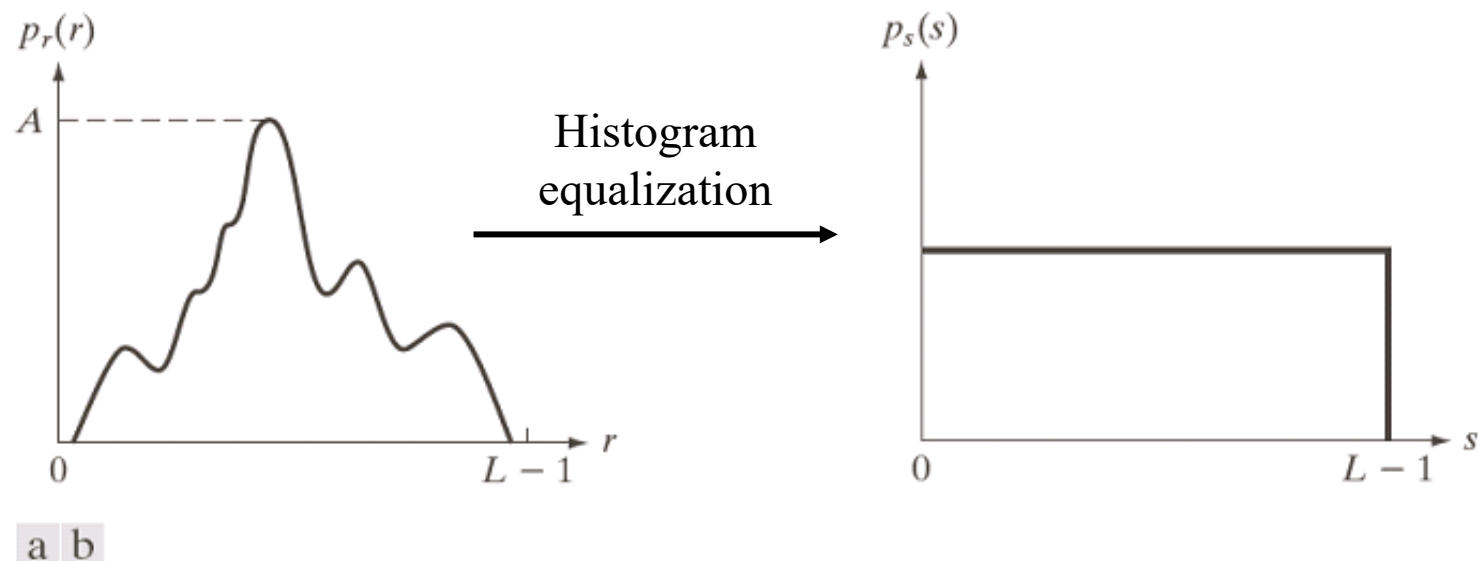
Histogram of a high contrast image

$$p_r(r_k) = \frac{n_k}{N}$$



- We want an image with output intensity values that approx. follow a *uniform distribution*.
- That is, a flat histogram, where each gray level,  $r_k$ , appears an equal number of times, i.e.,  $N/L$  times,
  - $L$  is the number of gray levels
  - $N$  is the total number of pixels in the image
- The intensity transformation is  $s = T(r)$ , such that
  - (a)  $T(r)$  is single-valued and non-decreasing in the interval  $0 \leq r \leq 1$
  - (b)  $0 \leq T(r) \leq 1$  for  $0 \leq r \leq 1$

## Histogram equalization



**FIGURE 3.18** (a) An arbitrary PDF. (b) Result of applying the transformation in Eq. (3.3-4) to all intensity levels,  $r$ . The resulting intensities,  $s$ , have a uniform PDF, independently of the form of the PDF of the  $r$ 's.



## Histogram equalization transform

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- The intensity transformation is the cumulative distribution function (CDF) of  $r$ , which is represented by

$$s = T(r) = \int_0^r p_r(w)dw$$

- The discrete implementation is given by

$$s_k = T(r_k) = \sum_{j=0}^k \frac{n_j}{N} = \sum_{j=0}^k p_r(r_j)$$

where

$s_k$  is the output intensity

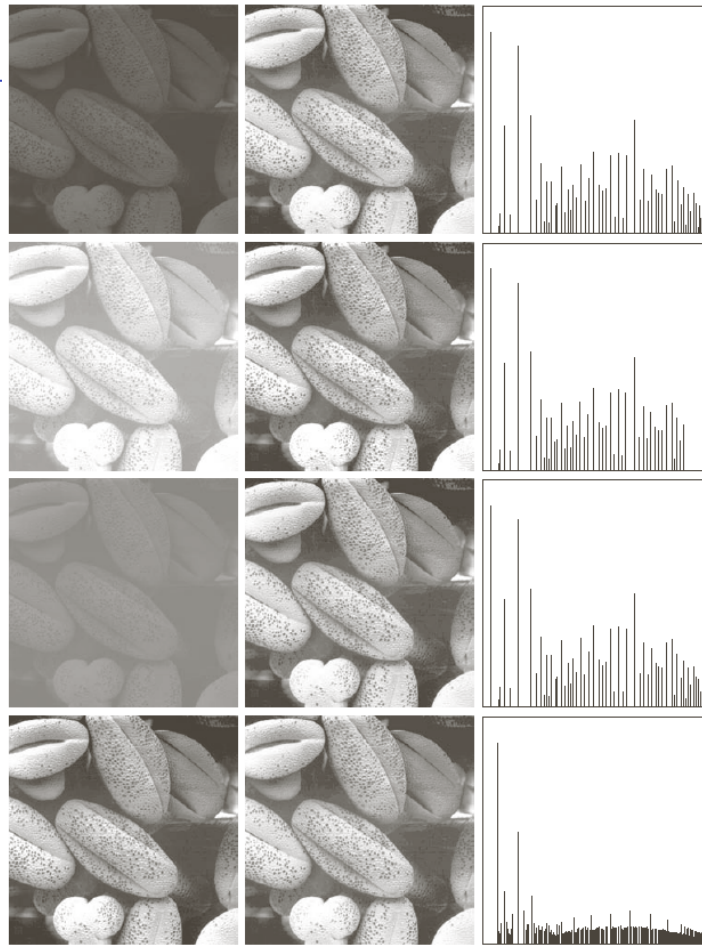
$r_k$  is the input intensity

$n_j$  is the number of pixels with gray level  $r_j$

$k = 0, 1, 2, 3, \dots, L-1$  (total number of gray levels is  $L$ )

## Example 2

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**FIGURE 3.20** Left column: images from Fig. 3.16. Center column: corresponding histogram-equalized images. Right column: histograms of the images in the center column.

- Histogram equalization can significantly improve image appearance
  - Automatic
  - Computationally efficient
- Good pre-processing step
  - Account for different lighting conditions
  - Account for different camera/device properties

- Apply histogram equalization about a neighborhood around  $(x,y)$
- Transform the gray level for pixel  $(x,y)$
- Move the neighborhood over the rest of the image

## Local histogram equalization

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Reveals details in local areas



Original



Global Histogram



Local Histogram

- Image histogram
- Histogram equalization
- Local histogram equalization
- More examples and details can be found at
  - [https://en.wikipedia.org/wiki/Histogram\\_equalization](https://en.wikipedia.org/wiki/Histogram_equalization)