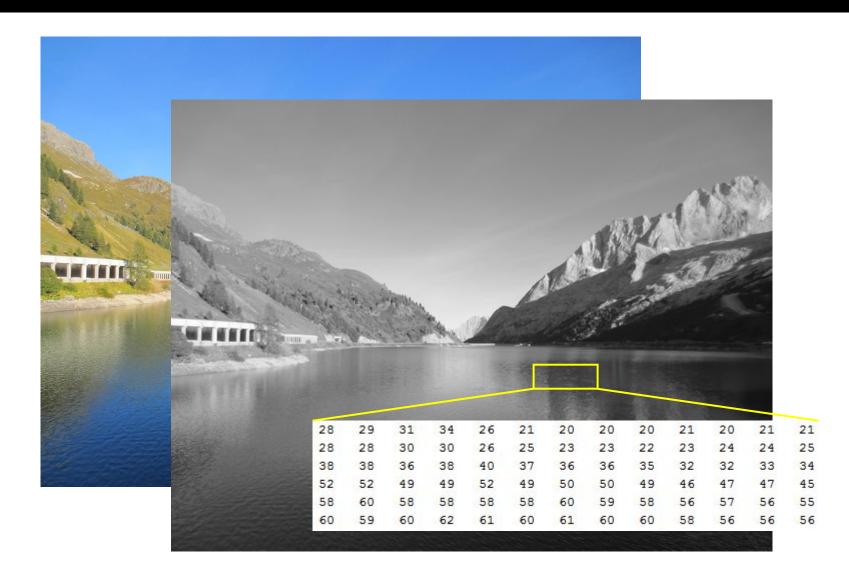
Recitation 1

Histograms

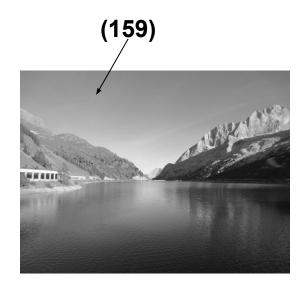
Agenda

- 1. Images
- 2. Intensity Transformations
- 3. Histogram Equalization
- 4. Histogram Matching
- 5. Ex 1

Intensity transformations
Histogram equalization
Histogram Matching
Ex 1



- Pixel (picture element) is the smallest single component of a digital image
- The number of pixels in an image is called the resolution
- In most cases:
 - Grayscale images have 8 bpp (256 intensities)
 - Color images have 24 bpp (each channel has 8 bpp)



- The value of a pixel describes how bright that pixel is.
- In cases of 8 bpp the range is between 0 and 255.



Representing Common Image Types

• Intensity (or grayscale) images -Either uint8 in the range {0,...,255} or floats in the range [0,1]



 Binary images boolean matrices of 0s and 1s

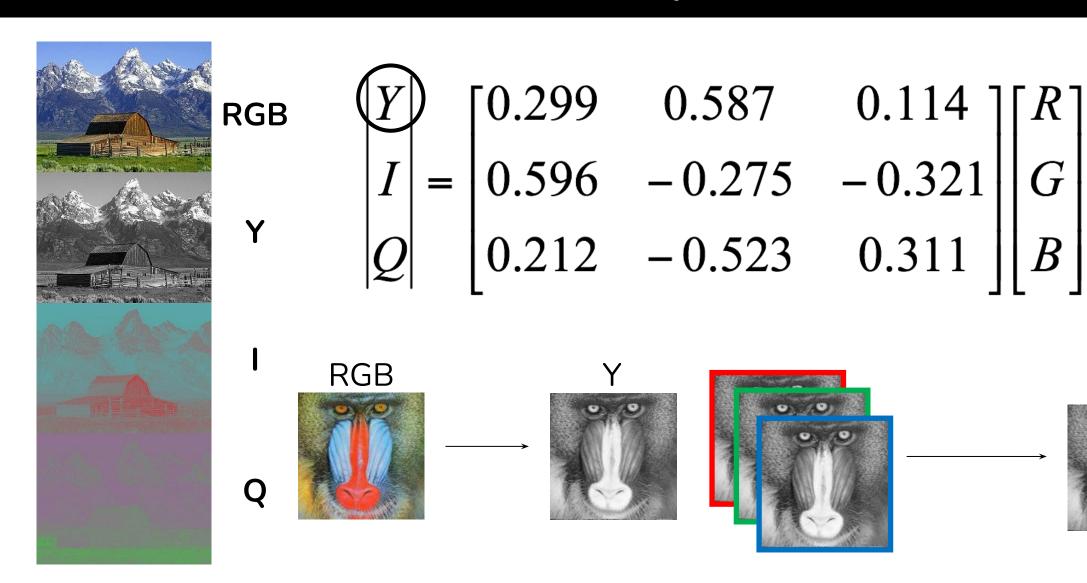


 RGB images three channels

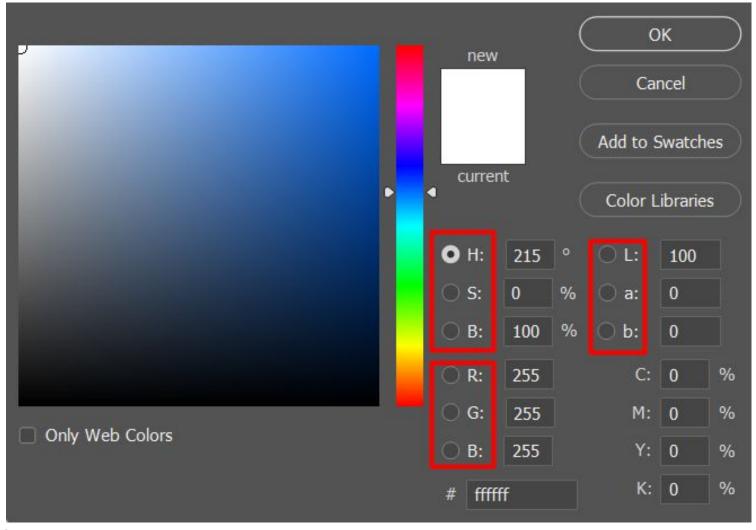




Color Spaces



Color Spaces



Adobe Phtotoshop 2019

lmages

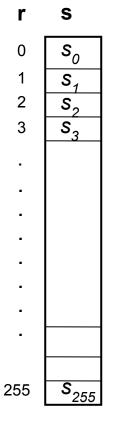
Intensity Transformation

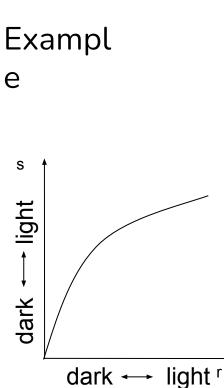
Histogram equalization
Histogram Matching
Ex 1

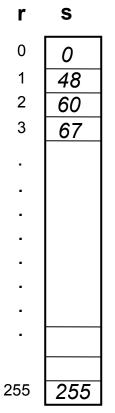
Intensity Transformation function

 In the case of a single-pixel transformation – we can use a lookup table to represent it

$$s = T(r)$$







$$s, r \in \{0, \dots, 255\}$$

$$s = 255 \left(\frac{r}{255}\right)^{0.3}$$

$$\downarrow$$

$$s, r \in [0,1]$$

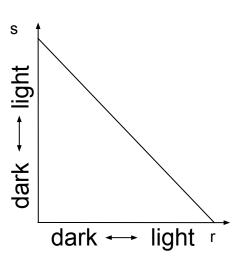
$$s = r^{0.3}$$

Simple Examples of Intensity Transformations

Image negative

$$s = 1 - r$$



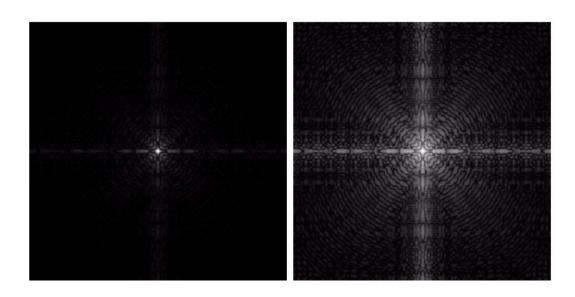


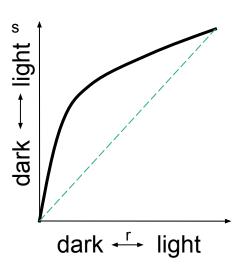
Simple Examples of Intensity Transformations

Log Transform

$$s = c * \log(1 + r)$$

$$c = \frac{255}{\log(1 + maxInpVal)}$$





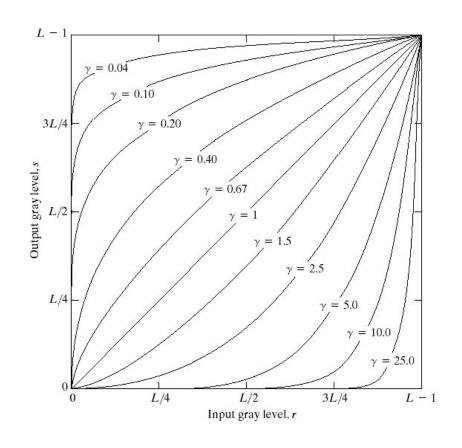
Simple Examples of Intensity Transformations

Power Law Transform

$$S = c \cdot r^{\gamma}$$

$$\gamma = 2$$

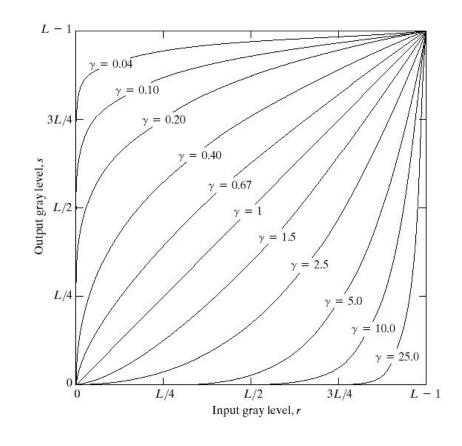




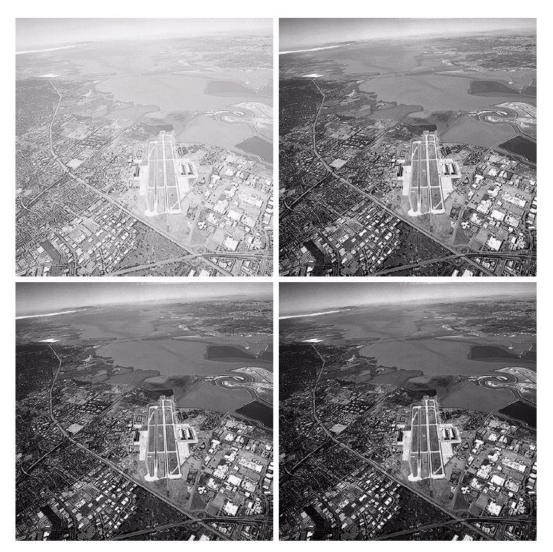
Gamma Correction - Example

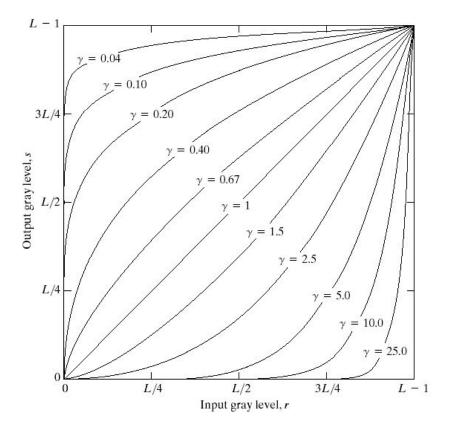
What kind of correction should we try here?



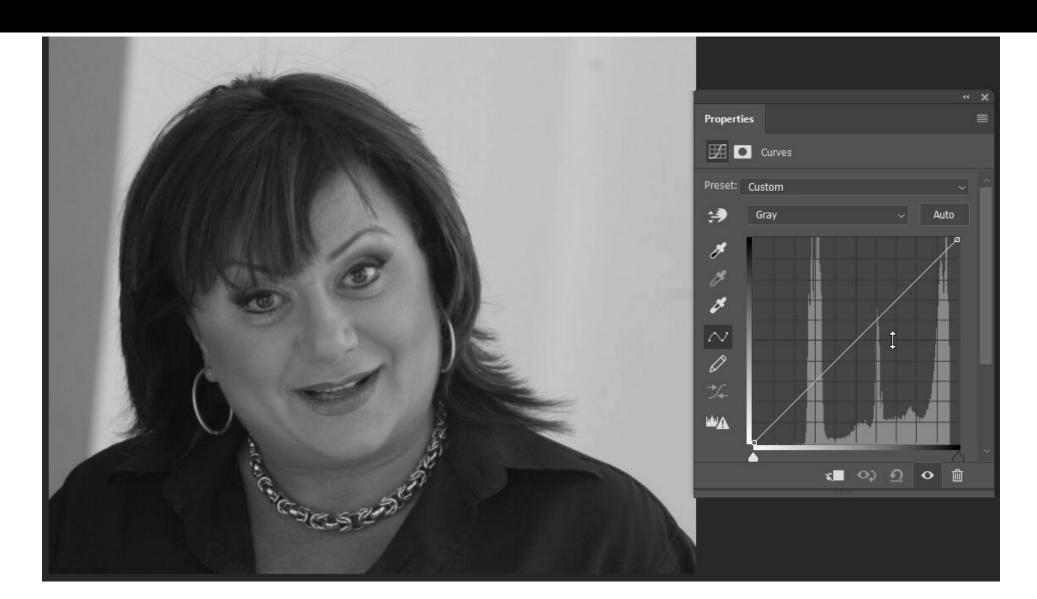


Gamma Correction - Example





Gamma Correction



Images Intensity Transformation

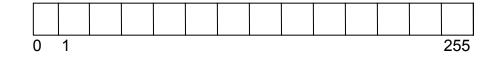
Histogram equalization

Histogram Matching

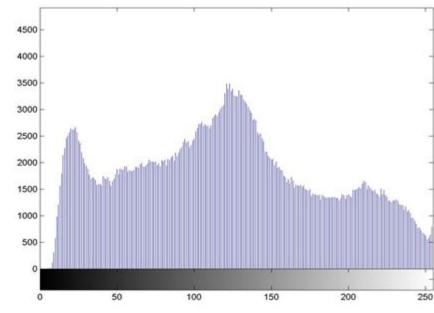
Ex 1

Histogram Processing

How many pixels intensity X has?







Histogram Processing

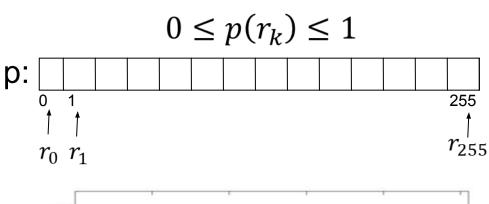
Image made of N pixels.

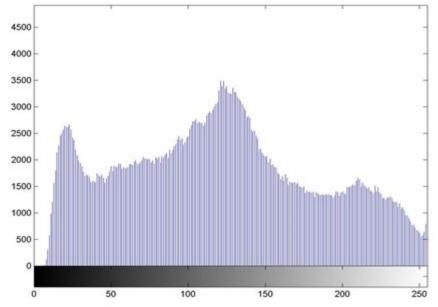
 n_k — How many pixels with intensity k

$$h(r_k) = n_k$$

$$\sum h(r_k) = N$$

To normalize: $p(r_k) = \frac{n_k}{N}$





Sample of Image Histograms

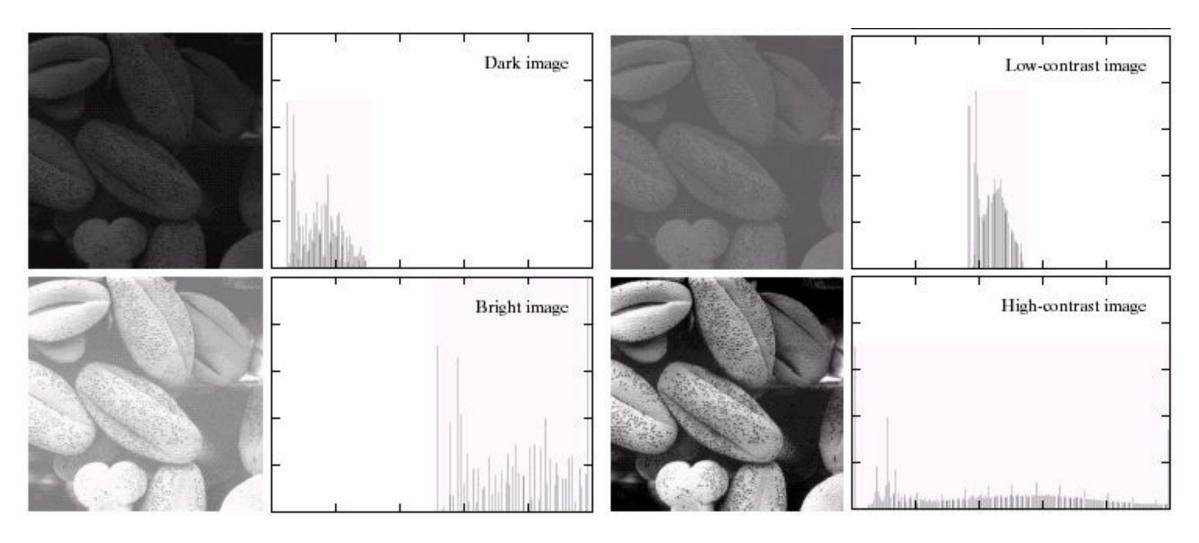


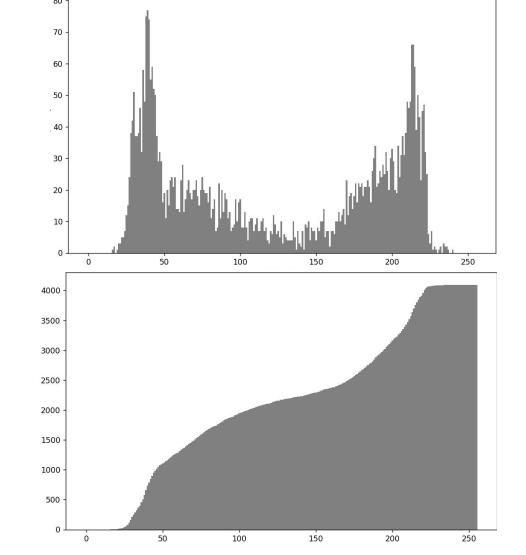
Image made of N pixels.

 n_k – How many pixels with intensity k

$$h(r_k) = n_k$$
 $\sum_{k=0}^{255} h(r_k) = N$ $p(r_k) = \frac{n_k}{N}$

cumulative histogram

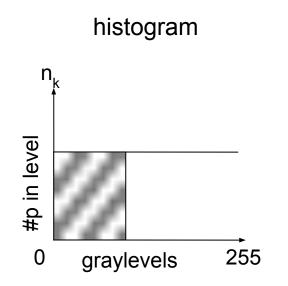
$$C(k) = \sum_{i=0}^{k} n_i$$
 $C(255) = \sum_{i=0}^{255} n_i = N$



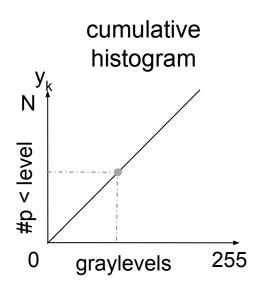
Motivation: improve the image contrast.

Goal: make equal use of all gray levels.

N pixels in range 0, ..., K (assume K=255).



Uniform histogram ↔ linear cumulative histogram



cumulative

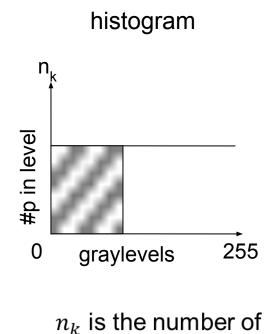
histogram

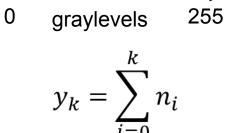
Ideally, we would like to equally use all gray levels

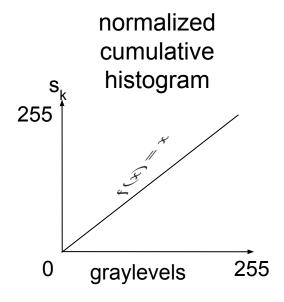
< level

d#

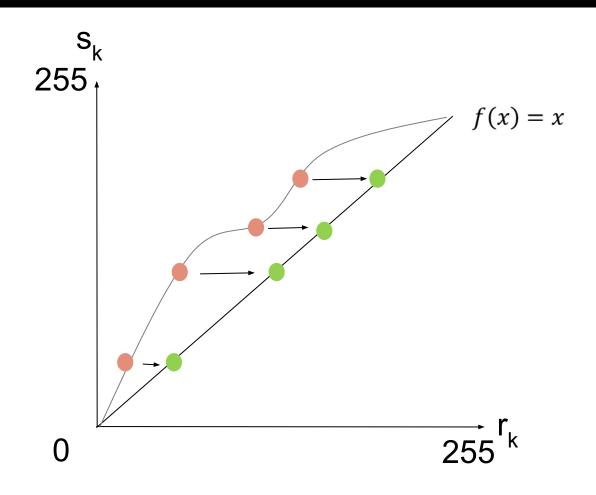
N pixels in range 0, ... K (assume K=255).







$$s_k = \frac{255}{N} y_k = \frac{255}{N} \sum_{i=0}^k n_i$$



- 1. Compute the image histogram (np.histogram)
- 2. Compute the cumulative histogram (np.cumsum)
- 3. Normalize the cumulative histogram (divide by the total number of pixels)
- 4. Multiply the normalized histogram by the maximal gray level value (Z-1)

cumulative histogram C(k)Let m be first grey level for which $C(m) \neq 0$

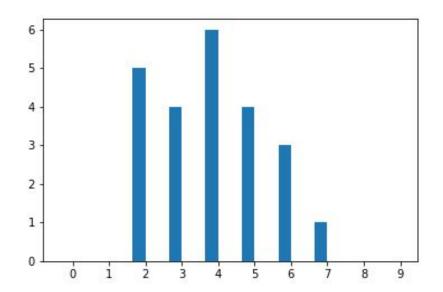
$$T(k) = round\left(255 * \frac{C(k)}{C(255)}\right) = round\left(255 * \frac{C(k)}{N}\right)$$

cumulative histogram C(k)

$$T(k) = round\left(9 * \frac{C(k)}{C(9)}\right)$$

h:

0	1	2	3	4	5	6	7	8	9
0	0	5	4	6	4	3	1	0	0



cumulative histogram C(k)

$$T(k) = round\left(9 * \frac{C(k)}{C(9)}\right)$$

h:

0	1	2	3	4	5	6	7	8	9
0	0	5	4	6	4	3	1	0	0

h cumulative:

0	1	2	3	4	5	6	7	8	9
0	0	5	9	15	19	22	23	23	23

cumulative histogram C(k)

$$T(k) = round\left(9 * \frac{C(k)}{C(9)}\right)$$

h:

0	1	2	3	4	5	6	7	8	9
0	0	5	4	6	4	3	1	0	0

h

cumulative:

0	1	2	3	4	5	6	7	8	9
0	0	5	9	15	19	22	23	23	23

T

0	1	2	3	4	5	6	7	8	9
0	0	2	4	6	7	9	9	9	9

New h:

0	1	2	3	4	5	6	7	8	9
0	0	5	0	4	0	6	4	0	4

Example

$$T(2) = round\left(9 * \frac{C(2)}{C(9)}\right)$$

$$= round\left(9 * \frac{5}{23}\right) = round(1.9) = 2$$

Not חח"ע

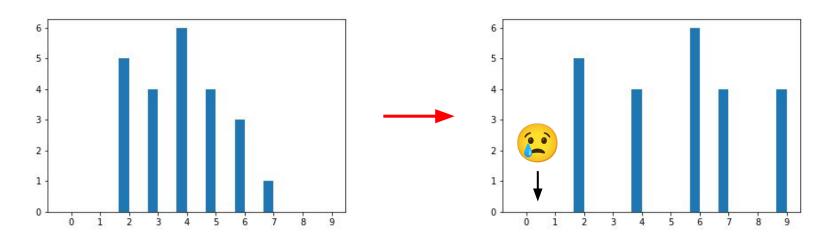
cumulative histogram C(k)

$$T(k) = round\left(9 * \frac{C(k)}{C(9)}\right)$$

h:

0	1	2	3	4	5	6	7	8	9
0	0	5	4	6	4	3	1	0	0

0	1	2	3	4	5	6	7	8	9
0	0	5	0	4	0	6	4	0	4



cumulative histogram C(k)Let m be first grey level for which $C(m) \neq 0$

$$T(k) = round\left(9 * \frac{C(k) - C(m)}{C(9)}\right)$$

h:

0	1	2	3	4	5	6	7	8	9
0	0	5	4	6	4	3	1	0	0

h cumulative:

0	1	2	3	4	5	6	7	8	9
0	0	5	9	15	19	22	23	23	23

T:

0	1	2	3	4	5	6	7	8	9
0	0	0	2	4	5	7	7	7	7

0	1	2	3	4	5	6	7	8	9
5	0	4	0	6	4	0	4	0	0

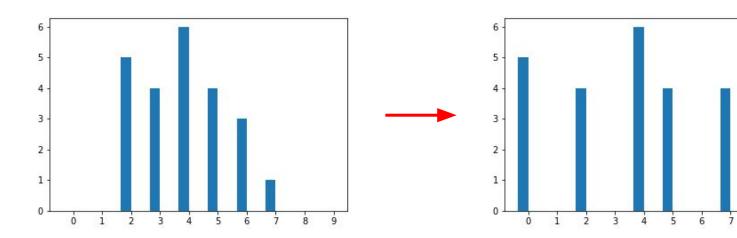
cumulative histogram C(k)Let m be first grey level for which $C(m) \neq 0$

$$T(k) = round\left(9 * \frac{C(k) - C(m)}{C(9)}\right)$$

h:

0	1	2	3	4	5	6	7	8	9
0	0	5	4	6	4	3	1	0	0

0	1	2	3	4	5	6	7	8	9
5	0	4	0	6	4	0	4	0	0



cumulative histogram C(k)Let m be first grey level for which $C(m) \neq 0$

$$T(k) = round\left(9 * \frac{C(k) - C(m)}{C(9) - C(m)}\right)$$

h:

0	1	2	3	4	5	6	7	8	9
0	0	5	4	6	4	3	1	0	0

h cumulative:

0	1	2	3	4	5	6	7	8	9
0	0	5	9	15	19	22	23	23	23

T

0	1	2	3	4	5	6	7	8	9
0	0	0	2	5	7	8	9	9	9

0	1	2	3	4	5	6	7	8	9
5	0	4	0	0	6	0	4	3	1

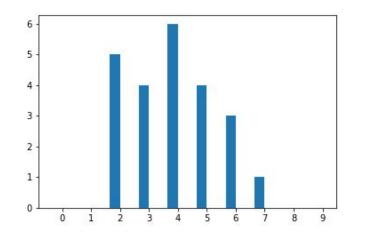
cumulative histogram C(k)Let m be first grey level for which $C(m) \neq 0$

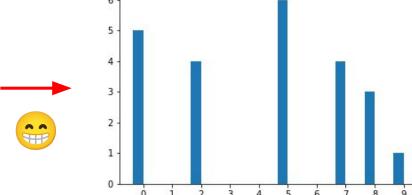
$$T(k) = round\left(9 * \frac{C(k) - C(m)}{C(9) - C(m)}\right)$$

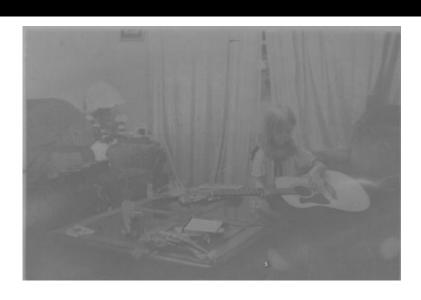
h:

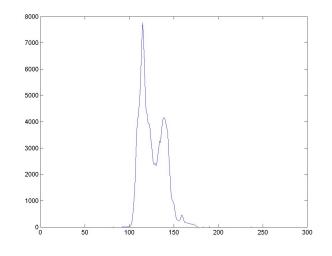
0	1	2	3	4	5	6	7	8	9
0	0	5	4	6	4	3	1	0	0

0	1	2	3	4	5	6	7	8	9
5	0	4	0	0	6	0	4	3	1

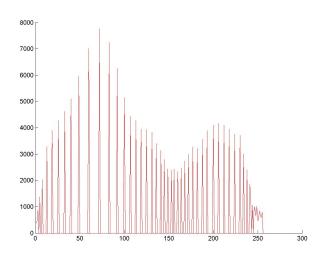


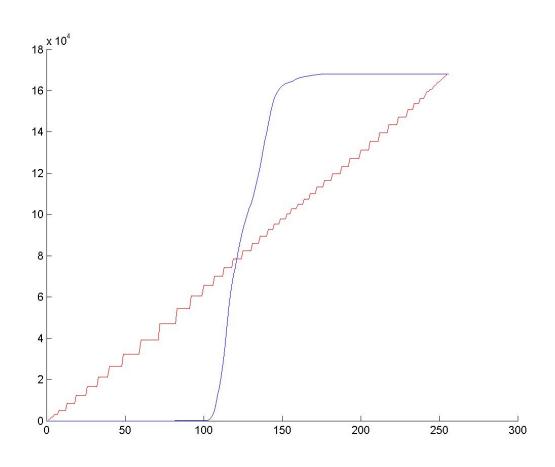












Histogram Equalization: Algorithm

- 1. Compute the image histogram (np.histogram)
- 2. Compute the cumulative histogram (np.cumsum)
- 3. Normalize the cumulative histogram (divide by the total number of pixels)
- 4. Multiply the normalized histogram by the maximal gray level value (Z-1)
- 5. Verify that the minimal value is 0 and that the maximal is Z-1, otherwise stretch the result linearly in the range [0,Z-1].
- 6. Round the values to get integers
- 7. Map the intensity values of the image using the result of step 6.

cumulative histogram
$$C(k)$$

Let m be first grey level for which $C(m) \neq 0$
$$T(k) = round \left(255 * \frac{C(k) - C(m)}{C(255) - C(m)}\right)$$

Equalization Properties

- In most discrete images it is impossible to achieve a completely uniform histogram – only an approximation.
- The histogram equalization transformation is **monotonic** (as the process involves the cumulative sum), and therefore the relative brightness of a pixel is preserved.
- The number of different pixel values (=full bins in the histogram) can only decrease. We might **merge bins**, but we can't split them.

When will this fail?

- When the desired gray level distribution is not uniform (e.g. text)
- When content from different sources is equalized together (e.g. two non related images)



Equalization Results

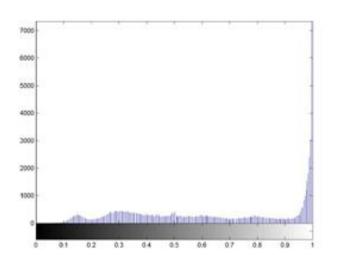
OFFICIAL NEWSLETTER FOR THE LEONARD PELTIER DEFENSE COMMITTEE



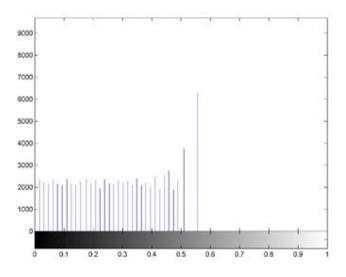
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Images Intensity Transformation Histogram Equalization

Histogram Matching

What is Histogram Matching?

A process to adjust the histogram of one image to resemble the histogram of another (target) image.

Used to transfer the appearance characteristics (such as contrast and brightness) of one image to another.

Steps in Histogram Matching

- 1. Compute histograms of both the source and target images.
- 2. Calculate the Cumulative Distribution Function (CDF) for both histograms.
- **3. Match intensities** by mapping pixel values from the source image to corresponding values in the target histogram using the CDFs.
- 4. Transform the image based on the mapping.

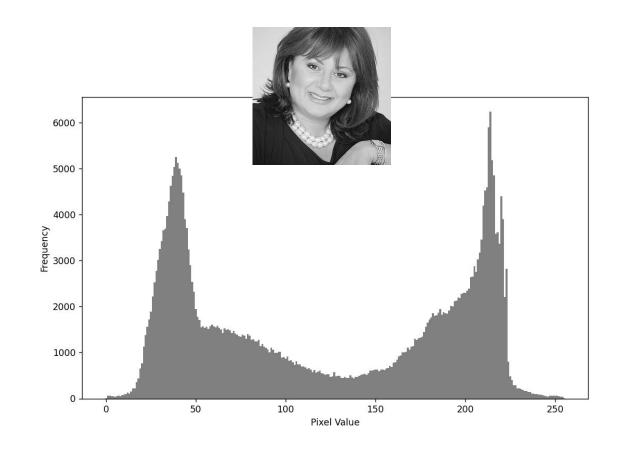
Source

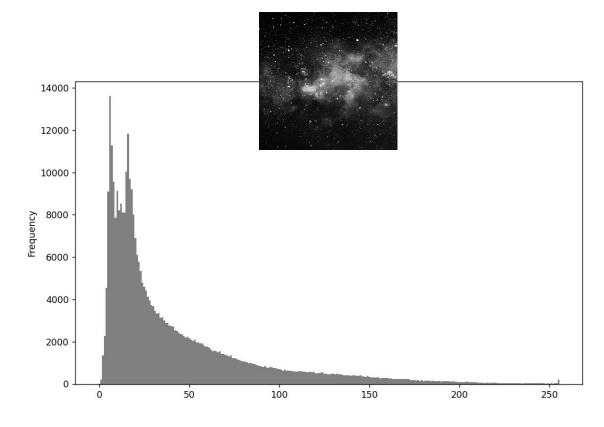


Target

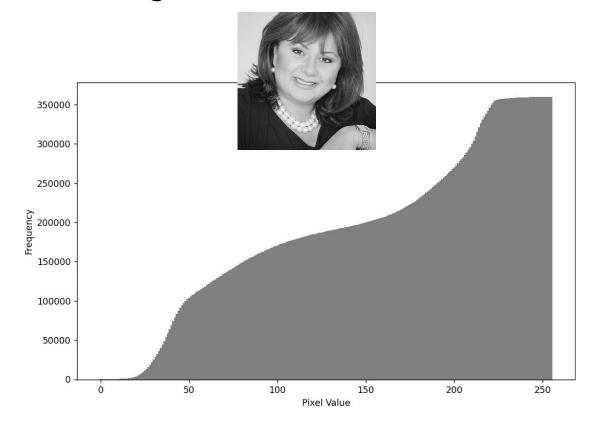


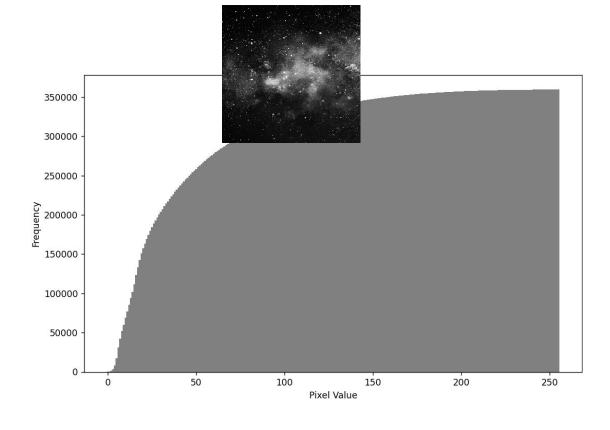
1. Compute histograms of both the source and target images.





2. Calculate the Cumulative Distribution Function (CDF) for both histograms.





3. Match intensities by mapping pixel values from the source image to corresponding values in the target histogram using the CDFs.

$$C_{source}(i) = \frac{1}{N_s} \sum_{j=0}^{i} H_s(j)$$

$$C_{target}(i) = \frac{1}{N_t} \sum_{j=0}^{i} H_t(j)$$

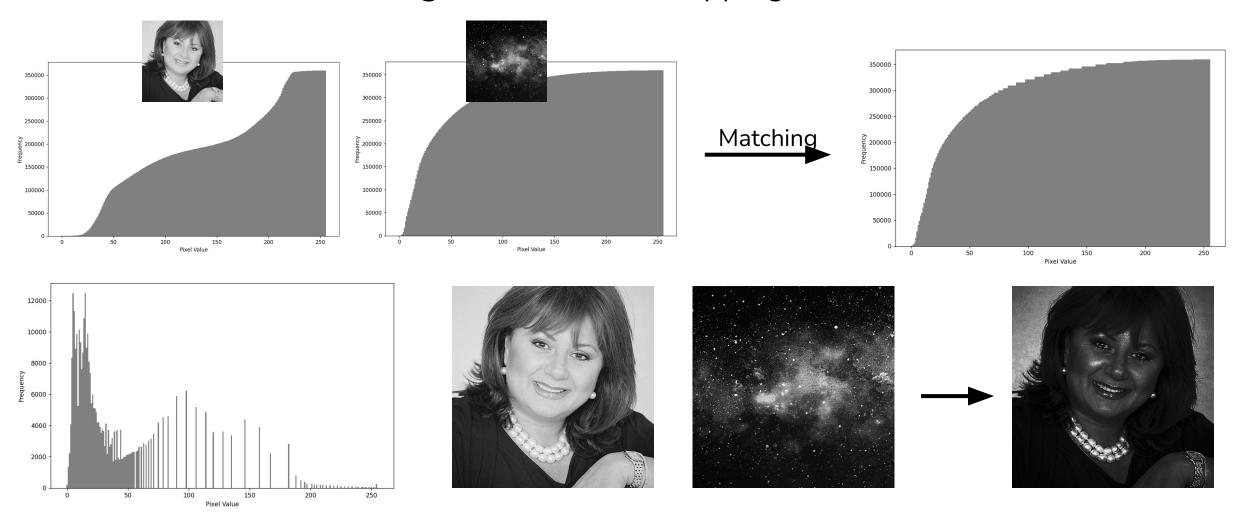
For matching the intensities from the source to the target, find the intensity *j* in the target image such that:

$$C_{source}(i) \approx C_{target}(j)$$

Transformation Function

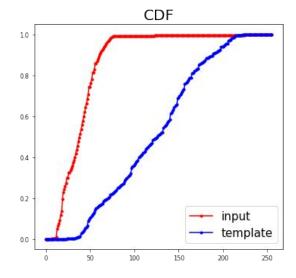
$$T(i) = arg \min_{j} |C_{source}(i) - C_{target}(j)|$$

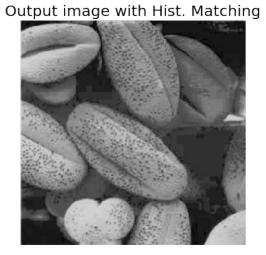
4. Transform the image based on the mapping.

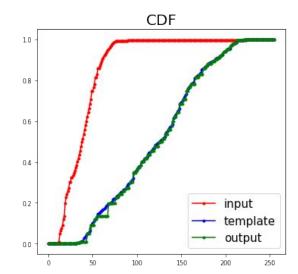


Input image









Intensity Transformation
Histogram equalization
Histogram Matching

Ex 1

Ex1

- •COLAB Bootcamp IMPORTANT
- cut detection

Good Luck!

Next week: 1D Fourier