





Histogram Processing, Equalization, Matching

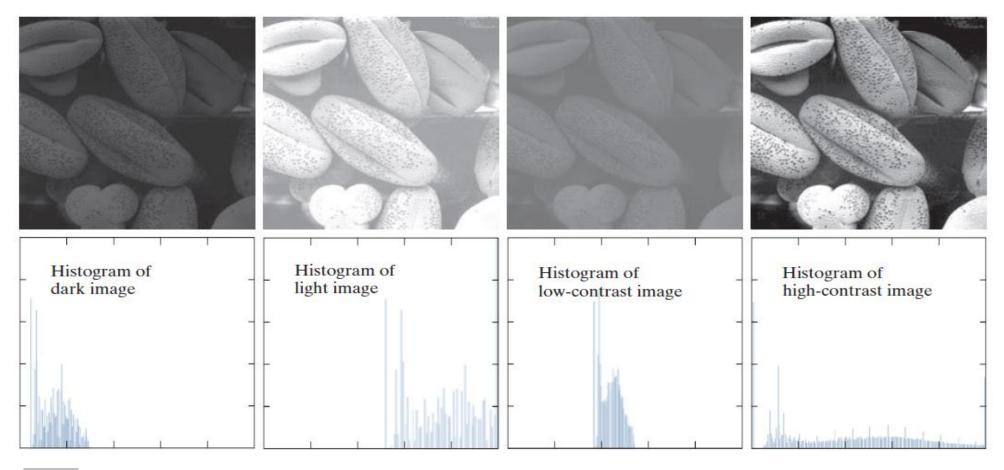
Image Histogram



- Let r_k , for k = 0,1,2,...,L-1 denote the intensities of an L level digital image f(x,y).
- For 4-level digital image we get r_0 , r_1 , r_2 , r_3 .
- The unnormalized histogram of f is defined as $h(r_k) = n_k$, where n_k is the number of pixels in f with the intensity r_k , and the subdivision of the intensity scale are called histogram bins.
- The normalized histogram of f is defines as $p(r_k) = \frac{n_k}{MN}$, where M and N are the number of rows and columns in the image.







Abcd Four image types and their corresponding histograms. (a) dark; (b) light; (c) low contrast; (d) high contrast. The horizontal axis of the histograms are values of r_k and the vertical axis are values of $p(r_k)$.

Image Histogram



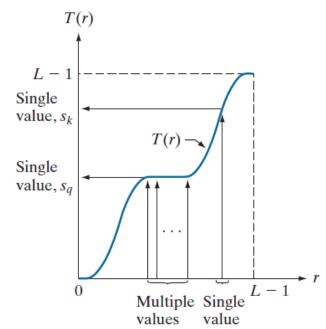
- Image histogram refers normalized histogram
- Histogram shape is related to image appearance
- High contrast image cover a wide range of the intensity scale and distribution of pixels is not too far from uniform.
- Developing an transformation function that can achieve this effect automatically, using only the histogram of an image will help to enhance image.

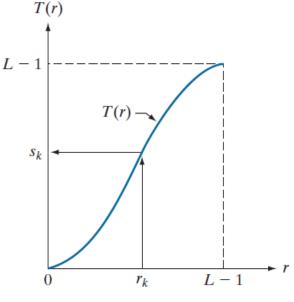


- Assuming initially continuous intensity values, let the variable r denote the intensities of an image to be processed. As usual, we assume that r is in the range [0, L-1], with r=0 representing black and r=L-1 presenting white.
- The transformations (intensity mappings) function of the form s = T(r), $0 \le r \le L 1$, produce an output intensity value s, for a given intensity value r in the input image.



- T(r) is a monotonic increasing function in the interval $0 \le r \le L-1$
- $0 \le T(r) \le L 1$ for $0 \le r \le L 1$
- If we required inverse transformation then the function T(r) should be strictly monotonic increasing function in the interval $0 \le r \le L-1$







- Let $p_r(r)$ and $p_s(s)$ denote the PDF of intensity values r and s in two different image.
- A fundamental result from probability theory is that if $p_r(r)$ and T(r) known, and T(r) is continuous and differentiable over the range of values of interest, then the PDF of transformation variable s can be obtained as $p_s(s) = p_r(r) \frac{dr}{ds}$
- The PDF of the output intensity variable, s, is determined by the PDF of the input intensities and the transformation function used.



- A transformation function of particular importance in image processing is $s = T(r) = (L-1) \int_0^r p_r(w) dw$
- The integral on the right side is the CDF of the random variable r.
- Leibniz rule says that the derivative of a definite integral with respect to its upper limit is the integrated evaluates at the limit

$$\frac{dr}{ds} = \frac{dT(r)}{dr} = (L-1)\frac{d}{dr}\left(\int_0^r p_r(w)dw\right) = (L-1)p_r(r)$$



- For discrete values $p_r(r_k) = \frac{n_k}{MN}$ it refers to the normalized histogram
- The discrete form of transformation

$$s_k = T(r_k) = (L-1) \sum_{\{j=0\}}^{\kappa} p_r(r_j), k = 0,1,2.3,...,L$$

A processed image is obtained by using above equation to map each pixel in the input image with intensity r_k into a corresponding pixel with level s_k in the output image.

This is this is called histogram equalization or histogram linearization.



- Take an image
- Calculate pdf of the intensity values
- Calculate cdf of the intensity values
- Calculate the discrete form of the cdf
- Round off discrete values and map to the new intensity values

3

4



6	6	7	7	6
5	2	2	3	4
3	3	4	4	5
5	7	3	6	2
7	6	5	5	4
	-	lacksquare	-	
6	6	7	7	6
4	1	1	2	3
2	2	3	3	5

6

Intensity values	frequenc ies	Pdf	cdf	(L-1)*cdf	Histogra m equal level
0	0	0	0	0	0
1	0	0	0	0	0
2	3	0.12	0.12	0.84	1
3	4	0.16	0.28	1.96	2
4	4	0.16	0.44	3.08	3
5	5	0.20	0.64	4.48	4
6	5	0.20	0.84	5.88	6
7	4	0.16	1	7	7



4	4	4	4	4
3	4	5	4	3
3	5	5	5	3
3	4	5	4	3
4	4	4	4	4
5	5	5	5	5
2	5	7	5	2
2	7	7	7	2
2	5	7	5	2
5	5	5	5	5

Intensity values	frequenc ies	Pdf	cdf	(L-1)*cdf	Histogra m equal level
0	0	0	0	0	0
1	0	0	0	0	0
2	0	0	0	0	0
3	6	0.24	0.24	1.68	2
4	14	0.56	0.80	5.6	5
5	5	0.20	1	7	7
6	0	0	1	7	7
7	0	0	1	7	7

Histogram Matching (Specification)

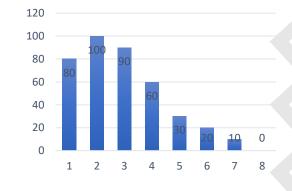


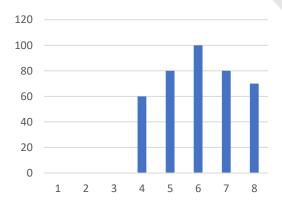
• Given below are two histograms (i) and (ii). Modify the histogram (i)

as given by histogram (ii).

Gray Level	0	1	2	3	4	5	6	7
No of pixel	80	100	90	60	30	20	10	0

Gray Level	0	1	2	3	4	5	6	7
No of pixel	0	0	0	60	80	100	80	70









Gray levels	m_k	$p(r_k) = m_k/n$	s_k	$s_k * (L-1)$	Histogram Equalization	New
0	80	0.20	0.20	1.4	1	80
1	100	0.25	0.45	3.15	3	100
2	90	0.23	0.68	4.76	5	90
3	60	0.15	0.83	5.81	6	90
4	30	0.07	0.90	6.3	6	-
5	20	0.05	0.95	6.65	7	30
6	10	0.02	0.97	6.79	7	-
7	0	0	0.97	6.79	7	-

N=390





Gray levels	m_k	$p(r_k) = m_k/n$	s_k	$s_k * (L-1)$	Histogram Equalization	New
0	0	0	0	0	0	
1	0	0	0	0	0	
2	0	0	0	0	0	
3	60	0.15	0.15	1.05	1	
4	80	0.20	0.35	2.45	2	
5	100	0.25	0.6	4.2	4	
6	80	0.20	0.8	5.6	6	
7	70	0.17	0.97	6.79	7	

Histogram Matching (Specification)



Desired Histogram

Gray Level n_k 0 0 0 0 6

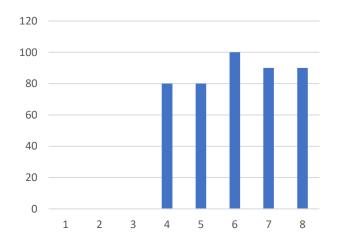
Equalized Histogram

Gray Level	n_k
0	0
-1	80
2	0
3	100
4	0
5	90
- 6	90
≻ 7	30





Gray Level	0	1	2	3	4	5	6	7
No of pixel	0	0	0	80	80	100	90	90



Reference

TECHNO INDIA
SILIGURI INSTITUTE
OF TECHNOLOGY
A Sayam Boychoudhury Indiative

 E Woods, Richard, and Rafael C Gonzalez. "Digital image processing." (2008).

