

Digital Image Processing

Lecture 3

Image Pre-processing (Image Color Transformations)

Lecturer: Associate Prof. Lý Quốc Ngọc

3. Image Pre-processing

3.1. Color Transformations

3.2. Geometric Transformations

3.3. Local Pre-processing

3.1. Color Transformations

3.1.1. Linear mapping

3.1.2. Non-linear mapping

3.1.3. Probability Density Function-based mapping

3.1.1. Linear mapping



(a)



(b)

3.1.1. Linear mapping



(a)



3.1.1. Linear mapping

❖ **Brightness modification**

$$g(x, y) = f(x, y) + b$$

❖ **Contrast modification**

$$g(x, y) = a.f(x, y)$$

❖ **Brightness+ Contrast modification**

$$g(x, y) = a.f(x, y) + b$$

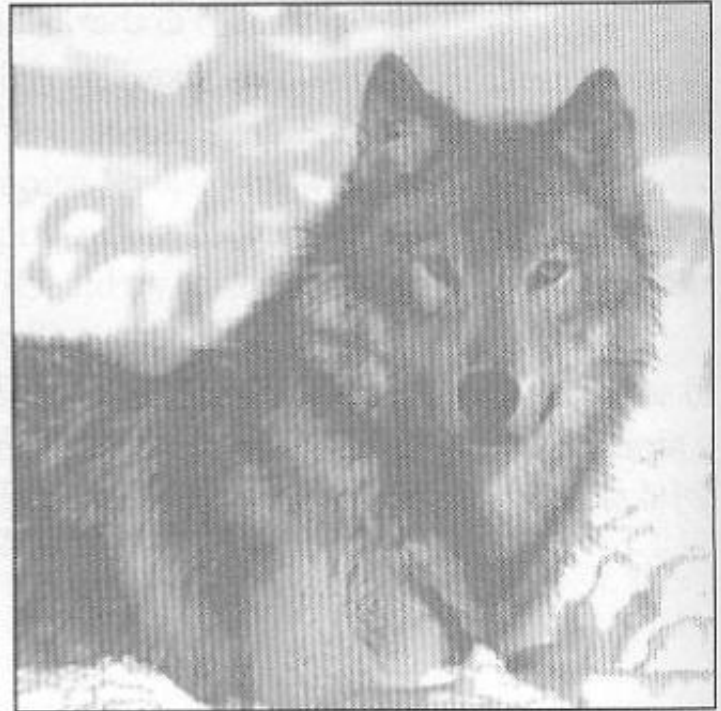
3.1.1. Linear mapping

❖ Brightness modification

$$g(x, y) = f(x, y) + b$$



(a)



(b)

3.1.1. Linear mapping

❖ Contrast modification

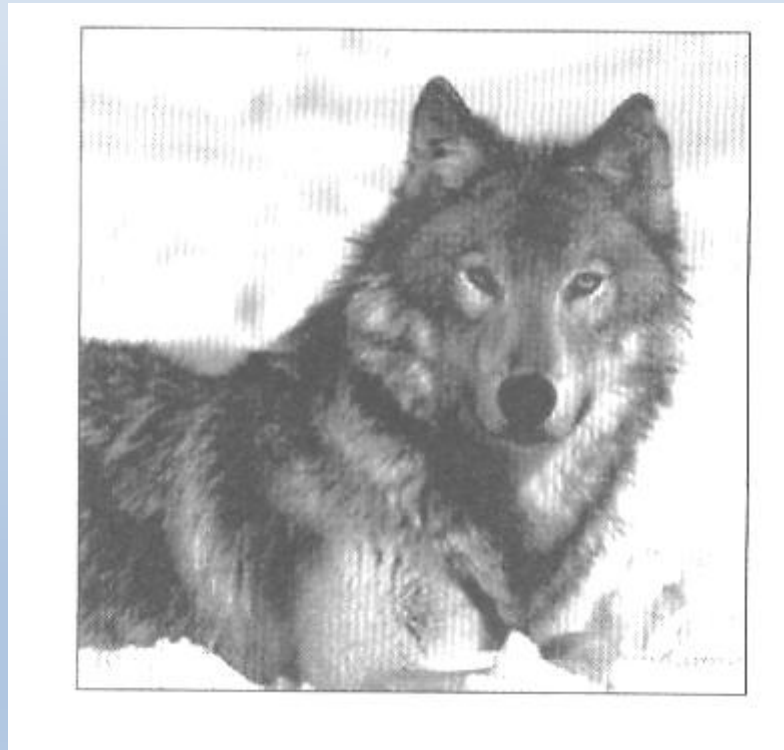
$$g(x, y) = a.f(x, y)$$



3.1.1. Linear mapping

❖ Brightness+ Contrast modification

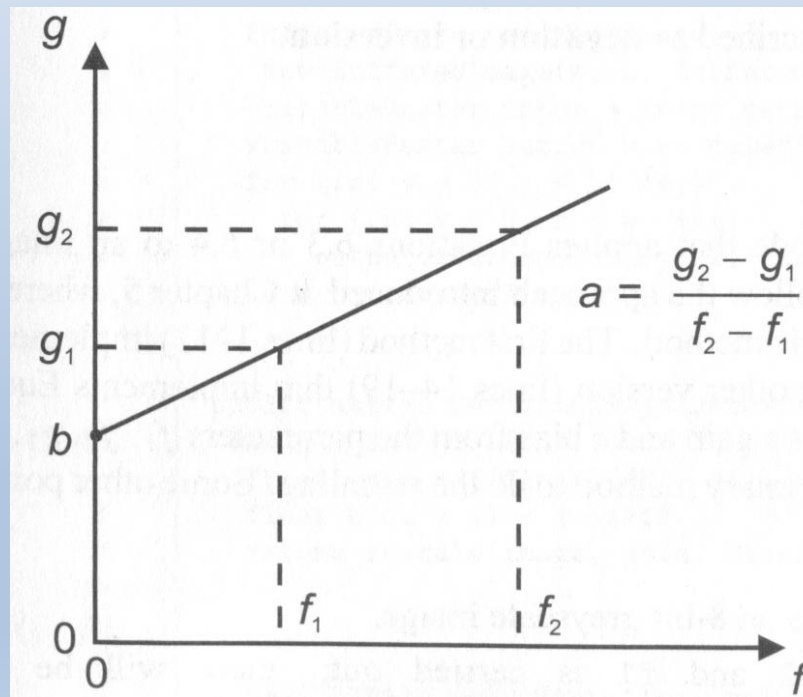
$$g(x, y) = a.f(x, y) + b$$



3.1.1. Linear mapping

❖ Map a particular range of grey levels onto a new range

Ex: map image from range of grey levels $[f_1 \ f_2]$ onto range $[g_1 \ g_2]$.



3.1.2. Non-linear mapping



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3.1.2. Non-linear mapping



?



3.1.2. Non-linear mapping

❖ Logarithmic mapping function

$$g(x, y) = c \log f(x, y)$$

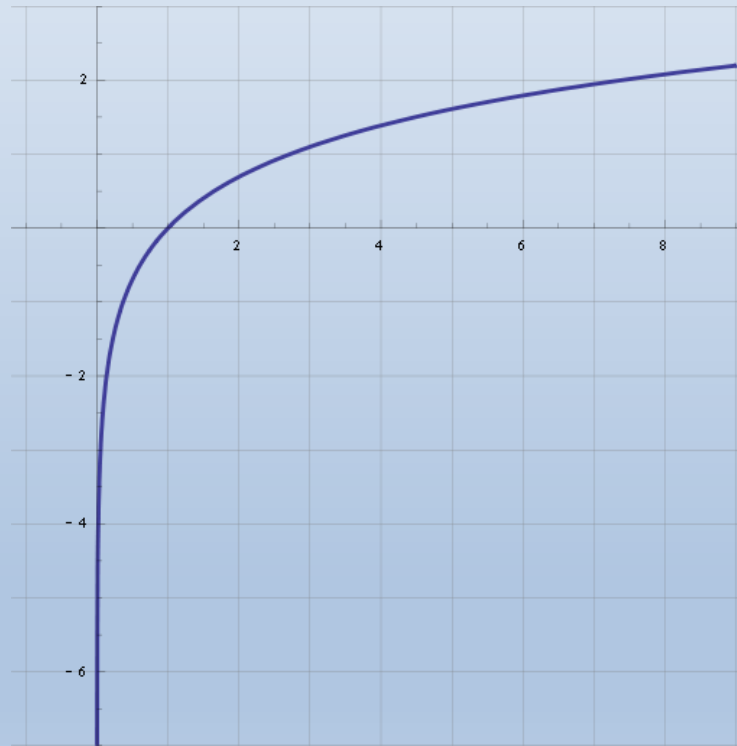
❖ Exponential mapping function

$$g(x, y) = e^{f(x, y)}$$

3.1.2. Non-linear mapping

❖ Logarithmic mapping function

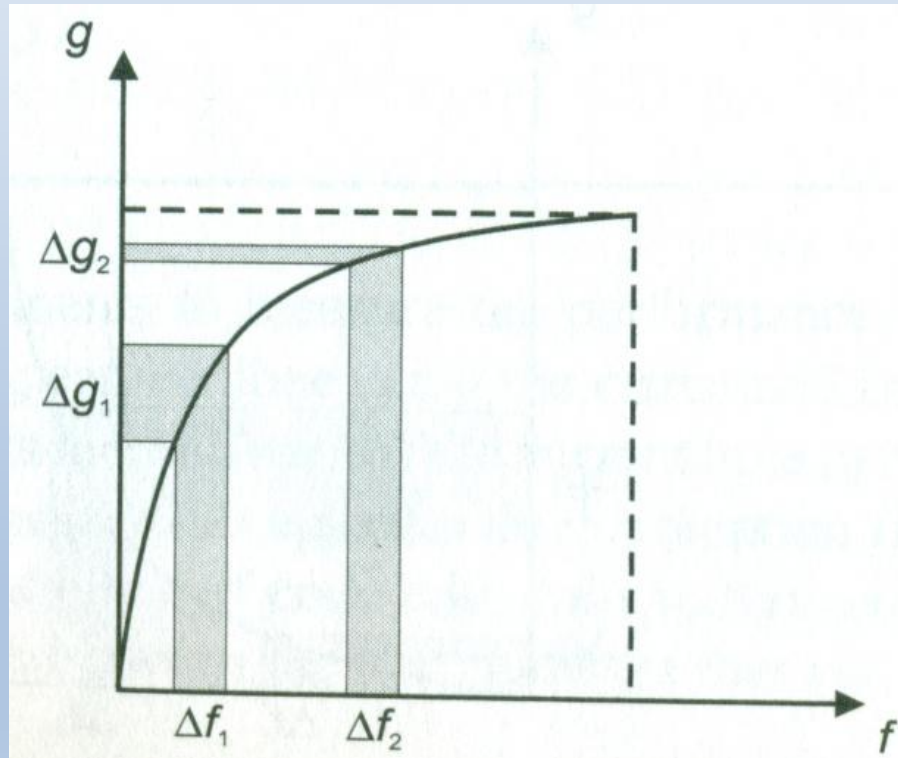
$$g(x, y) = c \log f(x, y)$$



3.1.2. Non-linear mapping

❖ Logarithmic mapping function

$$g(x, y) = c \log f(x, y)$$



3.1.2. Non-linear mapping

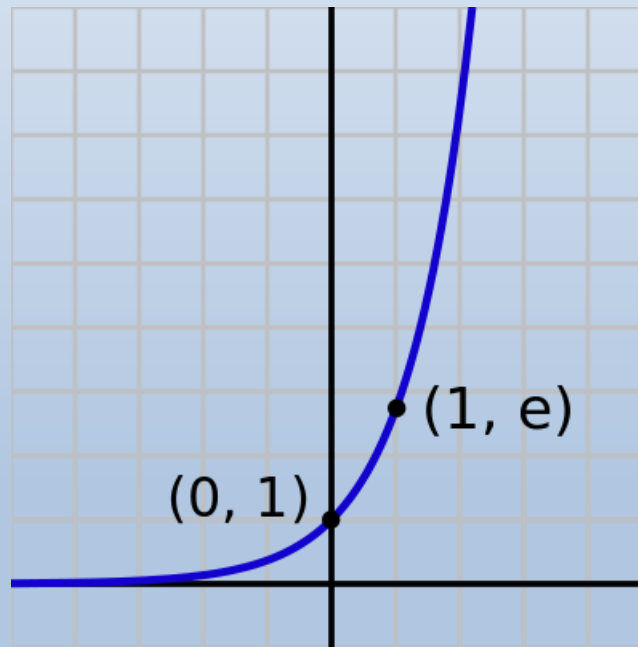
❖ Logarithmic mapping function



3.1.2. Non-linear mapping

❖ Exponential mapping function

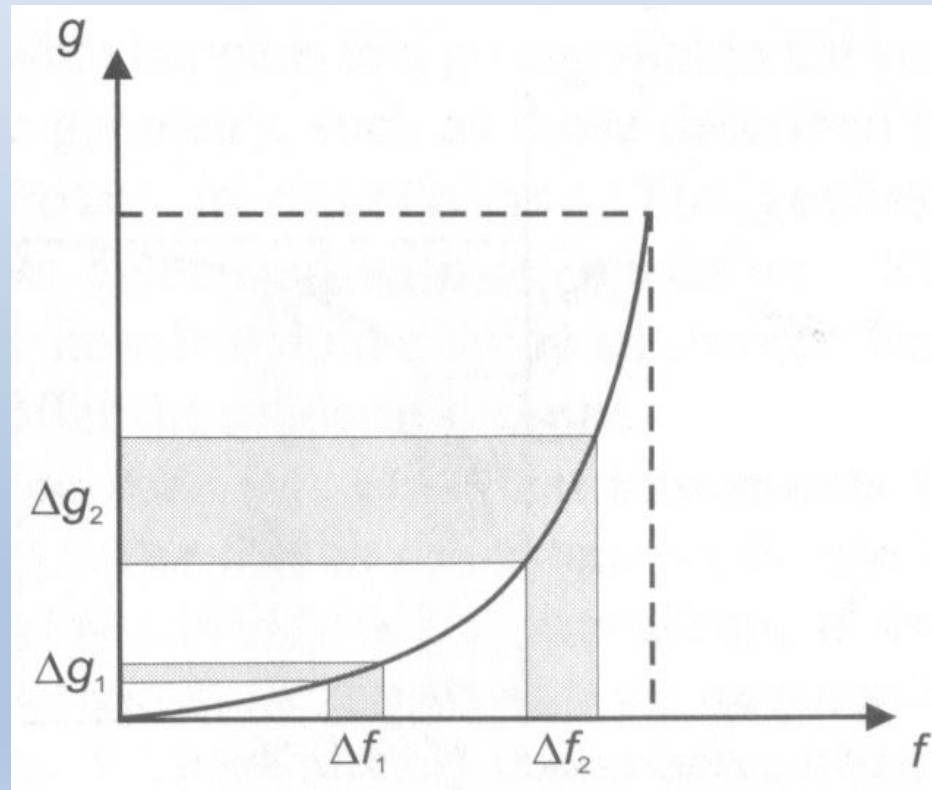
$$g(x, y) = e^{f(x, y)}$$



3.1.2. Non-linear mapping

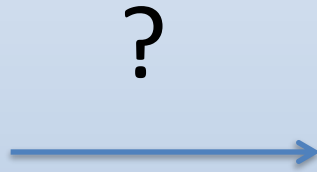
❖ Exponential mapping function

$$g(x, y) = e^{f(x, y)}$$



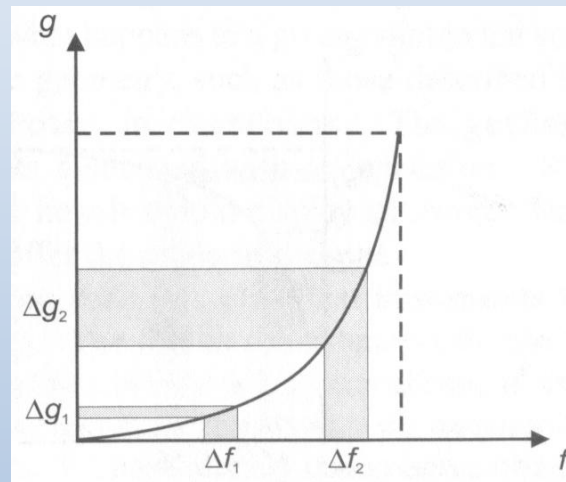
3.1.2. Non-linear mapping

❖ Ex: Exponential mapping function



3.1.2. Non-linear mapping

❖ Ex: Exponential mapping function



3.1.2. Non-linear mapping

❖ Ex: Logarithmic mapping function

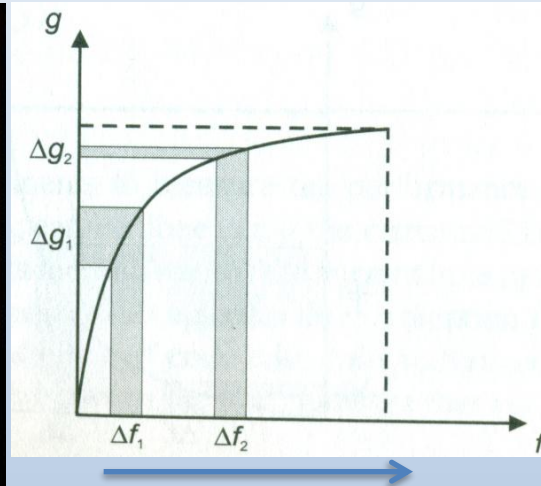


?



3.1.2. Non-linear mapping

❖ Ex: Logarithmic mapping function



3.1.3. Probability Density Function-based mapping



3.1.3. Probability Density Function - based mapping

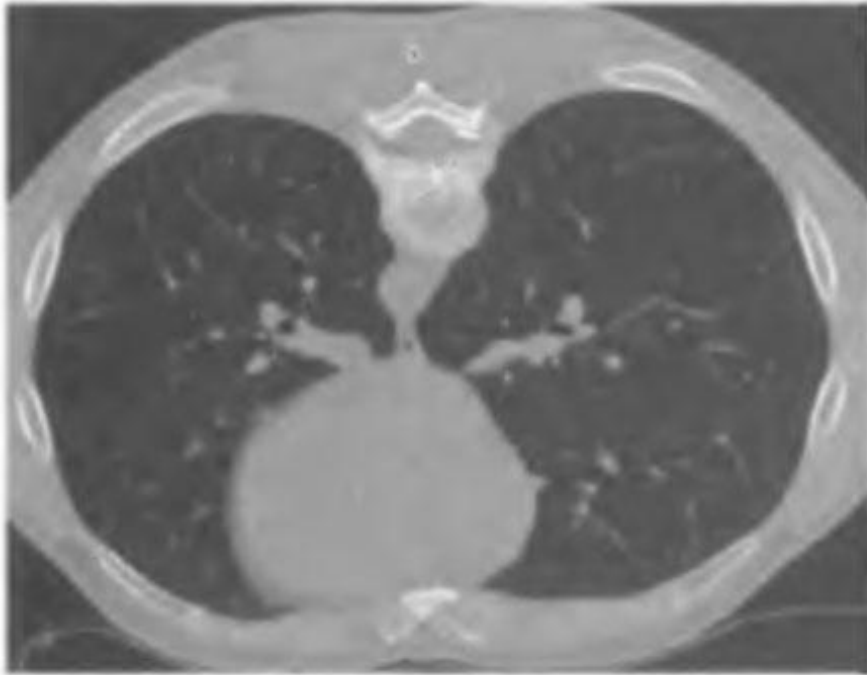
3.1.3.1. Histogram Equalization

Problem statement

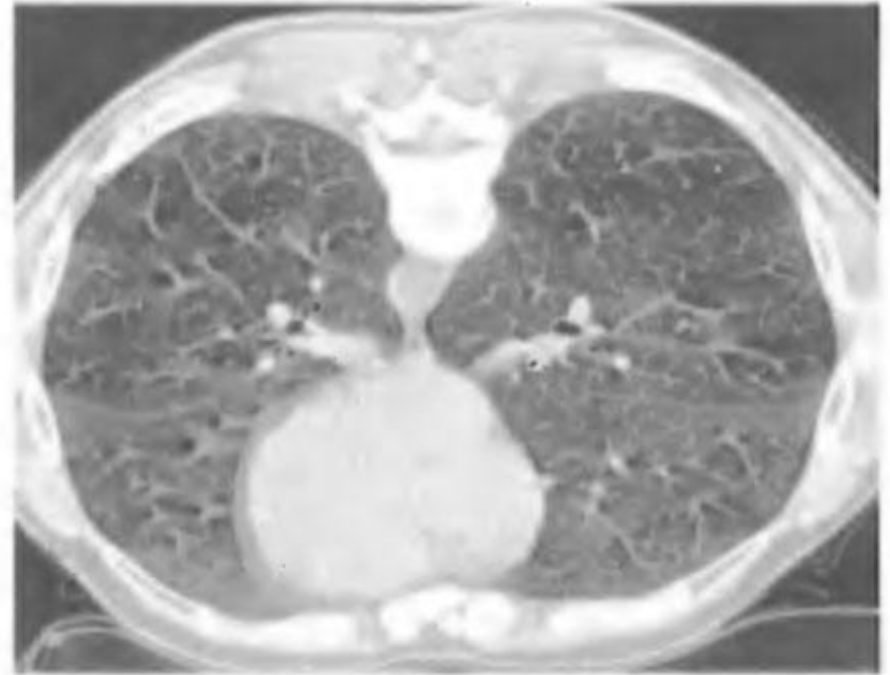
Given image $f(x,y)$, it is necessary to define the nonlinear transformation T such that the output image $g=T(f)$ whose **pdf is uniform**.

3.1.3. Probability Density Function - based mapping

3.1.3.1. Histogram Equalization



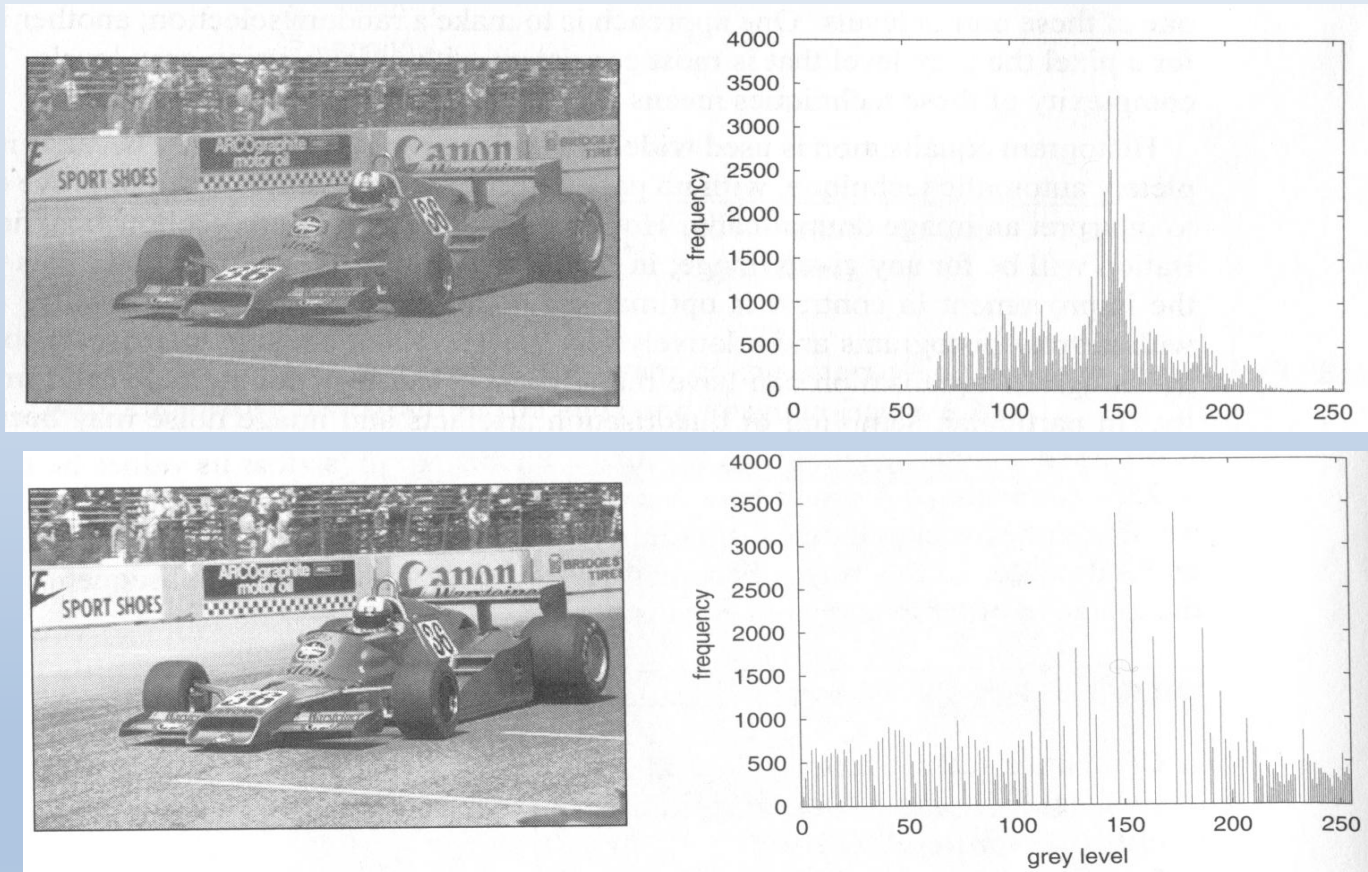
(a)



(b)

3.1.3. Probability Density Function - based mapping

3.1.3.1. Histogram Equalization



3.1.3. Probability Density Function - based mapping

3.1.3.1. Histogram Equalization

Method

$$g = T(f) \Rightarrow p_g(g) = \frac{p_f(f)}{|dT(f)/df|}$$

$$p_g(g) = 1 \Rightarrow T(f) = \int_0^f p_f(w)dw$$

$$g_k = T(f_k) = \sum_{j=0}^k p_f(f_j) = \sum_{j=0}^k \frac{n_j}{n}$$

3.1.3. Probability Density Function-based mapping

3.1.3.1. Histogram Equalization (**Algorithm**)

Step1. Create an array H of length nG initialized with 0 values (for an $N \times M$ image f of nG grey-levels).

Step2. Form **the image histogram** of f , save to H

$$H[f(x, y)] + = 1$$

Step3. Form **the cumulative image histogram** of f , save to T

$$T[0] = H[0]; \quad T[p] = T[p-1] + H[p], \quad p = 1, 2, \dots, nG-1$$

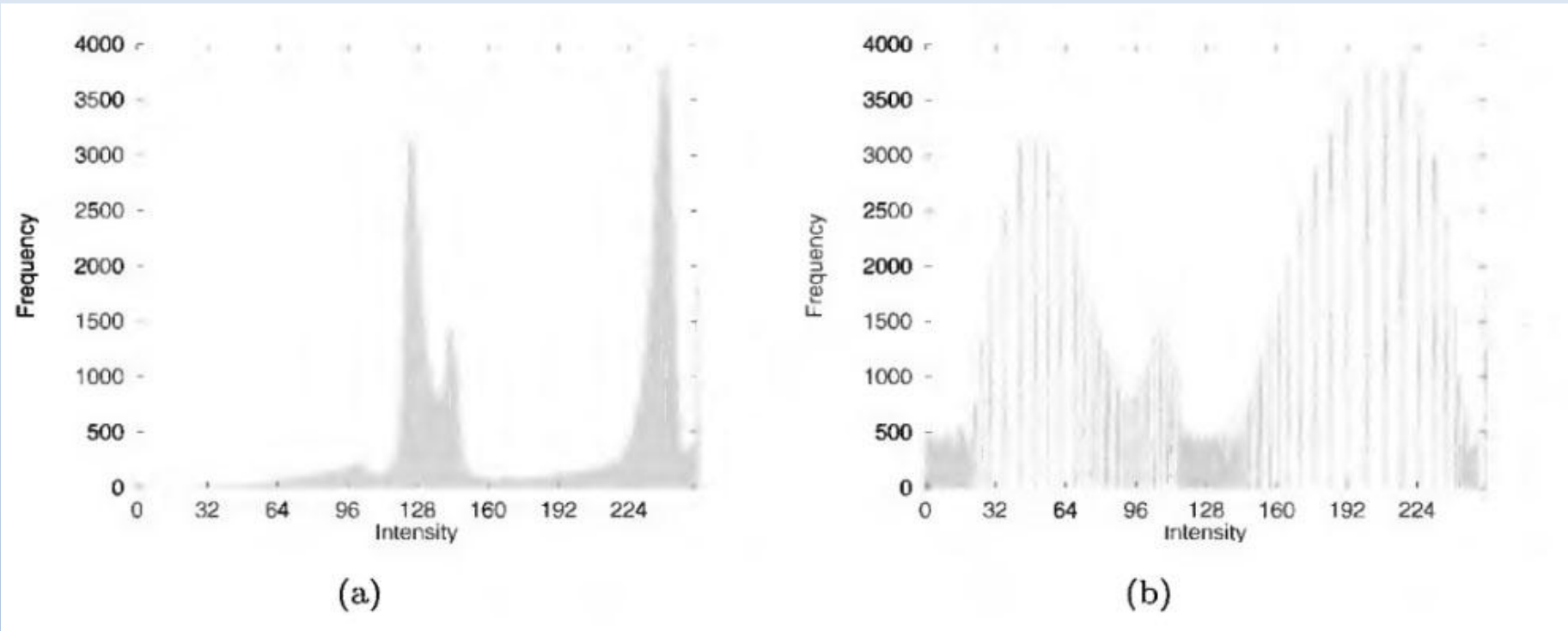
Step4. Constructing **a lookup table T** in range $[0; nG-1]$

$$T[p] = \text{round}((nG-1) / NM) T[p]$$

Step5. Form the **output image g** : $g(x, y) = T[f(x, y)]$

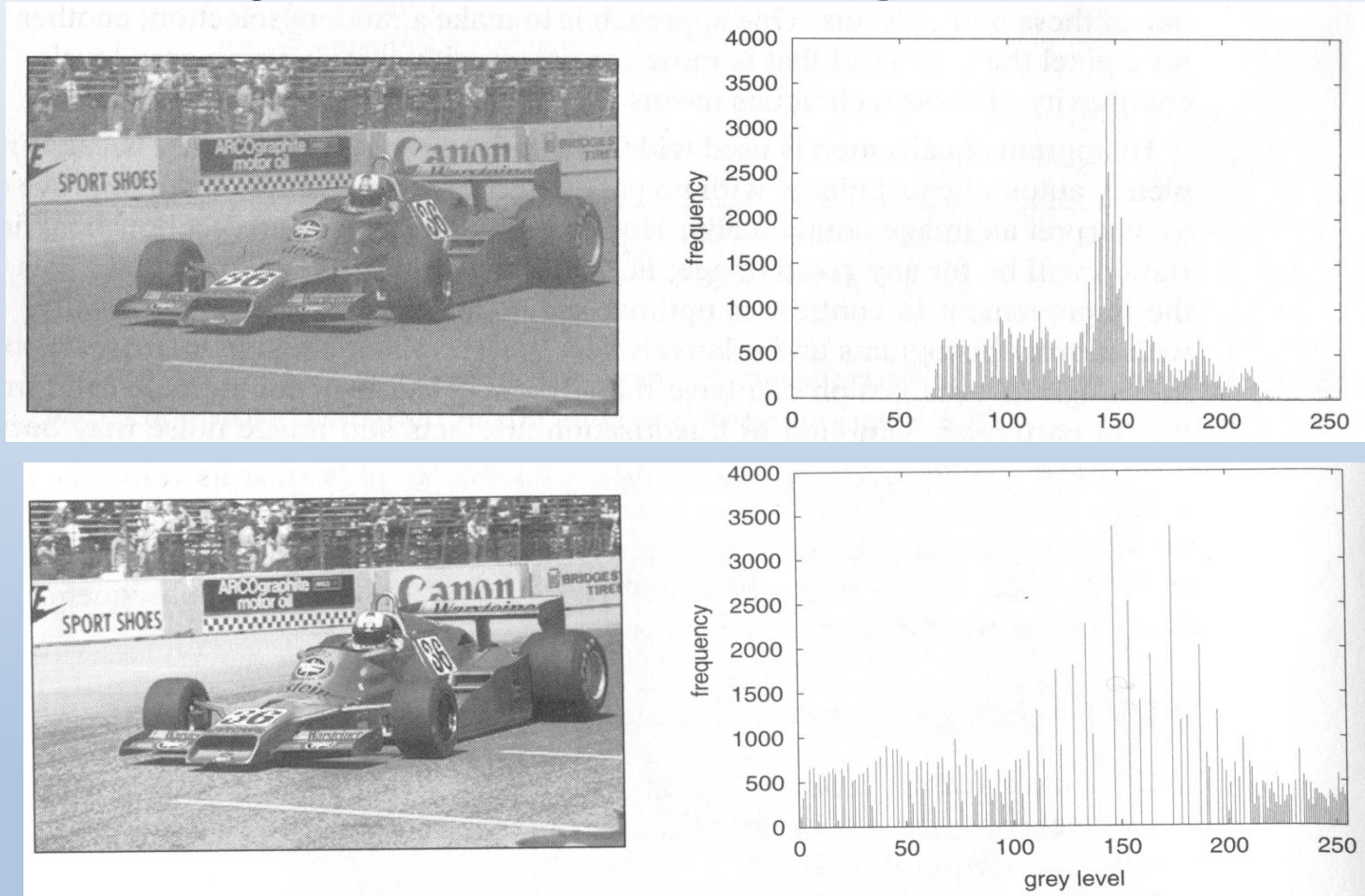
3.1.3. Probability Density Function-based mapping

3.1.3.1. Histogram Equalization (**Algorithm**)



3.1.3. Probability Density Function-based mapping

3.1.3.1. Histogram Equalization (**Algorithm**)



3.1.3. Probability Density Function-based mapping

3.1.3.1. Histogram Equalization

Ex: Histogram equalization of f

2	2	2	2	2	2	2	5
2	6	6	6	6	5	5	5
1	7	7	6	6	5	5	5
1	7	7	1	1	8	6	4
2	9	8	8	8	8	6	4
2	9	10	10	11	12	12	3
2	9	9	10	10	14	13	3
2	2	2	2	2	3	3	3

3.1.3. Probability Density Function-based mapping

3.1.3.1. Histogram Equalization

$nG=15$, $N.M=64$

f	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14
H	0	4	16	5	2	7	8	4	5	4	4	1	2	1	1
T	0	4	20	25	27	34	42	46	51	55	59	60	62	63	64
TR	0	1	4	5	6	7	9	10	11	12	13	13	14	14	14

$$T[0] = H[0]$$

$$T[p] = T[p-1] + H[p], \quad p = 1, 2, \dots, nG - 1$$

$$T[p] = \text{round}((nG - 1 / NM) T[p])$$

3.1.3. Probability Density Function-based mapping

3.1.3.1. Histogram Equalization

2	2	2	2	2	2	2	5
2	6	6	6	6	5	5	5
1	7	7	6	6	5	5	5
1	7	7	1	1	8	6	4
2	9	8	8	8	8	6	4
2	9	10	10	11	12	12	3
2	9	9	10	10	14	13	3
2	2	2	2	2	3	3	3

4	4	4	4	4	4	4	7
4	9	9	9	9	7	7	7
1	10	10	9	9	7	7	7
1	10	10	1	1	11	9	6
4	12	11	11	11	11	9	6
4	12	13	13	13	14	14	5
4	12	12	13	13	14	14	5
4	4	4	4	4	5	5	5

$$g(x, y) = T[f(x, y)]$$

3.1.3. Probability Density Function-based mapping

3.1.3.1. Histogram Equalization

g	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14
Hg	0	4	0	0	16	5	0	7	0	8	4	4	4	5	4

4	4	4	4	4	4	4	7
4	9	9	9	9	7	7	7
1	10	10	9	9	7	7	7
1	10	10	1	1	11	9	6
4	12	11	11	11	11	9	6
4	12	13	13	13	14	14	5
4	12	12	13	13	14	14	5
4	4	4	4	4	5	5	5

3.1.3. Probability Density Function - based mapping

3.1.3.2. Histogram Specification

Problem statement

Given an image $f(x,y)$, a predefined histogram $p_g(g)$, it is necessary to define the non-linear transformation F such that the output image $g=F(f)$.

3.1.3. Probability Density Function - based mapping

3.1.3.2. Histogram Specification

Method

$$s = T(f) = \int_0^f p_f(w)dw$$

$$z = G(g) = \int_0^g p_g(w)dw$$

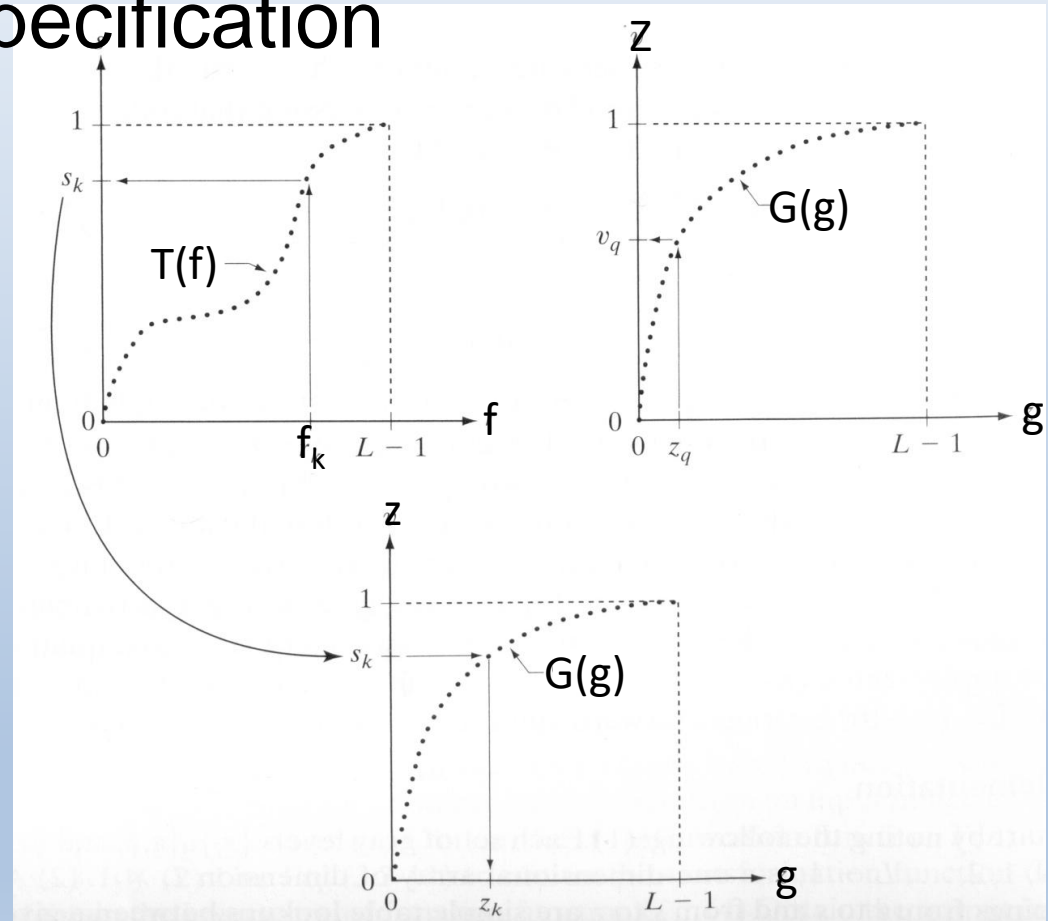
$$g = G^{-1}(z) = G^{-1}(s) = G^{-1}[T(f)]$$

$$g = F(f), F = G^{-1} \circ T$$

3.1.3. Probability Density Function - based mapping

3.1.3.2. Histogram Specification

Method



3.1.3. Probability Density Function-based mapping

3.1.3.2. Histogram Specification (**Algorithm**)

Step1. Create an array H of length nG initialized with 0 values (for an $N \times M$ image f of nG grey-levels).

Step2. Form **the image histogram** of f , save to H_f

$$H_f[f(x, y)] + = 1$$

Step3. Form **the cumulative image histogram** of f , save to T

$$T[0] = H[0]; \quad T[p] = T[p-1] + H_f[p], \quad p = 1, 2, \dots, nG-1$$

Step4. Constructing **a lookup table T** in range $[0; nG-1]$

$$T[p] = \text{round}((nG-1) / NM) T[p]$$

3.1.3. Probability Density Function-based mapping

3.1.3.2. Histogram Specification (**Algorithm**)

Step5. Form **the cumulative image histogram of g, save to G**

$$G[0] = Hg[0]$$

$$G[p] = G[p - 1] + Hg[p], \quad p = 1, 2, \dots, nG - 1$$

Step 6. Constructing **a lookup table G** in range $[0; nG-1]$

$$G[p] = \text{round}((nG - 1 / NM)G[p])$$

Step 7. Form the output image **g**:

$$g(x, y) = G^{-1}[T[f(x, y)]]$$