# Histogram Specification

**CHEN Siyu** 

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#### **Abstract**

This article describes Histogram Specification, whose purpose is to obtain a function f(p) to map the given data into a specified histogram.

This is useful when raw images directly read from sensors are not ideal in brightness or contrast.

#### 1 Problem definition

Given a series of data  $P = \{p_1, p_2, ..., p_N\}, p \in \mathbb{N}$ , its histogram can be expressed as

$$h_P(p) = n_p, (1)$$

in which,  $n_p$  denotes the number of element that values p in sequence P.

Given a desired target histogram  $h_Q(q) = n_q$ , the objective is to find a monotonic mapping function f(p) = q such that:

$$P' = \{f(q_1), f(q_2), ..., f(q_N)\} = \{p'_1, p'_2, ..., p'_N\},$$
(2)

$$h_{P'}(p) = h_Q(p). (3)$$

### 2 Derivations

The former problem definition is in discrete form. In some extreme but not rare cases, optimal solution may not exist for discrete histogram specification. For example:

$$P = \{1, 1, 1, 1, 1\},\tag{4}$$

$$Q = \{1, 2, 3, 4, 5\}. \tag{5}$$

In the above scenario, a desired f(p) does not exist.

To simplify the problem, we assume continuous distribution on set P and Q such that  $P, Q \in [0, 1]^N$ . In other word,  $h_P(x)$  and  $h_Q(x)$  are both continuous.

According to the definition of the problem, several constrains hold as listed below:

$$h_P(x) = h_O(f(x)), \tag{6}$$

$$\frac{\partial f(x)}{\partial x} \ge 0,\tag{7}$$

in which  $x \in [0, 1]$ .

According to Eq. 6, if the distribution density is identical everywhere, then the cumulative distribution shall also be equal-valued.

$$\int_{0}^{x} h_{P}(z) \, \mathrm{d}z = \int_{0}^{f(x)} h_{Q}(z) \, \mathrm{d}z. \tag{8}$$

The above equation can be transformed into the following form:

$$\int_{0}^{x} h_{P}(z) \, dz = \int_{0}^{f(x)} h_{Q}(z) \, dz, \tag{9}$$

$$H_P(x) = H_Q(f(x)), (10)$$

where the original functions  $H_P(x), H_Q(x)$  are the cumulative intensity functions of  $h_P(x), h_Q(x)$ . To obtain f(x), the inverse function of  $H_Q$  can be applied to both side:

$$H_Q^{-1}(H_P(f(x))) = H_Q^{-1}(H_Q(f(x)))$$
 (11)

$$H_Q^{-1}(H_P(x)) = f(x),$$
 (12)

which is the desired optimal solution.

## 3 Histogram Equalization

A mapping function f(x) that equalizes the histogram can be computed given that:

$$h_O(x) = C, (13)$$

where C is a constant value.

Under continuous distribution assumption, we have:

$$H_Q(x) = x, (14)$$

$$H_Q(x) = x,$$
 (14)  
 $H_Q^{-1}(x) = x,$  (15)  
 $f(x) = H_P(x).$  (16)

$$f(x) = H_P(x). (16)$$