

Mathematical Aspects/Derivation of Histogram Equalization

- Consider continuous values of intensity, rather than discretized.
- r = grey level of image T to be equalized
 r normalized to $[0, 1]$
 $0 \rightarrow$ dark = Black
 $1 \rightarrow$ Bright = White.

Goal: Design a Transformation.

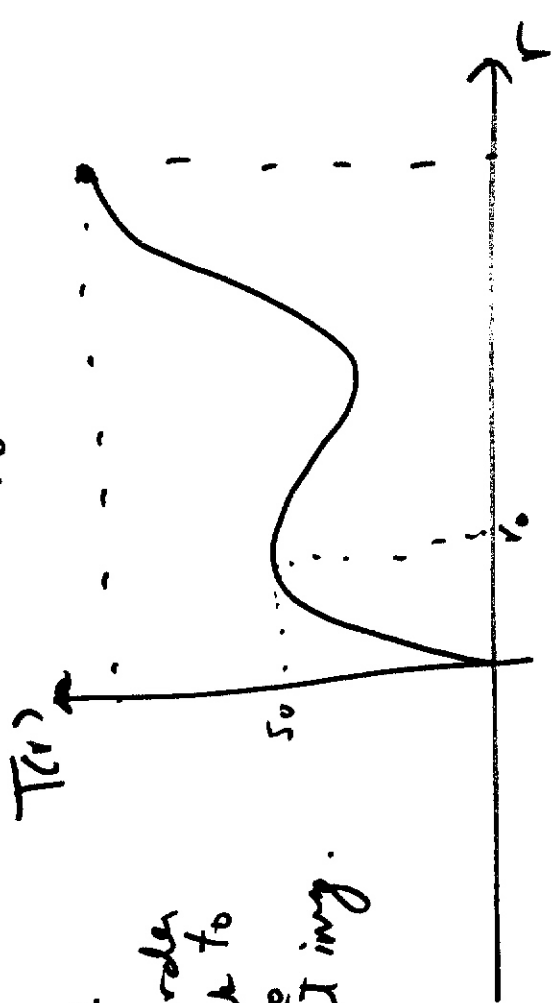
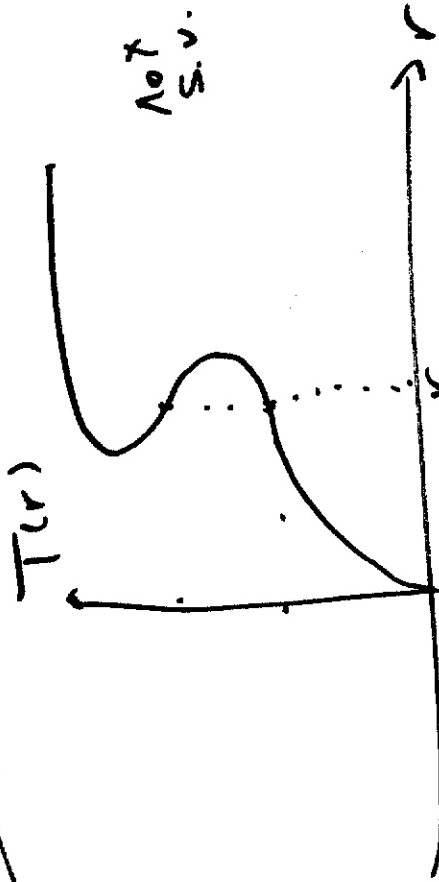
$$\begin{array}{l} \text{histogram} \\ \text{method of} \\ \text{equalization} \end{array} \leftarrow T(r) = S \quad \begin{array}{l} 0 \leq r \leq 1 \\ 0 \leq S \leq 1 \end{array}$$

- Assumptions about $T(r)$.

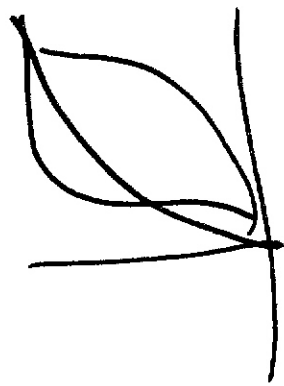
(a) $T(r)$ single valued and monotonically increasing

in the interval $0 < r < 1$

→ for inverse Transform to exist



→ preserve.
increasing order
from black to
the white
in outputting.



(b) $0 \leq T(r) \leq 1$ for $0 \leq r \leq 1$

\rightarrow output grey levels in the same range as input levels.

$$T(r) = S$$

$$T^{-1}(s) = r$$

$r =$ input intensity
 $s =$ output intensity.

Result from prob Theory:

if $T^{-1}(s)$ satisfies condition (a) then.

$$P_S(s) = P_r(r) \left| \frac{dr}{ds} \right|$$

Consider CDF: as a Transformation.

$$S = T(r) = \int_0^r P_r(\omega) d\omega$$

$$\frac{ds}{dr} = \frac{d}{dr} \left\{ \int_0^r P_r(\omega) d\omega \right\} = P_r(r)$$

$$P_s(s) = P_r(r) \left| \frac{1}{P_r(r)} \right| = 1 \quad 0 \leq s \leq 1$$

Words:

If $T(r)$ is just a CDF
or just the integral of input
pdf ($P_r(r)$) Then ~~and~~ applying
 $T(r)$ results in a image whose
pdf ($P_s(s)$) is Uniform

Discrete Case

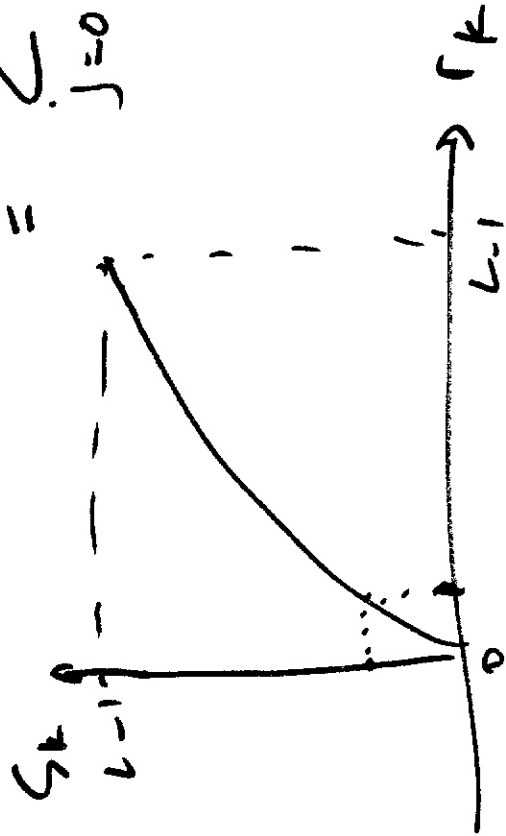
r_k = discrete intensity values. $k=0, \dots, L-1$

$$P_r(r_k) = \frac{n_k}{N} \quad k=0, \dots, L-1$$

n_k = # of pixels that have intensity r_k .

$$S_{k=T}(r_k) = \sum_{j=0}^k P_r(r_j) \leftarrow$$

$$= \sum_{j=0}^k \frac{n_j}{N} \quad k=0, \dots, L-1$$



Histogram Matching

$$T(r) = S$$

Rather than $P_3(s)$ uniform, we want

$P_3(s)$ To match a "desired" pdf given

~~spatial~~ matching

$r =$ pixel values before

$z =$ pixel " after

Can compute $P_r(r)$ from given image.

know, given $P_z(z)$

Goal: what is the transformation $r \rightarrow z$?

Approach :

$$S = T(r) = \int_0^r P_1(\omega) d\omega$$

↗ CDF of r .

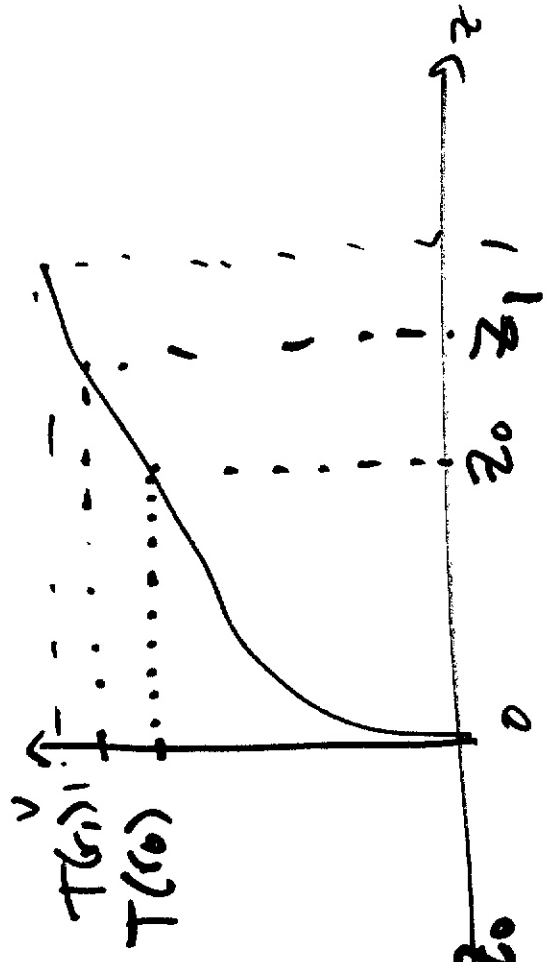
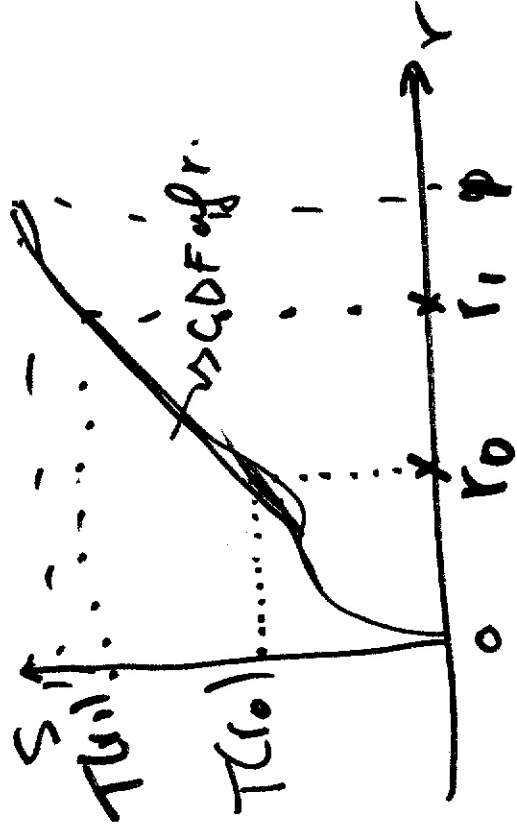
$$v = G(z) = \int_0^z P_2(t) dt$$

↖ CDF of z .

Histogram matching

$$G(z) = T(r)$$

$$z = G^{-1} \{ T(r) \}$$



r_0 → how to find z_0

lookuptable

r_0	z_0
r_1	z_1

→ Matched
histogram of
 $P_T(z)$ desired

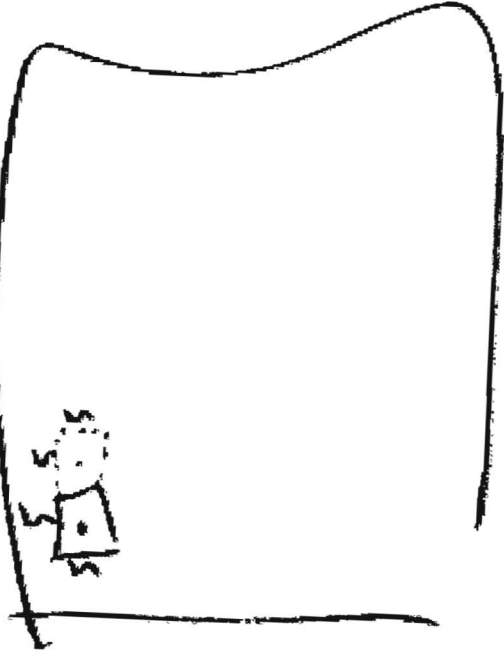
Local Enhancement

Using Local Histogram Analysis

Define square or rectangle neighborhood over each pixel, move center of "window" across the image.

At each location: compute histogram in the window

Do hist. eq.



Use local stats for Edman

stat \rightarrow mean, variance:

$$\text{global mean} = M_G = \sum_{i=0}^{L-1} r_i P(r_i)$$

$r_i \rightarrow$ intensity level i

$i=0, \dots, L-1$

$$\text{global variance} = \text{Var}_G = \sum_{i=0}^{L-1} (r_i - M_G)^2 P(r_i)$$

Local mean: Let S_{xy} neighborhood, subimage centered around (x, y) .

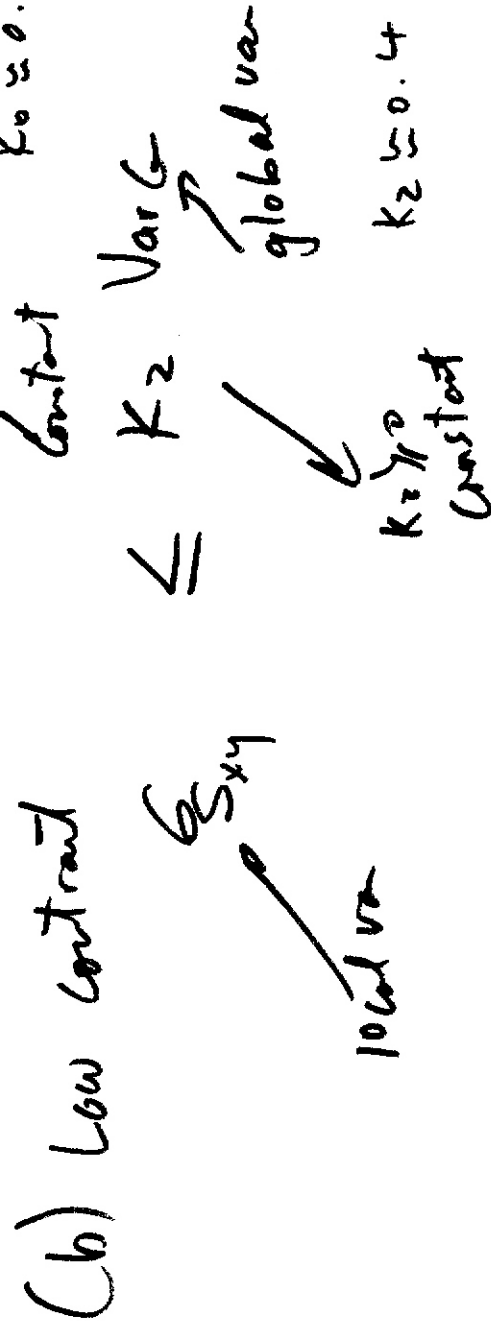
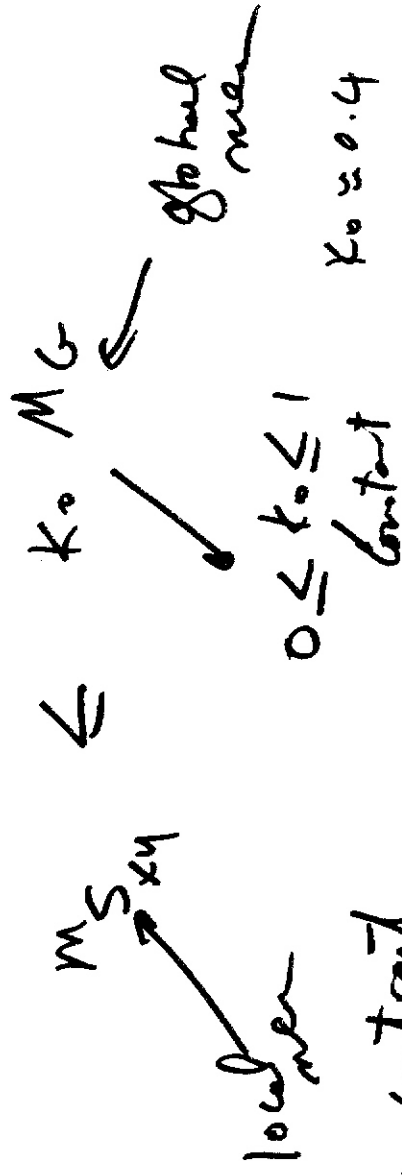
$$\text{local mean} = m_{S_{xy}} = \sum_{(s,t) \in S_{xy}} r_{st} P(r_{st})$$

$$\text{local variance} = \sigma_{S_{xy}}^2 = \sum_{(s,t) \in S_{xy}} (r_{st} - m_{S_{xy}})^2 P(r_{st})$$

Goal: Enhance dark areas while leaving the bright areas unchanged as possible.

- Detect regions that have both.

(a) dark \rightarrow local mean has to be small compared to global mean.



(c) not too low of constraint.
leave flat region unchanged.

$$\sigma_{Sxy} \Rightarrow K_1 \text{ Var } G$$

$$K_1 \approx 0.01$$

$$g(x, y) = \text{proceed} = \left\{ \begin{array}{l} E f(x, y) \text{ if } M_{Sxy} \leq K_0 M_G \\ K_1 \text{ Var } G \leq \sigma_{Sxy} \leq K_2 \text{ Var } G \\ 0 \end{array} \right.$$

otherwise.

Enhancement isig Arithmetic / Logic Operat:

Logic: AND OR NOT \rightarrow functionally complete.

NAND \rightarrow functionally complete.

NOT \rightarrow negative inversion

AND OR \rightarrow Masking.

Imex \rightarrow Subtraction

$$g(x, y) = f(x, y) - h(x, y) \rightarrow \text{mask.}$$