

Image Restoration

1. Subjunctive \rightarrow image "look" better

✓ subjective improvement.

2. Restoration

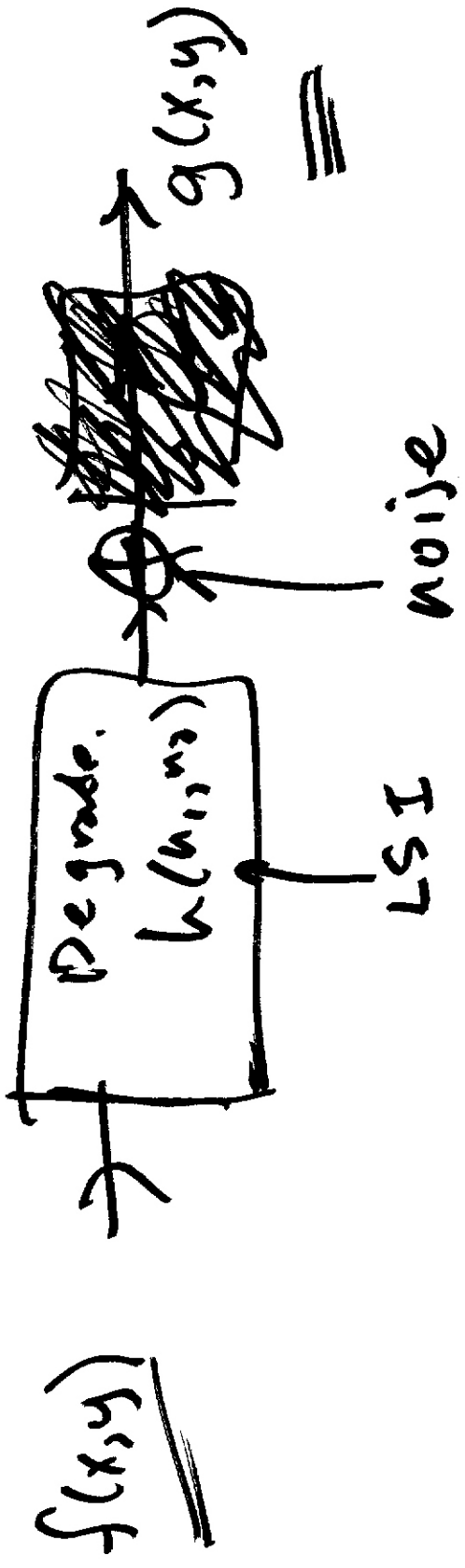
Image has been degraded by something.

- noise

- blur

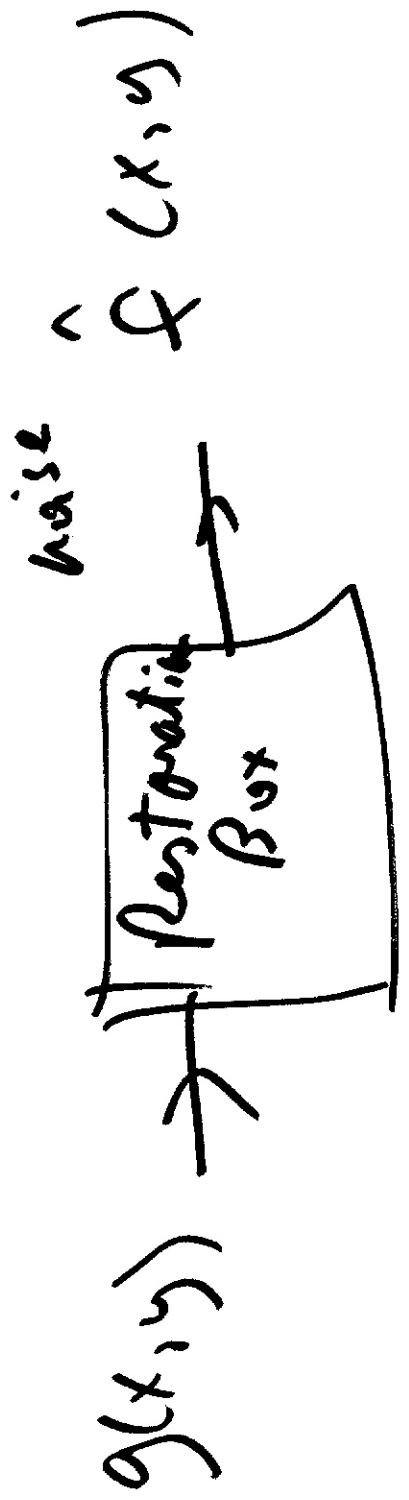
- Atmospheric Turbulence.

objective models \rightarrow Error \rightarrow MSE



$$g(x,y) = h * f + n$$

$$g(x,y) = h(x,y) * f(x,y) + n(x,y)$$



$$\text{minimize } E[(f(x,y) - \hat{f}(x,y))^2]$$

Today.

μ is identity.

corruption \rightarrow noise.

only

$$g(x,y) = f(x,y) + \eta(x,y)$$

observed.

original
input

noise

Assume noise is independent of spatial coordinates
uncorrelated w.r.t. input itself.

① Gaussian

$$= \frac{-(z - \mu)^2}{2\sigma^2} e$$

$$p(z) = \frac{1}{\sqrt{2\pi}\sigma}$$

$$\text{var} = \sigma^2 \quad \text{mean} = \mu$$

electronic noise

- sensor noise

due to poor illumination

or high temperature



70% of values are within one sigma
[μ - σ, μ + σ]

95% within 2 sigma.

② Rayleigh:

$$e^{-\frac{(z-a)^2}{b}} \quad z \geq a$$

$$\frac{z}{b} \quad (z-a)$$

$$p(z) = \begin{cases} \frac{z}{b} & z \geq a \\ 0 & z < a \end{cases}$$

$$z < a$$

$$\sigma^2 = \frac{b(4-\pi)}{4}$$

$$\mu = a + \sqrt{\frac{b\pi}{4}}$$

noise in range imaging app.

③ Erlang (Gauss) noise:

$$p(z) = \begin{cases} \frac{a^b z^{b-1} e^{-az}}{(b-1)!} & z > 0 \\ 0 & z < 0 \end{cases}$$

$z > 0$

$z < 0$

$$\sigma^2 = \frac{b}{a^2}$$

$$\mu = b/a$$

laser imaging:

④ Exponential special case of Erlang.

$$p(z) = \begin{cases} a e^{-az} & z > 0 \\ 0 & z < 0 \end{cases}$$

$z > 0$

$z < 0$

Network

uniform: $a \leq z \leq b$

$$p(z) = \begin{cases} \frac{1}{b-a} \\ 0 \end{cases} \text{ otherwise.}$$

$$\mu = \frac{a+b}{2} \quad \text{var} = \frac{(b-a)^2}{12}$$

random # generation.

Impulse, salt/pepper noise, quick transmit, faulty switching.

$$z = a$$

$$z = b$$

otherwise.

$$p(z) = \begin{cases} p_a \\ p_b \end{cases}$$

if $b > a$, show up as light dots
 if $a > b$, show up as dark dots.

① Mean Filters

1. Arithmetic Mean:

$S_{xy} \rightarrow$ window.

$m \times n$.

$$\hat{f}(x, y) = \frac{1}{mn}$$

$$\sum_{(s,t) \in S_{xy}} g(s, t)$$

2. Geometric Mean: $\left[\prod_{(s,t) \in S_{xy}} g(s, t) \right]^{1/mn}$

similar to arithmetic but less potent

3. ~~to~~ Harmonic mean filter.

works for salt & pepper
 otherwise:
 salt & pepper
 for

$$\hat{f}(x, y) = \frac{m^v}{\sum_{(s,t) \in S_{xy}} \frac{1}{g(s,t)}}$$

4. Center Harmonic.

$$\hat{f}(x, y) = \frac{\sum_{(s,t) \in S_{xy}} g(s,t)}{Q}$$

$Q > 0 \rightarrow$ remove pepper noise

$Q < 0 \rightarrow$ remove salt noise

Order Statistics

1. Median:

look at S_{xy} . find median
replace \bar{x} (x, y) with the
median value.

2. Max

$$\hat{f}(x, y) = \max_{S_{xy}} g(s, t)$$

also removes
some data points.

3. Min

$$\hat{f}(x, y) = \min_{S_{xy}} g(s, t)$$

also removes some white points

$$\frac{1}{2} \{ \max + \min \}$$

4. Midpoint

Alpha Trimmed mean filter

$$\hat{f}(x, y) = \frac{1}{mn - d} \sum_{(s,t) \in S_{xy}} g(s, t) \quad (G(x, y))$$

S_{xy} m x n.

excluding.

$$g_r(s, t) \rightarrow g(s, t)$$

$d/2$ brightest grayscale

and $d/2$ darkest " "

5	4	3	2
1	1	.	.
.	.	.	.
.	.	.	.

reference

~~kernel~~ ~~kernel~~

" "

Adaptive local noise reduction

$$\hat{f}(x, y) = g(x, y) - \frac{\sigma_N^2}{\sigma_L^2} (g(x, y) - m_L)$$

$$m_L = \text{local mean} = \frac{1}{mn} \sum_{(s,t) \in S_{xy}} g(s, t)$$

$\sigma_L^2 = \text{local variance}$.

$\sigma_N^2 = \text{noise variance}$.

if $\sigma_N^2 < \sigma_L^2 \rightarrow \hat{f} \approx g$ good

if $\sigma_N^2 > \sigma_L^2 \rightarrow$ then, approximate $\frac{\sigma_N^2}{\sigma_L^2} \approx 1$

$\rightarrow \hat{f} \approx m_L$

Image ~~Restoration~~ Restoration

Frequency domain

Spatial domain

salt/pepper

"Additive Gaussian noise"

Adaptive Median filters

Adaptive Median Filter

looking at neighborhood $S_{x,y}$ around pixel (x,y)

$Z_{x,y}$ = pixel value at center of $S_{x,y}$
 Z_{max} = max value of intensity over $S_{x,y}$

- z_{min} = min value in S_{xy}

- z_{med} = median value in S_{xy}

outline

- keep increasing window size until z_{med} is not an impulse.
 $z_{min} > z_{med} < z_{max}$

- When this happens, check z_{xy} .

- If z_{xy} is not an impulse \rightarrow output z_{xy}

- If z_{xy} is an impulse \rightarrow output z_{med}

z_{med} is not an impulse.

Pseudo code

Part A

if $z_{min} < z_{med} < z_{max}$
Then go to part B. $\rightarrow z_{med}$ is not an impulse.

else. if $z_{window} < z_{max}$
window \leftarrow window + 1, go to part A
output z_{xy} .

else.
if $z_{min} < z_{xy} < z_{max}$ $\rightarrow z_{xy}$ is not an impulse
output z_{xy}

else. output z_{med}