

Lec 19

Image Restoration

Degraded Images:

Cause of blur

- ↳ Camera translation
- ↳ Environment
- ↳ Device noise
- ↳ Quantisation noise

Optical Blur → Caused due to limited depth of field

Motion Blur

Spatial Quantisation (Discrete pixels)

Additive intensity noise

→ Solution:

→ Light field camera

→ Multifocus Image Fusion

Restoration from camera shakes:

$$B = K * L + N \rightarrow \text{Sensor Noise}$$



Considers blurred image as convolution with degradation
Deconvolution is done to restore img.

Image enhancement is subjective but image restoration is objective

Degradation Model

Model as convolution with shift-invariant filter $h(x,y)$

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

↓
Gaussian is preferred
 $\frac{1}{2\pi\sigma^2} e^{-x^2/2\sigma^2}$

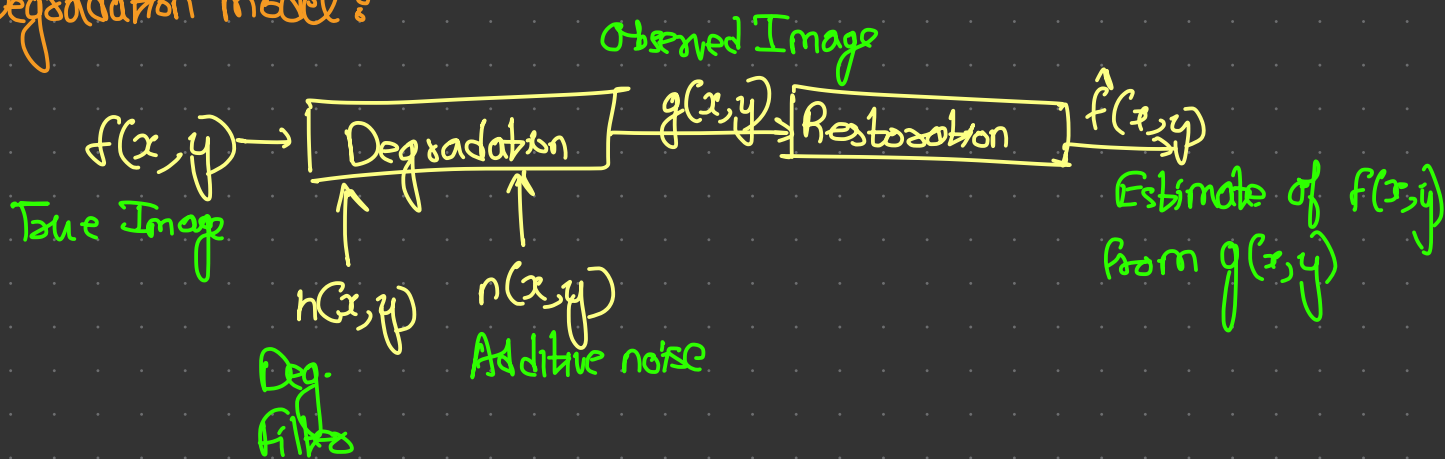
Loss of info (blurring attenuates high spatial frequency and noise)

→ Out of focus is modelled as gaussian

Limitations

Loss of info (blurring attenuates high spatial freq.) & noise

Degradation model:



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

$$\updownarrow$$

$$G(u,v) = H(u,v) F(u,v) + N(u,v)$$

Assumptions

→ Imp. of spatial location
↳ Except periodic noise
→ Uncorrelated with image

→ Degradation function (H)
↳ Linear
↳ Position-invariance

→ Doesn't depend on pixel coordinate but rather on functional value

Noise Models:

→ Gaussian

↳ Sensor Noise
↳ low/high beam

→ Uniform Noise

Random Num

Generators

→ Rayleigh Noise

↳ Noise in range
image

→ Gamma Noise
Loses Images

→ Exponential Noise

Loses Imaging

→ Impulse Noise
(salt & pepper)

Quick Transients
(faulty switching)

→ How to study system noise

↳ Case I: Imaging system available

↳ Noise calibration [capture set of "flat image"]

↳ Select model with better statistical test score

Case 2:

Only images available

↳ Estimate parameters of pdf from patches of constant by intensity

↳ Compute variance & mean from intensity levels

Restoration: (In presence of noise)

↳ Use spatial filtering techniques

Mean filters → Geometric mean filter

↳ Salt & Gaussian [Bad for peppers]

Harmonic filter:

Contraharmonic mean filter

↳ Salt-pepper noise

$Q < 0$ → salt

$Q > 0$ → Peppers

$Q = -1$ → harmonic mean filter

$Q = 0$ → arith. mean filter

Midpoint filter → Alpha trimmed filter

Remove the lowest & highest intensities

$d = 0$ [Arith. Mean]

$d = mn - 1$ [Median]

Adaptive Local Noise Reduction → Variance of overall noise

$$f(x, y) = g(x, y) - \frac{\sigma_n^2}{\sigma_z^2} [g(x, y) - m_z]$$

↳ Local Mean

$\sigma_z^2 > \sigma_n^2$ → Edge, preserve it

↳ Local Variance

$\sigma_z^2 = \sigma_n^2$ → Return mean local area

Periodic noise is

↳ Band pass / band reject

Estimation of degradation

3 ways

- ↳ Observation
- ↳ Experimentation
- ↳ Modelling

$$H_S(u, v) = \frac{G_S(u, v)}{\hat{F}_S(u, v)}$$

$$H(u, v) = \frac{G(u, v)}{\hat{F}_A} \quad \text{↳ length of impulse}$$

Motion blur

→ Amb of light hitting sensor changes fast

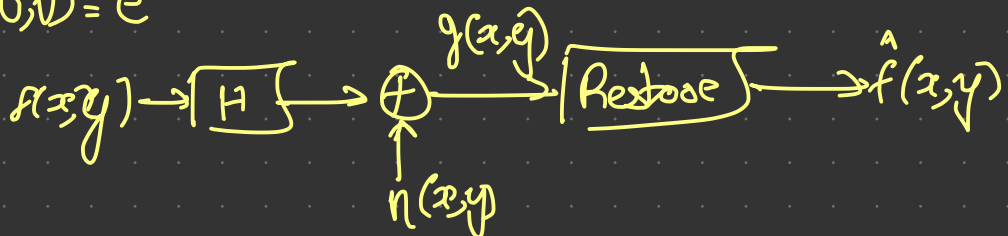
→ Camera motion

$$g(x, y) = \int_0^T f(x - x_0(t), y - y_0(t)) dt$$

$$x_0(t) = \frac{at}{T}, y_0(t) = \frac{bt}{T} \rightarrow \text{Linear motion}$$

Modelling Atmospheric turbulence

$$H(u, v) = e^{-k^2(u^2 + v^2)^{5/6}}$$



Direct Inverse Filtering

Assume H is known,

$$\hat{F}(u, v) = \frac{G(u, v)}{H(u, v)} = F(u, v) + \frac{N(u, v)}{H(u, v)}$$

→ Limit Noise upto certain value