

Restoration

Monday, January 6, 2020 10:47 AM

- reduction or elimination of degradations in an image

- Ex:
- camera out of focus
 - camera or object in motion
 - noisy acquisition system
 - geometric warp due to lens distortion

Applications:

- forensics (blurry picture of license plate of getaway car)
- military reconnaissance
- Hubble space telescope (spherical aberration)
- consumer photos

Image observation model

$$v(x,y) = g[w(x,y)] + n(x,y)$$

$g[\]$ is a pointwise nonlinearity

$n(x,y)$ is noise (possibly image-dependent)

$$w(x, y) = \iint h(x, y; x', y') u(x', y') dx' dy'$$

(blurring operation)

$u(x, y)$ is the original image

For shift-invariant system, point-spread function (PSF) is constant over image.

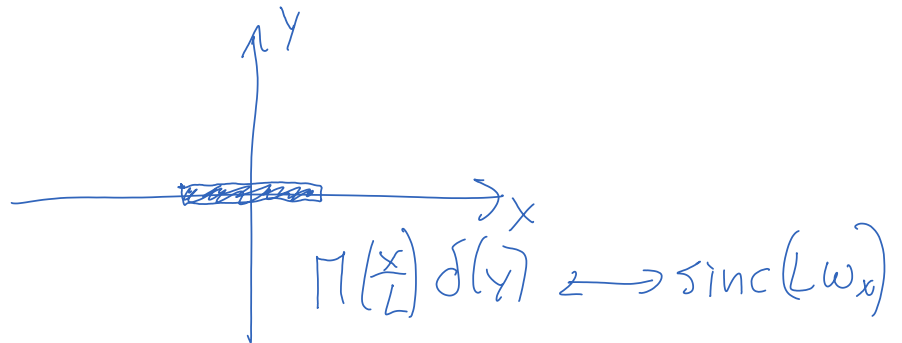
PSF \equiv impulse response

$$h(x, y; x', y') = h(x - x', y - y'; 0, 0) \triangleq h(x - x', y - y')$$

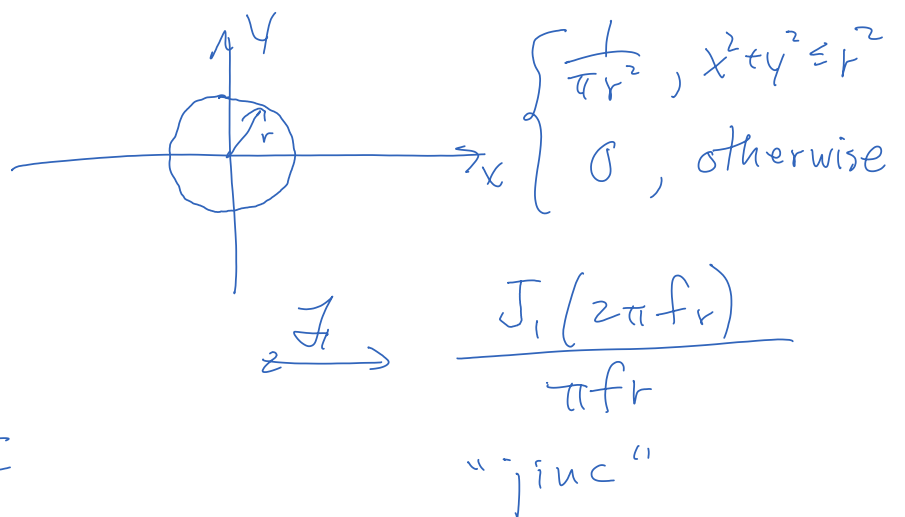
$$\text{Then } w(x, y) = h(x, y) * u(x, y)$$

blur models

horizontal
uniform
motion



out of focus
lens

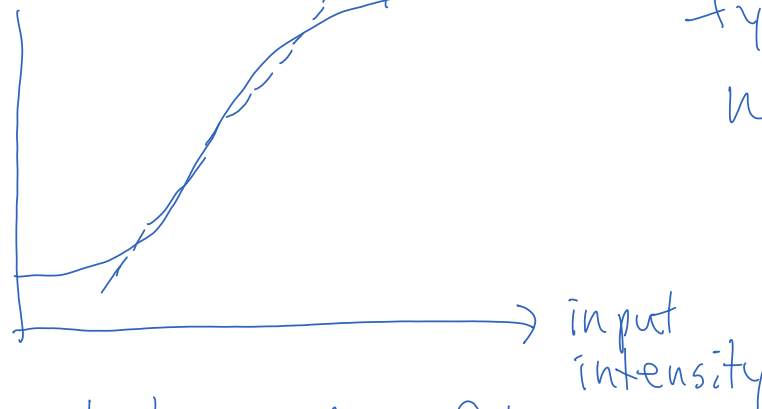


$$f = \sqrt{f_x^2 + f_y^2}$$

J_1 is Bessel function

of first kind of order 1

* sensor nonlinearity



typical sensor nonlinearity

- photographic film
- some CMOS sensors
- either include nonlinear model in restoration or assume image acquired in linear region

* noise

$$\eta(x, y) = \sqrt{w(x, y)} \eta_1(x, y) + \eta_2(x, y)$$

$\sqrt{w(x, y)} \eta_1(x, y)$ is Poisson-distributed —

for example, random photon emissions

$\eta_2(x, y)$ is Gaussian distributed —

electronic (thermal) noise

* can also model quantization, although this is actually a uniform distribution

From this point, we assume:

- shift-invariant blur
- linear sensor
- image-independent noise

Inverse filter

$$v(x, y) = h(x, y) * u(x, y) + n(x, y)$$

$$v(m, n) = h(m, n) * u(m, n) + n(m, n)$$

$$V(\omega_m, \omega_n) = H(\omega_m, \omega_n) U(\omega_m, \omega_n) + N(\omega_m, \omega_n)$$

(stack pixels
into tall vectors)

$$v = H u + n$$

inverse
filter :

$$\hat{U}(\omega_m, \omega_n) = \frac{1}{H(\omega_m, \omega_n)} V(\omega_m, \omega_n)$$

$$= U(\omega_m, \omega_n) + \frac{N(\omega_m, \omega_n)}{H(\omega_m, \omega_n)}$$

two problems:

- noise amplification
- may not be invertible

\Rightarrow pseudo inverse

$$\frac{1}{H(\omega_m, \omega_n)}, |H(\omega_m, \omega_n)| > \epsilon$$

$$H^-(\omega_m, \omega_n) = \begin{cases} 0 & , \text{ otherwise} \end{cases}$$

Wiener filter

Minimize $E\{[u(m,n) - \hat{u}(m,n)]^2\}$

for choice of restoration filter $g(m,n)$

Where $\hat{u}(m,n) = g(m,n) * v(m,n)$

- assumes we have a statistical description of $u(m,n)$ + noise

$$r_{uv}(m,n) = E[u(m',n') v(m'-m, n'-n)]$$

$$r_{uv}(m,n) = g(m,n) * r_{vv}(m,n)$$

crosscorrelation
of u & v

autocorrelation
of v

\Rightarrow In freq. domain,

$$S_{uv}(\omega_m, \omega_n) = G(\omega_m, \omega_n) S_{vv}(\omega_m, \omega_n)$$

$$G(\omega_m, \omega_n) = S_{uv}(\omega_m, \omega_n) S_{vv}^{-1}(\omega_m, \omega_n)$$

If $v(m,n) = h(m,n) * u(m,n) + n(m,n)$

then

$$S_{vv}(\omega_m, \omega_n) = |H(\omega_m, \omega_n)|^2 S_{uu}(\omega_m, \omega_n) + S_{nn}(\omega_m, \omega_n)$$