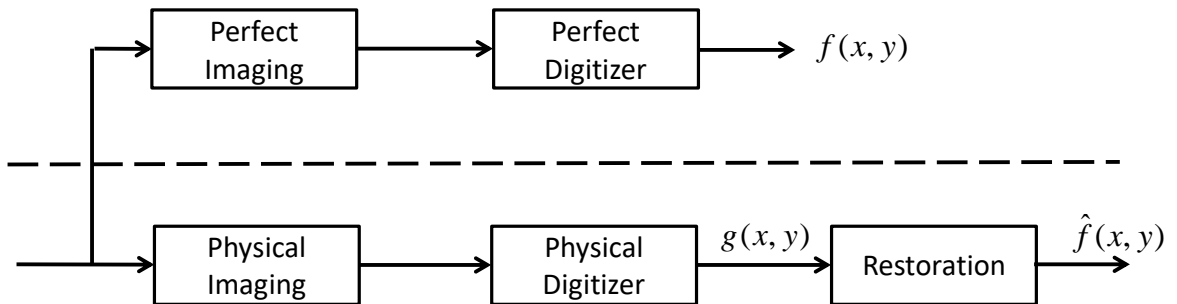


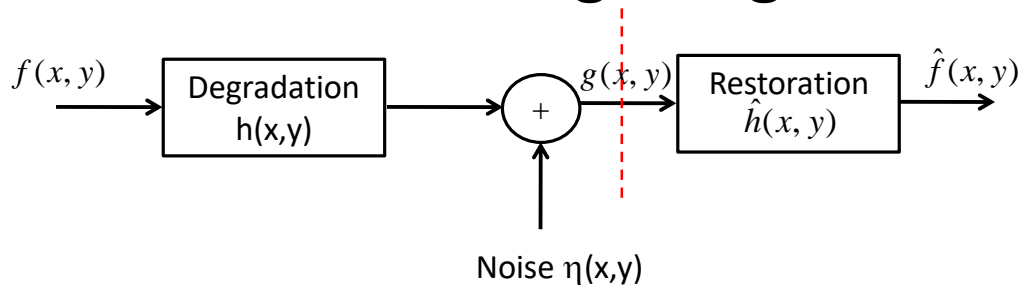
图像复原 Image Restoration

Imaging Modelling



线性模糊模型

Linear Model of Image Degradation



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

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

$$\hat{f}(x, y) = \hat{h}(x, y) * g(x, y) = \hat{h}(x, y) * [h(x, y) * f(x, y) + \eta(x, y)]$$

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

逆滤波（Inverse Filter）

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

$$\text{if } \hat{H}(u, v) \cdot H(u, v) = 1 \implies \hat{H}(u, v) = \frac{1}{H(u, v)}$$

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

Wiener Filter

$$E\{|f(x, y) - \hat{f}(x, y)|^2\} \rightarrow \min$$

$$\hat{F}(u, v) = \frac{H^*(u, v)S_f(u, v)}{S_f(u, v)|H(u, v)|^2 + S_n(u, v)}G(u, v)$$

$$\hat{F}(u, v) = \frac{H^*(u, v)}{|H(u, v)|^2 + S_n(u, v)/S_f(u, v)}G(u, v)$$

$$S_n(u, v) = |N(u, v)|^2$$

$$S_f(u, v) = |F(u, v)|^2$$

$$K = S_n(u, v)/S_f(u, v)$$

运动模糊（匀速直线运动）

Model of Uniform Linear Motion

$$g(x) = \int_0^{\tau} f(x - vt) dt$$

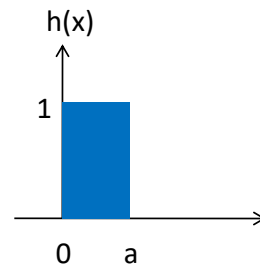
τ : exposure time / shutter opened
 v : velocity of motion

$$g(x) = \int_0^a f(x - t) dt$$

a : distance

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

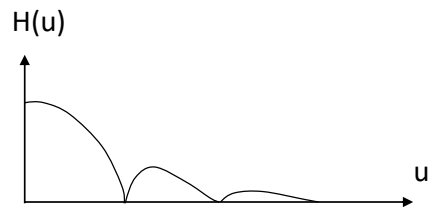
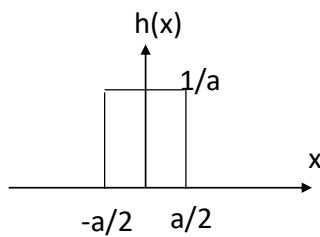
$$h(x) = \begin{cases} 1 & (0 \leq x \leq a) \\ 0 & \text{otherwise} \end{cases}$$



运动参数估计

Estimation of Motion Distance

Cannon Method



$$H(u) = \int_{-a/2}^{a/2} \frac{1}{a} e^{-j2\pi ux} dx = \frac{1}{-j2\pi u} e^{-j2\pi ux} \Big|_{-a/2}^{a/2} = \frac{e^{-j\pi ua} - e^{j\pi ua}}{-j2\pi u}$$

$$H(u) = \frac{\sin(\pi ua)}{\pi ua} = \text{sinc}(\pi ua)$$

Identification of blur parameters from motion blurred images

Y. Yitzhaky and N.S. Kopeika

Graphical Models and Image Processing, 59(5), 1997:310-320

Estimation of a correct point spread function (PSF)

$f(x,y)$: given ideal image

$g(x,y)$: the motion blurred image

$h(x,y)$: linear shift-invariant PSF

$n(x,y)$: random noise

degradation model:

$$g(x, y) = \iint h(x - x', y - y') f(x', y') dx' dy' + n(x, y)$$

assumption:

Exposure time : about 1/30 s

→ no extreme changes in motion velocity

Identification of Motion Direction

Motion Blurring → Low pass filter

⇒ intensity at low frequencies increased
resolution decreased in motion direction

derivative → High pass filter

⇒ intensity at low frequencies suppressed
intensity at high frequencies increased

⇒ Derivative in motion direction should suppress more of image intensity than that in other directions

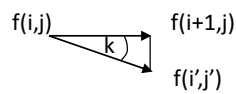
motion direction → lower power spectrum of derivative

implementation of derivative

$$\Delta f(i, j)_{[0^\circ]} = f(i+1, j) - f(i, j)$$

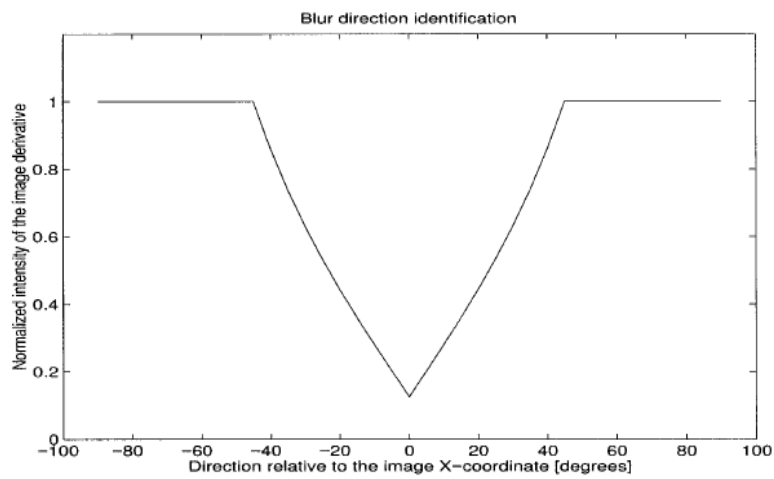
$$\Delta f(i, j)_{[k^\circ]} = f(i', j') - f(i, j)$$

$$0 \leq k \leq 45^\circ \quad f(i', j') = f(i+1, j) \cdot (1 - \tan(k)) + f(i+1, j+1) \cdot \tan(k)$$



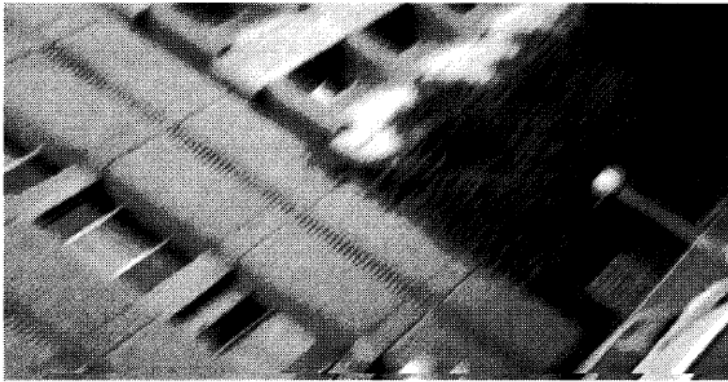
f(i,j)	f(i+1,j)
	f(i',j')
	f(i+1,j+1)

$$I(\Delta f)_{[k^\circ]} = \sum_i \sum_j |\Delta f(i, j)_{[k^\circ]}|$$



$$I(\Delta f)_{[k^\circ]} = \sum_i \sum_j |\Delta f(i, j)_{[k^\circ]}|$$

blurred image



Blur direction identification

