

浙江大学



本科实验报告

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Consider a linear, position invariant image degradation system with impulse response

$$h(p, q) = e^{-(p^2 + q^2)}$$

- Suppose that the input to the system is a binary image consisting of a white vertical line of infinite small width located at $x = a$, on a black background.

Such an image can be modeled as $f(x, y) = \delta(x - a)$.

Assume negligible noise and use $g(x, y) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(\alpha, \beta) h(x - \alpha, y - \beta) d\alpha d\beta$ to calculate the output image $g(x, y)$.

$$f(x, y) \rightarrow \boxed{h(x, y)} \rightarrow g(x, y)$$

(noise neglected)

$$\begin{aligned} \therefore \text{退化图像 } g(x, y) &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(\alpha, \beta) h(x - \alpha, y - \beta) d\alpha d\beta \\ &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} b(\alpha - a) h(x - \alpha, y - \beta) d\alpha d\beta \end{aligned}$$

由 $b(t)$ 筛选性质: $\int_{-\infty}^{\infty} f(t) b(t - t_0) dt = f(t_0)$

$$\begin{aligned} g(x, y) &= \int_{-\infty}^{\infty} h(x - \alpha, y - \beta) d\beta \\ &= \int_{-\infty}^{\infty} e^{-(x - \alpha)^2 + (y - \beta)^2} d\beta \\ &= e^{-(x - \alpha)^2} \int_{-\infty}^{\infty} e^{-(y - \beta)^2} d\beta \end{aligned}$$

令 $u = y - \beta$, 则 $d\beta = -du$

由高斯积分性质, $\int_{-\infty}^{\infty} e^{-u^2} du = \sqrt{\pi}$

$$\begin{aligned} g(x, y) &= e^{-(x - \alpha)^2} \left[\int_{-\infty}^{\infty} e^{-u^2} du \right] \\ &= \sqrt{\pi} e^{-(x - \alpha)^2} \end{aligned}$$

Assume that we have noisy image $g_i(x, y)$ from the image $f(x, y)$, i.e.

$$g_i(x, y) = f(x, y) + \eta_i(x, y)$$

where the noise η_i is zero-mean and at all point-pairs (x, y) are uncorrelated. Then reduce noise by taking the mean of all the noisy images:

$$\bar{g}(x, y) = \frac{1}{M} \sum_{i=1}^M g_i(x, y)$$

Prove that:

$$E\{\bar{g}(x, y)\} = f(x, y)$$

and:

$$\sigma_{\bar{g}(x, y)}^2 = \frac{1}{M} \sigma_{\eta(x, y)}^2$$

where $\sigma_{\eta(x, y)}^2$ is the variance of η and $\sigma_{\bar{g}(x, y)}^2$ is the variance of $\bar{g}(x, y)$.

对于图像 $g_i(x, y) = f(x, y) + \eta_i(x, y)$

平均图像 $\bar{g}(x, y) = \frac{1}{M} \sum_{i=1}^M [f(x, y) + \eta_i(x, y)]$

$$= f(x, y) + \frac{1}{M} \sum_{i=1}^M \eta_i(x, y)$$

\therefore noise η_i 是 zero-mean

$$\therefore E\{\eta_i(x, y)\} = 0, \text{ 即 } E\left\{\frac{1}{M} \sum_{i=1}^M \eta_i(x, y)\right\} = 0$$

$$\therefore E\{\bar{g}(x, y)\} = f(x, y) + 0 = f(x, y)$$

平均图像 $\bar{g}(x, y)$ 的方差 $\sigma_{\bar{g}(x, y)}^2 = E\{(\bar{g}(x, y) - E\{\bar{g}(x, y)\})^2\}$

$$= E\{(\bar{g}(x, y) - f(x, y))^2\}$$

$$= E\left\{\left(\frac{1}{M} \sum_{i=1}^M \eta_i(x, y)\right)^2\right\}$$

$$= \frac{1}{M^2} E\left\{\sum_{i=1}^M \eta_i(x, y)^2 + 2 \sum_{i < j} \eta_i(x, y) \eta_j(x, y)\right\}$$

$\therefore (x, y)$ 不相关

\therefore 交叉项期望为 0

$$\sigma_{\bar{g}(x, y)}^2 = \frac{1}{M^2} \sum_{i=1}^M E\{\eta_i(x, y)^2\}$$

$$= \frac{1}{M^2} M \cdot \sigma_{\eta(x, y)}^2$$

$$= \frac{1}{M} \sigma_{\eta}^2(x, y)$$

Suppose an image's degradation function is the convolution of original image and $h(x, y) = \frac{x^2+y^2-2\sigma^2}{\sigma^4} e^{-(x^2+y^2)/2\sigma^2}$ and here assume x, y are continuous.

1. Prove that the degradation in frequency filed is

$$H(u, v) = -8\pi^4 \sigma^2 (u^2 + v^2) e^{-2\pi^2 \sigma^2 (u^2 + v^2)}$$

2. Assume the power spectrum ratio of noise and undegraded image (S_η/S_f) is the constant parameter K , give the transfer function expression of Wiener filter for this image.

$$(1) H(u, v) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} h(x, y) e^{-j2\pi(ux+vy)} dx dy$$

$$\text{其中 } h(x, y) = \frac{x^2+y^2-2\sigma^2}{\sigma^4} e^{-\frac{x^2+y^2}{2\sigma^2}}$$

$$\text{设二维高斯函数为 } g(x, y) = e^{-\frac{x^2+y^2}{2\sigma^2}}$$

$$\text{则注意到 } \nabla^2 g = \frac{\partial^2 g}{\partial x^2} + \frac{\partial^2 g}{\partial y^2} = \left(\frac{x^2}{\sigma^4} - \frac{1}{\sigma^2}\right) e^{-\frac{x^2+y^2}{2\sigma^2}} + \left(\frac{y^2}{\sigma^4} - \frac{1}{\sigma^2}\right) e^{-\frac{x^2+y^2}{2\sigma^2}} \\ = h(x, y)$$

$$\text{由FT的微分性质, } F\left\{\frac{\partial^2 f}{\partial x^2}\right\} = (-j2\pi u)^2 F(u, v)$$

$$F\left\{\frac{\partial^2 f}{\partial y^2}\right\} = (-j2\pi v)^2 F(u, v)$$

$$\therefore F\{\nabla^2 g(x, y)\} = F\left\{\frac{\partial^2 g}{\partial x^2} + \frac{\partial^2 g}{\partial y^2}\right\} = -4\pi^2(u^2 + v^2) G(u, v)$$

$$\text{由高斯函数性质, } \int_{-\infty}^{\infty} e^{-\frac{x^2}{2\sigma^2}} e^{-j2\pi ux} dx = \sqrt{2\pi}\sigma^2 e^{-2\pi^2\sigma^2 u^2}$$

$$\int_{-\infty}^{\infty} e^{-\frac{y^2}{2\sigma^2}} e^{-j2\pi vy} dy = \sqrt{2\pi}\sigma^2 e^{-2\pi^2\sigma^2 v^2}$$

$$\therefore G(u, v) = 2\pi\sigma^2 e^{-2\pi^2\sigma^2(u^2+v^2)}$$

$$\therefore H(u, v) = -4\pi^2(u^2+v^2) G(u, v)$$

$$= -8\pi^3\sigma^2(u^2+v^2) e^{-2\pi^2\sigma^2(u^2+v^2)}$$

$$(2) \text{ Wiener filter } H_w(u, v) = \frac{H^*(u, v)}{|H(u, v)|^2 + \frac{S_\eta}{S_f}}, \quad \frac{S_\eta}{S_f} = \frac{\text{噪声功率}}{\text{未退化功率}} = K$$

$$\therefore H_w(u, v) = \frac{-8\pi^3\sigma^2(u^2+v^2) e^{-2\pi^2\sigma^2(u^2+v^2)}}{64\pi^6\sigma^4(u^2+v^2)^2 e^{-4\pi^2\sigma^2(u^2+v^2)} + K}$$