

2019-Chapter5 solution

Problem 1:

Given $f(x) = \frac{1}{\sqrt{2\pi\sigma_x^2}} \exp\left(-\frac{(x-\mu_x)^2}{2\sigma_x^2}\right)$ and $\exp(-\pi t^2) \Leftrightarrow \exp(-\pi f^2)$, then by

applying the time-shifting and scaling properties:

$$\begin{aligned} F(f) &= \frac{1}{\sqrt{2\pi\sigma_x^2}} \left| \sqrt{2\pi\sigma_x^2} \right| \exp\left(-\pi \left(\sqrt{2\pi\sigma_x^2}\right)^2 \pi f^2\right) \exp(j2\pi f \mu_x) \\ &= \exp\left(-\pi^2 2\sigma_x^2 f^2 + j\mu_x 2\pi f\right) \text{ and let } v=2\pi f \\ &= \exp\left(jv\mu_x - \frac{1}{2}v^2\sigma_x^2\right) \end{aligned}$$

另解

$$\begin{aligned} \phi_X(v) &= \int_{-\infty}^{\infty} e^{ivx} f(x) dx \\ &= \int_{-\infty}^{\infty} e^{ivx} \frac{1}{\sqrt{2\pi\sigma_x^2}} e^{\left(-\frac{(x-\mu_x)^2}{2\sigma_x^2}\right)} dx \\ &= \frac{1}{\sqrt{2\pi\sigma_x^2}} \int_{-\infty}^{\infty} e^{\frac{-2\sigma_x^2}{-2\sigma_x^2} ivx} e^{\left(\frac{-1}{2\sigma_x^2}\right)(x^2 - 2x\mu_x + \mu_x^2)} dx \\ &= \frac{1}{\sqrt{2\pi\sigma_x^2}} \int_{-\infty}^{\infty} e^{\frac{1}{-2\sigma_x^2}(x^2 - (2\mu_x + 2\sigma_x^2 iv)x) + \frac{\mu_x^2}{-2\sigma_x^2}} dx \\ &= \frac{1}{\sqrt{2\pi\sigma_x^2}} \int_{-\infty}^{\infty} e^{\frac{1}{-2\sigma_x^2}\left(x^2 - 2(\mu_x + \sigma_x^2 iv)x + (\mu_x + \sigma_x^2 iv)^2\right) + \frac{\mu_x^2}{-2\sigma_x^2} - \frac{-(\mu_x + \sigma_x^2 iv)^2}{-2\sigma_x^2}} dx \\ &= \frac{1}{\sqrt{2\pi\sigma_x^2}} \int_{-\infty}^{\infty} e^{\frac{1}{-2\sigma_x^2}\left(x - (\mu_x + \sigma_x^2 iv)\right)^2 + \frac{\mu_x^2 - \mu_x^2 - 2\mu_x\sigma_x^2 iv + \sigma_x^4 v^2}{-2\sigma_x^2}} dx e^{\frac{-2\mu_x\sigma_x^2 iv + \sigma_x^4 v^2}{-2\sigma_x^2}} \\ &= e^{\frac{\mu_x iv - \frac{1}{2}\sigma_x^2 v^2}{-2\sigma_x^2}} \end{aligned}$$

Problem 2:

For a complex random process $Z(t)$, $R_Z(\tau) = E[Z^*(t)Z(t+\tau)]$, then

- i. The mean square of a complex process is given by

$$R_Z(0) = E[Z^*(t)Z(t)] = E[|Z(t)|^2]$$

- ii. We show $R_Z(\tau)$ has conjugate symmetry by the following

$$R_Z(-\tau) = E[Z^*(t)Z(t-\tau)] = E[Z^*(s+\tau)Z(s)] = E[Z^*(s)Z(s+\tau)]^* = R_Z^*(\tau)$$

where we have used the change of variable $s = t - \tau$.

- iii. Taking an approach similar to that of $|R_X(\tau)| \leq R_X(0)$

$$\begin{aligned} 0 &\leq E[|Z(t) \pm Z(t+\tau)|^2] \\ &= E[(Z(t) \pm Z(t+\tau))(Z^*(t) \pm Z^*(t+\tau))] \\ &= E[Z(t)Z^*(t) \pm Z(t)Z^*(t+\tau) \pm Z^*(t)Z(t+\tau) + Z(t+\tau)Z^*(t+\tau)] \\ &= E[|Z(t)|^2] \pm E[Z(t)Z^*(t+\tau)] \pm E[Z^*(t)Z(t+\tau)] + E[|Z(t+\tau)|^2] \\ &= 2E[|Z(t)|^2] \pm 2\operatorname{Re}\{E[Z^*(t)Z(t+\tau)]\} \\ &= 2R_Z(0) \pm 2\operatorname{Re}\{R_Z(\tau)\} \end{aligned}$$

Thus $|\operatorname{Re}\{R_Z(\tau)\}| \leq R_Z(0)$

Problem 3:

- (a) $R_{XY}(\tau) = E[X(t+\tau)Y(t)]$, replacing $t+\tau$ with t' :

$$R_{XY}(\tau) = E[X(t')Y(t'-\tau)] = R_{YX}(-\tau)$$

- (b)

$$\begin{aligned} E[(X(t+\tau) \pm Y(t))^2] &= E[X^2(t+\tau)] \pm 2E[X(t+\tau)Y(t)] + E[Y^2(t)] \\ &= R_X(0) \pm 2R_{XY}(\tau) + R_Y(0) \geq 0 \end{aligned}$$

Thus, $\frac{1}{2}[R_X(0) + R_Y(0)] \geq |R_{XY}(\tau)|$

Problem 4:

(a) The filter output is $Y(t) = \int_{-\infty}^{\infty} h(\tau)X(t-\tau)d\tau = \frac{1}{T} \int_0^T X(t-\tau)d\tau$

Put $t - \tau = u$. Then, the sample value of $Y(t)$ at $t = T$ equals $Y = \frac{1}{T} \int_0^T X(u)du$

The mean of Y is therefore $E[Y] = E\left[\frac{1}{T} \int_0^T X(u)du\right] = \frac{1}{T} \int_0^T E[X(u)]du = 0$

The variance of Y is

$$\begin{aligned}\sigma_Y^2 &= E[Y^2] - \{E[Y]\}^2 = R_Y(0) \\ &= \int_{-\infty}^{\infty} S_Y(f)df = \int_{-\infty}^{\infty} S_X(f)|H(f)|^2 df\end{aligned}$$

But

$$\begin{aligned}H(f) &= \int_{-\infty}^{\infty} h(t)\exp(-j2\pi ft)dt \\ &= \frac{1}{T} \int_0^T \exp(-j2\pi ft) dt \\ &= \frac{1}{T} \frac{\exp(-j2\pi ft)}{-j2\pi f} \Big|_0^T \\ &= \frac{1}{j2\pi fT} [1 - \exp(-j2\pi fT)] = \text{sinc}(fT) \exp(-j\pi fT)\end{aligned}$$

Therefore, $\sigma_Y^2 = \int_{-\infty}^{\infty} S_X(f) \text{sinc}^2(fT) df$

(b) Since the filter input is Gaussian, it follows that Y is also Gaussian. Hence, the probability density function of Y is

$$f_Y(y) = \frac{1}{\sqrt{2\pi\sigma_Y^2}} \exp\left(\frac{-y^2}{2\sigma_Y^2}\right), \text{ where } \sigma_Y^2 \text{ is defined above.}$$