

1.

$$X_t \longrightarrow \boxed{H(f)} \longrightarrow Y_t$$

Given: $H(f) = \frac{j2\pi f + a}{j2\pi f + 2a}$

$$K_Y(\tau) = a^2 e^{-2a|\tau|}$$

$$m_Y = \frac{1}{2}$$

a)

$$R_Y(\tau) = K_Y(\tau) + m_Y^2$$

$$= a^2 e^{-2a|\tau|} + \frac{1}{4}$$

$$S_Y(f) = \int_{-\infty}^{\infty} R_Y(\tau) \cdot e^{-j2\pi f\tau} d\tau$$

$$= \int_{-\infty}^{\infty} \left(\frac{1}{4} + a^2 e^{-2a|\tau|} \right) e^{-j2\pi f\tau} d\tau$$

$$= \frac{1}{4} \delta(f) + \frac{4a^3}{4a^2 + 4\pi^2 f^2}$$

b) $S_Y(f) = S_X(f) \cdot |H(f)|^2$

$$m_Y = H(0) \cdot m_X$$

$$\Rightarrow S_X(f) = \frac{S_Y(f)}{|H(f)|^2}$$

$$\Rightarrow m_X = \frac{m_Y}{H(0)}$$

$$|H(f)|^2 = \frac{4\pi^2 f^2 + a^2}{4\pi^2 f^2 + 4a^2}$$

$$H(0) = \frac{1}{2}$$

$$S_X(f) = \frac{1/4 \delta(f)}{|H(0)|^2} + \frac{4a^3}{\frac{4\pi^2 f^2 + a^2}{4\pi^2 f^2 + 4a^2}}$$

$$m_X = \frac{1/2}{1/2} = 1$$

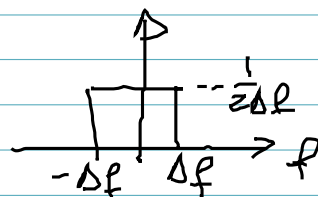
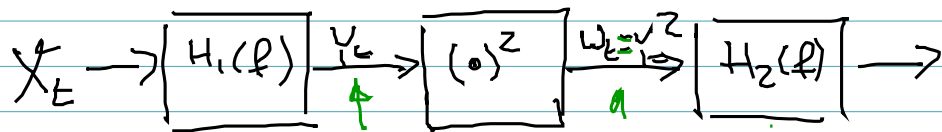
$$= \frac{1/4}{|1/2|^2} \delta(f) + \frac{4a^3}{a^2 + 4\pi^2 f^2}$$

$$R_x(\tau) = 1 + Za^2 \cdot e^{-a|\tau|}$$

$$K_x(\tau) = Za^2 e^{-a|\tau|}$$

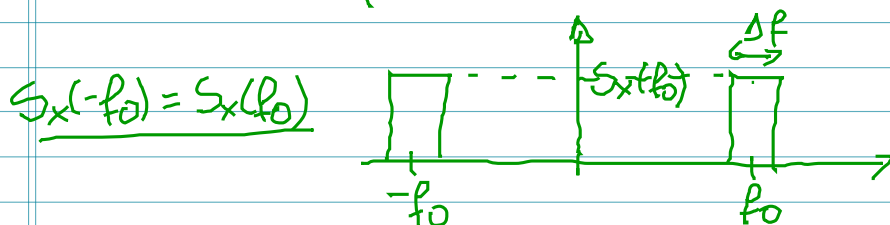
$$m_x = 1$$

2.



$S_X(f)$ is PSD of X_L

$$a) \quad S_Y(f) \approx \begin{cases} S_X(f_0) & f_0 - \frac{\Delta f}{2} \leq |f| \leq f_0 + \frac{\Delta f}{2} \\ 0 & \text{else} \end{cases}$$



$$\begin{aligned} R_Y(\tau) &= \int_{-\infty}^{\infty} S_Y(f) \cdot e^{j2\pi f\tau} df \\ &= \int_{-f_0 - \frac{\Delta f}{2}}^{-f_0 + \frac{\Delta f}{2}} S_X(f_0) \cdot e^{j2\pi f\tau} df + \int_{f_0 - \frac{\Delta f}{2}}^{f_0 + \frac{\Delta f}{2}} S_X(f_0) \cdot e^{j2\pi f\tau} df \\ &= S_X(f_0) \cdot 2 \cdot \int_{f_0 - \frac{\Delta f}{2}}^{f_0 + \frac{\Delta f}{2}} \cos(2\pi f\tau) df \\ &= S_X(f_0) \cdot 2\Delta f \cdot \cos(2\pi f_0\tau) \cdot \text{sinc}(\pi \Delta f \tau) \end{aligned}$$

$$W_t = Y_t^2 \quad \text{second order description}$$

$$E[W_t] = E[Y_t^2] = R_Y(0) = \underline{S_X(f_0) \cdot 2\Delta f}$$

$$R_W(\tau) = E[W_t \cdot W_{t+\tau}] = E[Y_t^2 \cdot Y_{t+\tau}^2]$$

Note that Y_t and $Y_{t+\tau}$ are jointly Gaussian noise

Isserlin's theorem: X, Y jointly Gaussian

$$\text{then } E[X^2 Y^2] = E[X^2] E[Y^2] + 2E^2[XY]$$

$$E[Y_t^2 \cdot Y_{t+\tau}^2] = E[Y_t^2] \cdot E[Y_{t+\tau}^2] + 2E^2[Y_t \cdot Y_{t+\tau}]$$

$$= R_Y(0) \cdot R_Y(0) + 2R_Y^2(\tau)$$

$$= (\underline{S_X(f_0) \cdot 2\Delta f})^2 +$$

$$2 \cdot (S_X(f_0) \cdot 2\Delta f \cdot \cos(2\pi f_0 \tau) \cdot \text{sinc}(\pi \Delta f \tau))^2$$

$$\cos^2 x =$$

$$\frac{1}{2} + \frac{1}{2} \cos 2x$$

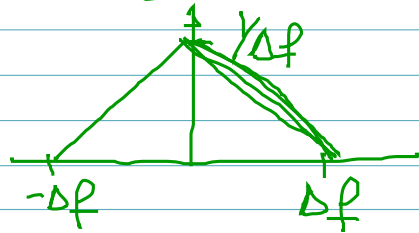
$$= \left((S_X(f_0) \cdot 2\Delta f)^2 \cdot (1 + \text{sinc}^2(\pi \Delta f \tau)) \right) +$$

$$(S_X(f_0) \cdot 2\Delta f)^2 \cdot (\cos(4\pi f_0 \tau) \cdot \text{sinc}^2(\pi \Delta f \tau))$$

rejected by LPF

$$\begin{aligned}
 b) \quad R_z(\tau) &= \left(\frac{1}{2\Delta f}\right)^2 \cdot (S_x(f_0) \cdot 2\Delta f)^2 \cdot (1 + \text{sinc}^2(\pi \Delta f \tau)) \\
 &= \underline{S_x^2(f_0)} \cdot \underline{(1 + \text{sinc}^2(\pi \Delta f \tau))}
 \end{aligned}$$

$$S_z(f) = S_x^2(f_0) \cdot \left(\delta(f) + \underbrace{\frac{1}{\Delta f} \cdot \left(1 - \frac{|f|}{\Delta f}\right)}_{\text{for } |f| < \Delta f} \right)$$



$$\begin{aligned}
 c) \quad E[z_t] &= H(0) \cdot E[Y_t^2] \\
 &= H(0) \cdot E[w_t] \\
 &= \frac{1}{2\Delta f} \cdot S_x(f_0) \cdot 2\Delta f = \underline{S_x(f_0)}
 \end{aligned}$$

Good! Unbiased estimate of $S_x(f_0)$

$$\begin{aligned}
 d) \quad \text{Var}[z_t] &= R_z(0) - m_z^2 \\
 &= S_x^2(f_0) \cdot (1 + 1) - S_x^2(f_0) \\
 &= S_x^2(f_0)
 \end{aligned}$$

Bad! Standard deviation equals $S_x(f_0)$. As large as $S_x(f_0)$.

3.

$$X_t = \int_0^t w_s ds$$

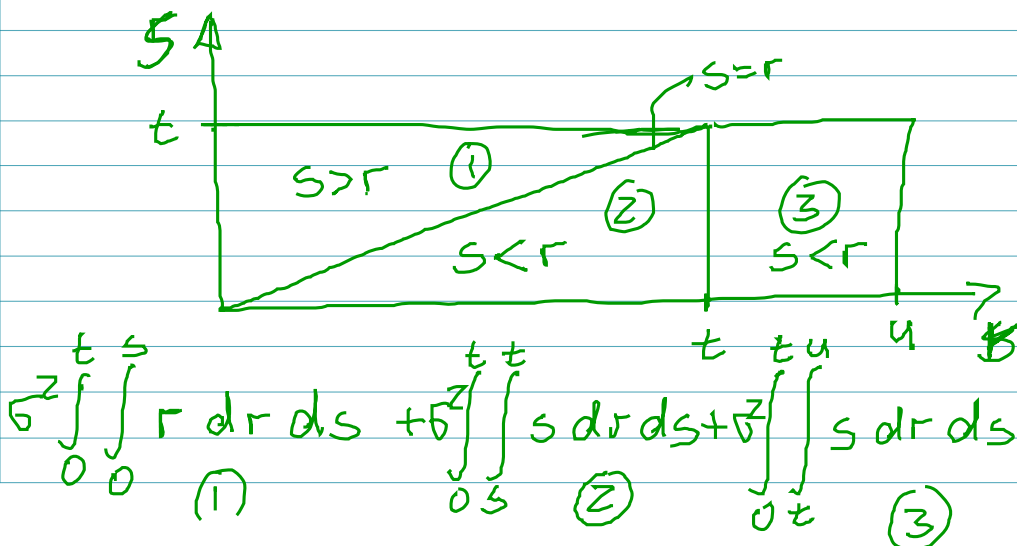
w_t - Wiener Process

Recall: $E[w_t] = 0$ $R_w(t, u) = \sigma^2 \min(t, u)$

$$\begin{aligned} a) E[X_t] &= E\left[\int_0^t w_s ds\right] \\ &= \int_0^t E[w_s] ds = \underline{\underline{0}} \end{aligned}$$

$$\begin{aligned} b) R_X(t, u) &= E[X_t \cdot X_u] \\ &= E\left[\int_0^t w_s ds \cdot \int_0^u w_r dr\right] \\ &= \int_0^t \int_0^u E[w_s w_r] ds dr \\ &= \int_0^t \int_0^u \sigma^2 \min(s, r) ds dr \end{aligned}$$

Assume first that $u \geq t$:



$$\begin{aligned} \textcircled{1} \int_0^t \int_0^s r \, dr \, ds &= \int_0^t \left. \frac{r^2}{2} \right|_0^s ds = \int_0^t \frac{s^2}{2} ds \\ &= \left. \frac{s^3}{6} \right|_0^t = \frac{t^3}{6} \end{aligned}$$

$$\begin{aligned} \textcircled{2} \int_0^t \int_s^t s \, dr \, ds &= \int_0^t r s \Big|_s^t ds = \int_0^t (ts - s^2) ds \\ &= \left. \frac{ts^2}{2} - \frac{s^3}{3} \right|_0^t = \frac{t^3}{6} \end{aligned}$$

$$\begin{aligned} \textcircled{3} \int_0^t \int_t^u s \, dr \, ds &= \int_0^t r \cdot s \Big|_t^u ds = \int_0^t (u-t) \cdot s \, ds \\ &= (u-t) \cdot \left. \frac{s^2}{2} \right|_0^t = \frac{t^2}{2} (u-t) \end{aligned}$$

For $u \geq t$:

$$R_x(t, u) = \sigma^2 \cdot \frac{t^2 u}{2} - \frac{\sigma^2 \cdot t^3}{6}$$

Analogously, for $t \geq u$:

$$R_x(t, u) = \sigma^2 \cdot \frac{u^2 t}{2} - \frac{\sigma^2 u^3}{6}$$

Combined:

$$R_x(t, u) = \sigma^2 \cdot u \cdot t \cdot \min(t, u) - \frac{\sigma^2}{6} (\min(t, u))^3$$

c) X_t is not wss

$$R_x(t, u) \neq R_x(t-u)$$

d) $\Pr\{|X_t| > 5 \cdot t\}$

pdf of X_t ?

- X_t is Gaussian

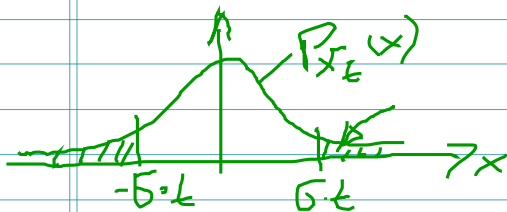
- linear operation on
White Gaussian noise

- Needs: $E[X_t] = 0$

$$\text{Var}[X_t] = E[X_t \cdot X_t]$$

$$= R_x(t, t)$$

$$= 5^2 \cdot \frac{t^3}{3}$$



$$\Pr\{|X_t| > 5 \cdot t\} = 2 \cdot \Pr\{X_t > 5t\}$$

$$= 2 \cdot \int_{5t}^{\infty} \frac{1}{\sqrt{2\pi} \cdot \frac{5^2 t^3}{3}} \cdot \exp\left(-\frac{1}{2} \frac{x^2}{\frac{5^2 t^3}{3}}\right) dx$$

$$z = \frac{x-0}{\sqrt{5^2 t^3/3}}$$

$$= 2 \cdot \int_{\frac{5t}{\sqrt{t/3}}}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{z^2}{2}} dz$$

Lower: $\frac{5 \cdot t}{\sqrt{5^2 t^3/3}}$

$$= 2 \cdot Q(\sqrt{3/t})$$