

## Solution to the exam in TMS115 Probability and Stochastic Processes 2004-10-22

**Problem 1.**  $Y_{2n} \sim \text{Binomial}(2n, p)$

$$\begin{aligned} P\{Y_{2n} \leq n\} & P\left\{ \frac{Y_{2n} - 2np}{\sqrt{2np(1-p)}} \leq \frac{n - 2np}{\sqrt{2np(1-p)}} \right\} \\ & \sim \Phi\left(\frac{n(1-2p)}{\sqrt{2np(1-p)}}\right) = \Phi\left(\frac{\sqrt{n}(1-2p)}{\sqrt{2p(1-p)}}\right). \end{aligned}$$

Then

$$\lim_{n \rightarrow \infty} P\{Y_{2n} \leq n\} = \begin{cases} \frac{1}{2}, & p = \frac{1}{2}, \\ 1, & p < \frac{1}{2}, \\ 0, & p > \frac{1}{2}. \end{cases}$$

**Problem 2.** The power of the exponent in the joint pdf can be written as

$$\frac{1}{2(1-\rho^2)} \left[ \frac{z_1^2}{\frac{1}{2(1-\rho^2)}} - \frac{2}{\sqrt{2}} \frac{z_1}{\sqrt{\frac{1}{2(1-\rho^2)}}} \cdot \frac{z_2}{\sqrt{\frac{1}{2(1-\rho^2)}}} + \frac{z_2^2}{\frac{1}{2(1-\rho^2)}} \right].$$

Thus

$$\rho(Z_1, Z_2) = \rho = \frac{1}{\sqrt{2}}, \quad E[Z_i] = 0, \quad \text{Var}(Z_i) = \sigma_i^2 = \frac{1}{2(1-\rho^2)} = 1, \quad i = 1, 2.$$

$$\begin{aligned} \text{a) } \text{Cov}(Z_1 - Z_2/\sqrt{2}, Z_2) &= \text{Cov}(Z_1, Z_2) - \frac{1}{\sqrt{2}} \text{Cov}(Z_2, Z_2) \\ &= \rho \sigma_1 \sigma_2 - \frac{1}{\sqrt{2}} \sigma_2^2 = 0. \end{aligned}$$

b) The random variables  $Z_1 - Z_2/\sqrt{2}$  and  $Z_2$  are jointly Gaussian and uncorrelated. Hence they are independent and we thus have

$$\begin{aligned} E[Z_1^2 Z_2] &= E\left[(Z_1 - Z_2/\sqrt{2} + Z_2/\sqrt{2})^2 Z_2\right] \\ &= E\left[(Z_1 - Z_2/\sqrt{2})^2 Z_2\right] - 2E\left[(Z_1 - Z_2/\sqrt{2}) Z_2^2/\sqrt{2}\right] + E\left[Z_2^3/2\right] \\ &= E\left[(Z_1 - Z_2/\sqrt{2})^2\right] E[Z_2] - 2E\left[Z_1 - Z_2/\sqrt{2}\right] E\left[Z_2^2/\sqrt{2}\right] + E\left[Z_2^3/2\right] = 0, \end{aligned}$$

since

$$E[Z_2] = E\left[Z_1 - Z_2/\sqrt{2}\right] = E\left[Z_2^3/2\right] = 0.$$

**Problem 3.**  $X \sim \text{Binomial}(M, p)$

a)  $Y =$  the number of discarded voices

$$Y = \begin{cases} 0, & \text{if } X \leq N, \\ X - N & \text{if } X > N. \end{cases}$$

$$E[Y] = \sum_{j=1}^{M-N} j P\{X = N + j\} = \sum_{j=1}^{M-N} j \binom{M}{N+j} p^{N+j} (1-p)^{M-N-j}.$$

$$\begin{aligned} \text{b) } P\{Y = 0\} &= P\{X \leq N\} \approx \Phi\left(\frac{N - M \cdot p}{\sqrt{M p (1-p)}}\right) \\ &= \Phi\left(\frac{16 - 45 \cdot 1/3}{\sqrt{45 \cdot 1/3 \cdot 2/3}}\right) = 1 - Q(0.31) = 0.62 \end{aligned}$$

**Problem 4.** Let  $N_1(t)$  and  $N_2(t)$  be the two Poisson processes on line 1 and line 2, and  $T_1$  and  $T_2$  be the waiting times for the first event in  $N_1(t)$  and  $N_2(t)$ , respectively.

a) The random variables  $T_1$  and  $T_2$  are independent and exponentially distributed of rates  $\lambda$  and  $\mu$ , correspondingly. We have

$$\begin{aligned} P\{T_1 < T_2\} &= \int_0^\infty P\{T_1 < T_2 | T_2 = x\} f_{T_2}(x) dx \\ &= \int_0^\infty P\{T_1 < x | T_2 = x\} f_{T_2}(x) dx = \int_0^\infty P\{T_1 < x\} f_{T_2}(x) dx \\ &= \int_0^\infty (1 - e^{-\lambda x}) \mu e^{-\mu x} dx = \frac{\lambda}{\lambda + \mu}. \end{aligned}$$

b) The waiting time for the first message to come is  $T = \min(T_1, T_2)$ .

$$\begin{aligned} P\{\min(T_1, T_2) \leq t\} &= 1 - P\{\min(T_1, T_2) > t\} \\ &= 1 - P\{T_1 > t, T_2 > t\} = 1 - P\{T_1 > t\} P\{T_2 > t\} = 1 - e^{-(\lambda + \mu)t}. \end{aligned}$$

Then  $T$  is an exponential random variable with

$$f_T(t) = (\lambda + \mu) e^{-(\lambda + \mu)t}.$$

- c) Let  $N(t) = N_1(t) + N_2(t)$ . Since  $N_1(t)$  and  $N_2(t)$  are independent,  $N(t)$  is a Poisson process of rate  $2\lambda$ . We have

$$\begin{aligned} & P\{N(t/2) \geq 1, N(t) - N(t/2) \geq 1 | N(t) = 3\} \\ &= \frac{P\{N(t/2) \geq 1, N(t) - N(t/2) \geq 1, N(t) = 3\}}{P\{N(t) = 3\}} \\ &= \frac{2 \cdot P\{N(t/2) = 1\} P\{N(t/2) = 2\}}{P\{N(t) = 3\}} = \frac{2 \cdot \frac{\lambda t}{2} e^{-\frac{\lambda t}{2}} \cdot \frac{(\lambda t)^2}{8} e^{-\frac{\lambda t}{2}}}{\frac{(\lambda t)^3}{6} e^{-\lambda t}} = \frac{3}{4}. \end{aligned}$$

**Problem 5.**

$$\begin{aligned} m_Z(t) &= 0, \\ E[Z(t + \tau)Z(t)] &= E[(bX(t + \tau) - Y(t + \tau - b))(bX(t) - Y(t - b))] \\ &= b^2 R_X(\tau) - bR_{X,Y}(\tau + b) - bR_{Y,X}(\tau - b) + R_Y(\tau). \end{aligned}$$

Hence  $Z(t)$  is WSS.

**Problem 6.**

a)  $R_{Y,X}(n + k, n) = E[(X_{n+k} + \beta X_{n+k-1})X_n] = R_X(k) + \beta R_X(k - 1)$

$$S_{Y,X}(f) = (1 + \beta e^{-j2\pi f})S_X(f)$$

$$S_X(f) = \sigma^2 \sum_{k=-\infty}^{\infty} \alpha^{|k|} e^{-j2\pi f k} = \frac{\sigma^2(1 - \alpha^2)}{1 + \alpha^2 - 2\alpha \cos 2\pi f}$$

- b) The unit impuls response is

$$h_n = \begin{cases} 1, & n = 0, \\ \beta, & n = 1, \\ 0, & \text{otherwise,} \end{cases}$$

and the transfer function is  $H(f) = 1 + \beta e^{-j2\pi f}$ .

Thus

$$\begin{aligned} S_Y(f) &= |1 + \beta e^{-j2\pi f}|^2 S_X(f) = \sigma^2(1 - \alpha^2) \cdot \frac{1 + \beta^2 + 2\beta \cos 2\pi f}{1 + \alpha^2 - 2\alpha \cos 2\pi f} \\ &= \sigma^2(1 - \alpha^2) \cdot \frac{\beta}{\alpha} \cdot \frac{\frac{1+\beta^2}{2\beta} + \cos 2\pi f}{\frac{1+\alpha^2}{2\alpha} - \cos 2\pi f} \\ &= \sigma^2(1 - \alpha^2) \cdot \frac{\beta}{\alpha} \cdot \left[ -1 + \frac{\frac{1+\beta^2}{2\beta} + \frac{1+\alpha^2}{2\alpha}}{\frac{1+\alpha^2}{2\alpha} - \cos 2\pi f} \right]. \end{aligned}$$

$S_Y(f)$  is constant in  $f$  if and only if

$$\begin{aligned} \frac{1 + \beta^2}{2\beta} &= -\frac{1 + \alpha^2}{2\alpha} \\ &\Downarrow \\ \alpha(1 + \beta^2) + \beta(1 + \alpha^2) &= 0 \\ &\Downarrow \\ (\alpha + \beta)(1 + \alpha\beta) &= 0 \end{aligned}$$

$Y_n$  is a white-noise process if and only if either  $\beta = -\alpha$  or  $\beta = -1/\alpha$ .

**Problem 7.**

- a) We have  $P(D)Y = W$ , where  $P(D) = D^0 - \frac{1}{2}D$  is the polynomial shift operator. Its characteristic polynomial is  $P(z) = 1 - \frac{1}{2}z$ ,  $|z| > 1$ , and

$$P(z)^{-1} = \frac{1}{1 - \frac{1}{2}z} = \sum_{k=0}^{\infty} 2^{-k} z^k.$$

Thus

$$P(D)^{-1} = \sum_{k=0}^{\infty} 2^{-k} D^k \quad \text{and} \quad Y_n = \sum_{k=0}^{\infty} 2^{-k} W_{n-k}.$$

The unit-impulse response of the system producing  $Y_n$  is

$$h_n = \begin{cases} 2^{-n}, & n \geq 0, \\ 0, & n < 0. \end{cases}$$

For  $k \geq 0$ , we have

$$\begin{aligned} R_Y(k) &= \sum_{r=0}^{\infty} \sum_{s=0}^{\infty} 2^{-r} \cdot 2^{-s} R_W(k + r - s) \\ &= \sum_{r=0}^{\infty} 2^{-r} \cdot 2^{-k-r} \sigma_W^2 = \sigma_W^2 2^{-k} \sum_{r=0}^{\infty} 2^{-2r} = \frac{4}{3} \sigma_W^2 2^{-k}. \end{aligned}$$

Then

$$R_Y(k) = \frac{4}{3} \sigma_W^2 \left(\frac{1}{2}\right)^{|k|}, \quad k = 0, \pm 1, \dots$$

- b)  $\hat{Y}_n = h_2 Y_{n-2} + h_3 Y_{n-3}$  - the estimation. The orthogonality condition

$$Y_n - \hat{Y}_n \perp Y_{n-2} \quad \text{and} \quad Y_n - \hat{Y}_n \perp Y_{n-3}$$

implies

$$h_2 R_X(0) + h_3 R_X(1) = R_X(2)$$

$$h_2 R_X(1) + h_3 R_X(0) = R_X(3)$$

or

$$4h_2 + 2h_3 = 1$$

$$2h_2 + 4h_3 = 1/2.$$

Thus  $h_2 = \frac{1}{4}$ ,  $h_3 = 0$ , and

$$E[e_n^2] = E[(Y_n - \hat{Y}_n)e_n] = E[Y_n(Y_n - \hat{Y}_n)] = R_X(0) - h_2 R_X(2) = 4 - \frac{1}{4} = 3.75$$