## Final Exam - ECE 503 Fall 2017

• Date: Wednesday, December 13, 2017.

• Time: 10:30 am -12:30 pm

• Maximum Credit: 100 points

- 1. [10 points] A jet aircraft's autopilot has conditional probability 1/3 of failure given that it employs a faulty microprocessor chip. The autopilot has conditional probability of failure of 1/10 given that it employs a non-faulty chip. According to the chip manufacturer, the probability of a customer receiving a faulty chip is 1/4.
  - (a) Find the probability of failure of autopilot.
  - (b) Given that an autopilot failure has occurred, find the conditional probability that a faulty chip was used.

Let 
$$A_F = \{autopilot \ fails \}$$
;  $C_F = \{fautty \ chip \ was used \}$ .  
We are given:  $P(A_F|C_F) = \frac{1}{3}$ ;  $P(C_F) = \frac{1}{4}$ ;  $P(A_F|C_F) = \frac{1}{10}$   
(a)  $P(A_F) = P(C_F) \cdot P(A_F|C_F) + P(C_F) \cdot P(A_F|C_F)$   
 $= \frac{1}{4} \times \frac{1}{3} + \frac{3}{4} \times \frac{1}{10} = \frac{19}{120}$   
(b)  $P(C_F|A_F) = \frac{P(A_F,C_F)}{P(A_F)} = \frac{P(C_F) \cdot P(A_F|C_F)}{P(A_F)}$   
 $= \frac{(1/4) \times (\sqrt{3})}{19/120} = \frac{120}{12 \times 19} = \frac{10}{19}$ 

## 2. [20 points]

- (a) Let X and Y be independent exponential random variables (both distributed as  $\exp(\lambda)$ ). Find  $E[\max(X,Y)]$ .
- (b) Let f(x) denote a probability density function. We define a sequence of random variables  $X_n$  which have the following density (PDF):

$$f_n(x) = n f(nx)$$

Find the CDF of  $X_n$  and its limiting behavior as  $n \to \infty$ . Does this sequence of random variables converge in distribution? If yes, then which random variable does this sequence converge to?

(a) Let 
$$Z = max(x, Y)$$
 CDF of  $Z : f_{z}(3) = P(Z \le 3)$   
 $= P(max(x, Y) \le 3) = P(x \le 3, Y \le 3)$   
 $= P(x \le 3) \cdot P(Y \le 3).$   
 $= F_{x}(3) \cdot F_{y}(3)$   
 $\Rightarrow f_{z}(3) = f_{x}(3) F_{y}(3) + f_{y}(3) F_{x}(3) = 2 f_{x}(3) F_{y}(3).$   
 $= 2 \lambda e^{\lambda 3} (1 - e^{\lambda 3})$   
 $= 2 \lambda e^{\lambda 3} (1 - e^{\lambda 3})$   
 $= 2 \lambda e^{\lambda 3} (1 - e^{\lambda 3})$ 

(b). 
$$F_n(x) = P(x_n \le x) = \int_{-\infty}^{x} f(nt)dt = \int_{-\infty}^{nx} f(y)dy = F(nx).$$

$$\Rightarrow F_n(x) = F(nx).$$

$$f(x) = F(nx).$$

(for 
$$x = 0$$
,  $F_n(o) = F(o)$ ).

 $\Rightarrow \times_n \longrightarrow o$  in

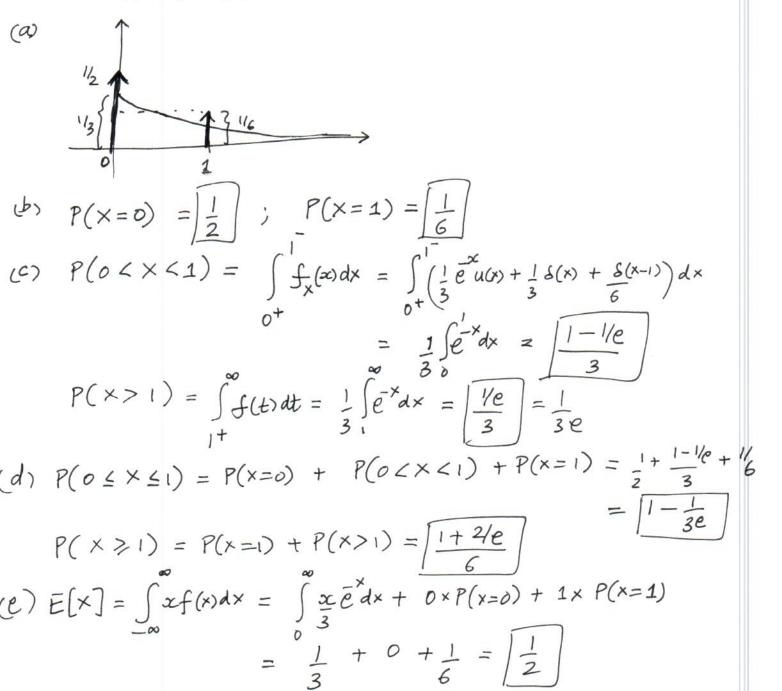
distribution as  $n \to \infty$  2

3. [15 points] A mixed random variable X has the following density:

$$f_X(x) = \frac{1}{3}e^{-x}u(x) + \frac{1}{2}\delta(x) + \frac{1}{6}\delta(x-1)$$

where u(x) is the unit step function, and  $\delta(x)$  is the Dirac delta function.

- (a) Sketch the density function  $f_X(x)$ .
- (b) Compute P(X = 0) and P(X = 1).
- (c) Compute P(0 < X < 1) and P(X > 1).
- (d) Compute  $P(0 \le X \le 1)$  and  $P(X \ge 1)$ .
- (e) Compute E[X].



4. [20 points] Consider the random process

$$W(t) = X\cos(2\pi t) + Y\sin(2\pi t)$$

where X and Y are uncorrelated Gaussian random variables. Both X and Y have zero mean, and unit variance.

- (a) Find the auto correlation function of the random process W(t).
- (b) Is W(t) a wide sense stationary random process?
- (c) What is the average power of this random process?
- (d) Find the probability  $P(W(1) + W(\frac{3}{4}) \le 0)$ .

$$(\omega) \quad E[N(t)N(t+v)] = E[(x (66(2\pi t) + y \sin(2\pi t))(x (66(2\pi (t+v) + y \sin(2\pi (t+v))))] \\ = E[x^{2}] \cdot (66(2\pi t) (66(2\pi (t+v)) + E[x^{2}] \sin(2\pi t) \sin(2\pi (t+v)))] \\ + E[x^{2}] \cdot (66(2\pi t) + (66(2\pi (2t+v)))] \\ = E[x^{2}] \cdot (66(2\pi t) + (66(2\pi (2t+v)))] \\ = E[x^{2}] \cdot (66(2\pi t) + (66(2\pi (2t+v)))] \\ = \frac{1}{2} \times (66(2\pi t) \times 2) = [(66(2\pi t) \times 2) + E[x^{2}] = 1] \\ = \frac{1}{2} \times (66(2\pi t) \times 2) = [(66(2\pi t) \times 2) + E[x^{2}] = 1] \\ = \frac{1}{2} \times (66(2\pi t) \times 2) = [(66(2\pi t) \times 2) + E[x^{2}] = 1] \\ = \frac{1}{2} \times (66(2\pi t) \times 2) = [(66(2\pi t) \times 2) + E[x^{2}] = 1] \\ = \frac{1}{2} \times (66(2\pi t) \times 2) = [(66(2\pi t) \times 2) + E[x^{2}] = 1] \\ = \frac{1}{2} \times (66(2\pi t) \times 2) = [(66(2\pi t) \times 2) + E[x^{2}] = 1] \\ = \frac{1}{2} \times (66(2\pi t) \times 2) = [(66(2\pi t) \times 2) + E[x^{2}] = 1] \\ = \frac{1}{2} \times (66(2\pi t) \times 2) = [(66(2\pi t) \times 2) + E[x^{2}] \times (66(2\pi t) \times 2) = 1] \\ = \frac{1}{2} \times (66(2\pi t) \times 2) = [(66(2\pi t) \times 2) + E[x^{2}] \times (66(2\pi t) \times 2) = 1] \\ = \frac{1}{2} \times (66(2\pi t) \times 2) = [(66(2\pi t) \times 2) + E[x^{2}] \times (66(2\pi t) \times 2) = 1] \\ = \frac{1}{2} \times (66(2\pi t) \times 2) = [(66(2\pi t) \times 2) + E[x^{2}] \times (66(2\pi t) \times 2) = 1] \\ = \frac{1}{2} \times (66(2\pi t) \times 2) = [(66(2\pi t) \times 2) + E[x^{2}] \times (66(2\pi t) \times 2) = 1] \\ = \frac{1}{2} \times (66(2\pi t) \times 2) = [(66(2\pi t) \times 2) + E[x^{2}] \times (66(2\pi t) \times 2) = 1] \\ = \frac{1}{2} \times (66(2\pi t) \times 2) = [(66(2\pi t) \times 2) + E[x^{2}] \times (66(2\pi t) \times 2) = 1] \\ = \frac{1}{2} \times (66(2\pi t) \times 2) = [(66(2\pi t) \times 2) + E[x^{2}] \times (66(2\pi t) \times 2) = 1] \\ = \frac{1}{2} \times (66(2\pi t) \times (66(2\pi t) \times 2) = 1] \\ = \frac{1}{2} \times (66(2\pi t) \times (66(2\pi t) \times 2) = 1] \\ = \frac{1}{2} \times (66(2\pi t) \times (66(2\pi t) \times 2) = 1] \\ = \frac{1}{2} \times (66(2\pi t) \times (66(2\pi t) \times 2) = 1] \\ = \frac{1}{2} \times (66(2\pi t) \times (66(2\pi t) \times 2) = 1] \\ = \frac{1}{2} \times (66(2\pi t) \times (66(2\pi t) \times 2) = 1] \\ = \frac{1}{2} \times (66(2\pi t) \times (66(2\pi t) \times 2) = 1] \\ = \frac{1}{2} \times (66(2\pi t) \times (66(2\pi t) \times 2) = 1] \\ = \frac{1}{2} \times (66(2\pi t) \times (66(2\pi t) \times 2) = 1] \\ = \frac{1}{2} \times (66(2\pi t) \times (66(2\pi t) \times 2) = 1] \\ = \frac{1}{2} \times (66(2\pi t) \times (66(2\pi t) \times 2) = 1] \\ = \frac{1}{2} \times (66(2\pi t) \times (66(2\pi t) \times 2) = 1] \\ = \frac{1}{2} \times (66(2\pi t) \times (66(2\pi t) \times 2) = 1] \\ = \frac{1}{2} \times (66(2\pi t) \times (66(2\pi t) \times 2) = 1] \\ = \frac{1}{2} \times (66(2\pi t) \times (66(2$$

5. [20 points] Suppose  $X_n$  is a random sequence satisfying

$$X_n = cX_{n-1} + Z_{n-1}$$

where  $Z_1, Z_2, ...$  is an i.i.d. sequence with  $E[Z_n] = 0$  and  $Var[Z_n] = \sigma^2$ , and c is a constant with |c| < 1. We also know that  $E[X_0] = 0$ , and  $Var[X_n] = \sigma^2/(1 - c^2)$ . We make noisy observations as follows:

$$Y_{n-1} = dX_{n-1} + W_{n-1}$$

where  $W_1, W_2, \ldots$  is a i.i.d. sequence of measurement noises, with  $E[W_n] = 0$  and  $Var[W_n] = \eta^2$ . We also know that  $W_n$ 's are independent of  $X_n$  and  $Z_n$ .

- (a) Find the optimal linear MMSE estimator  $\hat{X}_n$  of  $X_n$  using the noisy observation  $Y_{n-1}$ .
- (b) Find the resulting mean square estimation error  $E[(X_n \hat{X}_n)^2]$ .

(a) 
$$X_{n} = X_{n}(X_{n-1}) = \begin{cases} X_{n}X_{n-1}(X_{n}) & X_{n}X_{n-1}(X_{n}) \\ Y_{0n}(X_{n}) & Y_{0n}(X_{n-1}) \end{cases} + E[X_{n}]$$

optimum

Vine at MMSE-Estimater.

$$E[Y_{n-1}] = 0 \Rightarrow \hat{X}_{n} = \begin{cases} X_{n}X_{n-1}(X_{n}(X_{n})) & X_{n-1}(X_{n}(X_{n})) \\ Y_{0n}(X_{n-1}) & Y_{0n}(X_{n-1}) + Y_{0n}(W_{n-1}) \\ Y_{0n}(X_{n-1}) & Y_{0n}(X_{n-1}) + Y_{0n}(W_{n-1}) \\ Y_{0n}(X_{n-1}) & Y_{0n}(X_{n}) & Y_{0n}(X_{n-1}) + Y_{0n}(W_{n-1}) \\ Y_{0n}(X_{n}) & Y_{0n}(X_{n}) & Y_{0n}(X_{n}) & Y_{0n}(X_{n-1}) \\ Y_{0n}(X_{n}) & Y_{0n}(X_{n}) & Y_{0n}(X_{n}) & Y_{0n}(X_{n}) & Y_{0n}(X_{n}) \\ Y_{0n}(X_{n}) & Y_{0n}(X_{n}) & Y_{0n}(X_{n}) & Y_{0n}(X_{n}) \\ Y_{0n}(X_{n}) & Y_{0n}(X_{n}) & Y_{0n}(X_{n}) & Y_{0n}(X_{n}) \\ Y_{0n}(X_{n}) & Y_{0n}(X_{n}) & Y_{0n}(X_{n}) & Y_{0n}(X_{n}) & Y_{0n}(X_{n}) \\ Y_{0n}(X_{n}) & Y_{0n}(X_{n}) & Y_{0n}(X_{n}) & Y_{0n}(X_{n}) \\ Y_{0n}(X_{n}) & Y_{0n}(X_{n}) & Y_{0n}(X_{n}) & Y_{0n}(X_{n}) & Y_{0n}(X_{n}) \\ Y_{0n}(X_{n}) & Y_{0n}(X_{n}) & Y_{0n}(X_{n}) & Y_{0n}(X_{n}) & Y_{0n}(X_{n}) & Y_{0n}(X_{n}) \\ Y_{0n}(X_{n}) & Y_{0n}(X_{n}) & Y_{0n}(X_{n}) & Y_{0n}(X_{n}) & Y_{0n}(X_{n}) \\ Y_{0n}(X_{n}) & Y_{0n}(X_{n}) & Y_{0n}(X_{n}) & Y_{0n}(X_{n}) & Y_{0n}(X_{n}) & Y_{0n}(X_{n}) \\ Y_{0n}(X_{n}) & Y_{0n}($$

6. [15 points] Consider the following Markov chain with three states  $S = \{A, B, C\}$ , which has the following one-step transition matrix:

$$P = \left[ \begin{array}{ccc} 1/2 & 1/2 & 0 \\ 0 & 1/2 & 1/2 \\ 1/2 & 0 & 1/2 \end{array} \right]$$

- (a) Draw the state transition diagram for this Markov chain.
- (b) If you are given that  $P(X_1 = A) = 1/2$ ,  $P(X_1 = C) = 1/6$ , find the probability  $P(X_1 = B, X_2 = C, X_3 = A)$ .
- (c) Is this MC irreducible and aperiodic?
- (d) Find the limiting distribution of this Markov chain. Is this distribution unique? Does it depend on the initial distribution?

(a)  $\frac{1}{2}$   $\frac{1}{2}$   $\frac{1}{2}$   $\frac{1}{2}$   $\frac{1}{2}$   $\frac{1}{2}$   $\frac{1}{2}$ 

- (C) Yes, MC is irreducible & aperiodic
- (d) Unique limiting distrib  $\Rightarrow$  T = TPSolving gives  $T = [1/3] \frac{1}{3} \frac{1}{3}$