NATIONAL UNIVERSITY OF SINGAPORE Department of Electrical & Computer Engineering

EXAMINATION FOR

(Semester II, 2013/14)

EE2012 ANALYTICAL TECHNIQUES FOR ECE

April/May 2014 Time Allowed: 2.5 hours

INSTRUCTIONS FOR CANDIDATES:

- This is a CLOSED BOOK exam.
- This paper contains five (5) questions and one formula sheet, printed on four (4) pages.
- Answer ALL questions.
- A non-programmable calculator may be used.

Examiner: Professor Lim Teng Joon

- Q1. Answer the following short questions.
 - (a) Let T be exponentially distributed with E[T] = 2. Find P[T > 2]. (2 marks) Ans: Given that $\lambda = 0.5$, we have

$$P[T > 2] = \int_{2}^{\infty} 0.5e^{-0.5t} dt = e^{-1}.$$

(b) Consider a probability space (Ω, \mathcal{F}, P) , and two independent events A and B in \mathcal{F} . What is $P[A \cup B]$ in terms of P[A] and P[B]? (2 marks) Ans: Since in general $P[A \cup B] = P[A] + P[B] - P[A \cap B]$ and $P[A \cap B] = P[A]P[B]$ because A and B are independent, we have

$$P[A \cup B] = P[A] + P[B] - P[A]P[B].$$

(c) If the CDF of X is

$$F_X(x) = 0.1u(x) + 0.3u(x-2) + 0.6u(x-5),$$

where u(x) is the unit step function, write down the range of X, and the PDF of X. (2 marks) Ans: $S_X = \{0, 2, 5\}$, and $f_X(x) = 0.1\delta(x) + 0.3\delta(x-2) + 0.6\delta(x-5)$.

- (d) Two zero-mean unit-variance random variables X and Y have a correlation coefficient of 0.4. Find E[XY]. (2 marks) Ans: From the formula for $\rho_{X,Y}$, and E[X] = E[Y] = 0, $\sigma_X = \sigma_Y = 1$, we have E[XY] = 0.4.
- (e) If X is Gaussian with E[X] = 1 and var(X) = 4, find P[X > 2] in terms of the Q function. (2 marks)

Ans: It should be clear that

$$P[X > 2] = P\left[Z > \frac{2-1}{2}\right]$$
 (1)
= $Q(0.5)$.

(f) The joint PMF of (X,Y) is given as follows:

$$p_{X,Y}(0,0) = 0.1$$
 $p_{X,Y}(2,0) = 0.1$
 $p_{X,Y}(1,1) = 0.2$ $p_{X,Y}(0,1) = 0.2$
 $p_{X,Y}(2,1) = 0.25$ $p_{X,Y}(1,2) = 0.15$

Find the marginal PMF of X and of Y. (4 marks)

Ans: Using the marginalization rule, we have

$$p_X(k) = \begin{cases} 0.3 & k = 0\\ 0.35 & k = 1\\ 0.35 & k = 2\\ 0 & \text{otherwise} \end{cases}$$
 (3)

$$p_Y(j) = \begin{cases} 0.2 & j = 0\\ 0.65 & j = 1\\ 0.15 & j = 2\\ 0 & \text{otherwise} \end{cases}$$
 (4)

- (g) The random variable X is uniform in $\{1, 2, 3\}$, and Y is Bernoulli with p = 0.2. If X and Y are independent, find E[XY]. (3 marks) Ans: Since X and Y are independent, E[XY] = E[X]E[Y]. We know that E[X] = 2, and E[Y] = 0.2, and hence E[XY] = 0.4.
- (h) If Y = 2X + 3, and $X \sim \mathcal{N}(0, 1)$, find the mean and variance of Y, and write down the PDF of Y. (3 marks) Ans: E[Y] = 2E[X] + 3 = 3, and $\sigma_Y^2 = 4\sigma_X^2 = 4$. Since Y has the form aX + b and X is Gaussian, $Y \sim \mathcal{N}(3, 4)$ and has the PDF

$$f_Y(y) = \frac{1}{\sqrt{8\pi}} \exp\left(-\frac{(y-3)^2}{8}\right).$$

Q2. A random variable X has the PDF

$$f_X(x) = \frac{5}{4}(1 - x^4), \quad 0 < x \le 1.$$

(a) Find the CDF of X. (5 marks) Ans: By definition, for $0 < x \le 1$,

 $F_X(x) = \int_0^x \frac{5}{4} (1 - t^4) dt \tag{5}$

$$= \frac{5}{4} \left[t - \frac{1}{5} t^5 \right]_0^x \tag{6}$$

$$= \frac{x}{4}(5-x^4). (7)$$

The complete CDF is

$$F_X(x) = \begin{cases} 0 & x \le 0\\ \frac{x}{4}(5 - x^4) & 0 < x \le 1\\ 1 & x > 1 \end{cases}$$

(b) Find E[X] and var(X). (5 marks) Ans: Since X is non-negative, we can use $E[X] = \int_0^\infty 1 - F_X(x) dx$ to obtain

$$E[X] = \int_0^1 1 - \frac{x}{4} (5 - x^4) dx \tag{8}$$

$$= \frac{5}{12}.\tag{9}$$

(Alternatively, we can also find $\int_0^\infty x f_X(x) dx$ directly.)

The second moment of X is

$$E[X^2] = \frac{5}{4} \int_0^1 x^2 - x^6 dx = \frac{5}{21},\tag{10}$$

and hence $var(X) = \frac{5}{21} - \frac{25}{144} = 0.0645$.

(c) Find $f_X(x|X > 0.5)$. (5 marks) Ans: We have

$$P[X > 0.5] = 1 - F_X(0.5) = 1 - \frac{79}{128} = \frac{49}{128}.$$

Therefore,

$$f_X(x|X>0.5) = \frac{128}{49}f_X(x), \quad x>0.5$$
 (11)

$$= \frac{160}{49}(1 - x^4), \quad 0.5 < x \le 1. \tag{12}$$

- Q3. Let the discrete random variable X be uniformly distributed in $\{-3, -1, 1, 3\}$, and let N be a Gaussian random variable with mean 0 and variance σ_n^2 . X and N are independent.
 - (a) If Y = X + N, find the conditional PDF $f_{Y|X}(y|x)$. (3 marks) Ans: Given X = x, we have Y = x + N because N is independent of X. Therefore,

$$f_{Y|X}(y|x) = \frac{1}{\sqrt{2\pi\sigma_n^2}} \exp\left(-\frac{(y-x)^2}{2\sigma_n^2}\right).$$

(b) Find P[Y > 2|X = 1] in terms of the Q function. (3 marks) Ans: From part (a), Y conditioned on X = 1 is $\mathcal{N}(1, \sigma_n^2)$, and thus

$$P[Y > 2|X = 1] = Q\left(\frac{2-1}{\sigma_n}\right) = Q\left(\frac{1}{\sigma_n}\right).$$

(c) Given that Y > 2, what is the most likely value of X? (6 marks) Ans: Proceeding as in part (b), we can obtain

$$P[Y > 2|X = -3] = Q\left(\frac{5}{\sigma_n}\right) \tag{13}$$

$$P[Y > 2|X = -1] = Q\left(\frac{3}{\sigma_n}\right) \tag{14}$$

$$P[Y > 2|X = 1] = Q\left(\frac{1}{\sigma_n}\right) \tag{15}$$

$$P[Y > 2|X = 3] = Q\left(\frac{-1}{\sigma_n}\right). \tag{16}$$

By Bayes rule we have

$$P[X = x|Y > 2] = \frac{P[Y > 2|X = x]P[X = x]}{P[Y > 2]} \propto P[Y > 2|X = x]$$
 (17)

since P[X = x] = 0.25 for all $x \in S_X$, and P[Y > 2] is not a function of x. Noting that the Q(x) function is $P[\mathcal{N}(0,1) > x]$, and that the complementary CDF of $\mathcal{N}(0,1)$ is monotonically decreasing, the largest value of P[Y > 2|X = x] is obtained at x = 3. Therefore the most likely value of X, given that Y > 2, is 3.

(d) Define Z as the indicator function of the event $\{Y > 2\}$. Find the PMF of Z, in terms of the Q function. (3 marks) Ans: We have Z = 1 if Y > 2, and 0 otherwise. Therefore, P[Z = 1] = P[Y > 2], which by total probability is

$$P[Y > 2] = 0.25 \sum_{k \in S_X} Q\left(\frac{2-k}{\sigma_n}\right). \tag{18}$$

P[Z=0]=1-P[Z=1] and thus we have the PMF of Z.

Q4. Consider two i.i.d. exponential random variables X and Y, with E[X] = E[Y] = 2.

(a) Find the PDF of Z = X + Y using convolution. (5 marks) Ans: Both X and Y have the same PDF:

$$f_X(t) = f_Y(t) = 0.5e^{-0.5t}, \quad t > 0.$$

The PDF of Z is $f_Z(t) = f_X(t) * f_Y(t) = \int_{-\infty}^{\infty} f_X(u) f_Y(t-u) du$. With a sketch of the two functions in the convolution integral, we can obtain

$$f_Z(t) = \int_0^t 0.25e^{-u+u-t}du$$
 (19)

$$= 0.25e^{-t} \int_0^t du \tag{20}$$

$$= 0.25te^{-t}, \quad t > 0. \tag{21}$$

 $f_Z(t) = 0$ for $t \le 0$.

(b) Find and sketch the CDF of V = X/Y. (7 marks) Ans: We can find the CDF $F_V(v) = P[X/Y \le v]$ as follows:

$$F_V(v) = \int_0^\infty \int_0^{vy} 0.25e^{-0.5(x+y)} dx dy \tag{22}$$

$$= \int_0^\infty 0.5e^{-0.5y} (1 - e^{-0.5vy}) dy \tag{23}$$

$$\vdots (24)$$

$$= 1 - \frac{1}{1+v}, \quad v > 0. \tag{25}$$

(c) Find and sketch the PDF of V = X/Y. (3 marks) Ans: Differentiating the CDF from part (b), we obtain

$$f_V(v) = \frac{1}{(1+v)^2}, \quad v > 0.$$

- Q5. Consider a Poisson arrival process with average rate $\lambda = 2$ arrivals per minute. The number of arrivals in t minutes is denoted N(t).
 - (a) Find the probability mass function of N(t). (2 marks) Ans: N(t) is Poisson with mean λt , therefore

$$P[N(t) = k] = \frac{(\lambda t)^k}{k!} e^{-\lambda t}, \quad k = 0, 1, 2, \dots$$
 (26)

(b) Suppose that each arrival is independently tagged, with probability p. Find the PMF of M(t), the number of tagged arrivals in t minutes. (7 marks) Ans: Conditioned on N(t) = n, M(t) will be binomial with parameters n and p, i.e.

$$P[M(t) = k | N(t) = n] = \binom{n}{k} p^k (1-p)^{n-k}, \quad k = 0, 1, \dots, n.$$

By total probability, we then have, for k = 0, 1, 2, ...,

$$\begin{split} P[M(t) = k] &= \sum_{n=0}^{\infty} P[M(t) = k | N(t) = n] P[N(t) = n] \\ &= \sum_{n=k}^{\infty} \binom{n}{k} p^k (1-p)^{n-k} \frac{(\lambda t)^n}{n!} e^{-\lambda t} \\ &= e^{-\lambda t} \frac{(p\lambda t)^k}{k!} \sum_{n=k}^{\infty} \frac{[(1-p)\lambda t]^{n-k}}{(n-k)!} \\ &= e^{-\lambda t} \frac{(p\lambda t)^k}{k!} e^{(1-p)\lambda t} \\ &= e^{-p\lambda t} \frac{(p\lambda t)^k}{k!}. \end{split}$$

The second line comes from noting that $k \leq n$; the third from expanding the binomial coefficient and splitting $(\lambda t)^n$ into $(\lambda t)^{n-k}(\lambda t)^k$; and the final line from the power series expansion of e^x . In other words, the random tagging of Poisson arrivals results in another Poisson process.

(c) Let T_i be the waiting time until the *i*-th arrival, i = 1, 2, ... Derive the CDF of T_i . (Hint: What is the event $\{T_i > t\}$ equivalent to?) (6 marks) Ans: T_1 has already been shown in class to be exponential with parameter λ , because of the equivalence of $\{T_1 > t\}$ and $\{N(t) = 0\}$. Similarly, we can reason that $\{T_i > t\}$ is equivalent to $\{N(t) \le i - 1\}$: if the *i*-th arrival occurs more than t minutes later, than within t minutes there must have been fewer than t arrivals; conversely, if within t minutes there are fewer than t arrivals, then the t-th arrival must occur more than t minutes later. Therefore,

$$1 - F_{T_i}(t) = P[N(t) \le i - 1]$$
$$= \sum_{k=0}^{i-1} \frac{(\lambda t)^k}{k!} e^{-\lambda t}$$

and hence

$$F_{T_i}(t) = 1 - \sum_{k=0}^{i-1} \frac{(\lambda t)^k}{k!} e^{-\lambda t}, \quad t > 0.$$

List of Formulae and Notation

Definitions

Indicator Function: $I_A = 1$ if A occurs, 0 otherwise.

Marginal PMF/PDF:
$$p_X(x_j) = \sum_k p_{X,Y}(x_j, y_k); \quad f_X(x) = \int_{-\infty}^{\infty} f_{X,Y}(x, y) dy.$$

Marginal CDF:
$$F_X(x) = F_{X,Y}(x, \infty)$$
.

Joint Moments:
$$\rho_{X,Y} = \frac{\text{cov}(X,Y)}{\sigma_X \sigma_Y}, \text{ cov}(X,Y) = E[XY] - E[X]E[Y].$$

Discrete Random Variables

Bernoulli:
$$p_X(1) = p = 1 - p_X(0), E[X] = p, var[X] = p(1 - p)$$

Binomial:
$$p_X(k) = \binom{n}{k} p^k (1-p)^{n-k}, \quad k = 0, 1, \dots, n. \quad E[X] = np, \text{var}[X] = np(1-p)$$

Geometric:
$$p_X(k) = p(1-p)^{k-1}, k = 1, 2, \dots$$
 $E[X] = \frac{1}{p}, \text{var}[X] = \frac{1-p}{p^2}$

Poisson:
$$p_X(k) = \frac{\alpha^k}{k!} e^{-\alpha}, \quad k = 0, 1, \dots \quad E[X] = \alpha = \text{var}[X]$$

Continuous Random Variables

Uniform:
$$f_X(x) = \frac{1}{b-a}$$
, $a < x < b$. $E[X] = \frac{a+b}{2}$, $var[X] = \frac{(b-a)^2}{12}$

Exponential:
$$f_X(x) = \lambda e^{-\lambda x}, \ x \ge 0. \ E[X] = \frac{1}{\lambda}, \text{var}[X] = \frac{1}{\lambda^2}$$

Gaussian:
$$f_X(x) = \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{(x-\mu)^2}{2\sigma^2}\right), \ x \in \mathbb{R}. \ E[X] = \mu, \text{var}[X] = \sigma^2$$

Gaussianity

Q fm.:
$$Q(x) = \frac{1}{\sqrt{2\pi}} \int_{x}^{\infty} e^{-t^2/2} dt$$
, $Q(x) = 1 - Q(-x)$

CDF:
$$X \sim \mathcal{N}(\mu, \sigma^2) \Rightarrow P[X > t] = Q\left(\frac{t - \mu}{\sigma}\right)$$

Joint PDF:
$$f_{\mathbf{X}}(\mathbf{x}) = \frac{1}{2\pi\sqrt{\det(\mathbf{C})}} \exp\left[-\frac{1}{2}(\mathbf{x} - \boldsymbol{\mu})^T \mathbf{C}^{-1}(\mathbf{x} - \boldsymbol{\mu})\right], \quad \mathbf{x} \in \mathbb{R}^2$$

where
$$\mathbf{C} = E[(\mathbf{X} - \boldsymbol{\mu})(\mathbf{X} - \boldsymbol{\mu})^T]$$

Result 1: If X is jointly Gaussian, then Y = AX + b is jointly Gaussian.

Result 2: If X is jointly Gaussian, then its components are marginally Gaussian.

Other Useful Results

Bayes:
$$P(B_j|A) = \frac{P(A|B_j)P(B_j)}{\sum_{k=1}^n P(A|B_k)P(B_k)}$$
, where $\{B_k\}_{k=1}^n$ is a partition of \mathcal{S} .

Total Prob.:
$$P(A) = \sum_{k=1}^{n} P(A|B_k)P(B_k)$$
, where $\{B_k\}_{k=1}^{n}$ is a partition of \mathcal{S} .

$$E[X] = \sum_{k=1}^{n} E[X|B_k]P[B_k]; \quad f_X(x) = \sum_{k=1}^{n} f_X(x|B_k)P[B_k].$$

Functions of X:
$$Y = aX + b \Rightarrow f_Y(y) = \frac{1}{|a|} f_X\left(\frac{y-b}{a}\right)$$

$$Y = g(X) \Rightarrow f_Y(y) = \sum_{k=1}^n f_X(x_k) \left| \frac{dx_k}{dy} \right|$$
, where $g(x_k) = y$, $k = 1, \dots, n$.

Independence (rv's):
$$X, Y$$
 independent $\Leftrightarrow F_{X,Y}(x,y) = F_X(x)F_Y(y)$,

$$f_{X,Y}(x,y) = f_X(x)f_Y(y), \ p_{X,Y}(x,y) = p_X(x)p_Y(y).$$

Sum of rv's:
$$X, Y$$
 independent $\Rightarrow f_{X+Y}(z) = f_X(z) * f_Y(z)$.