## Probability and Stochastic Processes

## **Exam 2018 Solutions**

- 1. part a) is book work; surprisingly, quite a number of students couldn't derive it.
- a) It can be derived that

$$f_Y(y) = \begin{cases} \frac{1}{2\sqrt{y}} \left( f_X(\sqrt{y}) + f_X(-\sqrt{y}) \right), & y > 0, \\ 0, & \text{otherwise.} \end{cases}$$
 [3]

Since the Gaussian pdf is even,

$$f_Y(y) = \frac{1}{\sqrt{y}} f_X\left(\sqrt{y}\right) , \quad y > 0.$$
 [3]

Substituting  $f_X(x) = \frac{1}{\sqrt{2\pi}\sigma} e^{-x^2/2\sigma^2}$ , we obtain

$$f_Y(y) = \frac{1}{\sqrt{2\pi y}\sigma} e^{-y/2\sigma^2} \quad y > 0.$$
 [4]

- b) students did well in part b), but some didn't know the difference between these bounds
- i) Markov

$$P(|X| > a) \le \frac{E(|X|)}{a} = \frac{\sqrt{\frac{2}{\pi}}\sigma}{a} = \frac{\sqrt{\frac{2}{\pi}}\sigma}{3\sigma} = 0.265$$
 [4]

ii) Chebyshev

$$P(|X| > a) \le \frac{\sigma^2}{a^2} = \frac{1}{9} = 0.111$$
 [4]

iii) Chernoff

$$P(|X| > a) = 2P(X > a) \le 2e^{-\frac{a^2}{2\sigma^2}} = 2e^{-4.5} = 0.022$$
 [4]

Discussion: The accuracy of estimation improves since more moments are used from (i) to (iii), because more moments are used. [3]

## students did well in Question 2, probably because of similar questions in previous years

2.

a)

For n samples we have
$$f(\underline{\chi},c) = c^{4n} (\underline{\chi}, \dots \underline{\chi}_n)^{3n} e^{-c(\underline{\chi}_1 + \dots + \underline{\chi}_n)}$$

$$\frac{2f(\underline{\chi},c)}{\partial c} = 4n \cdot c^{4n+1} (\underline{\chi}, \dots \underline{\chi}_n)^{3n} e^{-c(\underline{\chi}_1 + \dots + \underline{\chi}_n)}$$
[2]

$$= \left[\frac{4n}{c} - (x_1 + \dots + x_n)\right] f(x, c)$$

$$= 0$$

$$C = \frac{4n}{\chi_1 + \dots + \chi_n}$$
 [2]

In this problem, n = 4, so

$$c = \frac{4 \times 4}{16} = 1.$$
 [2]

b) Recall the Wiender-Hopf equation standard question of linear prediction

$$\mathbf{c} = \mathbf{R}^{-1}\mathbf{r}$$

$$\sigma^2 = r_0 - \mathbf{r}^T \mathbf{R}^{-1} \mathbf{r}$$

i) If n=1, the Wiener-Hopf equation trivially reads

$$R_Y(0)c_1 = R_Y(1)$$

Therefore,

$$c_1 = \frac{R_Y(1)}{R_Y(0)} = \frac{2}{3}$$
 [3]

[2]

Mean-square error

$$\sigma^2 = 3 - 4/3 = 5/3$$

ii) When n = 2, we have

$$\mathbf{R} = \begin{bmatrix} 3 & 2 \\ 2 & 3 \end{bmatrix}$$

$$\mathbf{r} = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$$
[3]

Thus the coefficient vector

$$\mathbf{c} = \begin{bmatrix} c_1 \\ c_2 \end{bmatrix} = \mathbf{R}^{-1}\mathbf{r} = \frac{1}{5} \begin{bmatrix} 3 & -2 \\ -2 & 3 \end{bmatrix} \quad \begin{bmatrix} 1 \\ 2 \end{bmatrix} = \begin{bmatrix} -1/5 \\ 4/5 \end{bmatrix}$$
[4]

Mean-square error

$$\sigma^2 = 3 - \begin{bmatrix} 1 & 2 \end{bmatrix} \frac{1}{5} \begin{bmatrix} 3 & -2 \\ -2 & 3 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \end{bmatrix} = 8/5$$
 [3]

- 3. a) 1 mark each except those marked otherwise
- (1) linearity
- (2),(3), convolution
- (4) definition of power
- (5) due to definition of power spectral density
- (6) spectrum of  $y_s(t)$  is  $S(\omega)H(\omega)$  from (2); power spectrum of n(t) is  $S_{ww}(\omega)|H(\omega)|^2$  from (3) [2]
- (7) w(t) is white noise with power spectral density  $N_0$
- (8) because of Cauchy-Schwarz inequality some student didn't know Cauchy-Schwarz

$$\left| \int_{-\infty}^{+\infty} S(\omega) H(\omega) e^{j\omega t_0} d\omega \right|^2 \le \int_{-\infty}^{+\infty} |S(\omega) e^{j\omega t_0}|^2 d\omega \int_{-\infty}^{+\infty} |H(\omega)|^2 d\omega$$
 [2]

(9) Parseval's theorem

$$^{(10)}\int_{0}^{+\infty}s(t)^{2}dt=E_{s}$$

- b) part b) is a common question for Poisson processes.
- i) This is the same as the average time when the fourth student arrives.

$$E[T_4] = \frac{4}{\lambda} = \frac{4}{0.2} = 20 \text{ minutes}$$
 [3]

ii) This means that the number of students arriving in the first half an hour is less than 4.

Let 
$$t = 30$$
 minutes. Recall  $N(t)$  has Poisson distribution:

$$P(N(t) = k) = e^{-\lambda t} \frac{(\lambda t)^k}{k!}, \qquad k = 0,1,2,...$$

Therefore,

$$P(N(t) < 4) = P(N(t) = 0) + P(N(t) = 1) + P(N(t) = 2) + P(N(t) = 3)$$

$$= e^{-6} + 6e^{-6} + 18e^{-6} + 36e^{-6} = 0.151$$
[3]

iii) Let  $t_1 = 10$  minutes,  $t_2 = 20$  minutes. Since a Poisson process is memoryless,

$$P(N(t_1) \ge 1)P(N(t_2) - N(t_1) \le 2)$$
 [1]

$$= P(N(t_1) \ge 1)P(N(t_2 - t_1) \le 2)$$

$$= [1 - P(N(t_1) = 0)][P(N(t_2 - t_1) = 0) + P(N(t_2 - t_1) = 1) + P(N(t_2 - t_1) = 2)$$
 [2]

$$= [1 - e^{-2}][e^{-2} + 2e^{-2} + 2e^{-2}] = 0.585$$

[2]

[2]

## A small number of student didn't know what is a martingale.

4. a) i) Using the identity cos(A + B) + cos(A - B) = 2 cos A cos B, we have

$$E[X_{n+1}] = \frac{\cos\left\{\lambda\left[S_n + 1 - \frac{1}{2}(b - a)\right]\right\} + \cos\left\{\lambda\left[S_n - 1 - \frac{1}{2}(b - a)\right]\right\}}{2(\cos\lambda)^{n+1}}$$

$$= \frac{\cos\left\{\lambda\left[S_n - \frac{1}{2}(b - a)\right]\right\} \cos\{\lambda\}}{(\cos\lambda)^{n+1}}$$

$$= \frac{\cos\{\lambda\left[S_n - \frac{1}{2}(b - a)\right]\}}{(\cos\lambda)^n} = X_n \quad \text{remember, we need check the definition}$$

$$E[X_n + 1] = X_n$$

Thus,  $X_n$  forms a martingale

[2]

ii) Since  $X_n$  forms a martingale, we know

$$E[X_T] = X_0$$

Obviously

$$X_0 = \cos\{\frac{1}{2}\lambda(b-a)\}\tag{2}$$

Now

$$E[X_T] = E_T E_{S_T}[X_T | T]$$

$$= E_T \frac{E_{S_T} \cos\left\{\lambda \left[S_T - \frac{1}{2}(b - a)\right]\right\}}{(\cos \lambda)^T}$$
[2]

$$= E_T \frac{P(S_T = -a)\cos\left\{\lambda\left[-a - \frac{1}{2}(b - a)\right]\right\} + (S_T = b)\cos\left\{\lambda\left[b - \frac{1}{2}(b - a)\right]\right\}}{(\cos\lambda)^T}$$
 [2]

$$= E_T \frac{P(S_T = -a)\cos\left\{\lambda\left[\frac{1}{2}(b+a)\right]\right\} + P(S_T = b)\cos\left\{\lambda\left[\frac{1}{2}(b+a)\right]\right\}}{(\cos\lambda)^T}$$

$$= E_T \frac{\cos\left\{\lambda \left[\frac{1}{2}(a+b)\right]\right\}}{(\cos \lambda)^T}$$

$$= \frac{\cos\left\{\lambda \left[\frac{1}{2}(a+b)\right]\right\}}{E[(\cos\lambda)^T]}$$
 part ii) is a bit tricky [2]

Note that  $P(S_T = -a) + P(S_T = b) = 1$ .

(b) The stationary distribution satisfies

$$\pi_1 = \pi_0, \pi_2 = \frac{1}{2}\pi_1, \pi_3 = \frac{1}{3}\pi_2, \pi_4 = \frac{1}{4}\pi_3, \dots$$
 [2]

from which we get

[2]

[3]

[2]

$$\pi_i = \frac{1}{i!}\pi_0$$

Since

$$\sum_{i=0}^{\infty} \pi_i = \sum_{i=0}^{\infty} \frac{1}{i!} \pi_0 = e \pi_0 = 1,$$

we find  $\pi_0 = e^{-1}$  and

$$\pi_i = \frac{e^{-1}}{i!}, \quad i = 1, 2, 3 \dots$$
 [3]

Common mistakes in part (b) are:

5

Some students treat it as a finite state space Markov chain;

Some couldn't calculate the infinite sum if 1/i!.