

# Solutions to EE5137 Exam (Semester 1 2018/9)

December 3, 2018

## 1 Problem 1

- (a) (i) By the question, we know that

$$\mathbb{E}[X|Y = 1] = 3, \quad \mathbb{E}[X|Y = 2] = \mathbb{E}[X] + 5, \quad \mathbb{E}[X|Y = 3] = \mathbb{E}[X] + 7,$$

Hence,  $a_1 = 0, b_1 = 3, a_2 = 1, b_2 = 5, a_3 = 1, b_3 = 7$ .

- (ii) By the law of iterated expectations,

$$\begin{aligned} \mathbb{E}[X] &= \mathbb{E}[\mathbb{E}[X|Y]] \\ &= \frac{1}{3}(3) + \frac{1}{3}(\mathbb{E}[X] + 5) + \frac{1}{3}(\mathbb{E}[X] + 7) \end{aligned}$$

This yields after solving the equation

$$\frac{1}{3}\mathbb{E}[X] = 1 + \frac{5}{3} + \frac{7}{3} = 5, \quad \implies \quad \mathbb{E}[X] = 15.$$

- (a) (i) We claim that  $b = 0$ . Let's prove it. For fixed  $\epsilon > 0$

$$\mathbb{P}(|X_n - 0| > \epsilon) = \mathbb{P}(X_n = n^2) = \frac{1}{n} \rightarrow 0$$

Hence,  $X_n$  converges to 0 in probability and so  $b = 0$ .

- (ii) We claim that  $c = \infty$ . Let's prove it. Consider

$$\mathbb{E}[X_n] = \Pr(X_n = n^2) \cdot n^2 + \Pr(X_n = 0) \cdot 0 = \frac{1}{n} \cdot n^2 = n$$

Clearly,  $\mathbb{E}[X_n]$  diverges to  $\infty$  and so  $c = \infty$ .

## 2 Problem 2

(a) Consider the likelihood ratio test

$$\frac{f_{X|H_0}(x)}{f_{X|H_1}(x)} \stackrel{\text{decide } H_0}{\gtrless} \frac{p_1}{p_0} =: \eta,$$

where  $p_1$  and  $p_0$  are the prior probabilities of  $H_0$  and  $H_1$  resp. Then we have

$$\frac{f_{X|H_0}(x)}{f_{X|H_1}(x)} = \frac{\frac{1}{\sqrt{2\pi\sigma_0^2}} \exp(-\frac{x^2}{2\sigma_0^2})}{\frac{1}{\sqrt{2\pi\sigma_1^2}} \exp(-\frac{x^2}{2\sigma_1^2})} = \frac{\sigma_1}{\sigma_0} \exp\left(-\frac{x^2}{2}(\sigma_0^{-2} - \sigma_1^{-2})\right).$$

Comparing this to the threshold  $\eta$ , we obtain

$$-\frac{x^2}{2}(\sigma_0^{-2} - \sigma_1^{-2}) \stackrel{\text{decide } H_0}{\gtrless} \ln\left(\eta \frac{\sigma_0}{\sigma_1}\right)$$

Hence, by using the fact that  $\sigma_0 < \sigma_1$

$$x^2 \stackrel{\text{decide } H_1}{\gtrless} 2 \frac{1}{\sigma_1^{-2} - \sigma_0^{-2}} \ln\left(\eta \frac{\sigma_0}{\sigma_1}\right) =: \gamma^2$$

So we decide in favor of  $H_1$  iff

$$|x| > \gamma.$$

(b) (i) We have

$$L(X_1, X_2, X_3) = \frac{P_0(X_1)P_0(X_2)P_0(X_3)}{P_1(X_1)P_1(X_2)P_1(X_3)} = \frac{\prod_{i=1}^3 (\frac{1}{2})^{X_i} (\frac{1}{2})^{1-X_i}}{\prod_{i=1}^3 p^{X_i} (1-p)^{1-X_i}} = \frac{1/8}{(2/3)^T (1/3)^{3-T}}$$

Since  $L(X_1, X_2, X_3)$  depends only on  $T$ ,  $T$  is a sufficient statistic.

(ii) Note that  $T \in \{0, 1, 2, 3\}$ . Evaluating the likelihood ratio,

$$L(X_1, X_2, X_3) = \begin{cases} 27/8 & T = 0 \\ 27/16 & T = 1 \\ 27/32 & T = 2 \\ 27/64 & T = 3 \end{cases}$$

(iii) For probability of false alarm to be  $1/8$ , we need to put the threshold at  $(27/64, 27/32)$  and declare that if  $T > 2$ , then  $H_1$  is declared. This is because  $P_0(T > 2) = P_0(T = 3) = 1/8$ . Hence, the best probability of detection is  $P_1(T > 2) = P_1(T = 3) = (2/3)^3 = 8/27$ .

(iv) For probability of false alarm to be  $1/4$ , we consider that  $P_0(T > 1) = 1/2$  and the corresponding probability of detection is  $P_1(T > 1) = (2/3)^3 + 3(2/3)^2(1/3) = 20/27$ . Hence, we need to randomize between the strategy that places the threshold at  $T > 2$  and  $T > 1$ . Now we find  $\alpha \in [0, 1]$  such that

$$\alpha \frac{1}{8} + (1 - \alpha) \frac{1}{2} = \frac{1}{4}, \quad \implies \quad \alpha = \frac{2}{3}$$

Thus, the best probability of detection is

$$\alpha \frac{8}{27} + (1 - \alpha) \frac{20}{27} = \frac{12}{27}.$$

The best test in terms of  $T$  would be to randomize between  $T > 2$  and  $T > 1$  where the former has probability  $2/3$ .

### 3 Problem 3

- (a) States 1 and 4 are transient. States 2 and 3 constitute a class of recurrent states and state 5 constitutes another class (a singleton class) of recurrent states.
- (b) Let  $\pi^{(1)}$  be the steady-state vector for class  $\{2, 3\}$  and let  $\pi^{(2)}$  be the steady state vector for class  $\{5\}$ . For the class  $\{5\}$ ,  $\pi^{(2)} = (0, 0, 0, 0, 1)$ . For the class  $\{2, 3\}$ , the steady-state equations (written just for the recurrent class) are

$$\pi_3^{(1)} P_{32} = \pi_2^{(1)}; \quad \pi_2^{(1)} P_{23} + \pi_3^{(1)} P_{33} = \pi_3^{(1)}; \quad \pi_2^{(1)} + \pi_3^{(1)} = 1$$

Solving these, we obtain

$$\pi_3^{(1)} = \frac{1}{P_{32} + 1}, \quad \pi_2^{(1)} = \frac{P_{32}}{P_{32} + 1}.$$

- (c) (i)  $P_{44}^n = (1/3)^n$  since  $n$  successive self-loop transitions, each of probability  $1/3$ , are required.
- (ii)  $P_{45}^n = (1/3)^1 + (1/3)^2 + \dots + (1/3)^n = \frac{1}{2}(1 - 3^{-n})$ . The reason for this is that there are  $n$  walks going from 4 to 5 in  $n$  steps; each such walk contains the 4 to 5 transition at a different time. If it occurs at time  $i$  then there are  $i - 1$  self-transitions from 4 to 4, so the probability of that walk is  $(1/3)^i$ .
- (iii)  $P_{41}^n = n(1/3)^n$  since there are  $n$  walks that go from 4 to 1 in  $n$  steps, one for each step in which the  $4 \rightarrow 1$  transition can be made. Each walk has probability  $(1/3)^n$ .
- (iv)  $P_{43}^n + P_{42}^n = 1 - P_{44}^n - P_{45}^n - P_{41}^n = \frac{1}{2} - \frac{2n+1}{2}3^{-n}$  since  $\sum_j P_{4j}^n = 1$ .
- (v)  $\lim_{n \rightarrow \infty} P_{43}^n$ : From (iv) note that  $\lim_{n \rightarrow \infty} P_{43}^n + P_{42}^n = 1/2$ . Given that  $X_n \in \{2, 3\}$ ,

$$\lim_{m \rightarrow \infty} \Pr(X_m = 3 | X_n \in \{2, 3\}) = \pi_3^{(1)} = \frac{1}{P_{32} + 1}$$

Since  $\lim_{n \rightarrow \infty} \Pr(X_n \in \{2, 3\} | X_0 = 4) = 1/2$ , we see that

$$\lim_{m \rightarrow \infty} P_{43}^m = \frac{1}{2(P_{32} + 1)}.$$

- (d) We only have to find the eigenvalue with the second largest magnitude. For this purpose, consider

$$\det(P - \lambda I) = 0 \quad \Rightarrow \quad \det \left( \begin{bmatrix} 0.2 - \lambda & 0.8 \\ 0.5 & 0.5 - \lambda \end{bmatrix} \right) = 0.$$

Multiplying out, we see that

$$(0.2 - \lambda)(0.5 - \lambda) - 0.4 = 0.$$

This quadratic equation has two roots,  $\lambda_1 = 1$  and  $\lambda_2 = -0.3$ . Hence,  $\phi = -0.3$ .

## 4 Problem 4

- (a) (i) Independent increments property of the Poisson process.
- (ii) This distribution is exponential with rate  $\lambda$ .
- (iii)  $k = 0$ .
- (iv) We need to evaluate

$$\Pr(\text{no arrivals in interval } [t^* - x, t^*]) = \Pr(N(x) = 0) = e^{-\lambda x}.$$

We have shown that

$$\Pr(t^* - U > x) = e^{-\lambda x}, \quad \Pr(t^* - U \leq x) = 1 - e^{-\lambda x}$$

so  $t^* - U$  is also exponential with rate  $\lambda$ .

- (v) Since  $t^* - U$  and  $V - t^*$  are independent exponentials with rate  $\lambda$ , their sum  $L$  is Erlang of order 2 with rate  $\lambda$ .
  - (vi) The interarrival time of a Poisson process is an exponential with rate  $\lambda$ . It is more likely that  $t^*$ , being observed, is in a longer interarrival interval.
- (b) We have

$$\Pr(S_n \leq t) = \Pr((X_1 - \mu) + \dots + (X_n - \mu) \leq t - n\mu) = \Pr\left(\frac{1}{n}((X_1 - \mu) + \dots + (X_n - \mu)) \leq \frac{t}{n} - \mu\right)$$

For  $n$  large enough  $\frac{t}{n} - \mu \leq -\frac{\mu}{2}$ . Hence,

$$\Pr(S_n \leq t) \leq \Pr\left(\frac{1}{n}((X_1 - \mu) + \dots + (X_n - \mu)) \leq -\frac{\mu}{2}\right)$$

Since each of the summands on the right-hand-side have zero mean, by Chebyshev's inequality or the weak law of large numbers, the right-hand-side probability converges to zero so  $\Pr(S_n \leq t) \rightarrow 0$ .