

IMPERIAL COLLEGE LONDON

DEPARTMENT OF ELECTRICAL AND ELECTRONIC ENGINEERING  
EXAMINATIONS 2010

MSc and EEE PART IV: MEng and ACGI

Corrected Copy

4) a) ii)

**PROBABILITY AND STOCHASTIC PROCESSES**

Thursday, 6 May 10:00 am

Time allowed: 3:00 hours

Q4,a,ii Correction announced  
at the start.

**There are SIX questions on this paper.**

**Answer FOUR questions.**

*All questions carry equal marks*

**Any special instructions for invigilators and information for candidates are on page 1.**

Examiners responsible      First Marker(s) :      M.M. Draief  
Second Marker(s) :      R.B. Vinter

## PROBABILITY AND STOCHASTIC PROCESSES

1. We consider two random variables  $X$  and  $Y$  such that the joint distribution of the random vector  $(X, Y)$  is given by, for  $1 \leq j \leq i \leq 5$ ,

$$p_{ij} = \mathbf{P}(X = i, Y = j) = \frac{1}{15}.$$

- a) i) Show that  $(p_{ij})_{i,j \in \{1, \dots, 5\}}$  is a probability distribution. [ 1 ]  
 ii) Compute the marginal distributions of  $X$  and  $Y$ . [ 2 ]  
 iii) For  $1 \leq j \leq i \leq 5$ , compute the conditional distribution

$$\mathbf{P}(X = i \mid Y = j).$$

[ 2 ]

- b) i) Compute  $\mathbf{E}(X \mid Y = j)$ ,  $j = 1, \dots, 5$ . Show that  $\mathbf{E}(X \mid Y) = \frac{Y+5}{2}$ . [ 3 ]  
 ii) Compute  $\mathbf{E}(Y \mid X = i)$ ,  $i = 1, \dots, 5$ , and  $\mathbf{E}(Y \mid X)$ . [ 3 ]  
 iii) Using questions 1. b) i) and ii), show that

$$\mathbf{E}(X) = \frac{1}{2}\mathbf{E}(Y) + \frac{5}{2}, \quad \mathbf{E}(Y) = \frac{1}{2}\mathbf{E}(X) + \frac{1}{2}.$$

[ 3 ]

- iv) Using the previous question, compute  $\mathbf{E}(X)$  and  $\mathbf{E}(Y)$ . [ 2 ]  
 c) By analogy to the argument above, derive  $\mathbf{E}(X)$  and  $\mathbf{E}(Y)$  for  $X$  and  $Y$  having joint distribution

$$\mathbf{P}(X = i, Y = j) = \frac{2}{n(n+1)},$$

where  $n$  is a positive integer and  $1 \leq j \leq i \leq n$ .

[ 4 ]

2. We consider an urn containing  $n$  balls numbered  $1, \dots, n$ . We uniformly draw a ball from the urn, we write down its number and put it back in the urn. We suppose that the draws are independent from each other. We stop the process when we see each ball (number) at least once. This process is known as the *coupon collector problem*.

Let  $\tau$  be the (random) time by which we complete the coupon collection.

- a) i) Let  $\tau_i$  be the number of draws before you see  $i$  different balls,  $i = 1, \dots, n$ . We have  $\tau_n = \tau$ .  
Derive the distribution of  $X_i = \tau_i - \tau_{i-1}$ ,  $i = 2, \dots, n$ , and compute  $\mathbb{E}(X_i)$ . [ 3 ]
- ii) Show that  $\mathbb{E}(\tau) \sim n \log(n)$ , for  $n$  large. [ 3 ]  
*Hint:* Use the fact that  $\sum_{i=1}^n \frac{1}{i} \sim \log(n)$ , for  $n$  large.

- b) For  $j = 1, \dots, n$ , let  $Y_{k,j}$  be the indicator function that the ball  $j$  is **not** chosen in one of the first  $k$  draws.

- i) Compute  $\mathbb{E}(Y_{k,j})$  and  $\mathbb{E}(Y_{k,i}Y_{k,j})$ , for  $i \neq j$ . [ 3 ]
- ii) Compute the covariance of  $Y_{k,i}$  and  $Y_{k,j}$ . Note that  $\text{Cov}(Y_{k,i}, Y_{k,j})$  is negative. Comment. [ 3 ]

- c) Let  $U_k = \sum_{i=1}^n Y_{k,i}$ .

- i) Using 2. b) ii), show that  $\text{Var}(U_k) \leq \mathbb{E}(U_k)$ . [ 3 ]
- ii) Let  $k = cn$ , for  $c > 0$ , show that  $\lim_{n \rightarrow \infty} \mathbb{E}(U_k)/n = e^{-c}$ . [ 1 ]  
*Hint:* Use the fact that  $\lim_{n \rightarrow \infty} (1 - \frac{1}{n})^{cn} = e^{-c}$ .
- iii) Using Chebyshev's inequality, show that, for any  $\varepsilon > 0$ ,

$$\lim_{n \rightarrow \infty} \mathbf{P}(|U_n - \mathbb{E}(U_n)| \geq n\varepsilon/2) = 0.$$

Comment. [ 4 ]

*Chebyshev's inequality:* Let  $X$  with mean  $\mathbb{E}(X)$  and variance  $\text{Var}(X)$ . Then, for  $x > 0$ ,

$$\mathbf{P}(|X - \mathbb{E}(X)| \geq x) \leq \frac{\text{Var}(X)}{x^2}.$$

3. Let  $T$  be an integer-valued random variable distributed according to the following distribution, for  $k = 0, 1, 2, \dots$ ,

$$\mathbf{P}(T = k) = (1 - a)a^k,$$

where  $a \in (0, 1)$ .

- a) The *characteristic function* of a random variable  $X$  is given by, for  $|z| \leq 1$ ,

$$\phi_X(z) = \mathbf{E}(z^X).$$

- i) Compute  $\phi_T(z)$  and deduce the mean and the variance of  $T$ . [ 3 ]  
 ii) Compute  $\mathbf{P}(T \geq n)$ , for  $n \geq 1$  and show that, for  $k \geq 0$ ,

$$\mathbf{P}(T \geq n + k \mid T \geq n) = \mathbf{P}(T \geq k).$$

Comment. [ 3 ]

- b) Let  $T_1, T_2, \dots$  be a sequence of independent and identically distributed random variables following the same distribution as  $T$  above.

Derive the characteristic function, mean and variance of the random variable  $S_n = T_1 + \dots + T_n$ ,  $n \geq 1$ . [ 2 ]

- c) Let  $\nu$  be an integer-valued random variable with distribution, for  $n = 1, 2, \dots$ ,

$$\mathbf{P}(\nu = n) = (1 - b)b^{n-1},$$

where  $b \in (0, 1)$ . We assume that  $\nu$  is independent of the random variables  $T_1, T_2, \dots$  in question b).

- i) Show that

$$\mathbf{P}(S_\nu = k) = \sum_{n \geq 1} \mathbf{P}(\nu = n) \mathbf{P}(S_n = k).$$

[ 2 ]

*Hint:*  $S_\nu$  is equal to  $S_n$  when  $\nu = n$ .

- ii) Show that

$$\mathbf{E}(S_\nu) = \mathbf{E}(T)\mathbf{E}(\nu) \quad \text{and} \quad \text{Var}(S_\nu) = \text{Var}(\nu)\mathbf{E}(T)^2 + \mathbf{E}(\nu)\text{Var}(T).$$

[ 5 ]

- iii) Show that the characteristic distribution of  $S_\nu$  is given by

$$\phi_{S_\nu}(z) = \frac{1 - \alpha}{1 - \alpha z},$$

where  $\alpha = \frac{a}{1 - b(1 - a)}$ . [ 4 ]

- iv) What is the distribution of  $S_\nu$ ? [ 1 ]

4. Let  $X_1, X_2, \dots$  be a sequence of independent and identically distributed random variables such that

$$\mathbf{P}(X_i = -1) = 1 - \mathbf{P}(X_i = 1) = p \quad \text{for all } i \geq 1,$$

where  $p \in (0, 1)$ . Define the following sequence, for  $n \geq 1$ ,

$$Z_n = X_1 X_2 \dots X_n.$$

- a) i) Compute the conditional probabilities  $\mathbf{P}(Z_n = y \mid Z_{n-1} = x)$ , for  $x, y \in \{-1, 1\}$  and  $n \geq 1$ . [ 2 ]  
 → ii) Show that the random process  $(Z_n)_{n \geq 0}$  is a homogeneous Markov process. Carefully justify your answer. [ 4 ]  
 iii) Derive a necessary and sufficient condition on the parameter  $p$  so that the random variables  $Z_1, Z_2, \dots$  are independent. [ 2 ]

- b) i) Show that, for  $n, k \geq 1$ , the transition probabilities

$$\mathbf{P}(Z_{n+k} = y \mid Z_n = x), \quad x, y \in \{-1, 1\}$$

are given by the matrix

$$\begin{pmatrix} 1 - p_k & p_k \\ p_k & 1 - p_k \end{pmatrix}$$

where  $1 - 2p_k = (1 - 2p)^k$ . [ 5 ]

- ii) Compute  $\lim_{n \rightarrow \infty} \mathbf{P}(Z_n = x)$ ,  $x \in \{-1, 1\}$  using two different methods: (1) by question 4. b) i), and (2) by solving the equation for the invariant distribution. [ 5 ]  
 iii) For  $x, y \in \{-1, 1\}$ , compute

$$\lim_{n \rightarrow \infty} \mathbf{P}(Z_{n+1} = x, Z_n = y).$$

[ 2 ]

5. In this problem, we analyse the *Acceptance-Rejection Method*. More precisely, we assume that we know how to generate random variables with probability density function  $g$  and we wish to find a method to generate a random variable with probability density function  $f$ .

We suppose that there exists a constant  $c > 1$  such that, for all  $x \in \mathbb{R}$ , we have

$$f(x) \leq cg(x).$$

We also assume that we can generate uniformly distributed random variables on the interval  $[0, 1]$ .

- a) Let  $U$  be a uniformly distributed random variable on  $[0, 1]$ .
- i) Give the expression of the probability density function of  $U$  and compute its mean and its variance. [ 2 ]
  - ii) Let  $x \in (0, 1)$ , compute  $\mathbf{P}(U \leq x)$ . [ 1 ]
- b) Let  $U_1, U_2, \dots$  be a sequence of independent and identically distributed random variables uniformly distributed on  $[0, 1]$  and let  $Y_1, Y_2, \dots$  be a sequence of independent and identically distributed random variables with probability density function  $g$ . We assume that the two sequences are independent.

We are interested in the sequence of events, for  $n \geq 1$ ,

$$A_n = \left\{ U_n > \frac{f(Y_n)}{cg(Y_n)} \right\}.$$

- i) Compute  $\mathbf{P}(A_n^c)$ , where  $A_n^c$  is the complement of  $A_n$ . [ 3 ]
- ii) Show that, for  $x \in \mathbb{R}$ ,

$$\mathbf{P}(A_n^c, Y_n \leq x) = \frac{1}{c} \int_{-\infty}^x f(y) dy. \quad [ 2 ]$$

- iii) Using the two previous questions, compute  $\mathbf{P}(A_1, \dots, A_{n-1}, A_n^c, Y_n \leq x)$ . [ 4 ]

- c) To generate a random variable according to  $f$ , we generate the random variables  $Y_n$  and  $U_n$  until the instant  $\tau$  which is the first (random) instant  $n$  where we have

$$U_n \leq \frac{f(Y_n)}{cg(Y_n)}.$$

In what follows we will show that  $Z = Y_\tau$  is distributed according to  $f$ .

- i) Show that

$$\mathbf{P}(Z \leq x) = \sum_{n \geq 1} \mathbf{P}(A_1, \dots, A_{n-1}, A_n^c, Y_n \leq x). \quad [ 3 ]$$

- ii) Conclude that  $Z$  has  $f$  as probability density function. [ 5 ]



6. a) Let us assume that  $(X_t)_{t \geq 0}$  is a given continuous-time Markov chain on the state space  $S$  with rate matrix  $Q = (q_{ij})_{i,j \in S}$ . Show that if there exists a probability distribution  $\pi = (\pi_i)_{i \in S}$  such that, for all  $i, j \in S$ , we have

$$\pi_i q_{ij} = \pi_j q_{ji},$$

then  $\pi$  is the invariant distribution of  $(X_t)_{t \geq 0}$ . In this case, the Markov chain is said to be *reversible*. [ 3 ]

*Hint:* Use the fact that if  $Q$  is a rate matrix then  $\sum_{j \in S} q_{ij} = 0$ , for all  $i \in S$ .

- b) We consider the dynamics of a Markovian single server queue in continuous time. Customers join the queue and are served on a first-in-first-out basis, i.e., according to the order in which they join the queue. We suppose that the time between two successive arrivals is exponentially distributed with parameter  $\lambda > 0$  and that each customer requires a service time that is exponentially distributed with parameter  $\mu > 0$ .

Let  $X_t$ ,  $t \geq 0$ , be the random process that describes the number of customers waiting in the queue including the one being served.

- i) Derive the transition matrix  $Q$  of  $(X_t)_{t \geq 0}$ . [ 2 ]
- ii) Using question 6. a), show that, if  $\rho = \frac{\lambda}{\mu} < 1$ , the stationary (invariant) distribution of  $(X_t)_{t \geq 0}$  is given by

$$\pi_i = (1 - \rho) \rho^i.$$

Comment on the *stability condition*  $\rho < 1$ . [ 4 ]

- iii) Compute the average number of customers in the queue in the stationary regime. [ 2 ]

- c) Let us now assume that we have a post office with two cashiers. The two queues at the cashiers run (separately) in parallel. Each of these queues operates following the dynamics in 6) b) with the same parameters  $\lambda$  and  $\mu$ .

- i) Derive the average number of customers in the post office. [ 1 ]
- ii) A clever employee suggests merging the two queues so that customers arrive at rate  $2\lambda$  wait until one of the two cashiers is available and is then served. Describe the underlying continuous-time Markov chain and discuss its stability. [ 2 ]
- iii) Using similar arguments as in 6. b), show that the average number of customers in the post office in stationary regime is given by

$$\frac{2\mu\lambda}{(\mu + \lambda)(\mu - \lambda)}.$$

[ 4 ]

- iv) Is the suggestion of the employee better than running the two queues separately? Justify your answer. [ 2 ]

PROBABILITY & STOCHASTIC PROCESSES.

SOLUTIONS 2010

É 4.10  
SC4  
CS5.1

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a/  
e/

$$i/ \sum_{1 \leq j \leq 5} \frac{1}{15} = \sum_{i=1}^5 \sum_{j=1}^5 \frac{1}{15} = \sum_{j=1}^5 \frac{5}{15} = \frac{6 \times 5}{2} \times \frac{1}{15} = 1$$

$$ii) i, j: 1 \dots 5 \quad P(X=i) = \sum_{j=1}^5 P(X=i, Y=j) = \frac{i}{15}$$

$$P(Y=j) = \sum_{i=j}^5 P(X=i, Y=j) = \frac{6-j}{15}$$

$$iii/ \quad P(X=i | Y=j) = \frac{P(X=i, Y=j)}{P(Y=j)} \quad 1 \leq j \leq 5$$

$$= \frac{\frac{1}{15}}{\frac{6-j}{15}} = \frac{1}{6-j}$$

b/  
i/

$$E(X | Y=j) = \sum_{i=j}^5 i P(X=i | Y=j)$$

$$= \frac{1}{6-j} \sum_{i=j}^5 i = \frac{1}{6-j} \frac{30 - j(j-1)}{2}$$

$$= \frac{j+5}{2}$$

$$E(X | Y) = \frac{Y+5}{2}$$



a/

b/

ii)  $1P(\cancel{Y=j} | X=i) = \frac{1}{i}$   $1 \leq j \leq i$

$$1E(Y | X=i) = \frac{1}{i} \sum_{j=1}^i j = \frac{i+1}{2}$$

2/19

$$1E(Y | X) = \frac{X+1}{2}$$

iii)  $1E(X) = 1E[1E(X|Y)] = \frac{1E(Y)}{2} + 1/2$

$$1E(Y) = 1E(1E(Y|X)) = \frac{1E(X)}{2} + 1/2$$

iv) Solving the system  $\Rightarrow \begin{cases} 1E(X) = \frac{11}{3} \\ 1E(Y) = \frac{7}{3} \end{cases}$

c). By analogy:

$$1P(X=i) = \frac{2i}{n(n+1)}$$

$$1P(X=j) = \frac{2(n+1-j)}{n(n+1)}$$

$$1P(X=i | Y=j) = \frac{1}{n+1-j} \quad 1 \leq j \leq i \leq n$$

$$1E(X|Y) = \frac{Y}{2} + \frac{n}{2}; \quad 1E(Y|X) = \frac{X}{2} + \frac{1}{2}$$

$$\Rightarrow 1E(X) = \frac{1E(Y)}{2} + \frac{n}{2}, \quad 1E(Y) = \frac{1E(X)}{2} + \frac{1}{2} \Rightarrow \begin{cases} 1E(X) = \frac{2n+1}{3} \\ 1E(Y) = \frac{n+2}{3} \end{cases}$$

2/

a)

$$i) \quad X_i \sim \text{Geo}\left(1 - \frac{i-1}{n}\right).$$

$$E(X_i) = \frac{n}{n-i+1}.$$

Detailed derivation  
provided in solutions  
of Coursework I.

$$ii) \quad E(Z) = E\left(\sum_{i=1}^n X_i\right) = \sum_i E(X_i) = n \sum_{i=1}^n \frac{1}{n-i+1} \\ = n \sum_{i=1}^n \frac{1}{i}$$

$$\text{for } n \text{ large} \quad \sim n \log(n).$$

b)

$$P(Y_{k,j} = 1) = P(j \text{ does not appear in draw } 1 \dots k) \\ = P(j \text{ does not appear in draw } 1)^k \\ = \left(\frac{n-1}{n}\right)^k.$$

$$E(Y_{k,j}) = \left(\frac{n-1}{n}\right)^k$$

$$E(Y_{k,i} Y_{k,j}) = \left(\frac{n-2}{n}\right)^k \leftarrow \# \text{ draws.}$$

$\nearrow$   
 neither  $i$  or  $j$   
 appear in a draw

for

$$P(Y_{k,i}, Y_{k,j} = 1) = P(Y_{k,i} = 1 \& Y_{k,j} = 1).$$

(4/19)

2/

b/

ii).

$$\text{Cov}(Y_{k,i}, Y_{k,j}) = \left(1 - \frac{2}{n}\right)^k - \left(1 - \frac{1}{n}\right)^{2k} < 0.$$

$Y_{k,i}$  &  $Y_{k,j}$  are negatively correlated. This is not surprising as if ball  $j$  is not chosen then it is more likely for that ball  $i$  will be chosen.

Not required:

However the  $\text{Var}(U_n)$  to be defined in Q1 is smaller than the variance of the sum of  $n$  independent ident. dis. Bernoulli r.v.

c/

i/

$$\mathbb{E}(U_k) = n \left(1 - \frac{1}{n}\right)^k$$

$$\text{Var}(U_k) = n \mathbb{E}(Y_k) [1 - \mathbb{E}(Y_k)] + \sum_j \sum_{k \neq j} \text{Cov}(Y_j, Y_k).$$

$$\leq n \mathbb{E}(Y_k) (1 - \mathbb{E}(Y_k))$$

$$\leq n \mathbb{E}(Y_k) = \mathbb{E}(U_k).$$

$$ii) \quad \frac{E(U_h)}{h} = \left(1 - \frac{1}{n}\right)^{cn} \xrightarrow{n \rightarrow \infty} e^{-c}. \quad (5/19)$$

$$iii) \quad IP\left(\left|\frac{U_h}{h} - E(U_h)\right| \geq n\epsilon/2\right) \leq 4 \frac{\text{var}(U_h)}{n^2 \epsilon^2} \leq 4 \frac{E(U_h)}{h^2 \epsilon^2} \xrightarrow{n \rightarrow \infty} 0.$$

This means  $\lim_{n \rightarrow \infty} IP\left(\left|\frac{U_h}{h} - e^{-c}\right| \geq \epsilon\right) = 0.$

The above limit shows that it is not enough to draw  $cn$  times to have a reasonable chance of seeing all  $n$  balls. In fact, with  $cn$  draws, there is a high probability that  $ne^{-c}$  balls will not be seen.

NOT REQUIRED  
In fact, we can see from the above that if we let  $h = \alpha n \log(n)$

$$\left. \begin{array}{l} \lim_{n \rightarrow \infty} IP(U_h = 0) = 1 \\ \lim_{n \rightarrow \infty} IP(U_h = n) = 0. \end{array} \right\} \begin{array}{l} \text{if } \alpha > 1 \\ \alpha < 1 \end{array}$$

3/

(6/19)

$$a / \quad i) \quad \phi_T(z) = E(z^T) = \sum_{k=0}^{\infty} (1-a) a^k z^k$$

$$= \frac{(1-a)}{1-az} \quad |z| < 1$$

$$E(T) = \phi'_T(1) = \frac{a}{1-a}.$$

$$E(T(T-1)) = \phi''(1) = 2 \left( \frac{a}{1-a} \right)^2$$

$$\text{Var}(T) = E(T(T-1)) + E(T) - [E(T)]^2$$

$$= \frac{a}{(1-a)^2}$$

ii/

$$P(T \geq n) = \sum_{k=n}^{\infty} (1-a) a^k = a^n.$$

$$P(T \geq n+k | T \geq n) = \frac{P(T \geq n+k, T \geq n)}{P(T \geq n)}$$

$$= \frac{P(T \geq n+k)}{P(T \geq n)}$$

$$= \frac{a^{n+k}}{a^n} = a^k.$$

$$= P(T \geq k).$$



3/  
/ii/ Theorem property of the geometric distribution

b/  $\phi_{S_n}(z) = (\phi_T(z))^n$  by ind property

$$E(S_n) = n E(T)$$

$$\text{Var}(S_n) = n \text{Var}(T).$$

$$(7/19)$$

c/ i/  $P(S_V = k) = \sum_{n \geq 1} P(V=n) P(S_1 = k | V=n)$

$$= \cancel{P(V=0)} \sum_{n \geq 1} P(V=n) P(S_n = k)$$

ii/  $E(S_V) = \sum_{k \geq 0} k P(S_V = k)$

$$= \sum_{k \geq 0} k \sum_{n \geq 1} P(V=n) P(S_n = k)$$

$$= \sum_{n \geq 1} P(V=n) \sum_{k \geq 0} k P(S_n = k)$$

$$= \sum_{n \geq 0} P(V=n) n E(T)$$

$$= E(T) E(V).$$

$$\text{ii/ } \text{Var}(S_V) = \mathbb{E}(S_V^2) - \mathbb{E}(S_V)^2$$

$$\mathbb{E}(S_n^2) = \overset{\text{from above}}{n^2 \mathbb{E}(T)^2} + n \text{Var}(T).$$

$$\text{Var}(S_V) = \text{Var}(V) \mathbb{E}(T)^2 + \mathbb{E}(V) \text{Var}(T).$$

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$$\begin{aligned} \text{iii) } \mathbb{E}(2^{S_V}) &= \mathbb{E}_{\mathcal{V}}(\phi_T(2)) \\ &= \sum_{n \geq 1} \mathbb{E}(2^{S_n} | V=n) \mathbb{P}(V=n) \\ &= \sum_{n \geq 1} \phi_T(2)^n \mathbb{P}(V=n) \\ &= \mathbb{E}(\phi_T(2)^V) = \phi_V(\phi_T(2)). \end{aligned}$$

$$= \phi_V\left(\frac{1-a}{1-a^2}\right)$$

$$\& \phi_V(z) = \frac{(1-b)z}{1-bz}$$

$$\mathbb{E}(2^{S_V}) = \frac{(1-b) \frac{1-a}{1-a^2}}{1-b \frac{1-a}{1-a^2}} = \frac{(1-b)(1-a)}{1-a^2 - b + ab}$$

$$= \frac{1-a}{1-a^2} = 1-\alpha$$

$$\alpha = \frac{a}{1-b(1-a)}.$$

3/  
c/  
iv)

$S_U$  is a geometric distribution

$$P(S_U \geq k) = \alpha^k$$

do T.

9/19

4/

$$X_i = \begin{cases} +1 & \text{w.p. } 1-p \\ -1 & \text{w.p. } p \end{cases}$$

 $p \in (0, 1).$ 

$$Z_n = X_1 \dots X_n$$

$$Z_0 = 1.$$

a) i)  $IP(Z_n = y \mid Z_{n-1} = x)$

$$\begin{matrix} & -1 & 1 \\ -1 & 1-p & p \\ 1 & p & 1-p \end{matrix} = Q$$

$$\begin{pmatrix} 1-p \\ p \end{pmatrix}$$

ii)  $IP(Z_{n+1} = y \mid Z_0 = z_0, \dots, Z_n = z_n)$

$$= IP(X_{n+1} = x \mid Z_0, \dots, Z_n = z_n)$$

$$= IP(X_{t+1} = x) = p \text{ or } 1-p \text{ independently of } t.$$

$Z_n$  is homogeneous.

with transition matrix  $Q$ .

iii) if the above  $Z_n$  are independent

$$\text{if } p = 1-p = 1/2.$$

$$\text{if } p = 1/2 \Rightarrow IP(Z_n = z_n, \dots, Z_0 = 1) = \left(\frac{1}{2}\right)^n$$

$\Rightarrow Z_i$  are independent.

4/

b/

i/

$$IP(Z_{n+k}=y \mid Z_n=x) = (Q^k)_{xy}$$

$$\det Q^k = (1-p_k)^2 - p_k^2 = 1-2p_k$$

$$\frac{11}{19}$$

$$\det Q^k = [\det(Q)]^k = (1-2p)^k$$

$$\Rightarrow 1-2p_k = (1-2p)^k$$

Go to pre induction.

p/

$$ii/ \lim_{n \rightarrow \infty} IP(Z_n = \text{?}) = ?$$

$$\begin{aligned} \lim_{n \rightarrow \infty} IP(Z_n = -1) &= \lim_{n \rightarrow \infty} p_k = 1/2 \\ &= \lim_{n \rightarrow \infty} IP(Z_n = +1). \end{aligned}$$

~~Alternatively stationary dist~~  
 ~~$\pi_1(1-p) + \pi_2 p = \pi_1$~~   
 ~~$\pi_1(1-p) = \pi_2$~~   
 ~~$\pi_1 = \pi_2 = 1/2$~~



4/

b/ ii/

1<sup>st</sup> method:

Using the stationary distribution;

$$\pi p = \pi$$

$$\left\{ \begin{array}{l} \pi_{-1}(1-p) + \pi_1 p = \pi_{-1} \\ \pi_{-1} p + \pi_1(1-p) = \pi_1 \end{array} \right.$$

$$\Rightarrow$$

$$\pi_{-1} = \pi_1 = 1/2$$

$$\Rightarrow \lim_{n \rightarrow \infty} P(Z_n = \pm 1) = 1/2.$$

$$\text{iii/} \quad \lim_{n \rightarrow \infty} P(Z_{n+1} = x, Z_n = y)$$

$$\Rightarrow \lim_{n \rightarrow \infty} P(Z_{n+1} = x | Z_n = y) P(Z_n = y)$$

		y	
		-1	1
x	-1	$1/2 (1-p)$	$1/2 p$
	1	<del><math>1/2 p</math></del>	$1/2 (1-p)$

$$\frac{1/2}{1/2}$$

S/  
a/  
i/

$$f_U(u) = \begin{cases} 1 & u \in [0, 1) \\ 0 & \text{otherwise} \end{cases}$$

$$E(U) = \int x f_U(u) = \int_0^1 u \cdot 1 \, du = 1/2.$$

$$E(U^2) = \int_0^1 u^2 \cdot 1 \, du = 1/3.$$

$$\text{Var}(U) = 1/3 - 1/4 = 1/12.$$

$$\frac{1^3/12}{1/2}$$

ii/.  $IP(U \leq u) = \int_0^u 1 \, du = u.$

b/  $IP(A_n^c) = \int \cancel{IP(U_n \leq \frac{f(Y_n)}{cg(Y_n)})} \, dP =$

$$= \int_{\mathbb{R}} IP(U_n \leq \frac{f(y)}{cg(y)}) g(y) \, dy$$

$$= \int_{\mathbb{R}} \frac{f(y)}{cg(y)} g(y) \, dy$$

$f(y) \leq 1$   
 $0 \leq \frac{f(y)}{cg(y)}$   
 by assumption.

$$= \int_{\mathbb{R}} f(y) \cdot \frac{1}{c} \, dy = 1/c$$

5/

b/

ii/

By previous question.

$$P(A_n^c, Y_n \leq n) = \int_{-\infty}^n \frac{1}{c} f(y) dy$$

iii/

$$P(A_1 - A_{n-1}, A_n^c, Y_n \leq n)$$

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$$= P(A_1) \dots P(A_{n-1}) P(A_n^c, Y_n \leq n)$$

by independence

$$= \left(1 - \frac{1}{c}\right)^{n-1} \frac{1}{c} \int_{-\infty}^n \frac{1}{c} f(y) dy$$

5/

c/

$$\tau = \min \{n \geq 1 ; U_n \leq \frac{f(Y_n)}{c f(Y_n)}\}$$

i)

$$P(Z \leq n) = P(\tau \leq n)$$

$$= \sum_{n \geq 1} P(\tau = n ; Y_n \leq n)$$

$$= \sum_{n \geq 1} P(A_1 - A_{n-1}, A_n^c, Y_n \leq n)$$

ii).

$$P(Z \leq n) = \sum_{n \geq 1} \left(1 - \frac{1}{c}\right)^{n-1} \frac{1}{c} \int_{-\infty}^n f(y) dy = \int_{-\infty}^n f(y) dy$$

6)  
a)

if  $(\pi_i)$  is such that

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$$\pi_i q_{ij} = \pi_j q_{ji}$$

then

$$(\pi Q)_i = \sum_j \pi_j q_{ij} = \sum_j \pi_i q_{ij}$$

$$= \pi_i \sum_j q_{ij}$$

$$= 0$$

since  $Q$  is  
a rate matrix.

b)

i)

~~$$q_{ij} = \dots$$~~

$$Q = \begin{pmatrix} -(\lambda + \mu) & \lambda & 0 \\ \mu & -(\lambda + \mu) & \lambda \\ 0 & \mu & -(\lambda + \mu) \end{pmatrix}$$

ii)

$$\pi_i q_{ij} = \pi_j q_{ji}$$

$$\pi_i q_{i,i+1} = \pi_{i+1} q_{i+1,i}$$

$$\pi_{i+1} = \pi_i \frac{\lambda}{\mu}$$

$$\Rightarrow \pi_i = \pi_0 \left( \frac{\lambda}{\mu} \right)^i$$

$$\Rightarrow \sum \pi_i = 1$$

if  $\rho = \frac{\lambda}{\mu} < 1 \Rightarrow$

$$\pi_i = \frac{\rho^i (1 - \rho)}{1 - \rho^{i+1}}$$

6)

$\lambda < \mu$

- b) ii)  $\rho < 1$  the rate of arrival  $\lambda$  is smaller than rate of departure  $\mu$  hence the queue is stable  
if  $\rho > 1$  then the queue will build up until becoming infinitely large.

iii)  $IE(N_1) = \sum_i i \pi_i = \sum_{i=0}^{\infty} i \rho^i (1-\rho)$   
 $= \frac{\rho}{1-\rho} = \frac{\lambda}{\mu-\lambda}$

c) i)  $IE(N_2) = 2 IE(N_1) = \frac{2\lambda}{\mu-\lambda}$

ii).  $M/M/2$  queue.

$$\begin{bmatrix} -\lambda & \lambda & 0 & 0 & \dots \\ \mu & -(\lambda+\mu) & \lambda & 0 & \dots \\ 0 & 2\mu & -(\lambda+2\mu) & \lambda & \dots \\ 0 & 0 & 2\mu & -(\lambda+2\mu) & \dots \\ \vdots & \vdots & \vdots & \vdots & \ddots \end{bmatrix}$$



b) c).

ii)

~~q~~

$(\lambda + \mu)$

$$\text{if } \lambda < \mu \Rightarrow \lambda < 2\mu \Rightarrow \lambda < 2\mu$$

if the initial system is stable  $\rho$   
is the  $n/n/2$

~~in fact~~, ~~probability~~

iii)

by

$$\pi_i q_{ij} = \pi_j q_{ji} ; s=2$$

$\Rightarrow$

$$\pi_i = \begin{cases} e^n/n! \pi_0 & n=0,1,2 \\ \frac{e^n}{s! s^{n-s}} \pi_0 & \forall n \geq s. \end{cases}$$

$\Rightarrow$

$$\pi_0 = \frac{1}{2} \left( \mu/\lambda \right)^2 \pi_2$$

$$\pi_1 = \left( \mu/\lambda \right) \pi_2$$

$$\pi_n = \left( \lambda/\mu \right)^{n-2} \pi_2 \quad n \geq 2.$$

$$\text{with } \sum_i \pi_i = 1 \Rightarrow \pi_2 = \frac{2\lambda^2(\mu-\lambda)}{\mu^2/(\mu+\lambda)}$$

6)

c) iii)  $E(N_2) = \sum_i i \pi_i = \pi_1 + \pi_2 \sum_{n \geq 2} i \left(\frac{\lambda}{\mu}\right)^{i-2}$

$$= \frac{\lambda \mu \lambda}{(\mu + \lambda)(\mu - \lambda)}$$

(18)  
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iv)  
It is not difficult to see that

Since  $\frac{\lambda \lambda}{\mu + \lambda} < 1 \Rightarrow E(N_2) < E(N_1)$

From the post office view point, the place is less crowded with the suggested strategy.

NOT REQUIRED

One could also look at the customer's view point; it is not difficult to show that the cv. waiting time in strategy 1 is  $E(W_1) = \frac{\lambda}{\mu(\mu - \lambda)}$  & for the 2nd strategy is  $E(W_2) = \frac{\lambda^2}{\mu(\mu + \lambda)(\mu - \lambda)} < E(W_1)$  with a similar conclusion.

(19/19)

Finally; if we look at activity rate of employers in the two systems.

Then  $\rho_1 = 1 - \pi_0 = \frac{\lambda}{\mu}$ . 1<sup>st</sup> system.

$$\begin{aligned}\rho_2 &= 1 - (\pi_0 - \pi_1/2) && 2^{nd} \text{ system} \\ &= \frac{\lambda}{\mu} = \rho_1\end{aligned}$$

In the case, strategy 2 does not improve on strategy 1.