



Choose only 5 questions from the following:

Q1: [4+4]

(a) For the Markov process $\{X_t\}$, $t=0,1,2,\dots,n$ with states $i_0, i_1, i_2, \dots, i_{n-1}, i_n$

Prove that: $\Pr\{X_0 = i_0, X_1 = i_1, X_2 = i_2, \dots, X_n = i_n\} = p_{i_0} P_{i_0 i_1} P_{i_1 i_2} \dots P_{i_{n-1} i_n}$ where $p_{i_0} = \Pr\{X_0 = i_0\}$

(b) A Markov chain X_0, X_1, X_2, \dots has the transition probability matrix

$$\mathbf{P} = \begin{matrix} & \begin{matrix} 0 & 1 & 2 \end{matrix} \\ \begin{matrix} 0 \\ 1 \\ 2 \end{matrix} & \left\| \begin{array}{ccc} 0.7 & 0.2 & 0.1 \\ 0 & 0.6 & 0.4 \\ 0.5 & 0 & 0.5 \end{array} \right\| \end{matrix}$$

Find $\Pr\{X_1 = 1, X_2 = 1 | X_0 = 0\}$.

Q2: [4+4]

Consider the Markov chain whose transition probability matrix is given by

$$\mathbf{P} = \begin{matrix} & \begin{matrix} 0 & 1 & 2 & 3 \end{matrix} \\ \begin{matrix} 0 \\ 1 \\ 2 \\ 3 \end{matrix} & \left\| \begin{array}{cccc} 1 & 0 & 0 & 0 \\ 0.1 & 0.6 & 0.1 & 0.2 \\ 0.2 & 0.3 & 0.4 & 0.1 \\ 0 & 0 & 0 & 1 \end{array} \right\| \end{matrix}$$

(a) Starting in state 1, determine the probability that the Markov chain ends in state 0.

(b) Determine the mean time to absorption.

Q3: [5+3]

(a) A Markov chain X_0, X_1, X_2, \dots has the transition probability matrix

$$\mathbf{P} = \begin{matrix} & \begin{matrix} 0 & 1 & 2 \end{matrix} \\ \begin{matrix} 0 \\ 1 \\ 2 \end{matrix} & \begin{vmatrix} 0.3 & 0.2 & 0.5 \\ 0.5 & 0.1 & 0.4 \\ 0.5 & 0.2 & 0.3 \end{vmatrix} \end{matrix}$$

Every period that the process spends in state 0 incurs a cost \$2. Every period that the process spends in state 1 incurs a cost of \$5. Every period that the process spends in state 2 incurs a cost of \$3. What is the long run cost per period associated with this Markov chain?

(b) Let $X(t)$ be a Yule process that is observed at a random time U , where U is uniformly distributed over $[0,1)$. Show that $pr\{X(U) = k\} = p^k / (\beta k)$ for $k = 1, 2, \dots$, with $p = 1 - e^{-\beta}$.

Q4: [4+4]

(a) Let $\{X_n\}$ be a Markov chain with state space $S = \{1, 2\}$ has the transition probability matrix $\mathbf{P} = \begin{vmatrix} 0.5 & 0.5 \\ 1 & 0 \end{vmatrix}$, find $pr\{X_5 = 2 | X_2 = 1\}$.

(b) The probability of the thrower winning in the dice game is $p = 0.4929$. Suppose player A is the thrower and begins the game with \$5, and player B, his opponent, begins with \$10. What is the probability that player A goes bankrupt before player B? Assume that the bet is \$1 per round.

Q5: [8]

Suppose that the weather on any day depends on the weather conditions for the previous 2 days. Suppose also that if it was sunny today and yesterday, then it will be sunny tomorrow with probability 0.8; if it was sunny today but cloudy yesterday, then it will be sunny tomorrow with probability 0.6; if it was cloudy today but sunny yesterday, then it will be sunny tomorrow with probability 0.4; if it was cloudy for the last 2 days, then it will be sunny tomorrow with probability 0.1. Transform this model into a Markov chain, and then find the transition probability matrix. Find also the long run fraction of days in which it is sunny.

Q6: [4+4]

(a) Using the differential equations

$$\frac{dp_0(t)}{dt} = -\lambda p_0(t) \quad (1)$$

$$\frac{dp_n(t)}{dt} = \lambda p_{n-1}(t) - \lambda p_n(t), \quad n=1,2,3, \dots \quad (2)$$

where all birth parameters are the same constant λ with initial condition $X(0)=0$,

Show that $p_n(t) = \frac{(\lambda t)^n e^{-\lambda t}}{n!}$, $n=0,1,2,\dots$

(b) Suppose that customers arrive at a facility according to a Poisson process having rate $\lambda=2$. Let $X(t)$ be the number of customers that have arrived up to time t . Determine the following conditional probabilities

$pr\{X(3)=6|X(1)=2\}$ and $pr\{X(1)=2|X(3)=6\}$.

Model Answer For

M 380 - Stochastic Processes

$$\underline{Q1}$$

$$(a) \text{ pr } \{X_0 = i_0, X_1 = i_1, X_2 = i_2, \dots, X_{n-1} = i_{n-1}, X_n = i_n\}$$

$$= \text{pr} \{X_0 = i_0, X_1 = i_1, X_2 = i_2, \dots, X_{n-1} = i_{n-1}\}$$

$$\cdot \text{pr} \{X_n = i_n | X_0 = i_0, X_1 = i_1, \dots, X_{n-1} = i_{n-1}\}$$

$$\text{pr}(x, y) = \text{pr}(x|y) \text{pr}(y)$$

$$= \text{pr} \{X_0 = i_0, X_1 = i_1, \dots, X_{n-1} = i_{n-1}\} \cdot \text{pr} \{X_n = i_n | X_{n-1} = i_{n-1}\}$$

(joint prob
by using Markov prop.)

$$= \text{pr} \{X_0 = i_0, X_1 = i_1, \dots, X_{n-1} = i_{n-1}\} P_{i_{n-1} i_n}$$

$$\text{where } P_{ij} = \text{pr} \{X_{n+1} = j | X_n = i\}$$

By repeating this argument $n-1$ times, we obtain

$$\text{pr} \{X_0 = i_0, X_1 = i_1, \dots, X_{n-1} = i_{n-1}, X_n = i_n\}$$

$$= P_{i_0 i_1} P_{i_1 i_2} \dots P_{i_{n-1} i_n} \text{ where } P_{i_0 i_1} = \text{pr} \{X_1 = i_1 | X_0 = i_0\}$$

$$b) \text{ pr} \{X_1 = 1, X_2 = 1 | X_0 = 0\}$$

$$= \text{pr} \{X_2 = 1 | X_1 = 1, X_0 = 0\} \cdot \text{pr} \{X_1 = 1 | X_0 = 0\}$$

$$= \text{pr} \{X_2 = 1 | X_1 = 1\} \cdot \text{pr} \{X_1 = 1 | X_0 = 0\} \text{ Markov prop}$$

$$= P_{11} P_{01} = 0.6(0.2) = \boxed{0.12}$$

#

Q2

$$P = \begin{array}{c|cccc} & 0 & 1 & 2 & 3 \\ \hline 0 & 1 & 0 & 0 & 0 \\ 1 & 0.1 & 0.6 & 0.1 & 0.2 \\ 2 & 0.2 & 0.3 & 0.4 & 0.1 \\ 3 & 0 & 0 & 0 & 1 \\ \hline \end{array}$$

0 and 3 are 2 absorbing states
but 1 and 2 are not abs.

$$\Rightarrow u_i = \text{pr} \{X_T = 0 \mid X_0 = i\}, i=1,2$$

الحالة $X=1$
 $u_1 = \text{pr} \{X_T = 0 \mid X_0 = 1\}$
 $= P_{10} + P_{11} u_1 + P_{12} u_2$
 $u_1 = 0.1 + 0.6 u_1 + 0.1 u_2$
 $\Rightarrow 0.4 u_1 - 0.1 u_2 = 0.1$ ①

الحالة $X=2$
 $u_2 = \text{pr} \{X_T = 0 \mid X_0 = 2\}$
 $= P_{20} + P_{21} u_1 + P_{22} u_2$
 $= 0.2 + 0.3 u_1 + 0.4 u_2$
 $\Rightarrow 0.3 u_1 - 0.6 u_2 = -0.2$ ②

Solving ①, ②, $6 \times ① - ② \Rightarrow$

$$\therefore u_1 = u_{10} = \frac{8}{21}$$

$$u_i = E[T \mid X_0 = i]$$

$i=1,2$

$$u_1 = 1 + P_{11} u_1 + P_{12} u_2$$

$$u_1 = 1 + 0.6 u_1 + 0.1 u_2$$

$$\Rightarrow 0.4 u_1 - 0.1 u_2 = 1$$
 ①

$$u_2 = 1 + P_{21} u_1 + P_{22} u_2$$

$$u_2 = 1 + 0.3 u_1 + 0.4 u_2$$

$$\Rightarrow 0.3 u_1 - 0.6 u_2 = -1$$
 ②

Solving ①, ②, $① \times 6 - ② \Rightarrow$

$$u_1 = \frac{10}{3}$$

(a)
$$P = \begin{array}{c|ccc} & 0 & 1 & 2 \\ \hline 0 & 0.3 & 0.2 & 0.5 \\ \hline 1 & 0.5 & 0.1 & 0.4 \\ \hline 2 & 0.5 & 0.2 & 0.3 \end{array}$$

process X_0, X_1, X_2
 $C_0 = \$2$ $C_1 = \$5$ $C_2 = \$3$

long run mean cost per unit period

$$= \sum_{j=0}^2 \pi_j C_j$$

$$= \pi_0 C_0 + \pi_1 C_1 + \pi_2 C_2 \quad \text{①}$$

Also, we have

$$\begin{cases} \pi_0 = 0.3\pi_0 + 0.5\pi_1 + 0.5\pi_2 \\ \pi_1 = 0.2\pi_0 + 0.1\pi_1 + 0.2\pi_2 \\ \pi_0 + \pi_1 + \pi_2 = 1 \end{cases}$$

$$\Rightarrow \begin{cases} 7\pi_0 - 5\pi_1 - 5\pi_2 = 0 & \text{①} \\ 2\pi_0 - 9\pi_1 + 2\pi_2 = 0 & \text{②} \\ \pi_0 + \pi_1 + \pi_2 = 0 & \text{③} \end{cases}$$

Solving ①, ② and ③ by using

Cramer's rule

$$\Delta = \begin{vmatrix} 7 & -5 & -5 \\ 2 & -9 & 2 \\ 1 & 1 & 1 \end{vmatrix} = -132$$

$$\Delta_1 = -55, \Delta_2 = -24, \Delta_3 = -53$$

\Rightarrow

$$\pi_0 = \frac{\Delta_1}{\Delta} = \frac{5}{12}$$

$$\pi_1 = \frac{\Delta_2}{\Delta} = \frac{2}{11}$$

$$\pi_2 = \frac{\Delta_3}{\Delta} = \frac{53}{132}$$

②

Subs. ② in ①

long run mean cost per unit period
 $= 308/132$
 ≈ 2.95

(b) For Yuh process

$$P_n(t) = e^{-\beta t} (1 - e^{-\beta t})^{n-1}, n \geq 1$$

$$Pr\{X(U) = k\} = \int_0^1 e^{-\beta u} (1 - e^{-\beta u})^{k-1} du$$

, $k = 1, 2, \dots$

$$= \frac{1}{\beta} \int_0^1 (1 - e^{-\beta u})^{k-1} \beta e^{-\beta u} du$$

$$= \frac{1}{\beta} \left[\frac{(1 - e^{-\beta u})^k}{k} \right]_0^1$$

$$= \frac{1}{\beta k} (1 - e^{-\beta})^k$$

$$= \frac{p^k}{\beta k}, k = 1, 2, \dots$$

where $p = 1 - e^{-\beta}$

#

Q4 (a) P_{12}^3 ?

$$P^2 = \begin{pmatrix} \frac{1}{2} & \frac{1}{2} \\ 1 & 0 \end{pmatrix} \begin{pmatrix} \frac{1}{2} & \frac{1}{2} \\ 1 & 0 \end{pmatrix} = \begin{pmatrix} \frac{3}{4} & \frac{1}{4} \\ \frac{1}{2} & \frac{1}{2} \end{pmatrix}$$

$$P^3 = \begin{pmatrix} \frac{3}{4} & \frac{1}{4} \\ \frac{1}{2} & \frac{1}{2} \end{pmatrix} \begin{pmatrix} \frac{1}{2} & \frac{1}{2} \\ 1 & 0 \end{pmatrix}$$

$$P_{12}^3 = \frac{3}{8}$$

(b) $i = \$5$ fortune for player A

$$N = \$5 + \$10 = \$15$$

$$p = 0.4929 \Rightarrow q = 1 - p = 0.5071$$

$$u_i = \Pr \{ X_n \text{ reaches state 0 before state } N \mid X_0 = i \}$$

$$u_i = \frac{(q/p)^i - (q/p)^N}{1 - (q/p)^N} \quad , p \neq q$$

$$u_i = \frac{\left[\left(\frac{0.5071}{0.4929} \right)^5 - \left(\frac{0.5071}{0.4929} \right)^{15} \right]}{1 - \left(\frac{0.5071}{0.4929} \right)^{15}}$$

$$\therefore u_i \approx 0.71273$$

#

Q5

Weather states = $\{(S, S), (S, C), (C, S), (C, C)\}$

S \rightarrow Sunny \rightarrow C \rightarrow cloudy

\Rightarrow The transition prob. π_{ij} is given by

	(S, S)	(S, C)	(C, S)	(C, C)
(S, S)	0.8	0.2	0	0
(S, C)	0	0	0.4	0.6
(C, S)	0.6	0.4	0	0
(C, C)	0	0	0.1	0.9
	\downarrow π_0	\downarrow π_1	\downarrow π_2	\downarrow π_3

For long run the limiting distn is

$$\pi = (\pi_0, \pi_1, \pi_2, \pi_3)$$

$$0.8\pi_0 + 0.6\pi_2 = \pi_0 \Rightarrow \pi_2 = \frac{1}{5}\pi_0 \quad \left. \begin{array}{l} \pi_1 = \pi_2 = \frac{1}{5}\pi_0 \\ \pi_3 = 6\pi_1 = 2\pi_0 \end{array} \right\}$$

$$0.2\pi_0 + 0.4\pi_2 = \pi_1 \Rightarrow \pi_1 = \frac{1}{5}\pi_0$$

$$0.4\pi_1 + 0.1\pi_3 = \pi_2 \Rightarrow \pi_3 = 6\pi_1 = 2\pi_0$$

$$\sum \pi_i = \pi_0 + \pi_1 + \pi_2 + \pi_3 = 1 \Rightarrow (1 + \frac{1}{5} + \frac{1}{5} + 2)\pi_0 = 1$$

$$\therefore \pi_0 = \frac{3}{11}, \pi_1 = \pi_2 = \frac{1}{11}, \pi_3 = \frac{6}{11}$$

* the long run fraction of days in which it is sunny

$$= \pi_0 + \pi_1 = \frac{4}{11}$$

Q6

-6-

(a) $X(t)$ represents the size of the population
 $X(0) = 0$ is the initial condition

$$\Rightarrow P_n(t) = \begin{cases} 1 & , n=0 \\ 0 & , \text{otherwise} \end{cases}$$

$$(1) \Rightarrow \frac{dP_0(t)}{dt} = -\lambda P_0(t)$$

$$\int_0^t \frac{dP_0(u)}{P_0(u)} = -\lambda \int_0^t du$$

$$\therefore [\ln P_0(u)]_0^t = -\lambda [u]_0^t$$

$$\ln P_0(t) - \ln P_0(0) = -\lambda t$$

$$P_0(0) = 1 \quad \text{initial condition} \Rightarrow \ln(1) = 0$$

$$\therefore \ln P_0(t) = -\lambda t \Rightarrow \boxed{P_0(t) = e^{-\lambda t}} \quad (3)$$

$$(2) \Rightarrow \frac{dP_n(t)}{dt} = \lambda P_{n-1}(t) - \lambda P_n(t)$$

$$\therefore \frac{dP_n(t)}{dt} + \lambda P_n(t) = \lambda P_{n-1}(t), \quad n=1, 2, \dots$$

Multiply both sides by $e^{\lambda t}$

$$e^{\lambda t} \left[\frac{dP_n(t)}{dt} + \lambda P_n(t) \right] = e^{\lambda t} [\lambda P_{n-1}(t)]$$

$$\therefore \frac{d}{dt} [P_n(t) e^{\lambda t}] = \lambda P_{n-1}(t) e^{\lambda t}$$

$$d[P_n(t) e^{\lambda t}] = \lambda P_{n-1}(t) e^{\lambda t} dt$$

$$\Rightarrow \int_0^t d[P_n(u) e^{\lambda u}] = \lambda \int_0^t P_{n-1}(u) e^{\lambda u} du$$

$$\therefore [P_n(x) e^{\lambda x}]_0^t = \lambda \int_0^t P_{n-1}(x) e^{\lambda x} dx$$

$$P_n(t) e^{\lambda t} - P_n(0) = \lambda \int_0^t P_{n-1}(x) e^{\lambda x} dx$$

$$\therefore P_n(t) = e^{-\lambda t} \left[\lambda \int_0^t P_{n-1}(x) e^{\lambda x} dx \right], n=1, 2, \dots$$

$$P_n(t) = \lambda e^{-\lambda t} \int_0^t P_{n-1}(x) e^{\lambda x} dx \quad (4), n=1, 2, \dots$$

which is a recurrence relation

$$\begin{aligned} \text{at } n=1 \quad (4) \Rightarrow P_1(t) &= \lambda e^{-\lambda t} \int_0^t P_0(x) e^{\lambda x} dx \\ (3) \Rightarrow &= \lambda e^{-\lambda t} \int_0^t e^{-\lambda x} e^{\lambda x} dx \\ &= \lambda e^{-\lambda t} \int_0^t dx \end{aligned}$$

$$P_1(t) = \lambda t e^{-\lambda t} \quad (5)$$

$$\begin{aligned} \text{at } n=2 \quad (4) \Rightarrow P_2(t) &= \lambda e^{-\lambda t} \int_0^t P_1(x) e^{\lambda x} dx \\ (5) \Rightarrow &= \lambda e^{-\lambda t} \int_0^t \lambda x e^{-\lambda x} e^{\lambda x} dx \\ P_2(t) &= \lambda^2 e^{-\lambda t} \int_0^t x dx = \frac{1}{2} \lambda^2 e^{-\lambda t} t^2 \end{aligned}$$

$$\therefore P_2(t) = \frac{1}{2} (\lambda t)^2 e^{-\lambda t} \dots (6)$$

Similarly, $P_n(t) = \frac{(\lambda t)^n}{n!} e^{-\lambda t}, n=0, 1, 2, \dots$
which is called Poisson process #

$$(b) P\{X(3)=6 | X(0)=0\}$$

$$= P\{X(3)=6\}$$

$$= P\{X(3)-X(0)=6\}$$

$$= \frac{(2t)^k e^{-2t}}{k!}$$

$$= \frac{4^3 e^{-4}}{4!} = \frac{32e^{-4}}{3}$$

$$\rightarrow P\{X(0)=2 | X(3)=6\}$$

$$= \binom{n}{x} p^x q^{n-x}$$

$$= \binom{6}{2} \left(\frac{1}{3}\right)^2 \left(\frac{2}{3}\right)^4$$

$P\{X(3)=6 | X(0)=0\}$
 when $X=0$ and $t=3$ are
 independent



$$2t = 2(2) = 4$$

Binomial distn

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