
JUSTIFY YOUR ANSWERS

Allowed: material handed out in class and *handwritten* notes (*your handwriting*)

NOTE:

- The test consists of five questions plus one bonus problem.
 - The score is computed by adding all the credits up to a maximum of 10
-
-

Exercise 1. Let X_i , $1 = 1, \dots, n$ be independent normal random variables with respective means μ_i and variances σ_i^2 . Consider its mean

$$\bar{X} = \frac{1}{n} \sum_{i=1}^n X_i$$

- (a) (0.5 pts.) Prove that \bar{X} is also normally distributed.
- (b) (0.5 pts.) Determine the mean and variance of \bar{X} .

Answers: The moment-generating function $\Phi_{\bar{X}}(t)$ of \bar{X} factorizes, due to the independence of the X_i , in the following way:

$$\Phi_{\bar{X}}(t) = E(e^{t\bar{X}}) = \prod_{i=1}^n E(e^{tX_i/n}) = \prod_{i=1}^n \Phi_{X_i}(t/n)$$

where Φ_{X_i} is the moment-generator function of the variable X_i . As each X_i is normal,

$$\Phi_{X_i}(s) = \exp\left[\mu_i s + \frac{\sigma_i^2 s^2}{2}\right],$$

hence

$$\Phi_{\bar{X}}(t) = \exp\left[t\left(\frac{1}{n} \sum_{i=1}^n \mu_i\right) + \frac{t^2}{2}\left(\frac{1}{n^2} \sum_{i=1}^n \sigma_i^2\right)\right].$$

This is the moment-generating function of a normal variable with mean $\frac{1}{n} \sum_{i=1}^n \mu_i$ and variance $\frac{1}{n^2} \sum_{i=1}^n \sigma_i^2$. This identifies \bar{X} as a variable with such a law.

Exercise 2. (1 pt.) Consider a branching process with offspring number with mean μ and variance σ . That means, a sequence of random variables $(X_n)_{n \geq 0}$ with $X_0 = 1$ and

$$X_n = \sum_{i=1}^{X_{n-1}} Z_i \quad n \geq 1$$

where Z_n are iid random variables (offspring distribution) independent of the (X_n) with mean μ Show that $E(X_n) = \mu^n$. [Hint: Start by showing that $E(X_n) = \mu E(X_{n-1})$.]

Answer: Start with

$$E(X_n) = E(E(X_n | X_{n-1})).$$

Now

$$\begin{aligned}
 E(X_n | X_{n-1} = x_{n-1}) &= E\left(\sum_{n \geq 1}^{x_{n-1}} Z_i \mid X_{n-1} = x_{n-1}\right) \\
 &= \sum_{n \geq 1}^{x_{n-1}} E(Z_i | X_{n-1} = x_{n-1}) \\
 &= \sum_{n \geq 1}^{x_{n-1}} E(Z_i) \quad (\text{independence of } Z_i \text{ and } X_{n-1}) \\
 &= x_{n-1} \mu .
 \end{aligned}$$

Hence $E(X_n | X_{n-1}) = \mu X_{n-1}$ and

$$E(X_n) = E(\mu X_{n-1}) = \mu E(X_{n-1}) .$$

By induction in n we get the proposed result.

Exercise 3.

(a) (0.8 pts.) Show that

$$\begin{pmatrix} p & 1-p \\ 1-p & p \end{pmatrix}^n = \begin{pmatrix} 1/2 + a^n/2 & 1/2 - a^n/2 \\ 1/2 - a^n/2 & 1/2 + a^n/2 \end{pmatrix}$$

for $n \geq 1$. Determine a .

Answer: *This is an easy proof by induction. Comparing for the case $n = 1$ we obtain $a = 2p - 1$.*

(b) A communication system transmits the digits 0 and 1. Each digit must pass through n stages, each of which independently transmits the digit correctly with probability p .

-i- (0.8 pts.) Find the probability that the final digit, X_n , is correct.

Answer: $P_{00}^n = P_{11}^n = 1/2 - (2p - 1)^n/2$.

-ii- (0.8 pts.) Find the probability that *all* the first n stages transmit correctly.

Answer: *By independence the probability is equal to p^n .*

Exercise 4. Consider a three-state Markov process $(X_n)_{n \geq 0}$ with two absorbing states. That is, a process with a three-symbol alphabet (=state space), say $\{0, 1, 2\}$, and transition matrix

$$\mathbb{P} = \begin{pmatrix} 1 & 0 & 0 \\ a & b & c \\ 0 & 0 & 1 \end{pmatrix}$$

with $a, b, c > 0$ and $a + b + c = 1$.

(a) (0.8 pts.) Show that $\mathbb{P}_{11}^n = b^n$.

Answer: *As $P_{x1} = 0$ for every $x \neq 1$,*

$$P_{11}^n = \sum_{x=0}^2 P_{1x}^{n-1} P_{x1} = P_{11}^{n-1} P_{11} = P_{11}^{n-1} b .$$

The result follows by induction.

(b) (0.8 pts.) Show that the state “1” is transient.

Answer:

$$\sum_{n \geq 0} P_{11}^n = \sum_{n \geq 0} b^n = \frac{1}{1-b} < \infty.$$

(c) (0.8 pts.) Let $T = \inf\{n > 0 : X_n = 0 \text{ or } X_n = 2\}$ be the time it takes the process to be absorbed in one of the absorbing states. Compute $E(T | X_0 = 1)$. [Hint: you may want to use that for a discrete random variable Z , $E[Z] = \sum_{k \geq 0} P(Z > k)$.]

Answer:

$$\begin{aligned} E(T | X_0 = 1) &= \\ \sum_{k \geq 0} P(T > k | X_0 = 1) &= \sum_{k \geq 0} P(X_n = 1, n = 1, \dots, k | X_0 = 1) = \sum_{k \geq 0} b^k = \frac{1}{1-b}. \end{aligned}$$

(d) (0.8 pts.) Let $T_0 = \inf\{n > 0 : X_n = 0\}$ and $T_2 = \inf\{n > 0 : X_n = 2\}$ be the absorption times at each of the absorbing states. Compute $P(T_0 < T_2 | X_0 = 1)$.

Answer:

$$\begin{aligned} P(T_0 < T_2 | X_0 = 1) &= \\ &= \sum_{k \geq 0} P(X_{k+1} = 0, X_n = 1, n = 1, \dots, k | X_0 = 1) = \sum_{k \geq 0} P(X_{k+1} = 0 | X_k = 1) P_{11}^k \\ &= \sum_{k \geq 0} a b^k = \frac{a}{1-b}. \end{aligned}$$

(e) (0.8 pts.) Compute *all* the invariant measures of the process.

Answer: Let $\pi = (\pi_0, \pi_1, \pi_2)$ be the invariant measure. The conditions $\sum_x \pi_x P_{xy} = \pi_x$ plus the normalisation condition $\pi_0 + \pi_1 + \pi_2 = 1$ become:

$$\begin{aligned} \pi_0 + a \pi_1 &= \pi_0 \\ b \pi_1 &= \pi_1 \\ c \pi_1 + \pi_2 &= \pi_2 \\ \pi_0 + \pi_1 + \pi_2 &= 1. \end{aligned}$$

All their solutions are of the form $\pi_1 = 0$, $\pi_1 + \pi_2 = 1$. That is, the invariant measures Π take the form

$$\Pi = (\lambda, 0, 1 - \lambda) = \lambda(1, 0, 0) + (1 - \lambda)(0, 0, 1) \quad \text{for } 0 \leq \lambda \leq 1,$$

That is, the invariant measures are convex combinations of the measure concentrated in the state “0” and the measure concentrated in the state “2”.

Exercise 5. At a certain beach resort a bad day is equally likely to be followed by a good or a bad day, while a good day is five times more likely to be followed by a good day than by a bad day. The number of interventions by lifesavers is Poisson distributed with mean 4 in good days and mean 1 in bad days. Find, in the long run,

(a) (0.8 pts.) The probability of the lifesavers not having any intervention in a given day.

(b) (0.8 pts.) The average number of interventions per day.

[Take $e^{-4} \sim 0.02$ and $e^{-1} \sim 0.4$.]

Answers: Associating “good days” $\rightarrow 1$ and “bad days” $\rightarrow 2$, the weather pattern is a Markov process with transition matrix

$$\begin{pmatrix} 5/6 & 1/6 \\ 1/2 & 1/2 \end{pmatrix}.$$

The proportion of good and bad days is determined, in the long run, by the invariant measure Π of this chain. This measure satisfies:

$$\left. \begin{array}{l} \frac{5}{6}\Pi_1 + \frac{1}{2}\Pi_2 = \Pi_1 \\ \Pi_1 + \Pi_2 = 1 \end{array} \right\} \implies \Pi = \left(\frac{3}{4}, \frac{1}{4} \right).$$

(a) Let S be the number of interventions per day.

$$\begin{aligned} P(S = 0) &= P(S = 0 \mid \text{good day})P(\text{good day}) + P(S = 0 \mid \text{bad day})P(\text{bad day}) \\ &= e^{-4} \frac{3}{4} + e^{-1} \frac{1}{4} \sim 0.11 \end{aligned}$$

(b)

$$\begin{aligned} E(S) &= E(S \mid \text{good day})P(\text{good day}) + E(S \mid \text{bad day})P(\text{bad day}) \\ &= 4 \cdot \frac{3}{4} + 1 \cdot \frac{1}{4} = \frac{13}{4} = 3.25. \end{aligned}$$

Bonus problem

Bonus Consider a homogeneous (or shift-invariant) Markov chain $(X_n)_{n \in \mathbb{N}}$ with finite state space S . Let us recall that the *hitting time* of a state y is

$$T_y = \min\{n \geq 1 : X_n = y\}.$$

(a) If $\ell \leq n \in \mathbb{N}$, $x, y \in S$, prove the following

-i- (0.5 pts.)

$$P(X_n = y, T_y = \ell \mid X_0 = x) = P_{yy}^{n-\ell} P(T_y = \ell \mid X_0 = x).$$

Answer: Decomposing in terms of trajectories,

$$\begin{aligned} &P(X_n = y, T_y = \ell \mid X_0 = x) \\ &= \sum_{x_1, \dots, x_{\ell-1} \neq y} P(X_n = y, X_\ell = y, X_{\ell-1} = x_{\ell-1}, \dots, X_1 = x_1 \mid X_0 = x) \\ &= \sum_{x_1, \dots, x_{\ell-1} \neq y} P_{yy}^{n-\ell} P(X_\ell = y, X_{\ell-1} = x_{\ell-1}, \dots, X_1 = x_1 \mid X_0 = x) \\ &= \sum_{x_1, \dots, x_{\ell-1} \neq y} P_{yy}^{n-\ell} P(T_y = \ell \mid X_0 = x) \end{aligned}$$

-ii- (0.5 pts.)

$$P_{xy}^n = \sum_{\ell=1}^n P_{yy}^{n-\ell} P(T_y = \ell \mid X_0 = x).$$

Answer: As

$$\{X_n = y\} = \bigcup_{\ell=1}^n \{X_n = y, T_y = \ell\},$$

the union being disjoint, we conclude that

$$\sum_{\ell=1}^n P(X_n = y, T_y = \ell \mid X_0 = x) = P(X_n = y \mid X_0 = x) = P_{xy}^n.$$

The result follows, hence, by summing both sides of -i- with respect to ℓ .

(b) Conclude the following:

-i- (0.5 pts.) If every state is transient, then for every $x, y \in S$.

$$\sum_{n \geq 0} P_{xy}^n < \infty.$$

Answer: By -ii- above,

$$\sum_{n \geq 0} P_{xy}^n = \sum_{n \geq 0} \sum_{\ell=1}^n P_{yy}^{n-\ell} P(T_y = \ell \mid X_0 = x).$$

Hence, interchanging the order of summation,

$$\begin{aligned} \sum_n P_{xy}^n &= \sum_{\ell \geq 1} \sum_{n \geq \ell} P_{yy}^{n-\ell} P(T_y = \ell \mid X_0 = x) \\ &= \sum_{\ell \geq 1} P(T_y = \ell \mid X_0 = x) \sum_{m \geq 0} P_{yy}^m \\ &= P(T_y < \infty \mid X_0 = x) \sum_{m \geq 0} P_{yy}^m. \end{aligned}$$

If y is transient, the last sum is finite.

-ii- (0.5 pts.) The previous result leads to a contradiction with the stochasticity property of the matrix \mathbb{P} . Hence not all states can be transient.

Answer: Summing over y the inequality in (b)-i- we get

$$\sum_y \sum_{n \geq 0} P_{xy}^n < \infty \tag{1}$$

(recall that S is finite). However, by stochasticity $\sum_y P_{xy}^n = 1$ for every $n \geq 0$. Hence,

$$\sum_y \sum_{n \geq 0} P_{xy}^n = \sum_{n \geq 0} \sum_y P_{xy}^n = \sum_{n \geq 0} 1 = \infty,$$

in contradiction with (1).