

**IEOR 4106, Midterm Exam, Spring 2018. 75 Minutes.
100 Points Total. Professor K. Sigman**

Open Notes (anything on the course website plus your notes from class), but **no books and no electronic devices of any kind.**

Make sure to show/justify your work, don't just write down an answer with no explanation!

1. (30 points, 10 each) A gambler George starts with $i = 1$ (dollar), and plays according to the Gambler's ruin problem, with $p = 2/3$, but with N a random variable: $P(N = 2) = P(N = 3) = 1/2$. The idea is that just before George starts, he first flips a fair coin (once) to decide the value of N (Heads = 2, Tails = 3). Then he plays until reaching that value of N or going broke, whichever happens first, then he stops and goes home. He starts with $i = 1$ (dollar).

- (a) Given that $N = 2$, what is the probability that the gambler goes home broke? (Exact numerical answer must be given.)

SOLUTION: Probability = $1/3$ as is clear: When $N = 2$ with $i = 1$, the game is over after the first gamble (either $1 \rightarrow 2$ or $1 \rightarrow 0$), and with probability $q = 1/3$ that gamble will end the game with the gambler broke ($1 \rightarrow 0$).

Formally (but not needed here for credit, the above argument is fine):

We want $1 - P_i(N)$, where

$$P_i(N) = \frac{1 - (q/p)^i}{1 - (q/p)^N}.$$

And we would use $i = 1, q/p = 1/2, N = 2$:

$$P_1(2) = \frac{1 - (1/2)}{1 - (1/2)^2} = 2/3;$$

$$1 - P_1(2) = 1/3.$$

If $N = 3$ (we need this computation for (b) below), then we do need to use the more general formula given above and we want

$$P_1(3) = \frac{1 - (1/2)}{1 - (1/2)^3} = 4/7;$$

$$\text{Answer} = 1 - P_1(3) = 3/7.$$

- (b) Compute the probability (exact numerical answer) that the gambler will go home broke.

SOLUTION: By conditioning on $N = 2$ and $N = 3$ (using what we computed in (a) for both cases), our answer will be a 50 – 50 weighted average of using $N = 2, 3$ because each case will occur with probability $1/2$ by the fair coin assumption:

$$(1/2)(1 - P_1(2)) + (1/2)(1 - P_1(3)) = (1/2)(1/3 + 3/7) = 8/21$$

- (c) Explain (but you do not need to carry out the computation) how to compute the probability that George will go home after at most (\leq) 7 gambles.

SOLUTION:

If $N = 2$, then the game is over in 1 gamble with certainty, hence over after at most 7 gambles with certainty. But if $N = 3$, then we must consider the transition matrix P for the Gambler's ruin problem Markov chain on $\{0, 1, 2, 3\}$, and compute $P^{(7)} = P^7$ and use as our answer $P_{1,0}^{(7)} + P_{1,3}^{(7)}$; both 0 and 3 are absorbing states. When $N = 3$, P is given by

$$P = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 1/3 & 0 & 2/3 & 0 \\ 0 & 1/3 & 0 & 2/3 \\ 0 & 0 & 0 & 1 \end{pmatrix}$$

Then one must compute P^7 (not required here on the exam; just let $P^7 = (P_{i,j}^{(7)})$ and express the answer in terms of this.) Thus our answer is the 50 – 50 weighted average:

$$(1/2)(1 + (P_{1,0}^{(7)} + P_{1,3}^{(7)}))$$

2. (10 points) A certain stochastic process $\{X_n : n \geq 0\}$ is believed by a researcher to be a Markov chain with state space $\mathcal{S} = \{1, 2\}$ and transition matrix of the form

$$P = \begin{pmatrix} 0.5 & 0.5 \\ p & 1-p \end{pmatrix},$$

for some $0 < p < 1$ unknown. By looking at the values $\{X_0, \dots, X_n\}$ for a very very large time n , the researcher estimated that the process visits state 1 approximately 40% of the time, and visits state 2 approximately 60% of the time. From this, give a very reasonable choice of what the numerical value of p should be.

SOLUTION:

For any $0 < p < 1$, the chain is irreducible and since it has a finite state space it must have a unique limiting distribution $\pi = (\pi_1, \pi_2)$ satisfying $\pi = \pi P$. From the information they told us: We can assume that (as a very precise approximation) $\pi_1 = 0.40$ and $\pi_2 = 0.60$, since indeed the π_i are (by definition) the long run proportions of time the chain visits states 1 and 2 respectively. Then we plug that into $\pi = \pi P$ to solve for p :

$$\pi_1 = (0.5)\pi_1 + p\pi_2 \tag{1}$$

$$\pi_2 = (0.5)\pi_1 + (1-p)\pi_2. \tag{2}$$

The first equation becomes $0.40 = (0.5)(0.40) + p(0.60)$, from which we easily solve; $p = 1/3$. (The second equation yields the same solution if you wish to use that instead.)

3. (20 points, 10 each)

Let $\psi = \{t_n : n \geq 1\}$ be a Poisson process at rate λ , with counting process $\{N(t) : t \geq 0\}$. For a fixed $t > 0$, let $T = N(t) + 1$.

- (a) Compute $E(T)$

SOLUTION:

$N(t)$ is Poisson distributed with mean λt , so $E(T) = E(N(t)) + 1 = \lambda t + 1$.

- (b) Compute $E(t_T)$; the expected time that the T^{th} point occurs.

SOLUTION:

There are 2 Methods:

METHOD I (Wald's Equation): Letting $\{X_i\}$ denote the iid interarrival times; they are distributed as exponential with rate λ (hence $E(X) = 1/\lambda$). $t_n = X_1 + \dots + X_n = \sum_{i=1}^n X_i$, and hence

$$t_T = \sum_{i=1}^T X_i,$$

$t_T = t_{N(t)} + 1$ is the *first point strictly after time t* ; $T = N(t) + 1 = \min\{n \geq 1 : t_n > t\}$ is a stopping time with respect to $\{t_n : n \geq 1\}$, equivalently with respect to $\{X_n : n \geq 1\}$, and $E(T) = \lambda t + 1 < \infty$, and $E(X) = 1/\lambda < \infty$. Thus by Wald's equation we have $E(t_T) = E(T)E(X) = (\lambda t + 1)/\lambda = t + 1/\lambda$.

NOTE: $N(t)$ is *not* a stopping time; $N(t) = \max\{n : t_n \leq t\}$.

$t_{N(t)}$ is the *last point before or at time t* :

$t_{N(t)} \leq t < t_{N(t)+1}$. But $N(t) + 1$ is a stopping time.

METHOD II: Memoryless property of the exponential distribution:

$t_T = t_{N(t)+1} = t + A(t)$, where $A(t) = t_{N(t+1)} - t$ is the *remaining* interarrival time from time t onwards. By the memoryless property of the exponential distribution, $A(t)$ is exponential at rate λ (and independent of the past), hence $E(t_T) = t + E(A(t)) = t + 1/\lambda$.

4. (40 points, 10 each) Consider an $M/G/\infty$ queue, with arrival rate $\lambda = 2$ and iid service times distributed as (CDF) $G(x) = P(S \leq x) = 1 - \frac{1}{(1+x)^2}$, $x \geq 0$, but at time $t = 0$, two initial customers $C_0(1), C_0(2)$ enter service with iid independent service times Y_1, Y_2 distributed as exponential at rate 1.

- (a) Let $X(t)$, $X(0) = 0$ denote the number of busy servers at time t *not including the 2 initial customers*. Compute $E(X(1))$.

SOLUTION:

$X(t)$ has a Poisson distribution with mean

$$\alpha(t) = \lambda \int_0^t P(S > u) du = 2 \int_0^t \frac{1}{(1+u)^2} du = 2\left(1 - \frac{1}{1+t}\right).$$

$$E(X(1)) = \alpha(1) = 1.$$

$$\text{Also, } E(S) = \int_0^\infty \frac{1}{(1+u)^2} du = 1; \rho = \lambda E(S) = 2(1) = 2.$$

- (b) *Continuation:*

Compute $E(X(\infty)) = \lim_{t \rightarrow \infty} E(X(t))$.

SOLUTION: $\lim_{t \rightarrow \infty} E(X(t)) = \lim_{t \rightarrow \infty} \alpha(t) = \rho = \lambda E(S) = \alpha(\infty) = 2(1) = 2$.

- (c) Let $Z(t)$, $Z(0) = 2$ denote the number of busy servers at time t *including the 2 initial customers*. Compute $E(Z(1))$.

SOLUTION: The 2 initial customers will depart at times Y_1 and Y_2 respectively; hence each is still in the system at time t if and only if $Y_1 > t$ and $Y_2 > t$ respectively. Thus

$$Z(t) = X(t) + I\{Y_1 > t\} + I\{Y_2 > t\}, \text{ and } E(I\{Y_1 > t\}) = E(I\{Y_2 > t\}) = P(Y_1 > t) = e^{-t}.$$

Thus

$$E(Z(t)) = E(X(t)) + 2P(Y_1 > t) = \alpha(t) + 2e^{-t}.$$

$$E(Z(1)) = \alpha(1) + 2e^{-1} = 1 + 2e^{-1}.$$

- (d) *Continuation:* Compute $E(Z(\infty)) = \lim_{t \rightarrow \infty} E(Z(t))$.

SOLUTION: From (c), $E(Z(t)) = \alpha(t) + 2e^{-t}$, and $e^{-t} \rightarrow 0$ as $t \rightarrow \infty$ hence the answer is the same as in (b); $\rho = 2$.