
STAT 150: STOCHASTIC PROCESSES

Fall 2017



HOMEWORK 1

DUE SEP 6 AT 11:59PM



Solutions by

JINHONG DU

3033483677

1.3.11

Let X and Y be independent random variables sharing the geometric distribution whose mass function is

$$p(k) = (1 - \pi)\pi^k \quad k = 0, 1, \dots$$

where $0 < \pi < 1$. Let $U = \min\{X, Y\}$, $V = \max\{X, Y\}$ and $W = V - U$. Determine the joint probability mass function for U and W and show that U and W are independent.

Proof.

$\therefore X, Y \stackrel{iid}{\sim} G(\pi)$

\therefore the mass distribution functions of U and V are

$$\begin{aligned} p_U(k) &= P(\min\{X, Y\} = k) \\ &= P(\min\{X, Y\} = k, \max\{X, Y\} \geq k) \\ &= P(\min\{X, Y\} = k, \max\{X, Y\} > k) + P(\min\{X, Y\} = k, \max\{X, Y\} = k) \\ &= \binom{2}{1} p(k) \sum_{j=k+1}^{+\infty} p(j) + p(k)p(k) \\ &= 2(1 - \pi)\pi^{2k+1} + (1 - \pi)^2\pi^{2k} \\ &= \pi^{2k}(1 - \pi^2) \\ p_V(k) &= P(\max\{X, Y\} = k) \\ &= P(\min\{X, Y\} \leq k, \max\{X, Y\} = k) \\ &= P(\min\{X, Y\} < k, \max\{X, Y\} = k) + P(\min\{X, Y\} = k, \max\{X, Y\} = k) \\ &= \binom{2}{1} \sum_{j=0}^{k-1} p(j)p(k) + p(k)p(k) \\ &= 2(1 - \pi)(1 - \pi^k)\pi^k + (1 - \pi)^2\pi^{2k} \\ &= \pi^k(\pi - 1)(\pi^k + \pi^{k+1} - 2) \end{aligned}$$

The joint probability mass function for U and V is

$$\begin{aligned} p_{U,V}(u, v) &= P(\min\{X, Y\} = u, \max\{X, Y\} = v) \\ &= \begin{cases} \binom{2}{1} p(u)p(v), & u < v, u, v \in N \\ p(u)p(v), & u = v, u, v \in N \\ 0, & \text{otherwise} \end{cases} \\ &= \begin{cases} 2(1 - \pi)^2\pi^{u+v}, & u < v, u, v \in N \\ (1 - \pi)^2\pi^{2u}, & u = v, u, v \in N \\ 0, & \text{otherwise} \end{cases} \end{aligned}$$

let $V = W + U$, we get

$$p_{U,W}(u, w) = \begin{cases} 2(1 - \pi)^2\pi^{2u+w}, & w > 0, u, w \in N \\ (1 - \pi)^2\pi^{2u}, & w = 0, u \in N \\ 0, & \text{otherwise} \end{cases}$$

Solution (cont.)

The mass distribution function of W is

$$\begin{aligned} p_W(w) &= \sum_{u=0}^{+\infty} p_{U,W}(u, w) \\ &= \begin{cases} \frac{2(1-\pi)^2\pi^w}{1-\pi^2}, & w > 0, w \in \mathbb{N} \\ \frac{(1-\pi)^2}{1-\pi^2}, & w = 0 \\ 0, & \text{otherwise} \end{cases} \end{aligned}$$

$$\therefore p_U(u)p_W(w) = p_{U,W}(u, w)$$

$\therefore U, V$ are independent. □

2.1.2

A card is picked at random from N cards labeled $1, 2, \dots, N$, and the number that appears is X . A second card is picked at random from cards numbered $1, 2, \dots, X$ and its number is Y . Determine the conditional distribution of X given $Y = y$, for $y = 1, 2, \dots$.

Proof.

$$\begin{aligned} P(X = x|Y = y) &= \frac{P(X = x, Y = y)}{P(Y = y)} \\ &= \frac{P(Y = y|X = x)P(X = x)}{\sum_{j=y}^N P(Y = y|X = j)} \\ &= \begin{cases} \frac{P(Y = y|X = x)P(X = x)}{\sum_{j=y}^N P(Y = y|X = j)P(X = j)}, & y \leq x \leq N \\ 0, & x < y \end{cases} \\ &= \begin{cases} \frac{\frac{1}{x} \cdot \frac{1}{N}}{\sum_{j=y}^N \frac{1}{j} \cdot \frac{1}{N}}, & y \leq x \leq N \\ 0, & x < y \end{cases} \\ &= \begin{cases} \frac{1}{x \sum_{j=y}^N \frac{1}{j}}, & y \leq x \leq N \\ 0, & x < y \end{cases} \end{aligned}$$
□

2.3.3

Suppose that ξ_1, ξ_2, \dots are independent and identically distributed with $Pr\{\xi_k = \pm 1\} = \frac{1}{2}$. Let N be independent of ξ_1, ξ_2, \dots and follow the geometric probability mass function

$$p_N(k) = \alpha(1 - \alpha)^k \quad \text{for } k = 0, 1, \dots$$

where $0 < \alpha < 1$. Form the random sum $Z = \xi_1 + \dots + \xi_N$.

(a) Determine the mean and variance of Z .

\therefore

$$\begin{aligned} E\xi_i &= 1 \cdot Pr\{\xi_i = 1\} - 1 \cdot Pr\{\xi_i = -1\} \\ &= \frac{1}{2} - \frac{1}{2} \\ &= 0 \end{aligned}$$

$$\begin{aligned} Var\xi_i &= 1^2 Pr\{\xi_i = 1\} + (0 - 1)^2 Pr\{\xi_i = -1\} \\ &= \frac{1}{2} + \frac{1}{2} \\ &= 1 \end{aligned}$$

$$\begin{aligned} EN &= \sum_{k=0}^{+\infty} k p_N(k) \\ &= \sum_{k=0}^{+\infty} k \alpha (1 - \alpha)^k \\ &= \frac{1 - \alpha}{\alpha} \end{aligned}$$

$$\begin{aligned} VarN &= EN^2 - (EN)^2 \\ &= \sum_{k=0}^{+\infty} k^2 p_N(k) - \left(\frac{1 - \alpha}{\alpha} \right)^2 \\ &= \sum_{k=0}^{+\infty} k^2 \alpha (1 - \alpha)^k - \left(\frac{1 - \alpha}{\alpha} \right)^2 \\ &= \frac{(2 - \alpha)(1 - \alpha)}{\alpha^2} - \left(\frac{\alpha}{1 - \alpha} \right)^2 \\ &= \frac{1 - \alpha}{\alpha^2} \end{aligned}$$

Assume that $Z = 0$ when $N = 0$.

\therefore

$$\begin{aligned} EZ &= \sum_{k=0}^{+\infty} E(Z|N = k) Pr\{N = k\} \\ &= \sum_{k=1}^{+\infty} E(\xi_1 + \dots + \xi_k) Pr\{N = k\} \\ &= \sum_{k=1}^{+\infty} \sum_{j=1}^k E\xi_j Pr\{N = k\} \\ &= 0 \end{aligned}$$

Solution (cont.)

$$\begin{aligned}
 \text{Var} Z &= E(Z - EZ)^2 \\
 &= EZ^2 \\
 &= \sum_{k=1}^{+\infty} E(Z^2 | N = k) \text{Pr}\{N = k\} \\
 &= \sum_{k=1}^{+\infty} \text{Var}(Z | N = k) \text{Pr}\{N = k\} \\
 &= \sum_{k=1}^{+\infty} \sum_{j=1}^k \text{Var}(\xi_j) \text{Pr}\{N = k\} \\
 &= \sum_{k=1}^{+\infty} k\alpha(1 - \alpha)^k \\
 &= \alpha(1 - \alpha) \sum_{k=1}^{+\infty} k(1 - \alpha)^{k-1} \\
 &= \alpha(1 - \alpha) \frac{d}{d(1 - \alpha)} \left[\sum_{k=1}^{+\infty} (1 - \alpha)^k \right] \\
 &= \alpha(1 - \alpha) \frac{1}{\alpha^2} \\
 &= \frac{1 - \alpha}{\alpha}
 \end{aligned}$$

(b) Evaluate the higher moments $m_3 = E[Z^3]$ and $m_4 = E[Z^4]$.

Hint: Express Z^4 in terms of the ξ_i 's where $\xi_i^2 = 1$ and $E[\xi_i \xi_j] = 0$.

\therefore

$$\begin{aligned}
 E\xi_i^2 &= \text{Var}\xi_i + (E\xi_i)^2 \\
 &= 1 \\
 E(\xi_i \xi_j) &= E\xi_i E\xi_j \\
 &= 0 \\
 E\xi_i^3 &= 1^3 \text{Pr}\{\xi_i = 1\} + (-1)^3 \text{Pr}\{\xi_i = 0\} \\
 &= \frac{1}{2} - \frac{1}{2} \\
 &= 0 \\
 \xi_i^2 &= 1
 \end{aligned}$$

Solution (cont.)

\therefore

$$\begin{aligned}
m_3 &= EZ^3 \\
&= \sum_{k=0}^{+\infty} E(Z^3|N=k)Pr\{N=k\} \\
&= \sum_{k=0}^{+\infty} E(\xi_1 + \dots + \xi_k)^3 Pr\{N=k\} \\
&= \sum_{k=0}^{+\infty} E \left(\sum_{j=1}^k \xi_j^3 + \binom{3}{1} \sum_{i=1}^k \sum_{\substack{j=1 \\ i \neq j}}^k \xi_i^2 \xi_j + \binom{3}{1} \binom{2}{1} \sum_{i=1}^k \sum_{\substack{j=1 \\ i \neq j}}^k \sum_{\substack{m=1 \\ i \neq m \\ j \neq m}}^k \xi_i \xi_j \xi_m \right) \\
&\quad \cdot Pr\{N=k\} \\
&= \sum_{k=0}^{+\infty} \left(\sum_{j=1}^k E\xi_j^3 + \sum_{i=1}^k \sum_{\substack{j=1 \\ i \neq j}}^k E\xi_i^2 E\xi_j + \sum_{i=1}^k \sum_{\substack{j=1 \\ i \neq j}}^k \sum_{\substack{m=1 \\ i \neq m \\ j \neq m}}^k E\xi_i E\xi_j E\xi_m \right) \\
&\quad \cdot Pr\{N=k\} \\
&= 0 \\
m_4 &= EZ^4 \\
&= \sum_{k=0}^{+\infty} E(Z^4|N=k)Pr\{N=k\} \\
&= \sum_{k=0}^{+\infty} E(\xi_1 + \dots + \xi_k)^4 Pr\{N=k\} \\
&= \sum_{k=0}^{+\infty} E \left(\binom{4}{1} \binom{3}{1} \binom{2}{1} \sum_{i=1}^k \sum_{\substack{j=1 \\ i \neq j}}^k \sum_{\substack{l=1 \\ i \neq l \\ j \neq l}}^k \sum_{\substack{m=1 \\ i \neq m \\ j \neq m \\ l \neq m}}^k \xi_i \xi_j \xi_l \xi_m + \binom{4}{1} \binom{3}{1} \sum_{i=1}^k \sum_{\substack{j=1 \\ i \neq j}}^k \xi_i \xi_j \cdot 1 + \right. \\
&\quad \left. \frac{1}{2} \binom{4}{2} \sum_{i=1}^k \sum_{\substack{j=1 \\ i \neq j}}^k 1 \cdot 1 + \binom{4}{1} \sum_{i=1}^k \sum_{\substack{j=1 \\ i \neq j}}^k \xi_i \xi_j \cdot 1 + \sum_{i=1}^k 1 \cdot 1 \right) \cdot Pr\{N=k\} \\
&= \sum_{k=0}^{+\infty} (3k^2 - 3k + k) \alpha (1 - \alpha)^k \\
&= \sum_{k=0}^{+\infty} (3k^2 - 2k) \alpha (1 - \alpha)^k \\
&= E(3N^2 - 2N) \\
&= 3EN^2 - 2EN
\end{aligned}$$

Solution (cont.)

$$\begin{aligned} &= 3[(EN)^2 + \text{Var}N] - 2EN \\ &= 3 \left[\left(\frac{1-\alpha}{\alpha} \right)^2 + \frac{1-\alpha}{\alpha^2} \right] - 2 \frac{1-\alpha}{\alpha} \\ &= \frac{5\alpha^2 - 11\alpha + 6}{\alpha^2} \end{aligned}$$

2.4.3

Let X be a Poisson distribution with parameter $\lambda > 0$. Suppose λ itself is random, following an exponential density with parameter θ .

(a) What is the marginal distribution of X .

$$\because \lambda \sim \text{Exp}(\theta)$$

\therefore

$$f_{X|\lambda}(k|y) = \begin{cases} \frac{y^k}{k!} e^{-y}, & k \in \mathbb{N} \\ 0, & \text{otherwise} \end{cases}$$

$$f_{\lambda}(y) = \begin{cases} \theta e^{-\theta y}, & t \geq 0 \\ 0, & t < 0 \end{cases}$$

$$\begin{aligned} f_{X\lambda}(k, y) &= f_{X|\lambda}(k|y) f_{\lambda}(y) \\ &= \begin{cases} \frac{\theta y^k}{k!} e^{-(\theta+1)y}, & k \in \mathbb{N}, \theta \geq 0 \\ 0, & \text{otherwise} \end{cases} \end{aligned}$$

$$\begin{aligned} f_X(k) &= \int_{\mathbb{R}} f_{X\lambda}(k, y) dy \\ &= \int_0^{+\infty} \frac{\theta y^k}{k!} e^{-(\theta+1)y} dy \\ &= \frac{\theta}{k!(\theta+1)^{k+1}} \int_0^{+\infty} [(\theta+1)y]^k e^{-(\theta+1)y} d[(\theta+1)y] \\ &= \frac{\theta}{k!(\theta+1)^{k+1}} \Gamma(k+1) \\ &= \frac{\theta}{(\theta+1)^{k+1}} \quad k \in \mathbb{N} \\ f_X(k) &= 0 \quad k \notin \mathbb{N} \end{aligned}$$

(b) Determine the conditional density for λ given $X = k$.

$$\begin{aligned}
f_{\lambda|X}(y|k) &= \frac{f_{X,\lambda}(k,y)}{f_X(k)} \\
&= \begin{cases} \frac{\frac{\theta y^k}{k!} e^{-(\theta+1)y}}{\theta}, & y \geq 0, k \in \mathbb{N} \\ 0, & \text{otherwise} \end{cases} \\
&= \begin{cases} \frac{(\theta+1)^{k+1} y^k}{k!} e^{-(\theta+1)y}, & y \geq 0, k \in \mathbb{N} \\ 0, & \text{otherwise} \end{cases}
\end{aligned}$$

2.4.4

Suppose X and Y are independent random variables having the same Poisson distribution with parameter λ , but where λ is also random, being exponentially distributed with parameter θ . What is the conditional distribution for X given that $X + Y = n$?

From Problem 2.4.3 we have

$$f_X(x) = \frac{\theta}{(\theta+1)^{x+1}} I_{\mathbb{N}}(x)$$

$$f_Y(y) = \frac{\theta}{(\theta+1)^{y+1}} I_{\mathbb{N}}(y)$$

\therefore X and Y are independent

\therefore

$$f_{XY}(x,y) = \frac{\theta^2}{(\theta+1)^{x+y+2}} I_{\mathbb{N}}(x) I_{\mathbb{N}}(y)$$

$$\text{Let } Z = X + Y, J = \begin{vmatrix} 1 & 0 \\ 1 & -1 \end{vmatrix} = 1$$

$$\begin{aligned}
f_{XZ}(x,n) &= f_{XY}(x,n-x) \frac{1}{|J|} \\
&= \frac{\theta^2}{(\theta+1)^{n+2}} I_{\mathbb{N}}(x) I_{\mathbb{N}}(n-x)
\end{aligned}$$

Solution (cont.)

$$\begin{aligned}
 f_Z(n) &= \int_R f_{XZ}(x, n) dx \\
 &= \begin{cases} \sum_{x=0}^n \frac{\theta^2}{(\theta+1)^{n+2}}, & n \in \mathbb{N} \\ 0, & \text{otherwise} \end{cases} \\
 &= \begin{cases} \frac{(n+1)\theta^2}{(\theta+1)^{n+2}}, & n \in \mathbb{N} \\ 0, & \text{otherwise} \end{cases} \\
 f_{X|Z}(x|n) &= \frac{f_{XZ}(x, n)}{f_Z(n)} \\
 &= \begin{cases} \frac{\frac{\theta^2}{(\theta+1)^{n+2}}}{\frac{(n+1)\theta^2}{(\theta+1)^{n+2}}}, & x, n \in \mathbb{N}, x \leq n \\ 0, & \text{otherwise} \end{cases} \\
 &= \begin{cases} \frac{1}{n+1}, & x, n \in \mathbb{N}, x \leq n \\ 0, & \text{otherwise} \end{cases}
 \end{aligned}$$

2.4.6

Let X_0, X_1, X_2, \dots be independent identically distributed nonnegative random variables having a continuous distribution. Let N be the first index k for which $X_k > X_0$. That is, $N = 1$ if $X_1 > X_0$, $N = 2$ if $X_1 \leq X_0$ and $X_2 > X_0$, etc. Determine the probability mass function for N and the mean $E[N]$. (Interpretation: X_0, X_1, \dots are successive offers or bids on a car that you are trying to sell. Then, N is the index of the first bid that is better than the initial bid.)

Suppose that the probability distribution function of X_i is $f(x)$, the cumulative distribution function of X_i is $F(x)$. Because X_i is nonnegative random variables with continuous distribution, $f(x) \equiv 0$ when $x < 0$ and $F(0) = 0$.

Given n , let $A_n = \{X_1 \leq X_0, X_2 \leq X_0, \dots, X_{n-1} \leq X_0, X_n > X_0\}$,

Solution (cont.)

$$\begin{aligned}f_N(n) &= P(A_n) \\&= \int_{A_n} f_{X_0 X_1 \dots X_n}(x_0, x_1, \dots, x_n) dx_0 \dots dx_n \\&= \int_{A_n} f_{X_0}(x_0) f_{X_1}(x_1) \dots f_{X_n}(x_n) dx_0 \dots dx_n \\&= \int_0^{+\infty} \int_0^{x_n} \left[\left(\int_0^{x_0} f(x_1) dx_1 \right) \dots \left(\int_0^{x_0} f(x_{n-1}) dx_{n-1} \right) \right] \\&\quad \cdot f(x_0) f(x_n) dx_0 dx_n \\&= \int_0^{+\infty} \int_0^{x_n} [F(x_0)]^{n-1} dF(x_0) dF(x_n) \\&= \int_0^{+\infty} \frac{1}{n} [F(x_0)]^n \Big|_0^{x_n} dF(x_n) \\&= \int_0^{+\infty} \frac{1}{n} [F(x_n)]^n dF(x_n) \\&= \frac{1}{n(n+1)} [F(x_n)]^{n+1} \Big|_0^{+\infty} \\&= \frac{1}{n(n+1)} \\EN &= \sum_{n=1}^{+\infty} \frac{1}{n+1} \\&= +\infty\end{aligned}$$