

應用機率模型作業 5 解答

Chapter5: #10, #12, #43

1.

(a) .

$$E(MX | M = X) = E(M^2 | M = X)$$

Since we know that $(M | M = X)$ has the same distribution with $M \sim \exp(\lambda + \mu)$

$$\therefore E(M^2 | M = X) = E(M^2) = \frac{2}{(\mu + \lambda)^2}.$$

(b) .

Given that $M = Y$ (i.e. $Y < X$), by the lack of memory property

X can be denote as $M + X'$ where X' has the same distribution with X , and is independent with M .

$$\begin{aligned} \therefore E(MX | M = Y) &= E(M(M + X') | M = Y) = E(M^2 | M = Y) + E(MX' | M = Y) \\ &= E(M^2) + E(M)E(X') = \frac{2}{(\lambda + \mu)^2} + \frac{1}{\lambda(\lambda + \mu)}. \end{aligned}$$

(c).

$$\text{Cov}(X, M) = E(XM) - E(X)E(M)$$

where,

$$\begin{aligned} E(XM) &= E(E(XM | M)) = E(E(XM | M = X)P(M = X) + E(XM | M = Y)P(M = Y)) \\ &= \frac{2}{(\lambda + \mu)^2} \frac{\lambda}{\lambda + \mu} + \left(\frac{2}{(\lambda + \mu)^2} + \frac{1}{\lambda(\lambda + \mu)} \right) \frac{\mu}{\lambda + \mu}. \end{aligned}$$

hence,

$$\text{Cov}(X, M) = \frac{2}{(\lambda + \mu)^2} \frac{\lambda}{\lambda + \mu} + \left(\frac{2}{(\lambda + \mu)^2} + \frac{1}{\lambda(\lambda + \mu)} \right) \frac{\mu}{\lambda + \mu} - \frac{1}{\lambda(\lambda + \mu)}.$$

2.

(a).

By the lack of memory property,

$$\begin{aligned}
 P(X_1 < X_2 < X_3) &= P(X_2 < X_3 | X_1 = \min(X_1, X_2, X_3)) P(X_1 = \min(X_1, X_2, X_3)) \\
 &= P(X_2 < X_3) P(X_1 = \min(X_1, X_2, X_3)) \\
 &= \frac{\lambda_2}{\lambda_2 + \lambda_3} \frac{\lambda_1}{\lambda_1 + \lambda_2 + \lambda_3}.
 \end{aligned}$$

(b).

$$\begin{aligned}
 P(X_1 < X_2 | X_3 = \max(X_1, X_2, X_3)) &= \frac{P(X_1 < X_2 < X_3)}{P(X_1 < X_2 < X_3) + P(X_2 < X_1 < X_3)} \\
 &= \frac{\frac{\lambda_2}{\lambda_2 + \lambda_3} \frac{\lambda_1}{\lambda_1 + \lambda_2 + \lambda_3}}{\frac{\lambda_2}{\lambda_2 + \lambda_3} \frac{\lambda_1}{\lambda_1 + \lambda_2 + \lambda_3} + \frac{\lambda_1}{\lambda_1 + \lambda_3} \frac{\lambda_2}{\lambda_1 + \lambda_2 + \lambda_3}} \\
 &= \frac{1}{\frac{1}{\lambda_2 + \lambda_3} + \frac{1}{\lambda_1 + \lambda_3}}.
 \end{aligned}$$

(c).

$$E(\max X_i | X_1 < X_2 < X_3) = E(X_3 | X_1 < X_2 < X_3)$$

by the lack of memory property, given on $X_1 < X_2 < X_3$,

$$X_3 \equiv X_2 + X_3'$$

, where $X_3' \sim \exp(\lambda_3)$ is independent with X_1, X_2, X_3 .

$$\begin{aligned}
 \therefore E(\max X_i | X_1 < X_2 < X_3) &= E(X_3') + E(X_2 | X_1 < X_2 < X_3) \\
 &= E(X_3') + E(X_2 | M_1 = X_1, X_2 < X_3)
 \end{aligned}$$

, where M_1 denote $\min(X_1, X_2, X_3)$.

also by the lack of memory property again, given on $M_1 = X_1, X_2 < X_3$,

$$\begin{aligned}
 E(X_2 | M_1 = X_1, X_2 < X_3) &= E(X_1 + X_2' | M_1 = X_1, X_2' < X_3) \\
 &= E(X_1 | M_1 = X_1) + E(X_2' | X_2' < X_3)
 \end{aligned}$$

, where $X_2' \sim \exp(\lambda_2)$ is independent with M_1, X_1, X_3 .

Hence,

$$\begin{aligned}
E(\max X_i | X_1 < X_2 < X_3) &= E(X_3') + E(X_1 | M_1 = X_1) + E(X_2' | X_2' < X_3) \\
&= E(X_3') + E(\min(X_1, X_2, X_3)) + E(\min(X_2, X_3)) \\
&= \frac{1}{\lambda_1 + \lambda_2 + \lambda_3} + \frac{1}{\lambda_2 + \lambda_3} + \frac{1}{\lambda_3}.
\end{aligned}$$

(d).

$$\begin{aligned}
E(\max X_i) &= \sum_{i \neq j \neq k} E(\max(X_1, X_2, X_3) | X_i < X_j < X_k) P(X_i < X_j < X_k) \\
&= \sum_{i \neq j \neq k} \left(\frac{1}{\lambda_1 + \lambda_2 + \lambda_3} + \frac{1}{\lambda_j + \lambda_k} + \frac{1}{\lambda_k} \right) \left(\frac{\lambda_j}{\lambda_j + \lambda_k} \frac{\lambda_i}{\lambda_1 + \lambda_2 + \lambda_3} \right).
\end{aligned}$$

3.

Let T denote the time until the next arrival, S_i denote the service time at server i , $i = 1, 2$ since the exponential random variable has lack of memory property, so

$$P(T > S_1 + S_2) = P(T > S_1 + S_2 | T > S_1) P(T > S_1) = P(T > S_2) P(T > S_1) = \frac{\mu_2}{\lambda + \mu_2} \frac{\mu_1}{\lambda + \mu_1}.$$