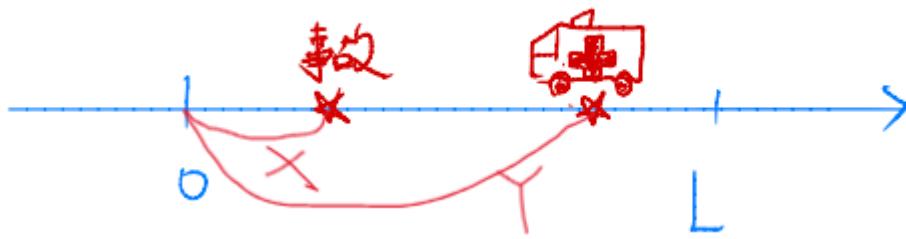


Probability_HW09_Solution

Problem 14

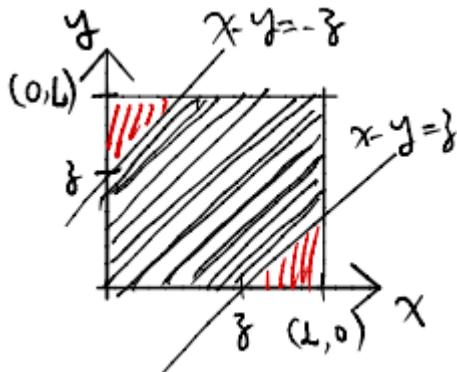


定義 X ：發生事故的位置 $\Rightarrow X \sim \text{uniform}(0, L)$

Y ：事故發生時救護車位置 $\Rightarrow Y \sim \text{uniform}(0, L)$

$Z = |X - Y|$ 是事故地點與救護車距離，則 Z 的可能值介於 $0 \sim L$

$$\text{因 } X, Y \text{ are independent, } f_{X,Y}(x,y) = f_X(x) \cdot f_Y(y) = \begin{cases} \frac{1}{L^2}, & 0 < x < L, 0 < y < L \\ 0, & \text{otherwise} \end{cases}$$



故 Z 的 cdf 是：

$$F_Z(z) = P(|X - Y| \leq z) = P(-z \leq X - Y \leq z) = \begin{cases} 0, & \text{if } z < 0 \\ 1 - 2 \cdot \underbrace{\frac{1}{L^2} \cdot \frac{(L-z)^2}{2}}_{\text{紅色的面積乘上 density}}, & \text{if } 0 < z < L \\ 1, & \text{if } z > L \end{cases}$$

而 Z 的 pdf 是：

$$f_Z(z) = \frac{\partial F_Z(z)}{\partial z} = \begin{cases} 2 \cdot \frac{(L-z)}{L^2}, & \text{if } 0 < z < L \\ 0, & \text{otherwise} \end{cases}$$

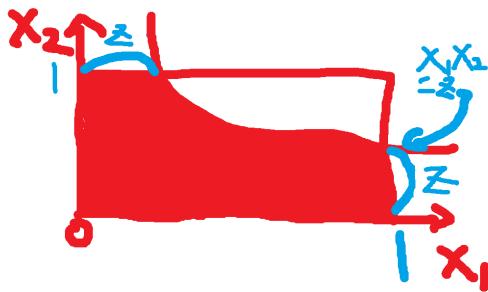
要回答 Z 的 distribution, 報告 cdf 或 pdf 皆可。

Problem 27

$$X_1, X_2 \stackrel{IID}{\sim} U(0,1)$$

\Rightarrow the joint pdf of X_1 and X_2 is $f_{X_1, X_2}(x_1, x_2) = \begin{cases} 1, & \text{if } 0 < x_1 < 1 \text{ and } 0 < x_2 < 1 \\ 0, & \text{otherwise} \end{cases}$

Let $Z = X_1 X_2 \in (0,1)$.



$$\begin{aligned} \text{The cdf of } Z \text{ is } F_Z(z) &= P(X_1 X_2 \leq z) = \int_0^1 \int_0^{\min(\frac{z}{x_1}, 1)} 1 dx_2 dx_1 = \int_0^1 \min\left(\frac{z}{x_1}, 1\right) dx_1 \\ &= \begin{cases} \int_0^z 1 dx_1 + \int_z^1 \frac{z}{x_1} dx_1 = (x_1) \Big|_0^z + (z \log(x_1)) \Big|_z^1 = z - z \log(z), & 0 < z < 1 \\ 1, & z \geq 1 \\ 0, & z \leq 0 \end{cases} \end{aligned}$$

$$\text{So } \Pr(Z > 0.5) = 1 - F_Z(0.5) = 1 - 0.5 + 0.5 \log(0.5) \approx 0.1534$$

Problem 30

Let X_1 be the service time of the first machine and
let X_2 be the service time of the second machine
Then $X_1 \sim N(1, 0.05^2)$, $X_2 \sim N(0.95, 0.02^2)$ and they are independent.

(a)

$$X_1 - X_2 \sim N(1 - 0.95, 0.05^2 + 0.02^2) = N(0.05, 0.0029).$$

$$P(X_1 < X_2) = P(X_1 - X_2 < 0) = P\left(Z < \frac{0 - (1 - 0.95)}{\sqrt{0.05^2 + 0.02^2}}\right) = P\left(Z < \frac{-0.05}{\sqrt{0.0029}}\right) = \Phi(-0.9284767) = 0.17658,$$

where $Z \sim N(0, 1)$, Φ is the $N(0, 1)$ cdf.

(b)

$$X_1 + X_2 \sim N(1 + 0.95, 0.05^2 + 0.02^2) = N(1.95, 0.0029).$$

$$P(X_1 + X_2 < 2.1) = P\left(Z < \frac{2.1 - 1.95}{\sqrt{0.0029}}\right) = \Phi\left(\frac{0.15}{\sqrt{0.0029}}\right) = \Phi(2.78543) = 0.9973,$$

where $Z \sim N(0, 1)$, Φ is the $N(0, 1)$ cdf.

Problem 42

The marginal distribution of X is $\sim N(0, 1), -\infty < X < \infty$.

Given $X = x$, the conditional pdf of $(Y \mid X = x)$ is

$$f_{Y|X}(y \mid x) = \begin{cases} \frac{1}{\Phi(\frac{x-0}{1})} \phi(\frac{y-0}{1}) = \frac{\phi(y)}{\Phi(x)} & , \text{ if } -\infty < y < x \\ 0 & , \text{ if } x \leq y < \infty \end{cases},$$

where ϕ is the pdf of $N(0, 1)$ and Φ is the cdf of $N(0, 1)$.

(a)

X, Y joint density: $f_{X,Y}(x, y) = f_{Y|X}(y|x) \cdot f_X(x)$

$$f_{X,Y}(x, y) = \begin{cases} \phi(x) \times \frac{\phi(y)}{\Phi(x)} = \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{x^2}{2}\right) \cdot \frac{\exp(-y^2/2)}{\sqrt{2\pi}\Phi(x)} & , \text{ if } -\infty < y < x < \infty \\ 0 & , \text{ otherwise} \end{cases}$$

(b)

$$\begin{aligned}
f_Y(y) &= \int f_{X,Y}(x,y)dx \\
&= \int_y^\infty \frac{\phi(x)}{\Phi(x)} \phi(y) dx \quad (\text{from } f_{X,Y}(x,y) \text{ as defined in (a)}). \\
&= \phi(y) \int_y^\infty \frac{\phi(x)}{\Phi(x)} dx \\
&= \phi(y) \cdot [\ln \Phi(x)]_y^\infty \\
&= \phi(y) \cdot (0 - \ln \Phi(y)) \\
&= -\phi(y) \ln \Phi(y) \\
&= -\frac{1}{\sqrt{2\pi}} \exp\left(-\frac{y^2}{2}\right) \ln \Phi(y), \quad \text{where } -\infty < y < \infty.
\end{aligned}$$

(c)

$$\begin{aligned}
&- \int_{-\infty}^\infty \phi(y) \ln \Phi(y) dy \\
&\stackrel{(\star)}{=} - \int_{-\infty}^0 ue^u du \\
&= - \left[(ue^u) \Big|_{-\infty}^0 - \int_{-\infty}^0 e^u du \right] \\
&= - \left[(0 - 0) - (e^u) \Big|_{-\infty}^0 \right] \\
&= (e^u) \Big|_{-\infty}^0 = 1 - 0 = 1.
\end{aligned}$$

$$(\star) : \begin{cases} u = \log(\Phi(y)) \implies e^u = \Phi(y), \quad -\infty < u < 0 \\ du = \frac{\phi(y)}{\Phi(y)} dy \implies ue^u du = \phi(y) \ln \Phi(y) dy. \end{cases}$$

Note: In (b), we derive $f_Y(y) = -\frac{1}{\sqrt{2\pi}} \exp\left(-\frac{y^2}{2}\right) \ln \Phi(y)$, where $-\infty < y < \infty$.

Since $0 < \Phi(y) \leq 1 \quad \forall -\infty < y < \infty$, thus $\ln \Phi(y) < 0 \quad \forall -\infty < y < \infty$.

And therefore $f_Y(y) > 0 \quad \forall -\infty < y < \infty$.

made by 林宸緯, 潘翠婷, 馬翌翔 助教

Problem 43

Let Z_1, Z_2 be IID random variables from $\text{Uniform}(0, 1)$.

Then, X, Y are order statistics of Z_1, Z_2 , i.e., $Z_{(1)}$ and $Z_{(n)}$ respectively.

For $n = 2$, the marginal pdf of X and Y are :

$$\begin{aligned} f_X(x) &= C_1^n \cdot f(x) \cdot [1 - F(x)]^{n-1} \\ &= n \cdot f(x) \cdot [1 - F(x)]^{n-1} \\ &= \begin{cases} 2(1-x) & , 0 < x < 1 \\ 0 & , \text{otherwise} \end{cases} \end{aligned}$$

$$\begin{aligned} f_Y(y) &= C_1^n \cdot f(y) \cdot [F(y)]^{n-1} \\ &= n \cdot f(y) \cdot [F(y)]^{n-1} \\ &= \begin{cases} 2y & , 0 < y < 1 \\ 0 & , \text{otherwise} \end{cases} \end{aligned}$$

In this example, f and F are the pdf and the cdf of $\text{Uniform}(0, 1)$, respectively.

We also have their joint pdf :

$$\begin{aligned} f_{X,Y}(x,y) &= C_2^n \cdot 2! \cdot f(x) \cdot f(y) \cdot [F(y) - F(x)]^{n-2} \\ &= n(n-1) \cdot f(x) \cdot f(y) \cdot [F(y) - F(x)]^{n-2} \\ &= \begin{cases} 2 & , 0 < x < y < 1 \\ 0 & , \text{otherwise} \end{cases} \end{aligned}$$

$$\begin{aligned} \text{So } f_{Y|X}(y|x) &= \frac{f_{X,Y}(x,y)}{f_X(x)} \\ &= \frac{n(n-1) \cdot f(x) \cdot f(y) \cdot [F(y) - F(x)]^{n-2}}{n \cdot f(x) \cdot [1 - F(x)]^{n-1}} \\ &= (n-1)f(y) \cdot \frac{[F(y) - F(x)]^{n-2}}{[1 - F(x)]^{n-1}}. \\ &= \begin{cases} 1/(1-x) & , x < y < 1 \\ 0 & , \text{otherwise} \end{cases} \end{aligned}$$

That is, $(Y \mid X = x) \sim \text{Uniform}(x, 1)$.

Because $f_{Y|X}(y \mid x) \neq f_Y(y)$, X and Y are not independent.

Problem 45

(a)

We first find the marginal densities.

$$\begin{aligned} f_X(x) &= \int_0^\infty f(x, y) dy \\ &= xe^{-x} \int_0^\infty e^{-xy} dy \\ &= xe^{-x} \left(-\frac{1}{x} e^{-xy} \right) \Big|_{y=0}^{y=\infty} \\ &= xe^{-x} \left(0 + \frac{1}{x} \right) \\ &= e^{-x}, \text{ for } x > 0. \end{aligned}$$

and $f_X(x) = 0$, for $x \leq 0$.

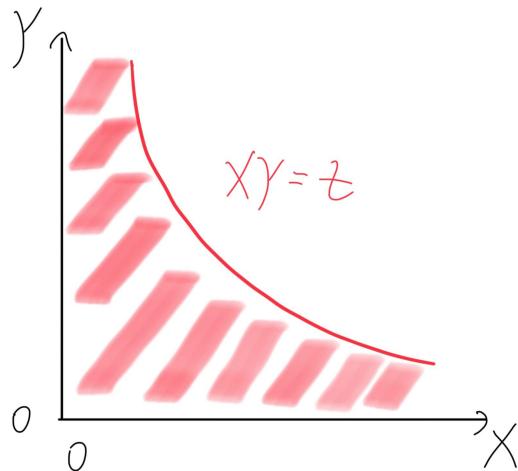
$$\begin{aligned} f_Y(y) &= \int_0^\infty f(x, y) dx \\ &= \int_0^\infty xe^{x(y+1)} dx \\ &\stackrel{(\star)}{=} \left(x \frac{-1}{y+1} e^{-x(y+1)} \right) \Big|_{x=0}^{x=\infty} - \int_0^\infty \frac{-1}{y+1} e^{-x(y+1)} dx \\ &= (0 - 0) - \left(\frac{1}{(y+1)^2} e^{-x(y+1)} \right) \Big|_{x=0}^{x=\infty} \\ &= \frac{1}{(y+1)^2}, \text{ for } y > 0. \end{aligned}$$

and $f_Y(y) = 0$, for $y \leq 0$.

(*) : Do integration by parts with $\begin{pmatrix} u = x & du = dx \\ v = e^{-x(y+1)} & dv = (-e^{-x(y+1)})/(y+1)dx \end{pmatrix}$ over $\{x \in \mathbb{R} \mid 0 < x < \infty\}$

$$\text{So we have } \begin{cases} f_{X|Y}(x|y) = \frac{f(x, y)}{f_Y(y)} = \begin{cases} x(y+1)^2 e^{-x(y+1)} & , x > 0, y > 0 \\ 0 & , \text{otherwise} \end{cases} \\ f_{Y|X}(y|x) = \frac{f(x, y)}{f_X(x)} = \begin{cases} xe^{-xy} & , x > 0, y > 0 \\ 0 & , \text{otherwise} \end{cases} \end{cases}.$$

(b)



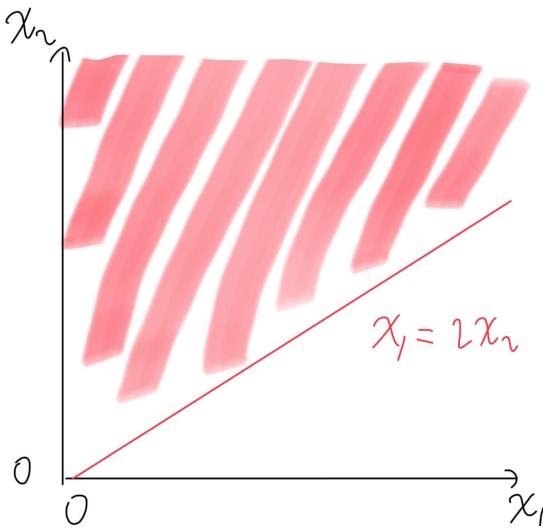
$$\begin{aligned}
 \text{The cdf of } Z \text{ is } F_Z(z) &= \Pr(Z \leq z) \\
 &= \Pr(XY \leq z) \\
 &= \underbrace{\int_0^\infty \int_0^{z/x} xe^{-x(y+1)} dy dx}_{(\star)} \quad (\star) : \left(\text{See the graph.} \right) \\
 &= \int_0^\infty \left(\left(-e^{-x(y+1)} \right) \Big|_{y=0}^{y=z/x} \right) dx \\
 &= \left(1 - e^{-z} \right) \int_0^\infty e^{-x} dx \\
 &= 1 - e^{-z}, \text{ for } z > 0.
 \end{aligned}$$

and $F_Z(z) = 0$, for $z \leq 0$.

Differentiating with respect to z gives :

$$f_Z(z) = \begin{cases} e^{-z}, & \text{if } z > 0. \\ 0, & \text{if } z \leq 0. \end{cases}$$

Problem 48



The desired quantity is :

$$\begin{aligned}
 & Pr(X_{(2)} > 2X_{(1)}) \\
 &= Pr(\max(X_1, X_2) > 2\min(X_1, X_2)) \\
 &= Pr(X_1 > 2X_2 | X_1 > X_2) Pr(X_1 > X_2) + Pr(X_2 > 2X_1 | X_2 > X_1) Pr(X_2 > X_1) \\
 &\stackrel{(*)}{=} 2 \left(\frac{Pr(X_1 > 2X_2)}{Pr(X_1 > X_2)} Pr(X_1 > X_2) \right) \quad (\star) : \text{(Since } X_1, X_2 \text{ are IID.)} \\
 &= 2 Pr(X_1 > 2X_2) \\
 &= 2 \underbrace{\int_0^\infty \int_{2x_2}^\infty f_{X_1, X_2}(x_1, x_2) dx_1 dx_2}_{(\star\star)} \quad (\star\star) : \text{(See the graph.)} \\
 &= 2 \int_0^\infty \int_{2x_2}^\infty \lambda e^{-\lambda x_1} \lambda e^{-\lambda x_2} dx_1 dx_2 \\
 &= 2 \int_0^\infty \lambda e^{-\lambda x_2} \left\{ \left(-e^{-\lambda x_1} \right) \Big|_{2x_2}^\infty \right\} dx_2 \\
 &= 2 \int_0^\infty \lambda e^{-\lambda x_2} e^{-2\lambda x_2} dx_2 \\
 &= 2 \lambda \left(-\frac{1}{3\lambda} e^{-3\lambda x_2} \right) \Big|_0^\infty \\
 &= \frac{2}{3}
 \end{aligned}$$

(Compare this problem with the example in LNp.7 – 17 !)

Problem 52

Note that $\Pr(X_i > a) = 1 - F_{X_i}(a) = (1-p)^a \forall a \in \mathbb{Z}^+$.

(a)

$$\begin{aligned} \Pr\left\{\min(X_1, \dots, X_n) > a\right\} &= \Pr(X_1 > a) \Pr(X_2 > a) \dots \Pr(X_n > a) = \left(\Pr(X_1 > a)\right)^n = (1-p)^{an} \\ \Rightarrow \Pr\left\{\min(X_1, \dots, X_n) \leq a\right\} &= 1 - (1-p)^{an} \end{aligned}$$

(b)

$$\begin{aligned} \Pr\left\{\max(X_1, \dots, X_n) \leq a\right\} &= \Pr(X_1 \leq a) \Pr(X_2 \leq a) \dots \Pr(X_n \leq a) = \left(\Pr(X_1 \leq a)\right)^n \\ \Rightarrow \Pr\left\{\max(X_1, \dots, X_n) \leq a\right\} &= \left(1 - (1-p)^a\right)^n \end{aligned}$$

Theoretical Exercise 22

Note that $\begin{cases} W \sim \Gamma(t, \beta) \Rightarrow f_W(w) = \frac{e^{-\beta w} w^{t-1} \beta^t}{\Gamma(t)}, w > 0. \\ (X_i | W = w) \stackrel{iid}{\sim} \text{Exp}(w) \Rightarrow f(x_i | w) = w e^{-wx_i}, x_i > 0. \end{cases}$

$$\begin{aligned} \Rightarrow f(x_1, \dots, x_n, w) &= f(x_1, \dots, x_n | w) f_W(w) = \prod_{i=1}^n f(x_i | w) f_W(w) \\ &= \left(w^n (e^{-w \sum_{i=1}^n x_i})\right) \frac{e^{-\beta w} w^{t-1} \beta^t}{\Gamma(t)} = \frac{1}{\Gamma(t)} \beta^t w^{n+t-1} \exp\left\{-w\left(\sum_{i=1}^n x_i + \beta\right)\right\}, w > 0, x_i \geq 0 \forall i = 1, \dots, n. \\ \Rightarrow f(x_1, \dots, x_n) &= \int_0^\infty \frac{1}{\Gamma(t)} \beta^t w^{n+t-1} e^{-w(\sum_{i=1}^n x_i + \beta)} dw = \frac{\beta^t}{\Gamma(t)} \frac{\Gamma(n+t)}{(\sum_{i=1}^n x_i + \beta)^{n+t}}, x_i \geq 0, i = 1, \dots, n. \end{aligned}$$

$$\Rightarrow f(w | x_1, \dots, x_n) = \begin{cases} \frac{\left(\frac{\beta^t}{\Gamma(t)} w^{n+t-1} e^{-w(\sum_{i=1}^n x_i + \beta)}\right)}{\left(\frac{\beta^t}{\Gamma(t)} \frac{\Gamma(n+t)}{(\sum_{i=1}^n x_i + \beta)^{n+t}}\right)} = \frac{(\sum_{i=1}^n x_i + \beta)^{n+t}}{\Gamma(n+t)} w^{n+t-1} e^{-w(\sum_{i=1}^n x_i + \beta)}, w > 0, x_i \geq 0, i = 1, \dots, n \\ 0, \text{ otherwise} \end{cases}.$$

$$\text{So } \left(W \mid X_1 = x_1, \dots, X_n = x_n\right) \sim \Gamma(t+n, \sum_{i=1}^n x_i + \beta).$$

Theoretical Exercise 24

Note that $X - \lfloor X \rfloor \in [0, 1)$, so $\begin{cases} \Pr\{\lfloor X \rfloor = n, X - \lfloor X \rfloor \leq x\} = \Pr\{\lfloor X \rfloor = n\} & , \text{ if } 1 \leq x \\ \Pr\{\lfloor X \rfloor = n, X - \lfloor X \rfloor \leq x\} = \Pr\{n \leq X \leq n + x\} & , \text{ if } 0 \leq x < 1 \end{cases}$.

$$\text{When } 1 \leq x, \Pr\{\lfloor X \rfloor = n, X - \lfloor X \rfloor \leq x\} = \Pr\{\lfloor X \rfloor = n\} \times \overbrace{\Pr\{X - \lfloor X \rfloor \leq x\}}^1 \quad (\Delta)$$

Now, we consider the case that $0 \leq x < 1$:

$$(\star) \left\{ \begin{array}{l} \Pr\{\lfloor X \rfloor = n\} = \Pr\{X \in [n, n+1)\} = \Pr\{X < n+1\} - \Pr\{X < n\} \\ = F_X(n+1) - F_X(n) = (1 - e^{-\lambda(n+1)}) - (1 - e^{-\lambda n}) \\ = e^{-\lambda n} - e^{-\lambda(n+1)} = e^{-\lambda n}(1 - e^{-\lambda}) \end{array} \right.$$

$$(\star\star) \left\{ \begin{array}{l} \Pr\{X - \lfloor X \rfloor \leq x\} = \Pr\{0 < X < x\} + \Pr\{1 < X < 1+x\} + \dots \\ = (e^0 - e^{-\lambda x}) + (e^{-\lambda} - e^{-\lambda(1+x)}) + \dots \\ = \sum_{i=0}^{\infty} e^{-\lambda i} (1 - e^{-\lambda x}) \\ = \frac{1 - e^{-\lambda x}}{1 - e^{-\lambda}} \end{array} \right.$$

$$\begin{aligned} \Pr\{\lfloor X \rfloor = n, X - \lfloor X \rfloor \leq x\} &= \Pr\{X \in [n, n+x]\} \\ &= \Pr\{X \leq n+x\} - \Pr\{X \leq n\} \\ &= F_X(n+x) - F_X(n) \\ &= (1 - e^{-\lambda(n+x)}) - (1 - e^{-\lambda n}) \\ &= e^{-\lambda n}(1 - e^{-\lambda x}) \\ &= (e^{-\lambda n}(1 - e^{-\lambda})) \left(\frac{1 - e^{-\lambda x}}{1 - e^{-\lambda}} \right) \\ &\stackrel{(\square)}{=} \Pr\{\lfloor X \rfloor = n\} \Pr\{X - \lfloor X \rfloor \leq x\} \quad \left(\text{By } (\star) \text{ and } (\star\star) \right) \end{aligned}$$

From (Δ) and (\square) , we can easily show that $\Pr\{\lfloor X \rfloor \leq n, X - \lfloor X \rfloor \leq x\} = \Pr\{\lfloor X \rfloor \leq n\} \times \Pr\{X - \lfloor X \rfloor \leq x\}$.

Since the joint cdf of $\lfloor X \rfloor$ and $X - \lfloor X \rfloor$ is a product of their marginal cdf's, $\lfloor X \rfloor$ and $X - \lfloor X \rfloor$ are independent.

To summarize, we have $\Pr\{\lfloor X \rfloor = n, X - \lfloor X \rfloor \leq x\} = \begin{cases} e^{-\lambda n}(1 - e^{-\lambda}) & , \text{ if } 1 \leq x \\ e^{-\lambda n}(1 - e^{-\lambda x}) & , \text{ if } 0 \leq x < 1 \end{cases}$,

and $\lfloor X \rfloor$ and $X - \lfloor X \rfloor$ are independent.