

HOMEWORK #13

SOLUTIONS

① Check Recitation 12 Notes

② X, Y, Z iid, $IP(X=1) = IP(X=2) = 1/2$

(2.a) $A = XYZ$

Given that $S_X = S_Y = S_Z = \{1, 2\}$, it is evident that $S_A = \{1, 2, 4, 8\}$

$$\begin{aligned} IP(A=1) &= IP(XYZ=1) \\ &= IP(X=1, Y=1, Z=1) \\ &= IP(X=1)IP(Y=1)IP(Z=1) \quad (\because X \perp Y \perp Z) \\ &= 1/8 // \end{aligned}$$

$$\begin{aligned} IP(A=2) &= IP(XZY=2) \\ &= IP(X=1, Y=2, Z=1) + IP(X=1, Y=1, Z=2) + IP(X=2, Y=1, Z=1) \\ &= IP(X=1)IP(Y=2)IP(Z=1) + IP(X=1)IP(Y=1)IP(Z=2) + IP(X=2)IP(Y=1)IP(Z=1) \\ &= 3 \times 1/8 = 3/8 // \end{aligned}$$

$$\begin{aligned} IP(A=4) &= IP(XYZ=4) + IP(XYZ=2) + IP(XYZ=1) \\ &= 3 \times 1/8 = 3/8 // \end{aligned}$$

$$\begin{aligned} IP(A=8) &= IP(XZY=8) \\ &= IP(X=2, Y=2, Z=2) \\ &= IP(X=2) \cdot IP(Y=2) \cdot IP(Z=2) \\ &= 1/8 // \end{aligned}$$

(2b) Consider the different combinations of X, Y, Z and observe the value A takes

$$B = XY + YZ + ZX$$

<u>x</u>	<u>y</u>	<u>z</u>	$\Rightarrow$	<u>B</u>
1	1	1		3
1	1	2		5
1	2	1		5
1	2	2		8
2	1	1		5
2	1	2		8
2	2	1		8
2	2	2		12

Since X, Y, Z are iid with $IP(X=1) = IP(X=2) = 1/2$,

$$IP(X=x, Y=y, Z=z) = 1/8 \quad \text{for } x, y, z \in \{1, 2\}$$

Hence by observing the above tabulation,

$$IP(B=3) = 1/8$$

$$IP(B=5) = 2/8$$

$$IP(B=8) = 3/8$$

$$IP(B=12) = 1/8 //$$

$$(4) \quad X \perp\!\!\!\perp Y, \quad X, Y \sim N(0, 1)$$

$$u = X \quad \text{and} \quad v = \begin{cases} Y/X & \text{if } X \neq 0 \\ 0 & \text{if } X = 0 \end{cases}$$

$$\begin{aligned} & \mathbb{P}(u \leq u, v \leq v) \\ &= \mathbb{P}(X \leq u, Y \leq v) \\ &= \mathbb{P}(X \leq u, Y/X \leq v, X \neq 0) + \underbrace{\mathbb{P}(X \leq u, v \leq v, X = 0)}_{\leq \mathbb{P}(X=0)=0} \\ &= \mathbb{P}(X \leq u, Y/X \leq v, X \neq 0) + 0 \\ &= \mathbb{P}(X \leq u, Y/X \leq v, X < 0) + \mathbb{P}(X \leq u, Y/X \leq v, X > 0) \\ &= \mathbb{P}(X \leq u, X < 0, Y \geq vX) + \mathbb{P}(0 < X \leq u, Y \leq vX) \end{aligned}$$

①

for $u < 0$:

$$\begin{aligned} \textcircled{1} &= \mathbb{P}(X \leq u, Y \geq vX) + 0 \\ &= \int_{-\infty}^{\infty} \mathbb{P}(X \leq u, Y \geq vX \mid X = x) f_X(x) dx \\ &= \int_{-\infty}^u \mathbb{P}(Y \geq vx) \frac{1}{\sqrt{2\pi}} e^{-x^2/2} dx \\ &= \int_{-\infty}^u (1 - \Phi(vx)) \frac{1}{\sqrt{2\pi}} e^{-x^2/2} dx, \quad \text{where } \Phi(t) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^t e^{-y^2/2} dy \end{aligned}$$

Clearly $\mathbb{P}(u \leq u, v \leq v)$ is differentiable and hence is of continuous type.

$$f_{u,v}(u, v) = \frac{\partial^2}{\partial v \partial u} \mathbb{P}(u \leq u, v \leq v)$$

$$\frac{\partial}{\partial u} \mathbb{P}(u \leq u, v \leq v) = (1 - \Phi(uv)) \frac{1}{\sqrt{2\pi}} e^{-u^2/2}$$

$$\begin{aligned} \frac{\partial^2}{\partial v \partial u} \mathbb{P}(u \leq u, v \leq v) &= -\frac{1}{\sqrt{2\pi}} e^{-u^2/2} \frac{\partial}{\partial v} (\Phi(uv)) \\ &= -\frac{1}{\sqrt{2\pi}} e^{-u^2/2} \frac{e^{-(vu)^2/2}}{\sqrt{1\pi}} u \end{aligned}$$

$$= \frac{e^{-u^2/2} e^{-(uv)^2/2}}{2\pi} |u| \quad \text{--- (2)}$$

$$\left\{ \text{where we use } \frac{\partial}{\partial v} \Phi(uv) = \frac{\partial}{\partial v} \left[\frac{1}{\sqrt{2\pi}} \int_{-\infty}^{uv} e^{-t^2/2} dt \right] = e^{-\frac{(uv)^2}{2}} u \right\}$$

For $u > 0$

$$\begin{aligned} \textcircled{1} &= P(X \leq u, X < 0, Y \geq vX) + P(X \leq u, Y \leq vX, X > 0) \\ &= P(X < 0, Y \geq vX) + P(0 < X \leq u, Y \leq vX) \\ &= \int_{-\infty}^0 P(X < 0, Y \geq vX | X=x) f_X(x) dx + \int_0^u P(0 < X \leq u, Y \leq vX | X=x) f_X(x) dx \\ &= \int_{-\infty}^0 P(Y \geq vx) f_X(x) dx + \int_0^u P(Y \leq vx) f_X(x) dx \\ &= \int_{-\infty}^0 (1 - \Phi(vx)) \frac{1}{\sqrt{2\pi}} e^{-x^2/2} dx + \int_0^u \Phi(vx) \frac{1}{\sqrt{2\pi}} e^{-x^2/2} dx \\ &= \int_{-\infty}^0 \Phi(-vx) \frac{1}{\sqrt{2\pi}} e^{-x^2/2} dx + \int_0^u \Phi(vx) \frac{1}{\sqrt{2\pi}} e^{-x^2/2} dx \\ &= \int_0^{\infty} \Phi(vx) \frac{1}{\sqrt{2\pi}} e^{-x^2/2} dx + \int_0^u \Phi(vx) \frac{1}{\sqrt{2\pi}} e^{-x^2/2} dx \end{aligned}$$

$$f_{u,v}(u,v) = \frac{\partial^2}{\partial v \partial u} P(u \leq u, V \leq v)$$

$$\frac{\partial}{\partial u} P(u \leq u, V \leq v) = 0 + \frac{e^{-u^2/2}}{\sqrt{2\pi}} \Phi(uv)$$

$$\begin{aligned} \frac{\partial^2}{\partial v \partial u} P(u \leq u, V \leq v) &= \frac{1}{\sqrt{2\pi}} e^{-u^2/2} \frac{1}{\sqrt{2\pi}} e^{-(uv)^2/2} u \\ &= \frac{e^{-u^2/2} e^{-(uv)^2/2}}{2\pi} |u| \quad \text{--- (3)} \end{aligned}$$

from (2) and (3),

$$f_{u,v}(u,v) = \frac{e^{-u^2/2} e^{-(uv)^2/2}}{2\pi} |u|$$

$$f_{uv}(u,v) = \underbrace{\frac{e^{-u^2/2}}{\sqrt{2\pi}}}_{f_u(u)} \cdot \frac{e^{-\frac{(uv)^2}{2}}}{\sqrt{2\pi}} |u|$$

$$\left(\text{Since } U = X, \quad f_v(u) = \frac{1}{\sqrt{2\pi}} e^{-u^2/2} \right)$$

$$\therefore f_{uv}(u,v) = f_u(u) \cdot \frac{e^{-\frac{(uv)^2}{2}}}{\sqrt{2\pi}} |u|$$

Clearly, $f_{uv}(u,v)$ cannot be equal to $f_v(u) \cdot f_v(v)$.
Hence, U and V are not independent.

Note 1

$V = Y/X$, ratio of two iid $N(0,1)$ variables is a Cauchy distribution.

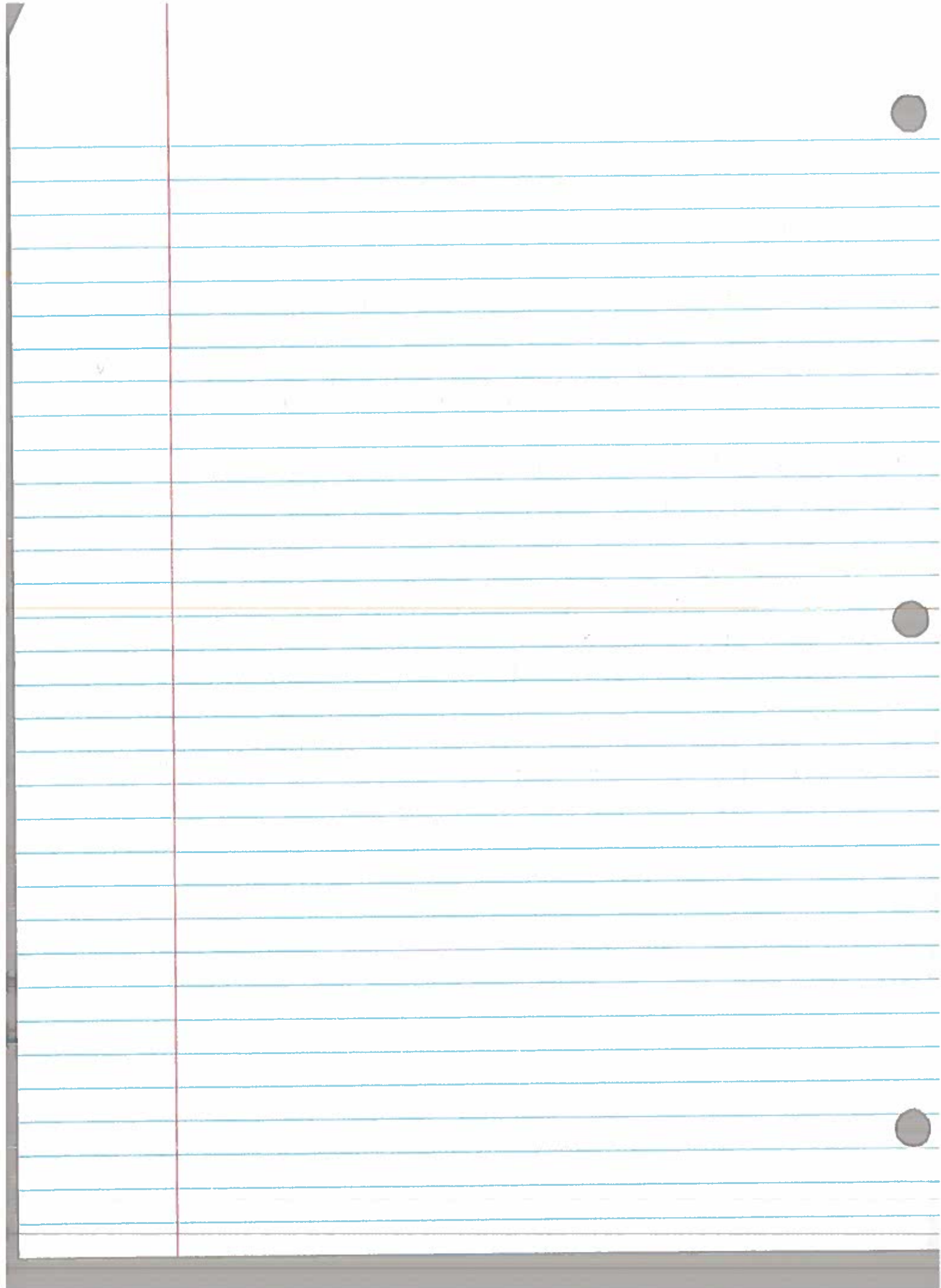
Note 2

In the above calculations, we used differentiation under integral sign.

$$\frac{d}{d\alpha} \int_{f(\alpha)}^{g(\alpha)} \phi(x, \alpha) dx = \int_{f(\alpha)}^{g(\alpha)} \frac{\partial \phi(x, \alpha)}{\partial \alpha} dx + \phi(g(\alpha), \alpha) \frac{dg(\alpha)}{d\alpha} - \phi(f(\alpha), \alpha) \frac{df(\alpha)}{d\alpha}$$

In particular, for 'a' constant,

$$\frac{d}{d\alpha} \int_a^{g(\alpha)} \phi(x) dx = \phi(g(\alpha)) \cdot \frac{dg(\alpha)}{d\alpha} //$$



5) $X \perp\!\!\!\perp Y$

$$X \sim \text{Unif}(0,1) \Rightarrow f_x(x) = 1, \text{ if } 0 < x < 1$$

$$Y \sim \text{Unif}(0,1) \Rightarrow f_y(y) = 1, \text{ if } 0 < y < 1$$

$$\therefore f_{x,y}(x,y) = f_x(x) \cdot f_y(y) = 1, \text{ if } 0 < x < 1, 0 < y < 1$$

Note that $Y > X$ is an event and its complement is $Y \leq X$.

So we can write

$$\begin{aligned} & \mathbb{E} \left[\frac{\max(X,Y)}{\min(X,Y)} \cdot \mathbb{1}[\min(X,Y) \neq 0] \right] \\ &= \mathbb{E} \left[\frac{\max(X,Y)}{\min(X,Y)} \cdot \mathbb{1}[\min(X,Y) \neq 0] \mid Y > X \right] \cdot \mathbb{P}[Y > X] + \mathbb{E} \left[\frac{\max(X,Y)}{\min(X,Y)} \cdot \mathbb{1}[\min(X,Y) \neq 0] \mid Y \leq X \right] \cdot \mathbb{P}[Y \leq X] \end{aligned}$$

$$\mathbb{P}[Y = X] = \iint_B f(x,y) dx dy, \text{ where } B = \{(x,y) \in \mathbb{R}^2 : x=y\}$$

= 0. [The set B has zero area]

$$\therefore \mathbb{E} \left[\frac{\max(X,Y)}{\min(X,Y)} \cdot \mathbb{1}[\min(X,Y) \neq 0] \right]$$

$$= \mathbb{E} \left[\frac{\max(X,Y)}{\min(X,Y)} \cdot \mathbb{1}[\min(X,Y) \neq 0] \mid Y > X \right] \cdot \mathbb{P}[Y > X] + \mathbb{E} \left[\frac{\max(X,Y)}{\min(X,Y)} \cdot \mathbb{1}[\min(X,Y) \neq 0] \mid Y < X \right] \cdot \mathbb{P}[Y < X]$$

Since the rvs X and Y are identically distributed and the above expectation is symmetric in x, y ,

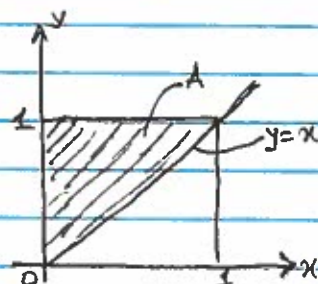
$$\mathbb{E} \left[\frac{\max(X,Y)}{\min(X,Y)} \cdot \mathbb{1}[\min(X,Y) \neq 0] \right]$$

$$= 2 \mathbb{E} \left[\frac{\max(X,Y)}{\min(X,Y)} \cdot \mathbb{1}[\min(X,Y) \neq 0] \mid Y > X \right] \cdot \mathbb{P}[Y > X] = 2 \mathbb{E} \left[\frac{Y}{X} \cdot \mathbb{1}[X \neq 0] \mid Y > X \right] \cdot \mathbb{P}[Y > X]$$

$$= 2 \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \frac{y}{x} \cdot \mathbb{1}[x \neq 0] \cdot f_{x,y|A}(x,y) dx dy \cdot \mathbb{P}[A] \quad [\text{where } A = [Y > X]]$$

$$= 2 \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \frac{y}{x} f_{x,y|A}(x,y) dx dy \cdot \mathbb{P}[A] \quad [\text{by continuity}]$$

$$\begin{aligned} \text{Now, } f_{x,y|A}(x,y) &= \begin{cases} \frac{f_{x,y}(x,y)}{\mathbb{P}[A]}, & \text{if } (x,y) \in A \\ 0, & \text{o/w} \end{cases} \\ &= \begin{cases} \frac{1}{\mathbb{P}[A]}, & \text{if } 0 < x < y < 1 \\ 0, & \text{o/w} \end{cases} \end{aligned}$$



$$\therefore E\left[\frac{\max(X,Y)}{\min(X,Y)} \mid [\min(X,Y) \neq 0]\right]$$

$$= 2 \int_{x=0}^1 \int_{y=x}^1 \frac{y}{x} \cdot \frac{1}{P[A]} dy dx \cdot P[A]$$

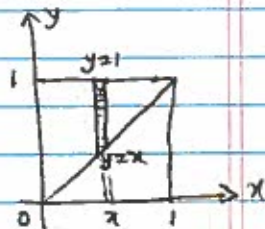
$$= 2 \int_{x=0}^1 \frac{1}{x} \cdot \int_{y=x}^1 y dy dx$$

$$= 2 \int_{x=0}^1 \frac{1}{x} \cdot \left[\frac{y^2}{2}\right]_x^1 dx$$

$$= 2 \int_{x=0}^1 \frac{1-x^2}{2x} dx$$

$$= \left[\ln|x| - \frac{x^2}{2} \right]_0^1$$

$$= \infty.$$



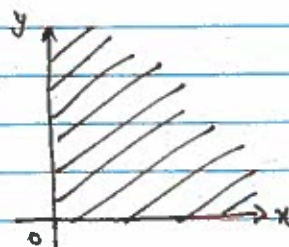
b) $X \perp Y$

$$X \sim \exp(\lambda) \Rightarrow f_X(x) = \lambda e^{-\lambda x}, \text{ if } x > 0$$

$$Y \sim \exp(\mu) \Rightarrow f_Y(y) = \mu e^{-\mu y}, \text{ if } y > 0$$

$$\therefore f_{X,Y}(x,y) = f_X(x) \cdot f_Y(y) \quad \forall x,y$$

$$= \begin{cases} \lambda \mu e^{-\lambda x} \cdot e^{-\mu y}, & \text{if } x > 0, y > 0 \\ 0, & \text{o/w} \end{cases}$$



$$T = \begin{cases} \frac{Y}{X}, & \text{if } X \neq 0 \\ 0, & \text{if } X = 0 \end{cases}$$

a) Clearly $F_T(t) = P[T \leq t] = 0$ if $t \leq 0$.

For $t > 0$,

$$F_T(t) = P[T \leq t] = P[T \leq t \mid X \neq 0] \cdot P[X \neq 0] + P[T \leq t \mid X = 0] \cdot P[X = 0]$$

$$= P[T \leq t \mid X \neq 0] \cdot P[X \neq 0] \quad [\because P[X = 0] = 0 \text{ as } X \text{ is continuous RV}]$$

$$= \mathbb{P}\left[\frac{Y}{X} \leq t \mid X \neq 0\right] \cdot \mathbb{P}[X \neq 0]$$

$$= \mathbb{P}\left[\frac{Y}{X} \leq t, X \neq 0\right]$$

$$= \mathbb{P}[Y \leq tX, X \neq 0] \quad [\because X \sim \exp(\lambda) \Rightarrow \mathbb{P}[X > 0] = 1]$$

$$= \mathbb{P}[Y \leq tX] \quad [\because \mathbb{P}[X \neq 0] = 1]$$

$$= \iint_B f_{X,Y}(x,y) dx dy$$

$$= \int_{y=0}^{\infty} \int_{x=\frac{y}{t}}^{\infty} \lambda \mu e^{-\lambda x} \cdot e^{-\mu y} dx dy$$

$$= \lambda \mu \int_{y=0}^{\infty} e^{-\mu y} \left[\frac{e^{-\lambda x}}{-\lambda} \right]_{x=\frac{y}{t}}^{x=\infty} dy$$

$$= \mu \int_{y=0}^{\infty} e^{-\mu y} \cdot e^{-\lambda y/t} dy$$

$$= \mu \int_{y=0}^{\infty} e^{-(\mu + \frac{\lambda}{t})y} dy$$

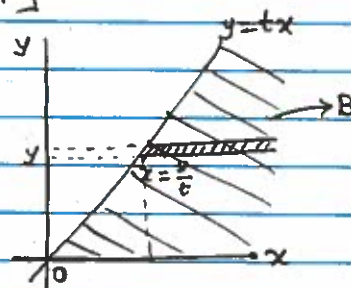
$$= \frac{\mu}{\mu + \frac{\lambda}{t}}$$

$$= \frac{\mu t}{\mu t + \lambda}$$

$$\therefore F_T(t) = \begin{cases} \frac{\mu t}{\mu t + \lambda}, & \text{if } t \geq 0 \\ 0, & \text{if } t \leq 0 \end{cases}$$

b) Clearly $F_T(t)$ is a continuous f.v. of t and differentiable w.r.t. t .

$$\therefore f_T(t) = \frac{d}{dt} F_T(t) = \begin{cases} \frac{\mu \lambda}{(\mu t + \lambda)^2}, & t \geq 0 \\ 0, & \text{o/w} \end{cases}$$



$$\begin{aligned}
c) \mathbb{E}[T] &= \int_{-\infty}^{\infty} t f_T(t) dt \\
&= \int_0^{\infty} \frac{\mu \lambda}{(\mu t + \lambda)^2} \cdot t dt \\
&= \int_{\lambda}^{\infty} \frac{\lambda(x-\lambda)}{\mu x^2} dx \quad [\text{with } x = \mu t + \lambda] \\
&= \frac{\lambda}{\mu} \int_{\lambda}^{\infty} \left(\frac{1}{x} - \frac{\lambda}{x^2} \right) dx \\
&= \frac{\lambda}{\mu} \left[\ln|x| + \frac{\lambda}{x} \right]_{x=\lambda}^{\infty} \\
&= \infty.
\end{aligned}$$

7) $R_1, \dots, R_n$ are mutually independent and identically distributed ; $n \geq 3, n \in \mathbb{N}$

$R_k \sim \exp(\lambda)$, $\forall k, k \geq 3$.

$$\begin{aligned}
a) \mathbb{P}[R_1 > 2R_2] &= \int_{-\infty}^{\infty} \mathbb{P}[R_1 > 2R_2 | R_2 = x] f_{R_2}(x) dx \\
&= \int_0^{\infty} \mathbb{P}[R_1 > 2R_2 | R_2 = x] \cdot \lambda e^{-\lambda x} dx \\
&= \int_0^{\infty} \mathbb{P}[R_1 > 2x | R_2 = x] \cdot \lambda e^{-\lambda x} dx \\
&= \int_0^{\infty} \mathbb{P}[R_1 > 2x] \cdot \lambda e^{-\lambda x} dx \quad [\because R_1 \perp R_2] \\
&= \int_0^{\infty} e^{-\lambda \cdot 2x} \cdot \lambda e^{-\lambda x} dx \quad [\text{Recall; if } X \sim \exp(\lambda), \\
&= \lambda \int_0^{\infty} e^{-3\lambda x} dx \quad \begin{aligned} &F_X(x) = \mathbb{P}[X \leq x] = 1 - e^{-\lambda x} \\ &\Rightarrow \mathbb{P}[X > x] = 1 - F_X(x) \\ &= e^{-\lambda x} \end{aligned} \\
&= \frac{1}{3}.
\end{aligned}$$

$$\begin{aligned}
b) \quad \mathbb{P}[R_1 > 2R_2 > 3R_3] &= \int_{-\infty}^{\infty} \mathbb{P}[R_1 > 2R_2 > 3R_3 \mid R_2 = x] f_{R_2}(x) dx \\
&= \int_0^{\infty} \mathbb{P}[R_1 > 2x > 3R_3 \mid R_2 = x] \lambda e^{-\lambda x} dx \\
&= \int_0^{\infty} \mathbb{P}[R_1 > 2x, R_3 < \frac{2}{3}x] \lambda e^{-\lambda x} dx \quad [\because R_1, R_2, R_3 \text{ are mutually independent}] \\
&= \int_0^{\infty} \mathbb{P}[R_1 > 2x] \cdot \mathbb{P}[R_3 < \frac{2}{3}x] \cdot \lambda e^{-\lambda x} dx \quad [\because R_1 \perp R_3] \\
&= \int_0^{\infty} e^{-\lambda \cdot 2x} \cdot (1 - e^{-\lambda \cdot \frac{2}{3}x}) \cdot \lambda e^{-\lambda x} dx \\
&= \lambda \int_0^{\infty} e^{-3\lambda x} (1 - e^{-\frac{2\lambda x}{3}}) dx \\
&= \lambda \int_0^{\infty} e^{-3\lambda x} - \lambda \int_0^{\infty} e^{-\frac{11\lambda x}{3}} dx \\
&= \lambda \cdot \frac{1}{3\lambda} - \lambda \cdot \frac{1}{\frac{11\lambda}{3}} \\
&= \frac{1}{3} - \frac{3}{11} \\
&= \frac{2}{33}
\end{aligned}$$

$$c) \quad R_1 > 2R_2 > \dots > nR_n \Leftrightarrow \lambda R_1 > \lambda \cdot 2R_2 > \dots > \lambda \cdot nR_n \quad [\because \lambda > 0]$$

[$\Leftrightarrow$ means if and only if]

$$\text{So, } \mathbb{P}[R_1 > 2R_2 > \dots > nR_n] = \mathbb{P}[\lambda R_1 > 2\lambda R_2 > \dots > n\lambda R_n]$$

Now, $\mathbb{P}[\lambda R_n \leq x]$, for $x \geq 0$ $ \begin{aligned} &= \mathbb{P}[R_n \leq \frac{x}{\lambda}] \\ &= 1 - e^{-\lambda \cdot \frac{x}{\lambda}} \quad [\because R_n \sim \exp(\lambda)] \\ &= 1 - e^{-x} \end{aligned} $	$\mathbb{P}[\lambda R_n \leq x]$, for $x < 0$ $ \begin{aligned} &= \mathbb{P}[R_n \leq \frac{x}{\lambda}] \\ &= 0 \quad [\because R_n \sim \exp(\lambda)] \end{aligned} $
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$$\therefore F_{R_n}(x) = \begin{cases} 1 - e^{-x}, & \text{if } x \geq 0 \\ 0, & \text{if } x < 0 \end{cases}$$

$$\Rightarrow \lambda R_n \sim \exp(1), \forall n \geq 3.$$

Also, $R_1, \dots, R_n$ are mutually independent $\Rightarrow \lambda R_1, \dots, \lambda R_n$ are mutually independent

$\therefore \mathbb{P}[\lambda R_1 > \dots > \lambda R_n]$ is independent of λ , as the λR_n 's are iid with exp(1) which does not depend on λ .

Hence, $\mathbb{P}[R_1 > 2R_2 > \dots > nR_n]$ does not depend on λ .

d) The R_k 's being iid are exchangeable.

So, $\mathbb{P}[R_1 < R_2 < \dots < R_n]$ is same as the probability of any ordering of the rvs in increasing order, i.e.,

$$\mathbb{P}[R_1 < R_2 < \dots < R_n] = \mathbb{P}[R_{k_1} < R_{k_2} < \dots < R_{k_n}]$$

where the index set $\{k_1, \dots, k_n\}$ is any reordering of $\{1, \dots, n\}$.

All such reorderings $\{R_{k_1} < \dots < R_{k_n}\}$ and the events where tie occurs between two or more R_k 's constitute the sample space Ω .

Since the R_k 's are all continuous rvs, the probability of ties are zero.

$$\text{So, } \sum_{\substack{\{k_1, \dots, k_n\} \\ \text{is any ordering} \\ \text{of } \{1, \dots, n\}}} \mathbb{P}[R_{k_1} < \dots < R_{k_n}] = 1$$

$$\Rightarrow \mathbb{P}[R_{k_1} < \dots < R_{k_n}] = \frac{1}{n!} \quad [\because \text{The no. of ways to reorder the set } \{1, \dots, n\} \text{ is } n!]$$

and all such $\mathbb{P}[R_{k_1} < \dots < R_{k_n}]$ has equal probability because the R_k 's are iid.

$$\text{Hence, } \mathbb{P}[R_1 < \dots < R_n] = \frac{1}{n!}$$

because $\{1, \dots, n\}$ is one such reordering of $\{1, \dots, n\}$.