

Probability 2 - Ex 0 Solution

Ex 1

(a) $0.4 + 0.2 + 0.1 + a + b + c = 1 \Rightarrow a + b + c = 0.3$

$$\begin{aligned} E[X] &= 0.4 + 2 \times 0.2 + 3 \times 0.1 + 4a + 5b + 6c \\ &= 0.4 + 0.4 + 0.3 + 4(a+b+c) + b + 2c \\ &= 2.3 + b + 2c = 2.3 + \underbrace{(b+c)}_{\leq 0.3} + c \leq 2.3 + 0.3 + 0.3 = 2.9 \end{aligned}$$

False

(b) False

(c) NO

(d) $E[X] = 2.3 + b + 2c$

$$a + b + c = b + 2c = 0.3$$

$$\text{So } E[X] = 2.6.$$

Ex 2

(a) $P(X=1) = P\left(\bigcup_{j \in \{0,1\}} X=1 \ \& \ Y=j\right) =$

$$= \sum_{j=1}^n P\left(X=1 \ \& \ Y \in \left[\frac{j-1}{n}, \frac{j}{n}\right)\right)$$

$$= \sum_{j=1}^n P\left(Y \in \left[\frac{j-1}{n}, \frac{j}{n}\right)\right) \cdot P\left(X=1 \mid Y \in \left[\frac{j-1}{n}, \frac{j}{n}\right)\right)$$

$$\leq \sum_{j=1}^n \frac{1}{n} \times \frac{j}{n} = \frac{1}{n^2} \sum_{j=1}^n j = \frac{n(n+1)}{2n^2} \xrightarrow{n \rightarrow +\infty} \frac{1}{2}$$

Similarly, $P(X=1) = \sum_{j=1}^n P\left(Y \in \left[\frac{j-1}{n}, \frac{j}{n}\right)\right) \cdot P\left(X=1 \mid Y \in \left[\frac{j-1}{n}, \frac{j}{n}\right)\right)$
 $\geq \sum_{j=1}^n \frac{1}{n} \times \frac{j-1}{n} = \frac{1}{n^2} \sum_{j=1}^n (j-1) = \frac{n(n-1)}{2n^2} \rightarrow \frac{1}{2}$

$$\text{So } P(X=1) = 1/2$$

Other way of computing $P(X=1)$

$$P(X=1) = \int P(X=1 \mid Y=y) f_Y(y) dy$$

$$= \int_0^1 y \, dy = \frac{1}{2}$$

Note: these two techniques are essentially the same

(b) Because Y is a continuous r.v. we have $P(Y=1)=0$.
So $P(Y=1|X=1) = P(X=1|Y=1) \frac{P(Y=1)}{P(X=1)} = 1 \times \frac{0}{1/2} = 0$

(c) $P(Y \geq 0.75 | X=1) = \frac{P(Y \geq 0.75 \text{ \& } X=1)}{P(X=1)}$

$$= \int_{0.75}^1 P(X=1|Y=y) f_Y(y) \, dy \frac{1}{P(X=1)} = \int_{0.75}^1 (1-y) \, dy \times 2 =$$

$$= 2 \left[\frac{1}{2} y^2 \right]_{y=3/4}^{y=1} = 2 \left(1 - \frac{1}{2} - \frac{3}{4} + \frac{9}{32} \right) = 2 - 1 - \frac{3}{2} + \frac{9}{16} = -\frac{1}{2} + \frac{9}{16} = \frac{1}{16}$$

Ex 3

(a) $Y \sim \text{Unif}(0,1)$, $f_X(u) = 2u$ ← density function for $u \in (0,1)$
for $t \in (0,1)$

$$P(X^2 \leq t) = P(|X| \leq \sqrt{t}) = P(X \leq \sqrt{t})$$

$$= \int_0^{\sqrt{t}} 2u \, du = \left[u^2 \right]_{u=0}^{u=\sqrt{t}} = \sqrt{t}^2 = t$$

$P(Y \leq t) = t$ for $t \in (0,1)$.

So $P(X^2 \leq t) = P(Y \leq t)$ is true.

(b) $P(X \leq t) = \int_0^t 2u \, du = t^2$ for $t \in (0,1)$

$P(Y^2 \leq t) = \int_0^{\sqrt{t}} 1 \, du = \sqrt{t}$ for $t \in (0,1)$

(c) $P(X^2 \leq 1) = P(Y^2 \leq 1)$

So $P(X \leq t) \neq P(Y \leq t)$ in general.
is false

(c) We just have information on the law of X^2 and Y , not the full information, so we don't know if $P(X^2=Y)=1$

(d) False, because $X=Y^2$ a.s. $\Rightarrow X \stackrel{(d)}{=} Y^2$, but $X \stackrel{(d)}{=} Y^2$ is false (see (b))

Ex 4

(a) False. Consider a r.v. $X \in (1, +\infty)$ with density

$$f_X(u) = \frac{1}{u^2}.$$

Note that $\int_1^{+\infty} f_X(u) du = \left[-u^{-1} \right]_1^{+\infty} = 1$ but

$$E[X] = \int_1^{+\infty} f_X(u) \cdot u du = \left[\ln u \right]_1^{+\infty} = +\infty.$$

Let $Y \perp\!\!\!\perp X$ be s.t. $P(Y=1) = P(Y=-1) = 1/2$

Then $E[X \cdot Y]$ is not well defined.

The expected value of a r.v. X is well defined in $\overline{\mathbb{R}}$ when:

- $X \geq 0$ or $X \leq 0$ or
- $E[|X|] < +\infty$ (that is, $X \in L^1$).

(b) True. This is a consequence of Hölder's inequality

(c) True. This is a consequence of (b)

(d) Because $X \leq 1$ a.s., $\|X\|_{\infty} = 1$ so $X \in L^1, L^2, L^\infty$

(e) Directly computing

$$\|X\|_\infty = \infty \quad \|X\|_2 = \infty \quad \|X\|_1 = 2$$

So $X \in L^1$, but $X \notin L^2, L^\infty$

$$\textcircled{f} \quad \mathbb{E}[|X|] = \sum_{j=0}^{+\infty} e^{-1} \frac{1}{j!} \cdot j = e^{-1} \sum_{j=1}^{+\infty} \frac{1}{(j-1)!} = e^{-1} \cdot e = 1$$

$$\begin{aligned} \mathbb{E}[X^2] &= \mathbb{E}[X(X-1) + X] = \\ &= e^{-1} \left(\sum_{j=0}^{+\infty} \frac{1}{j!} j(j-1) + \sum_{j=0}^{+\infty} \frac{1}{j!} j \right) \\ &= e^{-1} \left(\sum_{j=2}^{+\infty} \frac{1}{(j-2)!} + \sum_{j=1}^{+\infty} \frac{1}{(j-1)!} \right) = 2 \end{aligned}$$

$\|X\|_\infty =$ essential supremum of $X = +\infty$
So $X \in L^1, L^2$ but $X \notin L^\infty$.

$$\textcircled{g} \quad \mathbb{E}[|X|] = \sum_{k=1}^{+\infty} \frac{6}{\pi^2} \frac{1}{k} = +\infty \quad \text{so } X \notin L^1, L^2, L^\infty$$

Ex 5

$$\textcircled{a} \quad X_n \xrightarrow{L^p} Y \text{ if } \|X_n - Y\|_p \rightarrow 0.$$

$$\|X_n - Y\|_p = \|X_n\|_p = \mathbb{E}[|X_n|^p]^{\frac{1}{p}} = \mathbb{E}[X_n]^{\frac{1}{p}} = \frac{1}{n^{1/p}} \rightarrow 0$$

So $X_n \xrightarrow{L^p} Y$ for any $p \geq 1$.

\textcircled{b} We simply observe that $Y_n \sim \text{Ber}(1/n)$. That is

$$\mathbb{P}(Y_n = 1) = \mathbb{P}(Y \leq 1/n) = 1/n$$

$$\mathbb{P}(Y_n = 0) = \mathbb{P}(Y > 1/n) = 1 - 1/n$$

From (a) and the diagram (Figure 1), we have that

$$X_n \xrightarrow{L^0} Y \Rightarrow X_n \xrightarrow{(d)} Y$$

Because $X_n \stackrel{(d)}{=} Y_n$, $Y_n \xrightarrow{(d)} Y$.

(c) From (a) and the diagram (Figure 1), $X_n \xrightarrow{P} Y$

$$Y_n \xrightarrow{P} Y \quad \text{if} \quad P(|Y_n - Y| < \varepsilon) \xrightarrow{n} 1 \quad \text{for any } \varepsilon > 0.$$

$$P(|Y_n - Y| < \varepsilon) = P(|Y_n| < \varepsilon) = P(|Y_n| = 0) = 1 - \frac{1}{n} \rightarrow 1$$

(d) First, note that $|Y_n| \leq 1$ a.s., so Y_n is a dominated sequence of r.v.

Also, for any $\omega \in \Omega$, $Y_n = 0$ whenever $n \geq \frac{1}{U(\omega)}$, by definition of $Y_n = 11[U \leq \frac{1}{n}]$.

Since $P(U=0) = 0$, $P(\lim Y_n = Y) = 1$.

So by the dominated convergence theorem we have that $Y_n \xrightarrow{L^1} Y$.

(e) We use again the DCT. $|Z_n| \leq 1$ a.s.

and $Z_n = \frac{1}{n} Z \xrightarrow{n} 0$, so $Z_n \xrightarrow{L^1} 0$ and $E[Z_n] \rightarrow E[0] = 0$.