

1 Exercise 5.10.5 (a)

Proof:

(1) Show that $\int_{\mathbb{R}} F(x)F(dx) = 1/2$. Actually,

$$\begin{aligned}
 \int_{\mathbb{R}} F(x)F(dx) &= \int_{\mathbb{R}} \left[\int_{(-\infty, x]} F(dy) \right] F(dx) \\
 &= \int_{\mathbb{R}} \left[\int_{\mathbb{R}} 1_{(-\infty, x]}(y) F(dy) \right] F(dx) \\
 \text{(by Fubini)} &= \int_{\mathbb{R}} \left[\int_{\mathbb{R}} 1_{[y, \infty)}(x) F(dx) \right] F(dy) \\
 \text{(F is continuous)} &= \int_{\mathbb{R}} [1 - F(y)] F(dy) \\
 &= \int_{\mathbb{R}} F(dy) - \int_{\mathbb{R}} F(y)F(dy) \\
 &= 1 - \int_{\mathbb{R}} F(x)F(dx) \\
 \implies & 2 \int_{\mathbb{R}} F(x)F(dx) = 1 \\
 \implies & \int_{\mathbb{R}} F(x)F(dx) = \frac{1}{2}
 \end{aligned}$$

Note: An alternative proof is based on Exercise 3.4.5 on page 86.

$$\begin{aligned}
 X &\sim F, \text{ which is continuous} \\
 \implies Y = F(X) &\sim \text{Uniform}[0, 1] \\
 \implies \int_{\mathbb{R}} F(x)F(dx) &= E[F(X)] = F[Y] = \frac{1}{2}
 \end{aligned}$$

(2) Show that $P[X_1 \leq X_2] = 1/2$ if X_1, X_2 iid $\sim F$. Actually,

$$\begin{aligned}
 P[X_1 \leq X_2] &= \int_{\mathbb{R}^2} 1_{[x_1 \leq x_2]}(x_1, x_2) d(F \times F) \\
 \text{(by Fubini)} &= \int_{\mathbb{R}} \left[\int_{\mathbb{R}} 1_{(-\infty, x_2]}(x_1) F(dx_1) \right] F(dx_2) \\
 &= \int_{\mathbb{R}} F(x_2)F(dx_2) \\
 \text{(by (1))} &= \frac{1}{2}
 \end{aligned}$$

(3) Show that $E(F(X_1)) = 1/2$. Actually,

$$E(F(X_1)) = \int_{\mathbb{R}^2} F(x_1) d(F \times F)$$

$$\begin{aligned}
\text{(by Fubini)} &= \int_{\mathbb{R}} \left[\int_{\mathbb{R}} F(x_1) F(dx_1) \right] F(dx_2) \\
\text{(by (1))} &= \int_{\mathbb{R}} \frac{1}{2} F(dx_2) \\
&= \frac{1}{2}
\end{aligned}$$

□

2 Exercise 5.10.6 (a) (b) (d)

Proof: First note: since $X \in L_1$, that is, $E(|X|) < \infty$, then $|X| \in L_1$ and $P[|X| = \infty] = 0$. Otherwise, $P[|X| = \infty] > 0$ implies that $E(|X|) = \infty$.

(a) Based on the above, $\lim_{n \rightarrow \infty} \downarrow \{ |X| > n \} = N$ with $P(N) = 0$. Hence we have

$$\lim_{n \rightarrow \infty} X I_{\{|X| > n\}} = 0 \quad \text{a.s.}$$

By Dominated Convergence Theorem,

$$\lim_{n \rightarrow \infty} \int_{\{|X| > n\}} X dP = \lim_{n \rightarrow \infty} \int_{\Omega} X I_{\{|X| > n\}} dP = 0.$$

(b) Decompose

$$\int_{A_n} |X| dP = \int_{A_n \cap \{|X| \leq M\}} |X| dP + \int_{A_n \cap \{|X| > M\}} |X| dP = I_1 + I_2.$$

For any fixed $\varepsilon > 0$, by (a), we can choose M large enough such that $I_2 < \varepsilon$. With this fixed M , we look at I_1 . By monotonicity of integration, we have $I_1 < MP\{A_n\}$. Since $P\{A_n\} \rightarrow 0$, we know that for large enough n , we have $P\{A_n\} < \frac{\varepsilon}{M}$. This implied that when n large enough, we have $I_1 < \varepsilon$. Summing up, we conclude that for n large enough, we have

$$\int_{A_n} |X| dP < \varepsilon$$

for any fixed $\varepsilon > 0$. Since ε is arbitrary, we can conclude that

$$\lim_{n \rightarrow \infty} \int_{A_n} |X| dP = 0.$$

(d) Since $0 = \text{Var}(X) = E(X - E(X))^2 = \int_{\Omega} (X - E(X))^2 dP$, then by part (c),

$$0 = P(\Omega \cap [(X - E(X))^2 > 0]) = P(\Omega \cap [X \neq E(X)]) = P[X \neq E(X)].$$

Thus $P[X = E(X)] = 1$ so that X is equal to the constant $E(X)$ with probability 1. □

3 Exercise 5.10.15

(a) **Proof:**

$$0 \leq nE\left(\frac{1}{X}1_{[X>n]}\right) = E\left(\frac{n}{X}1_{[X>n]}\right) \leq E(1 \cdot 1_{[X>n]}) = P[X > n].$$

Since $P[0 \leq X < \infty] = 1$, then

$$P[X > n] = P([X > n] \cap [0 \leq X < \infty]) = P[n < X < \infty] \rightarrow 0$$

because $[n < X < \infty] \downarrow \emptyset$. Therefore,

$$\lim_{n \rightarrow \infty} nE\left(\frac{1}{X}1_{[X>n]}\right) = 0.$$

□

(b) **Proof:** First show that $\frac{1}{nX}1_{[X>n^{-1}]} \rightarrow 0$ as n goes to ∞ . Actually, $\forall \omega \in \Omega$, if $X(\omega) > 0$, then $\frac{1}{nX(\omega)} \rightarrow 0$ as n goes to ∞ ; otherwise, $X(\omega) \leq 0$, then $1_{[X>n^{-1}]}(\omega) = 0$ for each n .

On the other hand, $|\frac{1}{nX}1_{[X>n^{-1}]}| = \frac{1}{nX}1_{[nX>1]} \leq 1$. By DCT (Dominated Convergence Theorem on page 133), $E\left(\frac{1}{nX}1_{[X>n^{-1}]}\right) \rightarrow 0$ as n goes to ∞ . That is,

$$\lim_{n \rightarrow \infty} n^{-1}E\left(\frac{1}{X}1_{[X>n^{-1}]}\right) = 0.$$

□

4 Exercise 5.10.22

(a) **Proof:** Note that for a.e. $\omega \in \Omega$, we have $X(\omega) = X(\omega) - 0 = \int_{(0, X(\omega)]} dt$. Hence

$$\begin{aligned} E(X) &= \int_{\Omega} X(\omega)P(d\omega) = \int_{\Omega} \left(\int_{[0, X(\omega)]} dt \right) P(d\omega) \\ &= \int_{\Omega} \left(\int_{[0, \infty)} I_{[0, X(\omega)]}(t) dt \right) P(d\omega) = \int_{[0, \infty)} \left(\int_{\Omega} I_{[0, X(\omega)]}(t) P(d\omega) \right) dt \\ &= \int_{[0, \infty)} P[X(\omega) > t] dt. \end{aligned}$$

The proof is completed.

(b) The proof is similar to (a) once we note that $X^\alpha(\omega) = \alpha \int_{[0, X(\omega)]} t^{\alpha-1} dt$ for a.e. $\omega \in \Omega$.

5 Exercise 5.10.28

Proof: If

$$E(\sqrt[n=1]{X_n}) < \infty$$

simply define $Y = \bigvee_{n=1}^{\infty} |X_n|$.

Conversely, if there is a random variable $0 \leq Y \in L^1$, such that

$$P[|X_n| \leq Y] = 1,$$

then

$$\bigvee_{n=1}^{\infty} |X_n| \leq Y \quad \text{a.s.}$$

Hence,

$$E(\bigvee_{n=1}^{\infty} |X_n|) \leq EY < \infty.$$

The proof is completed.