Homework 4, MATH 9010

Due on Thursday, September 15 Shuai Wei

Problem 1 Let $X \in \mathcal{L}^1$ and let A_n be a sequence of events such that $\lim_{n\to\infty} P(A_n) = 0$. Show that $\lim_{n\to\infty} E\{X\mathbb{1}_{A_n}\} = 0$.

Proof. Since X is integrable, |X| is also integrable.

Then

$$E(|X|) < \infty$$
.

Let $n \in \mathbb{N}$, by Markov inequality,

$$P(|X| = \infty) \le P(|X| > n) \le \frac{E(|X|)}{n}$$

Since

$$\lim_{n \to \infty} \frac{E(|X|)}{n} = 0,$$

we have

$$P(|X| = \infty) = 0.$$

So

$$E(|X|\mathbb{1}_{\{|X|=\infty\}})=0$$

by what we have shown in class. Besides,

$$\begin{split} E(|X|\mathbb{1}_{A_n}) &= E(|X|\mathbb{1}_{A_n\{|X|=\infty\}}) + E(|X|\mathbb{1}_{A_n\{|X|<\infty\}}) \\ &\leq E(|X|\mathbb{1}_{\{|X|=\infty\}}) + E(|X|\mathbb{1}_{A_n\{|X|<\infty\}}) \\ &= E(|X|\mathbb{1}_{A_n\{|X|<\infty\}}) \end{split}$$

There exist M > 0 such that

$$|X|\mathbb{1}_{A_n\{|X|<\infty\}} \le M\mathbb{1}_{A_n\{|X|<\infty\}} = M\mathbb{1}_{A_n}.$$

So

$$E(|X|\mathbb{1}_{A_n}) \le E(|X|\mathbb{1}_{A_n\{|X|<\infty\}}) \le MP(A_n).$$

Then

$$\lim_{n \to \infty} E\{|X|\mathbb{1}_{A_n}\} \le M \lim_{n \to \infty} P(A_n) = 0.$$

Thus

$$\lim_{n \to \infty} E\{|X|\mathbb{1}_{A_n}\} = 0.$$

Since

$$\lim_{n\to\infty} E\{|X|\mathbbm{1}_{A_n}\} = \lim_{n\to\infty} E\{X^+\mathbbm{1}_{A_n}\} + \lim_{n\to\infty} E\{X^-\mathbbm{1}_{A_n}\} = 0,$$

we have

$$\lim_{n\to\infty} E\{X^+\mathbb{1}_{A_n}\} = \lim_{n\to\infty} E\{X^-\mathbb{1}_{A_n}\} = 0$$

Therefore,

$$\lim_{n \to \infty} E\{X \mathbb{1}_{A_n}\} = \lim_{n \to \infty} E\{X^+ \mathbb{1}_{A_n}\} - \lim_{n \to \infty} E\{X^- \mathbb{1}_{A_n}\} = 0$$

Problem 2 Suppose $X_n, n \ge 1$ and X are uniformly bounded random; i.e. there exists a constant K such that

$$|X_n| \lor |X| \le K$$
.

If $X_n \to X$ as $n \to \infty$, show by means of dominated convergence that

$$E|X_n - X| \to 0.$$

Proof. For any $n \in \mathbb{N}$,

$$|X_n - X| \le |X_n| + |X| \le M + M = 2M,$$

and

Constant M is integrable since

$$E(M) = E(M\mathbb{1}_{\Omega}) = M \times P(\Omega) = M < \infty.$$

Since $X_n \to X$, $X_n - X \to 0$.

By Dominated Convergence Theorem,

$$\lim_{n \to \infty} E(X_n - X) = E(0) = 0.$$

Namely,

$$E(X_n - X) \to 0.$$

Problem 3 For $X \geq 0$, let

$$X_n^* = \sum_{k=1}^{\infty} \frac{k}{2^n} \mathbb{1}_{\left[\frac{k-1}{2^n} \le X < \frac{k}{2^n}\right]}.$$

Show

$$E(X_n^*) \downarrow E(X)$$
.

Proof. We first show $\{X_n^*\}_{n\geq 1}$ is monotone decreasing. For $k, n \in \mathbb{N}$ and $k, n \geq 1$,

$$\left[\frac{k-1}{2^n}, \frac{k}{2^n}\right) = \left[\frac{k-1}{2^n}, \frac{k-1/2}{2^n}\right) \bigcup \left[\frac{k-1/2}{2^n}, \frac{k}{2^n}\right) \\
= \left[\frac{2k-2}{2^{n+1}}, \frac{2k-1}{2^{n+1}}\right) \bigcup \left[\frac{2k-1}{2^{n+1}}, \frac{2k}{2^{n+1}}\right)$$

So

$$X_n^* = \sum_{k=1}^{\infty} \frac{2k}{2^{n+1}} \mathbb{1}_{\left[\frac{2k-2}{2^{n+1}} \le X < \frac{2k-1}{2^{n+1}}\right]} + \frac{2k}{2^{n+1}} \mathbb{1}_{\left[\frac{2k-1}{2^{n+1}} \le X < \frac{2k}{2^{n+1}}\right]}.$$

On the other hand,

$$X_{n+1}^* = \sum_{k=1}^\infty \frac{2k-1}{2^{n+1}} \mathbb{1}_{\left[\frac{2k-2}{2^{n+1}} \le X < \frac{2k-1}{2^{n+1}}\right]} + \frac{2k}{2^{n+1}} \mathbb{1}_{\left[\frac{2k-1}{2^{n+1}} \le X < \frac{2k}{2^{n+1}}\right]}.$$

Thus, $X_n^* \ge X_{n+1}^*$. We claim $X_n^* \downarrow X$.

For any $\epsilon > 0$, there exists $N \in \mathbb{Z}$ such that $\frac{1}{2^N} < \epsilon$.

Then by the definition of X_N^* , $X_N^* \geq X$ and so

$$|X_N^* - X| = X_N^* - X < \frac{1}{2^N} < \epsilon.$$

So when n > N, $X_n < X_N$ and

$$|X_N^* - X| = X_n^* - X < X_N^* - X < \epsilon.$$

Therefore, $X_n^* \downarrow X$.

Let $Y_n^* = X_1^* - X_n^* \ge 0$, then Y_n^* is monotone increasing.

$$Y_n^* \uparrow X_1^* - X$$
.

By the Monotone Convergence Theorem, we have

$$E(Y_n^*) \uparrow E(X_1^* - X).$$

Then

$$E(X_1^*) - E(X_n^*) \uparrow E(X_1^*) - E(X).$$

since $E(Y_n^*) = E(X_1^*) - E(X_n^*)$.

As a result,

$$E(X_n^*) \downarrow E(X)$$
.

Problem 4 Suppose X is a non-negative random variable satisfying

$$P[0 \le X < \infty] = 1.$$

Show

(a)

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

(b)

$$\lim_{n\to\infty} n^{-1} E\Big(\frac{1}{X}\mathbb{1}_{[X>n^{-1}]}\Big) = 0.$$

(a) Proof.

$$0 \le nE\left(\frac{1}{X}\mathbb{1}_{[X>n]}\right) = n\int_{\mathbb{1}_{[X>n]}} \frac{1}{X(\omega)} dP(\omega)$$
$$= \int_{\mathbb{1}_{[X>n]}} \frac{n}{X(\omega)} dP(\omega)$$
$$\le \int_{\mathbb{1}_{[X>n]}} dP(\omega)$$
$$= P(\mathbb{1}_{[X>n]})$$

So

$$0 \leq \lim_{n \to \infty} nE\left(\frac{1}{X}\mathbb{1}_{[X>n]}\right) \leq \lim_{n \to \infty} P(\mathbb{1}_{[X>n]}) = P[X=\infty] = 1 - P[0 \leq X < \infty] = 0.$$

Thus

$$\lim_{n\to\infty} nE\Big(\frac{1}{X}\mathbb{1}_{[X>n]}\Big)=0.$$

(b) Proof.

When $X = 0, \mathbb{1}_{[X > n^{-1}]} = 0$, then $\frac{1}{X} \mathbb{1}_{[X > n^{-1}]} = 0$.

So $n^{-1}E(\frac{1}{X}\mathbb{1}_{[X>n^{-1}]})=0$. When $X \neq 0$, P(X=0)=0.

$$0 \le n^{-1} E\left(\frac{1}{X} \mathbb{1}_{[X > n^{-1}]}\right) = n^{-1} E\left(\frac{1}{X} \mathbb{1}_{\left[\frac{1}{X} < n\right]}\right)$$
$$\le n^{-1} E\left(n \mathbb{1}_{\left[\frac{1}{X} < n\right]}\right)$$
$$= E\left(\mathbb{1}_{\left[\frac{1}{X} < n\right]}\right)$$
$$= P\left(\frac{1}{X} < n\right)$$

So

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

Thus,

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

Problem 5 Suppose $\{p_k, k \geq 0\}$ is a probability mass function on $\{0, 1, ...\}$ and define the generating function

$$P(s) = \sum_{k=0}^{\infty} p_k s^k, \quad 0 \le s \le 1.$$

Prove using dominated convergence that

$$\frac{d}{ds}P(s) = \sum_{k=1}^{\infty} p_k k s^{k-1}, \quad 0 \le s \le 1,$$

that is, prove differentiation and summation can be interchanged.

Proof. Let X be the corresponding random variable. Then $P(X = k) = p_k$.

$$\frac{d}{ds}P(s) = \frac{d}{ds}E(s^X)$$

$$= \lim_{t \to s} \frac{E(s^X) - E(t^X)}{s - t}$$

$$= \lim_{t \to s} \frac{E(s^X - t^X)}{s - t}$$

$$= \lim_{t \to s} E\left(\frac{s^X - t^X}{s - t}\right).$$

Then we choose a sequence of numbers $\{t_n\}_{n=1}^{\infty}$ such that $t_n \downarrow s$ and $t_n \neq s$. So

$$\frac{d}{ds}P(s) = \lim_{t_n \to s} E\left(\frac{s^X - t_n^X}{s - t_n}\right)$$
$$= \lim_{t_n \to s} E(Y_n)$$

by letting $Y_n = \frac{s^X - t_n^X}{s - t_n}$. Then

$$\lim_{n \to \infty} Y_n = \lim_{n \to \infty} \frac{s^X - t_n^X}{s - t_n}$$

$$= \lim_{n \to \infty} \frac{-(X - 1)t_n^X}{-1}$$

$$= \lim_{n \to \infty} (X - 1)t_n^X$$

$$= (X - 1)s^X.$$

Let $f(y) = y^X$, then $f'(y) = Xy^{X-1}$. By Mean Value Theorem, there exists $\theta \in [t_n, s]$ such that $f(s) - f(t_n) = (s - t_n)(f'(\theta))$. So

$$|Y_n| = \left| \frac{s^X - t_n^X}{s - t_n} \right| = |f'(\theta)|$$
$$= |X\theta^{X-1}|$$
$$\le Xs^{X-1}.$$

Since Xs^{X-1} is an integrable random variable, by Dominated Convergence Theorem,

$$\lim_{t_n \to s} E(Y_n) = E(\lim_{n \to \infty} Y_n) = E\left((X - 1)s^X\right)$$

Since $\lim_{t_n \to s} E(Y_n) = \frac{d}{ds} P(s)$ and $E((X-1)s^X) = \sum_{k=1}^{\infty} p_k k s^{k-1}$, we have

$$\frac{d}{ds}P(s) = \sum_{k=1}^{\infty} p_k k s^{k-1}.$$