

Solution to DIY Problems in Homework 3.

1. **★★** Suppose X_1, X_2, \dots are independent nonnegative r.v.s.. Then, $\sum_n X_n < \infty$ a.s. if and only if $\sum_n P(X_n > c) < \infty$ and $\sum_n E(X_n \wedge c) > \infty$ for all or any $c > 0$.

Solution. As $\sum_n X_n < \infty$ is equivalent to $\sum_n X_n/c < \infty$ for all or any $c > 0$. We only need to consider $c = 1$ and cite Kolmogorov's three series theorem. For the sufficiency, the convergence of the first two series is trivial as the third also holds since $\text{var}(X_n \wedge 1) \leq E((X_n \wedge 1)^2) \leq E(X_n \wedge 1)$. For the necessary, the convergence of the first series is $\sum_n P(X_n > 1) < \infty$ and the $\sum_n E(X_n \wedge 1) = \sum_n E(X_n 1_{\{X_n \leq 1\}}) + \sum_n P(X_n > 1) < \infty$ by the convergence of the first and second series. \square

2. **★★** Suppose X_1, X_2, \dots are independent nonnegative r.v.s. Prove that $\sum_n X_n < \infty$ a.s. if and only if $\sum_n E[X_n/(1 + X_n)] < \infty$.

Solution. Notice that $E[X_n/(1 + X_n)1_{\{X_n > 1\}}] \leq P(X_n > 1) \leq 2E[X_n/(1 + X_n)1_{\{X_n > 1\}}]$ and $E(X_n/(1 + X_n)1_{\{X_n \leq 1\}}) > E(X_n 1_{X_n \leq 1}) > E(X_n/(1 + X_n)1_{\{X_n \leq 1\}})$. The the desired claim follows from Kolmogorov's three series theorem and that the X_i are nonnegative. \square

3. Let ξ_1, ξ_2, \dots be iid Cauchy r.v.s. with common density $1/[\pi(1 + x^2)]$ and let $X_n = |\xi_n|$. Find $b_n \uparrow \infty$ such that $S_n/b_n \rightarrow 1$ in probability.

Solution. Notice that

$$E(|\xi|1_{\{|\xi| < a\}}) = \int_0^a \frac{2t}{\pi(1 + t^2)} dt = \frac{1}{\pi} \log(1 + a^2) \approx \frac{2}{\pi} \log a, \quad \text{for large } a.$$

and

$$P(|\xi| > a) = \int_a^\infty \frac{2}{\pi(1 + t^2)} dt \approx \frac{2}{\pi a}, \quad \text{for large } a.$$

We check the two conditions of the WLLN. Let $b_n = (2/\pi)n \log(n)$ and $a_n = n \log n$. Then $nP(|\xi| > a_n) = 1/\log n \rightarrow 0$. and

$$nE(|\xi|^2 1_{|\xi| \leq a_n})/b_n^2 \leq \frac{1}{\pi} \frac{na_n}{b_n^2} = \frac{n^2 \log n}{\pi(n \log n)^2} \rightarrow 0.$$

Then, $S_n/b_n \rightarrow 1$ in probability. (In fact, no $b_n \uparrow \infty$ exists such that $S_n/b_n \rightarrow 1$ a.s..) \square

4. Suppose X_1, \dots, X_n, \dots are i.i.d. with mean 1. Show that $\max_{1 \leq k \leq n} |X_k|/|S_n| \rightarrow 0$ a.s.

Solution. By the SLLN, $S_n/n \rightarrow 1$ a.e. It suffices to show $\max_{1 \leq k \leq n} |X_k|/n \rightarrow 0$ a.e.. First, since X_i has finite mean, we know $E|X| < \infty$ and therefore $\sum_{i=1}^\infty P(|X_i| > i/\epsilon) < \infty$. It follows from the Borel-Cantelli lemma that, with probability 1, $X_n/n \leq \epsilon$ for all large n . In other words, $X_n/n \rightarrow 0$ a.e.. Write

$$\begin{aligned} \max_{1 \leq k \leq n} |X_k|/n &\leq \max_{1 \leq k \leq M} |X_k|/n + \max_{M \leq k \leq n} |X_k|/n \\ &\leq \max_{1 \leq k \leq M} |X_k|/n + \max_{M \leq k \leq n} |X_k|/k \leq \max_{1 \leq k \leq M} |X_k|/n + \max_{M \leq k \leq \infty} |X_k|/k \\ &\rightarrow \max_{M \leq k \leq \infty} |X_k|/k \quad \text{a.e.,} \quad \text{by letting } n \rightarrow \infty \\ &\rightarrow 0, \quad \text{a.e.,} \quad \text{by letting } M \rightarrow \infty. \end{aligned}$$

\square

5. Suppose X_1, \dots, X_n, \dots are independent with mean 0 and variance 1. Suppose c_n are constants such that $c_n/n^p \downarrow 0$ for some $0 < p < 1/2$. Show that $\sum_{j=1}^n c_j X_j/n \rightarrow 0$ a.s.

Solution.

$$\sum_{n=1}^{\infty} \text{var}(c_n X_n)/n^2 = \sum_{n=1}^{\infty} c_n^2/n^2 \leq \sum_{n=1}^{\infty} \left(\frac{c_n}{n^p}\right)^2 \frac{1}{n^{2-2p}} < \infty.$$

By Kolmogorov's criterion of SLLN, $\sum_{j=1}^n c_j X_j/n \rightarrow 0$, a.e.. \square

6. Prove, for iid r.v.s X, X_1, X_2, \dots , that $(S_n - C_n)/n \rightarrow 0$ a.s. for some constants C_n if and only if $E(|X|) < \infty$. Hint: Symmetrization.

Solution. “ \Leftarrow ” If $E|X| < \infty$, choose $C_n = nE(X)$. Then, The SLLN ensures $(S_n - C_n)/n \rightarrow 0$ a.s..

“ \Rightarrow ” Let $X_i^*, i = 1, 2, \dots$ be iid independent of $X_n, n = 1, 2, \dots$. Then $(\sum_{i=1}^n X_i^* - C_n)/n \rightarrow 0$ a.s.. As a result, $\sum_{i=1}^n (X_i - X_i^*)/n \rightarrow 0$, a.s.. Kolmogorov's SLLN implies $E(|X_1 - X_1^*|) < \infty$. Therefore $E(|X|) < \infty$. \square

7. If X, X_1, X_2, \dots are iid with $E(|X|^p) = \infty$ for some $p > 0$, then $\limsup |S_n|/n^{1/p} = \infty$ a.s.

Solution. $E(|X|^p) = \infty$ ensures $\limsup |X_n|/n^{1/p} \rightarrow \infty$ a.s., which also ensures $\limsup |S_n|/n^{1/p} = \infty$ a.s.. \square

8. Suppose X, X_1, X_2, \dots are iid with $E(|X|^p) < \infty$ for some $p \geq 1$. Then, $E(|S_n/n|^p) \rightarrow |E(X)|^p$. Hint: Fatou lemma.

Solution. It follows from SLLN that $S_n/n \rightarrow E(X)$ a.s. and $\sum_{i=1}^n |X_i|^p/n \rightarrow E(|X|^p)$ a.s.. Then,

$$\begin{aligned} \liminf [E|X|^p - E(|S_n/n|^p)] &= \liminf [E(\sum_{i=1}^n E|X_i|^p/n) - E(|S_n/n|^p)] \\ &\geq E[\liminf (\sum_{i=1}^n |X_i|^p/n)] - E(\limsup |S_n/n|^p) = E(|X|^p) - |E(X)|^p. \end{aligned}$$

As a result, $\limsup E(|S_n/n|^p) \leq |E(X)|^p$. On the other hand, $\liminf E(|S_n/n|^p) \geq E(|\liminf S_n/n|^p) = |E(X)|^p$. Then, $E(|S_n/n|^p) \rightarrow |E(X)|^p$. \square

9. Suppose X, X_1, X_2, \dots are iid with mean 1 and a_n are bounded real numbers. Then, $(1/n) \sum_{j=1}^n a_j \rightarrow 1$ if and only if $(1/n) \sum_{j=1}^n a_j X_j \rightarrow 1$ a.s.. Hint: Repeat the proof of Kolmogorov's SLLN.

Solution. Let $\xi = X - 1$ and $\xi_i = X_i - 1$. Then, ξ, ξ_1, ξ_2, \dots are iid with mean 0. We only need to show $(1/n) \sum_{j=1}^n a_j \xi_j \rightarrow 0$ a.s.. Let $\eta_i = \xi_i 1_{\{|\xi_i| \geq i\}}$.

(i) $P(\xi_i \neq \eta_i, i.o.) = 0$. Thus $(1/n) \sum_{i=1}^n a_i \xi_i \rightarrow 0$ a.s. is equivalent to Thus $(1/n) \sum_{i=1}^n a_i \eta_i \rightarrow 0$ a.s..

(ii) $E(\eta_n) \rightarrow 0$. Therefore $a_n \eta_n \rightarrow 0$ as a_n are bounded. Hence $(1/n) \sum_{i=1}^n a_i E(\eta_i) \rightarrow 0$.

(iii)

$$\sum_{i=1}^{\infty} \text{var}(a_i \eta_i/i) \leq \sup_k a_k^2 \sum_{i=1}^{\infty} \text{var}(\eta_i)/i^2 < \infty$$

Therefore $\sum_{i=1}^{\infty} a_i(\eta_i - E(\eta_i))/i < \infty$ a.s.. And the Kronecker lemma implies $(1/n) \sum_{i=1}^n a_i(\eta_i - E(\eta_i)) \rightarrow 0$ a.s..

Combine (i), (ii) and (iii), we have $(1/n) \sum_{j=1}^n a_j \xi_j \rightarrow 0$ a.s.. \square

10. Suppose X_1, \dots, X_n, \dots are iid with common density $f(x) = cg(x)$ with $g(x) = x^{-\alpha}$ if $x > 1$, 1 if $|x| \leq 1$ and $(-x)^{-\beta}$ if $x < -1$. $c > 0$ is some constant and $\alpha > 1$ and $\beta > 2$. Find the a.e. limit of S_n/n in terms of α and β . What condition on α and β would ensure $\sum_{n=1}^{\infty} X_n/n < \infty$ a.e.? (Hint: verify the three conditions in Kolmogorov's three series theorem; first make sure that $\alpha = \beta$.)

Solution.

$$\begin{aligned} E(X_n/n 1_{\{|X_n|/n \leq 1\}}) &= (c/n) \left[\int_1^n x x^{-\alpha} dx + \int_{-n}^{-1} x (-x)^{-\beta} dx + \int_{-1}^1 x dx \right] \\ &= (c/n) \left(\frac{n^{-\alpha+2} - 1}{-\alpha+2} - \frac{n^{-\beta+2} - 1}{-\beta+2} \right) \end{aligned}$$

(If $\alpha = 2$, $\frac{n^{-\alpha+2}-1}{-\alpha+2}$ is replaced by $\log n$.) $\sum_{n=1}^{\infty} E(X_n/n 1_{\{|X_n|/n \leq 1\}}) < \infty$ only if $\alpha = \beta$. Condition (ii) holds only when $\alpha = \beta$.

For condition (i), now that $\alpha = \beta$,

$$\sum_n P(|X_n|/n > 1) = \sum_n P(|X_1| > n) = 2c \sum_n \int_n^{\infty} x^{-\beta} dx = 2c \sum_n n^{-\beta+1}/(\beta-1) < \infty,$$

since $\beta > 2$.

For condition (iii),

$$\begin{aligned} \sum_n \text{var}(X_n/n 1_{\{|X_n| \leq n\}}) &= \sum_n (2c/n^2) \left[\int_1^n x^2 x^{-\beta} dx + \int_0^1 x^2 dx \right] \\ &= \sum_n (2c/n^2) \left(\frac{n^{-\beta+3} - 1}{-\beta+3} + 1/3 \right) < \infty, \end{aligned}$$

since $\beta > 2$. (Here $\frac{n^{-\beta+3}-1}{-\beta+3}$ is replaced by $\log(n)$ if $\beta = 3$.)

Overall, the condition on α and β is $\alpha = \beta$ to ensure $\sum_n X_n/n < \infty$, a.s.. \square

11. For any sequence of σ_n^2 satisfying $\sum_n \sigma_n^2/n^2 = \infty$, raise an example of a sequence of independent random variables X_1, X_2, \dots with mean 0 and $\text{var}(X_n) \leq \sigma_n^2$ such that X_n/n does not converge to 0 a.s., and, as a result, $\sum_n X_n/n \not\rightarrow \infty$ a.s..

Solution. Without loss of generality assume $0 < \sigma_n \leq n$. Let $a_n = \sigma_n/n$ and $X_n = \pm n$ with probability $a_n^2/2$ and 0 with probability $1 - a_n$. Then X_n are independent with mean 0 and variance σ_n^2 . And, for any $0 < \epsilon < 1$, $P(|X_n|/n > \epsilon) = a_n^2$ and therefore $\sum_n P(|X_n|/n > \epsilon) = \infty$. It follows from Borel-Cantelli lemma that $P(\{X_n/n \rightarrow 0\}) = 0$. \square

12. Suppose ξ_1, \dots, ξ_n is a sequence of r.v.s such that, for $p > 0$,

$$\sup_{i,j \geq n} E(|\xi_i - \xi_j|^p) \rightarrow 0, \quad \text{as } n \rightarrow \infty.$$

Then, there exists a r.v., denoted as ξ , such that $\xi_n \rightarrow \xi$ in L^p .

Solution. There exists $n_k \uparrow \infty$ such that $\sup_{i,j \geq n_k} E(|\xi_i - \xi_j|^p) \leq 2^{-2pk}$. Let $A_k = \{|\xi_{n_k} - \xi_{n_{k+1}}| \geq 2^{-k}\}$. By Markov inequality,

$$P(A_k) \leq \sup_{i,j \geq n_k} P(|\xi_i - \xi_j| \geq 2^{-k}) \leq \sup_{i,j \geq n_k} E(|\xi_i - \xi_j|^p)/2^{-pk} \leq 2^{-pk}.$$

Borel-Cantelli lemma implies $P(A_k \text{ i.o.}) = 0$. Hence, $P(\liminf_k A_k^c) = 1$. But on $\liminf_k A_k^c$, ξ_{n_k} is a Cauchy sequence. By the Cauchy criterion, there exists a limit, denoted as ξ , of ξ_{n_k} a.s.. Then,

$$\begin{aligned} \lim_n E(|\xi_n - \xi|^p) &= \lim_n E(|\xi_n - \lim_k \xi_{n_k}|^p) \\ &\leq \lim_n \liminf_k E(|\xi_n - \xi_{n_k}|^p) \quad \text{by Fatou's Lemma} \\ &\leq \lim_n \sup_{j \geq n} E(|\xi_j - \xi_n|^p) \rightarrow 0. \end{aligned}$$

□

13. ★★★ Suppose X_n are iid Poisson random variable with rate $\lambda > 0$. Prove that

$$\limsup_n \frac{X_n \log \log n}{\log n} = 1 \quad a.s..$$

Hint: first show $P(X = n) \leq P(X \geq n) \leq e^\lambda P(X = n)$.

Solution. Write

$$P(X \geq n) = \sum_{k=n}^{\infty} e^{-\lambda} \lambda^k / k! = \frac{e^{-\lambda} \lambda^n}{n!} \sum_{k=n}^{\infty} \frac{\lambda^{k-n}}{k!/n!} \leq \frac{e^{-\lambda} \lambda^n}{n!} \sum_{k=n}^{\infty} \frac{\lambda^{k-n}}{(k-n)!} = P(X = n) e^\lambda.$$

Let a_n be the integer part of $\delta \log n / \log \log n$ for a $\delta > 0$. Then,

$$P(X = a_n) = \frac{e^{-\lambda} \lambda^{a_n}}{a_n!} = e^{-\lambda} e^{a_n \log \lambda} e^{-\sum_{j=1}^{a_n} \log j} = e^{-a_n \log a_n (1+o(1))} = n^{-\delta+o(1)}.$$

Then, $\sum_{n=1}^{\infty} P(X_n \geq a_n) < \infty$ or $= \infty$ depending upon $\delta > 1$ or $\delta < 1$. This implies, by the Borel-Cantelli lemma,

$$\limsup \frac{X_n}{\log n / \log \log n} = 1 \quad a.s..$$

□