

Probability and Statistics

Solution sheet 6

Solution 6.1

(a) We have by Chebyshev inequality that

$$\begin{aligned}\mathbb{P}\left(\left|\frac{S_n}{n} - 1\right| \leq 0.5\right) &= 1 - \mathbb{P}\left(\left|\frac{S_n}{n} - 1\right| \geq 0.5\right) \\ &\geq 1 - \frac{\mathbb{E}\left(\left(\frac{S_n}{n} - 1\right)^2\right)}{0.25} \\ &= 1 - \frac{\text{Var}\left(\frac{S_n}{n}\right)}{0.25} \\ &= 1 - \frac{8}{n}.\end{aligned}$$

When $n = 40$, the bound is 0.8.

(b) Thanks to Central Limit Theorem, we have that:

$$\frac{\sqrt{n}}{\sqrt{\text{Var}(X_i)}} \left(\frac{S_n}{n} - \mathbb{E}(X_i)\right) \stackrel{(d)}{\rightarrow} N(0, 1).$$

Then, using this property we have

$$\begin{aligned}\mathbb{P}\left(\left|\frac{S_n}{n} - 1\right| \leq 0.5\right) &= \mathbb{P}\left(\frac{\sqrt{n}}{\sqrt{2}} \left|\frac{S_n}{n} - 1\right| \leq \frac{\sqrt{n}}{\sqrt{2}} 0.5\right) \\ &\approx \mathbb{P}(-\sqrt{5} \leq N(0, 1) \leq \sqrt{5}) \\ &= \phi(\sqrt{5}) - \phi(-\sqrt{5}) \approx 0.97.\end{aligned}$$

Solution 6.2

(a) Take $\epsilon > 0$, we know by continuity of f that there exists $\delta > 0$ so that for all $x \in [c - \delta, c + \delta]$, $|f(x) - f(c)| \leq \epsilon$. Then

$$\begin{aligned}|\mathbb{E}(f(Z_n) - f(c))| &\leq \mathbb{E}(|f(Z_n) - f(c)|) \\ &\leq \mathbb{E}(|f(Z_n) - f(c)| \mathbb{1}_{|Z_n - c| \leq \delta}) + \mathbb{E}(|f(Z_n) - f(c)| \mathbb{1}_{|Z_n - c| > \delta}) \\ &\leq \epsilon + 2\|f\|_\infty \mathbb{P}(|Z_n - c| > \delta) \xrightarrow{n \rightarrow \infty} \epsilon.\end{aligned}$$

(b) Take $\epsilon > 0$ and define

$$f_\epsilon(x) \mapsto \min\left\{\frac{1}{\epsilon}d(x, [c - \epsilon, c + \epsilon]), 1\right\}.$$

f_ϵ is clearly a continuous function. Note that $f_\epsilon(x) = 0$ if $x \in [c - \epsilon, c + \epsilon]$ and $f(x) = 1$ if $|x - c| \geq 2\epsilon$. Then, we have that:

$$\mathbb{P}(|X_n - c| \geq 2\epsilon) \leq \mathbb{E}[f_\epsilon(X_n)] \xrightarrow{n \rightarrow \infty} f_\epsilon(c) = 0.$$

Solution 6.3

(a) We know that for all $\epsilon \leq \frac{1}{2}$

$$\mathbb{P} \left[\bigcap_{0 < \epsilon \leq \frac{1}{2}} \{|X_n| \leq \epsilon\} \right] = \mathbb{P}[|X_n| = 0] = \mathbb{P}[A_n^c],$$

so $X_n \xrightarrow{\mathbb{P}} 0$ if and only if $\mathbb{P}(A_n^c) \rightarrow 1$.

(b) Given that X_n takes only values in $\{0, 1\}$, the sequence converges if and only if, from a point onward, it only takes the value 0. Therefore

$$\{\omega : \lim X_n(\omega) = 0\} = \bigcup_{k \in \mathbb{N}} \bigcap_{n \geq k} A_n^c = \liminf A_n^c.$$

(c) For $n \in \mathbb{N}$, define $r_n = \lfloor \log_2(n) \rfloor$ and $k_n = n - 2^{r_n}$. Take

$$A_n = \left[\frac{k_n}{2^{r_n}}, \frac{k_n + 1}{2^{r_n}} \right],$$

note that $\mathbb{P}(A_n) = 2^{-r_n} \rightarrow 0$, so $X_n \xrightarrow{\mathbb{P}} 0$. Moreover, there are 2^{r_n} integers n_i such that $r_n = \lfloor \log_2(n_i) \rfloor$, and we have

$$\mathbb{P} \left(\bigcup_{n:r_n=r} A_n \right) = 2^{r_n} \frac{1}{2^{r_n}} = 1,$$

so $\bigcup_{n:r_n=r} A_n = [0, 1]$. Then we know that for each $r \in \mathbb{N}$ and for all $x \in [0, 1]$ there exists $n \in \mathbb{N}$ so that $r_n = r$ and $x \in A_n$, so $X_n(x)$ is 1 infinitely many times. Thus, $\{\omega : X_n(\omega) \rightarrow 0\} = \emptyset$.

Solution 6.4

(a) With Chebyshev inequality

$$P \left[\left| \frac{S_n}{n} - \mu \right| \geq \epsilon \right] \leq \frac{1}{\epsilon^2} \text{Var} \left(\frac{S_n}{n} \right)$$

it is enough to prove that $\text{Var} \left(\frac{S_n}{n} \right) \rightarrow 0$ ($n \rightarrow \infty$).

Computing the variance we have:

$$\begin{aligned} \text{Var} \left(\frac{S_n}{n} \right) &= \text{Var} \left(\frac{1}{n} \sum_{i=1}^n X_i \right) \\ &= \frac{1}{n^2} \left(\sum_{i=1}^n \text{Var}(X_i) + 2 \sum_{i < j} \text{Cov}(X_i, X_j) \right) \\ &= \frac{1}{n^2} \left(n\sigma^2 + 2 \sum_{k=1}^{n-1} (n-k) R(k) \right) \\ &= \frac{1}{n} \left(\sigma^2 + 2 \sum_{k=1}^{n-1} \left(1 - \frac{k}{n} \right) R(k) \right) \end{aligned}$$

The result obtains if we can prove

$$\lim_{n \rightarrow \infty} \frac{2}{n} \sum_{k=1}^{n-1} \left(\frac{n-k}{n} \right) R(k) = 0.$$

This can be proved as Cesaro Lemma. Let $\epsilon > 0$. As $\lim_{k \rightarrow \infty} R(k) = 0$, there exists $N \in \mathbb{N}$ such that for $n \geq N$ it holds that $|R(n)| \leq \epsilon$. Then, for $n \geq N + 1$,

$$\frac{2}{n} \sum_{k=1}^{n-1} \left(\frac{n-k}{n}\right) R(k) = \frac{2}{n} \sum_{k=1}^N \left(\frac{n-k}{n}\right) R(k) + \frac{2}{n} \sum_{k=N+1}^{n-1} \left(\frac{n-k}{n}\right) R(k)$$

There exists $N' \in \mathbb{N}$ such that for $n \geq N'$, $\left| \frac{2}{n} \sum_{k=1}^N \left(\frac{n-k}{n}\right) R(k) \right| \leq \epsilon$. Then for $n \geq \max\{N, N'\}$, we have

$$\left| \frac{2}{n} \sum_{k=1}^{n-1} \left(\frac{n-k}{n}\right) R(k) \right| \leq \epsilon + 2\epsilon,$$

and we have proved the statement.

(b) We compute

$$\begin{aligned} \lim_{n \rightarrow \infty} n \operatorname{Var} \left(\frac{S_n}{n} \right) &= \lim_{n \rightarrow \infty} \left(\sigma^2 + 2 \sum_{k=1}^{n-1} \left(1 - \frac{k}{n} \right) R(k) \right) \\ &= \sigma^2 + 2 \sum_{k=1}^{\infty} R(k) - 2 \lim_{n \rightarrow \infty} \sum_{k=1}^{n-1} \frac{k}{n} R(k). \end{aligned}$$

Define:

$$a_n(k) := \begin{cases} \frac{k}{n} R(k) & (k < n) \\ 0 & (k \geq n) \end{cases}$$

it holds that $a_n(k) \rightarrow 0 (n \rightarrow \infty)$ for all k . Then we just have to use the dominated convergence theorem, to prove that this part converges to 0. Note that $|a_n(k)| \leq |R(k)|$ and $|R(k)|$ is absolutely convergent. Therefore

$$\lim_{n \rightarrow \infty} \sum_{k=1}^{n-1} \frac{k}{n} R(k) = \lim_{n \rightarrow \infty} \sum_{k=1}^{n-1} a_n(k) \leq \lim_{n \rightarrow \infty} \sum_{k=1}^{\infty} a_n(k) = \sum_{k=1}^{\infty} \lim_{n \rightarrow \infty} a_n(k) = 0$$

Then

$$\lim_{n \rightarrow \infty} n \operatorname{Var} \left(\frac{S_n}{n} \right) = \sigma^2 + 2 \sum_{k=1}^{\infty} R(k).$$

Solution 6.5

(a) Take $f : \mathbb{R} \rightarrow \mathbb{R}$ a continuous and bounded function

$$\begin{aligned} \left| \int f d((1 - \epsilon_n)\mu_n + \epsilon_n\nu_n) - \int f d\mu \right| &\leq \left| \int f d\mu_n - \int f d\mu \right| + \epsilon_n \left| \int f d\nu_n - \int f d\mu_n \right| \\ &\leq \left| \int f d\mu_n - \int f d\mu \right| + 2\epsilon_n \|f\|_{\infty} \rightarrow 0. \end{aligned}$$

(b) Take $\mu_n = \delta_0$, i.e. $\mu(A) = \mathbb{1}_{0 \in A}$ and $\nu_n = \delta_n$. It is clear that $\mu_n \rightarrow \delta_0$ (it is a constant sequence), so $(1 - \frac{1}{n})\mu_n + \frac{1}{n}\nu_n \rightarrow \delta_0$, but:

$$\int |x| d \left(\left(1 - \frac{1}{n} \right) \mu_n + \frac{1}{n} \nu_n \right) (x) = \frac{1}{n} n = 1 \neq 0 = \int |x| d\delta_0(x).$$

- (c) We prove first the two claims of the Hint. We know that $\min\{|\cdot|, M\}$ is a bounded continuous function, therefore

$$\int \min\{|x|, M\} d\mu_n(x) \rightarrow \int \min\{|x|, M\} d\mu(x),$$

and

$$\begin{aligned} & \int |x| d\mu_n(x) - \int \min\{|x|, M\} d\mu_n(x) \\ &= \int (|x| - M) \mathbb{1}_{|x| \geq M} d\mu_n(x) \\ &\leq \int |x| \mathbb{1}_{|x| \geq M} d\mu_n(x) \\ &\leq \sqrt{\int x^2 d\mu_n(x)} \sqrt{\int \mathbb{1}_{|x| \geq M} d\mu_n(x)} \\ &\leq \sqrt{K} \sqrt{\int \mathbb{1}_{|x|^2 \geq M^2} d\mu_n(x)} \\ &\leq \sqrt{K} \sqrt{K/M^2} \\ &= \frac{K}{M} \end{aligned}$$

where we used Cauchy-Schwarz and Chebychev inequalities successively. The above difference is clearly non-negative. By the monotone convergence theorem

$$\int \min\{|x|, M\} d\mu(x) \xrightarrow{M \rightarrow \infty} \int |x| d\mu(x)$$

To finish, take $\epsilon > 0$, and M so that $K/M \leq \epsilon$, and that

$$\left| \int \min\{|x|, M\} d\mu(x) - \int |x| d\mu(x) \right| \leq \epsilon.$$

Take n_0 such that for all $n \geq n_0$,

$$\left| \int \min\{|x|, M\} d\mu_n(x) - \int \min\{|x|, M\} d\mu(x) \right| \leq \epsilon.$$

Finally,

$$\begin{aligned} & \left| \int |x| d\mu_n(x) - \int |x| d\mu(x) \right| \\ &\leq \left| \int |x| d\mu_n(x) - \int \min\{|x|, M\} d\mu_n(x) \right| + \left| \int \min\{|x|, M\} d\mu_n(x) - \int \min\{|x|, M\} d\mu(x) \right| \\ &\quad + \left| \int \min\{|x|, M\} d\mu(x) - \int |x| d\mu(x) \right| \\ &\leq K/M + \epsilon + \epsilon = 3\epsilon. \end{aligned}$$

Since ϵ is arbitrary, the statement is proved.