

### Solution to exercise 3

Let  $(X_i)_{i \in \{1, 2, \dots, n\}}$  be  $n$  independent and equally distributed random variables in  $L^2(\Omega, \mathbb{R})$ , with expectation  $\mu$  and standard deviation  $\sigma$ . Call

$$X = \frac{1}{n} \sum_{i=1}^n X_i.$$

Note that  $X$  can be seen as the average of a sample of length  $n$  of drawings of a random variable with expectation  $\mu$  and standard deviation  $\sigma$ .

It holds

$$\mathbb{E}[X] = \mu, \quad \text{Var}(X) = \frac{1}{n^2} \text{Var}\left(\sum_{i=1}^n X_i\right) = \frac{1}{n^2} n \sigma^2 = \frac{\sigma^2}{n}. \quad (1)$$

We want to use first Chebyshev's inequality and then the Central limit theorem, together with (1), to determine the confidence interval at a given confidence level  $\ell \in (0, 1)$  for  $X$ , that is, we want to find  $a, b$  with  $a < \mu, b > \mu$ , such that

$$P(X \in [a, b]) \geq \ell. \quad (2)$$

We start with Chebyshev's inequality, that tells us that for a given  $\epsilon > 0$  it holds

$$P(|X - \mathbb{E}[X]| > \epsilon) \leq \frac{\text{Var}(X)}{\epsilon^2},$$

possibly with an approximation. Let  $\epsilon = \frac{\sqrt{\text{Var}(X)}}{\sqrt{1-\ell}}$ . Then we have

$$\begin{aligned} 1 - \ell &\geq P\left(|X - \mathbb{E}[X]| > \frac{\sqrt{\text{Var}(X)}}{\sqrt{1-\ell}}\right) \\ &= P\left(X > \mathbb{E}[X] + \frac{\sqrt{\text{Var}(X)}}{\sqrt{1-\ell}} \text{ or } X < \mathbb{E}[X] - \frac{\sqrt{\text{Var}(X)}}{\sqrt{1-\ell}}\right) \\ &= P\left(X > \mu + \frac{\sigma}{\sqrt{n(1-\ell)}} \text{ or } X < \mu - \frac{\sigma}{\sqrt{n(1-\ell)}}\right) \\ &= 1 - P\left(\mu - \frac{\sigma}{\sqrt{n(1-\ell)}} \leq X \leq \mu + \frac{\sigma}{\sqrt{n(1-\ell)}}\right). \end{aligned}$$

So we have

$$P\left(\mu - \frac{\sigma}{\sqrt{n(1-\ell)}} \leq X \leq \mu + \frac{\sigma}{\sqrt{n(1-\ell)}}\right) \geq \ell,$$

and we can write

$$a = \mu - \frac{\sigma}{\sqrt{n(1-\ell)}}, \quad b = \mu + \frac{\sigma}{\sqrt{n(1-\ell)}}$$

such that (2) holds true.

We now consider the Central limit theorem, which tells us that the distribution of  $X = \frac{1}{n} \sum_{i=1}^n X_i$  can be approximated by  $\mathcal{N}\left(\mu, \frac{\sigma^2}{n}\right)$  for large  $n$ . We want to find  $\delta > 0$  such that

$$\ell \approx P(\mu - \delta \leq X \leq \mu + \delta).$$

For large  $n$  it holds

$$\begin{aligned} P(|X - \mu| > \delta) &\approx P\left(\left|\left(\frac{\sigma}{\sqrt{n}}Z + \mu\right) - \mu\right| > \delta\right) = P\left(\left|\frac{\sigma}{\sqrt{n}}Z\right| > \delta\right) = 2 \cdot P\left(Z > \frac{\sqrt{n}}{\sigma}\delta\right) \\ &= 2 \cdot \left(1 - \Phi\left(\frac{\sqrt{n}}{\sigma}\delta\right)\right), \end{aligned}$$

where  $Z \sim \mathcal{N}(0, 1)$  and  $\Phi$  is the standard normal cumulative distribution function. Note that the second last inequality comes from the fact that the distribution of  $Z$  is symmetric. We have

$$\ell = P(\mu - \delta \leq X \leq \mu + \delta) = 1 - P(|X - \mu| > \delta) \approx -1 + 2\Phi\left(\frac{\sqrt{n}}{\sigma}\delta\right),$$

which implies

$$\Phi\left(\frac{\sqrt{n}}{\sigma}\delta\right) \approx \frac{\ell + 1}{2}.$$

Then we find

$$\delta \approx \frac{\sigma}{\sqrt{n}}\Phi^{-1}\left(\frac{\ell + 1}{2}\right).$$