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## Solution to exercise 3

Let  $(X_i)_{i\in\{1,2,\ldots,n\}}$  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

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

$$P\left(X \in [a,b]\right) \ge \ell. \tag{2}$$

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

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

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

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

So we have

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

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

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

For large n it holds

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

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

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

which implies

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

Then we find

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