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 μ and standard deviation σ. 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 μ and standard deviation σ.

It holds

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

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 $\delta > 0$ such that

$$P(X \in [\mu - \delta, \mu + \delta]) \geq \ell. \tag{2}$$

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

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

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

$$1 - \ell \geq P\left(|X - \mathbb{E}[X]| > \frac{\sqrt{Var(X)}}{\sqrt{1-\ell}}\right)$$

$$= P\left(X > \mathbb{E}[X] + \frac{\sqrt{Var(X)}}{\sqrt{1-\ell}} \text{ or } X < \mathbb{E}[X] - \frac{\sqrt{Var(X)}}{\sqrt{1-\ell}}\right)$$

$$= P\left(X > \mu + \frac{\sigma}{\sqrt{n(1-\ell)}} \text{ or } X < \mu - \frac{\sigma}{\sqrt{n(1-\ell)}}\right)$$

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

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

$$\delta = \frac{\sigma}{\sqrt{n(1-\ell)}},$$

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

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

For large n it holds

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

where $Z \sim \mathcal{N}(0,1)$ and Φ 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

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