

# MATH5835M Statistical Computing

## Exercise Sheet 5

<http://www1.maths.leeds.ac.uk/~voss/2018/MATH5835M/>

Jochen Voss, J.Voss@leeds.ac.uk

2018/19, semester 1

*This does not count towards your final mark, the questions are for self-study only. We will discuss the answers to these questions in the lecture on 13th December.*

**Exercise 17.** Let  $(X_j)_{j \in \mathbb{N}}$  be a sequence of i.i.d. random variables and  $f: \mathbb{R} \rightarrow \mathbb{R}$  a function. Assume  $\mathbb{E}(f(X_j)) = \mu$  and  $\text{Var}(f(X_j)) = \sigma^2$  for all  $j \in \mathbb{N}$ . The central limit theorem states that, for large  $N$ , the random variable

$$Y_N = \frac{1}{\sqrt{N}} \sum_{j=1}^N \frac{f(X_j) - \mu}{\sigma}$$

is approximately  $\mathcal{N}(0, 1)$ -distributed.

- a) Write the Monte-Carlo estimator

$$Z_N = \frac{1}{N} \sum_{j=1}^N f(X_j)$$

in terms of  $Y_N$ .

- b) Derive the approximate distribution of  $Z_N$  for large  $N$ .  
c) Show that

$$P\left(\mu \in \left[Z_N - \frac{1.96 \sigma}{\sqrt{N}}, Z_N + \frac{1.96 \sigma}{\sqrt{N}}\right]\right) \approx 0.95$$

for large  $N$ .

**Exercise 18.** Assume that we want to use MCMC to study the distribution with density

$$\pi(x) = \frac{1}{Z} \cdot \begin{cases} \frac{\sin(x)^2}{x^2}, & \text{if } x \in [-3\pi, 3\pi], \text{ and} \\ 0 & \text{else,} \end{cases}$$

where  $Z$  is the normalisation constant.

- a) Assuming that the proposals  $Y_j$  are constructed as  $Y_j = X_{j-1} + \varepsilon_j$  where  $\varepsilon_j \sim \mathcal{N}(0, \sigma^2)$  is independent of  $X_1, \dots, X_{j-1}$ , determine the corresponding transition densities  $p(x, y)$  and acceptance probabilities  $\alpha(x, y)$ .  
b) Let  $X \sim \pi$ . Using the Random Walk Metropolis algorithm from part (a), with  $\sigma = 6$ , determine the expectations  $\mathbb{E}(X^k)$  for  $k = 1, 2, 3, 4$  and the probability  $P(|X| \leq 2)$ .  
c) For  $\sigma = 1, 6, 36$ , use the output of the Random Walk Metropolis algorithm to estimate the lag  $k$  auto-correlations  $\rho_k = \text{Corr}(X_j, X_{j+k})$  for  $k = 1, 2, \dots, 100$ . Create plots of the auto-correlations  $\rho_k$  as a function of  $k$ .  
d) Use the results from part (c) to determine, for each value of  $\sigma$  considered, a sample size  $N$  such that the MCMC estimate for  $\mathbb{E}(X^4)$  has root-mean squared error smaller than 0.1.

**Exercise 19.** Let  $X_1, \dots, X_n$  be i.i.d. with  $\mathbb{E}(X_i) = \mu$  and  $\text{Var}(X_i) = \sigma^2$ . Let  $X^*$  be a bootstrap sample drawn from  $X = (X_1, \dots, X_n)$ . In lectures we have learned that

$$\text{Var}(X^*) = \mathbb{E}(\text{Var}_X^*(X^*)) + \text{Var}(\mathbb{E}_X^*(X^*)).$$

- a) Determine  $\mathbb{E}_X^*(X^*)$  and  $\text{Var}_X^*(X^*)$  in terms of  $X_1, \dots, X_n$  and  $n$ .
- b) Determine  $\mathbb{E}(\text{Var}_X^*(X^*))$  and  $\text{Var}(\mathbb{E}_X^*(X^*))$  in terms of  $\mu$ ,  $\sigma^2$ , and  $n$ .
- c) Explain the behaviour of  $\mathbb{E}(\text{Var}_X^*(X^*))$  and  $\text{Var}(\mathbb{E}_X^*(X^*))$  for  $n = 1$  and for  $n \rightarrow \infty$ .