

## Homework 2: Due Thursday February 14, 2019

1. Solve exercises 8.2, 9.2, 12.1 (without the absolute value, i.e. estimate  $P(\max_{1 \leq k \leq n} S_k > 2\sqrt{n})$ ), and 12.3 from the textbook.
2. Write a code to generate  $\{S_k : 1 \leq k \leq n\}$  from exercises 12.1 and 12.3 and use  $n = 10,000$  to check that your theoretical estimates are indeed good. Namely, for example for 12.1, generate say 10,000 samples of  $\{S_1, \dots, S_{10,000}\}$  and compute the fraction of samples for which  $\max_{1 \leq k \leq 10,000} S_k$  was larger than  $2\sqrt{10,000} = 200$ .
3. Write a code that does the following:
  - a) Given an integer  $N \geq 1$  (a parameter in your code), generate and plot a sample of an approximation to Wiener's Brownian motion:

$$W_N(t) = Z_0 t + \frac{\sqrt{2}}{\pi} \sum_{j=1}^N \frac{Z_j}{j} \sin(\pi j t), \text{ for } 0 \leq t \leq 1,$$

where  $\{Z_j : 0 \leq j \leq N\}$  are i.i.d. standard normals. (We know this converges (almost surely!) to Brownian motion as  $N \rightarrow \infty$ . To demonstrate the code, use  $N = 1000$ .)

- b) For a given  $T > 0$  (another parameter in the code), generate and plot a sample of an approximation of Brownian motion

$$B_N(t) = \sqrt{T} W_N(t/T), \text{ for } 0 \leq t \leq T.$$

(To demonstrate the code, use  $T = 2$  and  $T = 10$ .)

4. Write a code that does the following: For a given  $T > 0$  and a given integer  $n \geq 1$  (two parameters in the code), generate  $\{X_k : 1 \leq k \leq \lfloor nT \rfloor\}$  i.i.d. standard normal random variables. (Here,  $\lfloor x \rfloor$  is the round-down of  $x$ .) Then generate and plot a sample of an approximation of Brownian motion:

$$B_n(t) = \frac{1}{\sqrt{n}} \sum_{k=1}^{\lfloor nt \rfloor} X_k, \text{ for } 0 \leq t \leq T.$$

Use  $n = 10,000$  and  $T = 1, T = 2, T = 10$ , to demonstrate your code.

5. How should the formula for  $B_n$  be changed if we did not use standard normals? E.g. what does it become if we use instead a sequence of i.i.d. uniform(-1,1)? What about a sequence of i.i.d. random variables that each take the values  $\pm 1$  equally likely? What about if the random variables are not centered? E.g. if we use a sequence of i.i.d. uniform(0,1) random variables?
6. Check the arcsine law with a simulation. That is, generate say 10,000 samples of  $\{W_N(t) : 0 \leq t \leq 1\}$  (say with  $N = 1000$ ) and for each sample find the point  $\tau \in [0, 1]$  at which  $W_N$  reached its maximal value (for the first time, if there are ties). Then plot the cumulative histogram of the 10,000 samples of  $\tau$ . This should approximate the CDF of  $\tau$ . Now, on the same graph (with a different color) plot the actual CDF (coming from the arcsine law). Are they close?
7. Generate four data sets of 100 i.i.d. samples each from the following distributions: exponential(2), normal(6,2) (mean 6, variance 2), Cauchy, and gamma(5,1) (scaling is 1 and shape exponent is 5).
  - a) For all but Cauchy, plot the pdf from a parameter estimation. Use exponential( $\theta$ ), normal( $\mu, \sigma^2$ ), and Gamma( $\theta, 1$ ).
  - b) Plot a histogram for each of the data sets, varying the starting point and the smoothing parameter to study their effect.
  - c) Plot a kernel estimator for each of the data sets, using the kernels:  $0.5e^{-|x|}$  and  $e^{-x^2/2}/\sqrt{2\pi}$  and the two values 0.5 and 0.1 for the smoothing parameter. (16 plots total)