

MATH5835M Statistical Computing

Exercise Sheet 4 (answers)

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

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Answer 13.

a) Since $X \sim \mathcal{N}(0, 1)$, we have $\mathbb{E}(X^2) = \text{Var}(X^2) + \mathbb{E}(X)^2 = 1 + 0^1 = 1$, and thus

$$\mathbb{E}(1 - X^2/2) = 1 - \frac{1}{2}\mathbb{E}(X^2) = 1 - \frac{1}{2} = \frac{1}{2}.$$

b) Using Taylor expansion around $x = 0$ we find

$$\cos(x) \approx 1 + 0 \cdot x + \frac{1}{2}(-1)x^2 = 1 - \frac{x^2}{2}.$$

A control variates estimate can be obtained by using the analytical result from part a to get an exact value for the expectation of $1 - X^2/2$, and then to estimate the (smaller) difference between this and the expectation of $\cos(X)$ using Monte Carlo:

$$\begin{aligned}\mathbb{E}(\cos(X)) &= \mathbb{E}\left(1 - \frac{X^2}{2}\right) + \mathbb{E}\left(\cos(X) - 1 + \frac{X^2}{2}\right) \\ &= \frac{1}{2} + \mathbb{E}\left(\cos(X) - 1 + \frac{x^2}{2}\right) \\ &\approx \frac{1}{2} + \frac{1}{N} \sum_{j=1}^N \left(\cos(X_j) - 1 + \frac{X_j^2}{2}\right).\end{aligned}$$

Answer 14. The procedure described in the question generates $X \sim \mathcal{U}[a, b]$, and thus X has density

$$f(x) = \frac{1}{b-a} 1_{[a,b]}(x).$$

The value X is accepted (in step ii), if $f(X) < Y$, where $Y \sim \mathcal{U}[c, d]$, *i.e.* with probability $p(X)$, where

$$p(x) = P(Y > f(x)) = \frac{d - f(x)}{d - c}.$$

From lecture we know then, that accepted samples have density

$$\varphi(x) \propto f(x)p(x) = \frac{1}{b-a} 1_{[a,b]}(x) \frac{d - f(x)}{d - c} \propto (d - f(x)) 1_{[a,b]}(x).$$

We can find the normalising constant by integrating:

$$Z = \int_{-\infty}^{\infty} (d - f(x)) 1_{[a,b]}(x) dx = \int_a^b (d - f(x)) dx.$$

Thus, the numbers the algorithm outputs have density

$$\varphi(x) = \frac{1}{Z} (d - f(x)) 1_{[a,b]}(x) = \frac{d - f(x)}{\int_a^b (d - f(x)) dx} 1_{[a,b]}(x).$$

Answer 15. The initial distribution is the distribution of X_0 , and thus is the standard normal distribution $\mathcal{N}(0, 1)$, with density

$$\varphi(x) = \frac{1}{\sqrt{2\pi}} \exp(-x^2/2)$$

for all $x \in \mathbb{R}$. The transition density is the the density of X_{n+1} , given $X_n = x$. Since $X_{n+1} \sim \mathcal{N}(0, 1)$, irrespective of the value of X_n , we find the transition density as

$$p(x, y) = \varphi(y)$$

for all $x, y \in \mathbb{R}$.

Answer 16. This question anticipates results we later covered in the chapter about bootstrap estimates.

a) $P(Y = 1) = P(I \in \{2, 4\}) = 2/6 = 1/3.$

b) We have

$$\begin{aligned}\mathbb{E}(Y) &= \mathbb{E}(x_I) \\ &= 3P(I = 1) + 1P(I = 2) + 3P(I = 3) + 1P(I = 4) + 5P(I = 5) + 2P(I = 6) \\ &= (3 + 1 + 4 + 1 + 5 + 2) \frac{1}{6} \\ &= \frac{16}{6}.\end{aligned}$$

c) We have

$$\mathbb{E}(x_I^2) = (3^2 + 1^2 + 4^2 + 1^2 + 5^2 + 2^2) \frac{1}{6} = \frac{56}{6}$$

and thus

$$\text{Var}(Y) = \text{Var}(x_I^2) = \mathbb{E}(x_I^2) - \mathbb{E}(x_I)^2 = \frac{56}{6} - \left(\frac{16}{6}\right)^2 = \frac{20}{9}.$$