

Please note that these questions do **not** cover all the topics that may be on the test.

Question 1: Suppose we numerically evaluate the integral

$$\int_0^1 x^4 dx$$

using the midpoint rule. Using 100 points, the approximation we get is 0.1999833333625. Using 1000 points, the approximation we get is 0.19999833333363. The exact answer is of course $1/5$. Estimate what approximation we will get if we use the midpoint rule with 2000 points.

Question 2: Suppose we have i.i.d. data points a_1, \dots, a_n that are measurements of angles in radians, in the range of 0 to 2π . We decide to model this data with a form of the “von Mises” distribution that assigns probability density $K \exp(\cos(a_i - \theta))$ to data point a_i , where $K = 0.1257\dots$, and θ is an unknown model parameter. Suppose that our prior distribution for θ is uniform over the range 0 to 2π .

Given data a_1, \dots, a_n , we wish to compute the Bayes factor for this von Mises model versus the simple model (with no parameters) that says the a_i are uniformly distributed over the range 0 to 2π . The Bayes factor for model A versus model B is the ratio of the probability of the data under model A to the probability of the data under model B, integrating over the parameters (if any) of each model with respect to the prior.

Write an R function that will compute this Bayes factor using R’s `integrate` function.

Question 3: Suppose that we want to obtain points, (x, y) , that are uniformly distributed over the diamond shape with vertices at $(0, 1)$, $(-1, 0)$, $(0, -1)$, and $(1, 0)$. Write an R function to do this using Gibbs sampling. You should use $(0, 0)$ as the initial point to start the Markov chain, and then sample alternately for x given y and y given x for 1000 iterations. You should return the pairs of points as a list with elements `x` and `y`, each of which will be vectors 1000 long, containing all the points from the chain (except the initial point).

Question 4: Let the state variable, x , for a Markov chain consist of two components, x_1 and x_2 . The possible values for x_1 are 0 and 1. The possible values for x_2 are 0, 1, and 2. (There are therefore six possible values for the entire state: 00, 01, 02, 10, 11, and 12.) Define the distribution π by the probabilities

$$\pi(x) = \begin{cases} 1/4 & \text{if } x_2 = x_1 \\ 1/4 & \text{if } x_2 = x_1 + 1 \\ 0 & \text{otherwise} \end{cases}$$

- Sketch how the Gibbs sampling procedure would work for this distribution, giving in particular the details of what conditional distributions to sample from when, and what these conditional distributions are.
- Write down explicitly the transition probabilities for the Gibbs sampling Markov chain that you described above, in which first x_1 and then x_2 are updated. (Ie, write down the 6 by 6 matrix whose entries are $T(x'|x)$ for all x and x' .)
- Show explicitly from the definition of invariance that the Markov chain you described above leaves π invariant.