

(Strictly speaking we need to justify exchanging the limit and expectation. Although we won't do this here, it can be done provided  $f$  is bounded and continuous.) Thus if  $U_1, \dots, U_n$  are an iid sample of  $U(a, b)$  random variables, then our estimate of  $I$  is

$$\hat{I} = \frac{1}{n} \sum_{i=1}^n f(U_i)(b-a).$$

The following function performs Monte-Carlo integration of the function `ftn` over the interval  $[a, b]$ .

```
mc.integral <- function(ftn, a, b, n) {
  # Monte-Carlo integral of ftn over [a, b] using a sample of size n
  u <- runif(n, a, b)
  x <- sapply(u, ftn)
  return(mean(x)*(b-a))
}
```

### 19.2.2 Accuracy in higher dimensions

The big-O notation is used to describe how fast a function grows. We say  $f(x)$  is  $O(x^{-\alpha})$  if  $\limsup_{x \rightarrow \infty} f(x)/x^{-\alpha} = \limsup_{x \rightarrow \infty} f(x)x^\alpha < \infty$ .

Let  $d$  be the dimension of our integral and  $n$  the number of function calls used, then the accuracy of the different numerical integration techniques we have seen is as follows:

Method	Error
Trapezoid	$O(n^{-2/d})$
Simpson's rule	$O(n^{-4/d})$
Hit-and-miss Monte-Carlo	$O(n^{-1/2})$
Improved Monte-Carlo	$O(n^{-1/2})$

We see that the size of the error for the Monte-Carlo methods does not depend on  $d$  and that, asymptotically, they are preferable when  $d > 8$ .

## 19.3 Exercises

1. Suppose that  $X$  and  $Y$  are iid  $U(0, 1)$  random variables.

(a). What is  $\mathbb{P}((X, Y) \in [a, b] \times [c, d])$  for  $0 \leq a \leq b \leq 1$  and  $0 \leq c \leq d \leq 1$ ?

Based on your previous answer, what do you think you should get for  $\mathbb{P}((X, Y) \in A)$ , where  $A$  is an arbitrary subset of  $[0, 1] \times [0, 1]$ ?

(b). Let  $A = \{(x, y) \in [0, 1] \times [0, 1] : x^2 + y^2 \leq 1\}$ . What is the area of  $A$ ?

(c). Define the rv  $Z$  by

$$Z = \begin{cases} 1 & \text{if } X^2 + Y^2 \leq 1, \\ 0 & \text{otherwise.} \end{cases}$$

What is  $\mathbb{E}Z$ ?

(d). By simulating  $Z$ , write a program to estimate  $\pi$ .

2. Which is more accurate, the hit-and-miss method or the improved Monte-Carlo method? Suppose that  $f : [0, 1] \rightarrow [0, 1]$  and we wish to estimate  $I = \int_0^1 f(x) dx$ .

Using the hit-and-miss method, we obtain the estimate

$$\hat{I}_{HM} = \frac{1}{n} \sum_{i=1}^n X_i,$$

where  $X_1, \dots, X_n$  are an iid sample and  $X_i \sim \text{binom}(1, I)$  (make sure you understand why this is the case).

Using the improved Monte-Carlo method, we obtain the estimate

$$\hat{I}_{MC} = \frac{1}{n} \sum_{i=1}^n f(U_i),$$

where  $U_1, \dots, U_n$  are an iid sample of  $U(0, 1)$  random variables.

The accuracy of the hit-and-miss method can be measured by the standard deviation of  $\hat{I}_{HM}$ , which is just  $1/\sqrt{n}$  times the standard deviation of  $X_1$ . Similarly the accuracy of the basic Monte-Carlo method can be measured by the standard deviation of  $\hat{I}_{MC}$ , which is just  $1/\sqrt{n}$  times the standard deviation of  $f(U_1)$ .

Show that

$$\text{Var } X_1 = \int_0^1 f(x) dx - \left( \int_0^1 f(x) dx \right)^2,$$

and that

$$\text{Var } f(U_1) = \int_0^1 f^2(x) dx - \left( \int_0^1 f(x) dx \right)^2.$$

Explain why (in this case at least) the improved Monte-Carlo method is more accurate than the hit-and-miss method.

3. The previous exercise gave a theoretical comparison of the hit-and-miss and improved Monte-Carlo method. Can you verify this experimentally?

Repeat the example of Section 19.1 using the improved Monte-Carlo method. How many function calls are required to get 2 decimal places accuracy?

4. The trapezoidal rule for approximating the integral  $I = \int_0^1 f(x) dx$  can be broken into two steps

Step 1:  $I = \sum_{i=0}^{n-1} (\text{Area under the curve from } i/n \text{ to } (i+1)/n);$