

12.1

Exercise 12.1 Show that Monte Carlo integration converges almost surely to the true integral of a bounded function.

Solving it with study of Monte-Carlo method.

We want to compute the integral: $\int_{\Omega} g(x) dx$

where:

- $\Omega \subset \mathbb{R}^d$ bounded
- g bounded on Ω

We will sample: $X_1, X_2, \dots, X_n \stackrel{i.i.d.}{\sim} \text{Uniform}(\Omega)$

and if μ is the measure of the measure space $(\Omega, \mathcal{F}, \mu)$, we will also define $V_{\Omega} = \mu(\Omega)$.

Now, as $X_i \sim \text{Unif}(\Omega)$, the density of every X_i is $f_X(x) = \frac{1}{V_{\Omega}} \mathbb{I}_{\Omega}(x)$

We will now define $Y_i = g(X_i)$ and Y_i are also i.i.d.

If we compute the expectation:

$$\mathbb{E}[Y_i] = \mathbb{E}[g(X_i)] = \int_{\Omega} g(x) f_{X_i}(x) dx = \frac{1}{V_{\Omega}} \int_{\Omega} g(x) dx$$

We can rewrite now the integral as: $\int_{\Omega} g(x) dx = V_{\Omega} \mathbb{E}[Y_i] = V_{\Omega} \mathbb{E}[g(X_i)]$

We can now use the SLLN in Y_i because Y_i i.i.d and the mean is finite because g is bounded:

$$\frac{1}{n} \sum_{i=1}^n Y_i \xrightarrow{\text{a.s.}} \mathbb{E}[g(X_i)]$$

And this is the monte-carlo estimator:

$$\frac{1}{n} \sum_{i=1}^n Y_i \xrightarrow{\text{a.s.}} \mathbb{E}[g(X_i)] \Leftrightarrow \frac{1}{n} \sum_{i=1}^n g(X_i) \xrightarrow{\text{a.s.}} \frac{1}{V_{\Omega}} \int_{\Omega} g(x) dx \Leftrightarrow$$

$$\Leftrightarrow V_{\Omega} \frac{1}{n} \sum_{i=1}^n g(X_i) \xrightarrow{\text{a.s.}} \int_{\Omega} g(x) dx$$