

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  $\Omega$

We will sample :  $x_1, x_2, \dots, x_n \stackrel{i.i.d.}{\sim} \text{Uniform}(\Omega)$

and if  $\mu$  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  $f_{X_i}(x) = \frac{1}{V_\Omega} \mathbb{I}_\Omega(x)$

We will now define  $y_i = g(x_i)$  and  $y_i$  are also iid.

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 :  $\int_{\Omega} g(x) dx = V_\Omega \mathbb{E}[y_i] = V_\Omega \mathbb{E}[g(x)]$

We can now use the SLLN in  $y_i$  because  $y_i$  iid and the mean is finite because  $g$  is bounded :

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

And this is the monte-carlo estimator :

$$\begin{aligned} \frac{1}{n} \sum_{i=1}^n y_i \xrightarrow{\text{a.s.}} \mathbb{E}[g(x)] &\Leftrightarrow \frac{1}{n} \sum_{i=1}^n g(x_i) \xrightarrow{\text{a.s.}} \frac{1}{V_\Omega} \int_{\Omega} g(x) dx \Leftrightarrow \\ &\Leftrightarrow V_\Omega \frac{1}{n} \sum_{i=1}^n g(x_i) \xrightarrow{\text{a.s.}} \int_{\Omega} g(x) dx \end{aligned}$$