

Exercices (3)

Let $(Z_n)_{n \geq 1}$ be an i.i.d. sequence of $\mathcal{N}(0, 1)$ -distributed random variables defined on a probability space $(\Omega, \mathcal{A}, \mathbb{P})$.

- (1) Compute for every real numbers $a, b > 0$, the quantity $\mathbb{E}((1 + bZ_1)e^{aZ_1})$.

$$\begin{aligned}\mathbb{E}((1 + bZ_1)e^{aZ_1}) &= \int_{\mathbb{R}} (1 + bz) e^{az} \frac{e^{-z^2/2}}{\sqrt{2\pi}} dz \\ &= \int_{\mathbb{R}} (1 + ba + b(z - a)) \frac{e^{-\frac{1}{2}(z-a)^2} e^{a^2/2}}{\sqrt{2\pi}} dz \\ \text{(integral of odd integrable function is zero)} &= (1 + ba)e^{a^2/2} + 0\end{aligned}$$

- (2) Compute, for every integer $n \geq 1$,

$$\left\| e^{a(Z_1 + \dots + Z_n) - na^2/2} - \prod_{k=1}^n (1 + aZ_k) \right\|_2$$

(remember : for all real r.v. X , $\|X\|_2 = \sqrt{\mathbb{E}(X^2)}$).

$$\begin{aligned}\left\| e^{a(Z_1 + \dots + Z_n) - na^2/2} - \prod_{k=1}^n (1 + aZ_k) \right\|_2^2 &= \mathbb{E} \left(\left(e^{a(Z_1 + \dots + Z_n) - na^2/2} - \prod_{k=1}^n (1 + aZ_k) \right)^2 \right) \\ &= \mathbb{E} \left(e^{2a(Z_1 + \dots + Z_n) - na^2} \right) + \mathbb{E} \left(\prod_{k=1}^n (1 + aZ_k)^2 \right) - 2\mathbb{E} \left(e^{a(Z_1 + \dots + Z_n) - na^2/2} \prod_{k=1}^n (1 + aZ_k) \right) \\ &\quad \text{(} Z_1, \dots, Z_n \text{ independant)} \\ &= \left(\mathbb{E}(e^{2aZ_1 - a^2}) \right)^n + \left(\mathbb{E}((1 + aZ_1)^2) \right)^n - 2 \left(\mathbb{E}(e^{aZ_1 - a^2/2} (1 + aZ_1)) \right)^n \\ &\quad \text{(we use Question 1)} \\ &= \left(e^{\frac{4a^2}{2} - a^2} \right)^n + (1 + a^2)^n - 2((1 + a^2))^n \\ &= e^{na^2} - (1 + a^2)^n\end{aligned}$$

so

$$\left\| e^{a(Z_1 + \dots + Z_n) - na^2/2} - \prod_{k=1}^n (1 + aZ_k) \right\|_2 = \sqrt{e^{na^2} - (1 + a^2)^n}.$$

- (3) Let $\sigma > 0$. Show the existence of real constants c_1, c_2 such that

$$\left\| e^{\sigma(Z_1 + \dots + Z_n)/\sqrt{n} - \sigma^2/2} - \prod_{k=1}^n \left(1 + \frac{\sigma Z_k}{\sqrt{n}} \right) \right\|_2 = \frac{\sigma e^{\sigma^2/2}}{\sqrt{2n}} \left(1 - \frac{\sigma^2(c_1\sigma^2 + c_2)}{n} + O(1/n^2) \right).$$

Use the previous question:

$$\begin{aligned}
\left\| e^{\sigma(Z_1 + \dots + Z_n)/\sqrt{n} - \sigma^2/2} - \prod_{k=1}^n \left(1 + \frac{\sigma Z_k}{\sqrt{n}} \right) \right\|_2 &= \left(e^{\sigma^2} - \left(1 + \frac{\sigma^2}{n} \right)^n \right)^{1/2} \\
&= \left(e^{\sigma^2} - \exp \left(n \ln \left(1 + \frac{\sigma^2}{n} \right) \right) \right)^{1/2} \\
&= \left(e^{\sigma^2} - \exp \left(n \left(\frac{\sigma^2}{n} - \frac{1}{2} \frac{\sigma^4}{n^2} + \frac{1}{3} \frac{\sigma^6}{n^3} + O(1/n^4) \right) \right) \right)^{1/2} \\
&= e^{\sigma^2/2} \left(1 - \exp \left(-\frac{1}{2} \frac{\sigma^2}{n} + \frac{1}{3} \frac{\sigma^4}{n^2} + O(1/n^3) \right) \right)^{1/2} \\
&= e^{\sigma^2/2} \left(\frac{1}{2} \frac{\sigma^2}{n} - \frac{1}{3} \frac{\sigma^4}{n^2} - \frac{1}{2!} \frac{\sigma^4}{4n^2} + O(1/n^3) \right)^{1/2} \\
&= \frac{\sigma e^{\sigma^2/2}}{\sqrt{2n}} \left(1 - \frac{2\sigma^2}{3n} - \frac{\sigma^4}{4n} + O(1/n^2) \right)^{1/2} \\
&= \frac{\sigma e^{\sigma^2/2}}{\sqrt{2n}} \left(1 - \frac{1}{2} \left(\frac{2\sigma^2}{3n} + \frac{\sigma^4}{4n} \right) + O(1/n^2) \right)
\end{aligned}$$

(4) We are interested in the equation

$$(0.1) \quad dX_t = \sigma X_t dW_t$$

(where $(W_t)_{t \geq 0}$ is a standard Brownian motion). Show that the (strong) solution of this equation is

$$X_t = X_0 \exp(\sigma W_t - \sigma^2 t/2).$$

The equation is of the form $dX_t = b(X_t) dW_t$ with a b which is Lipschitz in x so there is a unique strong solution. Apply Itô's formula:

$$\begin{aligned}
e^{\sigma W_t - \sigma^2 t/2} &= e^{\sigma W_0} + \int_0^t \sigma e^{\sigma W_s - \sigma^2 s/2} dW_s + \int_0^t -\frac{\sigma^2}{2} e^{\sigma W_s - \sigma^2 s/2} ds + \frac{1}{2} \int_0^t \sigma^2 e^{\sigma W_s - \sigma^2 s/2} ds \\
&= e^{\sigma W_0} + \int_0^t \sigma e^{\sigma W_s - \sigma^2 s/2} dW_s.
\end{aligned}$$

So we have our solution.

(5) We suppose that the increments of (W_t) on $[0, 1]$ are such that $W_{\frac{k}{n}} - W_{\frac{k-1}{n}} = Z_k/\sqrt{n}$, for all $k \in \{1, 2, \dots, n\}$. Show that the Euler scheme of order n associated to Equation (0.1) is such that

$$\bar{X}_{\frac{k}{n}} = X_0 \prod_{i=1}^k \left(1 + \frac{\sigma Z_i}{\sqrt{n}} \right).$$

We prove this by recurrence (on k).

- Thus true for $k = 0$.
- Suppose this is true in k .

$$\begin{aligned}
\bar{X}_{\frac{k+1}{n}} &= \bar{X}_{\frac{k}{n}} + \sigma \bar{X}_{\frac{k}{n}} (W_{\frac{k+1}{n}} - W_{\frac{k}{n}}) \\
&= \bar{X}_{\frac{k}{n}} \left(1 + \frac{\sigma Z_{k+1}}{\sqrt{n}} \right) \\
(\text{by recurrence}) &= X_0 \prod_{i=1}^{k+1} \left(1 + \frac{\sigma Z_i}{\sqrt{n}} \right).
\end{aligned}$$

- (6) Show that the strong rate of convergence obtained in the course (in a theorem) is optimal.

The Theorem of the course tells us (if X_0 is constant)

$$\mathbb{E}(\sup_{0 \leq k \leq n} |X_k - \bar{X}_k|^2) \leq \frac{C(1 + X_0^2)}{n}$$

for some constant C . Here we have (by Questions 4,5)

$$\begin{aligned} \mathbb{E}((X_1 - \bar{X}_1)^2) &= \mathbb{E} \left(\left(X_0 e^{\sigma(Z_1 + \dots + Z_n)/\sqrt{n} - \sigma^2/2} - X_0 \prod_{k=1}^n \left(1 + \frac{\sigma Z_k}{\sqrt{n}} \right) \right)^2 \right) \\ (\text{by Question 3}) &= X_0^2 \frac{\sigma^2 e^{\sigma^2}}{2n} \left(\left(1 - \frac{1}{2} \left(\frac{2\sigma^2}{3n} + \frac{\sigma^4}{4n} \right) + O(1/n^2) \right) \right)^2. \end{aligned}$$

So the bound in $1/n$ is optimal.

- (7) Write a code computing the Monte-Carlo approximation of the L^2 error for the above scheme, n fixed.
- (8) Write a code producing a curve of the square of this L^2 error versus n (you can use a log-log scale and a linear regression to see the power of n). You should get something similar to Figure 0.1.

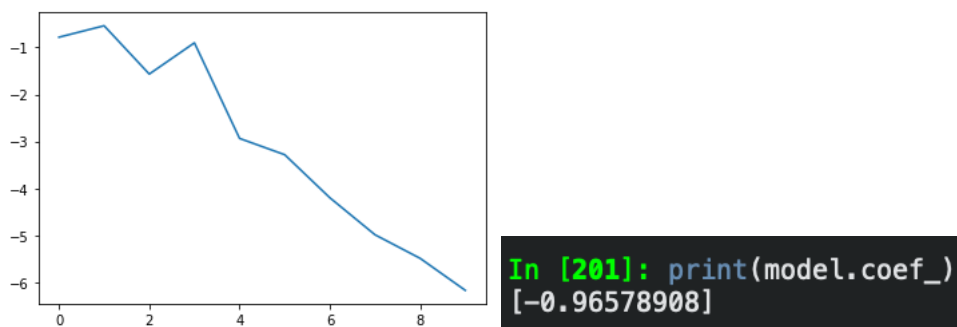


FIGURE 0.1. Rate of convergence in log-log scale.

PYTHON TIPS

Start by loading useful packages:

```
import numpy as np
import scipy.stats as sps
import matplotlib.pyplot as plt
import pandas as pd
from sklearn.linear_model import LinearRegression
```

Product/sum of terms in an array

```
np.prod(z1)
np.sum(z1)
```

Linear regression in dimension 1 (fit array y against array x) (get the slope)

```
model=LinearRegression()
model.fit(x.reshape((-1,1)),y)
print(model.coef_)
```