

Series 4

1. The price process S is a geometric Brownian motion with closed form solution

$$S_t = S_0 \exp \left((r - \sigma^2/2) t + \sigma W_t \right), \quad t \in [0, T].$$

In particular,

$$S_T = S_0 \exp \left((r - \sigma^2/2) T + \sigma W_T \right) = \exp \left(\ln(S_0) + (r - \sigma^2/2) T + \sigma W_T \right).$$

Furthermore, $W_T \sim \sqrt{T}Z$ with $Z \sim \mathcal{N}_{0,1}$, and consequently

$$S_T = \exp \left(\ln(S_0) + (r - \sigma^2/2) T + \sigma \sqrt{T}Z \right).$$

We denote

$$\alpha_T = \ln(S_0) + (r - \sigma^2/2) T, \quad \beta_T = \sigma \sqrt{T}.$$

Then, for every $K \in (0, +\infty)$ we have that

$$\mathbb{E}_P[f(S_T)] = \mathbb{E}_P \left[\max \left\{ e^{\alpha_T + \beta_T Z} - K, 0 \right\} \right] = \int_{\mathbb{R}} \max \left\{ e^{\alpha_T + \beta_T y} - K, 0 \right\} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}y^2} dy.$$

We observe that

$$e^{\alpha_T + \beta_T y} - K \geq 0 \iff \alpha_T + \beta_T y \geq \ln(K) \iff y \geq \frac{\ln(K) - \alpha_T}{\beta_T},$$

and consequently

$$\begin{aligned} \mathbb{E}_P[f(S_T)] &= \int_{\frac{\ln(K) - \alpha_T}{\beta_T}}^{\infty} (e^{\alpha_T + \beta_T y} - K) \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}y^2} dy \\ &= e^{\alpha_T} \int_{\frac{\ln(K) - \alpha_T}{\beta_T}}^{\infty} \frac{1}{\sqrt{2\pi}} e^{(\beta_T y - \frac{1}{2}y^2)} dy - K \int_{\frac{\ln(K) - \alpha_T}{\beta_T}}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}y^2} dy \\ &= e^{\alpha_T} \int_{\frac{\ln(K) - \alpha_T}{\beta_T}}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}(y^2 - 2\beta_T y + \beta_T^2 - \beta_T^2)} dy - K \Phi \left(\frac{\alpha_T - \ln(K)}{\beta_T} \right) \\ &= e^{\alpha_T + \frac{1}{2}\beta_T^2} \int_{\frac{\ln(K) - \alpha_T}{\beta_T}}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}(y - \beta_T)^2} dy - K \Phi \left(\frac{\alpha_T - \ln(K)}{\beta_T} \right). \end{aligned}$$

Finally, we make the change of variable $z = y - \beta_T$ and we obtain

$$\begin{aligned} \mathbb{E}_P[f(S_T)] &= e^{\alpha_T + \frac{1}{2}\beta_T^2} \int_{\frac{\ln(K) - \alpha_T}{\beta_T} - \beta_T}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}z^2} dz - K \Phi \left(\frac{\alpha_T - \ln(K)}{\beta_T} \right) \\ &= e^{\alpha_T + \frac{1}{2}\beta_T^2} \Phi \left(\frac{\alpha_T - \ln(K)}{\beta_T} + \beta_T \right) - K \Phi \left(\frac{\alpha_T - \ln(K)}{\beta_T} \right). \end{aligned}$$

Bitte wenden!

Since

$$e^{-rT} e^{\alpha_T + \frac{1}{2}\beta_T^2} = e^{\ln(S_0)} = S_0,$$

we arrive at the desired formula

$$\begin{aligned} e^{-rT} \mathbb{E}_P[f(S_T)] &= S_0 \Phi\left(\frac{\alpha_T - \ln(K)}{\beta_T} + \beta_T\right) - e^{-rT} K \Phi\left(\frac{\alpha_T - \ln(K)}{\beta_T}\right) \\ &= S_0 \Phi\left(\frac{\left(r - \frac{\sigma^2}{2}\right)T + \ln\left(\frac{S_0}{K}\right)}{\sigma\sqrt{T}} + \sigma\sqrt{T}\right) - e^{-rT} K \Phi\left(\frac{\left(r - \frac{\sigma^2}{2}\right)T + \ln\left(\frac{S_0}{K}\right)}{\sigma\sqrt{T}}\right) \\ &= S_0 \Phi\left(\frac{\left(r + \frac{\sigma^2}{2}\right)T + \ln\left(\frac{S_0}{K}\right)}{\sigma\sqrt{T}}\right) - e^{-rT} K \Phi\left(\frac{\left(r - \frac{\sigma^2}{2}\right)T + \ln\left(\frac{S_0}{K}\right)}{\sigma\sqrt{T}}\right). \end{aligned}$$

2. (i) We define the mappings

$$\mu: \mathbb{R} \rightarrow \mathbb{R}, \quad x \mapsto \log(1 + x^2)$$

and

$$\sigma: \mathbb{R} \rightarrow \mathbb{R}, \quad x \mapsto \mathbf{1}_{\{x>0\}}x.$$

In order to show existence and uniqueness of the solution process $X: [0, T] \times \Omega \rightarrow \mathbb{R}$, we prove that the conditions of Theorem 2.5.1 are fulfilled.

Since $\xi \in \mathcal{L}^p(P|_{\mathbb{R}_0}; |\cdot|)$ for some $p \geq 2$ by assumption, it remains to show that μ and σ are globally Lipschitz continuous. We start considering the mapping μ . By the mean value theorem, we obtain that for every $x, y \in \mathbb{R}$

$$|\mu(x) - \mu(y)| = |\log(1 + x^2) - \log(1 + y^2)| \leq \sup_{\xi \in \mathbb{R}} |\mu'(\xi)| |x - y| = \sup_{\xi \in \mathbb{R}} \left| \frac{2\xi}{1 + \xi^2} \right| |x - y|.$$

Moreover, we note that for every $\xi \in \mathbb{R}$

$$(1 + \xi)^2 \geq 0 \iff 1 + \xi^2 \geq |2\xi| \iff \frac{|2\xi|}{1 + \xi^2} \leq 1.$$

Hence, we have that

$$|\mu(x) - \mu(y)| \leq \sup_{\xi \in \mathbb{R}} \frac{|2\xi|}{1 + \xi^2} |x - y| \leq |x - y|.$$

In addition, it is immediate to check that σ satisfies for every $x, y \in \mathbb{R}$

$$|\sigma(x) - \sigma(y)| \leq |x\mathbf{1}_{\{x>0\}} - y\mathbf{1}_{\{y>0\}}| \leq |x - y|.$$

The above computations show that μ and σ are globally Lipschitz continuous. Therefore, the conditions of Theorem 2.5.1 are fulfilled and the SDE

$$dX_t = \log(1 + X_t^2)dt + \mathbf{1}_{\{X_t>0\}}X_t dW_t, \quad t \in [0, T], \quad X_0 = \xi,$$

admits a unique (up to indistinguishability) solution process $X: [0, T] \times \Omega \rightarrow \mathbb{R}$.

(ii) See file `EulerMaruyama.m`.

Siehe nächstes Blatt!

- (iii) See file `ErrorEM.m`. We know by item (i) that the coefficients of the SDE are globally Lipschitz continuous and $\xi = 1$. Hence, all assumptions of Theorem 3.3.10 are satisfied and consequently the Euler-Maruyama scheme converges in the strong L^2 -sense with order $\alpha = 1/2$. The expected convergence rate is realized in the experiment.

3. (i) See file `EulerMaruyamaGL.m`.

- (ii) See file `EulerMaruyamaGLEstimator.m`. The drift coefficient of X is not globally Lipschitz continuous and grows superlinearly. As a consequence, the Euler-Maruyama scheme does not converge in the strong L^2 -sense

- (iii) See files `TamedEulerMaruyamaGL.m` and `TamedEulerMaruyamaGLEstimator.m`.

Webpage: <https://moodle-app2.let.ethz.ch/course/view.php?id=17423>

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