

Oct 25,

p₁

[kp9.1] Euler approximation

Ⓢ[kp]: Kloeden & Platen 1999,

"Numerical Solutions of SDE"

SDE

$$dX_t = a(t, X_t) dt + b(t, X_t) dW_t$$

ex GBM is when

$$a(t, x) = r x$$

$$b(t, x) = \sigma x$$

Suppose $F: \mathbb{R}^d \mapsto \mathbb{R}$ is payoff function.

then we are interested in

$$\text{price} = \mathbb{E}[F(X_T)] \leftarrow \text{path indep.}$$

Suppose $F: \mathcal{D}[0, T] \mapsto \mathbb{R}$ is payoff, then

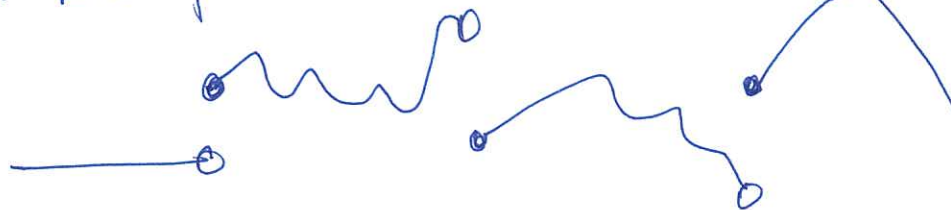
we are interested in

$$\text{price} = \mathbb{E}[F(x)] \leftarrow \text{path dep.}$$

where

$\mathcal{D}[0, T]$ is ^{set of} \wedge R-CLL (right cont. left limit exists)
~~processes~~ processes.

ex of RCLL process



ex of $F: D[0, T] \rightarrow \mathbb{R}$ is

$$F(\omega) = \max_{0 \leq t \leq T} (W_t - K)^+$$

Euler-Maruyama (EM)

Discrete process $\{(Y_n, T_n) : n \in \mathbb{N}\}$ is EM, if

- ① $0 = T_0 < T_1 < T_2 < \dots < T_N = T$
- ② $Y_0 = X_0$
- ③ $Y_{n+1} = Y_n + a(T_n, Y_n) \Delta_n + b(T_n, Y_n) \Delta_n W$

where

$$\Delta_n = T_{n+1} - T_n$$

$$\Delta_n W = W_{n+1} - W_n \sim \sqrt{\Delta_n} Z_n$$

iid std normal
↓

Goal $Y_n \approx X(T_n) \quad \forall n$

(in some sense?)

ex Demonstrate $(Y_N \approx X_{T_N} = X_T)$ by
following example.

p3

For GBM($r=0.0475$, $\sigma=0.2$, $X_0=100$),

Evaluate $\text{Put}(T=5, K=110)$ by Euler.

step1 Generate 1000 Euler trajectories.

$$\{Y^{(1)}, Y^{(2)}, \dots, Y^{(1000)}\}$$

step2 $\text{Put} \approx e^{-rT} \cdot \boxed{\hat{E}[(Y_T - K)^-]}$

$$\approx e^{-rT} \frac{Y_N^{(1)} + Y_N^{(2)} + \dots + Y_N^{(1000)}}{1000}$$

$$\approx e^{-rT} \frac{\sum_{i=1}^{1000} (Y_N^{(i)} - K)^-}{1000}$$

True $\text{Put} = e^{-rT} E[(Y_T - K)^-]$

= computed By BSM formula.

Prelim

P4

① max time step
$$\delta = \max_{1 \leq n \leq N} \Delta_n$$

② If $\delta = \frac{T}{N} = \Delta_n \quad \forall n$, then
EM is equidistant.

③ $E[\Delta_n W] = 0 \quad E[(\Delta_n W)^2] = \Delta_n$
i.e. $\Delta_n W \sim N(0, \Delta_n) = \sqrt{\Delta_n} N(0, 1)$.

ex ① plot EM of BM

② plot EM of GBM

{[KP 9.6]}. Strong convergence.

Given an approximation $\{Y^\delta: \delta \geq 0\}$ to
a process X , we say

Absolute Error

$$\varepsilon(\delta) = \sup_{0 \leq s \leq T} E |Y^\delta(s) - X(s)|^2$$

Def ① If $\lim_{\delta \rightarrow 0} \varepsilon(\delta) = 0$, then $Y^\delta \rightarrow X$ (strong)

② If $\varepsilon(\delta) \leq C \delta^{2r}$, then convergence order is r .

Consider 1-d/ homogeneous EM of P5
 equidistance process with

$$\begin{cases} \textcircled{1} & T_n = \frac{T}{N} (n-1) \\ \textcircled{2} & Y_0 = X_0 \\ \textcircled{3} & Y_{n+1} = Y_n + a(Y_n) \Delta_n + b(Y_n) \Delta_n W \end{cases}$$

for approximation of X of

$$\begin{cases} dX_t = a(X_t) dt + \sigma(X_t) dW_t \\ X_0. \end{cases} \quad (*)$$

[Assumption] $a(\cdot), b(\cdot)$ are Lipschitz cont.

Fact Under [A], $(*)$ has unique soln.

prop $\exists K$ constant. st.

$$\textcircled{1} \quad \mathbb{E} \left[\sup_{0 \leq s \leq t} |X_s|^2 \right] \leq K |X_0|^2 (1 + t e^{Kt})$$

$$\textcircled{2} \quad \mathbb{E} \left[\sup_{0 \leq s \leq t} |X_s - X_0|^2 \right] \leq K e^{Kt} |X_0|^2 \cdot t.$$

Ref [KP 1999], [FS06]

 \nwarrow Fleming
 \nearrow Soner

Interpolation

Let $N_t = \max \{ n \in \mathbb{N} : \tau_n \leq t \}$

Interpolation of $(Y_n^\delta : n=0, 1, \dots, N)$ is

~~$$\bar{Y}^\delta(s) = X_0 + \int_0^{\tau_{N_s}} a(\bar{Y}^\delta) ds + b$$~~

$$\bar{Y}^\delta(s) = Y_{N_s}^\delta$$

~~Then~~

Thm 9.6.2 With [A].

$$\bar{Y}^\delta \rightarrow X$$

convergence order
with $\delta = \frac{1}{2}$, which means

$$\sup_{0 \leq s \leq T} E |\bar{Y}_s^\delta - X_s|^2 \leq K\delta$$