学习报告

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考虑如下的随机微分方程

$$dX_{t} = a(t, X_{t}) dt + b(t, X_{t}) dW_{t}, t_{0} \le t \le T, X_{t_{0}} = X_{0}$$
(1)

给定时间划分 $t_0=\tau_0<\tau_1<\dots<\tau_n<\dots<\tau_N=T$,记 $\Delta_n=\tau_{n+1}-\tau_n,\delta=max\Delta_n$ 。

1 定义

定义 1.1. 称格式 Y_δ 在 T 处强收敛于 X 如果

$$\lim_{\delta \downarrow 0} E\left(\left|X_T - Y^{\delta}(T)\right|\right) = 0 \tag{2}$$

2 strong scheme

 $\mathcal{A}=\Lambda_k:=\left\{\alpha;\quad \alpha\in\mathcal{M}, l(\alpha)+n(\alpha)\leq 2k \text{ or } l(\alpha)=n(\alpha)=k+\frac{1}{2}\right\},$ 阶强格式为

$$X^{n+1} = \sum_{\alpha \in \Lambda_k} I_{\alpha} \left[f_{\alpha} \left(t_n, X^n \right) \right]_{t_n, t_{n+1}} = \sum_{\alpha \in \Lambda_k} f_{\alpha} \left(t_n, X^n \right) I_{\alpha, t_n, t_{n+1}}$$
 (3)

3 Euler 格式

Euler 格式是强 0.5 阶格式定义 A 为

$$\mathcal{A} = \{\phi, (0), (1)\} = \Lambda_{0.5} \tag{4}$$

由上面定义可知设 Brown 运动有 m 个分量, k 维的 Euler 格式为

$$Y_{n+1}^{k} = Y_n^{k} + a^k \Delta + \sum_{j=1}^{m} b^{k,j} \Delta W^j$$
 (5)

3.1 数值模拟

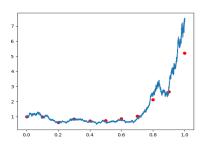
考虑如下的线性随机微分方程

$$dX(t) = \lambda X(t)dt + \mu X(t)dW(t), \quad X_{t_0} = X_0$$
 (6)

其真解为

$$X(t) = X_{t_0} \exp\left(\left(\lambda - \frac{1}{2}\mu^2\right)t + \mu W(t)\right)$$
 (7)

取区间为 [0,1] 将区间均匀剖分, $\delta=10^{-5}$,模拟出真解的情况,然后取 $\delta=0.1$ 应用欧拉格式,数值模拟结果如下 分别取 $\delta=10^{-i}, i=\{0,1,2,3,4\}$,各



重复实验 1000 次,取 T=1 处的误差,并求平均值来模拟 $E\left(\left|X_T-Y^\delta(T)\right|\right)$ 结果见下面表格

表 1: 误差						
1	2	3	4	5		
4.26748204	1.481642574	0.45718003	0.13571932	0.03757269		

4 Milstein 格式

Milstein 格式是强 1 阶格式,

$$\mathcal{A} = \{\phi, (0), (1), (1, 1)\} = \Lambda_1 \tag{8}$$

。代入 Ito - Taylor 展开得

$$X^{n+1} = \sum_{\alpha \in \Lambda_{1}} I_{\alpha} \left[f_{\alpha} \left(t_{n}, X^{n} \right) \right]_{t_{n}, t_{n+1}} = \sum_{\alpha \in \Lambda_{1}} f_{\alpha} \left(t_{n}, X^{n} \right) I_{\alpha, \text{tn}, t_{n+1}}$$

$$= f_{\phi} \left(t_{n}, X^{n} \right) I_{\phi, t_{n}, t_{n+1}} + f_{(0)} \left(t_{n}, X^{n} \right) I_{(0), t_{n}, t_{n+1}} + f_{(1)} \left(t_{n}, X^{n} \right) I_{(1), t_{n}, t_{n+1}}$$

$$+ f_{(1,1)} \left(t_{n}, X^{n} \right) I_{(1,1), t_{n}, t_{n+1}}$$

$$= X^{n} + a \left(t_{n}, X^{n} \right) \Delta t_{n} + b \left(t_{n}, X^{n} \right) \Delta W_{tn} + f_{(1,1)} \left(t_{n}, X^{n} \right) \int_{t_{n}}^{t_{n+1}} \int_{t_{n}}^{s} dW_{\tau}^{1} dW_{s}^{1} dW_{s}^{1$$

4.1 数值模拟

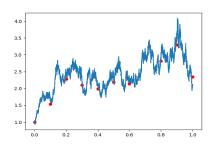
考虑如下的线性随机微分方程

$$dX(t) = \lambda X(t)dt + \mu X(t)dW(t), \quad X_{t_0} = X_0$$
 (10)

其真解为

$$X(t) = X_{t_0} \exp\left(\left(\lambda - \frac{1}{2}\mu^2\right)t + \mu W(t)\right)$$
(11)

取区间为 [0,1] 将区间均匀剖分, $\delta=10^{-5}$,模拟出真解的情况,然后取 $\delta=0.1$ 应用 Milstein 格式,数值模拟结果如下 分别取 $\delta=10^{-i},i=1$



 $\{0,1,2,3,4\}$,各重复实验 1000 次,取 T=1 处的误差,并求平均值来模拟 $E\left(\left|X_T-Y^\delta(T)\right|\right)$ 结果见下面表格

5 强 1.5 阶格式

$$\mathcal{A} = \left\{ \begin{array}{ccc} \emptyset, & (0), & (1), & (1,1) \\ (0,0), & (1,0), & (0,1), & (1,1,1) \end{array} \right\} = \Lambda_{1,5}$$
 (12)

表 2: 误差变化

1	2	3	4	5
4.576793014	1.481642574	0.16355771	0.01529428	0.00159989

 $\alpha = (0,0)$ 对应

$$f_{(0,0)} = L^0 L^0 f = L^0 a$$

$$= \frac{\partial a}{\partial t} + a \frac{\partial a}{\partial x} + \frac{1}{2} b^2 \frac{\partial^2 a}{\partial x^2}$$

$$I_{(0,0),t_n,t_{n+1}} = \int_{t_n}^{t_{n+1}} \int_{t_n}^{t} ds dt = \frac{1}{2} \Delta t_n$$

$$(13)$$

 $\alpha = (1,0)$ 对应

$$f_{(1,0)} = L^{1}L^{0}f = L^{1}a = b\frac{\partial a}{\partial x}$$

$$I_{(1,0),t_{n},t_{n+1}} = \int_{t_{n}}^{t_{n+1}} \int_{t_{n}}^{t} dW_{s}dt = \Delta Z_{n}$$
(14)

 $\alpha = (0,1)$ 对应

$$f_{(0,1)} = L^0 L^1 f = L^0 b$$

$$= \frac{\partial b}{\partial t} + a \frac{\partial b}{\partial x} + \frac{1}{2} b^2 \frac{\partial^2 b}{\partial x^2}$$

$$I_{(0,1)}, t_n, t_{n+1} = \int_t^{t_{n+1}} \int_t^t ds dW_t = \overline{\Delta Z_n}$$
(15)

 $\alpha = (1,1,1)$ 对应

$$f_{(1,1,1)} = L^{1}L^{1}L^{1}f = L^{1}L^{1}b = L^{1}\left(b\frac{\partial b}{\partial x}\right)$$

$$= b\frac{\partial}{\partial x}\left(b\frac{\partial b}{\partial x}\right) = b\left(\frac{\partial b}{\partial x}\right)^{2} + b^{2}\frac{\partial^{2}b}{\partial x^{2}}$$

$$= \int_{t_{n}}^{t_{n+1}} \int_{t_{n}}^{t} \int_{t_{n}}^{s} dW_{r}dW_{s}dW_{t}$$

$$I_{(1,1,1)}t_{n}, t_{n+1} = \frac{1}{2}\left(\frac{1}{3}\left(\Delta W_{t_{n}}\right)^{2} - \Delta t_{n}\right)\Delta W_{t_{n}}$$

$$(16)$$