

# Lecture 13: Strong Convergence Theorem and Derivative Free Schemes

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## Abstract

Convergence theorem for general strong Taylor schemes. Derivative free schemes.

Recall estimate in Lecture 12,

$$R_{0,u} = E\left(\sup_{s \in [0,u]} |g(s)|^2 \middle| A_0\right) < \infty, \quad (0.1)$$

$$F_t^\alpha = E\left(\sup_{z \in [0,t]} \left|\sum_{n=0}^{n_z-1} I_\alpha[g(\cdot)]_{\tau_n, \tau_{n+1}} + I_\alpha[g(\cdot)]_{\tau_{n_z}, z}\right|^2 \middle| A_0\right). \quad (0.2)$$

Then w.p. 1 for  $t \in [0, T]$ :

$$F_t^\alpha \leq T \delta^{2(l(\alpha)-1)} \int_0^t R_{0,u} du, \quad \text{if } l(\alpha) = n(\alpha), \quad (0.3)$$

and

$$F_t^\alpha \leq 4^{l(\alpha)-n(\alpha)+2} \delta^{l(\alpha)+n(\alpha)-1} \int_0^t R_{0,u} du, \quad l(\alpha) \neq n(\alpha), \quad (0.4)$$

Taking,

$$p(\alpha) = \begin{cases} 2(l(\alpha) - 1), & \text{if } l(\alpha) = n(\alpha) \\ l(\alpha) + n(\alpha) - 1, & \text{if } l(\alpha) \neq n(\alpha), \end{cases} \quad (0.5)$$

we have,

$$F_t^\alpha \leq C(T, \alpha) \delta^{p(\alpha)} \int_0^t R_{0,u} du. \quad (0.6)$$

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# 1 General Strong Convergence Theorem

Let  $\alpha = (\alpha_1, \alpha_2, \dots) \neq v$ ,  $v$  the empty index,  $\delta$  the time step of discretization over  $[0, T]$ ,  $\tau_n$ 's the uniform discrete time steps,  $Y(t) = Y^\delta(t)$  the strong Ito-Taylor approximation of order  $\gamma \geq 0.5$  satisfying:

$$Y(t) = Y_{n_t} + \sum_{\alpha \in A_\gamma} I_\alpha[f_\alpha(\tau_{n_t}, Y_{n_t})]_{\tau_{n_t}, t} = Y_{n_t} + \sum_{\alpha \in A_\gamma} f_\alpha(\tau_{n_t}, Y_{n_t}) I_\alpha[1]_{\tau_{n_t}, t}. \quad (1.7)$$

where

$$Y_{n+1} = Y_n + \sum_{\alpha \in A_\gamma} I_\alpha[f_\alpha(\tau_n, Y_n)]_{\tau_n, \tau_{n+1}} = Y_n + \sum_{\alpha \in A_\gamma} f_\alpha(\tau_n, Y_n) I_\alpha[1]_{\tau_n, \tau_{n+1}}, \quad (1.8)$$

$$n_z := \max\{n \in N \mid \tau_n \leq z\}. \quad (1.9)$$

Recall:

$$A_\gamma = \{\alpha : l(\alpha) + n(\alpha) \leq 2\gamma, \text{ or } l(\alpha) = n(\alpha) = \gamma + \frac{1}{2}\}, \quad (1.10)$$

with remainder set  $B = B(A) = \{\alpha \notin A : -\alpha \in A\}$ .

Assume that for all  $\alpha$  (detailed assumption can be found on KL's book):

$$|f_\alpha(t, x) - f_\alpha(t, y)| \leq K_1 |x - y|, \quad (1.11)$$

$$|f_\alpha(t, x)| \leq K_2(1 + |x|), \quad (1.12)$$

then:

$$E\left(\sup_{t \in [0, T]} |X_t - Y^\delta(t)|^2 | A_0\right) \leq K_3(1 + |X_0|^2)\delta^{2\gamma} + K_4|X_0 - Y^\delta(0)|^2. \quad (1.13)$$

The constants  $K_i$ 's are independent of  $\delta$ .

## 1.1 Proof

The exact Ito-Taylor expansion is:

$$X_\tau = X(\rho) + \sum_{\alpha \in A_\gamma} I_\alpha[f_\alpha(\rho, X_\rho)]_{\rho, \tau} + \sum_{\alpha \in B(A_\gamma)} I_\alpha[f_\alpha(\cdot, X_\cdot)]_{\rho, \tau}, \quad (1.14)$$

for any  $\rho < \tau$ . It follows that:

$$\begin{aligned} X_t &= X_0 + \sum_{\alpha \in A_\gamma} \left( \sum_{n=0}^{n_t-1} I_\alpha[f_\alpha(\tau_n, X_{\tau_n})]_{\tau_n, \tau_{n+1}} + I_\alpha[f_\alpha(\tau_{n_t}, X_{\tau_{n_t}})]_{\tau_{n_t}, t} \right) \\ &\quad + \sum_{\alpha \in B(A_\gamma)} \left( \sum_{n=0}^{n_t-1} I_\alpha[f_\alpha(\cdot, X_\cdot)]_{\tau_n, \tau_{n+1}} + I_\alpha[f_\alpha(\cdot, X_\cdot)]_{\tau_{n_t}, t} \right). \end{aligned} \quad (1.15)$$

Take the difference of (1.15) and  $Y(t)$  generate by (1.7):

$$\begin{aligned} Z(t) &= E\left(\sup_{s \in [0,t]} |X_s - Y^\delta(s)|^2 |A_0\right) \\ &\leq C(|X_0 - Y^\delta(0)|^2 + \sum_{\alpha \in A_\gamma} R_t^\alpha + \sum_{\alpha \in B(A_\gamma)} U_t^\alpha). \end{aligned} \quad (1.16)$$

The two type of terms  $R^\alpha$  and  $U^\alpha$  are bounded below. First by (0.6), then Lipschitz condition on  $f_\alpha$ :

$$\begin{aligned} R_t^\alpha &= E\left(\sup_{s \in [0,t]} \left| \sum_{n=0}^{n_s-1} I_\alpha[f_\alpha(\tau_n, X_{\tau_n}) - f_\alpha(\tau_n, Y_n^\delta)]_{\tau_n, \tau_{n+1}} \right.\right. \\ &\quad \left.\left. + I_\alpha[f_\alpha(\tau_{n_s}, X_{\tau_{n_s}}) - f_\alpha(\tau_{n_s}, Y_{n_s}^\delta)]_{\tau_{n_s}, s} \right|^2 |A_0\right) \\ &\leq C \int_0^t E\left(\sup_{s \in [0,u]} |f_\alpha(\tau_{n_s}, X_{\tau_{n_s}}) - f_\alpha(\tau_{n_s}, Y_{n_s}^\delta)|^2 |A_0\right) du \\ &\leq CK_1^2 \int_0^t Z(u) du. \end{aligned} \quad (1.17)$$

Note that  $\alpha \in A_\gamma$  here, so  $l(\alpha)$  starts from 1.

Next for  $\alpha \in B(A_\gamma)$ :

$$\begin{aligned} U_t^\alpha &= E\left(\sup_{s \in [0,t]} \left| \sum_{n=0}^{n_s-1} I_\alpha[f_\alpha(\cdot, X_\cdot)]_{\tau_n, \tau_{n+1}} + I_\alpha[f_\alpha(\cdot, X_\cdot)]_{\tau_{n_s}, s} \right|^2 |A_0\right) \\ &\leq C_1(1 + |X_0|^2)\delta^{p(\alpha)}. \end{aligned} \quad (1.18)$$

For  $\alpha \in B(A_\gamma)$ , recall,  $A_\gamma = \{\alpha : l(\alpha) + n(\alpha) \leq 2\gamma, \text{ or } l(\alpha) = n(\alpha) = \gamma + \frac{1}{2}\}$

- if  $l = n$ :
  - if  $\gamma$  is not integer,  $l \geq \gamma + 1/2 + 1$ ,
  - if  $\gamma$  is integer,  $l \geq \gamma + 1$ ;
- if  $l \neq n$ ,  $l + n \geq 2\gamma + 1$ .

So  $p(\alpha) \geq 2\gamma$ .

$$U^\alpha \leq C_2(1 + |X_0|^2)\delta^{2\gamma}. \quad (1.19)$$

Finally,  $Z$  obeys inequality:

$$Z(t) \leq C_2|X_0 - Y^\delta(0)|^2 + C_3(1 + |X_0|^2)\delta^{2\gamma} + C_4 \int_0^t Z(u) du, \quad (1.20)$$

by Gronwall inequality:

$$Z(T) \leq C_5(1 + |X_0|^2)\delta^{2\gamma} + C_6|X_0 - Y^\delta(0)|^2. \quad (1.21)$$

## 2 Remarks for General Strong Convergence

Now if we further assume,

$$E(|X_0|^2) < \infty, \quad (2.22)$$

and

$$\sqrt{E(|X_0 - Y^\delta(0)|^2)} \leq K_5 \delta^\gamma. \quad (2.23)$$

Then,

$$\sqrt{E\left(\sup_{0 \leq t \leq T} |X_t - Y^\delta(t)|^2\right)} \leq K_6 \delta^\gamma. \quad (2.24)$$

**$L_1$  convergence** Recall the Lyapunov inequality: if  $X$  is not concentrated on a single point and if  $E(|X|^s)$  exists for some  $s > 0$ , then for all  $0 < r < s$  and  $a \in \Re$

$$(E(|X - a|^r))^{1/r} \leq (E(|X - a|^s))^{1/s} \quad (2.25)$$

Also we should realize,

$$\sup_{0 \leq t \leq T} |X_t - Y^\delta(t)|^2 = \left( \sup_{0 \leq t \leq T} |X_t - Y^\delta(t)| \right)^2 \quad (2.26)$$

so

$$E\left(\sup_{0 \leq t \leq T} |X_t - Y^\delta(t)|\right) \leq K_6 \delta^\gamma \quad (2.27)$$

**$L_p$  convergence** In fact (need some derivation), given,

$$E(|X_0|^p) < \infty, \quad (E(|X_0 - Y^\delta(0)|^p))^{1/p} \leq K_5^* \delta^\gamma \quad (2.28)$$

we have

$$\left( E\left(\sup_{0 \leq t \leq T} |X_t - Y^\delta(t)|^p\right) \right)^{1/p} \leq K_6^* \delta^\gamma \quad (2.29)$$

**Selection of  $p$**  Suppose in addition, on each time interval  $[\tau_n, \tau_{n+1}]$  for each  $n = 0, 1, \dots$  and all  $\alpha \in \mathcal{A}_\gamma$ ,

$$E(|I_\alpha - I_\alpha^p|^2) \leq K_7 \delta^{2\gamma+1}, \quad (2.30)$$

then (left as project V),

$$E(|X_T - Y^\delta(T)|) \leq K_8 \delta^\gamma. \quad (2.31)$$

Recall,

$$E(|I_\alpha - I_\alpha^p|^2) \leq C \frac{\delta^2}{p}, \quad (2.32)$$

to obtain a strong scheme of order  $\gamma = 1.0, 1.5$  or  $2.0$  we need to choose  $p$  so that

$$p \geq p(\delta) = \frac{C}{K_7} \delta^{1-2\gamma}. \quad (2.33)$$

### 3 Derivative Free Schemes

#### 3.1 Convergence

RK (Runge-Kutta) schemes are constructed by approximating derivatives of drift and diffusion functions in Ito-Taylor expansion. Denote such a scheme as:

$$Y_{n+1} = Y_n + \sum_{\alpha \in A_\gamma} g_{\alpha,n} I_\alpha + R_n, \quad (3.34)$$

$$A_\gamma = \{\alpha : l(\alpha) + n(\alpha) \leq 2\gamma, \text{ or } l(\alpha) = n(\alpha) = \gamma + \frac{1}{2}\}. \quad (3.35)$$

Still

$$p(\alpha) = \begin{cases} 2(l(\alpha) - 1), & \text{if } l(\alpha) = n(\alpha) \\ l(\alpha) + n(\alpha) - 1, & \text{if } l(\alpha) \neq n(\alpha), \end{cases} \quad (3.36)$$

Assume that:

$$E(\max_{0 \leq n \leq n_T} |g_{\alpha,n} - f_\alpha(Y_n)|^2) \leq K \delta^{2\gamma-p(\alpha)}, \quad (3.37)$$

and

$$E(\max_{1 \leq n \leq n_T} |\sum_{0 \leq k \leq n-1} R_k|^2) \leq K \delta^{2\gamma}. \quad (3.38)$$

A scheme satisfying the above is called *strong Ito scheme of order  $\gamma$* .

**Proof** Compare Ito scheme with Ito-Taylor approximation of the same order. Denote the latter by  $\bar{Y}_n$ :

$$\bar{Y}_{n+1} = \bar{Y}_n + \sum_{\alpha \in A_\gamma} f_\alpha(\bar{Y}_n) I_\alpha,$$

converging with order  $\gamma$  to SDE solution  $X_t$ .

$$\begin{aligned}
H_t &= E(\max_{n \in [0, n_t]} |\bar{Y}_n - Y_n|^2) \\
&= E(\max_{n \in [0, n_t]} \left| \sum_{k=0}^{n-1} \sum_{\alpha} f_{\alpha}(\bar{Y}_k) I_{\alpha} - \sum_{k=0}^{n-1} \left( \sum_{\alpha} g_{\alpha, k} I_{\alpha} + R_k \right) \right|^2) \\
&\leq K_1 \sum_{\alpha} [E(\max_{n \in [0, n_t]} \left| \sum_{k=0}^{n-1} (f_{\alpha}(Y_k) - f_{\alpha}(\bar{Y}_n)) I_{\alpha} \right|^2) + E(\max_{n \in [0, n_t]} \left| \sum_{k=0}^{n-1} (f_{\alpha}(Y_k) - g_{\alpha, k}) I_{\alpha} \right|^2)] \\
&\quad + K_1 E(\max_{n \in [0, n_t]} \left| \sum_{k=0}^{n-1} R_k \right|^2),
\end{aligned} \tag{3.39}$$

for  $t \in [0, T]$ .

By (0.6) and (3.38),

$$\begin{aligned}
H_t &\leq K_2 \sum_{\alpha} \left[ \int_0^t E(\max_{n \in [0, n_u]} |f_{\alpha}(Y_n) - f_{\alpha}(\bar{Y}_n)|^2 du \right. \\
&\quad \left. + \int_0^t E(\max_{n \in [0, n_u]} |f_{\alpha}(Y_n) - g_{\alpha, n}|^2 du) \right] \delta^{p(\alpha)} + K_2 \delta^{2\gamma} \\
&\leq K_3 \int_0^t H_u du + K_4 \delta^{2\gamma},
\end{aligned} \tag{3.40}$$

where the second step comes from Lipschitz condition on  $f_{\alpha}$  and (3.37). Finally by Gronwall inequality,

$$E \left( \max_{0 \leq n \leq n_T} |Y_n - \bar{Y}_n|^2 \right) \leq K \delta^{2\gamma}. \tag{3.41}$$

Note,

$$|Y_n - X_{\tau_n}|^2 \leq 2 |Y_n - \bar{Y}_n|^2 + 2 |\bar{Y}_n - X_{\tau_n}|^2, \tag{3.42}$$

so

$$E \left( \max_{0 \leq n \leq n_T} |Y_n - X_{\tau_n}|^2 \right) \leq K \delta^{2\gamma} \tag{3.43}$$

## 3.2 Platen scheme

Platen scheme is:

$$\begin{aligned}
Y_{n+1} &= Y_n + a\Delta + b\Delta W_n + \frac{1}{2\sqrt{\Delta}}(b(Y_n^*) - b)((\Delta W_n)^2 - \Delta), \\
Y_n^* &= Y_n + a\Delta + b\sqrt{\Delta}.
\end{aligned} \tag{3.44}$$

It can be written as:

$$Y_{n+1} = Y_n + g_{(0),n} I_{(0)} + g_{(1),n} I_1 + g_{(1,1),n} I_{(1,1)} + R_n, \quad (3.45)$$

$$R_n = (b(Y_n^*) - b - bb'\sqrt{\Delta}) \times \frac{1}{2\sqrt{\Delta}}((\Delta W_n)^2 - \Delta) \quad (3.46)$$

$$= [ab'\Delta + \frac{1}{2}b''(Y_n + \theta(a\Delta + b\sqrt{\Delta}))(a\Delta + b\sqrt{\Delta})^2] \times \frac{1}{2\sqrt{\Delta}}((\Delta W_n)^2 - \Delta), \quad (3.47)$$

with:

$$g_{(0),n} = a = f_{(0)}, \quad g_{(1),n} = b = f_{(1)}, \quad g_{(1,1),n} = bb' = f_{(1,1)}.$$

Condition (3.37) holds exactly. To show (3.38), observe:  $\sum_{k=0}^n R_k$  is mean zero, bounded variance, and is martingale. With Doob inequality:

$$\begin{aligned} E(\max_{n \in [0, n_T]} |\sum_{k=0}^{n-1} R_k|^2) &\leq K\Delta E(\max_{n \in [0, n_T]} |\sum_{k=0}^{n-1} ((\Delta W_k)^2 - \Delta)|^2) \\ &\leq 4K\Delta \max_{n \in [0, n_T]} E(|\sum_{k=0}^{n-1} ((\Delta W_k)^2 - \Delta)|^2) \\ &\leq 4K\Delta \sum_{k=0}^{n_T-1} E(|(\Delta W_k)^2 - \Delta|^2) \\ &\leq 8KT\Delta^2. \end{aligned} \quad (3.48)$$

We conclude that Platen scheme is order one convergent.

### 3.3 Chang scheme

To simplify higher order schemes, consider  $b = \text{constant}$ , in the so called additive noise regime. Then the 2nd order Chang scheme is:

$$Y_{n+1} = Y_n + \frac{1}{2}(a(\bar{Y}_+) + a(\bar{Y}_-))\Delta + b\Delta W_n. \quad (3.49)$$

$$\bar{Y}_\pm = Y_n + a_n\Delta/2 + \frac{1}{\Delta}b(\Delta Z \pm \sqrt{2J_{(1,1,0)}\Delta - (\Delta Z)^2}), \quad (3.50)$$

$$\Delta Z = \int_0^\Delta W_s ds, \quad J_{(1,1,0)} = \frac{1}{2} \int_0^\Delta W_s^2 ds.$$

Cauchy-Schwartz:

$$2J_{(1,1,0)}\Delta - (\Delta Z)^2 \geq 0.$$

Note:  $E((\Delta Z)^2) = \Delta^3/3$ ,  $E(J_{(1,1,0)}) = \Delta^2/2$ .

The Ito-Taylor expansion up to 2nd order is:

$$Y_{n+1} = Y_n + a\Delta + b\Delta W + ba'\Delta Z + aa'\Delta^2/2 + b^2a''J_{(1,1,0)}, \quad (3.51)$$

Taylor expansion of  $a(\bar{Y}_\pm)$  about  $a(Y_n)$  allows one to write (3.49) in the general RK form with  $g_{\alpha,n} = f_{\alpha,n}$ ,  $\alpha \in A_2$ , and  $E(R_n^2) = O(\Delta^5)$ . Then (3.37) is true, and (3.38) holds with  $\gamma = 2$ . The Chang scheme is 2nd order accurate.

If  $b = b(t)$ ,  $a = a(t, x)$ , the scheme extends to:

$$\begin{aligned} Y_{n+1} = & Y_n + \frac{1}{2}(a(t_n + \frac{1}{2}\Delta, \bar{Y}_+) + a(t_n + \frac{1}{2}\Delta, \bar{Y}_-))\Delta \\ & + b(t_n)\Delta W + \frac{1}{\Delta}(b(t_{n+1}) - b(t_n))(\Delta W\Delta - \Delta Z). \end{aligned} \quad (3.52)$$

## 4 Project V: Due May 24 before lecture

**V-1** Finish the proof in (2.31).

**V-2** Consider, for  $t \geq t_0 = 0$ ,

$$dX_t = \left( \frac{2}{1+t} X_t + (1+t)^2 \right) dt + (1+t)^2 dW_t \quad (4.53)$$

with initial value  $X_0 = 1$ .

(a) Verify it has exact solution,

$$X_t = (1+t)^2 (1 + W_t + t). \quad (4.54)$$

- (b) Approximate  $X_T$  with scheme (3.52), for  $T = 0.5$  in which  $J_{(1,1,0)}$  is approximated by  $J_{(1,1,0)}^p$  with  $p = 15$ . Conduct such approximation for equal step sizes  $\delta = 2^{-1}, 2^{-2}, 2^{-3}, 2^{-4}, 2^{-5}$  and record the absolute error.
- (c) Plot  $\log_2$  of the absolute error against  $\log_2 \delta$  and explain what you see.