

1 Introduction

The content of these notes is also covered by chapter 3 section C of [1]. An alternative reference for Brownian motion and its properties is provided by Varadhan's lecture notes [2].

2 Kolmogorov–Čentsov theorem

Theorem 2.1 (Kolmogorov–Čentsov). *If $\xi_t(\omega)$ is a stochastic process on (Ω, \mathcal{F}, P) satisfying*

$$\langle |\xi_t - \xi_s|^\beta \rangle \leq C |t - s|^{1+\alpha}$$

for some positive constants α, β and C , then if necessary, $\xi_t(\omega)$ can be modified for each t on a set of measure zero, to obtain an equivalent version $\tilde{\xi}_t(\omega)$ that is almost surely continuous with exponent γ for every $\gamma \in [0, \alpha/\beta]$ and some $\delta > 0$:

$$P \left(\omega; \sup_{\substack{0 < t-s < h(\omega) \\ t, s \in [0, T]}} \frac{|\tilde{\xi}_t(\omega) - \tilde{\xi}_s(\omega)|}{|t-s|^\gamma} \leq \delta \right) = 1$$

Proof. Let

$$\mathcal{T}_n = \left\{ \frac{jT}{2^n} \right\}_{j=0}^{2^n}$$

The countable union $\mathcal{T} := \cup_{n=0}^{\infty} \mathcal{T}_n$ is a *countable dense* subset of $[0, T]$. By linear interpolation we can construct a sequence of approximations to the original process $\xi(t)$ coinciding with it on dyadic rationals e.g.

- the \mathcal{T}_n -based interpolation is

$$\xi_n(t) = \xi \left(\frac{jT}{2^n} \right) + \left(t - \frac{jT}{2^n} \right) \frac{\xi \left(\frac{(j+1)T}{2^n} \right) - \xi \left(\frac{jT}{2^n} \right)}{\frac{T}{2^n}} \quad \& \quad t \in \left[\frac{jT}{2^n}, \frac{(j+1)T}{2^n} \right] \equiv \mathcal{I}_j^n$$

- The \mathcal{T}_{n+1} -based interpolation is

$$\xi_{n+1}(t) = \xi \left(\frac{jT}{2^n} \right) + \left(t - \frac{jT}{2^n} \right) \frac{\xi \left(\frac{(2j+1)T}{2^{n+1}} \right) - \xi \left(\frac{jT}{2^n} \right)}{\frac{T}{2^{n+1}}} \quad \& \quad t \in \left[\frac{jT}{2^n}, \frac{(2j+1)T}{2^{n+1}} \right] \equiv \mathcal{I}_{2j}^{n+1}$$

so that $t \in \mathcal{T}_n$ we have

$$\xi(t) = \xi_n(t) = \xi_{n+1}(t) \quad t \in \mathcal{T}_n$$

and for $t \in \mathcal{I}_j^n \cap \mathcal{I}_{2j}^{n+1}$

$$|\xi_{n+1}(t) - \xi_n(t)| = \left(t - \frac{jT}{2^n} \right) \left| \frac{\xi \left(\frac{(2j+1)T}{2^{n+1}} \right) - \xi \left(\frac{jT}{2^n} \right)}{\frac{T}{2^{n+1}}} - \frac{\xi \left(\frac{(j+1)T}{2^n} \right) - \xi \left(\frac{jT}{2^n} \right)}{\frac{T}{2^n}} \right|$$

which can be rewritten as

$$|\xi_{n+1}(t) - \xi_n(t)| = \frac{2^{n+1}}{T} \left(t - \frac{jT}{2^n} \right) \left| \xi \left(\frac{(2j+1)T}{2^{n+1}} \right) - \frac{\xi \left(\frac{jT}{2^n} \right) + \xi \left(\frac{(j+1)T}{2^n} \right)}{2} \right|$$

so that by the triangular inequality:

$$\begin{aligned} |\xi_{n+1}(t) - \xi_n(t)| &\leq \frac{\left| \xi\left(\frac{2j+1}{2^{n+1}}T\right) - \xi\left(\frac{j}{2^n}T\right) \right| + \left| \xi\left(\frac{2j+1}{2^{n+1}}T\right) - \xi\left(\frac{j+1}{2^n}T\right) \right|}{2} \\ &\leq \left| \xi\left(\frac{2j+1}{2^{n+1}}T\right) - \xi\left(\frac{j+1}{2^n}T\right) \right| \vee \left| \xi\left(\frac{2j+1}{2^{n+1}}T\right) - \xi\left(\frac{j}{2^n}T\right) \right| \end{aligned}$$

The inequality entails that

$$\sup_{0 \leq t \leq T} |\xi_{n+1}(t) - \xi_n(t)| \leq \sup_{0 \leq j < 2^{n+1}} \left| \xi\left(\frac{2j+1}{2^{n+1}}T\right) - \xi\left(\frac{j+1}{2^n}T\right) \right| \vee \left| \xi\left(\frac{2j+1}{2^{n+1}}T\right) - \xi\left(\frac{j}{2^n}T\right) \right|$$

The following inequalities hold true:

- if the occurrence of the event B implies the occurrence of the event A we must have $A \subseteq B$ and therefore $P(A) \leq P(B)$ which translates for us into

$$\begin{aligned} &P\left(\sup_{0 \leq t \leq 1} |\xi_{n+1}(t) - \xi_n(t)| \geq 2^{-n\gamma}\right) \\ &\leq P\left(\sup_{0 \leq j \leq 2^{n+1}-1} \left| \xi\left(\frac{j+1}{2^{n+1}}T\right) - \xi_n\left(\frac{j}{2^{n+1}}T\right) \right| \geq 2^{-n\gamma}\right) \end{aligned}$$

- as $P(\cup_k A_k) \leq \sum_k P(A_k)$:

$$\begin{aligned} &P\left(\sup_{0 \leq j \leq 2^{n+1}-1} \left| \xi\left(\frac{j+1}{2^{n+1}}T\right) - \xi_n\left(\frac{j}{2^{n+1}}T\right) \right| \geq 2^{-n\gamma}\right) \\ &\leq 2^{n+1} \sup_{0 \leq j \leq 2^{n+1}-1} P\left(\left| \xi\left(\frac{j+1}{2^{n+1}}T\right) - \xi_n\left(\frac{j}{2^{n+1}}T\right) \right| \geq 2^{-n\gamma}\right) \end{aligned}$$

- By Čebyšev inequality and using the theorem's hypothesis

$$\begin{aligned} &P\left(\left| \xi\left(\frac{j+1}{2^{n+1}}T\right) - \xi_n\left(\frac{j}{2^{n+1}}T\right) \right| \geq 2^{-n\gamma}\right) \\ &\leq \frac{\prec \left| \xi\left(\frac{j+1}{2^{n+1}}T\right) - \xi_n\left(\frac{j}{2^{n+1}}T\right) \right|^\beta \succ}{2^{-n\beta\gamma}} \leq 2^{n\beta\gamma} C \left(\frac{T}{2^{n+1}}\right)^{\alpha+1} \end{aligned}$$

Gleaning the above information together we obtain

$$P\left(\sup_{0 \leq t \leq T} |\xi_n(t) - \xi_{n+1}(t)|\right) \leq C T^{\alpha+1} \left(\frac{1}{2}\right)^{n(\alpha-\gamma\beta)}$$

If

$$\gamma > \frac{\alpha}{\beta}$$

the inequality entitle us to apply Borel-Cantelli lemma and conclude

$$\sup_{0 \leq t \leq T} |\xi_n(t) - \xi_{n+1}(t)| \rightarrow 0 \quad a.s.$$

and consequently

$$\lim_{n \uparrow \infty} \xi_n(t) = \tilde{\xi}(t)$$

The limit $\tilde{\xi}(t)$ will be continuous on $[0, T]$ and will coincide with $\xi(t)$ on \mathcal{T} thereby establishing our result. \square

3 Summary of notions of convergence

There are free notion of convergence

- *Convergence in probability*: $\{\xi_i\}_{i=1}^{\infty}$ converges to ξ in probability if for every positive ϵ

$$\lim_{n \uparrow \infty} P(|\xi_n - \xi| < \epsilon) = 0$$

- *Mean square convergence*: $\{\xi_i\}_{i=1}^{\infty}$ converges to ξ in mean square if for every positive ϵ

$$\lim_{n \uparrow \infty} \langle (\xi_n - \xi)^2 \rangle = 0 \quad \xi_n \xrightarrow{n \uparrow \infty} \xi \quad m.s.$$

- *Almost sure convergence*: $\{\xi_i\}_{i=1}^{\infty}$ converges to ξ almost surely (i.e. $P = 1$) if the event

$$\left\{ \xi_n \xrightarrow{n \uparrow \infty} \xi \right\} \equiv \left\{ \omega \in \Omega \mid \xi_n(\omega) \xrightarrow{n \uparrow \infty} \xi \right\}$$

has probability one.

Convergence in probability is the weakest notion

$$\begin{aligned} \xi_n \xrightarrow{m.s.} \xi &\Rightarrow \xi_n \xrightarrow{P} \xi && \text{by Čebyšev} \\ \xi_n \xrightarrow{a.s.} \xi &\Rightarrow \xi_n \xrightarrow{P} \xi \end{aligned}$$

Almost sure convergence *does not* imply mean square convergence:

$$\xi_n \xrightarrow{a.s.} \xi \not\Rightarrow \xi_n \xrightarrow{m.s.} \xi$$

For example let $\{\xi_n\}_{n=0}^{\infty}$ independent uniformly distributed such that

$$\xi_n = \begin{cases} n & \omega \in [0, 1/n] \\ 0 & \text{otherwise} \end{cases}$$

then

$$\xi_n \xrightarrow{a.s.} 0$$

but

$$\langle \xi_n^2 \rangle = n$$

Conversely, mean square convergence *does not* almost sure convergence:

$$\xi_n \xrightarrow{m.s.} \xi \not\Rightarrow \xi_n \xrightarrow{a.s.} \xi$$

4 Non differentiability of the Brownian motion

Even if we adopt the weakest notion of convergence, i.e. convergence in probability, Brownian motion turns out to be non-differentiable.

$$p_{\frac{w_{t+h} - w_t}{h}}(x) = \int_{\mathbb{R}} dy \delta^{(1)}\left(x - \frac{y}{h}\right) \frac{e^{-\frac{y^2}{2\sigma^2 h}}}{\sqrt{2\sigma^2 h}} = \frac{\sqrt{|h|} e^{-\frac{h x^2}{2\sigma^2}}}{\sqrt{2\pi\sigma^2}} \xrightarrow{h \downarrow 0} 0$$

so that for any $c \in \mathbb{R}_+$

$$P\left(-c \leq \frac{w_{t+h} - w_t}{h} \leq c\right) \leq C \sqrt{\frac{|h|}{2\pi}} \xrightarrow{h \downarrow 0} 0$$

the measure concentrates for **infinite values** of the differential ratio.

References

- [1] L.C. Evans, *An Introduction to Stochastic Differential Equations*, lecture notes, <http://math.berkeley.edu/~evans/>
- [2] S.R.S. Varadhan, *Stochastic Calculus, lecture notes*, <http://www.math.nyu.edu/faculty/varadhan/stochastic.fall108.html>
- [3] T. Sapsis, and P. Lermusiaux, *Dynamically Orthogonal field equations for continuous stochastic dynamical systems*, *Physica D*, **238** 2347 (2009) and http://web.mit.edu/sapsis/www/papers/DO_fld_eqs.pdf