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# Statistics 212: Lecture 19 (April 14, 2025)

## Stochastic Differential Equations II

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### 1 Uniqueness of SDE Solutions

Typical set-up: we have a stochastic differential equation

$$dX_t = \sigma(X_t)dB_t + v_t dt$$

where  $\sigma, v$  are Lipschitz from  $\mathbb{R}$  to  $\mathbb{R}$ . Last time, we showed the existence of a stochastic process that satisfies this SDE using Picard iteration. To demonstrate uniqueness, we can suppose that  $X_t$  and  $Y_t$  are solutions to this SDE. Consider  $Z_t = X_t - Y_t$ . We have

$$Z_t = X_t - Y_t = \int_0^t [\sigma(X_s) - \sigma(Y_s)]dB_s + \int_0^t [v(X_s) - v(Y_s)]ds,$$

and note that  $|v(X_s) - v(Y_s)| \leq L|Z_t|$  from some Lipschitz bounding. Applying Ito's formula to  $Z_t^2$ , we also get

$$d(Z_t^2) = 2Z_t(\sigma(X_t) - \sigma(Y_t))dB_t + (\sigma(X_t) - \sigma(Y_t))^2 dt + 2Z_t(v(X_t) - v(Y_t))dt,$$

implying that

$$E[Z_t^2] \leq L \int_0^t E[Z_s^2]ds.$$

If we let  $f(t) = \int_0^t E[Z_s^2]ds$ , this implies  $f'(t) \leq CLf(t)$  by some Lipschitz bounding, which further implies  $f(t)e^{-CLt}$  is decreasing with  $f(0) = 0$ . Thus,  $f(t) = 0$  for all  $t$ . At any given time, we see  $X_t = Y_t$  almost surely.

**Definition 1.1** (Locally Lipschitz). We call  $\sigma$  locally Lipschitz if  $|\sigma(x) - \sigma(y)| \leq L(R)|x - y|$  if  $|x|, |y| \leq R$ .

**Question 1.2.** We assumed that  $\sigma, v$  are globally Lipschitz. What if they are locally Lipschitz, e.g.,  $dX_t = X_t^2 dB_t + X_t^3 dt$ ?

Existence and uniqueness will hold until an “explosion time.” For motivation behind this, let's take an ODE  $dX_t = X_t^2 dt$ . A solution to this is  $X_t = \frac{1}{1-t}$ , and this process will blow up when approaching  $t = 1$ . For any  $R > 0$ , we can modify coefficients outside  $B_R(0)$ , the ball centered at 0 with radius  $R$ . We can construct the following functions:

$$\sigma^R(x) = \begin{cases} \sigma(x), & |x| \leq R \\ \sigma(Rx/|x|), & |x| \geq R \end{cases}$$

$$v^R(x) = \begin{cases} v(x), & |x| \leq R \\ v(Rx/|x|), & |x| \geq R. \end{cases}$$

Note that  $\sigma^r, v^r$  are globally Lipschitz, and define  $\tau^R$  to be the first time  $|X_t| = R$ . Then  $\lim_{R \rightarrow \infty} \tau^R = \tau$ , which is the “explosion time.”

## 2 Another characterization of BM

**Definition 2.1.**  $X_t$  is a “strong solution” to a SDE if it is adapted to the filtration by the driving Brownian motion  $B_t$ , i.e.,  $\mathcal{F}_t^X \subseteq \mathcal{F}_t^B$ .  $X_t$  is a “weak solution” if it is adapted wrt  $\mathcal{F}_t \supseteq \mathcal{F}_t^B$  and  $B_t$  is BM with respect to  $\mathcal{F}_t$ .

*Example.*  $dX_t = \text{sign}(X_t)dB_t$  where  $\text{sign}(x) = 1$  for  $x \geq 0$  and  $-1$  otherwise.

This is an example of a SDE that has a weak solution but not a strong solution. By inspection,  $X_t$  should be a Brownian motion of some kind. So let  $X_t$  be BM, and  $B_t = \int_0^t \text{sign}(X_s) dX_s$  (we’ll show that this integral is a Brownian motion later). Since  $\text{sign}(X_s)^2 = 1$ , we see that this yields a solution. Define  $\mathcal{F}_t^X = \sigma(X_s : s \leq t)$  and  $\mathcal{F}_t^B = \sigma(B_s : s \leq t)$ . It turns out

$$\mathcal{F}_t^{|X|} = \mathcal{F}_t^B \subsetneq \mathcal{F}_t^X.$$

Some intuition behind this: if we look at  $\tilde{X}_t = -X_t$  for all  $t$ , this yields the same  $B_t$ . The existence of  $\tilde{X}_t$  means we don’t have uniqueness of solutions, and our solution isn’t even measurable with respect to the filtration generated by our Brownian motion. We see  $\int_0^t \text{sign}(X_s) dB_s$  is defined for any nice filtration  $\mathcal{F}$  such that  $B_t$  is a Brownian motion with respect to  $\mathcal{F}$ . For this example, we have to enlarge our state space to find a solution, so we have  $\mathcal{F}_t \supsetneq \mathcal{F}_t^B$ .

**Theorem 2.2** (Levy’s Characterization of BM). *BM is the unique continuous martingale such that  $B_t^2 - t$  is also a martingale. In other words, any such process that satisfies the above characteristics has the law of BM on  $C([0, 1])$ .*

**Question 2.3.** Is  $B_t = \int_0^t \text{sign}(X_s) dX_s$  BM?

**Question 2.4.** Can we show any weak solution  $X_t$  is BM?

Yes! Both questions can be shown with the above theorem.

*Proof.* We cover two proofs to Levy’s characterization of BM.

1. You can define Ito integration with respect to any continuous martingale  $X_t$ . You need a continuous increasing process  $A_t$  such that  $X_t^2 - A_t$  is a martingale. (Non-trivial fact that you can always construct an  $A_t$  given that  $X_t$  is a martingale). For example, if  $X_t = \int_0^t \sigma_s dB_s$ , then  $A_t = \int_0^t \sigma_s^2 ds$ .

Generalized Ito’s lemma:

$$f(X_t) = \int_0^t f'(X_s) dX_s + \frac{1}{2} \int_0^t f''(X_s) dA_s.$$

Working with characteristic functions, let  $Y_t = f(X_t) = \exp(i\omega X_t)$ . If  $X_t$  and  $X_t^2 - t$  are martingales, then Ito results in

$$Y_t - Y_0 = \int_0^t i\omega Y_s dX_s - \frac{\omega^2}{2} \int_0^t Y_s ds.$$

We have

$$E[Y_t] = 1 - \frac{\omega^2}{2} \int_0^t E[Y_s] ds \implies E[Y_t] = \exp(-\omega^2 t/2),$$

meaning that  $X_t \sim \mathcal{N}(0, t)$ .

2. Proving martingale CLT in discrete time. The idea is start with  $X_t$  and discretize time into buckets, resulting in a discrete-time martingale. Somehow using CLT to get a Gaussian from this martingale given our understanding of the variance of each bucket.

**Theorem 2.5** (Levy). *Suppose that we have a triangular array  $\{\xi_{n,i}\}_{i \leq m(n)}$  with  $E[\xi_{n,i} | \mathcal{F}_{n,i-1}] = 0$  and  $\xi_{n,i}$  adapted to  $\{\mathcal{F}_{n,i}\}_{i \leq m(n)}$ . Define  $S_{n,k} = \sum_{i=1}^k \xi_{n,i}$  and  $V_{n,k}^2 = \sum_{i=1}^k E[\xi_{n,i}^2 | \mathcal{F}_{n,i-1}]$  (defining the variance as  $\sigma_i^2$ ). Assume  $V_{n,m(n)}^2 = 1$ . Recall the Lindeberg condition which states for all  $\delta$ , we have  $\lim_{n \rightarrow \infty} \sum_{i=1}^{m(n)} E[\xi_{n,i}^2 \mathbf{1}_{|\xi_{n,i}| > \delta}] = 0$ .*

*Then  $S_{n,m(n)} \xrightarrow{d} \mathcal{N}(0, 1)$ .*

For our use case here, we can let  $\xi_{n,i} = X_{i/n} - X_{(i-1)/n}$  and  $m(n) = n$ . So  $S_{n,n} = X_1$ . By continuity, the Lindeberg condition follows (choosing  $n$  large enough for every  $\delta$ ). Thus,  $X_1 \sim \mathcal{N}(0, 1)$ .

□

Next time, we'll go through a proof of this martingale CLT using the Lindeberg replacement method!