

Math 562 Fall 2020

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Outline

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- 1 General Info
- 2 8.2 The Lipschitz Case

HW6 is due Friday, 11/20, at noon. Please submit your homework via the course Moodle page. Make sure that your HW is uploaded successfully.

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- 1 General Info
- 2 8.2 The Lipschitz Case**

Assumptions

The functions σ and b are continuous on $\mathbb{R}_+ \times \mathbb{R}^d$ and Lipschitz in the variable x : There exists a constant K such that, for every $t \geq 0, x, y \in \mathbb{R}^d$,

$$|\sigma(t, x) - \sigma(t, y)| \leq K|x - y|;$$

$$|b(t, x) - b(t, y)| \leq K|x - y|.$$

Theorem 8.3

Under the preceding assumptions, pathwise uniqueness holds for $E(\sigma, b)$, and, for every choice of the filtered probability space $(\Omega, \mathcal{F}, (\mathcal{F}_t)_{t \in [0, \infty]}, \mathbb{P})$ and of the (\mathcal{F}_t) -Brownian motion B , for every $x \in \mathbb{R}^d$, there exists a (unique) strong solution of $E_x(\sigma, b)$.

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Proof of Theorem 8.3

Last time we have proved the first assertion of the theorem, that is, pathwise uniqueness holds. Now we prove the second assertion. We construct a solution using Picard's approximation method.

Define inductively

$$X_t^0 = x,$$

$$X_t^1 = x + \int_0^t \sigma(s, x) dB_s + \int_0^t b(s, x) ds,$$

$$X_t^n = x + \int_0^t \sigma(s, X_s^{n-1}) dB_s + \int_0^t b(s, X_s^{n-1}) ds.$$

The stochastic integrals are well defined since one verifies by induction that, for every n , the process X^n is adapted and has continuous sample paths.

Proof of Theorem 8.3 (cont)

It suffices to show that, for every $T > 0$, there is a strong solution of $E_x(\sigma, b)$ on the time interval $[0, T]$. Indeed, the uniqueness part of the argument will then allow us to get a (unique) strong solution on \mathbb{R}_+ that will coincide with the solution on $[0, T]$ up to time T .

We fix $T > 0$ and, for every $n \geq 1$ and every $t \in [0, T]$, we set

$$g_n(t) = \mathbb{E} \left[\sup_{s \in [0, t]} |X_s^n - X_s^{n-1}|^2 \right].$$

We will bound the functions g_n by induction on n (at present, it is not yet clear that these functions are finite). The fact that the functions $\sigma(\cdot, x)$ and $b(\cdot, x)$ are continuous, hence bounded, over $[0, T]$ implies that there exists a constant C'_T such that $g_1(t) \leq C'_T$ for every $t \in [0, T]$ (use Doob's L^2 inequality for the stochastic integral term).

Proof of Theorem 8.3 (cont)

Then we observe that

$$\begin{aligned} X_t^{n+1} - X_t^n &= \int_0^t (\sigma(s, X_s^n) - \sigma(s, X_s^{n-1})) dB_s + \int_0^t (b(s, X_s^n) - b(s, X_s^{n-1})) ds. \end{aligned}$$

Hence, using the case $p = 2$ of the Burkholder-Davis-Gundy inequalities in the second bound (and writing $C_{(2)}$ for the constant in this inequality),

$$\begin{aligned} &\mathbb{E} \left[\sup_{s \in [0, t]} |X_s^{n+1} - X_s^n|^2 \right] \\ &\leq 2\mathbb{E} \left[\sup_{s \in [0, t]} \left| \int_0^s (\sigma(u, X_u^n) - \sigma(u, X_u^{n-1})) dB_u \right|^2 \right] \\ &\quad + 2\mathbb{E} \left[\sup_{s \in [0, t]} \left| \int_0^s (b(u, X_u^n) - b(u, X_u^{n-1})) du \right|^2 \right] \end{aligned}$$

Proof of Theorem 8.3 (cont)

$$\begin{aligned} &\leq 2C_{(2)}\mathbb{E}\left[\int_0^t (\sigma(u, X_u^n) - \sigma(u, X_u^{n-1}))^2 du\right] \\ &\quad + 2T\mathbb{E}\left[\int_0^t (b(u, X_u^n) - b(u, X_u^{n-1}))^2 du\right] \\ &\leq 2(C_{(2)} + T)K^2\mathbb{E}\left[\int_0^t |X_u^n - X_u^{n-1}|^2 du\right] \\ &\leq C_T\mathbb{E}\left[\int_0^t \sup_{r \in [0, u]} |X_r^n - X_r^{n-1}|^2 du\right], \end{aligned}$$

where $C_T = 2(C_{(2)} + T)K^2$. We have thus obtained that, for every $n \geq 1$,

$$g_{n+1}(t) \leq C_T \int_0^t g_n(u) du \tag{1}$$

Proof of Theorem 8.3 (cont)

Recall that $g_1(t) \leq C'_T$. An induction argument using (1) shows that, for every $n \geq 1$ and $t \in [0, T]$,

$$g_n(t) \leq C'_T (C_T)^{n-1} \frac{t^{n-1}}{(n-1)!}.$$

In particular, $\sum_{n=1}^{\infty} g_n(T)^{1/2} < \infty$, which implies that

$$\sum_{n=0}^{\infty} \sup_{t \in [0, T]} |X_t^{n+1} - X_t^n| < \infty, \quad \text{a.s.}$$

Hence the sequence $(X_t^n : t \in [0, T])$ converges uniformly on $[0, T]$ a.s., to a limiting process $(X_t : t \in [0, T])$, which has continuous sample paths. By induction, one also verifies that, for every n , X^n is adapted with respect to the (completed) canonical filtration of B , and the same holds for X .

Proof of Theorem 8.3 (cont)

Finally, from the fact that σ and b are Lipschitz in the variable x , we also get that, for every $t \in [0, T]$,

$$\lim_{n \rightarrow \infty} \left(\int_0^t \sigma(s, X_s) dB_s - \int_0^t \sigma(s, X_s^n) dB_s \right) = 0,$$

$$\lim_{n \rightarrow \infty} \left(\int_0^t b(s, X_s) ds - \int_0^t b(s, X_s^n) ds \right) = 0,$$

in probability (to deal with the stochastic integrals, we may use Proposition 5.8, noting that $|X_s^n - X_s|$ is dominated by $\sum_{k=0}^{\infty} \sup_{r \in [0, s]} |X_r^{k+1} - X_r^k|$). By passing to the limit in the induction equation defining X^n , we get that X solves $E_x(\sigma, b)$ on $[0, T]$. This completes the proof of the theorem.

In the following statement, $W(dw)$ stands for the Wiener measure on the canonical space $C(\mathbb{R}_+, \mathbb{R}^m)$ of all continuous functions from \mathbb{R}_+ to \mathbb{R}^m . ($W(dw)$ is the law of $(B_t : t \geq 0)$ if B is an m -dimensional Brownian motion started from 0).

Theorem 8.5

Under the assumptions of the preceding theorem, there exists, for every $x \in \mathbb{R}^d$, a mapping $F_x : C(\mathbb{R}_+, \mathbb{R}^m) \rightarrow C(\mathbb{R}_+, \mathbb{R}^d)$, which is measurable when $C(\mathbb{R}_+, \mathbb{R}^m)$ is equipped with the Borel σ -field completed by the W -negligible sets, and $C(\mathbb{R}_+, \mathbb{R}^d)$ is equipped with the Borel σ -field, such that the following properties hold:

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Theorem 8.5 (cont)

- (i) for every $t \geq 0$, $F_x(w)_t$ coincides $W(dw)$ a.s. with a measurable function of $(w_r : r \in [0, t])$;
- (ii) For every $w \in C(\mathbb{R}_+, \mathbb{R}^m)$, the mapping $x \mapsto F_x(w)$ is continuous;
- (iii) for every $x \in \mathbb{R}^d$, for every choice of the (complete) filtered probability space $(\Omega, \mathcal{F}, (\mathcal{F}_t), \mathbb{P})$ and of the m -dimensional (\mathcal{F}_t) -Brownian motion B with $B_0 = 0$, the process X_t defined $X_t = F_x(B)_t$ is the unique solution of $E_x(\sigma, b)$; furthermore, if U is an \mathcal{F}_0 -measurable real-valued random variable, the process $F_U(B)_t$ is the unique solution with $X_0 = U$.

Remark)

Assertion (iii) implies in particular that weak uniqueness holds for $E(\sigma, b)$: any solution of $E_x(\sigma, b)$ must be of the form $F_x(B)$ and its law is thus uniquely determined as the image of $W(dw)$ under F_x .

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Proof of Theorem 8.5

Again we consider only the case $d = m = 1$. Let \mathcal{N} be the class of all W -negligible sets in $C(\mathbb{R}_+, \mathbb{R})$, and, for every $t \in [0, \infty]$, define

$$\mathcal{G}_t = \sigma(w(s) : s \in [0, t]) \vee \mathcal{N}.$$

For every $x \in \mathbb{R}$, we write X^x for the solution of $E_x(\sigma, b)$ corresponding to the filtered probability space $(C(\mathbb{R}_+, \mathbb{R}), \mathcal{G}_\infty, (\mathcal{G}_t), W)$ and the (canonical) Brownian motion $B_t(w) = w(t)$. This solution exists and is unique (up to indistinguishability) by Theorem 8.3, noting that the filtration (\mathcal{G}_t) is complete by construction.

Let $x, y \in \mathbb{R}$ and let T_n be the stopping time defined by

$$T_n = \inf\{t \geq 0 : |X_t^x| \vee |X_t^y| \geq n\}.$$

Let $p \geq 2$ and $T \geq 1$. Using the Burkholder-Davis-Gundy inequalities, and then the Hölder inequality, we get, for $t \in [0, T]$,

Proof of Theorem 8.5 (cont)

$$\begin{aligned}
 & \mathbb{E} \left[\sup_{s \leq t} |X_{s \wedge T_n}^x - X_{s \wedge T_n}^y|^p \right] \\
 & \leq C_p \left(|x - y|^p + \mathbb{E} \left[\sup_{s \leq t} \left| \int_0^{s \wedge T_n} (\sigma(r, X_r^x) - \sigma(r, X_r^y)) dB_r \right|^p \right] \right. \\
 & \quad \left. + E \left[\sup_{s \leq t} \left| \int_0^{s \wedge T_n} (b(r, X_r^x) - b(r, X_r^y)) dr \right|^p \right] \right) \\
 & \leq C_p \left(|x - y|^p + C'_p \mathbb{E} \left[\left(\int_0^{t \wedge T_n} (\sigma(r, X_r^x) - \sigma(r, X_r^y))^2 dr \right)^{p/2} \right] \right. \\
 & \quad \left. + E \left[\left(\int_0^{t \wedge T_n} |b(r, X_r^x) - b(r, X_r^y)| dr \right)^p \right] \right)
 \end{aligned}$$

Proof of Theorem 8.5 (cont)

$$\begin{aligned}
 &\leq C_p \left(|x - y|^p + C'_p t^{\frac{p}{2}-1} \mathbb{E} \left[\int_0^t |\sigma(r \wedge T_n, X_{r \wedge T_n}^x) - \sigma(r \wedge T_n, X_{r \wedge T_n}^y)|^p dr \right] \right. \\
 &\quad \left. + t^{p-1} \mathbb{E} \left[\int_0^t |b(r \wedge T_n, X_{r \wedge T_n}^x) - b(r \wedge T_n, X_{r \wedge T_n}^y)|^p dr \right] \right) \\
 &\leq C''_p \left(|x - y|^p + T^p \int_0^t \mathbb{E}[|X_{r \wedge T_n}^x - X_{r \wedge T_n}^y|^p] dr \right),
 \end{aligned}$$

where the constants C_p, C'_p, C''_p depend on p (and on the constant K appearing in our assumption on σ and b) but not on n or on $x; y$ and T .

Proof of Theorem 8.5 (cont)

As the function $t \mapsto \mathbb{E}[\sup_{s \leq t} |X_{s \wedge T_n}^x - X_{s \wedge T_n}^y|^p]$ is bounded, Gronwall's lemma implies that, for $t \in [0, T]$,

$$\mathbb{E} \left[\sup_{s \leq t} |X_{s \wedge T_n}^x - X_{s \wedge T_n}^y|^p \right] \leq C_p'' |x - y|^p \exp(C_p'' T^p t),$$

hence, letting n tend to ∞ ,

$$\mathbb{E} \left[\sup_{s \leq t} |X_s^x - X_s^y|^p \right] \leq C_p'' |x - y|^p \exp(C_p'' T^p t).$$

The topology on the space $C(\mathbb{R}_+, \mathbb{R})$ is defined by the distance

$$d(w, w') = \sum_{k=1}^{\infty} \alpha_k \left(\sup_{s \leq k} |w(s) - w'(s)| \wedge 1 \right)$$

where the sequence of positive reals α_k can be chosen arbitrarily, provided that $\sum_{k=1}^{\infty} \alpha_k$ converges.

Proof of Theorem 8.5 (cont)

We may choose α_k so that

$$\sum_{k=1}^{\infty} \alpha_k \exp(C_p'' k^{p+1}) < \infty.$$

For every $x \in \mathbb{R}$, we consider X^x as a random variable with values in $C(\mathbb{R}_+, \mathbb{R})$. The preceding estimates and Jensen's inequality then show that

$$\mathbb{E}[(\mathbf{d}(X^x, X^y))^p] \leq \left(\sum_{k=1}^{\infty} \alpha_k \right)^{p-1} \sum_{k=1}^{\infty} \alpha_k \mathbb{E} \left[\sup_{s \leq k} |X_s^x - X_s^y|^p \right] \leq \bar{C}_p |x - y|^p,$$

with a constant \bar{C}_p independent of x and y . By Kolmogorov's continuity theorem applied to the process $(X^x : x \in \mathbb{R})$ with values in the space $E = C(\mathbb{R}_+, \mathbb{R})$ equipped with the distance \mathbf{d} , we get that $(X^x : x \in \mathbb{R})$ has a modification with continuous sample paths, which we denote by $(\tilde{X}^x : x \in \mathbb{R})$. We set $F_x(w) = \tilde{X}^x = (\tilde{X}_t^x)_{t \geq 0}$. Property (ii) is then obvious.

Proof of Theorem 8.5 (cont)

The mapping $w \mapsto F_x(w)$ is measurable from $C(\mathbb{R}_+, \mathbb{R})$ equipped with the σ -field \mathcal{G}_∞ into $C(\mathbb{R}_+, \mathbb{R})$ equipped with the Borel σ -field $\mathcal{C} = \sigma(w(s) : s \geq 0)$. Moreover, for every $t \geq 0$,

$F_x(w)_t = \tilde{X}_t^x(w) = X_t^x(w)$ a.s. is \mathcal{G}_t -measurable hence coincides $W(dw)$ a.s. with a measurable function of $(w(s) : 0 \leq s \leq t)$. Thus property (i) holds.