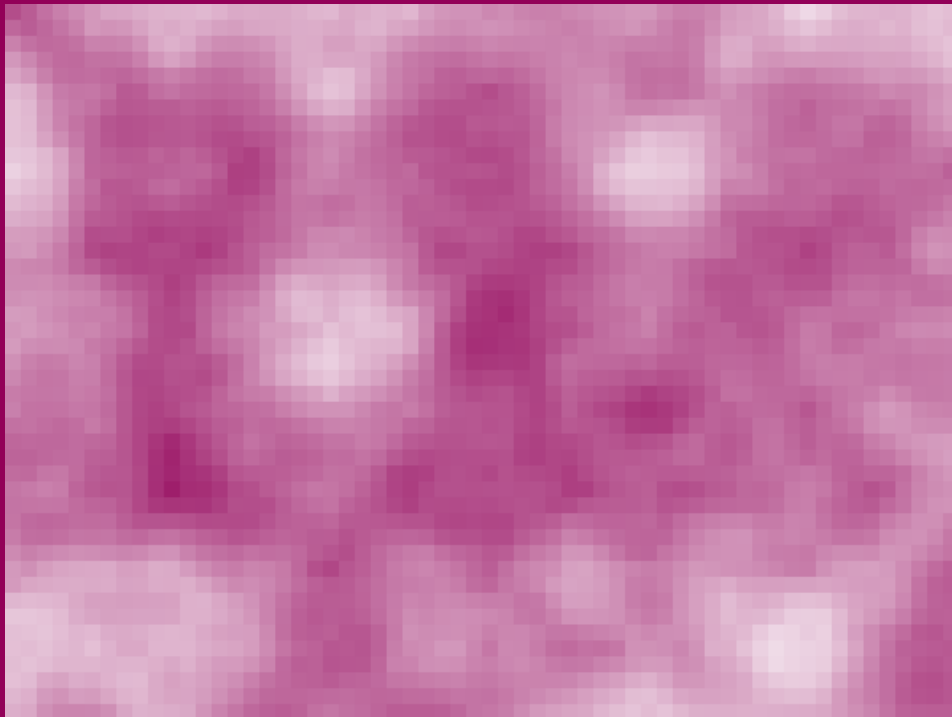

Lecture notes on Stochastic Analysis

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Based on the lectures of
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Chapter 0:

Manuel's notes

Warning

These are unofficial lecture notes written by a student. They are messy, will almost surely contain errors, typos and misunderstandings and may not be kept up to date! I do however try my best and use these notes to prepare for my exams. Feel free to email me any corrections to mh@mssh.dev or s6mlhinz@uni-bonn.de.
Happy learning!

General Information

- Ecampus: [Ecampus link](#)
- Basis: [Basis link](#)
- Website: None
- Time slot(s): Tuesday 12-14 and Thursday 12-14
- Exams: Oral: 23-25 July, 30 Sept. maybe 27. Sept
- Deadlines: ?
- Exercises: One of Tue 16-18(for now this is preferred), Friday 10-12. Starting 16.04.

- First halve based on Eberle and / or Gubinelli (be careful with Notation of dimensions!)

Start of lecture 01
(11.04.23)

Overview of the content

- Weak solutions of SDE
 - Martingale problem (characterization)
 - Time change (Dubin-Schwarz)
 - Change of measure (Girsonov)
- How to condition a diffusion to stay in a given domain forever (reflected BM, local time, ...)
- Lévy processes (a bit)
- Multiplicative stochastic heat-equation
 - relations with Kardar-Pavisi-Zhang class of growth models

Chapter 1:

Weak solutions of SDEs

SDE:

$$\begin{cases} dX_t = b(t, X_t)dt + \sigma(t, X_t)dB_t \\ X_0 = x_0 \text{ Initial condition} \end{cases} \quad (1.1)$$

- $X_t \in \mathbb{R}^d, B : n\text{-dim BM}$
- $b(t, x) : [b_k(t, x)]_{1 \leq k \leq d}$: **drift vector**
- $\sigma(t, x) = [\sigma_{k,l}(t, x)]_{\substack{1 \leq k \leq d \\ 1 \leq l \leq n}}$: **dispersion matrix**
- $a(t, x) = \sigma(t, x) \cdot \sigma(t, x)^\top$: **diffusion matrix**

1.1 Strong solutions

Definition 1.1. Let $(\Omega, \mathcal{F}, (\mathcal{F}_t)_{t \geq 0}, \mathbb{P})$ be a filtered probability space with $\mathcal{F}_t = \sigma(x_0, (B_s)_{0 \leq s \leq t})$

- a d -dim process X_t is a **strong solution** of equation 1.1 if:

- $X_t = x_0$ a.s.
- X_t is adapted to $\mathcal{F}_t \forall t \geq 0$
- X is a continuous semimartingale s.t. $\forall t \geq 0$:

$$\int_0^t (\|b(s, X_s)\| + \|\sigma(s, X_s)\|^2) ds < \infty \text{ a.s.}$$

- $X_t = x_0 + \int_0^t b(s, X_s) ds + \int_0^t \sigma(s, X_s) dB_s$

In the last semester we proved:

Theorem 1.2. Assume that b, σ are globally lipschitz with at most linear growth at ∞ (in space)
 $\implies \exists!$ strong solution of SDE.

Foundations of Stochastic Analysis Thm 8.6

Added remark. There exists $K > 0$ s.t. for all $x, y \in \mathbb{R}^d$: Globally Lipschitz:

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

Linear growth condition:

$$\|b(t, x)\| + \|\sigma(t, x)\| \leq K(1 + \|x\|)$$

Remark. For strong solutions, \mathcal{F}_t is given by the driving BM, which is given to us.
 $\implies X_t(\omega) = \Phi(x_0(\omega), (B_s)_{0 \leq s \leq t})$

1.2 Weak solutions

- For weak solutions we do not fix the driving brownian motion.

Definition 1.3. A **weak solution** of equation 1.1 is a **pair** of adapted processes (X, B) to a $(\Omega, \mathcal{F}, (\mathcal{F}_t)_{t \geq 0}, \mathbb{P})$ s.t.

- B is a n -dim BM
- X is a d -dim continuous semimartingale with
 1. $X_0 = x_0$ a.s.
 2. $\forall t \geq 0$

$$\int_0^t (\|b(s, X_t)\| + \|\sigma(t, X_t)\|^2) ds < \infty \text{ a.s.}$$

$$3. X_t = x_0 + \int_0^t b(s, X_s) ds + \int_0^t \sigma(s, X_s) dB_s$$

Remark. • The filtration $(\mathcal{F}_t)_{t \geq 0}$ is not necessarily the one generated by B

- If X is adapted to the filtration generated by the BM \implies we have strong solutions
- \exists weak solution which are not strong solutions
- Warning: To give a weak solution we really need to provide $(\Omega, \mathcal{F}, (\mathcal{F}_t)_{t \geq 0}, \mathbb{P}, X, B)$

Definition 1.4 (Uniqueness in law). An SDE 1.1 has **uniqueness in law** if given any two weak solutions $(\Omega, \mathcal{F}, (\mathcal{F}_t)_{t \geq 0}, \mathbb{P}, X, B)$ and $(\tilde{\Omega}, \tilde{\mathcal{F}}, (\tilde{\mathcal{F}}_t)_{t \geq 0}, \tilde{\mathbb{P}}, \tilde{X}, \tilde{B})$ satisfy:

$$\text{Law}_{\mathbb{P}}(X) = \text{Law}_{\tilde{\mathbb{P}}}(\tilde{X})$$

They agree on any set in the sigma algebra

Definition 1.5 (Pathwise uniqueness). The SDE 1.1 has pathwise uniqueness if, whenever the filtered space $(\Omega, \mathcal{F}, (\mathcal{F}_t)_{t \geq 0}, \mathbb{P})$ and $(B_t)_{t \geq 0}$ are fixed, then two solutions X, \tilde{X} with $X_0 = \tilde{X}_0$ are indistinguishable.

Example 1.6 (No strong solutions, no pathwise uniqueness, \exists weak solution & uniqueness in law by Tanaka).

$$\begin{cases} dX_t = \text{sgn}(X_t) dB_t \\ X_0 = 0 \end{cases} \quad (1.2)$$

or more generally $X_0 = Y$, where

$$\text{sgn}(x) = \begin{cases} 1 & x > 0 \\ -1 & x \leq 0 \end{cases}$$

Let W be a BM with $W_0 = Y$. Define

$$B_t := \int_0^t \text{sgn}(W_s) dW_s \text{ or } dB_t = \text{sgn}(W_t) dW_t$$

$$\implies dW_t = \text{sgn}(W_t) dB_t$$

$$\implies W_t = y + \int_0^t \text{sgn}(W_s) dB_s$$

B_t is a local martingale with

$$\langle B \rangle_t = \int_0^t \underbrace{(\text{sgn}(W_s))^2}_{=1} \underbrace{d\langle W \rangle_s}_{=ds} = t$$

Also $B_0 = 0$, therefore B is a BM (see Lévy characterization) $\implies W$ solves the SDE. For $Y = 0$, W **and** $-W$ solves the same SDE.

\implies

- exists weak solutions
- For $Y = 0$: no pathwise uniqueness
- Uniqueness in law (because the law is determined by X_t being a BM)
- No strong solution, because: $\mathcal{F}^B = \mathcal{F}^{|W|} \subsetneq \mathcal{F}^W$

Not a proof just yet, just the reason!

Example 1.7 (No solutions).

$$\begin{cases} dX_t = -\frac{1}{2X_t} 1_{X_t \neq 0} dt + dB_t \\ X_0 = 0 \end{cases} \quad (1.3)$$

Assume there exists a solution. Use Itô formula for X_t^2 , then:

$$\begin{aligned} X_t^2 &= 2 \int_0^t X_s dX_s + \int_0^t 1 ds \\ &= - \int_0^t 1_{X_s \neq 0} ds + 2 \int_0^t x_s dB_s + t \\ &= \int_0^t 1_{X_s = 0} ds + 2 \int_0^t X_s dB_s \end{aligned}$$

We will prove $\int_0^t 1_{X_s = 0} ds = 0 \implies X_t^2$ is a local martingale, $X_t^2 \geq 0$ (and therefore a supermartingale) and $X_0 = 0$ ($\implies \mathbb{E}(X_t^2) = 0$). If $X_t = 0 \implies \int_0^t 1_{X_s = 0} ds = t \implies 0 = dB_t$ which are contradictions!

Remark. If $X = \underbrace{M}_{\in \mathcal{M}_{loc}} + \underbrace{A}_{\in \mathcal{A}}$

$$\implies \forall b \in \mathbb{R} \int_0^t 1_{X_s = b} d\langle M \rangle_s = 0$$

Sheet 0:

- Mail: Alexander.Becker@uni-bonn.de
- Exercise sheet handed in Fr 10 am via eCampus
- Groups of 3

Motivation:

in the last semester: Introduction to stochastic analysis

- Diffusion processes & SDEs

Here: Deepen the knowledge & have fun

- Learn SDE techniques and get other results
- Modify diffusion processes to behave in a certain way
- Ex. diffusion bridge (Brownian bridge)
- Condition a diffusion to stay in a domain
 - Ex. Condition BM to stay positive
 - Old SDE: $dB_t = dB_t$

- New SDE: $dX_t = \frac{1}{X_t} dx + dB_t \rightarrow P(X_t \in \cdot) = P(B_t \in \cdot | B > 0 \text{ forever})$
- $D \subset \mathbb{R}^d$ open domain, X diffusion process, with generator $L = \Sigma b \partial_i + \frac{1}{2} \sum a^{ij} \partial_i \partial_j$

$$Y := (X | X \in D \text{ forever})$$

get drift term $\nabla \log \phi_0$, where ϕ_0 is the lowest eigenfunction of $-L$ on D with dirichlet boundary.

Recap:

Brownian motion:

Added definition. $B_0 = 0$, independent $\mathcal{E} \mathcal{N}(0, t_i - t_{i-1})$ increments, $t \mapsto B_t(\omega)$ continuous.

Regularity of path $t \mapsto B_t(\omega)$:

- nowhere differentiable
- α -locally Hölder continuous $\iff \alpha < \frac{1}{2}$
- Quadratic variation $\langle B \rangle_t = t$
- Generator $\frac{\Delta}{2}$
- Recurrent $\iff ds^2?$

Itô-Integral:

1. If X simple process \implies RS-Integral
2. Itô isometry $\mathcal{E} \rightarrow \{L^2 - \text{local martingale}\}, \mathcal{E} \subset L^2(d[M])$ dense
3. general $X : \int X dM$ as L^2 -limit

Added remark (Itô formula). •

$$df(t, X_t) = \partial_t f(t, X_t) dt + \partial_x f(t, X_t) dX_t + \frac{1}{2} f(t, X_t) d\langle X \rangle_t$$

- associative $\int X d(\int Y dZ) = \int XY dZ$
- If M local martingale $\implies \int X dM$ local martingale

SDEs:

$$dX_t = b(t, X_t) dt + \sigma(t, X_t) dB_t$$

- ex./ uniqueness: b, σ locally Lipschitz \implies strong ex.+pathwise uniqueness until explosion time
- globally Lipschitz (in space) and linear growth ($|b(t, x)| + |\sigma(t, x)| \leq C(1 + |x|)$), $e = \infty$.

Problem 00.1: SDE

Let B be a one-dimensional Brownian motion (starting from 0) and let $X_t = \sin(B_t)$.

1. Determine the SDE of X_t
2. Discuss the existence and/or uniqueness of strong solutions of the SDE
3. Determine whether the local martingale term is a martingale (Hint: for the integrability condition, think about the Itô isometry for instance)

Solution 00.1

1.:

Idea: Use Itô formula: $X_t = \sin(B_t) = f(B_t)$

$$dX_t = df(B_t) \stackrel{\text{Itô}}{=} \underbrace{\partial_x \cos(B_t)}_{\sqrt{1-X_t^2}} dB_t - \frac{1}{2} \underbrace{\sin(B_t)}_{X_t} dt$$

Careful: $\sqrt{1-X_t^2}$ is not inverse mapping, because it is always positive while $\cos(B_t)$ is not

2.:

Consider the SDE

$$dX_t = \begin{cases} -\frac{1}{2}X_t dt + \sqrt{1-X_t^2} dB_t \\ X_0 = x \in (-1, 1) \end{cases}$$

Coefficients: $b: [-1, 1] \rightarrow \mathbb{R}, b(x) = -\frac{1}{2}x$ and $\sigma: [-1, 1] \rightarrow \mathbb{R}, \sigma(x) = \sqrt{1-x^2}$

$$[\sigma]_{1,2} = \sup \frac{|\sigma(x) - \sigma(y)|}{|x - y|^{1/2}} < \infty$$

Probably σ^2 Lipschitz $\implies \sigma$ Hölder $\frac{1}{2}$

3.:

Problem 00.2: Time change

Let B be a one-dimensional Brownian motion (starting from 0). Let $Y_t = \int_0^t s^2 dB_s$.

1. Determine the SDE of Y_t
2. Find A_t such that Y_{A_t} is a (stopped) Brownian motion

Solution 00.2

1.: By Definition

$$dY_t = t^2 dB_t$$

2.: We use the Dubin-Schwarz theorem:

$$X_{\infty} \in \mathcal{M}_{\text{loc}}^0, \langle X \rangle_{\infty} = \infty, T_t := \inf\{s \geq 0 | \langle X \rangle_s \geq t\} = X_t^{[-1]}$$

$$\implies B_t := X_{T_t} \text{ 1 d BM w.r.t. } (F_{T_t})_{t \geq 0}, X_t = B_{\langle X \rangle_t}$$

$$\text{here: use } X_t = b(X_t)dt + \sigma(X_t)dB_t \implies d[X]_t = \sigma^2(X_t)dt$$

Problem 00.3: SDE and PDE

Let f be a function supported on $[0, 1]$, u the solution of

$$\frac{1}{2}u(t, x) = \frac{1}{2}x(1-x) + \frac{d^2}{dx^2}u(t, x), \quad u(0, x) = f(x).$$

Consider also the one-dimensional SDE (also known as Wright-Fisher diffusion)

$$dX_t = \sqrt{X_t(1-X_t)}dB_t$$

with $X_0 = x \in (0, 1)$.

1. For any fixed $t > 0$, define $M_s = u(t-s, X_s)$ for $s \in [0, t]$. Use Itô formula to show that M_s is a local martingale
2. Assume that f is bounded and there is a bounded solution of u . Show that $u(t, x) = \mathbb{E}_x(f(X_t))$.

Solution 00.3

1.:

$$\begin{aligned}
dM_s &= du(t-s, X_s) \\
&= -\partial_s u(t-s, X_s)ds + \underbrace{\partial_x u(t-s, X_s)}_{b(X_s)ds + \sigma(X_s)dB_s \text{ by asso.}} \underbrace{dX_s}_{\substack{= \sigma^2(X_s)ds \\ = \sigma^2(X_s)ds}} + \frac{1}{2} \partial_x^2 u(t-s, X_s) \underbrace{d[X]_s}_{= \sigma^2(X_s)ds} \\
&= \underbrace{(-\partial_s u + b\partial_x u + \frac{1}{2}\sigma^2 \partial_x^2 u)(t-s, X_s)}_{=0} ds + \partial_x u(t-s, X_s) \sigma(X_s) dB_s \\
&\implies dM_s = \partial_x u(t-s, X_s) \sigma(X_s) dB_s
\end{aligned}$$

(i.e.: $M_t - M_0 = \int_0^t \dots dB_s$)This is a purely stochastic integral against a (local) martingale \implies martingale.

2.:

- M_s true martingale:

1. $\forall t : \mathbb{E}(\sup_{[0,t]} |M|) < \infty$, for example: M bounded
2. $\forall t : \mathbb{E}(\sup_{[0,t]} [M]_t) < \infty$

 $M_s := u(t-s, X_s), u$ bounded $\implies M$ bounded $\implies M$ true martingale $w(s, x) := u(t-s, x)$

$$u(t, x) = w(0, x) = \mathbb{E}_x[w(0, X_0)] \stackrel{\text{martingale}}{=} \mathbb{E}_x[w(t, X_t)] = \mathbb{E}_x[u(0, X_t)] \stackrel{\text{PDE}}{=} \mathbb{E}_x[f(X_t)]$$

Feynman-Kac formula (later more general).

Also solvable by Kolmogorov Backward / forward equation

Start of lecture 02
(16.04.24)**Example 1.8** (Non-uniqueness of law, non pathwise uniqueness, with solutions).

$$\begin{cases} dX_t &= 1_{X_t \neq 0} dB_t \\ x_0 &= 0 \end{cases}$$

Then

$$X_t = 0 \forall t \geq 0$$

and

$$X_t = B_t \forall t \geq 0$$

both are solutions:

$$X_t - B_t = - \int_0^t 1_{X_s=0} dB_s \implies \langle X - B \rangle_t = \int_0^t 1_{X_s=0} d\langle B \rangle_s = 0$$

Let $\eta \sim \text{Ber}(\frac{1}{2})$ independent of $(B_t)_{t \geq 0}$ and define

$$\tilde{X}_t = \begin{cases} 0 & \eta = 0 \\ B_t & \eta = 1 \end{cases}$$

 $\implies \tilde{X}_t$ is adapted to $\sigma(\eta(B_s)_{0 \leq s \leq t})$, but not to $\sigma((B_s)_{0 \leq s \leq t})$ and therefore not a **strong solution**.

$$\begin{aligned} X_t &= \int_0^t 1_{X_s \neq 0} dB_s \\ &= \int_0^t (1 - 1_{X_s=0}) dB_s \\ B_t - \int_0^t 1_{X_s=0} dB_s \end{aligned}$$

Example 1.9 (No strong solution, weak solution, no pathwise uniqueness, no uniqueness in law).

$$\begin{cases} dX_t &= 1_{X_t \neq 1} \operatorname{sgn}(X_t) dB_t \\ X_0 &= 0 \end{cases}$$

Let Y_t be a solution of

$$\begin{cases} dY_t &= \operatorname{sgn}(Y_t) dB_t \\ Y_0 &= 0 \end{cases}$$

$\implies X_t := Y_{t \wedge \tau}$, where $\tau = \inf\{s \geq 0 \mid Y_s = 1\}$ is also a solution.

Theorem 1.10 (Yamada-Watanabe). If both existence of weak solutions and pathwise uniqueness hold, uniqueness in law also holds.

Moreover, \forall choices of $(\Omega, \mathcal{F}, (\mathcal{F}_t)_{t \geq 0}, \mathbb{P})$ and $(B_t)_{t \geq 0}$ then there exists a **strong solution**.

1.3 Lévy characterization

Example 1.11. Consider the SDE

$$\begin{cases} dX_t &= O_t dB_t \\ X_0 &= x_0 \end{cases}$$

where both X_t and B_t are d -dimensional and O_t is an adapted process (matrix) s.t. $O_t^\top O_t = 1 \forall t \geq 0$ i.e. O_t is a rotation

$$\begin{aligned} \implies X_t^k &= X_0^k + \sum_{l=1}^d \int_0^t O_s^{k,l} dB_s^l \\ \implies \langle X^k, X^{\bar{k}} \rangle_t &= \sum_{l, \bar{l}} \int_0^t O_s^{k,l} O_s^{\bar{k}, \bar{l}} d\langle B^l, B^{\bar{l}} \rangle_s = \sum_{l=1}^d \int_0^t O_s^{k,l} O_s^{\bar{k}, l} ds \\ &= \int_0^t \underbrace{(O_s O_s^\top)^{k, \bar{k}}}_{=1} ds = \delta_{k, \bar{k}} t \xrightarrow{\text{Lévy}} (X_t)_{t \geq 0} \text{ is also a } d\text{-dim BM starting from } x_0 \end{aligned}$$

Theorem 1.12 (Yamada, Watanabe). Let $dX_t = b(X_t)dt + \sigma(X_t)dB_t$ and assume that there exist both a increasing function $\rho(u) \geq 0$ s.t. $|\sigma(x) - \sigma(y)| \leq \rho(|x - y|) \forall x, y \in \mathbb{R}$ s.t.

$$\int_0^\infty \frac{1}{\rho(u)} du = \infty \implies 0 < u < C_1 \implies \rho(u) \leq C_2 u^{0.5}$$

and some increasing concave function $\gamma_1(u) \geq 0$ s.t.

$$|b(x) - b(y)| \leq \gamma_1(|x - y|) \forall x, y \in \mathbb{R}$$

and

$$\int_0^\infty \frac{1}{\gamma_1(u)} du = \infty.$$

Then pathwise uniqueness holds.

Theorem 1.13 (Storokhod). Assume that σ, b are continuous bounded functions

\implies there exist weak solutions to the SDE $dX_t = b(X_t)dt + \sigma(X_t)dB_t$.

1.4 Weak solutions and martingale problems

Let $(X_t)_{t \geq 0}$ be a weak solution of the SDE

$$\begin{cases} dX_t = b(t, X_t)dt + \sigma(t, X_t)dB_t \\ X_0 = 0 \end{cases}$$

on $(\Omega, \mathcal{F}, (\mathcal{F}_t)_{t \geq 0}, \mathbb{P})$.

$\implies X_t$ is a semimartingale s.t. $X_t^k = X_0^k + \int_0^t b(s, X_s)ds + \sum_{l=1}^n \int_0^t \sigma_{k,l}(s, X_s)dB_s^l$ and

$$\langle X^i, X^j \rangle_t = \int_0^t \underbrace{\sum_{l=1}^n \sigma_{i,l}(s, X_s) \sigma_{j,l}^T(s, X_s)}_{=a_{i,j}(s, X_s)} ds$$

1.4.1 Itô-Doeblin formula

Itô formula leads to

Proposition 1.14. For $f \in C^{1,2}(\mathbb{R}_+ \times \mathbb{R}^d)$, then

$$f(t, X_t) = f(0, X_0) + \int_0^t (\sigma^T \nabla f)(s, X_s) dB_s + \int_0^t \left[\left(\frac{\partial}{\partial s} + \mathcal{L} \right) \right] (s, x_s) ds$$

where $(\mathcal{L}f)(t, x) = \frac{1}{2} \sum_{k,l=1}^n a_{k,l}(t, x) \frac{\partial^2}{\partial x_k \partial x_l} f(t, x) + \sum_{k=1}^n b_k(t, x) \frac{\partial}{\partial x_k} f(t, x)$.

\mathcal{L} is called the **generator**

Remark. The Itô-Doeblin formula provides a connection between SDEs and PDEs as we have seen in the last part **Foundations of stochastic analysis**.

Example 1.15. Let $f \in C^{1,2}(\mathbb{R}_+ \times \mathbb{R}^d)$ be a solution of the PDE

$$\begin{aligned} \frac{\partial}{\partial t} f(t, x) + (\mathcal{L}f)(t, x) &= -g(t, x) & t \geq 0, x \in U \subseteq \mathbb{R}^d \\ f(t, x) &= \varphi(t, x) & t \geq 0, x \in \partial U. \end{aligned}$$

then $M_t := f(t, X_t) + \int_0^t g(s, X_s)ds \in \mathcal{M}_{loc}$ by proposition 1.14 and if f, g are bounded $M_t \in \mathcal{M}$.

$$T := \inf\{s \geq 0 | X_s \notin U\} \implies M_t^T := M_{T \wedge t} \in \mathcal{M}.$$

Furthermore, if we assume $T < \infty$ a.s. :

$$\mathbb{E}[M_t] = \mathbb{E}[\varphi(T, X_T)] + \mathbb{E}\left[\int_0^T g(s, X_s)ds\right] = \varphi(0, X_0)$$

where $X_0 = x_0$.

There are two special cases:

$g = 0 \implies$ yields the exit distributions, while

$\varphi = 0, g = 1$ yields the mean exit times.

Example 1.16 (Feynman-Kac formula). Let $t \in \mathbb{R}_+$ be finite. Assume $f : \mathbb{R}^d \rightarrow \mathbb{R}$ and $K : [0, t] \times \mathbb{R}^d \rightarrow \mathbb{R}_+$ be continuous functions. Assume that u is a $C^{1,2}$ -solution (bounded, for simplicity) of

$$\begin{cases} \frac{\partial u}{\partial s}(s, x) = \frac{1}{2} \Delta u(s, x) - K(s, x)u(s, x) & s \in [0, t], x \in \mathbb{R}^d \\ u(0, x) = f(x) & x \in \mathbb{R}^d \end{cases}$$

Then u has the stochastic representation $u(t, x) = \mathbb{E}_x \left[f(X_t) \exp \left(- \int_0^t K(t-s, X_s) ds \right) \right]$, where X_t is BM starting from $X_0 = x$.

Sketch. 1. Define $r(s, x) := u(t-s, x)$ for $s \in [0, t]$

2. Show: $M_s := \exp(-A_s)r(s, x)$ with $A_s = \int_0^s K(u, X_u)du$ is a local martingale. □

Remark. This is a reformulation of the formula from the last semester.

1.4.2 Martingale problem

A solution of an SDE is generically defined up to some **explosion time** ξ , where it either diverges or it exists a given domain $U \subset \mathbb{R}^d$ (open).

\implies For $k \in \mathbb{N}$ define $U_k := \{x \in U \mid |x| < k \wedge \text{dist}(x, U^c) \geq \frac{1}{k}\}$ with $U = \bigcup_{k \geq 1} U_k$ and

$$T_k := \{t \geq 0 \mid x_t \notin U_k\}.$$

A solution of the SDE $b(t, X_t)dt + \sigma(t, X_t)dB_t$ is defined up to $\xi = \sup_{k \geq 1} T_k$.

Start of lecture 03
(18.04.24)

Added remark. uniqueness of solution to the heat equation $\frac{1}{2}\Delta u - Ku$: not unique! Stochastic calculus: Paolo Baldi: 10.3-10.4

Define (\star) :

$$dX_t b(t, X_t)dt + \sigma(t, X_t)dB_t \text{ with } X_0 = x_0$$

Theorem 1.17 (Martingale problem). If X is a solution of (\star) up to time ζ , then $\forall f \in C^{1,2}(\mathbb{R}_+ \times U)$

$$M_t = f(t, X_t) - \int_0^t \left(\frac{\partial}{\partial s} + \mathcal{L} \right) f(s, x_s) ds, t < s$$

is a local martingale up to ζ and $M_t^{T_k}$ are localizing martingales.

Definition 1.18 (Martingale solutions). $(X_t)_{t \geq 0}$ is a **martingale solution** of (\star) if $\forall f \in C^2(\mathbb{R}^d)$

$$M_t^f = f(X_t) - f(X_0) - \int_0^t (\mathcal{L}f)(X_s) ds$$

is a continuous local martingale.

Theorem 1.19 (Equivalent definitions). The following are equivalent (for X_t being a solution of (\star)):

(a) $\forall f \in C^2(\mathbb{R}^d)$,

$$M_t^f := f(X_t) - f(X_0) - \int_0^t (\mathcal{L}f)(X_s) ds$$

is a local martingale

(b) The process in \mathbb{R}^d given by

$$M_t := X_t - X_0 - \int_0^t b(s, X_s) ds$$

is a d -dimensional local martingale with $\langle M^i, M^j \rangle_t = \int_0^t a_{i,j}(s, X_s) ds = \langle X^i, X^j \rangle_t$

(c) $\forall f \in C^{1,2}(\mathbb{R}_+ \times \mathbb{R}^d)$

$$\tilde{M}_t^f := f(t, X_t) - f(0, X_0) - \int_0^t \left(\frac{\partial}{\partial s} + \mathcal{L} \right) f(s, X - s) ds$$

is a local martingale

ds

Proof. $c \implies a$: by choosing f independent of t .

$a \implies b$: 1.: Choosing $f(X) = X_i$ implies

$$M_t^f = X_t^i - X_0^i - \int_0^t b_i(s, X_s) ds \in \mathcal{M}_{\text{loc}}$$

2.: $f(X) = X_i X_j$:

$$\begin{aligned} (\mathcal{L}f)(x) &= \frac{1}{2}a_{ij} + \frac{1}{2}a_{ji} + b_i X_j b_j X_i \\ a &= a^\top \implies a_{ij} b_i X_j b_j X_i \\ \implies M_t^f X_t^i X_t^j - X_0^i X_0^j - \int_0^t [a_{ij}(s, X_s) + b_i(s, X_s) X_s^j + b_j(s, X_s) X_s^i] ds \end{aligned}$$

$$\begin{aligned} X_t^i X_t^j - X_0^i X_0^j &\stackrel{\text{Integration by parts}}{=} \int_0^t X_s^i dX_s^j + \int_0^t X_s^j dX_s^i + \langle X^i, X^j \rangle_t \\ &= M_t^f + \int_0^t [a_{ij} + b_i X_s^j + b_j X_s^i] ds \end{aligned}$$

We can maybe also proof this by calculating X^2 ?

Here dX_s^i is the same as $b_i X_s^j$ is the same up to a local martingale term and $\langle X^i, X^j \rangle_t = \int_0^t a_{ij}(s, X_s) ds = \langle M^i, M^j \rangle_t$

$b \implies c$: By Proposition 1.14 If (use the next theorem) X was a weak solution $\implies \tilde{M}_t^f$ is a local martingale. \square

Theorem 1.20. Let $n = d$, assume $\sigma(t, x)$ is invertible $\forall t, x$ and $\sigma^{-1}(t, x)$ is uniformly bounded. T.f.a.e.:

- (a) $(X_t)_{t \geq 0}$ is a weak solution of the SDE (\star) on $(\Omega, \mathcal{F}, (\mathcal{F}_t)_{t \geq 0}, \mathbb{P}; B)$
- (b) $(X_t)_{t \geq 0}$ is a martingale solution of the SDE (\star)

This also works for $n \neq d$, but with a different proof

Proof. $a \implies b$: True

$b \implies a$: Goal construct a BM for the weak solution.

By proposition ?? $a \implies b$ $dM_t = dX_t - b(t, X_t)dt \in \mathcal{M}_{\text{loc}}$ and $d\langle M^i, M^j \rangle_t = a_{k,l}(t, X_t)dt$

$$\begin{aligned} \implies dX_t &= dM_t + b(t, X_t)dt \\ &= \sigma(t, X_t) d\tilde{B}_t + b(t, X - t)dt \end{aligned}$$

where $\tilde{B}_t := \sigma(s, X_s)^{-1} dM_s$

To see: \tilde{B}_t is a brownian motion.

$$\begin{aligned} \langle \tilde{B}^i, \tilde{B}^j \rangle_t &= \sum_{k,l} \int_0^t \sigma_{ij}^{-1} \sigma_{jl}^{-1} d\langle M^k, M^l \rangle_s \\ &\quad = \underbrace{a_{ij}}_{(\sigma^\top \sigma)_{kl}} ds \\ &= \sum_{k,l,p} \int_0^t \sigma_{ik}^{-1} \sigma_{kp} \sigma_{pl}^\top \sigma_{lj}^{-1} ds \\ &= \delta_{ij} \int_0^t 1 ds = \delta_{ij} t \end{aligned}$$

Then by the Lévy characterization \tilde{B} is a brownian motion. \square

Added remark. This is the first way to construct a weak solution: Solve a martingale problem! This is used a lot in practice.

1.5 Weak solutions and time change

1.5.1 Time change

For $d = 1$:

Theorem 1.21. [Dubins-Schwarz]

- Let $M \in \mathcal{M}_{loc}^0$ and $\langle M \rangle_\infty = \infty$ a.s.
- Let $T_t := \inf\{s \geq 0 : \langle M \rangle_s \geq t\}$

This implies

1. $t \mapsto M_{T_t}$ is a (\mathcal{F}_{T_t}) brownian motion
2. $M_t = B_{\langle M \rangle_t}$ for some standard brownian motion B

$$X_t = X_0 + \underbrace{\int_0^t b(s, X_s) ds + \int_0^t \sigma(s, X_s) dB_s}_{=M_t}. \text{ If } \langle M \rangle_\infty = \infty \text{ a.s.:}$$

$$X_t = X_0 + \int_0^t b(s, X_s) ds + \tilde{B}_{\int_0^t \sigma^2(s, X_s) ds}$$

1.5.2 Time change in a martingale problem

Consider $d = 1 = n$.

$$dY_t = \tilde{\sigma}(Y_t) dB_t \quad (\star\star)$$

and $\tilde{\sigma}$ strictly positive continuous function.

$$\langle Y \rangle_t = \int_0^t \tilde{\sigma}(Y_s)^2 ds =: A_t$$

By theorem 1.21 $\implies Y_t = W_{A_t}$ for some brownian motion W .

Assume $A_t \rightarrow \infty$ a.s.

$$T_t := \inf\{s \geq 0 : \langle Y \rangle_s \geq t\}$$

$$\implies T_{A_t} = \inf\{s \geq 0 : \langle Y \rangle_s \geq \langle Y \rangle_t\} = t$$

$$\begin{aligned} 1 &= \frac{d}{dt} (T_{A_t}) = T'_{A_t} \cdot A_t \\ &\implies \underbrace{T'_u}_{= \frac{dT_u}{du}} = \frac{1}{A'_u} \implies T_u = \int_0^u \frac{1}{\tilde{\sigma}(Y_s)^2} ds = \int_0^u \frac{1}{\tilde{\sigma}(W_s)^2} ds \end{aligned}$$

\implies to construct a solution of $(\star\star)$: Given $W \rightarrow$ compute $T_u \rightarrow$ determine $A - t = T_t^{-1} \implies Y_t = W_{A_t}$

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Theorem 1.22. Let $(X_u)_{u \geq 0}$ on $(\Omega, \mathcal{F}, (\mathcal{F}_t)_{t \geq 0}, \mathbb{P})$ be a weak solution of

$$dX_u = b(X_u) du + \sigma(X_u) dB_u$$

where $b : \mathbb{R}^d \rightarrow \mathbb{R}^d$, the drift and $\sigma : \mathbb{R}^d \rightarrow \mathbb{R}^d \times \mathbb{R}^d$ are locally bounded, σ^{-1} exists for a.e. x and is locally bounded.

Consider a **time change** $T_u := \int_0^u \rho(X_s) ds$, where $\rho : \mathbb{R}^d \rightarrow \mathbb{R}_+$ s.t.

$$T_u < \infty \forall u \geq 0 \text{ and } T_\infty = \infty \text{ a.s.}$$

\implies Then $Y_t := X_{A_t}$, where $A_t = T_t^{-1}$ is a weak solution of the SDE

$$dY_t = \frac{b(Y_t)}{\rho(Y_t)}dt + \frac{\sigma(Y_t)}{\sqrt{\rho(Y_t)}}dB_t$$

Remark. Special case: $d = 1, b = 0, \sigma = 1$: Then X is a BM and $\rho = \frac{1}{\tilde{\sigma}^2(x)} \implies Y_t = X_{T_t^{-1}}$ solves

$$dY_t = \tilde{\sigma}(Y_t)dB_t$$

Proof. By theorem 1.20, it is enough to see how the martingale problem after change of time becomes:

$$\mathcal{L} = \mathcal{L}_{X_t} \xrightarrow{\text{time change}} \mathcal{L}_{Y_t} = \tilde{\mathcal{L}}$$

$Y_t = X_{A_t}; Y_0 = X_{A_0}$. For $f \in C^2$: $M_t^f = f(X_t) - f(X_0) - \int_0^t (\mathcal{L}f)(X_s)ds$ is a local martingale w.r.t. $(\mathcal{F}_t)_{t \geq 0}$.

$$\implies N_t^f := M_{A_t}^f = f(\underbrace{X_{A_t}}_{=Y_t}) - f(\underbrace{X_{A_0}}_{=Y_0}) - \int_{A_0}^{A_t} (\mathcal{L}f)(X_s)ds$$

is also a local martingale w.r.t. $(\mathcal{F}_{A_t})_{t \geq 0}$.

Change of variable (to get rid of the X_s in the integral):

$$\begin{aligned} \tau = T_s \leftrightarrow A_t = s &\implies X_s = X_{A_t} = Y_\tau \\ d\tau &= \rho(X_s)ds \end{aligned}$$

$$\implies N_t^f = f(Y_t) - f(Y_0) - \int_0^t (\mathcal{L}f)(Y_t) \frac{1}{\rho(Y_t)} d\tau$$

Since $\mathcal{L}f(x) = \sum_k b_k(x) \frac{\partial}{\partial x_k} f(x) + \sum_{k,l} a_{k,l}(x) \frac{\partial^2}{\partial x_k \partial x_l} f(x)$

$$\implies (\tilde{\mathcal{L}}f)(x) = \sum_k \frac{b_k(x)}{\rho(x)} \frac{\partial}{\partial x_k} f(x) + \sum_{k,l} \frac{\overbrace{a_{k,l}}^{(\sigma\sigma^\top)_{k,l}}(x)}{\sqrt{\rho(x)\rho(x)}} \frac{\partial^2}{\partial x_k \partial x_l} f(x)$$

\implies It is a martingale problem for the SDE where the drift $\rightarrow \frac{\text{drift}}{\rho}$ and $\sigma \rightarrow \frac{\sigma}{\sqrt{\rho}}$

□

1.5.3 Weak solutions in d=1

We will do both time and “space” changes.

- 1-d SDE:

$$\begin{cases} dX_t &= b(X_t)dt + \sigma(X_t)dB_t \\ X_0 = x_0 \in (\alpha, \beta) \end{cases} \quad (1.4)$$

- X_t a process in (α, β)
- Assume $b, \sigma : (\alpha, \beta) \rightarrow \mathbb{R}$ continuous, $\sigma(x) > 0 \forall x \in (\alpha, \beta)$
- Do a change of coordinates $Y_t := s(X_t)$ where $s : (\alpha, \beta) \rightarrow (s(\alpha), s(\beta))$, C^2 with $S'(x) > 0, x \in (\alpha, \beta)$.
- $s(x)$ is called the **scale function** and it is given by

$$s(x) = \int_{x_0}^x \exp\left(-\int_{x_0}^y dz \frac{2b(z)}{\sigma(z)^2}\right) dy$$

- s satisfies $\frac{1}{2}\sigma^2(x)s''(x) + b(x)s'(x) = 0 \iff (\mathcal{L}s)(x) = 0$

The A_0 in the integral is probably 0, but it does not matter, we do a change of variables anyway.

Remark. If $b(z) = 0 \implies s(x) = x - x_0 \implies s'(x) = 1$. If $s'(x) = 1$, we say that the process is in its “natural scale”

By proposition 1.14: $\mathcal{L}s = 0, \dot{s} = 0$.

$\implies Y_t = s(X_t)$ is a local martingale satisfies $dY_t = s'(X_t)\sigma(X_t)dB_t$.

the other terms cancel

$\iff Y_t$ is a solution of

$$\begin{cases} dY_t &= \tilde{\sigma}(Y_t)dB_t \\ Y_0 &= s(X_0) \end{cases} \quad (1.5)$$

where $\tilde{\sigma}(y) = s'(s^{-1}s(y))\sigma(s^{-1}(y))$.

Therefore we can write the original SDE in terms of a BM

Theorem 1.23. The following are equivalent:

1. The process $(X_t)_{t < \xi}$, where ξ is the explosion time, on $(\Omega, \mathcal{F}, (\mathcal{F}_t)_{t \geq 0}, \mathbb{P}; (B_t)_{t \geq 0})$ is a solution of (1.4) up to the stopping time ξ
2. The process $Y_t = s(X_t)_{t < \xi}$ on $(\Omega, \mathcal{F}, (\mathcal{F}_t)_{t \geq 0}, \mathbb{P}; (B_t)_{t \geq 0})$ is a solution of (1.5) up to ξ
3. The process $(Y_t)_{t < \xi}$ has the representation $Y_t = \tilde{B}_{A_t}$, where \tilde{B} is a BM starting at $\tilde{B}_0 = s(X_0)$ and $A_t = T_t^{-1}$ and $T_t = \int_0^t \frac{1}{\tilde{\sigma}^2(\tilde{B}_u)} du$

s and A_t have the same definition as before

A degenerate case:

Let $\sigma(x) = |x|^\alpha$ for some $\alpha \in (0, \frac{1}{2})$. \implies

$$\begin{cases} dY_t &= |Y_t|^\alpha dB_t \\ Y_0 &= y \end{cases} \quad (1.6)$$

$\implies T_t = \int_0^t \frac{1}{\sigma(\tilde{B}_u)^2} du$, $A_t = \int_0^t \sigma(Y_s)^2 ds$ and $Y_t = \tilde{B}_{A_t} \implies \tilde{B}_0 = y$.
 $T_t < \infty$ a.s. ?

$$\begin{aligned} \mathbb{E}(T_t) &= \int_0^t \mathbb{E} \left(\frac{1}{\sigma(\tilde{B})^2} \right) du \\ &= \int_0^t \mathbb{E} \left(\frac{1}{|\tilde{B}|^{2\alpha}} \right) du \\ &= \int_0^t \int_{\mathbb{R}} dx \frac{1}{|x|^{2\alpha}} \frac{\exp(-\frac{(x-y)^2}{2u})}{\sqrt{2\pi u}} \quad 0 < 2\alpha < 1 < \infty \end{aligned}$$

$\implies A_t = T_t^{-1}$, then $Y_t = \tilde{B}_{A_t}$ is a solution of (1.6), i.e. $\forall y \in \mathbb{R} \exists$ a non-trivial solution of (1.6).

For $Y_0 = 0, Y_t = 0$ is also a solution \implies

- No uniqueness in law of (1.6)
- No pathwise uniqueness as well

Remark. In general: uniqueness in law of 1-d SDEs is not to be expected if $\sigma(x) = 0$ somewhere (and σ continuous ...) (i.e. if σ is degenerate).

By theorem 1.12 as soon as $\sigma(x) = |x|^\alpha$ for some $\alpha \geq \frac{1}{2}$, then one has pathwise uniqueness.

Hitting times and scale functions Bessel process:

$$\begin{cases} dR_t &= \frac{d-1}{2R_t} dt + dW_t \\ R_0 &= r_0 \in (0, \infty) \end{cases}$$

$\implies b(x) = \frac{d-1}{2x}, \sigma(x) = 1$.

The scale function satisfies $\mathcal{L}s(x) = \frac{1}{2}s''(x) + \frac{d-1}{2x}s'(x) = 0$

$$\implies s(x) = \begin{cases} \frac{x^{2-d}}{2-d} & d \neq 2 \\ \ln(x) & d = 2 \end{cases}$$

and therefore

$$(s_0, s_\infty) = \begin{cases} (0, \infty) & d < 2 \\ (-\infty, \infty) & d = 2 \\ (-\infty, 0) & d > 2 \end{cases}$$

Define the stopping time

$$T_a^R := \inf\{t \geq 0 \mid R_t = a\}$$

Choose an $\alpha < r_0 < \beta$

$$\implies \mathbb{P}(T_\alpha^R < T_\beta^R) \stackrel{s' \geq 0}{=} \mathbb{P}(T_{s(\alpha)}^{s(R)} < T_{s(\beta)}^{s(R)})$$

One computes

WS exercises

$$\mathbb{P}(T_a^R < T_\beta^R) = \frac{s(\beta) - s(r_0)}{s(\beta) - s(\alpha)} = \mathbb{P}(T_{s(\alpha)}^{s(R)} < T_{s(\beta)}^{s(R)})$$

This is generic, for Itô diffusions, provides that \exists no killing in (α, β) .

unlike in 1.16

List of Lectures

- **Lecture 01:** Introduction, reminder of strong solutions, definition of weak solutions, uniqueness in law, pathwise uniqueness, and some examples
- **Lecture 02:** Further examples, Yamata-Watanabe theorems and Skorohod theorem (no proof), reminder of Lévy characterization, Ito-Doeblin formula
- **Lecture 03:** The martingale problem and one-to-one relation with weak solutions (special case of $d = n$ proven); reminder of Dubins-Schwarz theorem
- **Lecture 04:** PLACEHOLDER