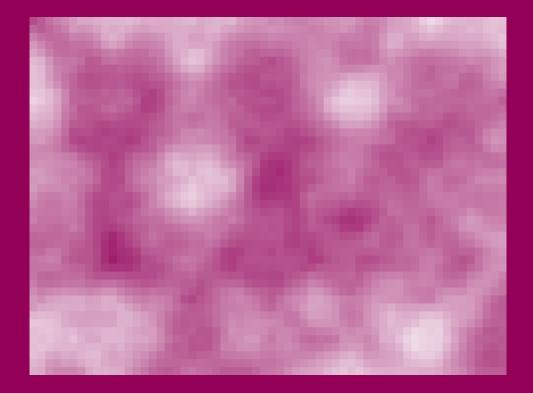
Lecture notes on Stochastic Analysis

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University of Bonn Summer semester 2024 Last update: May 2, 2024

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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

 \bullet 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:

$$\begin{cases} dX_t = b(t, X_t)dt + \sigma(t, X_t)dB_t \\ X_0 = x_0 \text{ Initial condition} \end{cases}$$
 (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
- $a(t,x) = \sigma(t,x) \cdot \sigma(t,x)^{\mathsf{T}}$: diffusion matrix

1.1 Strong solutions

Definition 1.1. Let $(\Omega, \mathcal{F}, (\mathcal{F})_{t \geq 0}, \mathbb{P})$ 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:
 - $\cdot X_t = x_0 \ a.s.$
 - · X_t is adapted to $\mathcal{F}_t \forall t \geq 0$
 - · X is a continuous semimartingale s.t. $\forall t \geq 0$:

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

 $X_t = X_0 + \int_0^t b(s, X_s) ds + \int_0^t \sigma(s, X_s) dB_s$

In the last semester we proved:

Theorem 1.2. Assume that b, σ are globally lipschitz with at most linear growth at ∞ (in space) $\Rightarrow \exists !$ 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:

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

Linear growth condition:

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

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

1.2 Weak solutions

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

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

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

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

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 <u>uniqueness in law</u> 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:

the sigma algebra

They agree on any set in

$$Law_{\mathbb{P}}(X) = Law_{\tilde{\mathbb{D}}}(\tilde{X})$$

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 & and uniqueness in law by Tanaka).

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

or more generally $X_0 = Y$, where

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

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

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

$$\implies dW_t = sgn(W_t)dB_t$$

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

 B_t is a local martingale with

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

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.

 \Longrightarrow

- 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).

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

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

$$X_{t}^{2} = 2 \int_{0}^{t} X_{s} dX_{s} + \int_{0}^{t} 1 ds$$

$$= -\int_{0}^{t} 1_{X_{s} \neq 0} ds + 2 \int_{0}^{t} x_{s} dB_{s} + t$$

$$= \int_{0}^{t} 1_{X_{s} = 0} ds + 2 \int_{0}^{t} X_{s} dB_{s}$$

We will prove $\int_0^t 1_{X_s=0} ds = 0 \implies X_t^2$ is a local martingale, $X_t^2 \ge 0$ (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 $L = \sum b \partial_i + \frac{1}{2} \sum a^{ij} \partial_i \partial_j$

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

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 & $\mathcal{N}(0, t_i - t_{i-1})$ 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 $\iff ds2$?

Itô-Integral:

- 1. If X simple process \implies RS-Integral
- 2. Itô isometry $\mathcal{E} \to \{L^2 \text{local martingale}\}, \mathcal{E} \subset L^2(d[M])$ dense
- 3. general $X: \int XdM$ 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 $\int Xd(\int YdZ) = \int XYdZ$
- If M local martingale $\implies \int XdM$ 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 $(|b(t,x)| + |\sigma(t,x)| \le C(1+|x|)), e = \infty.$

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.:

 $\overline{\text{Idea:}}$ Use Itô formula: $X_t = \sin(B_t) = f(B_t)$

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

<u>2.:</u>

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

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

Probably σ^2 Lipschitz $\implies \sigma$ Hölder $\frac{1}{2}$ 3.

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

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:

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

 $\Longrightarrow B_{t} := X_{T_{t}} \text{ 1 d BM w.r.t. } (F_{T_{t}})_{t \geq 0}, X_{t} = B_{\langle B \rangle_{t}}$
here: use $X_{t} = b(X_{t})dt + \sigma(X_{t})dB_{t} \Longrightarrow d[X]_{t} = \sigma^{2}(X_{t})dt$

Problem 00.3: SDE and PDE

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

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

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

<u>1.:</u>

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

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

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

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

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

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).

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

Then

$$X_t = 0 \forall t > 0$$

and

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

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

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

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

$$X_{t} = \int_{0}^{t} 1_{X_{s} \neq 0} dB_{s}$$

= $\int_{0}^{t} (1 - 1_{X_{s} = 0} dB_{s})$
 $B_{t} - \int_{0}^{t} 1_{X_{s} = 0} dB_{s}$

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

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

Let Y_t be a solution of

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

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

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. $O_t^{\intercal}O_t = 1 \forall t \geq 0$ i.e. O_t is a rotation

$$\implies X_t^k = X_0^k + \sum_{l=1}^d \int_0^t O_s^{k,l} dB_s^l$$

$$\implies \langle X^k, X^{\tilde{k}} \rangle_t = \sum_{l,\tilde{l}}^d \int_0^t O_s^{k,l} O_s^{\tilde{k},\tilde{l}} d\langle B^l, B^{\tilde{l}} \rangle_s = \sum_{l=1}^d \int_0^t O_s^{k,l} O_s^{\tilde{k},l} ds$$

$$= \int_0^t \underbrace{(O_s O_s^\intercal)^{k,\tilde{k}}}_{=1} ds = \delta_{k,\tilde{k}} t \stackrel{Lévy}{\Longrightarrow} (X_t)_{t \geq 0} \text{ is also a } d\text{-dim } BM \text{ starting from } x_0$$

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.

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

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

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

and

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

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 weka 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

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

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

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

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

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

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 $M_t \in \mathcal{M}$.

$$T \coloneqq \inf\{s \ge 0 | X_s \notin U\} \implies M_t^T \coloneqq M_{T \land t} \in \mathcal{M}.$$

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

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

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 $f : \mathbb{R}^d \to \mathbb{R}$ and $K : [0,t] \times \mathbb{R}^d \to \mathbb{R}_+$ 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 $s \in [0,t]$

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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 $U_k := \{x \in U \mid |x| < k \land \operatorname{dist}(x, U^c) \ge \frac{1}{k}\}$ with $U = \bigcup_{k \ge 1} U_k$ and

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

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

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) :

$$dX_t b(t, X_t) dt + \sigma(t, X_t) dB_t$$
 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) $\forall f \in C^{1,2}(\mathbb{R}_+ \times \mathbb{R}^d)$

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

is a local martingale

ds

Proof. $\underline{c} \Longrightarrow \underline{a}$: by choosing f independent of t. $\underline{a} \Longrightarrow \underline{b}$: 1.: Choosing $f(X) = X_i$ implies

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

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

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

$$\begin{split} (\mathcal{L}f)(x) &= \frac{1}{2}a_{ij} + \frac{1}{2}a_{ji} + b_iX_jb_jX_i \\ a &= a^{\mathsf{T}} \implies = a_{ij}b_iX_jb_jX_i \\ \implies M_t^fX_t^iX_t^j - X_0^iX_0^j - \int_0^t \left[a_{ij}(s,X_s) + b_i(s,X_s)X_s^j + b_j(s,X_s)X_s^i\right]ds \end{split}$$

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

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

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

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.:

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

- (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)

Proof. $a \implies b$: True

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

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

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

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

To see: B_t is a brownian motion.

$$\langle \tilde{B}^{i}, \tilde{B}^{j} \rangle_{t} = \sum_{k,l} \int_{0}^{t} \sigma_{ij}^{-1} \sigma_{jl}^{-1} \underbrace{d\langle M^{k}, M^{l} \rangle_{s}}_{= \underbrace{\alpha_{ij}}_{(\sigma^{\mathsf{T}}\sigma)_{kl}} ds}$$

$$= \sum_{k,l,p} \int_{0}^{t} \sigma_{ik}^{-\mathsf{T}} \sigma_{kp} \sigma_{pl}^{\mathsf{T}} \sigma_{lj}^{-\mathsf{T}} ds$$

$$= \delta_{ij} \int_{0}^{t} 1 ds = \delta_{ij} t$$

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

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

This implies

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

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

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

and $\tilde{\sigma}$ strictly positive 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 $A_t \infty = \infty$ a.s.

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

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

$$1 = \frac{d}{dt} (T_{A_t}) = T'_{\underbrace{A_t}} \cdot A_t$$

$$\Longrightarrow T'_u = \frac{1}{A'_{T_u}} \Longrightarrow T_u = \int_0^u \frac{1}{\tilde{\sigma}(Y_{T_s})^2} ds = \int_0^u \frac{1}{\tilde{\sigma}(W_s)^2} ds$$

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

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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 $b: \mathbb{R}^d \to \mathbb{R}^d$, the drift and $\sigma: \mathbb{R}^d \to \mathbb{R}^d \times \mathbb{R}^d$ 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 $\rho : \mathbb{R}^d \to \mathbb{R}_+$ s.t.

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

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 $\rho=\frac{1}{\tilde{\sigma}^2(x)} \implies Y_t=X_{T^{-1}}$ 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:

$$\mathcal{L} = \mathcal{L}_{X_t} \stackrel{ ext{time change}}{\longrightarrow} \mathcal{L}_{Y_t} = \tilde{\mathcal{L}}$$

 $Y_t = X_{A_t}; Y_0 = X_{A_0}$. 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. $(\mathcal{F}_{A_t})_{t>0}$.

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

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

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

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

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

 \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:

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

- X_t a process in (α, β)
- Assume $b, \sigma : (\alpha, \beta) \to \mathbb{R}$ continuous, $\sigma(x) > 0 \forall x \in (\alpha, \beta)$
- Do a change of coordinates $Y_t := s(X_t)$ where $s: (\alpha, \beta) \to (s(\alpha), s(\beta)), C^2$ 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.

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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 $dY_t = s'(X_t)\sigma(X_t)dB_t$.

 $\iff Y_t \text{ is a solution of}$

the other terms cancel

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

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

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 tje stopping time ξ
- 2. The process $Y_t = s(X_t)_{t < \xi}$ on $(\Omega, \mathcal{F}, (\mathcal{F}_t)_{t > 0}, \mathbb{P}; (B_t)_{t > 0})$ 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 $T_t = \int_0^t \frac{1}{\tilde{\sigma}^2(\tilde{B}_u)} du$

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

s and A_t have the same definition as before

A degenerate case:

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

$$\begin{cases} dY_t = |Y_t|^{\alpha} dB_t \\ Y_0 = y \end{cases} \tag{1.6}$$

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

$$\mathbb{E}(T_t) = \int_0^t \mathbb{E}\left(\frac{1}{\sigma(\tilde{B})^2}\right) du$$

$$= \int_0^t \mathbb{E}\left(\frac{1}{|\tilde{B}|^{2\alpha}}\right) du$$

$$= \int_0^t \int_{\mathbb{R}} dx \frac{1}{|x|^{2\alpha}} \frac{\exp(-\frac{(x-y)^2}{2u})}{\sqrt{2\pi u}} \stackrel{0<2\alpha<1}{<} \infty$$

 $\implies A_t = T_t^{-1}$, 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 $Y_=0, Y_t = 0$ 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 $\alpha \ge \frac{1}{2}$, 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

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

Choose an $\alpha < r_0 < \beta$

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

One computes

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

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

unlike in 1.16

WS exercises

Start of lecture 05 (25.04.24)

1.5.4 Uniqueness of martingale solution

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

$$\mathcal{L} = \sum_{k} b_{k} \frac{\partial}{\partial x_{k}} + \frac{1}{2} \sum_{k,l} (\sigma \sigma^{\mathsf{T}})_{k,l} \frac{\partial^{2}}{\partial x_{k} \partial x_{l}}$$

Definition 1.24. Let $C = C(\mathbb{R}_+, \mathbb{R}^d)$ with σ -algebra \mathcal{F} , canonical filtration $(\mathcal{F}_t)_{t \geq 0}$, canonical process $Z_t(\omega) := \omega$.

We say that \mathbb{P} on $(\mathcal{C}, \mathcal{F})$ is a <u>martingale solution</u> for the generator $\mathcal{L} \iff \forall f \in C^{1,2}(\mathbb{R}_+ \times \mathbb{R}^d \to \mathbb{R})$

$$M_t^f := f(t, Z_t) - f(0, Z_0) - \int_0^t \left(\frac{\partial}{\partial s} + \mathcal{L}\right) f(s, Z_s) ds \tag{1.7}$$

is a martingale w.r.t. \mathbb{P} .

Definition 1.25. A martingale problem (1.7) has a unique solution if for any two martingale solutions $\mathbb{P} = \mathbb{Q}$ s.t. $Law_{\mathbb{P}}(Z_0) = Law_{\mathbb{Q}}(Z_0)$

$$\implies \mathbb{P} = \mathbb{Q}$$

Remark. Uniqueness of martingale solutions corresponds to uniqueness in law of the weak solutions.

Backwards Kolmogorov Equation (BKE):

$$\frac{\partial}{\partial t}\varphi(t,x) = \mathcal{L}\varphi(t,x) \forall x \in \mathbb{R}^d, t \ge 0$$
(1.8)

Theorem 1.26. Assume that \forall initial condition

$$\varphi(0,x) = \Psi(x), \Psi \in C_0^{\infty}(\mathbb{R}^d)$$

the (1.8) has a solution and φ bounded for all finite time intervals. We have uniqueness of martingale solutions and therefore uniqueness of weak solutions!

The Kolmogorov forward equation is (related to) the Folker-Plank equation!

Proof. Prove that $\forall 0 \leq t_1 < t_2 < \cdots < t_n$:

$$\text{Law}_{\mathbb{P}}(Z_{t_1}, Z_{t_2}, \dots, Z_{t_n}) = \text{Law}_{\mathbb{Q}}(Z_{t_1}, Z_{t_2}, \dots, Z_{t_n})$$

1. One-time distribution:

 $\forall 0 \leq s \leq r$:

$$\left(\frac{\partial}{\partial s} + \mathcal{L}\right)\varphi(r - s, x) \stackrel{(1.8)}{=} 0$$

take $t \in [0, r]$:

$$M_t^r := \varphi(r - t, Z_t) - \varphi(r, Z_0) - \int_0^t \underbrace{(\partial_s + \mathcal{L})\varphi(r - s, Z_s)}_{=0} ds$$
$$= \varphi(r - t, Z_t) - \varphi(r, Z_0) \text{ is a martingale}$$

for any solution \mathbb{P} .

$$0 = \mathbb{E}_{\mathbb{P}} \left(M_r^r - M_t^r \mid \mathcal{F}_t \right) = \mathbb{E}(\varphi(0, Z_r) - \varphi(r - t, Z_t) \mid \mathcal{F}_t)$$

$$\Rightarrow \forall 0 \leq t \leq r : \mathbb{E}_{\mathbb{P}}(\varphi(0, Z_r) \mid \mathcal{F}_t) = \mathbb{E}_{\mathbb{P}}(\varphi(r - t, Z_t) \mid \mathcal{F}_t)$$

$$\stackrel{\text{a.s.}}{=} \varphi(r - t, Z_t)$$

$$\mathbb{E}(\underline{\varphi(0, Z_r)}) \stackrel{t=0}{=} \mathbb{E}_b P(\varphi(r, Z_0))$$

 \forall other martingale solutions \mathbb{Q} :

$$\mathbb{E}_{\mathbb{Q}}(\Psi(Z_r)) = \mathbb{E}_{\mathbb{Q}}(\varphi(r, Z_0))$$

By assumption this implies $\text{Law}_{\mathbb{P}}(Z_r) = \text{Law}_{\mathbb{Q}}(Z_r)$.

2. Multi-time distributions:

For $\Psi \in C_0^{\infty}$, denote φ_{Ψ} the solution of (1.8) with initial condition Ψ :

$$\mathbb{E}_{\mathbb{P}}(\Psi(Z_r) \mid \mathcal{F}_t) = \varphi_{\Psi}(r - t, Z_t)$$

 $0 \le r_2 \le r_1$ test for $g \in C_0^{\infty}$:

$$\mathbb{E}_{\mathbb{P}}(\Psi(Z_{r_1})g(X_{r_2})) = \mathbb{E}(\underbrace{\mathbb{E}(\varphi_{\Psi}(Z_{r_1})|\mathcal{F}_{r_2})}_{\varphi_{\Psi}(r_1 - r_2, Z_{r_2})} g(Z_{r_2}))$$

$$= \mathbb{E}_{\mathbb{P}}(\varphi_{\Psi}(r_1 - r_2, Z_{r_2})g(Z_{r_2}))$$

$$\stackrel{!}{=} \mathbb{E}_{\mathbb{Q}}(\varphi_{\Psi}(r_1 - r_2, Z_{r_2})g(Z_{r_2}))$$

$$= \mathbb{E}_{\mathbb{Q}}(\Psi(Z_{r_1})g(Z_{r_2}))$$

Iterating yields the statement.

This needs the boundedness of φ , otherwise it might only be a local martingale. There are softer contidions we can put on the coefficients to achieve the same result. This might not be needed, because φ is C^1 in time anyways

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Chapter 2: SDE techniques

Goal: Study process by changing the measure.

E.g.:
$$X_t = B_t$$
. Condition $X_t \ge 0 \forall t \ge 0$.

2.1 Girsanov theorem

Given $(\Omega, \mathcal{F}, (\mathcal{F}_t)_{t \geq 0})$ and two measures \mathbb{P}, \mathbb{Q} Assume $\mathbb{P} \ll \mathbb{Q}$ and $\mathbb{Q} \ll \mathbb{P}$ and let $H = \frac{d\mathbb{Q}}{d\mathbb{P}}$

$$\implies Z_t := \mathbb{E}_{\mathbb{P}}(H \mid \mathcal{F}_t) = \frac{d\mathbb{Q}}{d\mathbb{P}} \mid_{\mathcal{F}_t}$$

The notes of Eberle switches the roles of $\mathbb{P}, \mathbb{Q}!$

is a martingale.

From last semester: $\forall Y \in L^1(\mathbb{Q}) \cap L^1(\mathbb{P}), \mathcal{F}_t$ measurable:

$$\mathbb{E}_{\mathbb{Q}}(Y \mid \mathcal{F}_s) = \frac{bE_{\mathbb{P}}(Y \cdot Z_t \middle| \mathcal{F}_s)}{Z_s} \forall s < t$$

If Z > 0 is $\mathcal{M}_{loc}, \exists L \in \mathcal{M}_{loc}$ s.t.:

$$Z_t = e^{L_t - \frac{1}{2}\langle L \rangle_t} \to L_t = \ln(Z_0) + \int_0^t \frac{dZ_s}{Z_s}$$

There might be a problem, because $ln(Z_0)$ might not be integrable, and therefore not a local martingale!

Theorem 2.1 (Girsanov). Assume Z > 0 is a martingale. If M is a local martingale w.r.t. \mathbb{P} , then

$$\tilde{M}_t := M_t - \langle M, L \rangle_t$$

is a local martingale w.r.t. \mathbb{Q} and

$$\langle M \rangle_t = \langle \tilde{M} \rangle_t.$$

Moreover, if M is a BM w.r.t. to \mathbb{P} , then \tilde{M} is a BM w.r.t. \mathbb{Q} .

Remark. In applications, given \mathbb{P} , $(Z_t)_{t\geq 0}$ a positive continuous martingale, define \mathbb{Q} on $\mathcal{F}_{\infty} = \bigcup_{t\geq 0} \mathcal{F}_t$ s.t.

$$\frac{d\mathbb{Q}}{d\mathbb{P}}\Big|_{\mathcal{F}_t} = Z_t$$

Z is uniformly integrable $\iff \mathbb{Q} \ll \mathbb{P}$. **Problem:** In applications, Z is not necessarily uniformly integrable.

 \implies restrict to [0,T] \implies all fine.

 $\mathbb{Q} \to \mathbb{Q}_T$ as in the last semester.

Added example. Let $\gamma \in \mathbb{R}^d$, $(B_t)_{t\geq 0}$ standard BM. Let $L_t := \gamma \cdot B_t$ and $Z_t = \exp(L_t - \frac{1}{2}\langle L \rangle_t) = \exp(\gamma \cdot B_t - \frac{1}{2}|\gamma|^2 t)$

Remark. $\lim_{M\to\infty}\sup_{t\geq 0}\mathbb{E}(|Z_t|1_{|Z_t|>M})\neq 0.$ Define \mathbb{Q} on \mathcal{F}_{∞} s.t. $\tilde{B}_t^k=B_t^k-\langle L,B^k\rangle=B_t^k-\gamma^k t$ is a BM with drift. Show: $Q\not\ll \mathbb{P}$.

 $Construct \ Q \ via \ Z$

$$A = \left\{ \lim_{t \to \infty} \frac{\tilde{B}_t + \gamma t}{t} = \gamma \right\} \in \mathcal{F}_{\infty}$$

but: B is a BM w.r.t. $\mathbb{Q} \implies \mathbb{Q}(A) = 1$, B is a BM with drift $-\gamma$ w.r.t $\mathbb{P}(A) = 0$.

Start of lecture 06 (30.04.24)

2.1.1 **Drift transformation of SDE**

We once again start with

$$\star = \begin{cases} dX_t &= b(t, X_t) \\ X_0 &= x_0 \end{cases}$$

with drift b continuous.

Goal: Get a weak solution of \star .

Let $(X_t)_{t\geq 0}$ be a BM in $(\Omega, \mathcal{F}(\mathcal{F}_t)_{t\geq 0}, \mathbb{P}), X_0 = x_0$.

Assume: $Z_t := \exp\left(\underbrace{\int_0^t b(s, X_s) dX_s}_{\bullet} - \frac{1}{2} \int_0^t |b(s, X_s)|^2 ds\right)$ is a martingale w.r.t. \mathbb{P} .

Remark. The assumption holds if

$$|b(t,x)| \le C(1+||x||)$$
 for some C.

$$\implies \mathbb{E}_{\mathbb{P}}(Z_t) = 1 \forall t \ge 0$$

By Girsanov

$$\tilde{X}_t = X_t - \langle L, X \rangle_t$$

is a \mathbb{Q} -BM.

But

$$\begin{split} d\langle L, X\rangle_t &= b \cdot dX_t \cdot dX_t = b \cdot dt \\ \Longrightarrow & \tilde{X}_t = X_t - \int_0^t b(s, X_s) ds \\ \text{w.r.t. } \mathbb{Q} : dX_t = b(t, X_t) dt + \underbrace{d\tilde{B}_t}_{d\tilde{X}_t} \end{split}$$

Generalization: Start with

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

where X, B are d-dimensional.

Proposition 2.2. Assume (X, B) is a weak solution of $\star\star$ on $(\Omega, \mathcal{F}, (\mathcal{F}_t)_{t>0}, \mathbb{P})$. If

$$Z_t = \exp\left(\int_0^t c(s, X_s) dB_s - \frac{1}{2} \int_0^t c(s, X_s)^2 ds\right)$$

is a martingale w.r.t. \mathbb{P} , $\mathbb{Q} \ll \mathbb{P}$ and \mathcal{F}_t s.t. $Z_t = \frac{d\mathbb{Q}_t}{d\mathbb{P}} \operatorname{Vert}_{\mathcal{F}_t}$.

In practice this is surprisingly useful in practice!

this is an implicit condition for c

Then $(X_t)_{t\geq 0}$ on $(\Omega, \mathcal{F}, (\mathcal{F}_t)_{t\geq 0}, \mathbb{Q})$ is a weak solution of

$$dX_t = [b(t, X_t) + \sigma(t, X_t)c(t, X_t)dt] + \sigma(t, X_t)d\tilde{B}_t$$

where \tilde{B} is a d-dim BM.

Proof.

$$\begin{split} d\langle L,B\rangle_t &\overset{L_t = \int_0^t c(s,X_s)dB_s}{=} c(t,X_t)dt \\ &\Longrightarrow \tilde{B}_t \coloneqq B_t - \langle L,B\rangle_t \text{ is a } \mathbb{Q}\text{-BM} \\ &\Longrightarrow dB_t = c(t,X_t)dt + d\tilde{B}_t \end{split}$$

From $\star\star$: $dX_t = b(t, X_t)dt + \sigma(t, X_t)c(t, X_t)dt + \sigma(t, X_t)d\tilde{B}_t = \star\star\star$

Measure	SDE	Generators
\mathbb{P}	$dX_tb \cdot dt + \sigma \cdot dB_t$	$\mathcal{L}\sum_k b_k rac{\partial}{\partial x_k} + rac{1}{2}\sum_{k,l} (\sigma\sigma^\intercal)_{k,l} rac{\partial^2}{\partial x_k \partial x_l}$
Q	$[b + \sigma \cdot c]dt + \sigma d\tilde{B}_t$	$\tilde{\mathcal{L}} = \mathcal{L} + \sum_{k} \sum_{l} \sigma_{k,l} c_{l} \frac{\partial}{\partial x_{k}} = \mathcal{L} + c \sigma^{T} \nabla$

 $t \leq T$ for \mathbb{Q}_t .

2.2 Doob-h transform

- 1. From $B_t \to B_t$ conditioned on $\{B_1 = 0\}$ (measure zero set)
- 2. From $B_t \to B_t$ conditioned on $\{B_t \ge 0 \forall t \ge 0\}$ (measure zero set)
- 3. From $B_t \to B_t$ conditioned on $\{B t \ge 0 \forall t \in [0, 1], B_1 = 0\}$

Let $(X_t, B_t)_{t>0}$ be a weak solution of

$$(\star) = \begin{cases} dX_t &= b(t, X_t)dT + \sigma(t, X_t)dB_t \\ X_0 &= x_0 \text{ fixed} \end{cases}$$

Assume there exists $h \in C^{1,2}(\mathbb{R}_+ \times \mathbb{R}^d)$ s.t. h > 0 satisfying

$$(\frac{\partial}{\partial t} + \mathcal{L})h = 0 \forall t, \in [0, T], x \in \mathbb{R}^d$$

where \mathcal{L} is the generator of the SDE (\star) .

By theorem 1.14 $Z_t := h(t, X_t) = h(0, X_0) + \int_0^t (\sigma \nabla h)(s, X_s) dB_s$. Z_t is a positive local martingale. Assume Z_t is a martingale.

W.l.o.g. $Z_0 = 1$ (if not $h \mapsto \frac{h}{h(0,X_0)}$).

$$\implies d\mathbb{Q}_T = Z_T \cdot d\mathbb{P}$$
. Let L_t s.t. $Z_t = \exp(L_t - \frac{1}{2}\langle L \rangle_t)$.

Girsanov, since B_t is a \mathbb{P} -BM,

 $\tilde{B}_t := B_t - \langle L, B \rangle_t$ is a Q_T -BM, where

$$\langle L, B \rangle_t = \sigma^{\mathsf{T}} \cdot \nabla \ln h \cdot dB_t$$

because of

$$dL_t = \frac{dZ_t}{Z_t} = \frac{\sigma^{\mathsf{T}} \cdot \nabla h \cdot dB_t}{h(t, X_t)} = \sigma^{\mathsf{T}} \cdot \nabla \ln h \cdot dB_t$$

And therefore $dB_t = d\tilde{B}_t + \sigma^{\intercal} \nabla \ln h dt$.

From
$$(\star) \implies d\tilde{X}_t = (\underbrace{b(t, X_t) + \sigma(t, X_t)\sigma^{\mathsf{T}}(t, X_t)\nabla h(t, X_)}_{:=\tilde{b}})dt + \sigma(t, X - t)d\tilde{B}_t$$

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Essentially three cases in

1d

Proposition 2.3. Let X be a weak solution of $dX_t = b(t, X_t) + \sigma(t, X_t) dB_t$ on [0, T] under \mathbb{P} . Then under \mathbb{Q}_T , X is a weak solution of the SDE

$$\begin{cases} dX_t &= \tilde{b}(t, X_t)dt + \sigma(t, X_t)d\tilde{B}_t \\ X_0 &= x_0 \end{cases}$$

where $\tilde{b} = b + \sigma \sigma^{\dagger} \nabla \ln h$.

Remark. In applications, often $h(x) \not> 0$ for all x! $E.g.: h(t,x) = x \implies \left(\frac{\partial}{\partial t} + \mathcal{L}\right) h = 0 \ (X_t = B_t + x_0), \ but \ h(x) \le 0 \ for \ x \le 0.$ In this case first do the construction on $[0,\tau]$, where $\tau := \inf\{t > 0 \mid B_t = 0\}$ $\implies ok \ for \ B_t^{\tau} \stackrel{magically}{\Longrightarrow} the \ new \ SDE \ has \ drift:$

$$\frac{\partial}{\partial x}\ln(h(x)) = \frac{1}{h(x)}\frac{\partial h(x)}{\partial x} = \frac{1}{x}$$

Example 2.4. If $X_t = B_t$ and $h(t, t) = e^{\gamma x - \frac{1}{2}|\gamma|^2 t}$

$$\implies dX_t = \nabla \ln h(t, X_t) dt + dB_t$$
$$= \nabla (\gamma \cdot x) dt + dB_t$$
$$= \gamma d_t + dB_t$$

 $\implies X_t$ is a BM with drift γ w.r.t \mathbb{Q} .

2.3 Diffusion Bridges

Consider a Markov process $(X_t)_{t\geq 0}$ which is a diffusion starting from $X_0 = x_0 \in \mathbb{R}^d$. We want to condition on the event $\{X_T = y\}$ for some given $T > 0, y \in \mathbb{R}^d$. Goal: Find, $\forall y \in \mathbb{R}^d, \mathbb{Q}^y$ which is the conditional measure on $\{X_T = y\}$. Assume: $(X_t)_{t\geq 0}$ has transition density P s.t.

$$\forall 0 \le s \le t \le T, \mathbb{P}(X_t \in dz \mid X_s = x) = p(s, x; t, z)dz.$$

 X_t solves

$$\begin{cases} dX_t = b(t, X_t)dt + \sigma(t, X_t)dB_t \\ X_0 = x_0 \end{cases}.$$

Define $h^y(s,x) := \frac{p(s,x;T,y)}{p(0,x_0;T,y)}$ for $s \in [0,T), x \in \mathbb{R}^d$.

Remark. s < T since otherwise we get ingeneral singularities.

Lemma 2.5. Let $Z_t^y := h^y(t, X_t)$ is a martingale.

Proof. Let $0 \le < t < T$:

$$\mathbb{E}(Z_t^y \mid \mathcal{F}_s) = \mathbb{E}(h^y(t, X_t) \mid \mathcal{F}_s)$$

$$\stackrel{\text{MP}}{=} \mathbb{E}(h^y(t, X_t \mid X_s))$$

$$= \int_{\mathbb{R}^d} h^y(t, x) p(s, X_s; t, x) dx$$

$$= \int_{\mathbb{R}^d} \frac{p(t, x; T, y)}{p(0, x_0; T, y)} p(s, X_s; t, x) dx$$

$$= \frac{1}{p(0, x_0; T, y) \int_{\mathbb{R}^d} p(s, X_s; t, x) p(t, x; T, y) dx}$$

$$\stackrel{\text{Chapman-Kolmogorov}}{=} Z_s^y =$$

We realize there is no problem in the new measure

This is not true in general, if we don't have unique solutions

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: Transformation of SDE under time change, weak solutions for 1d SDEs, scale function and its relation to hitting times
- Lecture 05: Uniqueness of the solution of martingale problem, reminder of Girsanov theorem, changes of SDE under drift transformation
- Lecture 06: Drift transformation for SDE, Doob-h transform, start set-up for diffusion bridges