

Lecture Notes

# **Introduction to Stochastic Analysis**

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corrected until and including p. 130

This script has been written by Matthias Stephan on behalf of Prof. M. Röckner and has been revised by Dr. Judith Dohmann. Please note that despite several checks it may still contain mistakes and, therefore, there is no guarantee of correctness. A report of any kind of errors found in these notes is appreciated ([mstephan@math.uni-bielefeld.de](mailto:mstephan@math.uni-bielefeld.de)).

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# 1. Introduction to Pathwise Itô-Calculus

## 1.1. Preparation

**Definition 1.1.1.**  $X : [0, \infty[ \rightarrow \mathbb{R}$  has bounded variation, if for all  $t \geq 0$

$$\text{var}_t X := \sup_{\tau} \sum_{t_i \in \tau} |X(t_{i+1} \wedge t) - X(t_i \wedge t)| < \infty, \quad (1.1.1)$$

where  $\tau$  is a partition  $0 = t_0 < t_1 < \dots < t_N < \infty$ .

**Notation:** Since in Stochastics a process  $X$  also depends on  $\omega$ , we write  $X_t$  and  $X_t(\omega)$  instead of  $X(t)$  and  $X(t)(\omega)$  respectively.

As a reference see [dB03].

**Definition 1.1.2.** Suppose  $X$  to be right-continuous and of bounded variation. Let  $f \in \mathcal{C}(\mathbb{R})$  and  $(\tau_n)_{n \in \mathbb{N}}$  be a sequence of partitions, whose mesh

$$|\tau_n| := \sup_{1 \leq i \leq N_n} (t_{i+1}^{(n)} - t_i^{(n)})$$

converges to zero as  $n \rightarrow \infty$  and  $t_{N_n} \xrightarrow{n \rightarrow \infty} \infty$ .

Then, since  $X$  is of bounded variation, there exists the Lebesgue-Stieltjes-Integral defined by

$$\int_0^t f_s dX_s := \lim_{n \rightarrow \infty} \sum_{t_i^{(n)} \in \tau_n} f_{t_i^{(n)}} \cdot (X_{t_{i+1}^{(n)} \wedge t} - X_{t_i^{(n)} \wedge t}). \quad (1.1.2)$$

**Remark 1.1.3.** Note that for continuous  $X$  this definition is independent of the choice of  $(\tau_n)_{n \in \mathbb{N}}$  (Exercise!).

### 1.1.1. Quadratic Variation of Brownian Motion

We know (cf. [Röc11]) that a typical path of Brownian motion  $X$  on  $\mathbb{R}^1$  is of unbounded variation, since for its quadratic variation we have  $\langle X \rangle_t = t$  (see 1.1.4(i) below). Nonetheless, one wants to define

$$\int_0^t f_s dX_s$$

for a typical path of the Brownian motion  $X$ . More generally, we want to do this for every continuous path with continuous quadratic variation  $t \mapsto \langle X \rangle_t$ .

$(X_t)_{t \geq 0}$  is a (continuous)  $\mathbb{R}$ -valued Brownian motion on  $(\Omega, \mathcal{F}, P)$  if

- (i) The increments  $X_t - X_s$  are independent and  $N(0, t - s)$  distributed ( $t > s$ ).
- (ii)  $t \mapsto X_t(\omega)$  is continuous for all  $\omega \in \Omega$ .

**Theorem 1.1.4.** Let  $(X_t)_{t \geq 0}$  be a (continuous) Brownian motion and  $(\tau_n)_{n \in \mathbb{N}}$  a sequence of subdivisions with  $|\tau_n| \xrightarrow{n \rightarrow \infty} 0$ ,  $\tau_n \subset \tau_{n+1}$ ,  $t_{N_n}^{(n)} \xrightarrow{n \rightarrow \infty} \infty$ . Then, for all  $t \geq 0$

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(i)

$$\sum_{\substack{t_i^{(n)} \in \tau_n \\ t_i^{(n)} \leq t}} \left( X_{t_{i+1}^{(n)}} - X_{t_i^{(n)}} \right)^2 \xrightarrow{n \rightarrow \infty} t \quad P\text{-a.s.}$$

Here the zero set depends on the sequence  $(\tau_n)_{n \in \mathbb{N}}$  (and so far it could depend on  $t$ .)

(ii) Moreover,

$$P\left( \sum_{\substack{t_i^{(n)} \in \tau_n \\ t_i^{(n)} \leq t}} \left( X_{t_{i+1}^{(n)}} - X_{t_i^{(n)}} \right)^2 \xrightarrow{n \rightarrow \infty} t, \quad \forall t \geq 0 \right) = 1.$$

*Proof.* (i) See [Röc11, Satz 9.4.5].

(ii) Exercise (“sandwich argument”, notice that zero set in (i) a-priori depends on  $t$ ).

□

## 1.2. Quadratic Variation and Itô’s Formula

Fix a continuous and real-valued function  $t \mapsto X_t$  on  $[0, \infty[$  (in short  $(X_t)_{t \geq 0}$ ) with an existing sequence of subdivisions  $(\tau_n)$  with  $|\tau_n| \xrightarrow{n \rightarrow \infty} 0$  and  $t_{N_n}^{(n)} \xrightarrow{n \rightarrow \infty} \infty$  such that the quadratic variation (along  $(\tau_n)$ )

$$\langle X \rangle_t := \lim_{n \rightarrow \infty} \sum_{\substack{t_i^{(n)} \in \tau_n \\ t_i^{(n)} \leq t}} \left( X_{t_{i+1}^{(n)}} - X_{t_i^{(n)}} \right)^2 = \lim_{n \rightarrow \infty} \sum_{t_i^{(n)} \in \tau_n} (X_{t_{i+1}^{(n)} \wedge t} - X_{t_i^{(n)} \wedge t})^2, \quad t \geq 0, \quad (1.2.3)$$

exists for all  $t \geq 0$  and such that  $t \mapsto \langle X \rangle_t$  is continuous on  $[0, \infty)$ . Note that  $\langle X \rangle_t$  in (1.2.3) could depend upon the choice of  $(\tau_n)_{n \geq 1}$  in contrast to the limit in (1.1.2) (cf. [RY99, (2.3)-(2.5) p. 27/28]). By definition it is obvious that  $t \mapsto \langle X \rangle_t$  is increasing.

**Remark 1.2.5.** (i) If  $(X_t)_{t \geq 0}$  is of bounded variation, then

$$\sum_{t_i^{(n)} \in \tau_n} \left( X_{t_{i+1}^{(n)} \wedge t} - X_{t_i^{(n)} \wedge t} \right)^2 \leq \underbrace{\max_{t_i^{(n)} \in \tau_n} \left| X_{t_{i+1}^{(n)} \wedge t} - X_{t_i^{(n)} \wedge t} \right|}_{< \varepsilon \text{ uniformly for large } n} \underbrace{\sum_{t_i^{(n)} \in \tau_n} \left| X_{t_{i+1}^{(n)} \wedge t} - X_{t_i^{(n)} \wedge t} \right|}_{< \infty} \rightarrow 0,$$

hence,  $\langle X \rangle \equiv 0$ .

Therefore,  $\langle X \rangle \not\equiv 0$  implies that  $(X_t)_{t \geq 0}$  is not of bounded variation and the Lebesgue-Stieltjes-Integral can not be defined in the usual way.

(ii) If  $t \mapsto \langle X \rangle_t$  is increasing, continuous and  $\langle X \rangle_0 = 0$ , hence,  $t \mapsto \langle X \rangle_t$  is a distribution function of a measure  $\mu$  (i.e.  $d\mu = d\langle X \rangle_t$  on  $([0, \infty), \mathcal{B}([0, \infty))$ ), then (1.2.3) is equivalent to: The distribution function  $F_n$  of

$$\mu_n := \sum_{t_i^{(n)} \in \tau_n} \left( X_{t_{i+1}^{(n)}} - X_{t_i^{(n)}} \right)^2 \delta_{t_i^{(n)}}$$

converges pointwise to

$$F(t) := \mu([\!-\infty, t]) = \int 1_{[\!-\infty, t]} d\mu.$$

But, since for continuous  $\langle X \rangle_t$

$$\mu(\{t\}) = \lim_{n \rightarrow \infty} \mu([\!-\infty, t]) - \mu([\!-\infty, t - \frac{1}{n}]) = \lim_{n \rightarrow \infty} (\langle X \rangle_t - \langle X \rangle_{t - \frac{1}{n}}) = 0,$$

we have that the pointwise convergence of  $F_n$  to  $F$  is equivalent to  $\mu_n \rightarrow \mu$  weakly by Portemanteau.

(iii) Note that if  $X$  is an increasing function, then  $X$  is always of bounded variation:

$$\text{var}_t X = \sup_{\tau} \sum_{t_i \in \tau} |X_{t_{i+1} \wedge t} - X_{t_i \wedge t}| = \sup_{\tau} \sum_{t_i \in \tau} (X_{t_{i+1} \wedge t} - X_{t_i \wedge t}) = X_t - X_0 < \infty.$$

Therefore, we can define the integral with respect to  $\langle X \rangle$  as a Lebesgue-Stieltjes-Integral.

**Lemma 1.2.6** (Calculating integrals with respect to  $d\langle X \rangle_s$ ). *Let  $g \in \mathcal{C}([0, \infty))$ . Then*

$$\sum_{\substack{t_i^{(n)} \in \tau_n \\ t_i^{(n)} \leq t}} g(t_i) \left( X_{t_{i+1}^{(n)}} - X_{t_i^{(n)}} \right)^2 \rightarrow \int_0^t g(s) d\langle X \rangle_s.$$

*Proof.* The left hand side is equal to  $\int g 1_{[0, t]} d\mu_n$ , whereas the right hand side equals  $\int g 1_{[0, t]} d\mu$ . But the integrand  $g 1_{[0, t]}$  is  $\mu$ -a.e. continuous and bounded. Hence, convergence follows by the Portemanteau theorem and Remark 1.2.5 (ii).  $\square$

**Theorem 1.2.7** (Pathwise Itô-formula): Let  $F \in \mathcal{C}^2(\mathbb{R})$ . Then for all  $t \geq 0$  the Itô-Formula is given by

$$F(X_t) - F(X_0) = \int_0^t F'(X_s) dX_s + \frac{1}{2} \int_0^t F''(X_s) d\langle X \rangle_s, \quad (1.2.4)$$

where

$$\int_0^t F'(X_s) dX_s := \lim_{n \rightarrow \infty} \sum_{\substack{t_i^{(n)} \in \tau_n \\ t_i^{(n)} \leq t}} F'(X_{t_i^{(n)}}) \left( X_{t_{i+1}^{(n)}} - X_{t_i^{(n)}} \right).$$

$\int_0^t F'(X_s) dX_s$  is called the (pathwise) Itô-Integral and depends on  $(\tau_n)_{n \in \mathbb{N}}$ .

*Proof.* Consider  $(\tau_n)_{n \in \mathbb{N}}$  such that  $\langle X \rangle_t$  exists along  $(\tau_n)_{n \in \mathbb{N}}$ . We apply the Taylor formula to  $F$ . Hence, for all  $n \in \mathbb{N}$  there exist  $\theta_i^{(n)} \in ]0, 1[$  such that

$$\begin{aligned} & \sum_{\substack{t_i^{(n)} \in \tau_n \\ t_i^{(n)} \leq t}} F \left( X_{t_{i+1}^{(n)}} \right) - F \left( X_{t_i^{(n)}} \right) \xrightarrow{n \rightarrow \infty} F(X_t) - F(X_0) \\ &= \sum_{\substack{t_i^{(n)} \in \tau_n \\ t_i^{(n)} \leq t}} F' \left( X_{t_i^{(n)}} \right) \left( X_{t_{i+1}^{(n)}} - X_{t_i^{(n)}} \right) + \underbrace{\sum_{\substack{t_i^{(n)} \in \tau_n \\ t_i^{(n)} \leq t}} \frac{1}{2} F'' \left( X_{t_i^{(n)}} \right) \left( X_{t_{i+1}^{(n)}} - X_{t_i^{(n)}} \right)^2}_{\xrightarrow{1.2.6} \frac{1}{2} \int_0^t F''(X_s) d\langle X \rangle_s} \\ &+ \underbrace{\sum_{\substack{t_i^{(n)} \in \tau_n \\ t_i^{(n)} \leq t}} \frac{1}{2} \left[ F'' \left( X_{t_i^{(n)}} + \theta_i^{(n)} \left( X_{t_{i+1}^{(n)}} - X_{t_i^{(n)}} \right) \right) - F'' \left( X_{t_i^{(n)}} \right) \right] \cdot \left( X_{t_{i+1}^{(n)}} - X_{t_i^{(n)}} \right)^2}_{=: S} \end{aligned}$$

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Since  $F''$  is locally uniformly continuous and  $X$  is uniformly continuous,

$$\left| F'' \left( X_{t_i^{(n)}} + \theta_i^{(n)} \left( X_{t_{i+1}^{(n)}} - X_{t_i^{(n)}} \right) \right) - F'' \left( X_{t_i^{(n)}} \right) \right| < \varepsilon$$

holds uniformly in  $i$  for  $n$  big enough. Therefore,

$$S \leq \varepsilon \underbrace{\sum_{\substack{t_i^{(n)} \in \tau_n \\ t_i^{(n)} \leq t}} \left( X_{t_{i+1}^{(n)}} - X_{t_i^{(n)}} \right)^2}_{\xrightarrow{n \rightarrow \infty} \langle X \rangle_t} \xrightarrow{\varepsilon \rightarrow 0} 0,$$

which finishes the proof.  $\square$

**Remark 1.2.8.** If  $\langle X \rangle_t \equiv 0$  (e.g.  $(X_t)_{t \geq 0}$  is of bounded variation), then we are in the classical case:

$$F(X_t) - F(X_0) = \int_0^t F'(X_s) dX_s$$

is an ordinary Lebesgue-Stieltjes-Integral. (If  $X_t = t$ , this is the “Fundamental Theorem of Calculus”). We introduce a short-hand notation for (1.2.4)

$$dF(X) = F'(X) dX + \frac{1}{2} F''(X) d\langle X \rangle, \quad (1.2.5)$$

in contrast to the classical case (i.e.  $\langle X \rangle \equiv 0$ ), where

$$dF(X) = F'(X) dX. \quad (1.2.6)$$

**Example 1.2.9.** Consider the differential equation

$$dX^n = nX^{n-1} dX \quad \text{for } n \in \mathbb{N} \text{ fixed.}$$

If  $\langle X \rangle = 0$ , then a solution is  $X^n$ .

If  $\langle X \rangle \neq 0$ , this is not a solution, since by Itô

$$dX^n = nX^{n-1} dX + \frac{1}{2} n(n-1) X^{n-2} d\langle X \rangle.$$

We would like to find a function  $h_n : \mathbb{R} \rightarrow \mathbb{R}$  such that

$$dh_n(X) = nh_{n-1}(X) dX$$

in the general case  $\langle X \rangle \neq 0$ . Later (cf. Example 1.3.25(ii) below), we shall see that the  $n$ -th Hermite polynomial  $h_n$  will provide a solution to this problem.

**Remark 1.2.10.** We know that, if  $f \in \mathcal{C}^1(\mathbb{R})$ , then the Itô-integral  $\int_0^t f(X_s) dX_s$ ,  $t \geq 0$ , is (well-)defined. (Simply take  $F$  as a primitive of  $f$ , i.e.  $F' = f$ , and apply Itô's formula.)

**Definition 1.2.11** ( $\alpha$ -Integral). More generally we define for  $\alpha \in [0, 1]$  and  $f \in \mathcal{C}^1(\mathbb{R})$

$$\alpha\text{-} \int_0^t f(X_s) dX_s := \lim_{n \rightarrow \infty} \sum_{\substack{t_i^{(n)} \in \tau_n \\ t_i^{(n)} \leq t}} f \left( X_{t_i^{(n)}} + \alpha \left( X_{t_{i+1}^{(n)}} - X_{t_i^{(n)}} \right) \right) \cdot \left( X_{t_{i+1}^{(n)}} - X_{t_i^{(n)}} \right). \quad (1.2.7)$$

**Claim:** This limit exists and

$$\alpha \int_0^t f(X_s) dX_s = \int_0^t f(X_s) dX_s + \alpha \int_0^t f'(X_s) d\langle X \rangle_s. \quad (1.2.8)$$

*Proof.* Exercise (Compare “ $\alpha$ -sum” with “0-sum” (Itô-Integral) and use the mean-value-theorem for  $f$ ).  $\square$

**Special cases:**

$\alpha = 0$ : “Itô-integral”

$\alpha = 1$ : “Backward Ito-integral”.

$\alpha = \frac{1}{2}$ : “Stratonovich-Fisk-Integral”

**Notation:**  $\int \dots \circ dX_s := \frac{1}{2} \int \dots dX_s$ . Hence

$$\int_0^t f(X_s) \circ dX_s \left( = \frac{1}{2} \int_0^t f(X_s) dX_s \right) = \int_0^t f(X_s) dX_s + \frac{1}{2} \int_0^t f'(X_s) d\langle X \rangle_s$$

and we have by Itô the *Stratonovich-formula*

$$F(X_t) - F(X_0) = \int_0^t F'(X_s) \circ dX_s. \quad (1.2.9)$$

**Remark 1.2.12.** (i) An advantage of the Itô-integral is (see Section 1.4 below) that, if  $X$  is a martingale, then, again,  $\int f(X_s) dX_s$  is a martingale.

(ii) In the Stratonovich-formula one only has to deal with derivatives of first order and, therefore, it can be used for manifold-valued  $X$ .

### 1.2.1. Supplement on the Quadratic Variation

**Lemma 1.2.13.** (i) Let  $F \in C^1(\mathbb{R})$ . Then  $t \mapsto F(X_t)$  has (finite) quadratic variation (along fixed  $(\tau_n)_{n \in \mathbb{N}}$ )

$$\langle F(X) \rangle_t = \int_0^t (F'(X_s))^2 d\langle X \rangle_s \quad (\text{automatically continuous in } t).$$

(ii) If  $M_t := X_t + A_t, t \geq 0$ , for some  $t \mapsto A_t$  continuous and  $\langle A \rangle \equiv 0$  (again  $\langle A \rangle$  calculated along  $(\tau_n)$ ), then

$$\langle M \rangle_t = \langle X \rangle_t.$$

(iii) The Itô-integral  $t \mapsto \int_0^t f(X_s) dX_s =: M_t$  with  $f \in C^1(\mathbb{R})$ , has quadratic variation (along  $(\tau_n)$ ) and

$$\langle M \rangle_t = \left\langle \int_0^t f(X_s) dX_s \right\rangle_t = \int_0^t f(X_s)^2 d\langle X \rangle_s.$$

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*Proof.* (i) We first apply Taylor up to order 1, then take the square on both sides and finally apply the Binomial formula to get

$$\begin{aligned}
& \sum_{\substack{t_i^{(n)} \in \tau_n \\ t_i^{(n)} \leq t}} \left( F(X_{t_{i+1}^{(n)}}) - F(X_{t_i^{(n)}}) \right)^2 \\
&= \underbrace{\sum_{\substack{t_i^{(n)} \in \tau_n \\ t_i^{(n)} \leq t} F'(X_{t_i^{(n)}})^2 \left( X_{t_{i+1}^{(n)}} - X_{t_i^{(n)}} \right)^2}_{\xrightarrow{n \rightarrow \infty} \int_0^t (F'(X))^2 d\langle X \rangle_s \text{ by Lemma 1.2.6}} \\
&+ \sum_{\substack{t_i^{(n)} \in \tau_n \\ t_i^{(n)} \leq t}} \underbrace{\left( F' \left( X_{t_i^{(n)}} + \theta_i^{(n)} \left( X_{t_{i+1}^{(n)}} - X_{t_i^{(n)}} \right) \right) - F' \left( X_{t_i^{(n)}} \right) \right)^2}_{< \varepsilon \text{ for large } n, \text{ since } F' \text{ is uniformly continuous}} \cdot \left( X_{t_{i+1}^{(n)}} - X_{t_i^{(n)}} \right)^2 \\
&\quad \text{on the compact set } \{X_s | s \in [0, t+1]\} \\
&+ 2 \sum_{\substack{t_i^{(n)} \in \tau_n \\ t_i^{(n)} \leq t}} F'(X_{t_i^{(n)}}) \left( X_{t_{i+1}^{(n)}} - X_{t_i^{(n)}} \right) \\
&\cdot \left( F' \left( X_{t_{i+1}^{(n)}} + \theta_i^{(n)} \left( X_{t_{i+1}^{(n)}} - X_{t_i^{(n)}} \right) \right) - F' \left( X_{t_i^{(n)}} \right) \right) \left( X_{t_{i+1}^{(n)}} - X_{t_i^{(n)}} \right).
\end{aligned}$$

Since the second term goes to zero as  $n \rightarrow \infty$ , so, by Cauchy-Schwartz, does the third.

(ii)

$$\begin{aligned}
\sum_{\substack{t_i^{(n)} \in \tau_n \\ t_i^{(n)} \leq t}} \left( M_{t_{i+1}^{(n)}} - M_{t_i^{(n)}} \right)^2 &= \underbrace{\sum_{\substack{t_i^{(n)} \in \tau_n \\ t_i^{(n)} \leq t} \left( X_{t_{i+1}^{(n)}} - X_{t_i^{(n)}} \right)^2}_{\xrightarrow{n \rightarrow \infty} \langle X \rangle_t} + \underbrace{\sum_{\substack{t_i^{(n)} \in \tau_n \\ t_i^{(n)} \leq t} \left( A_{t_{i+1}^{(n)}} - A_{t_i^{(n)}} \right)^2}_{\xrightarrow{n \rightarrow \infty} 0 \text{ by assumption}} \\
&+ 2 \underbrace{\sum_{\substack{t_i^{(n)} \in \tau_n \\ t_i^{(n)} \leq t} \left( X_{t_{i+1}^{(n)}} - X_{t_i^{(n)}} \right) \left( A_{t_{i+1}^{(n)}} - A_{t_i^{(n)}} \right)}_{\xrightarrow{n \rightarrow \infty} 0 \text{ by Cauchy-Schwartz}}.
\end{aligned}$$

(iii) Let  $F \in \mathcal{C}^2(\mathbb{R})$  such that  $F' = f$  and apply Itô to get

$$M_t = F(X_t) - \underbrace{\left( F(X_0) + \frac{1}{2} \int_0^t F''(X_s) d\langle X \rangle_s \right)}_{=: A_t}.$$

But  $A_t$  can be written as a difference of increasing functions. Therefore,  $A_t$  is of bounded variation, hence,  $\langle A \rangle = 0$ . Thus, by (ii) and (i)

$$\langle M \rangle \stackrel{(ii)}{=} \langle F(X) \rangle_t \stackrel{(i)}{=} \int_0^t (F'(X_s))^2 d\langle X \rangle_s.$$

□

### 1.3. $d$ -Dimensional Itô-Formula and Covariation

Fix  $X, Y : [0, \infty) \rightarrow \mathbb{R}$  continuous with bounded quadratic variation  $\langle X \rangle, \langle Y \rangle$  (along the same  $(\tau_n)_{n \in \mathbb{N}}$ ) (cf. [RY99, (2.3)-(2.5) (p.27/28)]).

**Definition 1.3.14.** *If*

$$\langle X, Y \rangle_t := \lim_{n \rightarrow \infty} \sum_{\substack{t_i^{(n)} \in \tau_n \\ t_i^{(n)} \leq t}} \left( X_{t_{i+1}^{(n)}} - X_{t_i^{(n)}} \right) \left( Y_{t_{i+1}^{(n)}} - Y_{t_i^{(n)}} \right), \quad t \geq 0,$$

*exists, then it is called the covariation of  $X$  and  $Y$  (along  $(\tau_n)$ ).*

**Lemma 1.3.15.** *The following assertions are equivalent:*

(i)  $\langle X, Y \rangle$  exists and is continuous.

(ii)  $\langle X + Y \rangle$  exists and is continuous. In this case the Polarization identity holds:

$$\langle X, Y \rangle = \frac{1}{2} (\langle X + Y \rangle - \langle X \rangle - \langle Y \rangle).$$

*In particular,  $\langle X, Y \rangle$  is the distribution function of a signed measure on  $\mathbb{R}_+$*

$$d\langle X, Y \rangle = \frac{1}{2} d\langle X + Y \rangle - \frac{1}{2} d\langle X \rangle - \frac{1}{2} d\langle Y \rangle.$$

*Furthermore, if  $\langle X, Y \rangle$  exists, we have*

$$|\langle X, Y \rangle| \leq \langle X \rangle^{\frac{1}{2}} \langle Y \rangle^{\frac{1}{2}},$$

*i.e. a Cauchy-Schwartz inequality.*

*Proof.* Exercise. □

**Remark 1.3.16.**  $|\langle X, Y \rangle| \leq \langle X \rangle^{\frac{1}{2}} \langle Y \rangle^{\frac{1}{2}}$  is a special case of the Kunita-Watanabe-inequality (cf. 2.2.16 below).

**Example 1.3.17.** (i) Let  $(X_t)_{t \geq 0}, (Y_t)_{t \geq 0}$  be independent Brownian motions on  $(\Omega, \mathcal{F}, P)$ . Then there exists  $\langle X, Y \rangle(\omega)$  for  $P$ -a.e.  $\omega \in \Omega$  and

$$\langle X, Y \rangle(\omega) = 0 \quad P\text{-a.e. } \omega \in \Omega.$$

*Proof.* We know that

$$Z_t := \frac{1}{\sqrt{2}}(X_t + Y_t), \quad t \geq 0,$$

is a Brownian motion. Hence, (cf. Proposition 1.1.4(i))  $\langle Z \rangle_t = t$ , so there exists  $\langle X + Y \rangle_t = 2\langle Z \rangle_t = 2t$ . Then it follows by Lemma 1.3.15 applied to  $P$ -a.e.  $\omega \in \Omega$  that

$$\langle X, Y \rangle = \frac{1}{2} (\langle X + Y \rangle - \langle X \rangle - \langle Y \rangle) = 0.$$

□

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(ii) Let  $f, g \in \mathcal{C}(\mathbb{R})$  and

$$Y_t := \int_0^t f(X_s) d\langle X \rangle_s, \quad Z_t := \int_0^t g(X_s) d\langle X \rangle_s.$$

Then (again with respect to our  $(\tau_n)$ ) there exists

$$\langle Y, Z \rangle_t = \int_0^t f(X_s)g(X_s) d\langle X \rangle_s.$$

*Proof.* By 1.2.13(iii) the quadratic variation of

$$Y_t + Z_t = \int_0^t (f + g)(X_s) dX_s$$

along our  $(\tau_n)$  exists. Hence, by the polarization identity and Lemma 1.2.13(iii) we get

$$\begin{aligned} 2\langle Y, Z \rangle &= \langle Y + Z \rangle - \langle Y \rangle - \langle Z \rangle \\ &= 2 \int f(X_s)g(X_s) d\langle X \rangle_s + \int f(X_s)^2 d\langle X \rangle_s + \int g(X_s)^2 d\langle X \rangle_s - \langle Y \rangle - \langle Z \rangle \\ &= 2 \int f(X_s)g(X_s) d\langle X \rangle_s. \end{aligned}$$

□

**Proposition 1.3.18** (Itô's product rule). *Let  $X, Y$  be as above such that there exists  $\langle X, Y \rangle$  (with respect to  $(\tau_n)$ ) and is continuous in  $t \geq 0$ . If there exists either*

$$\lim_{n \rightarrow \infty} \sum_{\substack{t_i^{(n)} \in \tau_n \\ t_i^{(n)} \leq t}} X_{t_i^{(n)}} \left( Y_{t_{i+1}^{(n)}} - Y_{t_i^{(n)}} \right) =: \int_0^t X_s dY_s$$

or

$$\lim_{n \rightarrow \infty} \sum_{\substack{t_i^{(n)} \in \tau_n \\ t_i^{(n)} \leq t}} Y_{t_i^{(n)}} \left( X_{t_{i+1}^{(n)}} - X_{t_i^{(n)}} \right) =: \int_0^t Y_s dX_s,$$

then both of these limits exist and

$$X_t Y_t = X_0 Y_0 + \int_0^t X_s dY_s + \int_0^t Y_s dX_s + \langle X, Y \rangle_t.$$

*Proof.* We have

$$X_t Y_t = \frac{1}{2} \left( (X_t + Y_t)^2 - X_t^2 - Y_t^2 \right).$$

Furthermore, by Lemma 1.3.15  $\langle X + Y \rangle$  exists and is continuous, since by Itô with  $F(X) = \frac{1}{2}X^2$  we know that

$$\begin{aligned} X_t Y_t &= \underbrace{\frac{1}{2} \left( (X_0 + Y_0)^2 - X_0^2 - Y_0^2 \right)}_{=X_0 Y_0} + \int_0^t (X + Y)_s d\langle X + Y \rangle_s \\ &\quad - \int_0^t X_s dX_s - \int_0^t Y_s dY_s + \underbrace{\frac{1}{2} \left( \langle X + Y \rangle_t - \langle X \rangle_t - \langle Y \rangle_t \right)}_{=\langle X, Y \rangle_t}. \end{aligned}$$

By definition we have

$$\begin{aligned} & \int_0^t (X + Y)_s \, d(X + Y)_s \\ &= \lim_{n \rightarrow \infty} \sum_{\substack{t_i^{(n)} \in \tau_n \\ t_i^{(n)} \leq t}} \underbrace{\left( X_{t_i^{(n)}} + Y_{t_i^{(n)}} \right) \left( X_{t_{i+1}^{(n)}} - X_{t_i^{(n)}} + Y_{t_{i+1}^{(n)}} - Y_{t_i^{(n)}} \right)} \\ & \quad = X_{t_i^{(n)}} \left( X_{t_{i+1}^{(n)}} - X_{t_i^{(n)}} \right) + Y_{t_i^{(n)}} \left( Y_{t_{i+1}^{(n)}} - Y_{t_i^{(n)}} \right) \\ & \quad + X_{t_i^{(n)}} \left( Y_{t_{i+1}^{(n)}} - Y_{t_i^{(n)}} \right) + Y_{t_i^{(n)}} \left( X_{t_{i+1}^{(n)}} - X_{t_i^{(n)}} \right) \end{aligned}$$

Therefore, we get

$$X_t Y_t = X_0 Y_0 + \lim_{n \rightarrow \infty} \sum_{\substack{t_i^{(n)} \in \tau_n \\ t_i^{(n)} \leq t}} \left( X_{t_i^{(n)}} \left( Y_{t_{i+1}^{(n)}} - Y_{t_i^{(n)}} \right) + Y_{t_i^{(n)}} \left( X_{t_{i+1}^{(n)}} - X_{t_i^{(n)}} \right) \right) + \langle X, Y \rangle_t,$$

which implies the assertion.  $\square$

**Remark 1.3.19.** *If  $X$  or  $Y$  has bounded variation, e.g.  $X$ , then there already exists  $\int Y_s \, dX_s$  and all assumptions in Proposition 1.3.18 are fulfilled. In this case  $\langle X, Y \rangle = 0$  and, by substituting  $dY_s = Y'_s ds$ , we are in the classical case of integration by parts:*

$$X_t Y_t - X_0 Y_0 = \int X_s Y'_s \, ds + \int Y_s X'_s \, ds.$$

**Example 1.3.20.** *Suppose  $t \mapsto Y_t$  is of bounded variation, hence,  $\langle Y \rangle \equiv 0$  and  $\langle X, Y \rangle = 0$  (by Hölder). Then by Proposition 1.3.18*

$$\int_0^t Y_s \, dX_s = - \int_0^t X_s \, dY_s + X_t Y_t - X_0 Y_0.$$

Here, we can define the left hand side by the right hand side since  $\int_0^t X_s \, dY_s$  is a usual Lebesgue-Stieltjes integral. This approach was used by Paley-Wiener to define stochastic integrals, if  $X$  is a Brownian motion:

Let  $X(\omega)$  be a typical Brownian path (hence,  $X_0(\omega) = 0$ ) and  $h(s) (= Y_s)$  continuous, of bounded variation and independent of  $\omega$  with  $h(1) = 0$ . Define

$$\int_0^1 h(s) \, dX_s(\omega) := - \int_0^1 X_s(\omega) \, dh(s).$$

One can show that

$$E \left[ \left( \int_0^1 h(s) \, dX_s \right)^2 \right] = \int_0^1 h(s)^2 \, ds,$$

hence,

$$\begin{array}{ccc} \mathcal{L}^2([0, 1], ds) & \rightarrow & \mathcal{L}^2(\Omega, \mathcal{F}, P) \\ \cup & & \cup \\ h & \mapsto & \int h(s) \, dX_s(\omega) \end{array}$$

is an isometry. It is first defined for a dense subset of functions  $h$  in  $\mathcal{L}^2([0, 1], ds)$ , and then extended to the closure, i.e. for all  $h \in \mathcal{L}^2([0, 1], ds)$ , by this isometry.

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Fix now  $X_t = (X_t^1, \dots, X_t^d) : [0, \infty) \rightarrow \mathbb{R}^d$  continuous with continuous  $\langle X^i \rangle, \langle X^i, X^j \rangle$ ,  $i, j \in \{1, \dots, d\}$ ,  $i \neq j$  (along our  $(\tau_n)$ ).

**Proposition 1.3.21** (*d*-dimensional Itô-formula). *Let  $F \in \mathcal{C}^2(\mathbb{R}^d)$ . Then with  $(\cdot, \cdot)_t$  as Euklidian inner product on  $\mathbb{R}$*

$$F(X_t) - F(X_0) = \int_0^t (\nabla F(X_s), dX_s) + \frac{1}{2} \int_0^t \sum_{k,l=1}^d \frac{\partial^2 F}{\partial x_k \partial x_l}(X_s) d\langle X^k, X^l \rangle_s, \quad (1.3.10)$$

where

$$\int_0^t (\nabla F(X_s), dX_s) := \lim_{n \rightarrow \infty} \sum_{\substack{t_i^{(n)} \in \tau_n \\ t_i^{(n)} \leq t}} \left( \nabla F(X_{t_i^{(n)}}), X_{t_{i+1}^{(n)}} - X_{t_i^{(n)}} \right)$$

is called multidimensional Itô-integral.

*Proof.* Obviously, we have

$$\sum_{\substack{t_i^{(n)} \in \tau_n \\ t_i^{(n)} \leq t}} F(X_{t_{i+1}^{(n)}}) - F(X_{t_i^{(n)}}) \xrightarrow{n \rightarrow \infty} F(X_t) - F(X_0).$$

Furthermore, by *d*-dimensional Taylor formula we obtain

$$\begin{aligned} & \sum_{\substack{t_i^{(n)} \in \tau_n \\ t_i^{(n)} \leq t}} F(X_{t_{i+1}^{(n)}}) - F(X_{t_i^{(n)}}) \\ &= \sum_{\substack{t_i^{(n)} \in \tau_n \\ t_i^{(n)} \leq t}} \left( \nabla F(X_{t_i^{(n)}}), X_{t_{i+1}^{(n)}} - X_{t_i^{(n)}} \right) \\ &+ \frac{1}{2} \underbrace{\sum_{\substack{t_i^{(n)} \in \tau_n \\ t_i^{(n)} \leq t}} \left( A(X_{t_i^{(n)}}) \left( X_{t_{i+1}^{(n)}} - X_{t_i^{(n)}} \right), \left( X_{t_{i+1}^{(n)}} - X_{t_i^{(n)}} \right) \right)}_{=:S} \\ &+ \frac{1}{2} \sum_{\substack{t_i^{(n)} \in \tau_n \\ t_i^{(n)} \leq t}} \left( \left( A \left( X_{t_i^{(n)}} + \theta_i^n \left( X_{t_{i+1}^{(n)}} - X_{t_i^{(n)}} \right) \right) \right) - A \left( X_{t_i^{(n)}} \right) \right) \\ &\quad \cdot \left( X_{t_{i+1}^{(n)}} - X_{t_i^{(n)}} \right), \left( X_{t_{i+1}^{(n)}} - X_{t_i^{(n)}} \right) \right), \end{aligned}$$

where

$$A(x) = \left( \frac{\partial^2}{\partial x_i \partial x_j} F(x) \right)_{i,j}.$$

The third summand vanishes analogously to the 1-dimensional case. Moreover, since we can

interchange the sums, we have by polarization

$$\begin{aligned}
 S &= \sum_{j,k=1}^d \sum_{\substack{t_i^{(n)} \in \tau_n \\ t_i^{(n)} \leq t}} \frac{\partial^2 F}{\partial x_j \partial x_k} (X_{t_i^{(n)}}) \left( X_{t_{i+1}^{(n)}}^{(j)} - X_{t_i^{(n)}}^{(j)} \right) \left( X_{t_{i+1}^{(n)}}^{(k)} - X_{t_i^{(n)}}^{(k)} \right) \\
 &= \sum_{j,k=1}^d \sum_{\substack{t_i^{(n)} \in \tau_n \\ t_i^{(n)} \leq t}} \frac{\partial^2 F}{\partial x_j \partial x_k} (X_{t_i^{(n)}}) \\
 &\quad \cdot \frac{1}{2} \left( \left( \left( X_{t_{i+1}^{(n)}}^{(j)} + X_{t_{i+1}^{(n)}}^{(k)} \right) - \left( X_{t_i^{(n)}}^{(j)} + X_{t_i^{(n)}}^{(k)} \right) \right)^2 - \left( X_{t_{i+1}^{(n)}}^{(j)} - X_{t_i^{(n)}}^{(j)} \right)^2 - \left( X_{t_{i+1}^{(n)}}^{(k)} - X_{t_i^{(n)}}^{(k)} \right)^2 \right) \\
 &\xrightarrow{n \rightarrow \infty} \frac{1}{2} \sum_{j,k=1}^d \left( \int_0^t \frac{\partial^2 F}{\partial x_j \partial x_k} (X_s) d\langle X^{(j)} + X^{(k)} \rangle_s - \int_0^t \frac{\partial^2 F}{\partial x_j \partial x_k} (X_s) d\langle X^{(j)} \rangle_s \right. \\
 &\quad \left. - \int_0^t \frac{\partial^2 F}{\partial x_j \partial x_k} (X_s) d\langle X^{(k)} \rangle_s \right) \\
 &= \int_0^t \sum_{k,l=1}^d \frac{\partial^2 F}{\partial x_k \partial x_l} (X_s) d\langle X^k, X^l \rangle_s, \quad P\text{-a.s.}
 \end{aligned}$$

□

**Remark 1.3.22.** (i) We have defined

$$\int_0^t (\nabla F(X_s), dX_s) := \lim_{n \rightarrow \infty} \sum_{k=1}^d \sum_{\substack{t_i^{(n)} \in \tau_n \\ t_i^{(n)} \leq t}} \frac{\partial F}{\partial x_k} (X_{t_i^{(n)}}) \left( X_{t_{i+1}^{(n)}}^k - X_{t_i^{(n)}}^k \right),$$

but we cannot interchange  $\sum_{k=1}^d$  with the limit, since we do not know whether for every  $k$  the limit exists.

(ii) For  $F \in C^1(\mathbb{R}^d)$  the function  $t \mapsto F(X_t)$  has continuous quadratic variation and

$$\langle F(X) \rangle_t = \sum_{k,l=1}^d \int_0^t \frac{\partial F}{\partial x_k} (X_s) \frac{\partial F}{\partial x_l} (X_s) d\langle X^k, X^l \rangle_s.$$

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*Proof.* By Taylor

$$\begin{aligned}
& \sum_{\substack{t_i^{(n)} \in \tau_n \\ t_i^{(n)} \leq t}} \left( F \left( X_{t_{i+1}^{(n)}} \right) - F \left( X_{t_i^{(n)}} \right) \right)^2 \\
&= \underbrace{\sum_{k,l=1}^d \sum_{\substack{t_i^{(n)} \in \tau_n \\ t_i^{(n)} \leq t}} \frac{\partial F}{\partial x_k} \left( X_{t_i^{(n)}} \right) \frac{\partial F}{\partial x_l} \left( X_{t_i^{(n)}} \right) \left( X_{t_{i+1}^{(n)}}^k - X_{t_i^{(n)}}^k \right) \left( X_{t_{i+1}^{(n)}}^l - X_{t_i^{(n)}}^l \right)}_{\xrightarrow{1.2.6} \sum_{k,l=1}^d \int_0^t \frac{\partial F}{\partial x_k} (X_s) \frac{\partial F}{\partial x_l} (X_s) d\langle X^k, X^l \rangle_s \text{ by polarization}} \\
&+ \sum_{k,l=1}^d \sum_{\substack{t_i^{(n)} \in \tau_n \\ t_i^{(n)} \leq t}} \underbrace{\left( \frac{\partial F}{\partial x_k} \left( X_{t_i^{(n)}} + \theta_i^n \left( X_{t_{i+1}^{(n)}} - X_{t_i^{(n)}} \right) \right) - \frac{\partial F}{\partial x_k} \left( X_{t_i^{(n)}} \right) \right)}_{< \varepsilon \text{ uniformly in } i \text{ for } n \text{ big enough}} \\
&\cdot \underbrace{\left( \frac{\partial F}{\partial x_l} \left( X_{t_i^{(n)}} + \theta_i^n \left( X_{t_{i+1}^{(n)}} - X_{t_i^{(n)}} \right) \right) - \frac{\partial F}{\partial x_l} \left( X_{t_i^{(n)}} \right) \right)}_{< \infty} \left( X_{t_{i+1}^{(n)}}^k - X_{t_i^{(n)}}^k \right) \left( X_{t_{i+1}^{(n)}}^l - X_{t_i^{(n)}}^l \right) \\
&+ 2 \sum_{k,l=1}^d \sum_{\substack{t_i^{(n)} \in \tau_n \\ t_i^{(n)} \leq t}} \left( \frac{\partial F}{\partial x_k} \left( X_{t_i^{(n)}} + \theta_i^n \left( X_{t_{i+1}^{(n)}} - X_{t_i^{(n)}} \right) \right) - \frac{\partial F}{\partial x_k} \left( X_{t_i^{(n)}} \right) \right) \left( X_{t_{i+1}^{(n)}}^k - X_{t_i^{(n)}}^k \right) \\
&\cdot \sum_{\substack{t_i^{(n)} \in \tau_n \\ t_i^{(n)} \leq t}} \frac{\partial F}{\partial x_l} \left( X_{t_i^{(n)}} \right) \left( X_{t_{i+1}^{(n)}}^l - X_{t_i^{(n)}}^l \right).
\end{aligned}$$

By Cauchy Schwartz the last two double sums converge to 0 as  $n$  goes to  $\infty$ .  $\square$

(iii) Since

$$\frac{1}{2} \int_0^t \sum_{k,l=1}^d \frac{\partial^2 F}{\partial x_k \partial x_l} (X_s) d\langle X^k, X^l \rangle_s$$

is of bounded variation, its quadratic variation is 0. But then, by 5.2.7 and 1.2.13 (ii), we can conclude that

$$\left\langle \int_0^\cdot (\nabla F(X_s), dX_s) \right\rangle_t = \langle F(X) - F(X_0) \rangle_t = \langle F(X) \rangle_t \stackrel{(ii)}{=} \sum_{k,l=1}^d \int_0^t \frac{\partial F}{\partial x_k} (X_s) \frac{\partial F}{\partial x_l} d\langle X^k, X^l \rangle_s.$$

### 1.3.1. Important special cases

#### Brownian motion in $\mathbb{R}^d$ and Laplace-operator $\Delta$

The components of a  $d$ -dimensional Brownian motion are independent, hence, (by 1.3.17(i))

$$\langle X^k, X^l \rangle_t(\omega) = \delta_{kl} \cdot t \quad (\text{“covariation reflects independence”}),$$

which implies by Itô for  $F \in \mathcal{C}^2(\mathbb{R})$

$$F(X_t) - F(X_0) = \int_0^t (\nabla F(X_s), dX_s) + \frac{1}{2} \int_0^t \Delta F(X_s) ds.$$

In particular, if  $F$  is harmonic (i.e.  $\Delta F = 0$ ), then

$$F(X_t) = F(0) + \int_0^t (\nabla F(X_s), dX_s),$$

which is an Itô-integral of a Brownian motion. Hence, a harmonic function preserves the martingale property since the Itô-integral again is a martingale (see Section 1.4 below).

### Itô-formula for time dependent functions

**Proposition 1.3.23.** *Let  $F \in C^2(\mathbb{R}^2)$  and  $X : [0, \infty) \rightarrow \mathbb{R}$  be continuous with continuous  $\langle X \rangle$  (along  $(\tau_n)_{n \in \mathbb{N}}$ ). Then*

$$F(X_t, \langle X \rangle_t) = F(X_0, 0) + \int_0^t \frac{\partial F}{\partial x}(X_s, \langle X \rangle_s) dX_s + \int_0^t \left( \frac{1}{2} \frac{\partial^2 F}{\partial x^2} + \frac{\partial F}{\partial y} \right) (X_s, \langle X \rangle_s) d\langle X \rangle_s.$$

*Proof.* Apply Itô for  $d = 2$  with  $(X_t, \langle X \rangle_t)$  to get

$$\begin{aligned} & F(X_t, \langle X \rangle_t) - F(X_0, 0) \\ &= \underbrace{\int_0^t \frac{\partial F}{\partial x}(X_s, \langle X \rangle_s) dX_s + \int_0^t \frac{\partial F}{\partial y}(X_s, \langle X \rangle_s) d\langle X \rangle_s}_{=: S} + \frac{1}{2} \int_0^t \frac{\partial^2 F}{\partial x^2}(X_s, \langle X \rangle_s) d\langle X \rangle_s \\ &+ \underbrace{\frac{1}{2} \int_0^t \frac{\partial^2 F}{\partial y^2}(X_s, \langle X \rangle_s) d\langle \langle X \rangle_s, \langle X \rangle_s \rangle + \frac{1}{2} \int_0^t \frac{\partial^2 F}{\partial x \partial y}(X_s, \langle X \rangle_s) d\langle X_s, \langle X \rangle_s \rangle}_{=0 \text{ since } \langle X \rangle_s \text{ is of bounded variation and by 1.3.20}}. \end{aligned}$$

The second summand of  $S$  exists since  $\langle X \rangle_t$  is of bounded variation (along  $(\tau_n)_{n \in \mathbb{N}}$ ). As the whole sum  $S$  exists by Itô ( $d = 2$ ), so does the first summand.  $\square$

**Remark 1.3.24.** (i) *If  $X_t$  is a Brownian motion, then  $\langle X \rangle_t$  represents the time, i.e.*

$$F(X_t, \langle X \rangle_t) = F(X_t, t).$$

*Therefore,  $\langle X \rangle_t$  is also called inner clock of  $X_t$ .*

(ii) *If  $F$  is a solution to the backward heat equation, i.e.*

$$\frac{1}{2} \frac{\partial^2 F}{\partial x^2} + \frac{\partial F}{\partial t} = 0,$$

*then we have by above (for  $X$  as above)*

$$F(X_t, \langle X \rangle_t) = F(X_0, 0) + \int_0^t \frac{\partial F}{\partial x}(X_s, \langle X \rangle_s) dX_s = F(X_0, 0) + \text{“Itô-integral”}.$$

*Later we shall see that the Itô-integral is a local martingale, if  $X$  is one. Therefore, every solution to the backward heat equation provides a local martingale!*

**Example 1.3.25.** (i) *Let  $F(x, t) := \exp(\alpha x - \frac{1}{2}\alpha^2 t)$ . Then  $F$  solves the backward heat equation. Thus, if  $X_0 = 0$ , then*

$$G_t := F(X_t, \langle X \rangle_t) = \exp\left(\alpha X_t - \frac{1}{2}\alpha^2 \langle X \rangle_t\right)$$

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solves the differential equation

$$\begin{aligned} G_0 &= 1, \\ dG &= \alpha G dX, \end{aligned}$$

i.e.

$$G_t = 1 + \int_0^t \alpha G_s dX_s.$$

Recall, if  $\langle X \rangle \equiv 0$ , then  $G_t = \exp(\alpha X_t)$ . Hence, in the Itô-calculus  $G_t$  as above is the right analogon to the exponential function  $\exp(\alpha t)$  in the usual case.

**Application to Brownian motion:**

For  $X$  with  $\langle X \rangle_t = t$

$$G_t(\omega) := G_0 e^{\beta t} \exp\left(\alpha X_t(\omega) - \frac{1}{2}\alpha^2 t\right), \quad t \geq 0,$$

by Proposition 1.3.23 solves the linear SDE

$$dG = \alpha G dX + \beta G dt.$$

**Classical** ( $\alpha = 0$ ):

$$dG = \beta G dt \quad \Rightarrow \quad G_t = G_0 e^{\beta t}.$$

Here, for  $\beta > 0$   $G_t$  tends to  $\infty$  as  $t \rightarrow \infty$ .

**Stochastic** ( $\alpha \neq 0$ ):

$$G_t = G_0 \exp\left(\alpha X_t + \left(\beta - \frac{1}{2}\alpha^2\right)t\right).$$

By law of iterated logarithm

$$\frac{X_t}{t} \xrightarrow{t \rightarrow \infty} 0, \quad a.s.,$$

hence, for large  $t$

$$\left|\alpha \frac{X_t}{t}\right| < \frac{1}{2} \left(\frac{1}{2}\alpha^2 - \beta\right).$$

Thus, for  $\beta < \frac{1}{2}\alpha^2$

$$G_t \leq G_0 e^{-(\frac{1}{2}\alpha^2 - \beta)\frac{t}{2}} \xrightarrow{t \rightarrow \infty} 0, \quad (a.s. \text{ pathwise stable}).$$

But  $G_t$  is not uniformly integrable, since by the Residue theorem

$$E[e^{\alpha X_t}] = e^{\frac{1}{2}\alpha^2 t},$$

and therefore,

$$E[G_t] = G_0 e^{\beta t} E[e^{\alpha X_t}] e^{-\frac{1}{2}\alpha^2 t} = G_0 e^{\beta t} \rightarrow \infty, \quad \text{if } \beta > 0 \quad (\text{unstable in the mean}).$$

(ii) “Hermite- polynomials” (cf. Example 1.2.9):

Define  $h_n(x, t)$  by

$$e^{\alpha x - \frac{1}{2}\alpha^2 t} = \sum_{n=0}^{\infty} \frac{\alpha^n}{n!} h_n(x, t), \quad (1.3.11)$$

where the left hand side is analytic in  $\alpha$ , i.e.

$$h_n(x, t) = \frac{\partial^n}{\partial \alpha^n} \left( e^{\alpha x - \frac{1}{2} \alpha^2 t} \right) \Big|_{\alpha=0} \left( = e^{\alpha x} \sum_{k=0}^n \binom{n}{k} x^k \frac{\partial^{n-k}}{\partial \alpha^{n-k}} \left( e^{-\frac{1}{2} \alpha^2 t} \right) \Big|_{\alpha=0} \right). \quad (1.3.12)$$

Recall, that for all  $n \in \mathbb{N}$

$$H_n(\cdot, t) := \frac{1}{\sqrt{n! t^n}} h_n(\cdot, t)$$

is an orthonormal basis of  $\mathcal{L}^2(\mathbb{R}, N(0, t))$ .

*Proof.* Clearly, (by monotone classes)  $\text{span}\{H_n, n \in \mathbb{N}\}$  is a dense subset of  $\mathcal{L}^2(\mathbb{R}, N(0, t))$ . Hence, it is sufficient to show that it is an ONS:

$$\sum_{n,m} \frac{\alpha^n}{n!} h_n(x, t) \frac{\beta^m}{m!} h_m(x, t) \stackrel{(1.3.11)}{=} e^{(\alpha+\beta)x - \frac{1}{2}(\alpha+\beta)^2 t} = \exp((\alpha + \beta)x) e^{-\frac{1}{2}(\alpha+\beta)^2 t} e^{\alpha\beta t}.$$

By integration with  $N(0, t)$  in  $x$ , since sums interchange with integration and

$$\int e^{(\alpha+\beta)x} N(0, t)(dx) = e^{\frac{1}{2}(\alpha+\beta)^2 t},$$

we get by (1.3.12) that

$$\sum_{n,m} \alpha^m \beta^m \int \frac{1}{n!} h_n(x, t) \frac{1}{m!} h_m(x, t) N(0, t)(dx) = e^{\alpha\beta t} = \sum_n \alpha^n \beta^n \frac{t^n}{n!}, \quad \forall \alpha, \beta.$$

□

We have for  $n \geq 0$  by (1.3.12) and by interchanging order of differentiation

$$\frac{1}{2} \frac{\partial^2}{\partial x^2} h_n + \frac{\partial}{\partial t} h_n = 0.$$

Additionally, for  $n \geq 1$  we have

$$\frac{\partial}{\partial x} h_n(\cdot, t) \stackrel{(1.3.11)}{=} n \cdot h_{n-1}(\cdot, t).$$

Therefore, if  $X, \langle X \rangle$  are continuous and  $X_0 = 0$ , (since  $h_n(0, 0) = 0$ ), we get

$$\begin{aligned} h_n(X_t, \langle X \rangle_t) &= \int_0^t \frac{\partial}{\partial x} h_n(X_{t_1}, \langle X \rangle_{t_2}) dX_{t_1} = n \int_0^t h_{n-1}(X_{t_1}, \langle X \rangle_{t_1}) dX_{t_1} \\ &= \dots = n! \int_0^t dX_{t_1} \int_0^{t_1} dX_{t_2} \dots \int_0^{t_{n-1}} dX_{t_n}. \end{aligned}$$

## 1.4. Itô-Integrals as (Local) Martingales

**Recall:** If  $(\mathcal{F}_t)_{t \geq 0}$  is not right-continuous, then define

$$\mathcal{F}_{t+} := \bigcap_{s>t} \mathcal{F}_s, \quad t \geq 0.$$

Then  $(\mathcal{F}_{t+})_{t \geq 0}$  is right-continuous.

Let  $(\Omega, \mathcal{F}, P)$  be a probability space equipped with a right-continuous filtration  $(\mathcal{F}_t)_{t \geq 0}$ , i.e.

$$\mathcal{F}_t = \bigcap_{s>t} \mathcal{F}_s.$$

Let  $T : \Omega \rightarrow [0, \infty]$  and let  $X : \{(t, \omega) \in ]0, \infty[ \times \Omega : t < T(\omega)\} \cup \{0\} \times \Omega \rightarrow \mathbb{R}$  be a map such that  $X$  is a *continuous local martingale* (up to  $T$ ) i.e. there exists a localizing sequence  $T_1 \leq T_2 \leq \dots \leq T$  and  $\langle X \rangle$  on  $\{T > 0\}$ , i.e. they are  $(\mathcal{F}_t)$ -stopping times such that

- (i)  $(X_{t \wedge T_n})_{t \geq 0}$  is a continuous  $(\mathcal{F}_t)$ -martingale for all  $n \in \mathbb{N}$ ,
- (ii)  $\sup_{n \in \mathbb{N}} T_n = T$   $P$ -a.s..

Assume in addition, that  $X$  has a continuous quadratic variation  $[0, T(\omega)[ \ni t \rightarrow \langle X \rangle_t(\omega)$  along a sequence of partitions  $(\tau_n)$  as above for  $P$ -a.e.  $\omega \in \Omega$ .

**Remark 1.4.26.** *Later we shall see that for any  $(\tau_n)$  there exists  $(\tau_{n_k})$  such that  $X$  as above has continuous quadratic variation  $\langle X \rangle_t$  along  $(\tau_{n_k})$  for  $P$ -a.e.  $\omega \in \Omega$ .*

**Proposition 1.4.27.** *Let  $f \in \mathcal{C}^2(G \times \mathbb{R}_+)$ ,  $G \subset \mathbb{R}^1$  open (or  $f \in \mathcal{C}^1(G)$ ). Assume there exists a compact set  $K \subset G$  such that*

$$X_0(\omega) \in K \quad \text{for } P\text{-a.e. } \omega \in \Omega.$$

Define  $M_0(\omega) = 0$  for all  $\omega \in \Omega$  and

$$M_t(\omega) := \int_0^t f(X_s(\omega), \langle X \rangle_s(\omega)) dX_s(\omega), \quad 0 < t < T(\omega) \text{ for } \omega \in \{T > 0\}. \quad (\text{Itô-integral!})$$

Then  $M$  is a continuous local martingale up to

$$S := \underbrace{\inf\{t > 0 \mid X_t \notin G\}}_{=: \sigma_{G^c}} \wedge T.$$

*Proof. Step 1:* Assume that  $T = \infty$ ,  $X$  is a bounded martingale,  $G = \mathbb{R}^1$  and  $f$  is bounded. Then  $M_t(\omega)$  is defined for all  $t \geq 0$ , since  $S = +\infty$ .

**Claim:**  $(M_t)_{t \geq 0}$  is a continuous martingale.

To see this define for  $n \in \mathbb{N}$  fixed

$$M_t^{(n)} := \sum_{\substack{i^{(n)} \in \tau_n \\ t_i^{(n)} \leq t}} f\left(X_{t_i^{(n)}}, \langle X \rangle_{t_i^{(n)}}\right) \left(X_{t_{i+1}^{(n)}} - X_{t_i^{(n)}}\right).$$

To see that  $(M_t^{(n)})_{t \geq 0}$  is a martingale, (but not necessarily continuous), let  $s < t$  and take  $t_k^{(n)}, t_l^{(n)} \in \tau_n$  such that  $t_k^{(n)} \leq t < t_{k+1}^{(n)}, t_l^{(n)} \leq s < t_{l+1}^{(n)}$ . Then, we have  $M_t^{(n)} = M_{t_k}^{(n)}$  and

$M_s^{(n)} = M_{t_l}^{(n)}$  and therefore,

$$\begin{aligned} E[M_t^{(n)} - M_s^{(n)} | \mathcal{F}_s] &= E[M_{t_k}^{(n)} - M_{t_l}^{(n)} | \mathcal{F}_s] \\ &= \sum_{i=l+1}^k E \left[ E \left[ \underbrace{f \left( X_{t_i}^{(n)}, \langle X \rangle_{t_i}^{(n)} \right)}_{\mathcal{F}_{t_i}^{(n)}\text{-measurable}} \left( X_{t_{i+1}}^{(n)} - X_{t_i}^{(n)} \right) \middle| \mathcal{F}_{t_i}^{(n)} \right] \middle| \mathcal{F}_s \right] \\ &= \sum_{i=l+1}^k E \left[ f \left( X_{t_i}^{(n)}, \langle X \rangle_{t_i}^{(n)} \right) \underbrace{E \left[ \left( X_{t_{i+1}}^{(n)} - X_{t_i}^{(n)} \right) \middle| \mathcal{F}_{t_i}^{(n)} \right]}_{=0 \text{ since } X_t \text{ is a martingale}} \middle| \mathcal{F}_s \right] = 0 \end{aligned}$$

Hence by Proposition 2.1.4 below it is sufficient to show that  $M_t^{(n)} \rightarrow M_t$  in  $\mathcal{L}^1$  for all  $t \geq 0$ . We know that by definition,

$$M_t^{(n)} \rightarrow M_t, \quad P\text{-a.s. } \forall t \geq 0.$$

Furthermore, for all  $t \geq 0$ ,

$$E \left[ \left( M_t^{(n)} \right)^2 \right] = \sum_{\substack{t_i^{(n)} \in \tau_n \\ t_i^{(n)} \leq t}} E \left[ f^2 \left( X_{t_i}^{(n)}, \langle X \rangle_{t_i}^{(n)} \right) \left( X_{t_{i+1}}^{(n)} - X_{t_i}^{(n)} \right)^2 \right],$$

since all nondiagonal terms vanish by the martingale property of  $X_t$ . Furthermore,

$$\begin{aligned} E \left[ \left( M_t^{(n)} \right)^2 \right] &\leq \sup_{\mathbb{R} \times \mathbb{R}_+} f^2 \sum_{\substack{t_i^{(n)} \in \tau_n \\ t_i^{(n)} \leq t}} E \left[ \underbrace{E \left[ X_{t_{i+1}}^2 - 2X_{t_{i+1}} X_{t_i} + X_{t_i}^2 \middle| \mathcal{F}_{t_i}^{(n)} \right]}_{=E \left[ X_{t_{i+1}}^2 \middle| \mathcal{F}_{t_i}^{(n)} \right] - X_{t_i}^2} \right] \\ &= \sup_{\mathbb{R} \times \mathbb{R}_+} f^2 \sum_{\substack{t_i^{(n)} \in \tau_n \\ t_i^{(n)} \leq t}} E \left[ X_{t_{i+1}}^2 - X_{t_i}^2 \right] \\ &= \sup_{\mathbb{R} \times \mathbb{R}_+} f^2 E \left[ \underbrace{X_{t_{N_n}}^2 - X_0^2}_{\leq \sup_{t, \omega} X_t^2(\omega)} \right]. \end{aligned}$$

Hence,  $\sup_n E \left[ \left( M_t^{(n)} \right)^2 \right] < \infty$  and thus  $\left( M_t^{(n)} \right)_{n \in \mathbb{N}}$  is uniformly integrable. By the generalized Lebesgue dominated convergence theorem the claim follows. Therefore  $(M_t)_{t \geq 0}$  is a martingale. It still remains to show that  $M_t$  has  $P$ -a.s. continuous sample paths:

In order to see that consider

$$\tilde{M}_t^{(n)} := \sum_{i=0}^n f \left( X_{t_i}^{(n)}, \langle X \rangle_{t_i}^{(n)} \right) \left( X_{t_{i+1} \wedge t}^{(n)} - X_{t_i \wedge t}^{(n)} \right).$$

Then as above one shows that  $\tilde{M}_t(\omega)$  is a martingale. Note that  $\left( \tilde{M}_t^{(n)} \right)_{t \geq 0}$  has  $P$ -a.s. continuous sample paths and that

$$M_t^{(n)} - \tilde{M}_t^{(n)} \xrightarrow{n \rightarrow \infty} 0 \quad P\text{-a.s.}$$

and in  $\mathcal{L}^q$  (by the same argument as above). Hence,  $\tilde{M}_t^{(n)} \xrightarrow{n \rightarrow \infty} M_t$  in  $\mathcal{L}^q$  for all  $q \in [1, 2)$ . Then by Doob's maximal inequality one can show that  $(M_t)_{t \geq 0}$  has  $P$ -a.s. continuous sample paths

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(cf. below Proposition 2.1.4).

**Step 2** (“Localization by stopping times”):

Let  $(T_n)$  be a localizing sequence for  $X$ . For  $n \in \mathbb{N}$  define

$$\tilde{T}_n := \inf\{t > 0 \mid \langle X \rangle_t > n\} \wedge T$$

and

$$S_n := T_n \wedge \sigma_{G_n^c} \wedge \tilde{T}_n,$$

where  $G_n \nearrow G$ ,  $G_n$  relatively compact and open and  $\bar{G}_n \subset G_{n+1}$  for all  $n \in \mathbb{N}$ . Without loss of generality  $K \subset G_n \forall n \in \mathbb{N}$ . Then  $\sup_n \sigma_{G_n^c} = \sigma_{G^c}$  and  $S_n$  are stopping times such that  $S_n \leq T_n \leq T$  and  $S_n \leq T_n < T$  on  $\{T > 0\}$ . Furthermore,  $\tilde{T}_n \nearrow T$ , hence,  $S_n \nearrow T \wedge \sigma_{G^c}$ . By optional stopping  $(X_{t \wedge S_n})_{t \geq 0}$  is a continuous martingale taking values in  $G_n$ , hence  $(X_{t \wedge S_n})_{t \geq 0}$  is bounded in  $(t, \omega)$ . Furthermore, (exercise)

$$M_{t \wedge S_n}(\omega) \stackrel{!}{=} \int_0^t f(X_{s \wedge S_n}(\omega), \langle X \rangle_{s \wedge S_n}(\omega)) dX_{s \wedge S_n}(\omega). \quad (1.4.13)$$

Take  $\chi_n \in \mathcal{C}_0^2(G \times \mathbb{R}_+)$  such that

$$\chi_n = 1 \text{ on } G_n \times [0, n].$$

Then we can replace  $f$  in (1.4.13) by  $\chi_n f \in \mathcal{C}_b^2(G \times \mathbb{R}_+)$ . Therefore, the representation for  $(M_{t \wedge S_n})_{t \geq 0}$  in (1.4.13) and Step 1 imply that  $(M_{t \wedge S_n})_{t \geq 0}$  is a continuous martingale.  $\square$

**Corollary 1.4.28.** *Let  $X$  be a continuous local martingale up to  $T$  with continuous quadratic variation (later proved to always be the case). Then*

(i)  $X^2 - \langle X \rangle$  is a continuous local martingale up to  $T$ .

(ii) If  $\langle X \rangle = 0$  (which is particularly true if  $X$  has bounded variation), then for  $P$ -a.s.  $\omega \in \Omega$

$$X_t(\omega) \equiv X_0(\omega) \quad \forall t \in [0, T(\omega)[. \quad (1.4.14)$$

*Proof.* Without loss of generality assume that  $X_0 \equiv 0$ . (Otherwise, consider  $X_t - X_0$ ,  $t \leq T$ .)

(i) By Itô

$$X_t^2 = 2 \cdot \int_0^t X_s dX_s + \langle X \rangle_t \quad \text{on } \{t < T\}.$$

But the first term on the right hand side is a continuous local martingale up to  $T$  by Proposition 5.3.9.

(ii) By (i) it also follows that  $X^2$  is a continuous local martingale up to  $T$  if  $\langle X \rangle = 0$ . Hence, if  $T_n \nearrow T$  is a localising sequence for  $X^2$ , then

$$E[X_{t \wedge T_n}^2] = E[X_0^2] = 0 \quad \forall t \geq 0.$$

Therefore,  $1_{\{t < T\}} X_t = 0$   $P$ -a.s.  $\forall t \geq 0$ , (with zero set depending on  $t$ ). Hence,

$$P[1_{\{t < T\}} X_t = 0 \quad \forall t \in \mathbb{Q}] = 1,$$

and by  $P$ -a.s. continuity in  $t$  it follows that  $X_t = 0$  on  $\{t < T\}$  for all  $t \geq 0$   $P$ -a.s..

$\square$

**Proposition 1.4.29** (*d*-dimensional version of Proposition 5.3.9). *Let  $X = (X^1, \dots, X^d)$  with  $X^1, \dots, X^d$  continuous local martingales up to  $T$  such that  $\langle X_i, X_j \rangle$  exist for all  $1 \leq i, j \leq d$  and are continuous up to  $T$ . Let  $F \in \mathcal{C}^2(D)$ ,  $D \subset \mathbb{R}^d$ ,  $D$  open, with  $X_0(\omega) \subset K$  for  $P$ -a.e.  $\omega \in \Omega$  for some compact  $K \subset D$ . Define  $M_0 := 0$  and*

$$M_t := \int_0^t (\nabla F(X_s), dX_s)_{\mathbb{R}^d} \quad \text{on } \{t < T\} \quad (\textit{d-dimensional It\hat{o}-integral})$$

Then  $M$  is a continuous local martingale up to

$$S := T \wedge \underbrace{\inf\{t > 0 \mid X_t \notin D\}}_{=: \sigma_{D^c}}.$$

*Proof.* Exercise (Proceed as for Proposition 5.3.9). □

**Remark 1.4.30.** *By Remark 1.3.22 (iii) for  $M$  as in Proposition 1.4.29 we have*

$$\langle M \rangle_t = \sum_{k,l=1}^d \int_0^t \frac{\partial F}{\partial x_k}(X_s) \frac{\partial F}{\partial x_l}(X_s) d\langle X^k, X^l \rangle_s \quad \text{on } \{t < S\}.$$

Hence, if  $S \equiv \infty$ , then  $M$  is even a martingale provided  $\frac{\partial F}{\partial x_k}$ ,  $1 \leq k \leq d$ , are all bounded and  $X$  is a Brownian motion. This is a consequence of Corollary 1.4.32 below.

**Proposition 1.4.31.** *Let  $M$  be a continuous local martingale up to  $T$  (with continuous  $\langle M \rangle$  up to  $T$ ) and let  $T_0 \leq T$  with  $T_0 < T$  on  $\{T > 0\}$  be a stopping time such that*

$$E[\langle M \rangle_{T_0}] < \infty.$$

Then  $(M_{t \wedge T_0})_{t \geq 0}$  is a continuous martingale such that  $\sup_{t \geq 0} E[M_{T_0 \wedge t}^2] \leq E[\langle M \rangle_{T_0}]$ . Furthermore,

$$E[M_{T_0}] = E[M_0].$$

Note that  $T_0$  doesn't need to be bounded.

*Proof.* Without loss of generality assume  $M_0 \equiv 0$ . (Otherwise, consider  $M_t - M_0$ .) Let  $T_n \nearrow T$  be a localizing sequence for  $M$ . Then  $(M_{T_0 \wedge T_n \wedge t})_{t \geq 0}$  is a continuous martingale. Hence

$$\lim_{n \rightarrow \infty} M_{T_0 \wedge T_n \wedge t} = M_{T_0 \wedge t} \quad P\text{-a.s. } \forall t \geq 0$$

and

$$E[M_{T_0 \wedge T_n \wedge t}^2] \stackrel{\text{Cor. 5.3.10}}{=} E[\langle M \rangle_{T_0 \wedge T_n \wedge t}] \leq E[\langle M \rangle_{T_0}] < \infty.$$

Hence,  $\{M_{T_0 \wedge T_n \wedge t} \mid n \in \mathbb{N}\}$  is uniformly integrable  $\forall t \geq 0$ . In particular, by Lebesgue

$$\lim_{n \rightarrow \infty} M_{T_0 \wedge T_n \wedge t} = M_{T_0 \wedge t} \quad \text{in } \mathcal{L}^1 \forall t \geq 0.$$

Therefore,  $(M_{T_0 \wedge t})_{t \geq 0}$  is a continuous martingale. In addition,

$$E[M_{T_0 \wedge t}^2] \leq \liminf_{n \rightarrow \infty} E[M_{T_0 \wedge T_n \wedge t}^2] \leq E[\langle M \rangle_{T_0}] < \infty \quad \forall t \geq 0.$$

So, by the  $\mathcal{L}^2$ -martingale convergence theorem

$$\lim_{t \rightarrow \infty} M_{T_0 \wedge t} = M_{T_0} \quad \text{in } \mathcal{L}^2,$$

In particular,

$$E[M_{T_0}] = \lim_{t \rightarrow \infty} E[M_{T_0 \wedge t}] = E[M_0] = 0.$$

□

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**Corollary 1.4.32.** *Let  $M$  be a continuous local martingale up to  $T \equiv \infty$  (with continuous  $\langle M \rangle$ ) such that*

$$E[\langle M \rangle_t] < \infty \quad \forall t \geq 0.$$

*Then  $M$  is a square integrable martingale.*

**Proposition 1.4.33.** *Let  $T$  be a stopping time and  $X, A$  continuous processes with continuous  $\langle X \rangle$  and  $\langle A \rangle = 0$  up to  $T$ . Assume  $A_0 = 0$  and  $X_0 = 0$ . Then the following are equivalent:*

- (i)  $X$  is a local martingale up to  $T$  with  $\langle X \rangle = A$ .
- (ii) For all  $\alpha \geq 0$

$$G_t^\alpha := \begin{cases} 1 & \text{if } t = 0 \\ \exp[\alpha X_t - \frac{1}{2}\alpha^2 A_t] & \text{on } \{t < T\} \end{cases}$$

*is a local martingale up to  $T$ .*

*Proof.* (i)  $\Rightarrow$  (ii): By Itô's formula for time dependent functions (cf. Proposition 1.3.24)

$$G_t^\alpha = G_0^\alpha + \int_0^t \alpha G_s^\alpha dX_s + 0.$$

But the right hand side is a local martingale up to  $T$  by Proposition 5.3.9 for  $f(X_t, \langle X \rangle_t) = \exp(\alpha X_t - \frac{1}{2}\alpha^2 \langle X \rangle_t)$ .

(ii)  $\Rightarrow$  (i): By Itô's formula we get

$$\begin{aligned} X_t &= \frac{1}{\alpha} \left( \log G_t^\alpha - \underbrace{\log G_0^\alpha}_{=0} \right) + \frac{1}{2} \alpha A_t \\ &\stackrel{\text{Itô}}{=} \frac{1}{\alpha} \left( \int_0^t \frac{1}{G_s^\alpha} dG_s^\alpha - \frac{1}{2} \int_0^t \frac{1}{(G_s^\alpha)^2} d\langle G^\alpha \rangle_s \right) + \frac{1}{2} \alpha A_t \\ &\stackrel{1.3.22(ii)}{=} \frac{1}{\alpha} \left( \int_0^t \frac{1}{G_s^\alpha} dG_s^\alpha - \frac{1}{2} \int_0^t \frac{1}{(G_s^\alpha)^2} (\alpha G_s^\alpha)^2 d\langle X \rangle_s \right) + \frac{1}{2} \alpha A_t \\ &= \frac{1}{\alpha} \int_0^t \frac{1}{G_s^\alpha} dG_s^\alpha + \frac{\alpha}{2} (A_t - \langle X \rangle_t) \quad \forall \alpha \in \mathbb{R} \setminus \{0\}. \end{aligned} \tag{1.4.15}$$

Here,  $\frac{1}{\alpha} \int_0^t \frac{1}{G_s^\alpha} dG_s^\alpha$  is a local martingale up to  $T$  by Proposition 5.3.9 with  $G = ]0, \infty[$ . Consider  $\alpha \neq \alpha'$  and take the difference of the two corresponding equalities to get

$$0 = M + \frac{\alpha - \alpha'}{2} (A_t - \langle X \rangle_t),$$

where  $M$  is a local martingale up to  $T$ . By Corollary 5.3.10(ii) it follows that  $A_t - \langle X \rangle_t = A_0 - \langle X \rangle_0 = 0$  on  $\{t < T\}$   $P$ -a.s., and then (1.4.15) implies that  $X$  is a local martingale up to  $T$ .  $\square$

## 1.5. Levy's Characterization of Brownian Motion

Let  $X = (X_t)_{t \geq 0}$  be a stochastic process on  $(\Omega, \mathcal{F}, P)$  with continuous sample paths (and continuous quadratic variation  $\langle X \rangle$ ).

**Proposition 1.5.34** (Levy). *Assume that  $X$  is a continuous local martingale up to  $\infty$  with respect to some filtration  $(\mathcal{F}_t)_{t \geq 0}$  such that  $X$  is  $(\mathcal{F}_t)$ -adapted. If  $\langle X \rangle_t = t$  holds for all  $t \geq 0$ , then  $X$  is a Brownian motion.*

**Remark 1.5.35.** Every continuous martingale can be transformed by a time change into a Brownian motion provided its quadratic variation is strictly increasing (see later). Therefore, Brownian motion is the only local martingale, where the quadratic variation is the usual time.

*Proof of 1.5.34.* Corollary 1.4.32 implies that  $X$  is a martingale. For  $u \in \mathbb{R}$  we apply the Itô-formula to

$$F(x) = e^{iux} = \cos(ux) + i \sin(ux), \quad x \in \mathbb{R}.$$

Then, for all  $s < t$

$$e^{iuX_{t \wedge T_n}} - e^{iuX_{s \wedge T_n}} = \int_0^{t \wedge T_n} iue^{iuX_r} dX_r - \int_0^{s \wedge T_n} iue^{iuX_r} dX_r + \frac{1}{2} \int_{s \wedge T_n}^{t \wedge T_n} (-u^2) e^{iuX_r} \underbrace{d\langle X \rangle_r}_{= dr},$$

where  $T_n \nearrow \infty$  are stopping times such that

$$\left( \int_0^{t \wedge T_n} iue^{iuX_r} dX_r \right)_{t \geq 0}$$

is a martingale for all  $n$ . Now take  $E[\cdot | \mathcal{F}_s]$  of the equality above:

$$E[e^{iuX_{t \wedge T_n}} - e^{iuX_{s \wedge T_n}} | \mathcal{F}_s] = \frac{1}{2} E \left[ \int_{s \wedge T_n}^{t \wedge T_n} (-u^2) e^{iuX_r} dr | \mathcal{F}_s \right].$$

Hence, letting  $n \rightarrow \infty$  we obtain

$$\frac{1}{2} E \left[ \int_s^t (-u^2) e^{iuX_r} dr | \mathcal{F}_s \right] = E[e^{iuX_t} - e^{iuX_s} | \mathcal{F}_s].$$

Multiplication by  $e^{-iuX_s}$  yields

$$E[e^{iu(X_t - X_s)} | \mathcal{F}_s] - 1 = \frac{1}{2} E \left[ \int_s^t (-u^2) e^{iu(X_r - X_s)} dr | \mathcal{F}_s \right].$$

Therefore, for all  $A \in \mathcal{F}_s$

$$\underbrace{E[e^{iu(X_t - X_s)}, A]}_{=: \varphi(t)} - P(A) \stackrel{\text{Fubini}}{=} -\frac{1}{2} u^2 \int_s^t \underbrace{E[e^{iu(X_r - X_s)}, A]}_{=: \varphi(r)} dr.$$

Then,  $\varphi \in \mathcal{C}^1$  (since the right hand side is  $\mathcal{C}^1$ ) and by differentiation

$$\dot{\varphi}(t) = -\frac{1}{2} u^2 \varphi(t) \quad \forall t \geq s.$$

Solving this equation we get

$$\varphi(t) = C \cdot e^{-\frac{1}{2} u^2 t}, \quad t \geq s.$$

Substituting  $s$  for  $t$  yields

$$P(A) = \varphi(s),$$

which implies

$$C = P(A) e^{\frac{1}{2} u^2 s},$$

hence,

$$E[e^{iu(X_t - X_s)}; A] = e^{-\frac{1}{2} u^2 (t-s)} P(A) \quad \forall A \in \mathcal{F}_s. \quad (1.5.16)$$

Taking  $A := \Omega$  implies that  $X_t - X_s$  is  $N(0, (t-s))$  distributed by uniqueness of the Fourier-transform. By a monotone class argument, (1.5.16) implies that  $X_t - X_s$  is independent of  $\mathcal{F}_s$  (Exercise!). In particular,

$$X_{t_n} - X_{t_{n-1}}, \dots, X_{t_2} - X_{t_1}$$

are independent for all  $0 \leq t_1 < t_2 < \dots < t_n < \infty$ . □



## 2. (Semi-)Martingales and Stochastic Integration

In this chapter we want to define

$$\int g(t, \cdot) dM_t,$$

where  $g$  is defined “more general” and  $M_t$  is an arbitrary semimartingale. This is more general than so far, since we could only define

$$\int f(t, \mu_t) dM_t, \quad \forall f \in \mathcal{C}^1.$$

### 2.1. Review of Some Facts from Martingale Theory

Fix a probability space  $(\Omega, \mathcal{F}, P)$  and let  $(\mathcal{F}_t)_{t \geq 0}$  be a right-continuous filtration. (Sometimes assume in addition that  $N \in \mathcal{F}$ ,  $P(N) = 0 \Rightarrow N \in \mathcal{F}_0$ .) These conditions are also called “usual conditions”.

**Proposition 2.1.1.** *Let  $(M_t)_{t \geq 0}$  be a martingale. Then there exists a version*

$$(\tilde{M})_{t \geq 0} \quad (\text{i.e. } M_t = \tilde{M}_t \text{ on } N(t)^c \text{ such that } P(N(t)) = 0),$$

*such that  $t \mapsto \tilde{M}_t(\omega)$  is càdlàg (i.e. is right-continuous and has left limits) for all  $\omega \in \Omega$ .*

*Proof.* Doob’s upcrossing lemma (cf. [vWW90]). □

**Proposition 2.1.2** (Optional stopping theorem). *Let  $(M_t)_{t \geq 0}$  be a martingale and  $T$  be a stopping time. Then  $(M_{t \wedge T})_{t \geq 0}$  is a martingale.*

*Proof.* See [Röc11, Proposition VIII.4.2, Corollary VIII.4.3]. □

**Proposition 2.1.3** (Doob’s inequality). *Let  $p > 1$ . Then*

$$\left\| \sup_{s \leq t} |M_s| \right\|_p \leq \frac{p}{p-1} \|M_t\|_p,$$

where  $\|\cdot\|_p = \|\cdot\|_{\mathcal{L}^p}$ . In particular, if

$$M^* := \sup_{t \geq 0} |M_t|,$$

then

$$\|M^*\|_p \leq \frac{p}{p-1} \sup_{t \geq 0} \|M_t\|_p.$$

**Proposition 2.1.4.** *Let  $p \geq 1$  and let  $(M_t^{(n)})_{t \geq 0}$ ,  $n \in \mathbb{N}$ , be a sequence of martingales such that*

$$M_t^{(n)} \xrightarrow{n \rightarrow \infty} M_t \quad \text{in } \mathcal{L}^p, \quad \forall t \geq 0,$$

Then

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(i)  $(M_t)_{t \geq 0}$  is again a martingale in  $\mathcal{L}^p$ .

(ii) If  $p > 1$  and  $(\mathcal{F})_{t \geq 0}$  is such that all  $P$ -zero sets in  $\mathcal{F}_t$  are in  $\mathcal{F}_0$  and each  $(M_t^{(n)})_{t \geq 0}$  has  $P$ -a.s. (right-) continuous ((càdlàg)) sample paths, then  $(M_t)_{t \geq 0}$  has a (right-) continuous ((càdlàg))  $(\mathcal{F}_t)$ -adapted version again denoted by  $(M_t)_{t \geq 0}$  and

$$M_t^{(n)} \xrightarrow{n \rightarrow \infty} M_t$$

in  $\mathcal{L}^p(\Omega, \mathcal{F}, P; C([0, t]; \mathbb{R}))$  and hence locally uniformly in  $t$  and in  $\mathcal{L}^p$  and has a locally uniformly in  $t$   $P$ -a.s. convergent subsequence.

*Proof.* (i) Obvious.

(ii) Fix  $t > 0$ . Since  $M^n - M^m$  is a martingale, we have by Doob

$$\left\| \sup_{s \leq t} |M_s^{(n)} - M_s^{(m)}| \right\|_p \leq \frac{p}{p-1} \left\| M_t^{(n)} - M_t^{(m)} \right\|_p.$$

Since they are  $\mathcal{L}^p$ -convergent, for some subsequence  $(n_k)_{k \in \mathbb{N}}$

$$\sum_{k=1}^{\infty} \left\| \sup_{s \leq t} |M_s^{(n_{k+1})} - M_s^{(n_k)}| \right\|_p \leq \frac{p}{p-1} \sum_{k=1}^{\infty} \left\| M_t^{(n_{k+1})} - M_t^{(n_k)} \right\|_p < \infty.$$

Therefore,

$$P \left( \sum_{k=1}^{\infty} \sup_{s \leq t} |M_s^{(n_{k+1})} - M_s^{(n_k)}| < \infty \right) = 1.$$

Hence,  $M_s^{(n_k)} \xrightarrow{k \rightarrow \infty} M_s$  uniformly on  $[0, t]$   $P$ -a.s. (cf. Proof of Riesz-Fischer!). Then up to a  $P$ -zero set  $M_s$  is  $\mathcal{F}_s$ -measurable, since so is each  $M_s^{(n_k)}$ . But  $\mathcal{F}_s$  contains all  $P$ -zero sets in  $\mathcal{F}_0$ . Therefore,  $M_s$  is  $\mathcal{F}_s$ -measurable. □

**Remark 2.1.5** (Localization). Let  $(M_t)_{t \geq 0}$  be a continuous local martingale (up to  $\infty$ ) such that  $M_0 = 0$ . Set

$$R_n(\omega) := \inf \{ t > 0 \mid |M_t(\omega)| > n \} \xrightarrow{n \rightarrow \infty} \infty.$$

**Claim:** For all  $n \in \mathbb{N}$   $(M_{t \wedge R_n})_{t \geq 0}$  is a continuous bounded martingale.

*Proof.* Let  $(T_k)_{k \in \mathbb{N}}$  be a localizing sequence for  $M$ , such that  $T_k \leq k$ . (Otherwise consider  $T_k \wedge k$ .) Then for  $s \leq t$  and  $n$  fixed

$$E[M_{t \wedge R_n \wedge T_k}; A_s] = E[M_{s \wedge R_n \wedge T_k}; A_s] \quad \forall A_s \in \mathcal{F}_s.$$

By Lebesgue, for all  $A_s \in \mathcal{F}_s$ ,

$$E[M_{t \wedge R_n \wedge T_k}; A_s] \xrightarrow{k \rightarrow \infty} E[M_{t \wedge R_n}; A_s]$$

and, since  $|M_{t \wedge R_n \wedge T_k}| \leq n$  for fixed  $n$ ,

$$E[M_{s \wedge R_n \wedge T_k}; A_s] \xrightarrow{k \rightarrow \infty} E[M_{s \wedge R_n}; A_s].$$

So, we get  $E[M_{t \wedge R_n}; A_s] = E[M_{s \wedge R_n}; A_s]$ . □

## 2.2. Quadratic Variation and Covariation for Continuous Local Martingales

Let  $(M_t)_{t \geq 0}$  be a càdlàg martingale.

Approach to stochastic integration:

Assume initially  $M_t \in \mathcal{L}^2$  for all  $t \geq 0$ . (If  $M_t \notin \mathcal{L}^2$ , then localize.) By Jensen's inequality  $(M_t^2)_{t \geq 0}$  is a submartingale. Then one can show (by the Doob-Meyer decomposition) that there exists a unique adapted process  $(\langle M \rangle_t)_{t \geq 0}$  with  $\langle M \rangle_0 = 0$ , increasing, right-continuous, predictable (see below) such that

$$(M_t^2 - \langle M \rangle_t)_{t \geq 0} \quad (\star)$$

is a martingale. Then,  $\langle M \rangle$  is the variance process of  $M$ , i.e.

$$\begin{aligned} & E[(M_t - M_s - \overbrace{E[M_t - M_s]}^{=0})^2 | \mathcal{F}_s] \quad (\text{conditioned variance of } M_t - M_s \text{ given } \mathcal{F}_s) \\ &= E[M_t^2 - M_s^2 | \mathcal{F}_s] \stackrel{(\star)}{=} E[\langle M \rangle_t - \langle M \rangle_s | \mathcal{F}_s]. \end{aligned}$$

In [Röc11] we proved the Doob-Meyer decomposition in discrete time. In continuous time for càdlàg martingales this is much more difficult (cf. [Kry]).

We are going to prove the Doob-Meyer decomposition only for *continuous* martingales. In the stochastic integration theory below, we shall, however, allow càdlàg martingales, simply assuming Doob-Meyer without proof, and referring to [vWW90, Cor. 6.6.3] instead.

For a continuous martingale  $M$  we shall construct the process  $\langle M \rangle$  such that it pathwise coincides  $P$ -a.s. with the quadratic variation of  $M$  along  $(\tau_n)_{n \in \mathbb{N}}$  of the previous chapter. Let  $M$  be a continuous local martingale (up to  $\infty$ ) and let  $(\tau_n)_{n \in \mathbb{N}}$  be a sequence of partitions of  $[0, \infty)$  such that

$$|\tau_n| \xrightarrow{n \rightarrow \infty} 0$$

and

$$t_{N_n}^{(n)} \xrightarrow{n \rightarrow \infty} \infty.$$

**Definition 2.2.6.** *Let*

$$V_t^{(n)} := \sum_{s \in \tau_n} (M_{s' \wedge t} - M_{s \wedge t})^2, \quad n \in \mathbb{N}, t \geq 0$$

*be the quadratic variation of  $M$  along  $\tau_n$  on  $[0, t]$ . Here,  $s'$  denotes the successor of  $s$  in  $\tau_n$ .*

*Note that  $t \mapsto V_t^{(n)}$  is  $P$ -a.s. continuous for all  $n \in \mathbb{N}$ .*

**Remark 2.2.7.** *Since  $(a - b)^2 = a^2 - 2b(a - b) - b^2$ ,*

$$\begin{aligned} V_t^{(n)} &= M_t^2 - M_0^2 - 2 \sum_{u \in \tau_n} M_{u \wedge t} (M_{u' \wedge t} - M_{u \wedge t}) \\ &= M_t^2 - M_0^2 - 2 \underbrace{\sum_{u \in \tau_n, u \leq t} M_u (M_{u' \wedge t} - M_u)}_{\text{local martingale}}. \end{aligned}$$

**Proposition 2.2.8.** *Assume that  $\mathcal{F}_0$  contains all  $P$ -zero sets. Let  $M$  be a continuous (for all  $\omega \in \Omega$ ), local martingale (up to  $\infty$ ). Fix  $(\tau_n)$  as above and let  $(V^{(n)})_{n \in \mathbb{N}}$  be as in Definition 2.2.6. Then there exists a continuous increasing  $(\mathcal{F}_t)$ -adapted process  $\langle M \rangle$  with  $\langle M \rangle_0 = 0$  such that*

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(i)  $\langle M \rangle_t = \lim_{n \rightarrow \infty} V_t^{(n)}$  in  $P$ -measure locally uniformly in  $t \geq 0$ .  
(Convergence is even in  $\mathcal{L}^2(\Omega, \mathcal{F}, P; \mathcal{C}([0, t]; \mathbb{R}))$  for all  $t \geq 0$ , if  $M$  is a bounded martingale.)

(ii)  $M^2 - \langle M \rangle$  is a continuous local martingale.

(iii) If  $M_t \in \mathcal{L}^2$  for all  $t \geq 0$  and if  $M_t$  is a martingale, then  $M_t^2 - \langle M \rangle_t$  is a martingale. In particular,

$$E[M_t^2] = E[M_0^2] + E[\langle M \rangle_t] \quad \forall t \geq 0.$$

(Warning: At this stage  $\langle M \rangle = (\langle M \rangle_t)_{t \geq 0}$  depends on  $(\tau_n)$ !)

*Proof.* (i) **Case 1:**  $M$  is bounded (i.e.  $\sup_{t \geq 0, \omega \in \Omega} |M_t(\omega)| < \infty$ ) and a martingale.

**Claim:**  $(V_t^{(n)})_{n \in \mathbb{N}}$  is a Cauchy sequence in  $\mathcal{L}^2$  for all  $t \geq 0$ .

Take  $m, n \geq N$ ,  $N$  large so that  $t \leq t_{N_n}^{(n)}, t \leq t_{N_m}^{(m)}$ . Define

$$V_t^{(m,n)} := M_t^2 - M_0^2 - 2 \sum_{s \in \tau_n \cup \tau_m} M_s (M_{s' \wedge t} - M_{s \wedge t}) \quad (= V_t^{(n,m)}).$$

Here,  $s'$  denotes the successor of  $s$  in  $\tau_n \cup \tau_m$ . Then by Remark 2.2.7

$$V_t^{(m,n)} - V_t^{(n)} = -2 \sum_{u \in \tau_n} \left( \sum_{\substack{u \leq s < u' \\ s \in \tau_n \cup \tau_m}} M_s (M_{s' \wedge t} - M_{s \wedge t}) \right) - M_u (M_{u' \wedge t} - M_{u \wedge t}),$$

where  $u'$  is the successor in  $\tau_n$  and  $s'$  the one in  $\tau_n \cup \tau_m$ . Using the fact that

$$M_{u' \wedge t} - M_{u \wedge t} = \sum_{\substack{u \leq s < u' \\ s \in \tau_n \cup \tau_m}} (M_{s' \wedge t} - M_{s \wedge t})$$

we conclude that

$$V_t^{(m,n)} - V_t^{(n)} = -2 \sum_{u \in \tau_n} \sum_{\substack{u \leq s < u' \\ s \in \tau_n \cup \tau_m}} (M_s - M_u) (M_{s' \wedge t} - M_{s \wedge t}).$$

Hence,

$$E \left[ \left( V_t^{(m,n)} - V_t^{(n)} \right)^2 \right] = 4E \left[ \sum_{u \in \tau_n} \sum_{\substack{u \leq s < u' \\ s \in \tau_n \cup \tau_m}} (M_s - M_u)^2 (M_{s' \wedge t} - M_{s \wedge t})^2 \right]$$

Note that all occurring mixed terms are zero by the martingale property (exercise!). Furthermore,

$$E \left[ \left( V_t^{(m,n)} - V_t^{(n)} \right)^2 \right] \leq 4E \left[ \Delta_N \sum_{s \in \tau_n \cup \tau_m} (M_{s' \wedge t} - M_{s \wedge t})^2 \right],$$

where

$$\Delta_N := \sup_{\tilde{n} \geq N} \sup_{\tilde{m}} \sup \{ (M_s - M_u)^2 \mid s \in \tau_{\tilde{n}} \cup \tau_{\tilde{m}}, u \in \tau_{\tilde{n}}, u \leq s \leq u', s, u \leq t \} \xrightarrow{N \rightarrow \infty} 0 \text{ } P\text{-a.s.}$$

## 2.2. Quadratic Variation and Covariation for Continuous Local Martingales

is bounded in  $N$  and  $\omega$ , since  $M$  is bounded. Then, by Cauchy-Schwarz (since  $n, m \geq N$ )

$$E \left[ \left( V_t^{(m,n)} - V_t^{(n)} \right)^2 \right] \leq 4 \left( E \left[ \Delta_N^2 \right] \right)^{1/2} \left\| \sum_{s \in \tau_n \cup \tau_m} (M_{s' \wedge t} - M_{s \wedge t})^2 \right\|_2.$$

By Lebesgue's dominated convergence theorem,

$$\left( E \left[ \Delta_N^2 \right] \right)^{1/2} = \|\Delta_N\|_2 \xrightarrow{N \rightarrow \infty} 0,$$

since  $M(\omega)$  is uniformly continuous on  $[0, t]$  for all  $t$  and  $(M_t)_{t \geq 0}$  is bounded in  $(\omega, t)$ . But for  $c := \sup_{t, \omega} |M_t(\omega)|$  and for all  $n, m \in \mathbb{N}$

$$\begin{aligned} & E \left[ \left( \sum_{s \in \tau_n \cup \tau_m} (M_{s' \wedge t} - M_{s \wedge t})^2 \right)^2 \right] \\ = & E \left[ \sum_{s \in \tau_n \cup \tau_m} (M_{s' \wedge t} - M_{s \wedge t})^4 \right] \\ & + 2E \left[ \sum_{s \in \tau_n \cup \tau_m} \sum_{\substack{u \in \tau_n \cup \tau_m \\ s < u}} (M_{s' \wedge t} - M_{s \wedge t})^2 (M_{u' \wedge t} - M_{u \wedge t})^2 \right] \\ \leq & 4c^2 E \left[ \underbrace{\sum_{s \in \tau_n \cup \tau_m} (M_{s' \wedge t} - M_{s \wedge t})^2}_{=: S_1} \right] \\ & + 2 \underbrace{\sum_{s \in \tau_n \cup \tau_m} \sum_{\substack{u \in \tau_n \cup \tau_m \\ s < u}} E \left[ (M_{s' \wedge t} - M_{s \wedge t})^2 \cdot E \left[ (M_{u' \wedge t} - M_{u \wedge t})^2 | \mathcal{F}_{u \wedge t} \right] \right]}_{=: S_2} \end{aligned}$$

and

$$\begin{aligned} S_1 &= \sum_{s \in \tau_n \cup \tau_m} E \left[ M_{s' \wedge t}^2 - 2M_{s' \wedge t} M_{s \wedge t} + M_{s \wedge t}^2 \right] \\ &= \sum_{s \in \tau_n \cup \tau_m} E \left[ M_{s' \wedge t}^2 + M_{s \wedge t}^2 - 2E \left[ M_{s' \wedge t} M_{s \wedge t} | \mathcal{F}_s \right] \right] \\ &= \sum_{s \in \tau_n \cup \tau_m} E \left[ M_{s' \wedge t}^2 + M_{s \wedge t}^2 - 2M_{s \wedge t} \underbrace{E \left[ M_{s' \wedge t} | \mathcal{F}_s \right]}_{M_{s \wedge t}} \right] \\ &= \sum_{s \in \tau_n \cup \tau_m} E \left[ M_{s' \wedge t}^2 - M_{s \wedge t}^2 \right] = E \left[ M_t^2 - M_0^2 \right] \end{aligned}$$

In addition,

$$\begin{aligned} S_2 &= \sum_{\substack{u \in \tau_n \cup \tau_m \\ s < u}} E \left[ (M_{s' \wedge t} - M_{s \wedge t})^2 \cdot E \left[ M_{u' \wedge t}^2 - M_{u \wedge t}^2 | \mathcal{F}_{u \wedge t} \right] \right] \\ &= \sum_{\substack{u \in \tau_n \cup \tau_m \\ s < u}} E \left[ (M_{s' \wedge t} - M_{s \wedge t})^2 \cdot (M_{u' \wedge t}^2 - M_{u \wedge t}^2) \right] \\ &= E \left[ (M_{s' \wedge t} - M_{s \wedge t})^2 (M_t^2 - M_{s' \wedge t}^2) \right] \\ &\leq 2c^2 E \left[ E \left[ (M_{s' \wedge t} - M_{s \wedge t})^2 | \mathcal{F}_{s \wedge t} \right] \right] = 2c^2 E \left[ M_{s' \wedge t}^2 - M_{s \wedge t}^2 \right]. \end{aligned}$$

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Thus,

$$\begin{aligned} & E \left[ \left( \sum_{s \in \tau_n \cup \tau_m} (M_{s' \wedge t} - M_{s \wedge t})^2 \right)^2 \right] \\ & \leq 4c^2 E[M_t^2 - M_0^2] + 4c^2 \sum_{s \in \tau_n \cup \tau_m} E[M_{s' \wedge t}^2 - M_{s \wedge t}^2] \\ & \leq 8c^2 E[M_t^2 - M_0^2] \leq 16c^4. \end{aligned}$$

Alltogether, we obtain for all  $n, m \geq N$

$$\begin{aligned} \left( E \left[ (V_t^{(n)} - V_t^{(m)})^2 \right] \right) & \leq 2E \left[ (V_t^{(n)} - V_t^{(n,m)})^2 \right] + 2E \left[ (V_t^{(m,n)} - V_t^{(m)})^2 \right] \\ & \leq 8 \|\Delta_N\|_2 4c^2 + 8 \|\Delta_N\|_2 4c^2 \xrightarrow{N \rightarrow \infty} 0 \end{aligned}$$

and the claim is proved.

Let  $V_t := \mathcal{L}^2\text{-}\lim_{n \rightarrow \infty} V_t^{(n)}$ ,  $t \geq 0$ . Define

$$Y_t^{(n)} := 2 \sum_{s \in \tau_n} M_s (M_{s' \wedge t} - M_{s \wedge t}).$$

Then each  $(Y_t^{(n)})_{t \geq 0}$  is an  $(\mathcal{F}_t)$ -martingale and by Remark 2.2.7 and the above

$$Y_t^{(n)} = M_t^2 - M_0^2 - V_t^{(n)} \xrightarrow{n \rightarrow \infty} M_t^2 - M_0^2 - V_t =: Y_t \quad \text{in } \mathcal{L}^2. \quad (**)$$

Hence, by Proposition 2.1.4  $Y_t$  is a martingale and has a  $P$ -a.s. continuous version and, therefore,  $V_t$  has a continuous version  $\langle M \rangle$  and  $\langle M \rangle$  is  $(\mathcal{F}_t)$ -adapted since all  $P$ -zero sets in  $\mathcal{F}$  are in  $\mathcal{F}_0$ , hence in every  $\mathcal{F}_t$ . Furthermore, by Proposition 2.1.4(ii),  $Y^{(n)} \rightarrow Y$  as  $n \rightarrow \infty$  in  $\mathcal{L}^2(\Omega, \mathcal{F}, P; C([0, t]; \mathbb{R}))$ , hence  $V^{(n)} \rightarrow \langle M \rangle$  as  $n \rightarrow \infty$  in  $\mathcal{L}^2(\Omega, \mathcal{F}, P; C([0, t]; \mathbb{R}))$ , i.e.

$$E \left[ \sup_{s \leq t} |V_s^{(n)} - \langle M \rangle_s|^2 \right] \xrightarrow{n \rightarrow \infty} 0.$$

**Case 2:**  $M$  as in the assertion.

Without loss of generality  $M_0 \equiv 0$ . (Otherwise consider  $(M_t - M_0)_{t \geq 0}$ .) Let  $R_k$  be as in 2.1.5, i.e.

$$R_k := \inf\{t > 0 \mid |M_t| > k\}$$

and therefore,  $R_k$  is a stopping time. Define

$$M_t^k := M_{t \wedge R_k}, \quad t \geq 0.$$

By Remark 2.1.5  $M_t^k$  is a bounded martingale. Hence, by Case 1 there exists  $V_t^k := \langle M^k \rangle_t$ ,  $t \geq 0$ . Let  $(V_t^k)^{(n)}$ ,  $t > 0$ , be the corresponding approximations along  $(\tau_n)$ . Then there exists a subsequence  $(n_l)_{l \in \mathbb{N}}$  and  $\Omega_0 \in \mathcal{F}_0$ ,  $P(\Omega_0) = 1$  such that for all  $k \in \mathbb{N}$  (cf. Proposition 2.1.4(ii))

$$(V_s^k)^{(n_l)}(\omega) \xrightarrow{l \rightarrow \infty} \langle M^k \rangle_s(\omega)$$

locally uniformly for  $s \in [0, t]$  and for  $P$ -a.e.  $\omega \in \Omega_0$ . But

$$1_{\{t \leq R_k\}} V_t^{(n_l)} = 1_{\{t \leq R_k\}} V_{t \wedge R_k}^{(n_l)} = 1_{\{t \leq R_k\}} (V_t^k)^{(n_l)} \xrightarrow{l \rightarrow \infty} 1_{\{t \leq R_k\}} \langle M^k \rangle_t \quad P - \text{a.s.} \quad (***)$$

Hence, we can (well-)define

$$\langle M \rangle_t(\omega) := \begin{cases} \langle M^k \rangle_t(\omega), & \text{with } k \in \mathbb{N} \text{ such that } \omega \in \{t \leq R_k\} \cap \Omega_0, \\ 0, & \text{otherwise.} \end{cases}$$

Recall that by the continuity, hence local boundedness of  $t \rightarrow M_t(\omega)$  we have that  $R_k(\omega) \nearrow \infty \forall \omega \in \Omega$ , hence  $\Omega_0 = \bigcup_{k \in \mathbb{N}} \{t < R_k\} \cap \Omega_0$ , so  $\langle M \rangle_t(\omega)$  is defined  $\forall \omega \in \Omega$  by the above and by (\*\*\*) this definition is independent of  $k$ . Then  $\langle M \rangle$  is  $P$ -a.s. continuous and  $(\mathcal{F}_t)$ -adapted.

Furthermore, since  $R_k \rightarrow \infty$  as  $k \rightarrow \infty$ , for  $\delta > 0$  we can fix  $k = k(\delta)$  large enough, such that

$$P[R_k < t] < \delta.$$

Then

$$\begin{aligned} & P\left(\sup_{0 \leq s \leq t} |V_s^{(n)} - \langle M \rangle_s| \geq \varepsilon\right) \\ &= P\left(\sup_{0 \leq s \leq t} |V_s^{(n)} - \langle M \rangle_s| \geq \varepsilon, R_k < t\right) + P\left(\sup_{0 \leq s \leq t} |V_s^{(n)} - \langle M \rangle_s| \geq \varepsilon, R_k \geq t\right) \\ &\leq P(R_k < t) + P\left(\sup_{0 \leq s \leq t} |(V_s^k)^{(n)} - \langle M^k \rangle_s| \geq \varepsilon, R_k \geq t\right) \\ &\leq P(R_k < t) + P\left(\sup_{0 \leq s \leq t} |(V_s^k)^{(n)} - \langle M^k \rangle_s| \geq \varepsilon\right). \end{aligned}$$

But by the first part of the proof and Chebychev's inequality

$$P\left(\sup_{0 \leq s \leq t} |(V_s^k)^{(n)} - \langle M^k \rangle_s| \geq \varepsilon\right) \xrightarrow{n \rightarrow \infty} 0.$$

Therefore,

$$\limsup_{n \rightarrow \infty} P\left(\sup_{0 \leq s \leq t} |V_s^{(n)} - \langle M \rangle_s| \geq \varepsilon\right) \leq P[R_k < t] < \delta.$$

It remains to show that  $t \mapsto \langle M \rangle_t$  is increasing  $P$ -a.s.. But for fixed  $t \geq 0$  we have that

$$V_t^{(n)} = \sum_{s \in \tau_n, s \leq t} (M_{s'} - M_s)^2 + N_t^{(n)} \xrightarrow{n \rightarrow \infty} \langle M \rangle_t,$$

where the sum is increasing in  $t$  and

$$N_t^{(n)} := (M_t - M_{\delta_t^{(n)}})^2 \xrightarrow{n \rightarrow \infty} 0$$

and

$$\delta_t^{(n)} := \sup\{s \in \tau_n | s \leq t\}.$$

Hence, (i) is completely proved.

(ii) Continuity is clear. By (\*\*) on p. 28 we know, that  $P$ -a.s.

$$M_{t \wedge R_k}^2 - \langle M \rangle_{t \wedge R_k} = (M_t^k)^2 - \langle M^k \rangle_t = Y_t^k + M_0^2$$

and the right hand side is a martingale. Hence,  $M^2 - \langle M \rangle$  is a local martingale for  $T$  up to  $\infty$ .

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(iii) Without loss of generality  $M_0 = 0$ . We have to show that  $M_T^2 - \langle M \rangle_T \in \mathcal{L}^1(P)$  and that

$$E[\langle M \rangle_T] = E[M_T^2]$$

for all bounded stopping times  $T$ . By the monotone convergence theorem, we get that

$$E[\langle M \rangle_T] = \lim_{n \rightarrow \infty} E[\langle M \rangle_{T \wedge R_n}] \stackrel{\text{proof of (ii)}}{=} \lim_{n \rightarrow \infty} E[M_{T \wedge R_n}^2] \stackrel{\text{Fatou}}{\geq} E[M_T^2].$$

On the other hand, by the submartingale property,  $\lim_{n \rightarrow \infty} E[M_{T \wedge R_n}^2] \leq E[M_T^2]$ , so

$$E[\langle M \rangle_T] = E[M_T^2].$$

Finally, by assumption,

$$E[M_T^2] \leq \left[ M_{\sup_{\omega \in \Omega} T(\omega)}^2 \right] < \infty,$$

so that

$$E[M_T^2 - \langle M \rangle_T] = 0.$$

□

**Remark 2.2.9.** (i) One can drop the assumption that “all  $P$ -zero sets in  $\mathcal{F}$  are in  $\mathcal{F}_0$ ”, but one only gets that the particular version of  $\langle M \rangle$  of  $V$  is only right-continuous and adapted. But it is continuous only for  $\omega \in N^c$ , with a  $P$ -zero set  $N \in \mathcal{F}$ .

(ii) Since convergence in probability implies  $P$ -a.s. convergence of a subsequence, it follows by Proposition 2.2.8(i) that for some subsequence  $(n_k)_{k \in \mathbb{N}}$

$$P(V_s^{(n_k)} \xrightarrow{k \rightarrow \infty} \langle M \rangle_s \text{ locally uniformly on } [0, t], \forall t \geq 0) = 1.$$

(iii) Since  $\langle M \rangle$  is exactly the pathwisely defined continuous quadratic variation process of  $M$  in Chapter I, we can apply all results from Chapter I for  $P$ -a.e.  $\omega \in \Omega$  fixed, i.e.

$$\langle M(\omega) \rangle = \langle M \rangle(\omega).$$

**Corollary 2.2.10.** Assume  $M_0 \equiv 0$ . Then

$$M_t^2 - \langle M \rangle_t = 2 \int_0^t M_s dM_s.$$

Hence, the continuous local martingale  $M^2 - \langle M \rangle$  is an Itô-integral (cf. Corollary 5.3.10(i)).

**Corollary 2.2.11.**  $P$ -a.s. the paths of a continuous local martingale are either constant or of unbounded variation.

*Proof.* Apply Corollary 5.3.10(ii). □

**Corollary 2.2.12.**  $\langle M \rangle$  is the unique increasing continuous adapted process such that  $\langle M \rangle_0 = 0$  and  $M^2 - \langle M \rangle$  is a continuous local martingale. In particular,  $\langle M \rangle$  is (up to a  $P$ -zero set in  $\mathcal{F}$ ) independent of the chosen sequence of partitions  $(\tau_n)$  in 2.2.6.

*Proof.* Let  $A, B$  be two such processes. Then  $M^2 - A, M^2 - B$  are continuous local martingales. Hence,  $B - A$  is a continuous local martingale (with respect to  $(\mathcal{F}_t)$  since  $A, B$  are  $(\mathcal{F}_t)$ -adapted) of bounded variation. Therefore, by Corollary 2.2.11

$$B - A = c = B_0 - A_0 = 0 \quad P\text{-a.s.}$$

□

## 2.2. Quadratic Variation and Covariation for Continuous Local Martingales

**Definition 2.2.13.** (cf. 5.2.3, 1.3.15) Let  $M, N$  be continuous local martingales. Then define the covariation process of  $M, N$  by

$$\langle M, N \rangle_t := \frac{1}{2} (\langle M + N \rangle_t - \langle M \rangle_t - \langle N \rangle_t).$$

**Remark 2.2.14.** Since  $M + N$  is a continuous local martingale,  $\langle M + N \rangle$  exists as a continuous process by Proposition 2.2.8. Therefore, since  $\frac{1}{2}[(a+b)^2 - a^2 - b^2] = ab \quad \forall a, b \in \mathbb{R}$  („polarization identity“)

$$\langle M, N \rangle = \lim_{n \rightarrow \infty} \sum_{s \in \tau_n} (M_{s' \wedge t} - M_{s \wedge t})(N_{s' \wedge t} - N_{s \wedge t}).$$

(Exercise!)

**Proposition 2.2.15.** Let  $M, N$  be continuous local martingales.  $\langle M, N \rangle$  is uniquely determined by the following:

- (i) Its paths are ( $P$ -a.s.) continuous and of bounded variation and  $\langle M, N \rangle_0 = 0$ .
- (ii)  $M \cdot N - \langle M, N \rangle$  is a continuous local martingale. In particular,  $\langle M, N \rangle \equiv 0$  if and only if  $M \cdot N$  is a local martingale.

*Proof.* Analogous to the case  $M = N$  in Corollary 2.2.12 (or use polarization). □

**Lemma 2.2.16.** Let  $M, N$  be a continuous local martingale. Let  $G (= G_s(\omega), s \geq 0, \omega \in \Omega)$  and  $H$  be  $\mathcal{B}(\mathbb{R}_+) \otimes \mathcal{F}$ -measurable. Then

$$\left| \int_0^t H_s(\omega) G_s(\omega) d\langle M, N \rangle_s(\omega) \right| \leq \left( \int_0^t H_s^2 d\langle M \rangle_s(\omega) \right)^{\frac{1}{2}} \left( \int_0^t G_s^2 d\langle N \rangle_s(\omega) \right)^{\frac{1}{2}}.$$

In particular, we obtain (by Cauchy) the Kunita-Watanabe inequality

$$E \left[ \left| \int_0^t H_s(\omega) G_s(\omega) d\langle M, N \rangle_s \right|^2 \right] \leq E \left[ \int_0^t H_s^2 d\langle M \rangle_s \right] E \left[ \int_0^t G_s^2 d\langle N \rangle_s(\omega) \right].$$

*Proof.* Exercise (cf. [RW87, Vol II, p.50]). □

### 2.3. Construction of stochastic integrals

Fix a probability space  $(\Omega, \mathcal{F}, P)$  with right-continuous filtration  $(\mathcal{F}_t)$  such that all  $P$ -zero sets in  $\mathcal{F}$  are in  $\mathcal{F}_0$ . We want to define

$$H.M := \int_0^\cdot H_s dM_s \quad \text{as a martingale}$$

for càdlàg  $(\mathcal{F}_t)$ -martingales and most general  $H$ , where càdlàg means right-continuous for all  $\omega \in \Omega$  with left limits  $P$ -a.s.. (The latter is automatic by [vWW90, p.47, 3.25, 3.26].)  $M_s$  is called *integrator process* and  $H_s$  *integrand process*.

“Admissible integrators” are given by

$$\mathcal{M}^2 := \mathcal{M}^2(\Omega, \mathcal{F}, P) := \left\{ M \mid M \text{ is a càdlàg martingale, } M_0 = 0, \|M\|^2 := \sup_{t \geq 0} E[M_t^2] < \infty \right\}.$$

We define

$$\mathcal{M}_c^2 := \{M \in \mathcal{M}^2 \mid M \text{ has } P\text{-a.s. continuous sample paths}\}$$

We know that for each  $M \in \mathcal{M}^2$  the Doob-Meyer decomposition holds, that is:

For all  $M \in \mathcal{M}^2$  there exists a process  $\langle M \rangle$  such that it is the unique predictable, right continuous increasing, adapted process and that  $M_0 \equiv 0$  and  $M^2 - \langle M \rangle$  is a martingale. We only proved that this  $\langle M \rangle$  exists and is unique if  $M \in \mathcal{M}_c^2$  (see Proposition 2.2.8 and Corollary 2.2.12). For the general case see [Kry] or [vWW90, p.130, Cor. 6.6.3]. However, if  $M \in \mathcal{M}^2 \setminus \mathcal{M}_c^2$ , then  $\langle M \rangle$  is NOT equal to the quadratic variation of  $M$ .

**Remark 2.3.17.** (i) Define  $\mathcal{F}_\infty := \sigma\left(\bigcup_{t \geq 0} \mathcal{F}_t\right)$ . Clearly,  $M_t \in \mathcal{L}^2(\Omega, \mathcal{F}_\infty, P)$  for all  $M \in \mathcal{M}^2$  and by the  $\mathcal{L}^2$ -martingale convergence theorem we have for any càdlàg martingale such that  $M_0 = 0$ :

$$M \in \mathcal{M}^2 \Leftrightarrow \exists M_\infty := \lim_{t \rightarrow \infty} M_t \in \mathcal{L}^2(\Omega, \mathcal{F}_\infty, P).$$

In this case

$$M_t = E[M_\infty | \mathcal{F}_t], \quad t \geq 0.$$

So  $(M_t)_{t \in [0, \infty]}$  is a martingale.

(ii) Starting with the norm  $\|\cdot\|$  define an inner product on  $\mathcal{M}^2$  by polarization:

$$(M, N) := \frac{1}{4}(\|M + N\|^2 - \|M - N\|^2), \quad \forall M, N \in \mathcal{M}.$$

It remains to check that it is really an inner product. This could be done by checking that  $\|\cdot\|$  on  $\mathcal{M}^2$  satisfies

$$\|M + N\|^2 + \|M - N\|^2 = 2\|M\|^2 + 2\|N\|^2 \quad (\text{parallelogram identity})$$

and using J. von Neumann’s theorem. But (iii) implies this more easily.

(iii) Let  $M \in \mathcal{M}^2$ . Since  $M^2$  is a submartingale and because of (i) we have

$$E[M_\infty^2] \stackrel{(i)}{=} \lim_{t \rightarrow \infty} E[M_t^2] = \sup_{t \geq 0} E[M_t^2] = \|M\|^2.$$

Hence, for  $N \in \mathcal{M}^2$ , by polarization

$$\begin{aligned} E[M_\infty N_\infty] &= \frac{1}{4} (E[M_\infty + N_\infty]^2 - E[M_\infty - N_\infty]^2) \\ &= \frac{1}{4} (\|M + N\|^2 - \|M - N\|^2) = (M, N). \end{aligned}$$

Therefore,  $(\cdot, \cdot)$  is really an inner product with corresponding norm  $\|\cdot\|$ .

(iv) Note that  $\|M\|^2 = \sup_{t \geq 0} E[M_t^2] \leq E[\sup_{t \geq 0} M_t^2] \stackrel{Doob}{\leq} 4 \sup_{t \geq 0} E[M_t^2] \leq 4 \|M\|^2$ . So, on  $\mathcal{M}^2$  the norms  $\|\cdot\|$  and  $\|\cdot\|_{L^2(\Omega, \mathcal{F}, P; C_b([0, \infty); \mathbb{R}))}$  are equivalent.

**Proposition 2.3.18.** (i)  $\mathcal{M}^2$  is a Hilbert space and

$$(M, N) = E[M_\infty N_\infty] = E[\langle M, N \rangle_\infty],$$

where

$$\langle M, N \rangle_\infty := \lim_{t \rightarrow \infty} \langle M, N \rangle_t.$$

(ii)  $\mathcal{M}_c^2$  is a closed subspace of  $\mathcal{M}^2$ .

*Proof.* (i)  $(\mathcal{M}^2, \|\cdot\|)$  is an inner product space (Pre-Hilbert space) by Remark 2.3.17. Furthermore, it is complete by Proposition 2.1.4(ii) and because

$$\mathcal{L}^2(\Omega, \mathcal{F}, P; C_b([0, \infty), \mathbb{R})),$$

is complete.

Finally,

$$\|M\|^2 = E[M_\infty^2] = \lim_{t \rightarrow \infty} E[M_t^2] = \lim_{t \rightarrow \infty} E[\langle M \rangle_t] \stackrel{B.Levi}{=} E[\underbrace{\lim_{t \rightarrow \infty} \langle M \rangle_t}_{=: \langle M \rangle_\infty}].$$

So, by polarization the last assertion follows.

(ii) Apply Proposition 2.1.4(ii). □

Now we define stochastic integrals with  $M \in \mathcal{M}^2$  as integrators, but first for elementary functions:

**Definition 2.3.19.** Define  $\mathcal{E}$  to be the set of all processes  $H$  on  $]0, \infty[ \times \Omega$  which are of the following form:

For  $t > 0$ ,  $\omega \in \Omega$

$$H_t(\omega) := \sum_{i=0}^{n-1} h_{t_i}(\omega) 1_{]t_i, t_{i+1}]}(t), \quad \text{“elementary (predictable) adapted processes”}$$

where  $n \in \mathbb{N}$ ,  $0 = t_0 < t_1 < \dots < t_n < \infty$  and  $h_{t_i}$  are  $\mathcal{F}_{t_i}$ -measurable bounded. For  $M \in \mathcal{M}^2$  and  $H$  as above define

$$\int_0^t H_s dM_s := \sum_{i=0}^{n-1} h_{t_i} (M_{t_{i+1} \wedge t} - M_{t_i \wedge t}), \quad t \geq 0.$$

An easy exercise shows that this is independent of the representation of  $H$ ! Set

$$(H.M)_t := \int_0^t H_s dM_s, \quad t \geq 0.$$

Then  $H.M$  is called the stochastic integral of  $H$  with respect to  $M$ .

**Lemma 2.3.20.** Let  $H \in \mathcal{E}$ .

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(i)  $H.M \in \mathcal{M}^2$  and if  $M \in \mathcal{M}_c^2$ , then  $H.M \in \mathcal{M}_c^2$ .

(ii)

$$\begin{aligned}\langle H.M \rangle_t &= \int_0^t H_s^2 d\langle M \rangle_s \quad \forall t \geq 0 \\ &= \sum_{i=0}^{n-1} h_{t_i}^2 (\langle M \rangle_{t_{i+1} \wedge t} - \langle M \rangle_{t_i \wedge t}).\end{aligned}$$

In particular,

$$\|H.M\|^2 = E \left[ \int_0^\infty H_s^2 d\langle M \rangle_s \right]. \quad \text{“Isometry”}$$

Note that here we always set

$$\int_0^t H_s^2 d\langle M \rangle_s := \int_{]0,t]} H_s^2 d\langle M \rangle_s.$$

*Proof.* (i) By definition we see that:

- if  $M$  is a cadlag (continuous) martingale, then  $H.M$  is càdlàg (continuous),
- $H.M$  is adapted,
- $(H.M)_0 = 0$ .

Furthermore, since  $M$  is a martingale,

$$\begin{aligned}\sup_{t \geq 0} E[(H.M)_t^2] &= \sup_{t \geq 0} E \left[ \sum_{i=0}^{n-1} h_{t_i}^2 (M_{t_{i+1} \wedge t} - M_{t_i \wedge t})^2 \right] \\ &= \sup_{t \geq 0} \sum_{i=0}^{n-1} E \left[ h_{t_i}^2 (M_{t_{i+1} \wedge t} - M_{t_i \wedge t})^2 \right] \\ &\leq \sup_{0 \leq i \leq n} \|h_{t_i}^2\|_\infty \sup_{t \geq 0} \sum_{i=0}^{n-1} E \left[ (M_{t_{i+1} \wedge t}^2 - M_{t_i \wedge t}^2) \right] \\ &\leq \sup_{0 \leq i \leq n} \|h_{t_i}^2\|_\infty \sup_{t \geq 0} E [M_{t_n \wedge t}^2] \\ &\leq \sup_{0 \leq i \leq n} \|h_{t_i}^2\|_\infty \|M\|^2 < \infty.\end{aligned}$$

It remains to show the martingale property of  $H.M$ :

Let  $T$  be a bounded stopping time. Then

$$\begin{aligned}E[(H.M)_T] &= \sum_{i=0}^{n-1} E \left[ h_{t_i} (M_{t_{i+1} \wedge T} - M_{t_i \wedge T}) \right] \\ &= \sum_{i=0}^{n-1} E \left[ h_{t_i} \underbrace{E [M_{t_{i+1} \wedge T} - M_{t_i \wedge T} | \mathcal{F}_{t_i}]}_{=0, \text{ since } (M_{t \wedge T})_{t \geq 0} \text{ is an } (\mathcal{F}_t)\text{-martingale}} \right] \\ &= 0.\end{aligned}$$

Thus,  $H.M$  is a martingale.

- (ii) By uniqueness of the Doob-Meyer decomposition (see Corollary 2.2.12 for  $M \in \mathcal{M}_c^2$ ) it is enough to show that

$$(H.M)_t^2 - \int_0^t H_s^2 d\langle M \rangle_s, \quad t \geq 0,$$

is a martingale. (Then  $\langle H.M \rangle_t = \int_0^t H_s^2 d\langle M \rangle_s$ .) Let  $T$  be a bounded stopping time. Then, by defining

$$\Delta_i M := M_{t_{i+1} \wedge T} - M_{t_i \wedge T}$$

and since all mixed terms dissappear, we get that

$$\begin{aligned} E[(H.M)_T^2] &= \sum_{i,j} E [h_{t_i} h_{t_j} (\Delta_i M)(\Delta_j M)] \\ &= \sum_i E [h_{t_i}^2 E[(\Delta_i M)^2 | \mathcal{F}_{t_i}]] \\ &= \sum_i E \left[ h_{t_i}^2 E[M_{t_{i+1} \wedge T}^2 - M_{t_i \wedge T}^2 | \mathcal{F}_{t_i}] \right] \\ &= \sum_i E \left[ h_{t_i}^2 E[\langle M \rangle_{t_{i+1} \wedge T} - \langle M \rangle_{t_i \wedge T} | \mathcal{F}_{t_i}] \right] \\ &= E \left[ \int_0^T H_s^2 d\langle M \rangle_s \right]. \end{aligned}$$

In particular,

$$\|H.M\|^2 = \sup_{t \geq 0} E[(H.M)_t^2] = \sup_{t \geq 0} E \left( \int_0^t H_s^2 d\langle M \rangle_s \right) \stackrel{\text{B.Levi}}{=} E \left( \int_0^\infty H_s^2 d\langle M \rangle_s \right).$$

□

We now want to consider the above ‘‘Isometry Property’’ more closely:

Let  $\bar{\Omega} := ]0, \infty[ \times \Omega$ ,  $\bar{\omega} := (t, \omega)$ ,  $\bar{\mathcal{F}} = \mathcal{B}(]0, \infty[) \otimes \mathcal{F}$ .  $\omega \mapsto \langle M \rangle_t(\omega)$  is  $\mathcal{F}$ -measurable for fixed  $t$  and for all  $\omega \in \Omega$   $t \mapsto \langle M \rangle_t(\omega)$  is right-continuous, nonnegative, increasing, with  $\langle M \rangle_0(\omega) = 0$ . Therefore, for fixed  $\omega \in \Omega$  there exists a unique, positive measure

$$\langle M \rangle(\omega, dt) := d\langle M \rangle_t(\omega).$$

Thus,  $\langle M \rangle(\omega, dt)$  defines a transition kernel from  $(\Omega, \mathcal{F})$  to  $(]0, \infty[, \mathcal{B}(]0, \infty[))$  and it induces a measure

$$P_M(d\bar{\omega}) := P(d\omega) \otimes \langle M \rangle(\omega, dt)$$

on  $(\bar{\Omega}, \bar{\mathcal{F}})$ . Explicitly,

$$P_M(A) = E \left[ \int_0^\infty 1_A(\cdot, t) d\langle M \rangle_t(\cdot) \right], \quad A \in \bar{\mathcal{F}}.$$

Note that this is not a probability measure, but it is finite since  $M \in \mathcal{M}_{(c)}^2$ . In particular, defining

$$E_M[\cdot] = \int \cdot dP_M,$$

we have

$$E_M[H^2] = E \left[ \int_0^\infty H_s^2 d\langle M \rangle_s \right], \quad (2.3.1)$$

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which is the  $\mathcal{L}^2(\bar{\Omega}, \bar{\mathcal{F}}, P_M)$ -norm of  $H \in \mathcal{E}$ . ( $H : \bar{\Omega} \rightarrow \mathbb{R}$ ,  $H(\omega, t) = H_t(\omega)$ .) Note that the map

$$\begin{aligned} \mathcal{E} &\rightarrow \mathcal{M}^2, \\ H &\mapsto H.M \end{aligned}$$

is obviously linear and by Lemmas 2.3.20(ii) and (2.3.1), an isometry from  $\mathcal{E} \subset \mathcal{L}^2(\bar{\Omega}, \bar{\mathcal{F}}, P_M)$  to  $(\mathcal{M}_{(c)}^2, \|\cdot\|)$ . Therefore, there exists a unique isometric extension to the closure of  $\mathcal{E}$  in  $\mathcal{L}^2(\bar{\Omega}, \bar{\mathcal{F}}, P_M)$  denoted by

$$\bar{\mathcal{E}} := \bar{\mathcal{E}}^M$$

(which depends on  $M$ ).

**Definition 2.3.21.** For  $H \in \bar{\mathcal{E}}^M$ , let  $H.M$  denote the uniquely determined element in  $\mathcal{M}_{(c)}^2$  with

$$\lim_{n \rightarrow \infty} \|H^n.M - H.M\| = 0$$

for every sequence  $(H^n)_{n \in \mathbb{N}} \subset \mathcal{E}$  which converges in  $\mathcal{L}^2(\bar{\Omega}, \bar{\mathcal{F}}, P_M)$  to  $H$ .

Automatically, we have

$$\|H.M\|^2 = E_{P_M}[H^2].$$

But we also have the analogue of 2.3.20(ii), namely:

**Proposition 2.3.22.** Let  $H \in \bar{\mathcal{E}}^M$ ,  $M \in \mathcal{M}_{(c)}^2$ , and therefore,  $H.M \in \mathcal{M}_{(c)}^2$ . Then

$$\langle H.M \rangle_t = \int_0^t H_s^2 d\langle M \rangle_s, \quad t \geq 0,$$

and hence by polarization for  $G \in \bar{\mathcal{E}}^M$

$$\langle H.M, G.M \rangle_t = \int_0^t H_s G_s d\langle M \rangle_s, \quad t > 0.$$

*Proof.* Let  $T$  be a bounded stopping time. Then

$$E[(H.M)_T^2] = \lim_{n \rightarrow \infty} E[(H^n.M)_T^2],$$

since

$$\begin{aligned} E[((H.M)_T - (H^n.M)_T)^2] &\leq E \left[ \sup_t (H.M - H^n.M)_t^2 \right] \\ &\stackrel{\text{Doob}}{\leq} 4 \|H.M - H^n.M\|^2 \xrightarrow{n \rightarrow \infty} 0. \end{aligned}$$

Hence,

$$\begin{aligned} E[(H.M)_T^2] &= \lim_{n \rightarrow \infty} E[(H^n.M)_T^2] \\ &\stackrel{2.3.20(ii)}{=} \lim_{n \rightarrow \infty} E \left[ \int_0^T (H^n)_s^2 d\langle M \rangle_s \right] \\ &= E \left[ \int_0^T H_s^2 d\langle M \rangle_s \right], \end{aligned}$$

since  $H^n \xrightarrow{n \rightarrow \infty} H$  in  $\mathcal{L}^2(P_M)$ , so,  $1_{]0, T]} H^n \rightarrow 1_{]0, T]} H$  in  $\mathcal{L}^2(P_M)$ . Therefore,  $(H.M)_t^2 - \int_0^t H_s^2 d\langle M \rangle_s$  is a martingale and the assertion follows by the uniqueness of the Doob-Meyer decomposition.  $\square$

In our next step we want to determine the size of  $\bar{\mathcal{E}}^M$ . We want to characterize admissible integrands in dependence of  $M$ .

**Definition 2.3.23.**

$$\mathcal{P} := \sigma(H = (H_t)_{t>0} | H \text{ is a (real-valued) left-continuous, adapted process})$$

$$\stackrel{\text{exercise}}{=} \sigma(H = (H_t)_{t>0} | H \text{ is a (real-valued) continuous, adapted process})$$

is called  $\sigma$ -algebra (on  $]0, \infty[ \times \Omega$ ) of predictable sets. A process  $H = (H_t)_{t>0}$  is called predictable, if it is  $\mathcal{P}$ -measurable.

Let  $\mathcal{P}_M :=$  the completion of  $\mathcal{P}$  with respect to  $P_M$  in  $\bar{\mathcal{F}}$ .

**Remark 2.3.24.** Every predictable process  $(H_t)_{t>0}$  is progressively measurable, i.e.  $H_{]0,t] \times \Omega}$  is  $\mathcal{B}(]0,t]) \otimes \mathcal{F}_t$ -measurable for all  $t > 0$ , hence in particular  $(\mathcal{F}_t)$ -adapted (cf. [vWW90, Lemma 6.1.6]).

**Remark 2.3.25.** Let  $\tau_n$  be a partition of  $[0, \infty)$  and

$$\mathcal{P}_n := \sigma(]t, t'[\times A_t | t \in \tau_n, t' \text{ is the successor of } t \text{ in } \tau_n \text{ and } A_t \text{ is } \mathcal{F}_t\text{-measurable})$$

$\mathcal{P}_n$  is called the  $\sigma$ -algebra of predictable rectangles (with respect to  $\tau_n$ ).

(i) If  $|\tau_n| \xrightarrow{n \rightarrow \infty} 0$  and  $t_{N_n} \xrightarrow{n \rightarrow \infty} \infty$ , then

$$\mathcal{P} = \sigma\left(\bigcup_n \mathcal{P}_n\right).$$

(ii)  $H$  is  $\mathcal{P}_n$ -measurable and bounded if and only if

$$H_t(\omega) = \sum_{t_i \in \tau_n} h_{t_i}(\omega) 1_{]t_i, t_{i+1}]}(t), \quad t > 0,$$

where  $h_{t_i}$  is  $\mathcal{F}_{t_i}$ -measurable and bounded.

*Proof.* (i) By definition we have  $\mathcal{P}_n \subset \mathcal{P}$ . Hence,  $\sigma(\bigcup_n \mathcal{P}_n) \subset \mathcal{P}$ . To show that  $\mathcal{P} \subset \sigma(\bigcup_n \mathcal{P}_n)$ , let  $(H_t)_{t>0}$  be a left-continuous adapted process. It suffices to show that  $H$  is  $\sigma(\bigcup_n \mathcal{P}_n)$ -measurable.  $H$  is adapted and left-continuous. Therefore,

$$H_t(\omega) = \lim_{n \rightarrow \infty} \sum_{s \in \tau_n} \underbrace{H_s(\omega) 1_{]s, s']}(t)}_{\sigma(\mathcal{P}_n)\text{-measurable}}, \quad t > 0,$$

Hence,

$$\mathcal{P} = \sigma\left(\bigcup_n \mathcal{P}_n\right).$$

(ii) The assertion follows by a monotone class argument. □

**Proposition 2.3.26.**

$$\bar{\mathcal{E}}^M = \mathcal{L}^2(\bar{\Omega}, \mathcal{P}_M, P_M).$$

In particular, for all  $H \in \mathcal{L}^2(\bar{\Omega}, \mathcal{P}_M, P_M)$  there exists

$$H.M = \int H \, dM.$$

In particular,

$$\mathcal{M}_M^2 := \{H.M | H \in \mathcal{L}^2(\bar{\Omega}, \mathcal{P}_M, P_M)\}$$

is a closed subspace of  $\mathcal{M}^2$ .

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*Proof.* Clearly,  $\bar{\mathcal{E}}^M \subset \mathcal{L}^2(\bar{\Omega}, \mathcal{P}_M, P_M)$ . To prove the dual inclusion, let  $H \in \mathcal{L}^2(\bar{\Omega}, \mathcal{P}_M, P_M)$ . Without loss of generality  $H \geq 0$ . (Otherwise consider  $H^+$  and  $H^-$ .) Since

$$H^{(m)} := (m \wedge H) \xrightarrow{m \rightarrow \infty} H$$

in  $\mathcal{L}^2(\bar{\Omega}, \mathcal{P}_M, P_M)$ , we may assume  $H \geq 0$ , bounded. Now let  $\tau_n$  be a sequence of partitions as in Remark 2.3.25(i). Then by the  $\mathcal{L}^1$ -martingale convergence theorem (see [Röc11, Korollar 8.5.4 (i)]; ok, since  $\mathcal{P}_M$  is a finite measure!)

$$H_n := E_{P_M}[H|\mathcal{P}_n] \xrightarrow{n \rightarrow \infty} E_{P_M} \left[ H \middle| \sigma \left( \bigcup_n \mathcal{P}_n \right) \right] \text{ in } \mathcal{L}^1(\mathcal{P}_M).$$

Since  $H$  is bounded, this convergence also holds in  $\mathcal{L}^2(\mathcal{P}_M)$ . By Remark 2.3.25(i) this implies that in  $\mathcal{L}^2(\mathcal{P}_M)$

$$\lim_{n \rightarrow \infty} H_n = \lim_{n \rightarrow \infty} E_{P_M}[H|\mathcal{P}_n] = E_{P_M}[H|\mathcal{P}] = E_{P_M}[H|\mathcal{P}_M] = H.$$

Since  $H_n$  is  $\mathcal{P}_n$ -measurable, we have by Remark 2.3.25(ii), that

$$H_n \in \mathcal{E},$$

and the assertion follows.  $\square$

**Proposition 2.3.27.** (i) Let  $t \mapsto \langle M \rangle_t$  be continuous (which is e.g. the case if  $M \in \mathcal{M}_c^2$ ). Then

$$\left( \mathcal{L}^2(\bar{\Omega}, \mathcal{P}_M, P_M) \stackrel{2.3.26}{=} \bar{\mathcal{E}}^M = \mathcal{L}^2(\bar{\Omega}, \mathcal{O}_M, P_M), \right)$$

where the optional  $\sigma$ -algebra  $\mathcal{O}$  is defined by

$$\mathcal{O} := \sigma(\{H \mid H \text{ is an adapted càdlàg process on } ]0, \infty[ \times \Omega\})$$

and  $\mathcal{O}_M$  its completion with respect to  $P_M$  in  $\bar{\mathcal{F}}$ .

(ii) If  $M \in \mathcal{M}_c^2$  and  $d\langle M \rangle$  is absolutely continuous with respect to Lebesgue measure  $dt$ , then

$$\bar{\mathcal{E}}^M = \{H \in \mathcal{L}^2(\bar{\Omega}, \bar{\mathcal{F}}, P_M) \mid H \text{ has an adapted version}\}.$$

*Proof.* (i) We have that  $\mathcal{O} \supset \mathcal{P}$ . Hence, also  $\mathcal{O}_M \supset \mathcal{P}_M$ . It remains to prove  $\mathcal{O}_M \subset \mathcal{P}_M$ . So, let  $H$  be càdlàg and adapted. Consider for  $\omega \in \Omega$

$$U_\omega := \{t > 0 \mid s \mapsto H_s(\omega) \text{ is discontinuous in } t\}.$$

Then  $U_\omega$  is countable, since  $H$  is càdlàg (exercise). Define for  $\omega \in \Omega$

$$H_t^-(\omega) := \lim_{s \nearrow t} H_s(\omega) =: H_{t-}(\omega) \quad (\in \mathbb{R}).$$

Then  $H^-$  is left-continuous. Hence, it is  $\mathcal{P}$ -measurable and

$$U_\omega = \{t > 0 \mid H_t(\omega) \neq H_{t-}(\omega)\}.$$

Note that  $\{(t, \omega) \in ]0, \infty[ \times \Omega \mid H_t(\omega) \neq H_{t-}(\omega)\} \in \mathcal{O} \subset \bar{\mathcal{F}}$ . Since  $\langle M \rangle$  is continuous, we have

$$\int_0^\infty 1_{U_\omega}(t) d\langle M \rangle_t = 0 \quad \text{for } P\text{-a.e. } \omega \in \Omega.$$

But then

$$E_{P_M}[1_{\{H \neq H^-\}}] = \int \int_{]0, \infty[} 1_{U_\omega}(t) d\langle M \rangle_t(\omega) P(d\omega) = 0.$$

Hence,  $H$  is  $P$ -a.e. equal to a predictable process, therefore,  $H$  is  $\mathcal{P}_M$ -measurable and (i) is proved.

(ii) One has to prove that  $\bar{\mathcal{F}} \subset \mathcal{P}_M$  (cf. [CW90, p.60] or [vWW90, p.124]).

□

**Remark 2.3.28.** *By a modification of the proof of Proposition 2.3.27(i) one can even show that every progressively measurable process  $(H_t)_{t>0}$  (see Remark 2.3.24 above) has a predictable  $\mathcal{P}_M$ -version (of course still assuming that  $\langle M \rangle$  is  $P$ -a.s. continuous).*

**Remark 2.3.29** (Extending stochastic integration via localization). *Note that, if  $M$  is a Brownian motion, then  $M \notin \mathcal{M}^2$ . Therefore, we have to stop. Define*

$$\mathcal{M}_{loc}^2 := \{M \mid \exists \text{ a sequence } T_n \nearrow \infty \text{ of stopping times such that } M^{T_n} (= M_{T_n \wedge \cdot}) \in \mathcal{M}^2 \forall n \in \mathbb{N}\}.$$

Then for all  $H \in \mathcal{L}_{loc}^2(\bar{\Omega}, \mathcal{P}_M, P_M)$

$$H.M = \int H dM$$

is defined via localization. We have to check that

- the definition is consistent on  $\{T_n = T_{n+1}\}$ ,
- $H.M \in \mathcal{M}_{loc}^2$  (We need Lemma 2.4.33 below).

**Remark 2.3.30** (Semi-martingales as integrals). *If  $A$  is predictable and of bounded variation and  $M \in \mathcal{M}_{loc}^2$  we can define for  $H \in \mathcal{L}_{loc}^2(\bar{\Omega}, \mathcal{P}_M, P_M)$*

$$\int H d(M + A) := \int H dM + \int H dA,$$

where  $M + A$  is a semi-martingale and  $\int H dA$  is a pathwise defined Lebesgue-Stieltjes-integral. Axiomatic considerations show that, if  $\int H d(M + A)$  is required to have reasonable properties, then this cannot be generalized. As a consequence, reasonable stochastic integrators are semi-martingales (Dellacherie-Mokobodzki-Bichteler) (cf. [vWW90]).

## 2. (Semi-)Martingales and Stochastic Integration

### 2.4. Characterization of $H.M$ in $\mathcal{M}^2$

Fix  $M \in \mathcal{M}^2$ ,  $H \in \mathcal{L}^2(\bar{\Omega}, \mathcal{P}_M, P_M)$ .

**Proposition 2.4.31.**  *$H.M$  is the unique element  $L \in \mathcal{M}^2$  such that*

$$(i) \quad d\langle L, N \rangle = H d\langle M, N \rangle \quad P\text{-a.s. } \forall N \in \mathcal{M}^2,$$

that is

$$\langle L, N \rangle_t = \int_0^t H_s d\langle M, N \rangle_s := \int 1_{]0,t]}(s) H_s d\langle M, N \rangle_s, \quad \forall t \geq 0, P\text{-a.s. } \forall N \in \mathcal{M}^2, \forall t > 0,$$

respectively, such that the following weaker property holds:

$$(ii) \quad E[L_\infty N_\infty] = E \left[ \int_0^\infty H_s d\langle M, N \rangle_s \right] \left( := \int 1_{]0,\infty[}(s) H_s d\langle M, N \rangle_s \right)$$

for all  $N \in \mathcal{M}^2$ .

**Remark 2.4.32.** (i) Because of  $E[L_\infty N_\infty] = E[\langle L, N \rangle_\infty]$  (cf 2.3.18(i)), it follows that in 2.4.31, (i) implies (ii).

(ii) By 2.4.31(i) with  $N$  replaced by  $G.N$  for  $G \in \mathcal{L}^2(\bar{\Omega}, \mathcal{P}_M, P_M)$  we particularly have

$$d\langle H.M, G.N \rangle = H d\langle M, G.N \rangle = HG d\langle M, N \rangle.$$

*Proof of 2.4.31.* (a) Uniqueness if (ii) holds:

Let  $L, L' \in \mathcal{M}^2$  such that both satisfy 2.4.31(ii). Then

$$E[(L_\infty - L'_\infty)N_\infty] = 0 \quad \forall N_\infty \in \mathcal{L}^2(\Omega, \mathcal{F}_\infty, P).$$

Hence  $L_\infty = L'_\infty$   $P$ -a.s.. Therefore,

$$L_t = E[L_\infty | \mathcal{F}_t] = L'_t \quad P\text{-a.s..}$$

Now let us first prove (ii).

(b)  $L := H.M$  satisfies (ii):

**Step 1:** Assume for fixed  $s > 0$  that

$$H_t(\omega) := h_s(\omega) 1_{]s,\infty[}(t),$$

where  $h_s$  is bounded and  $\mathcal{F}_s$ -measurable. Note that  $H \in \mathcal{L}^2(\bar{\Omega}, \mathcal{P}_M, P_M)$ . Therefore, for  $L := H.M$  we have

$$\begin{aligned} L_\infty &= (H.M)_\infty = \int_0^\infty h_s 1_{]s,\infty[}(t) dM_t \\ &= \lim_{N \rightarrow \infty} \int_0^N h_s 1_{]s,\infty[}(t) dM_t \\ &= \lim_{N \rightarrow \infty} h_s (M_N - M_s) = h_s (M_\infty - M_s) \end{aligned}$$

Thus, for all  $N \in \mathcal{M}^2$

$$\begin{aligned}
 E[L_\infty N_\infty] &= E[h_s(M_\infty - M_s)N_\infty] \\
 &\stackrel{(N_t)_{t \in [0, \infty]} \text{ mart.}}{=} E[h_s(M_\infty N_\infty - M_s N_s)] \\
 &= E[h_s(M_\infty N_\infty - \langle M, N \rangle_\infty - (M_s N_s - \langle M, N \rangle_s))] \\
 &+ E[h_s(\langle M, N \rangle_\infty - \langle M, N \rangle_s)] \\
 &= E[h_s(\langle M, N \rangle_\infty - \langle M, N \rangle_s)] \\
 &= E\left[h_s \int_{]s, \infty[} d\langle M, N_t \rangle\right] \\
 &= E\left[\int_0^\infty H_t d\langle M, N \rangle_t\right].
 \end{aligned}$$

**Step 2:** Let  $H \in \mathcal{E}$ .

In this case (ii) is clear by Step 1 and linearity.

**Step 3:** (ii) holds for all  $H \in \mathcal{L}^2(\bar{\Omega}, \mathcal{P}_M, P_M)$  :

Let  $H^{(n)} \in \mathcal{E}$ ,  $n \in \mathbb{N}$ ,  $H^{(n)} \rightarrow H$  in  $\mathcal{L}^2(\bar{\Omega}, \mathcal{P}_M, P_M)(= \bar{\mathcal{E}})$ . Then

$$H^{(n)}M \rightarrow H.M \text{ in } (\mathcal{M}^2, \|\cdot\|).$$

Therefore, by 2.3.17(iii) for all  $N \in \mathcal{M}^2$

$$\begin{aligned}
 E[(H.M)_\infty N_\infty] &= \lim_{n \rightarrow \infty} E[(H^{(n)}M)_\infty N_\infty] \\
 &\stackrel{\text{Step 2}}{=} \lim_{n \rightarrow \infty} E\left[\int_0^\infty H_s^{(n)} d\langle M, N \rangle_s\right] \\
 &= E\left[\int_0^\infty H_s d\langle M, N \rangle_s\right],
 \end{aligned}$$

because by Kunita-Watanabe-inequality (cf. 2.2.16) we have

$$E\left[\left|\int_0^\infty (H^{(n)} - H) \cdot 1 d\langle M, N \rangle\right|\right] \leq \left(E\left[\int_0^\infty (H^{(n)} - H)^2 d\langle M \rangle\right]\right)^{\frac{1}{2}} \cdot (E[\langle N \rangle_\infty])^{\frac{1}{2}} \xrightarrow{n \rightarrow \infty} 0.$$

This proves (ii).

(c)  $L := H.M$  satisfies (i):

By Proposition 2.2.15 we have to show that  $L_t N_t - \int_0^t H_s d\langle M, N \rangle_s$ ,  $t \geq 0$ , is a martingale. Let  $T$  be a bounded stopping time. Then, since  $(L_t)_{t \in [0, \infty]}$  is a martingale, for  $N_t^T := N_{T \wedge t}$ ,  $t \geq 0$ , we have

$$\begin{aligned}
 E[L_T N_T] &= E[L_\infty N_T] = E[L_\infty N_\infty^T] \\
 &\stackrel{(ii)}{=} E\left[\int_0^\infty H_s d\langle M, N^T \rangle_s\right] \\
 &\stackrel{2.4.33 \text{ below}}{=} E\left[\int_0^\infty H_s d\langle M, N \rangle_{s \wedge T}\right] \\
 &= E\left[\int_0^T H_s d\langle M, N \rangle_s\right]
 \end{aligned}$$

(where we used that obviously  $N^T \in \mathcal{M}^2$ , since  $N \in \mathcal{M}^2$ ).

□

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**Lemma 2.4.33.** *Let  $M, N \in \mathcal{M}^2$  and  $T$  be a stopping time. Then*

$$\langle M, N^T \rangle_t = \langle M, N \rangle_{T \wedge t} \quad \forall t \geq 0.$$

*Proof.* Exercise. (!) □

**Corollary 2.4.34.** *Let  $M \in \mathcal{M}^2$ ,  $H_1, H_2 \in \mathcal{L}^2(\bar{\Omega}, \mathcal{P}_M, P_M)$  such that  $H_1 \cdot H_2 \in \mathcal{L}^2(\bar{\Omega}, \mathcal{P}_M, P_M)$ . Then*

$$H_1 \cdot (H_2 \cdot M) = (H_1 \cdot H_2) \cdot M,$$

*i.e.*

$$\int_0^\infty H_1(s) d(H_2 \cdot M)_s = \int_0^\infty H_1(s) d\left(\int_0^s H_2(r) dM_r\right) = \int_0^\infty H_1(r)H_2(r) dM_r.$$

*Proof.* Let  $L := H_1 \cdot (H_2 \cdot M)$ . Then by 2.4.31 for all  $N \in \mathcal{M}^2$

$$d\langle L, N \rangle = H_1 d\langle H_2 \cdot M, N \rangle = H_1 H_2 d\langle M, N \rangle,$$

and

$$d\langle (H_1 \cdot H_2) \cdot M, N \rangle = (H_1 \cdot H_2) d\langle M, N \rangle.$$

Hence (again by 2.4.31)

$$(H_1 \cdot H_2) \cdot M = H_1 \cdot (H_2 \cdot M).$$

□

**Corollary 2.4.35.** *Let  $M \in \mathcal{M}^2$  and let  $T$  be a stopping time.*

(i) *Then  $(H \cdot M)^T = H 1_{]0, T]} \cdot M$ , i.e.*

$$(H \cdot M)_t^T = \int_0^t H_s 1_{]0, T]}(s) dM_s, \quad t \geq 0.$$

*In particular  $(H \cdot M)^T \in \mathcal{M}_M^2$ , hence,  $\mathcal{M}_M^2$  is “stopping stable”, that is,  $N \in \mathcal{M}_M^2$  implies  $N^T \in \mathcal{M}_M^2$  for all stopping times  $T$ .*

(ii)

$$(H \cdot M^T)_t = (H \cdot M)_t^T, \quad t \geq 0.$$

*Proof.* (i) Let  $N \in \mathcal{M}^2$ . Then

$$\begin{aligned} E[(H \cdot M)_\infty^T N_\infty] &\stackrel{2.3.18(i)}{=} E[\langle (H \cdot M)^T, N \rangle_\infty] \\ &\stackrel{2.4.33}{=} E[\langle H \cdot M, N \rangle_T] \\ &\stackrel{2.4.31(i)}{=} E\left[\int_0^\infty H_s 1_{]0, T]}(s) d\langle M, N \rangle_s\right]. \end{aligned}$$

Hence, the assertion follows by 2.4.31(ii).

(ii) Since  $M^T \in \mathcal{M}^2$ , the proof of (ii) is by 2.4.31(i) analogous to (i)

□

### 2.4.1. Orthogonality in $\mathcal{M}^2$

**Definition 2.4.36.** Let  $M, N \in \mathcal{M}^2$ .

(i)  $M, N$  are called weakly orthogonal if  $E[M_\infty N_\infty] (= (M, N)) = 0$  (i.e. orthogonal in  $(\mathcal{M}^2, \|\cdot\|)$ ).

(ii)  $M, N$  are called strongly orthogonal, denoted by  $M \perp N$ , if  $\langle M, N \rangle = 0, P$ -a.s..

**Remark 2.4.37.** (i)  $M \perp N \stackrel{2.2,15(i)}{\Leftrightarrow} M \cdot N$  is a martingale  $\Leftrightarrow E[M_T N_T] = 0$  for all bounded stopping times  $T$ . But  $M_t \xrightarrow{t \rightarrow \infty} M_\infty$  and  $N_T \xrightarrow{t \rightarrow \infty} N_\infty$  in  $\mathcal{L}^2(P)$ . Hence if  $M \perp N$ , then

$$E[M_\infty N_\infty] = 0,$$

i.e.  $M, N$  are weakly orthogonal.

(ii)  $M \perp N \stackrel{2.4,31(i)}{\Leftrightarrow} \mathcal{M}_M^2 \perp \mathcal{M}_N^2$  (since by 2.4.31  $d\langle H.M, \tilde{H}.N \rangle = H\tilde{H} d\langle M, N \rangle$ ).

(iii) Since  $(\mathcal{M}^2, \|\cdot\|)$  is a Hilbert space and  $\mathcal{M}_M^2$  is a closed linear subspace of  $\mathcal{M}^2$  (see 2.3.26), for all  $N \in \mathcal{M}^2$  there exists an  $H \in \mathcal{L}^2(\bar{\Omega}, \mathcal{P}_M, P_M)$  and  $L \in \mathcal{M}^2$  such that

$$N = \underbrace{H.M}_{\in \mathcal{M}_M^2} + \underbrace{L}_{\in \mathcal{M}^2}$$

and  $L$  is weakly orthogonal to  $\mathcal{M}_M^2$ .

But in (iii) we have more because of the following proposition.

**Proposition 2.4.38.** Let  $M, L \in \mathcal{M}^2$ ,  $L$  weakly orthogonal to  $\mathcal{M}_M^2$ . Then  $L \perp \mathcal{M}_M^2$  (since  $\mathcal{M}_M^2$  is stopping stable). Hence,  $L$  is weakly orthogonal to  $\mathcal{M}_M^2$  if and only if  $L \perp \mathcal{M}_M^2 (\Leftrightarrow L \perp M)$ .

*Proof.* Because of 2.4.37(i), it remains to show that  $E[L_T M_T] = 0$  for all bounded stopping times  $T$ . But by 2.4.35 (stopping stability) we have  $M^T \in \mathcal{M}_M^2$ . Hence,

$$0 = E[L_\infty M_\infty^T] = E[L_\infty M_T^T] \stackrel{L \text{ mart.}}{=} E[L_T M_T].$$

□

**Corollary 2.4.39** (Kunita-Watanabe-decomposition). (i) Let  $M, N \in \mathcal{M}^2$ . Then there exist a unique  $L \in \mathcal{M}^2$  and a unique  $H \in \mathcal{L}^2(\bar{\Omega}, \mathcal{P}_M, P_M)$  such that  $L \perp \mathcal{M}_M^2$  and  $N = H.M + L$ .

(ii) Suppose  $\mathcal{L}^2(\Omega, \mathcal{F}_\infty, P)$  separable (e.g. true, if  $\mathcal{F}_\infty$  is countably generated). Then there exist  $M_i \in \mathcal{M}^2, i \in \mathbb{N}$ , such that

$$M_i \perp M_j, i \neq j,$$

and for all  $N \in \mathcal{M}^2$  there exist  $H^i \in \mathcal{L}^2(\bar{\Omega}, \mathcal{P}, P_M), i \in \mathbb{N}$ , (uniquely determined by  $M_i$ ) such that

$$N = \sum_{i=1}^{\infty} H^i M_i, \quad (\text{the series converges in } \|\cdot\|_{\mathcal{M}^2}!)$$

that is,

$$\mathcal{M}^2 = \bigoplus_{i=1}^{\infty} \mathcal{M}_{M_i}^2$$

and

$$\mathcal{M}_{M_i}^2 \perp \mathcal{M}_{M_j}^2, \quad \text{for } i \neq j.$$

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*Proof.* (i) 2.4.37(iv) and 2.4.38.

(ii) By assumption and 2.4.37(iii)  $(\mathcal{M}^2, \|\cdot\|)$  is separable. So, we can apply (i) and the Gram-Schmidt orthogonalization procedure (cf. [Wei87],[RS80]). □

**Corollary 2.4.40.** *Let  $F \in \mathcal{L}^2(\Omega, \mathcal{F}_\infty, P)$ ,  $M \in \mathcal{M}^2$ . Then there exist uniquely determined  $H \in \mathcal{L}^2(\bar{\Omega}, \mathcal{P}_M, P_M)$  and  $L \in \mathcal{M}^2$ ,  $L \perp M$ , such that,*

$$F = E[F|\mathcal{F}_0] + \int_0^\infty H_s dM_s + L_\infty.$$

*Proof.* Let  $F_t := E[F|\mathcal{F}_t] - E[F|\mathcal{F}_0]$ . Since  $F_t$  is a martingale, by 2.1.1 there exists a càdlàg version  $(\tilde{F}_t)_{t \geq 0}$  of  $(F_t)_{t \geq 0}$  (i.e.  $P[\{F_t = \tilde{F}_t\}] = 1 \forall t \geq 0$ ). Then  $(\tilde{F}_t) \in \mathcal{M}^2$ . Hence, by 2.4.39(i) there exist unique  $L \in \mathcal{M}^2$ ,  $L \perp M$ , and  $H \in \mathcal{L}^2(\bar{\Omega}, \mathcal{P}_M, P_M)$  such that  $P$ -a.s.

$$\tilde{F}_t = (H.M)_t + L_t, \quad t \geq 0.$$

In particular, (since  $\tilde{F}_\infty = F - E[F|\mathcal{F}_0]$ ), we have  $P$ -a.s.

$$F = E[F|\mathcal{F}_0] + \int_0^\infty H_s dM_s + L_\infty.$$

□

## 2.5. Itô's Representation Theorem

Let  $(\Omega, \mathcal{F}, P)$  be a probability space with filtration  $(\mathcal{F}_t)$  and let  $(W_t)_{t \geq 0}$  be a (real-valued) Wiener process on  $(\Omega, \mathcal{F}, (\mathcal{F}_t), P)$ , that is,  $(W_t)$  is a continuous  $(\mathcal{F}_t)$ -adapted process such that, for all  $s \leq t$ ,  $W_t - W_s$  is independent of  $\mathcal{F}_s$  and  $N(0, t - s)$  distributed. Let for  $t \geq 0$

$$\mathcal{F}_t^W := \sigma(\{W_s | 0 \leq s \leq t\})$$

and  $\mathcal{F}_{t+}^W$ ,  $t \geq 0$ , the corresponding right-continuous filtration.

**Remark 2.5.41.** Compared with our definition of Brownian motion for a Wiener process the increment  $W_t - W_s$  is independent of the larger  $\sigma$ -algebra  $\mathcal{F}_s (\supseteq \mathcal{F}_s^W)$ . In this space of  $(\mathcal{F}_t^W)$ -martingales belonging to  $\mathcal{L}^2(\Omega, \mathcal{F}, P)$  the Kunita-Watanabe decomposition (cf. 4.9(i) with  $M = W$ ) has a particularly simple form:

**Theorem 2.5.42** (Itô's representation theorem). Let  $M = (M_t)_{t \geq 0} \subseteq \mathcal{L}^2(\Omega, \mathcal{F}, P)$  be a right-continuous martingale with respect to  $(\mathcal{F}_{t+}^W)$ , hence with respect to

$$(\mathcal{F}_{t+}^W)^P := \sigma(\mathcal{F}_{t+}^W, \sigma(\{N \in \mathcal{F} | P(N) = 0\})), t \geq 0.$$

Then

$$M_t = M_0 + \int_0^t H_s dW_s, t \geq 0, P\text{-a.s.}, \quad (2.5.2)$$

where  $H$  is  $(\mathcal{F}_t)$ -adapted and  $H \cdot 1_{]0, T]} \in \mathcal{L}^2(\bar{\Omega}, \bar{\mathcal{F}}, P_W)$  for all  $T > 0$  (and through this it is uniquely determined). In particular,  $M$  has  $P$ -a.s. continuous sample paths and  $\mathcal{M}^2 = \mathcal{M}_{W}^2$  (where  $\mathcal{M}^2$  is defined as in the previous section with respect to  $\mathcal{F}_t = (\mathcal{F}_{t+}^W)^P$ ,  $t \geq 0$ , which by the backward martingale convergence theorem is again right-continuous) and

$$\mathcal{M}_W^2 := \left\{ \left( \int_0^t H_s dW_s \right)_{t \geq 0} \left| H \cdot 1_{]0, t]} \in \mathcal{L}^2(\bar{\Omega}, \bar{\mathcal{F}}, P_W) \quad \forall t \geq 0, \sup_{t \geq 0} \left\| \int_0^t H_s dW_s \right\|_{\mathcal{L}^2(P)} < \infty \right\}.$$

*Proof.* Without loss of generality  $M_0 = 0$ . Fix  $k \in \mathbb{N}$  and set

$$\mathcal{F}_{k+}^W := (\mathcal{F}_{k+}^W)^P, M_t^{(k)} := M_{t \wedge k}, W_t^{(k+1)} := W_{t \wedge (k+1)}, t \geq 0.$$

Let  $(M^n)_{n \in \mathbb{N}} \subseteq \mathcal{L}^2(\Omega, \mathcal{F}_{k+}^W, P)$ , such that  $\|M^n\|_\infty < \infty$  and

$$\lim_{n \rightarrow \infty} M^n = M_k = M_\infty^{(k)} \quad \text{in } \mathcal{L}^2(\Omega, \mathcal{F}_{k+}^W, P).$$

Let  $M_t^n := E[M^n | \mathcal{F}_{t+}^W]$ ,  $t \geq 0$ . Then by the backward martingale convergence theorem (see [Röc11])  $(M_t^n)_{t \geq 0}$  is right continuous. Then for each  $n \in \mathbb{N}$

$$\sup_{t, \omega} |M_t^n(\omega)| < \infty.$$

In particular,  $M^n \in \mathcal{M}^2$  for all  $n$ . Hence, by the Kunita-Watanabe decomposition (cf. 2.4.39(i)) there exist adapted  $H^n \in \mathcal{L}^2(\bar{\Omega}, \bar{\mathcal{F}}, P_W)$  and  $N^n \in \mathcal{M}^2$ ,  $N^n \perp \mathcal{M}_{W^{(k+1)}}^2$ , such that

$$M_t^n = \left( H^n W^{(k+1)} \right)_t + N_t^n, t \geq 0, \quad \forall n \in \mathbb{N}.$$

We would like to prove that  $N^n = 0$  for all  $n \in \mathbb{N}$ .

**Step 1: Claim:** Let  $N \in \mathcal{M}^2$ ,  $N \perp W^{(k+1)}$ . If  $N$  is bounded, then  $N_t = 0$  for all  $t < k + 1$ .

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*Proof of Claim.* Let  $c := \sup_{t,\omega} |N_t(\omega)| < \infty$ . Define

$$D := 1 + \frac{N_\infty}{2c}.$$

Then  $D \geq \frac{1}{2}$  and

$$E[D] = 1 + \frac{1}{2c}E[N_\infty] = 1 + \frac{1}{2c}E[N_0] = 1.$$

Thus,  $\tilde{P} := DP$  is a probability measure on  $(\Omega, \mathcal{F})$  equivalent to  $P$ .

**Claim':**  $(W_t)_{t \leq k+1}$  is a Brownian motion under  $\tilde{P}$ .

Suppose the claim is true. Then the finite dimensional distributions of  $(W_t)_{t \leq k+1}$  under  $P$  and  $\tilde{P}$  are the same. Hence, by Radon-Nikodym,

$$P = \tilde{P} \quad \text{on } \mathcal{F}_{k+1}^W.$$

Therefore,  $P = \tilde{P}$  on  $\mathcal{F}_{t+}^W$  for all  $t < k+1$ . Thus,  $E[D|\mathcal{F}_{t+}^W] = 1$  for all  $t < k+1$ . Hence,

$$N_t = E[N_\infty|\mathcal{F}_{t+}^W] = 0, \quad \forall t < k+1.$$

□

*Proof of Claim'.*

(i)  $(W_t^{(k+1)})$  is a martingale under  $\tilde{P}$ , because for all bounded stopping times  $T$  we have

$$E_{\tilde{P}}[W_T^{(k+1)}] = E_P[W_T^{(k+1)}D] = \underbrace{E_P[W_T^{(k+1)}]}_{=0} + \frac{1}{2c} \underbrace{E_P[W_T^{(k+1)}N_\infty]}_{=:A}.$$

But by 2.4.37  $W_T^{k+1}N_T$  is a martingale, since  $W^{k+1} \perp N$ . Hence,

$$A = E_P[W_T^{k+1}N_T] = E_P[W_0^{(k+1)}N_0] = 0,$$

and thus

$$E_{\tilde{P}}[W_T^{(k+1)}] = 0.$$

(ii) For  $P$ -a.e.  $\omega \in \Omega$  we have

$$\langle W \rangle_t(\omega) = t \quad \forall t \geq 0.$$

Hence, for  $\tilde{P}$ -a.e.  $\omega \in \Omega$

$$\langle W \rangle_t(\omega) = t \quad \forall t \geq 0.$$

Thus,  $\tilde{P}$ -a.s.

$$\langle W_t^{(k+1)} \rangle = t \wedge (k+1) \quad \forall t \geq 0.$$

Therefore, by (i), (ii) and Levy's characterization (cf. Proposition 1.5.34)  $(W_t)_{t \leq k+1}$  is a Brownian motion under  $\tilde{P} = DP$  (up to  $k+1$ ). □

**Step 2:** Now fix  $n_0 \in \mathbb{N}$  and set  $N := N^{n_0}$ . Define for  $n \in \mathbb{N}$

$$T_n := \inf \left\{ t > 0 \mid \left| \int_0^t H_s^{n_0} dW_s^{(k+1)} \right| > n \right\} \wedge (k+1).$$

Then  $T_n$  are  $(\mathcal{F}_{t+}^W)$ -stopping times such that  $T_n \nearrow (k+1)$  as  $n \rightarrow \infty$ . We know that  $N^{T_n} (= N_{\cdot \wedge T_n}) \perp \mathcal{M}_{W^{(k+1)}}^2$  (cf. 2.4.33, 2.4.37(ii)).

Furthermore,

$$|N_t^{T_n}| = |N_{t \wedge T_n}| \leq |M_{t \wedge T_n}^{n_0}| + \left| \int_0^{t \wedge T_n} H_s^{n_0} dW_s^{(k+1)} \right| \leq \|M^{n_0}\|_\infty + n.$$

Hence,

$$\sup_{t, \omega} |N_t^{T_n}(\omega)| \leq \|M^{n_0}\|_\infty + n.$$

So, we can apply Step 1 to conclude that  $N_t^{T_n} = 0$ , for all  $t < k + 1$  and for all  $n \in \mathbb{N}$ . Letting  $n \rightarrow \infty$ , we get  $N_t = 0$  for all  $t < k + 1$   $P$ -a.s. (since  $N$  is right-continuous and the zero set does not depend on  $t$ ). Therefore,  $N^n = 0$  for all  $n$  and  $t < k + 1$ . But for all  $n, m \in \mathbb{N}$

$$\begin{aligned} E_{P_{W^{(k+1)}}} [(H^n - H^m)^2] &= E \left( \int_0^{k+1} (H_s^n - H_s^m)^2 ds \right) \\ &= E_P \left[ \left( (H^n W^{(k+1)})_{k+1} - (H^m W^{(k+1)})_{k+1} \right)^2 \right] \\ &= E \left[ (M_{k+1}^n - M_{k+1}^m)^2 \right] \xrightarrow{n, m \rightarrow \infty} 0. \end{aligned}$$

This is true, since  $P$ -a.e.

$$M_{k+1}^n = E \left[ M^n | \mathcal{F}_{(k+1)+}^W \right] \xrightarrow{n \rightarrow \infty} E \left[ M_k | \mathcal{F}_{(k+1)+}^W \right] = M_k \quad \text{in } \mathcal{L}^2(\Omega, \mathcal{F}, P).$$

Therefore,  $(H^n)_{n \in \mathbb{N}}$  is a Cauchy sequence in  $\mathcal{L}^2(\bar{\Omega}, \bar{\mathcal{F}}, P_{W^{(k+1)}})$ .

Hence, there exists  $H \in \mathcal{L}^2(\bar{\Omega}, \bar{\mathcal{F}}, P_{W^{(k+1)}})$  such that  $H^n \rightarrow H$  in  $\mathcal{L}^2(\bar{\Omega}, \bar{\mathcal{F}}, P_{W^{(k+1)}})$ . In particular, we can assume  $H$  to be adapted. Furthermore, for  $0 \leq t \leq k + 1$ ,

$$M_t = \lim_{n \rightarrow \infty} M_t^n = \lim_{n \rightarrow \infty} (H^n \cdot W^{(k+1)})_t = (H \cdot W^{(k+1)})_t = \int_0^t H_s dW_s^{(k+1)} \quad P\text{-a.s.}$$

Since  $k \in \mathbb{N}$  was arbitrary, the decomposition (2.5.2) follows, since  $M$  is right-continuous and the right hand side of (2.5.2) is continuous,  $M$  is  $P$ -a.s. continuous.  $\square$

We now include an independent direct proof that every  $M \in \mathcal{M}^2$  (with underlying filtration  $(\mathcal{F}_{t+}^W)^P$ ,  $t \geq 0$ ) is  $P$ -a.s. continuous:

**Proposition 2.5.43.** *Every  $(\mathcal{F}_{t+}^W)$ -adapted  $(M_t) \in \mathcal{M}_{loc}^2$  is  $P$ -a.s. continuous.*

*Proof.* Without loss of generality  $M \in \mathcal{M}^2$  (localization). It suffices to consider the case

$$M_t := E[F | \mathcal{F}_{t+}^W], \quad t \geq 0,$$

where

$$F := \prod_{i=1}^n f_i(W_{t_i}), \quad f_i \in \mathcal{C}_b(\mathbb{R}) \text{ uniformly continuous, } 0 \leq t_1 < \dots < t_n < \infty \quad (\star)$$

and to prove that  $(M_t)$  has a continuous modification. This is enough because  $M_t = E[M_\infty | \mathcal{F}_{t+}^W]$  (since  $M \in \mathcal{M}^2$ ) and  $F$  of type  $(\star)$  are dense in  $\mathcal{L}^2(\Omega, \mathcal{F}_\infty^W, P)$ .

(Exercise, by a monotone class argument: By 2.1.4  $M$  has a continuous  $(\mathcal{F}_{t+}^W)^P$ -adapted version. Define

$$V := \{G \in \mathcal{L}^2(\Omega, \mathcal{F}_\infty^W, P) | \exists F_n \text{ of type } (\star) \text{ such that } F_n \rightarrow G \text{ in } \mathcal{F}_\infty^W, P\}$$

and prove that  $V = \mathcal{L}^\infty(\Omega, \mathcal{F}_\infty^W, P)$ .)

## 2. (Semi-)Martingales and Stochastic Integration

Let  $t \geq 0$  and  $i \in \{1, \dots, n\}$  such that  $t \in [t_i, t_{i+1}[$  where  $t_{n+1} := +\infty$ . Then (cf. section of Markov property of Brownian motion),

$$\begin{aligned}
M_t &= E[F | \mathcal{F}_{t+}^W] \\
&= \prod_{j \leq i} f_j(W_{t_j}) E \left[ \prod_{j > i} f_j(W_{t_j}) \middle| \mathcal{F}_{t+}^W \right] \\
&= \prod_{j \leq i} f_j(W_{t_j}) E_{W_t} \left[ \prod_{j > i} f_j(W_{t_j-t}) \right] \\
&= \prod_{j \leq i} f_j(W_{t_j}) p_{t_{i+1}-t} \underbrace{(f_{i+2}(p_{t_{i+1}-t_{i+2}} f_{i+2} \cdots p_{t_n-t_{n-1}} f_n) \cdots)}_{=: g \in C_b^\infty(\mathbb{R})}(\omega_t) \quad P\text{-a.s.},
\end{aligned}$$

where

$$p_s f(x) := \frac{1}{\sqrt{2\pi s}} \int f(y) e^{-\frac{1}{2s}(x-y)^2} dy.$$

So, we have to prove that  $p_{t_{i+1}-t} g(\omega_t)$  is  $P$ -a.s. continuous in  $t$ . Set

$$\mathcal{C}_{b,u}(\mathbb{R}) := \{\varphi \in \mathcal{C}_b(\mathbb{R}) \mid \varphi \text{ uniformly continuous}\}.$$

(i) Let  $f \in \mathcal{C}_{b,u}(\mathbb{R})$ . Take  $x, y \in \mathbb{R}$ . Then

$$|p_t f(x) - p_t f(y)| \leq \frac{1}{\sqrt{2\pi s}} \int |f(x+z) - f(y+z)| e^{-\frac{|z|^2}{2s}} dz.$$

Let  $\varepsilon > 0$ . Then there exists  $\delta = \delta(\varepsilon) > 0$  such that  $|x-y| < \delta \Rightarrow |f(x) - f(y)| < \varepsilon$ . So,  $|p_t f(x) - p_t f(y)| \leq \varepsilon$  and  $p_s f \in \mathcal{C}_{b,u}(\mathbb{R})$ .

(ii) Let  $f \in \mathcal{C}_{b,u}(\mathbb{R})$ . Then

$$\lim_{s \rightarrow 0} \|p_s f - f\|_\infty = 0.$$

*Proof.* Let  $\varepsilon > 0$  and  $\delta > 0$  such that  $|x-y| < \delta \Rightarrow |f(x) - f(y)| < \varepsilon$ . Since

$$f(x) = \frac{1}{\sqrt{2\pi s}} \int f(y) e^{-\frac{|y|^2}{2s}} dy$$

we have

$$\begin{aligned}
&|p_s f(x) - f(x)| \\
&\leq \frac{1}{\sqrt{2\pi s}} \int |f(x+y) - f(y)| e^{-\frac{|y|^2}{2s}} dy \\
&\leq \frac{1}{\sqrt{2\pi s}} \int_{-\delta}^{\delta} \underbrace{|f(x+y) - f(x)|}_{< \varepsilon} e^{-\frac{|y|^2}{2s}} dy + \frac{1}{\sqrt{2\pi s}} \int_{\{|y| > \delta\}} |f(x+y) - f(x)| e^{-\frac{|y|^2}{2s}} dy \\
&\leq \varepsilon + 2 \|f\|_\infty \frac{1}{\sqrt{2\pi s}} \int_{\{|y| > \delta\}} e^{-\frac{|y|^2}{2s}} dy.
\end{aligned}$$

Therefore,

$$\limsup_{s \rightarrow 0} \|p_s f - f\|_\infty \leq \varepsilon + 2 \|f\|_\infty \limsup_{s \rightarrow 0} \frac{1}{\sqrt{2\pi s}} \int_{\{|y| > \delta\}} e^{-\frac{|y|^2}{2s}} dy \stackrel{y' = \frac{y}{\sqrt{s}}}{=} \varepsilon.$$

□

(iii)

$$p_t(p_s f) = p_{t+s} f,$$

in short,

$$p_t p_s = p_{t+s}.$$

*Proof.* Exercise (by Fourier transform and by use of

$$p_s f(x) = (\varrho_s * f)(x),$$

 where  $\varrho_s(z) = \frac{1}{\sqrt{2\pi s}} e^{-\frac{|z|^2}{2s}}$  and recall  $\hat{\varrho}_s(\xi) = e^{-\frac{s\xi^2}{2}}$ . □

 (iv) For  $t < t_{i+1}$  we have

$$\lim_{s \searrow t} \|p_{t_{i+1}-s} g - p_{t_{i+1}-t} g\|_\infty = 0 \quad (\text{right-continuity}).$$

*Proof.* Let  $h > 0$  such that  $s := t + h < t_{i+1}$ . Then

$$\begin{aligned} \|p_{t_{i+1}-t-h} g - p_{t_{i+1}-t} g\|_\infty &\stackrel{(iii)}{=} \|p_{t_{i+1}-t-h}(g - p_h g)\|_\infty \\ &\leq \|p_{t_{i+1}-t-h}\|_\infty \cdot \|g - p_h g\|_\infty \\ &\stackrel{p_s 1=1}{\leq} \|g - p_h g\|_\infty \xrightarrow{s \rightarrow t} 0 \quad \text{by (ii)}. \end{aligned}$$

□

 (v) Assume  $t_i < t$  and let  $h > 0$  such that  $s := t - h > t_i$ . Then

$$\lim_{s \nearrow t} \|p_{t_{i+1}-s} g - p_{t_{i+1}-t} g\|_\infty = 0.$$

*Proof.* Let  $h > 0$  such that  $s := t - h > t_i$ . Then

$$\|p_{t_{i+1}-t+h} g - p_{t_{i+1}-t} g\|_\infty \stackrel{(iii)}{=} \|p_{t_{i+1}-t}(p_h g - g)\|_\infty \leq \|p_h g - g\|_\infty \xrightarrow{h \rightarrow 0} 0.$$

□

 (vi) Consider  $t := t_i$ :

 Let  $h > 0$ 

$$p_{t_i-(t_i-h)}(f_i p_{t_{i+1}-t_i} f_{i+1} \cdots p_{t_n-t_{n-1}} f_n) = p_h \underbrace{(f_i p_{t_{i+1}-t_i} f_{i+1} \cdots p_{t_n-t_{n-1}} f_n)}_{=\tilde{g} \in C_{b,u}(\mathbb{R})} \xrightarrow{h \rightarrow 0} \tilde{g}$$

 uniformly in  $x$ . Hence, also continuous in  $t = t_i$  from the left uniformly in  $x$ .

**Summary:** The function  $G$  defined by

$$G_t(x) := \prod_{j \leq i} f_j(x) p_{t_{i+1}-t}(f_{i+1} p_{t_{i+2}-t_{i+1}} f_{i+2} \cdots p_{t_n-t_{n-1}} f_n(x) \cdots) \quad \text{for } t \in [t_i, t_{i+1}[$$

 is continuous in  $t \geq 0$  uniformly in  $x$ .

 (vii)  $t \mapsto G(t, \omega_t)$  is continuous  $P$ -a.s..

## 2. (Semi-)Martingales and Stochastic Integration

*Proof.* Take  $|G(t, \omega_t) - G(s, \omega_s)| \leq |G(t, \omega_t) - G(s, \omega_t)| + |G(s, \omega_t) - G(s, \omega_s)|$ . Fix  $s > 0$ . Then the right hand side is dominated by

$$\underbrace{\lim_{t \rightarrow s} \|G(t, \cdot) - G(s, \cdot)\|_\infty}_{=0} + \underbrace{\lim_{t \rightarrow s} \|G(s, \omega_t) - G(s, \omega_s)\|_\infty}_{=0},$$

because  $t \mapsto \omega_t$  is continuous and  $G$  is continuous in  $x$ . □

□

**Corollary 2.5.44** (cf. 2.5.42). *Let  $F \in \mathcal{L}^2(\Omega, \mathcal{F}_{t_0+}^W, P)$ ,  $t_0 > 0$  fixed. Such  $F$  are sometimes called “Wiener functional”. Then there exists an  $H \in \mathcal{L}^2(\bar{\Omega}, \bar{\mathcal{F}}, P_{W(t_0+1)})$ ,  $(\mathcal{F}_{t_+}^W)$ -adapted, such that*

$$F - E[F] = \int_0^{t_0} H_s dW_s.$$

*Proof.* (cf. 2.4.40) Let

$$F_t^W := E[F | \mathcal{F}_{t_+}^W] - E[F | \mathcal{F}_{0+}^W], \quad 0 \leq t \leq t_0.$$

Then  $(F_t^W)_{1 \leq t \leq t_0}$  is a martingale, which by the backward martingale convergence theorem (cf. [Röc11]) is right continuous. Hence, by 2.5.42,

$$F - E[F | \mathcal{F}_{0+}^W] = E[F | \mathcal{F}_{t_0+}^W] - E[F | \mathcal{F}_{0+}^W] = F_{t_0}^W \stackrel{2.5.42}{=} \int_0^{t_0} H_s dW_s.$$

The uniqueness is also clear by martingale property. □

**Example 2.5.45** (Special case: the canonical Model). *Let  $(\Omega, \mathcal{F}, P)$  be the (classical) Wiener space, i.e.  $\Omega = \mathcal{C}([0, 1], \mathbb{R})$ ,  $\mathcal{F}$  its Borel  $\sigma$ -algebra and  $P$  the Wiener measure. Define  $X_t(\omega) := \omega(t)$ ,  $t \in [0, 1]$ ,  $\omega \in \Omega$ , hence  $X$  is Brownian motion,*

$$\mathcal{F}_t := \bigcap_{s>t} \sigma(\sigma(X_r | r \leq s \leq 1), \text{ } P\text{-zero sets in } \mathcal{F}),$$

and

$$\mathcal{F} = \bigcup_{0 < t \leq 1} \mathcal{F}_t.$$

Then, by 2.5.44

$$F \in \mathcal{L}^2(\Omega, \mathcal{F}, P) \Rightarrow F = E[F] + \int_0^1 H_s dX_s.$$

**Example 2.5.46.** *Consider the situation of 2.5.45 the Wiener functional*

$$F := \int_0^1 X_t dt.$$

Because of  $E[F] = \int_0^1 \underbrace{E(X_t)}_{=0} dt = 0$ ,

$$F = \int_0^1 H_s dX_s$$

for some  $H \in \mathcal{L}^2(\bar{\Omega}, \bar{\mathcal{F}}, P_W)$ .

Identification of  $H$ :

Let

$$M_t := E[F|\mathcal{F}_t] = \int_0^t E[X_s|\mathcal{F}_t] ds + \int_t^1 E[X_s|\mathcal{F}_t] ds = \underbrace{\int_0^t X_s ds + X_t(1-t)}_{\text{continuous!}} \quad t \geq 0,$$

and the latter is clearly in  $\mathcal{M}_c^2$ .

**Claim:**  $H_t = 1 - t$ .

*Proof.* In view of 2.4.31(i) it is enough to show

$$\langle M, X \rangle = \int_0^t H_s d\langle X \rangle_s \leq \int_0^t (1-s) ds,$$

since then, by  $\mathcal{M}^2 \stackrel{2.5.42}{=} \mathcal{M}_X^2$ , we have  $M = H.X$ , in particular,

$$\int_0^1 X_t dt = F = M_1 = \int_0^1 (1-s) dX_s.$$

But by Itô's product rule

$$X_t t = \int_0^t X_s ds + \int_0^t s dX_s.$$

Hence

$$d\langle M, X \rangle_t = d\langle X - \int_0^\cdot s dX_s, X \rangle_t = dt - t dt,$$

thus  $\langle M, X \rangle_t = \int_0^t (1-s) ds$ . □



## 3. Markov Processes

### 3.1. Markov Processes and Semigroups

**Definition 3.1.1.** A family  $(p_t)_{t \geq 0}$  of kernels on a measurable space  $(S, \mathcal{S})$  is called a semigroup of kernels if

$$p_t p_s = p_{t+s} \quad \forall t, s \geq 0 \quad (\text{Chapman-Kolmogorov equation}),$$

i.e.

$$p_t p_s(x, A) = \int p_t(x, dy) p_s(y, A) \left( := \int p_s(y, A) p_t(x, dy) \right).$$

If for all  $t, x$   $p_t(x, S) = 1$ ,  $(p_t)_{t \geq 0}$  is called Markovian and, if  $p_t(x, S) \leq 1$  for all  $x, t$ , sub-Markovian, respectively.

**Definition 3.1.2.** Let  $(S, \mathcal{S})$  be a measurable space. A family of stochastic processes

$$(\Omega, \mathcal{F}, (X_t)_{t \geq 0}, (P_x)_{x \in S})$$

with state space  $(S, \mathcal{S})$  is called a Markov process, if

(M1)  $x \mapsto P_x(\Gamma)$  is  $\mathcal{S}$ -measurable for all  $\Gamma \in \mathcal{F}$ ,

(M2) there exists a filtration  $(\mathcal{F}_t)_{t \geq 0}$  such that each  $X_t$  is  $\mathcal{F}_t$ -measurable and

$$P_x(X_{s+t} \in B | \mathcal{F}_s) = P_{X_s}(X_t \in B) \quad P_x\text{-a.e.} \quad \forall s, t \geq 0, B \in \mathcal{S}, x \in S$$

(Markov property with respect to  $(\mathcal{F}_t)_{t \geq 0}$ ).

**Remark 3.1.3.** If (M2) is true for  $(\hat{\mathcal{F}}_t)_{t \geq 0}$  with  $\hat{\mathcal{F}}_t \supset \mathcal{F}_t \quad \forall t \geq 0$ , then (M2) is true for  $(\mathcal{F}_t)_{t \geq 0}$  if  $(X_t)$  is  $(\mathcal{F}_t)$ -adapted!

The following theorem shows, at least if  $(S, \mathcal{S})$  is polish, that Markovian semigroups and Markov processes are in one-to-one correspondence to each other.

**Theorem 3.1.4.** (i) Let  $(S, \mathcal{S})$  be a measurable space and let  $\mathbb{M} = (\Omega, \mathcal{F}, (X_t)_{t \geq 0}, (P_x)_{x \in S})$  be a family of stochastic processes with state space  $(S, \mathcal{S})$  and  $\mathcal{F} = \sigma(X_t | t \geq 0)$ .

a) Suppose there exists a Markovian semigroup of kernels  $(p_t)_{t \geq 0}$ , such that for all  $0 \leq t_0 < t_1 < \dots < t_n$ ,  $f : S^{n+1} \rightarrow \mathbb{R}$ , bounded and  $\mathcal{S}^{n+1}$  measurable, and all  $x \in S$ ,

$$E_x[f(X_{t_0}, \dots, X_{t_n})] = \int_S p_{t_0}(x, dx_0) \cdots \int_S p_{t_n - t_{n-1}}(x_{n-1}, dx_n) f(x_0, \dots, x_n) \quad (3.1.1)$$

Then  $\mathbb{M}$  is a Markov process with respect to  $\mathcal{F}_t := \sigma(X_s | s \leq t)$ ,  $t \geq 0$  (and even (3.1.3) below holds).

b) Suppose  $\mathbb{M}$  is a Markov process and set

$$p_t f(x) := E_x(f(X_t)), \quad x \in S, f : S \rightarrow \mathbb{R} \text{ bounded, } \mathcal{S}\text{-measurable, } t \geq 0. \quad (3.1.2)$$

Then  $(p_t)_{t \geq 0}$  is a Markovian semigroup of kernels on  $(S, \mathcal{S})$  and we have (3.1.1).

### 3. Markov Processes

(ii) If  $(S, \mathcal{S})$  is polish and  $(p_t)_{t \geq 0}$  is a Markovian semigroup of kernels on  $(S, \mathcal{S})$ . Then there exists a Markov process  $\mathbb{M}$  with (3.1.2) and  $\Omega = S^{[0, \infty]}$ .

*Proof.* (i): a) (M1) follows by ‘‘monotone classes’’ from (3.1.1), since  $\mathcal{F} = \sigma(\{X_t | t \geq 0\})$ .  
For (M2) we show an a-priori stronger fact:

$$E_x[f(X_{t_1+s}, \dots, X_{t_n+s}), \Gamma] = E_x[E_{X_s}[f(X_{t_1}, \dots, X_{t_n})], \Gamma] \quad (3.1.3)$$

for all  $x \in S$ , bounded and  $\mathcal{S}^n$ -measurable  $f : S^n \rightarrow \mathbb{R}$ ,  $0 \leq s_0 < s_1 < \dots < s_m = s$  and  $\Gamma \in \mathcal{F}_s, s \geq 0$ . By monotone classes (applied to  $\Gamma$ ) this follows from

$$\begin{aligned} & E_x[f(X_{t_1+s}, \dots, X_{t_n+s})g(X_{s_0}, \dots, X_{s_m})] \\ &= E_x[E_{X_s}[f(X_{t_1}, \dots, X_{t_n})]g(X_{s_0}, \dots, X_{s_m})] \end{aligned}$$

for all bounded  $\mathcal{S}^{m+1}$ -measurable  $g : S^{m+1} \rightarrow \mathbb{R}$ ,  $0 \leq s_0 < s_1 < \dots < s_m = s$ . By (3.1.1) the left hand side is equal to

$$\begin{aligned} & \int p_{s_0}(x, dx_0) \cdot \dots \cdot \int p_{s_m-s_{m-1}}(x_{m-1}, dx_m) \cdot \\ & \underbrace{\int p_{t_1}(x_m, dy_1) \dots \int p_{t_n-t_{n-1}}(y_{n-1}, dy_n) f(y_1, \dots, y_n) g(x_0, x_1, \dots, x_m)}_{=E_{x_m}[f(X_{t_1}, \dots, X_{t_n})]} \\ &= E_x[E_{X_{s_m}}[f(X_{t_1} \dots X_{t_n})]g(X_{s_0}, X_{s_1}, \dots, X_{s_m})]. \end{aligned}$$

b) By (M1) it is clear that (3.1.2) defines a Markovian kernel for all  $t \geq 0$ . Furthermore,

$$\begin{aligned} p_{t+s}f(x) &= E_x[f(X_{t+s})] = E_x[E_x[f(X_{t+s})|\mathcal{F}_s]] \\ &\stackrel{\text{M.P.}}{=} E_x[E_{X_s}[f(X_t)]] = E_x[p_t f(X_s)] \\ &= p_s(p_t f)(x) = (p_s p_t)f(x) \end{aligned}$$

To prove (3.1.1) let  $f = f_0 \otimes f_1 \otimes \dots \otimes f_n$ ,  $f_i : S \rightarrow \mathbb{R}$ , bounded and  $\mathcal{S}$ -measurable. Then for  $0 \leq t_0 < t_1 < \dots < t_n$

$$\begin{aligned} & E_x[f_0(X_{t_0})f_1(X_{t_1}) \dots f_n(X_{t_n})] \\ &= E_x[f_0(X_{t_0})f_1(X_{t_1}) \dots f_{n-1}(X_{t_{n-1}})E_x[f_n(X_{(t_n-t_{n-1})+t_{n-1}})|\mathcal{F}_{t_{n-1}}]] \\ &\stackrel{\text{M.P.}}{=} E_x[f_0(X_{t_0})f_1(X_{t_1}) \dots f_{n-1}(X_{t_{n-1}})E_{X_{t_{n-1}}}[f_n(X_{t_n-t_{n-1}})]] \\ &\stackrel{(3.1.2)}{=} E_x[f_0(X_{t_0})f_1(X_{t_1}) \dots f_{n-1}(X_{t_{n-1}})p_{t_n-t_{n-1}}f_n(X_{t_{n-1}})] \\ &\stackrel{\text{Ind.Hyp.}}{=} \int \dots \int p_{t_0}(x, dx_0)p_{t_1-t_0}(x_0, dx_1) \dots p_{t_{n-1}-t_{n-2}}(x_{n-2}, dx_{n-1}) \\ & \quad f_0(x_0)f_1(x_1) \dots f_{n-1}(x_{n-1})p_{t_n-t_{n-1}}f_n(x_{n-1}) \\ &= \int \dots \int p_{t_0}(x, dx_0)p_{t_1-t_0}(x_0, dx_1) \dots p_{t_{n-1}-t_{n-2}}(x_{n-2}, dx_{n-1})p_{t_n-t_{n-1}}(x_{n-1}, dx_n) \\ & \quad f_0(x_0)f_1(x_1) \dots f_{n-1}(x_{n-1})f_n(x_n). \end{aligned}$$

By monotone classes this equality extends to all  $f : S^n \rightarrow \mathbb{R}$ , bounded, measurable. Hence, (3.1.1) holds.

(ii): see Bauer!

□

Now consider the “canonical model”:

Let  $S$  be a topological space,  $\mathcal{S}$  the Borel- $\sigma$ -algebra and  $\Omega \subset S^{[0, \infty[}$  (e.g.  $\Omega = S^{[0, \infty[}$  or  $\Omega = \mathcal{C}([0, \infty[, S)$  or  $\Omega$  are all bounded continuous paths in  $S$ ). Define

$$X_t(\omega) := \omega(t), \quad t \geq 0, \quad \omega \in \Omega,$$

$$\mathcal{F} := \sigma(X_t | t \geq 0),$$

$$\mathcal{F}_t^0 := \sigma(X_s | s \leq t) \quad (\text{“past”})$$

$$\hat{\mathcal{F}}_t^0 := \sigma(X_s | s \geq t) \quad (\text{“future”}).$$

Let  $\mathbb{M} = (\Omega, \mathcal{F}, (\mathcal{F}_t^0)_{t \geq 0}, (X_t)_{t \geq 0}, \underbrace{(P_x)_{x \in S}}_{\text{all information here!}})$  be a Markov process. Then  $\mathbb{M}$  is called the *canonical model*.

**Definition 3.1.5.** In the canonical model define the shift operator  $\vartheta_t : \Omega \rightarrow \Omega$  for  $t \geq 0$  by

$$\vartheta_t(\omega)(s) := \omega(s + t),$$

i.e.

$$\vartheta_t(\omega) = \omega(\cdot + t).$$

It is obvious that  $\vartheta_t : \Omega \rightarrow \Omega$  is  $\hat{\mathcal{F}}_t^0 / \mathcal{F}$ -measurable and moreover, (exercise)

$$\vartheta_t^{-1}(\mathcal{F}) = \hat{\mathcal{F}}_t^0 \quad \forall t \geq 0.$$

**Lemma 3.1.6.** (i)  $\psi$  is  $\hat{\mathcal{F}}_t^0$ -measurable if and only if there exists an  $\mathcal{F}$ -measurable  $\varphi$  such that

$$\psi = \varphi \circ \vartheta_t.$$

(ii) Suppose  $P_x, x \in S$ , are given and that

$$\mathbb{M} := (\Omega, \mathcal{F}, (\mathcal{F}_t^0)_{t \geq 0}, (X_t)_{t \geq 0}, (P_x)_{x \in S})$$

satisfies (M1). Then (M2) holds if and only if

$$E_x[\varphi \circ \vartheta_t | \mathcal{F}_t^0] = E_{X_t}[\varphi] \quad P_x\text{-a.s. } \forall x \in S, \forall \varphi : \Omega \rightarrow \mathbb{R} \text{ bounded and } \mathcal{F}\text{-measurable.} \quad (3.1.4)$$

*Proof.* (i): Exercise (Factorization lemma).

(ii): “ $\Leftarrow$ ” is clear. (Consider  $\varphi = 1_{\{X_s \in B\}}, B \in \mathcal{S}$ .)

“ $\Rightarrow$ ”: By 3.1.4 (i)(b) we know that (3.1.1) holds. But in the proof of 3.1.4(i) we have shown that (3.1.1) implies (3.1.3). Now Lemma 3.1.6 follows by “monotone classes”.  $\square$

It is clear that (3.1.4) implies (elementary Markov property)

$$E_x[\varphi \circ \vartheta_t | \mathcal{F}_t^0] = E_x[\varphi \circ \vartheta_t | \sigma(X_t)] \quad \forall t \geq 0. \quad (3.1.5)$$

Interpretation of (3.1.5): The conditional expectation of a future observable, given the past, only depends on the present at time  $t$ . The equivalent formulations of (3.1.5) in the following lemma have corresponding interpretations:

**Lemma 3.1.7.** Fix  $x \in S$ . Then the following statements are equivalent to (3.1.5):

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(i)

$$E_x[\varphi_t^0 | \hat{\mathcal{F}}_t^0] = E_x[\varphi_t^0 | \sigma(X_t)] \quad P\text{-a.s.}, \forall t \geq 0, \forall \varphi_t^0 : \Omega \rightarrow \mathbb{R} \text{ bounded and } \mathcal{F}_t^0\text{-measurable,}$$

(ii)

$$E_x[\varphi_t^0 \hat{\varphi}_t^0 | \sigma(X_t)] = E_x[\varphi_t^0 | \sigma(X_t)] E_x[\hat{\varphi}_t^0 | \sigma(X_t)] \quad P_x\text{-a.s.},$$

for all  $t \geq 0$  and for all  $\varphi_t^0, \hat{\varphi}_t^0 : \Omega \rightarrow \mathbb{R}$ , bounded,  $\varphi_t^0$   $\mathcal{F}_t^0$ -measurable and  $\hat{\varphi}_t^0$   $\hat{\mathcal{F}}_t^0$ -measurable.

*Proof.* (3.1.5)  $\Rightarrow$  (i):

$$\begin{aligned} E_x[\varphi_t^0 \hat{\varphi}_t^0] &= E_x[\varphi_t^0 E_x[\hat{\varphi}_t^0 | \mathcal{F}_t^0]] \\ &\stackrel{3.1.6(i),(3.1.5)}{=} E_x[\varphi_t^0 E_x[\hat{\varphi}_t^0 | \sigma(X_t)]] \\ &= E_x[E_x[\varphi_t^0 | \sigma(X_t)] E_x[\hat{\varphi}_t^0 | \sigma(X_t)]] \\ &= E_x[\hat{\varphi}_t^0 E_x[\varphi_t^0 | \sigma(X_t)]] \quad \forall \text{ bounded } \hat{\mathcal{F}}_t^0\text{-measurable } \hat{\varphi}_t^0 : \Omega \rightarrow \mathbb{R}. \end{aligned}$$

Therefore,  $E_x[\varphi_t^0 | \sigma(X_t)]$  is a  $P_x$ -version of  $E_x[\varphi_t^0 | \hat{\mathcal{F}}_t^0]$ . Hence, (i) holds.

(i)  $\Rightarrow$  (ii):

Let  $f : S \rightarrow \mathbb{R}$  be bounded and  $\mathcal{S}$ -measurable and  $\hat{\varphi}_t^0 : \Omega \rightarrow \mathbb{R}$  be bounded,  $\hat{\mathcal{F}}_t^0$ -measurable. Then

$$\begin{aligned} E_x[\varphi_t^0 \hat{\varphi}_t^0 f(X_t)] &= E_x[E_x[\varphi_t^0 | \hat{\mathcal{F}}_t^0] \hat{\varphi}_t^0 f(X_t)] \stackrel{(i)}{=} E_x[E_x[\varphi_t^0 | \sigma(X_t)] \hat{\varphi}_t^0 f(X_t)] \\ &= E_x[E_x[\varphi_t^0 | \sigma(X_t)] E_x[\hat{\varphi}_t^0 | \sigma(X_t)] f(X_t)]. \end{aligned}$$

(ii)  $\Rightarrow$  (3.1.5):

$$\begin{aligned} E_x[\varphi_t^0 \hat{\varphi}_t^0] &\stackrel{(ii)}{=} E_x[E_x[\varphi_t^0 | \sigma(X_t)] E_x[\hat{\varphi}_t^0 | \sigma(X_t)]] \\ &= E_x[\varphi_t^0 E_x[\hat{\varphi}_t^0 | \sigma(X_t)]] \quad \forall \text{ bounded } \mathcal{F}_t^0\text{-measurable } \varphi_t^0 : \Omega \rightarrow \mathbb{R}. \end{aligned}$$

Hence,

$$E_x[\hat{\varphi}_t^0 | \sigma(X_t)] = E_x[\hat{\varphi}_t^0 | \mathcal{F}_t^0],$$

and (3.1.5) follows by Lemma 3.1.6(i).  $\square$

**Remark 3.1.8.** Consider the situation of 3.1.4(i)(b). Then  $P_x(X_0 = x) = 1$  if and only if  $p_0(x, \cdot) = \delta_x$  (Dirac-measure in  $x$ ). If this holds for every  $x \in S$ , then  $M$  is called normal. In this case,  $p_t(x, A) = P_x(X_t \in A)$  is the probability to be at time  $t$  in  $A$  starting in  $x$  (transition probability).

## 3.2. The Strong Markov Property

Consider the ‘‘canonical model’’. ( $\leftarrow$  only needed to have  $\vartheta_t$ , can be done abstractly!)

**Recall:** Let  $(\mathcal{F}_t)$  be some filtration and  $T$  an  $(\mathcal{F}_t)$ -stopping time. Define the  $\sigma$ -field of the  $T$ -past by

$$\mathcal{F}_T := \{A \in \mathcal{F} | A \cap \{T \leq t\} \in \mathcal{F}_t \forall t \geq 0\}.$$

**Definition 3.2.9.** Let  $\mathbb{M} = (\Omega, \mathcal{F}, (X_t)_{t \geq 0}, (P_x)_{x \in S}, (\vartheta_t)_{t \geq 0})$  be a canonical Markov process. Then  $\mathbb{M}$  satisfies the strong Markov property (SMP), if there exists a right-continuous filtration  $(\mathcal{F}_t)_{t \geq 0}$  on  $(\Omega, \mathcal{F})$  (i.e.  $\mathcal{F}_t = \bigcap_{\varepsilon > 0} \mathcal{F}_{t+\varepsilon}$ ) such that for all  $(\mathcal{F}_t)$ -stopping times  $T$  we have that  $\vartheta_T |_{\{T < \infty\}}$  is  $\mathcal{F} \cap \{T < \infty\} / \mathcal{F}$ - and  $\mathcal{F}_{T+s} \cap \{T < \infty\} / \mathcal{F}_s$ -measurable for all  $s \leq 0$  and we have (SMP)

$$E_x[1_{\{T < \infty\}} \varphi \circ \vartheta_T | \mathcal{F}_T] = 1_{\{T < \infty\}} E_{X_T}[\varphi] \quad P\text{-a.s. } \forall \varphi \text{ bounded, } \mathcal{F}\text{-measurable and } \forall x \in S.$$

**Remark 3.2.10.** (i) It is clear that (SMP) implies the Markov property with respect to  $(\mathcal{F}_t)$  and therefore, with respect to  $(\mathcal{F}_t^0)$  by 3.1.6(ii). In general, the converse is false.

(ii) (SMP) holds if and only if for all  $(\mathcal{F}_t)$ -stopping times  $T$  we have

$$E_x[\varphi \circ \vartheta_T; T < \infty] = E_x[E_{X_T}[\varphi]; T < \infty] \quad P_x\text{-a.s.} \quad (3.2.6)$$

for all  $\varphi : \Omega \rightarrow \mathbb{R}$ , bounded and  $\mathcal{F}$ -measurable, and for all  $x \in S$ .

*Proof.* “ $\Rightarrow$ ” is clear.

“ $\Leftarrow$ ”: Let  $A \in \mathcal{F}_T$  and  $\tilde{T} := T \cdot 1_A + \infty \cdot 1_{A^c}$ . Then  $\tilde{T}$  is a  $(\mathcal{F}_t)$ -stopping time because  $\forall t \geq 0$

$$\{\tilde{T} \leq t\} = A \cap \{T \leq t\} \in \mathcal{F}_t.$$

Hence

$$\begin{aligned} E_x[\varphi \circ \vartheta_T; A \cap \{T < \infty\}] &= E_x[\varphi \circ \vartheta_T; \tilde{T} < \infty] \\ &= E_x[\varphi \circ \vartheta_{\tilde{T}}; \tilde{T} < \infty] \\ &\stackrel{(3.2.6)}{=} E_x[E_{X_{\tilde{T}}}(\varphi); \tilde{T} < \infty] \\ &= E_x[E_{X_T}(\varphi); A \cap \{T < \infty\}]. \end{aligned}$$

□

**Proposition 3.2.11.** Suppose  $S$  is a topological space and  $\mathcal{S}$  Borel  $\sigma$ -algebra such that  $\mathcal{S} = \sigma(\mathcal{C}_b(S))$ . Let  $\mathbb{M} = (\Omega, \mathcal{F}, (X_t)_{t \geq 0}, (P_x)_{x \in S}, (\vartheta_t)_{t \geq 0})$  be a (canonical) Markov process with

(i) right-continuous paths,

(ii)  $p_t(\mathcal{C}_b(S)) \subset C_b(S)$  (“Feller property”),

where  $p_t$  is the corresponding semigroup. Then  $\mathbb{M}$  fulfills the (SMP) with respect to  $\mathcal{F}_t := \bigcap_{\varepsilon > 0} \mathcal{F}_{t+\varepsilon}^0$ ,  $t \geq 0$  (right-continuous).

*Proof.* Let  $T$  be an  $(\mathcal{F}_t)$ -stopping time. By [Röc11]  $X_T$  is  $\mathcal{F}_T$ -measurable since  $X$  has right-continuous sample paths. Similarly, one can prove that  $\vartheta_T$  has the required measurability properties. Define

$$T_n := \sum_{k=1}^{\infty} \frac{k}{2^n} 1_{\{\frac{k-1}{2^n} \leq T < \frac{k}{2^n}\}} + \infty \cdot 1_{\{T=\infty\}}.$$

Then  $T_n \searrow T$ . We have to show that (3.2.6) holds:

Without loss of generality (otherwise use monotone classes)

$$\varphi = f_0(X_{t_0})f_1(X_{t_1}) \dots f_n(X_{t_n}), \quad f_i \in \mathcal{C}_b(S), \quad t_0 \leq t_1 \leq \dots \leq t_n.$$

Furthermore,

$$\{T < \infty\} = \bigcup_{k \in \mathbb{N}} \left\{ \frac{k-1}{2^n} \leq T < \frac{k}{2^n} \right\} \quad (3.2.7)$$

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Then

$$\begin{aligned}
E_x[\varphi \circ \vartheta_T, T < \infty] &\stackrel{(i)}{=} \lim_{n \rightarrow \infty} E_x[\varphi \circ \vartheta_{T_n}, T < \infty] \\
&\stackrel{(3.2.7)}{=} \lim_{n \rightarrow \infty} \sum_{k=1}^{\infty} E_x \left[ \varphi \circ \underbrace{\vartheta_{T_n}}_{=\vartheta_{k2^{-n}}}, \frac{k-1}{2^n} \leq T < \frac{k}{2^n} \right] \\
&\stackrel{M.P.}{=} \lim_{n \rightarrow \infty} \sum_{k=1}^{\infty} E_x \left[ E_{X_{k2^{-n}}}[\varphi]; \frac{k-1}{2^n} \leq T < \frac{k}{2^n} \right] \\
&\stackrel{(3.2.7)}{=} \lim_{n \rightarrow \infty} E_x[E_{X_{T_n}}[\varphi], T < \infty] \\
&= E_x[E_{X_T}[\varphi]; T < \infty] \quad (\text{Lebesgue}),
\end{aligned}$$

since  $X_{T_n} \rightarrow X_T$  on  $[T < \infty]$  and

$$x \mapsto E_x[\varphi] = E_x[f_0(X_{t_0}), \dots, f_n(X_{t_n})] = p_{t_0}(f_0 p_{t_1-t_0} f_1 p_{t_2-t_1} f_2 \dots p_{t_n-t_{n-1}} f_n)(x)$$

is continuous on  $S$  by (ii). □

**Remark 3.2.12** (Blumenthal's 0-1 law). *Let  $\mathbb{M}$  be a canonical normal Markov process with respect to  $\mathcal{F}_t := \bigcap_{s>t} \mathcal{F}_s^0$ ,  $t \geq 0$  (on measurable state space  $(S, \mathcal{S})$ ). Then  $P_x = 0$  or  $P_x = 1$  on  $\mathcal{F}_0$  for every  $x \in S$ .*

*Proof.* Let  $\varphi$  be  $\mathcal{F}_0$ -measurable and bounded. Then for all  $x \in S$  we have

$$\varphi = E_x[\varphi | \mathcal{F}_0] = E_x[\varphi \circ \vartheta_0 | \mathcal{F}_0] \stackrel{M.P.}{=} E_{X_0}[\varphi] = E_x[\varphi] \quad P_x\text{-a.s.}$$

Hence,  $\varphi$  is constant  $P_x$ -a.e.. □

### 3.3. Application to Brownian Motion

Let  $S = \mathbb{R}^d$ ,  $\mathcal{S} = \mathcal{B}(\mathbb{R}^d)$  and

$$\begin{aligned}
p_t(x, A) &= \int_A \underbrace{\frac{1}{(2\pi t)^{\frac{d}{2}}} e^{-\frac{|y-x|^2}{2t}}}_{=: g_t(y-x)} dy, \quad x \in \mathbb{R}^d, t > 0, A \in \mathcal{B}(\mathbb{R}^d), \\
p_0(x, \cdot) &= \delta_x.
\end{aligned}$$

**Lemma 3.3.13.** *For all  $t, x \geq 0$ , we have*

$$p_t p_s = p_{t+s}.$$

*Proof.* Exercise (by Fourier transform). □

Let  $\Omega := \mathcal{C}([0, \infty[, \mathbb{R}^d)$ ,  $X_t(\omega) := \omega(t)$ ,  $P_0 := P$  be the Wiener measure on  $\Omega$ ,  $\mathcal{F} := \sigma(X_t | t \geq 0)$  and  $P_x$  the image measure of  $P_0$  under  $\omega \mapsto \omega + x$ . Then we have (cf. [Röc11])

$$E_x[f(X_{t_0}, \dots, X_{t_n})] = \int p_{t_0}(x, dx_0) p_{t_1-t_0} p_{t_n-t_{n-1}}(x_{n-1}, dx_n) f(x_0, \dots, x_n),$$

for all  $x \in \mathbb{R}^d$ , for all  $\mathcal{B}((\mathbb{R}^d)^{n+1})$ -measurable, bounded functions  $f : (\mathbb{R}^d)^{n+1} \rightarrow \mathbb{R}$  and for  $0 \leq t_0 \leq t_1 \leq \dots \leq t_n$ . Hence,

$$\mathbb{M} := (\Omega, \mathcal{F}, (X_t)_{t \geq 0}, (P_x)_{x \in \mathbb{R}^d})$$

is by Proposition 3.1.4(i)(a) a Markov Process on  $(\mathbb{R}^d, \mathcal{B}(\mathbb{R}^d))$ .  $\mathbb{M}$  has continuous sample paths and is normal. In particular, Remark 3.2.12 is fulfilled. But also, by Proposition 3.2.11  $\mathbb{M}$  has SMP with respect to

$$\mathcal{F}_t = \bigcap_{s>t} \mathcal{F}_s^0, \quad t \geq 0,$$

because of the following Proposition.

**Proposition 3.3.14.**  $(p_t)_{t \geq 0}$  is strong Feller, that is,

$$p_t f \in \mathcal{C}_b(\mathbb{R}^d) \quad \forall f \in \mathcal{B}_b(\mathbb{R}^d).$$

(Moreover, we have  $p_t f \in \mathcal{C}^\infty$  for all  $t > 0$  and  $f \in \mathcal{B}_b(\mathbb{R}^d)$ .)

*Proof.* Fix  $t > 0$ . Let  $x \in \mathbb{R}^d$  and  $x_n \rightarrow x$ . We have to show that  $p_t f(x_n) \rightarrow p_t f(x)$ . Let  $n_0$  such that  $x_n \in B_1(x)$  for all  $n \geq n_0$ . There exists  $h \in \mathcal{L}^1(\mathbb{R}) := \mathcal{L}^1(\mathbb{R}, dx)$  (with  $dx =$  Lebesgue measure) such that  $e^{-\frac{(x'-y)^2}{2t}} \leq h(y)$  for all  $y, x' \in \mathbb{R}$  such that  $|x' - x| \leq 1$ . Define  $g(\tilde{x}_1, \dots, \tilde{x}_d) = h(\tilde{x}_1) \cdots h(\tilde{x}_d)$ ,  $\tilde{x} = (\tilde{x}_1, \dots, \tilde{x}_d) \in \mathbb{R}^d$ , then

$$e^{-\frac{|x'-y|^2}{2t}} \leq g(y) \quad \forall x' \in B_1(x) \forall y \in \mathbb{R}^d.$$

Hence

$$\lim_{n \rightarrow \infty} p_t f(x_n) = \frac{1}{\sqrt{(2\pi t)^d}} \lim_{n \rightarrow \infty} \int f(y) e^{-\frac{|y-x_n|^2}{2t}} dy = \frac{1}{\sqrt{(2\pi t)^d}} \int f(y) e^{-\frac{|y-x|^2}{2t}} dy = p_t f(x)$$

by Lebesgue's dominated convergence theorem since  $f$  is bounded and  $e^{-\frac{|x'-y|^2}{2t}} \leq g(y)$  and  $g \in \mathcal{L}^1(\mathbb{R})$ .  $\square$

**Corollary 3.3.15.** (i) Let  $\mathbb{M}$  be a normal (canonical) Markov-Process with respect to  $\mathcal{F}_t := \bigcap_{s>t} \mathcal{F}_s^0$ ,  $t \geq 0$ . Then, by Blumenthal for  $t$  not fixed,

$$P_x \left[ \underbrace{\limsup_{t \searrow 0} 1_{\{X_t=x\}}}_{\in \mathcal{F}_0} = 1 \right] \in \{0, 1\}.$$

For Brownian motion this probability is equal to 1 because of the law of iterated logarithm.

(ii) Let  $X$  be Brownian motion starting at  $x = 0$  and let

$$N(\omega) := \{0 \leq s \leq 1 \mid X_s(\omega) = 0\}.$$

Then for  $P_0$ -a.e.  $\omega$  we have that

- a)  $N(\omega)$  is closed,
- b)  $N(\omega)$  has Lebesgue-measure zero,
- c)  $N(\omega)$  has no isolated points,

that is  $N(\omega)$  is a Cantor set for  $P_0$ -a.e.  $\omega$ , i.e. every point of  $N(\omega)$  is a cluster point of  $N(\omega)$ .

(iii) Let  $U \subset \mathbb{R}^d$ ,  $U$  open and bounded and let  $X$  be Brownian motion on  $\mathbb{R}^d$ . Define the first exit time of  $U$

$$T := \inf\{t > 0 \mid X_t \notin U\} = \sigma_{U^c}.$$

$T$  is an  $(\mathcal{F}_t)$ -stopping time. Then there exists  $\varepsilon > 0$  such that

$$\sup_{x \in U} E_x[e^{\varepsilon T}] < \infty.$$

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*Proof.* (i) Clear.

(ii) a)  $X$  is  $P$ -a.s. continuous.

b) By Fubini, we get

$$\begin{aligned} \int \int_0^1 1_{N(\omega)}(s) ds P_0(d\omega) &= E_0 \left[ \int_0^1 1_{\{0\}} \circ X_s ds \right] = \int_0^1 P_0 \circ X_s^{-1}(\{0\}) ds \\ &= \int_0^1 p_s(0, \{0\}) ds = 0. \end{aligned}$$

c) (Intuitively clear: by law of (local) iterated logarithm  $t = 0$  is not an isolated point of  $N(\omega)$ ). By (SMP) this is true for any  $s \in N(\omega)$ ). Define

$$\begin{aligned} J &:= \{\omega | N(\omega) \text{ has isolated points}\} \\ &= \bigcup_{r,s \in \mathbb{Q}, r < s} \underbrace{\{\omega | N(\omega) \cap ]r, s[ \text{ contains exactly one (isolated) point}\}}_{A_{r,s}}. \end{aligned}$$

Let  $T_{\{0\}} := \inf\{t > 0 | X_t = 0\}$ . Then  $T_{\{0\}}$  is an  $(\mathcal{F}_t)$  ( $= \bigcap_{s>t} \mathcal{F}_s^0$ )-stopping time (exercise). Therefore, for  $T := T_{\{0\}} \circ \vartheta_r + r$  (first hitting time of 0 after time  $r$ ,  $(\mathcal{F}_t)$ -stopping time!)

$$\begin{aligned} P_0[A_{r,s}] &\leq P_0[r + T_{\{0\}} \circ \vartheta_r < s, T_{\{0\}} \circ \vartheta_{r+T_{\{0\}} \circ \vartheta_r} > 0] \\ &= E_0[1_{\{T_{\{0\}} > 0\}} \circ \vartheta_T; T < s] \\ &= E_0[E_0[1_{\{T_{\{0\}} > 0\}} \circ \vartheta_T | \mathcal{F}_T]; T < s] \\ &\stackrel{\text{SPM}}{=} E_0[E_{X_T}[1_{\{T_{\{0\}} > 0\}}], T < s] \\ &\stackrel{X_T=0}{=} E_0[\underbrace{P_0[T_{\{0\}} > 0]}_{=0}; T < s] = 0, \end{aligned}$$

by (i).

(iii) Let  $R > 0$  such that  $U \subset B_R$  (closed ball with radius  $R$  around 0). Define

$$T_R := \inf\{t > 0 | |X_t| > R\} = \sigma_{B_R^c}.$$

By Ito's formula (or Doob-Meyer),  $|X_t|^2 - d \cdot t$  is a martingale under  $P_x$ . Hence,

$$\begin{aligned} 0 \leq |x|^2 &= E_x[|X_0|^2 - d \cdot 0] = E_x[|X_{T_R \wedge n}|^2 - d \cdot T_R \wedge n] \\ &= E_x[|X_{T_R \wedge n}|^2] - d \cdot E_x[T_R \wedge n] \leq R^2 - d \cdot E_x[T_R \wedge n], \quad \forall x \in U, \end{aligned}$$

Therefore, taking  $n$  to  $\infty$ , we get

$$E_x[T] \leq E_x[T_R] \leq \frac{R^2}{d} < \infty, \quad \forall x \in U.$$

Hence, for large  $t_0$

$$\sup_{x \in U} P_x[T > t_0] \leq \sup_{x \in U} \frac{1}{t_0} E_x[T] < 1.$$

Define

$$\varphi(t) := \sup_{x \in U} P_x[T > t] \quad (\text{decreasing in } t!)$$

Then, we have

$$\begin{aligned}
 \varphi(t+s) &= \sup_{x \in U} E_x[1_{\{T > t+s\}}] \\
 &= \sup_{x \in U} E_x[1_{\{T > s\}} \cdot 1_{\{T > t\}} \circ \vartheta_s] \\
 &= \sup_{x \in U} E_x[E_x[1_{\{T > t\}} \circ \vartheta_s | \mathcal{F}_s]; T > s] \quad (\text{since } \{T > s\} \in \mathcal{F}_s) \\
 &\stackrel{\text{MP}}{=} \sup_{x \in U} E_x[P_{X_s}[T > t]; T > s] \\
 &\leq \sup_{x \in U} P_x[T > t] \cdot \sup_{x \in U} P_x[T > s] = \varphi(t)\varphi(s).
 \end{aligned}$$

**Claim:** These two conditions, i.e.

- a) there exists a  $t_0 > 0$  such that  $\varphi(t_0) < 1$ ,
- b)

$$\varphi(t+s) \leq \varphi(t)\varphi(s), \quad (\star)$$

imply that  $\varphi$  is subexponential, i.e. there exist  $K > 0$  and  $\lambda > 0$  such that

$$\varphi(t) \leq K e^{-\lambda t} \quad \forall t \geq t_0.$$

*Proof.* We have

$$\varphi(s) = \varphi\left(\frac{s}{t_0} t_0\right) \stackrel{\varphi \text{ is decreasing}}{\leq} \varphi\left(\left\lfloor \frac{s}{t_0} \right\rfloor \cdot t_0\right)$$

Hence, inequality  $(\star)$  implies

$$\varphi(s) \leq \underbrace{\varphi(t_0 + \dots + t_0)}_{\lfloor \frac{s}{t_0} \rfloor \text{-times}} \leq \underbrace{\varphi(t_0) \cdot \dots \cdot \varphi(t_0)}_{\lfloor \frac{s}{t_0} \rfloor \text{-times}} = \varphi(t_0)^{\lfloor \frac{s}{t_0} \rfloor}.$$

From now on, without loss of generality we may assume  $\varphi(t_0) > 0$ . Then, since  $\left\lfloor \frac{s}{t_0} \right\rfloor \geq \frac{s}{t_0} - 1$ , it follows that

$$\varphi(t_0)^{\lfloor \frac{s}{t_0} \rfloor} = e^{\lfloor \frac{s}{t_0} \rfloor \ln \varphi(t_0)} \leq \exp\left(-s \frac{|\ln \varphi(t_0)|}{t_0}\right) \exp(|\ln \varphi(t_0)|) = K e^{-s\lambda},$$

where

$$K := \exp(|\ln \varphi(t_0)|) \text{ and } \lambda := \frac{|\ln \varphi(t_0)|}{t_0}.$$

□

Then for all  $x \in U$  and any  $\varepsilon > 0$

$$\begin{aligned}
 E_x[e^{\varepsilon T}] &= E_x\left[\int_0^{e^{\varepsilon T}} 1 \, ds\right] = E_x\left[\int_0^\infty 1_{[0, e^{\varepsilon T}]}(s) \, ds\right] \\
 &\stackrel{\text{Fubini}}{=} \int_0^\infty P_x[e^{\varepsilon T} > s] \, ds \\
 &= \int_0^{e^{t_0 \varepsilon}} P_x[e^{\varepsilon T} > s] \, ds + \int_{e^{t_0 \varepsilon}}^\infty P_x[e^{\varepsilon T} > s] \, ds.
 \end{aligned}$$

### 3. Markov Processes

For the first term we have easily

$$\int_0^{e^{t_0\varepsilon}} P_x[e^{\varepsilon T} > s] ds \leq e^{t_0\varepsilon} < \infty.$$

But, we can estimate the second:

$$\begin{aligned} \int_{e^{t_0\varepsilon}}^{\infty} P_x[e^{\varepsilon T} > s] ds &\leq \int_{e^{t_0\varepsilon}}^{\infty} \sup_{x \in U} P_x[e^{\varepsilon T} > s] ds \\ &= \int_{t_0}^{\infty} \varepsilon e^{\varepsilon u} \underbrace{\sup_{x \in U} P_x[e^{\varepsilon T} > e^{\varepsilon u}]}_{=P_x[T > u]} du \\ &= \varepsilon \int_{t_0}^{\infty} e^{\varepsilon u} \varphi(u) du \\ &\leq \varepsilon \cdot K \int_{t_0}^{\infty} e^{u(\varepsilon - \lambda)} du < \infty, \end{aligned}$$

if  $\varepsilon < \lambda$ .

□

### 3.4. Sojourn Time

Let  $S$  be a polish space,  $\mathcal{S}$  the Borel- $\sigma$ -algebra and  $\mathbb{M} = (\Omega, \mathcal{F}, (X_t)_{t \geq 0}, (P_x)_{x \in S})$  a normal Markov process with respect to a right-continuous filtration  $(\mathcal{F}_t)_{t \geq 0}$  with state space  $S$  and continuous paths.

**Definition 3.4.16.** Fix  $x \in S$ . The time

$$\tau_x := \inf\{t > 0 | X_t \neq x\} = \sigma_{\{x\}}^c$$

is called sojourn time.

Clearly, by (right) continuity of the sample paths

$$\tau_x = 0 \text{ on } \{X_0 = y\}, y \neq x.$$

So,  $\tau_x$  is only non-trivial under  $P_x$ .

**Remark 3.4.17.**  $\tau_x$  is an  $(\mathcal{F}_t)_{t \geq 0}$ -stopping time because for all  $t \geq 0$

$$\{\tau_x \geq t\} = \{X_s = x | 0 \leq s \leq t, s \in \mathbb{Q}\} \in \mathcal{F}_t,$$

hence  $\{\tau_x < t\} \in \mathcal{F}_t$ . Since  $\mathcal{F}_t = \mathcal{F}_{t+}$ , also

$$\{\tau_x \leq t\} = \bigcap_{k \in \mathbb{N}} \underbrace{\{\tau_x < t + \frac{1}{k}\}}_{\in \mathcal{F}_{t + \frac{1}{k}}} \in \mathcal{F}_{t+} = \mathcal{F}_t.$$

**Proposition 3.4.18.**

$$P_x[\tau_x > t] = \exp(-c_x \cdot t), t \geq 0,$$

where

$$c_x := -\ln P_x[\tau_x > 1] \in [0, \infty],$$

i.e.  $\tau_x$  is exponentially distributed with constant  $c_x$  (cf. [KL99]).

In particular, if the Markov process started at  $x$  does not move for sure before time  $t = 1$ , it for sure never moves. If it moves for sure before time  $t = 1$ , it for sure moves immediately.

*Proof.* (By Markov property.) Define for  $t \geq 0$

$$f(t) := P_x[\tau_x > t].$$

**Claim:**  $f(t + s) = f(t)f(s)$  for all  $t, s \geq 0$ .

Indeed,

$$\begin{aligned} E_x[1_{\{\tau_x > t+s\}} | \mathcal{F}_s] &= E_x[1_{\{\tau_x > t\}} \circ \vartheta_s | \mathcal{F}_s] \cdot 1_{\{\tau_x > s\}} \\ &\stackrel{\text{M.P.}}{=} E_{X_s}[1_{\{\tau_x > t\}}] 1_{\{\tau_x > s\}} \stackrel{X_s=x \text{ on } \{\tau_x > s\}}{=} P_x[\tau_x > t] \cdot 1_{\{\tau_x > s\}}. \end{aligned}$$

Hence,

$$\begin{aligned} f(t + s) &= P_x[\tau_x > t + s] = E_x[E_x[1_{\{\tau_x > t+s\}} | \mathcal{F}_s]] \\ &= E_x[P_x[\tau_x > t] 1_{\{\tau_x > s\}}] \\ &= P_x[\tau_x > t] P_x[\tau_x > s] = f(t)f(s). \end{aligned}$$

In particular,

$$f(1) = f\left(2^n \frac{1}{2^n}\right) = f\left(\frac{1}{2^n}\right)^{2^n}$$

and then

$$f\left(\frac{k}{2^n}\right) = f\left(\frac{1}{2^n}\right)^k = f(1)^{\frac{k}{2^n}}.$$

Hence, by the right-continuity of  $f$  we have  $f(t) = f(1)^t = e^{t \ln f(1)}$ . □

**Corollary 3.4.19.** *Let  $T_x = \inf\{t > 0 | X_t = x\} = \sigma_{\{x\}}$ . Let  $\mathbb{M}$  be as in Proposition 3.4.18, but assume in addition, that  $\mathbb{M}$  has (SMP) with respect to  $(\mathcal{F}_t)_{t \geq 0}$  (right-continuity!). Let  $y \in S$  such that*

$$P_y[T_x < \infty] > 0.$$

*Then for all  $u \geq 0$ , we have*

$$P_y[X_s = x, \forall s \in [T_x, T_x + u); T_x < \infty] = e^{-c_x u} \cdot P_y[T_x < \infty],$$

*where  $c_x$  is as in Proposition 3.4.18.*

*Proof.* Let  $\varphi := 1_{\{X_s = x, \forall s \in [0, u)\}}$ . Then

$$\begin{aligned} P_y[X_s = x, \forall s \in [T_x, T_x + u); T_x < \infty] &= E_y[\varphi \circ \vartheta_{T_x}; T_x < \infty] \\ &= E_y[E_y[\varphi \circ \vartheta_{T_x} | \mathcal{F}_{T_x}]; T_x < \infty] \\ &= E_y[E_{X_{T_x}}[\varphi]; T_x < \infty] \\ &= E_x[\varphi] \cdot P_y[T_x < \infty] \\ &= P_x[X_s = x, \forall s \in [0, u)] \cdot P_y[T_x < \infty] \\ &\stackrel{3.4.18}{=} e^{-c_x u} P_y[T_x < \infty]. \end{aligned}$$

□



## 4. Girsanov Transformation

### 4.1. Problem Outline ( $d = 1$ )

We want to construct a process such that it solves (in a “weak” sense) the following equation (“law of motion for the stochastic dynamics  $(X_t)_{t \geq 0}$ ”):

$$\begin{aligned} dX_t &= b(X_t, t) dt + dW_t, \\ X_0 &= x_0 \in \mathbb{R}^d, \end{aligned}$$

that is,

$$X_t(\omega) = x_0 + \int_0^t b(X_s(\omega), s) ds + W_t(\omega).$$

Here,  $X_t = x_0 + \int_0^t b(X_s, s) ds$  denotes the deterministic part and  $W_t$  the stochastic perturbation, i.e.  $W_t$  is a Wiener process.

One possible strategy of solving this equation is to find a strong solution, that is, for a *given* Wiener process  $(W_t)_{t \geq 0}$  on a given probability space  $(\Omega, \mathcal{F}, P)$  construct the paths  $(X_t)_{t \geq 0}$  of the solution by classical methods (e.g. Picard-Lindelöf or Euler scheme).

**Example:** The *Ornstein-Uhlenbeck process* has the “law of motion”

$$\begin{aligned} dX_t &= -\alpha X_t dt + dW_t, \quad \alpha > 0, \\ X_0 &= x_0 \in \mathbb{R}. \end{aligned} \tag{4.1.1}$$

**Heuristics:**

$$\begin{aligned} (4.1.1) \quad &\stackrel{\text{Ito's product rule}}{\Rightarrow} d(e^{\alpha t} X_t) = \alpha e^{\alpha t} X_t dt + e^{\alpha t} dX_t \\ &\stackrel{(4.1.1)}{=} e^{\alpha t} dW_t \\ &\Rightarrow e^{\alpha t} X_t = x_0 + \int_0^t e^{\alpha s} dW_s \\ &\Rightarrow X_t = e^{-\alpha t} x_0 + \int_0^t e^{-\alpha(t-s)} dW_s. \end{aligned}$$

**Claim:** (4.1.1) has a strong solution

$$X_t := e^{-\alpha t} x_0 + \underbrace{\int_0^t e^{-\alpha(t-s)} dW_s}_{\text{“stochastic convolution”}} = F(x_0, (W_s)_{s \leq t})(t),$$

hence,  $X_t$  is adapted to the Wiener filtration.

*Proof.* We apply Itô’s product rule to  $X_t = e^{-\alpha t} \cdot \left(x_0 + \int_0^t e^{\alpha s} dW_s\right)$  to get

$$\begin{aligned} X_t &= x_0 + \int_0^t e^{-\alpha s} d \left( x_0 + \int_0^s e^{\alpha u} dW_u \right) + \int_0^t \underbrace{\left( x_0 + \int_0^s e^{\alpha u} dW_u \right)}_{= X_s \cdot e^{\alpha s} \text{ by definition}} (-\alpha) e^{-\alpha s} ds \\ &= x_0 - \alpha \int_0^t X_s ds + \int_0^t 1 dW_s = x_0 - \alpha \int_0^t X_s ds + W_t. \end{aligned}$$

□

#### 4. Girsanov Transformation

Instead of strong solutions one can construct “(probabilistically) weak solutions”. We want to construct a Brownian motion  $(W_t)_{t \geq 0}$  and a process  $(X_t)_{t \geq 0}$  on some probability space  $(\Omega, \mathcal{F}, P)$  such that

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

that is, construct  $(X_t)_{t \geq 0}$  on a suitable probability space  $(\Omega, \mathcal{F}, P)$  such that

$$W_t := X_t - x_0 - \int_0^t b(X_s, s) ds$$

is a Brownian motion, e.g. take  $(\Omega, \mathcal{F})$ ,  $(X_t)_{t \in [0, t]}$  canonical, i.e.,

$$\Omega := \mathcal{C}([0, 1]),$$

$$X_t(\omega) := \omega(t),$$

$$\mathcal{F} := \sigma(X_t | t \geq 0),$$

such that

$$W_t(\omega) := X_t(\omega) - x_0 - \int_0^t b(X_s(\omega), s) ds (= G((X_s(\omega))_{s \leq t})) \quad (4.1.2)$$

is a Brownian motion under  $P$ .

But we have to identify  $P$ !

One technique to find  $P$  is using the Girsanov transformation. This approach has the following advantages:

- One can do this even if the dependence on the past is very complicated (i.e.  $b(X_s(\omega), s)$  replaced by  $b_s$ ).
- One can do this for very irregular  $b_s$ .

From now on for simplicity  $x_0 = X_0 = 0$ .

**Method:** Let  $\Omega = \mathcal{C}([0, 1])$ ,  $(X_t)_{t \geq 0}$  be the coordinate process, i.e.  $X_t(\omega) = \omega(t)$  and  $P_0$  is the Wiener measure on  $C([0, 1]) = \Omega$ . Then define

$$P := \exp \left( \int_0^1 b_t dX_t - \frac{1}{2} \int_0^1 b_t^2 dt \right) P_0.$$

We can check that

$$W_t := X_t - \int_0^t b_s ds$$

is a Brownian motion under  $P$ , where  $b_s$  denotes the drift. ( $(X_t)_{t \geq 0}$  is a Brownian motion under  $P_0$ , hence, a martingale under  $P_0$ , but *not* a martingale under  $P$ !)

**Catch:** We have to check, that  $P$  is a probability measure, i.e. we have to check, that

$$\int e^{\int_0^1 b_t dX_t - \frac{1}{2} \int_0^1 b_t^2 dt} dP_0 = 1.$$

This is the hard work in applications.

**Relation with the Transformation Rule for Lebesgue Measure**

Define  $T : \mathcal{C}([0, 1]) \rightarrow \mathcal{C}([0, 1])$  by

$$T(\omega) := \underbrace{X(\omega)}_{=\omega} - \int_0^{\cdot} b(X_s(\omega), s) \, ds.$$

Then by Girsanov (under the conditions specified below)

$$\left( \underbrace{e^{\int_0^1 b_s(X_s, s) \, dX_s - \frac{1}{2} \int_0^1 b(X_s, s)^2 \, ds}}_{="|\det DT|"} P_0 \right) \circ T^{-1} = P_0.$$

**4.2. The General Girsanov Transformation**

Let  $(\Omega, \mathcal{F}, P)$  be a probability space and  $(\mathcal{F}_t)_{t \geq 0}$  be a right-continuous filtration (not necessarily “completed”). Then let  $\tilde{P}$  be another probability measure such that

$$\tilde{P} \stackrel{\text{loc.}}{\ll} P \quad (\text{i.e. } \tilde{P}|_{\mathcal{F}_t} \ll P|_{\mathcal{F}_t} \, \forall t \geq 0).$$

Then the Radon-Nikodym densities

$$Z_t := \frac{d\tilde{P}}{dP} \Big|_{\mathcal{F}_t} := \frac{d\tilde{P}|_{\mathcal{F}_t}}{dP|_{\mathcal{F}_t}}$$

exist and  $(Z_t)_{t \geq 0}$  is a martingale (since for all  $t > s$  and  $F_s \in \mathcal{F}_s$

$$\int_{F_s} Z_t \, dP \stackrel{F_s \in \mathcal{F}_s \subset \mathcal{F}_t}{=} \int_{F_s} d\tilde{P} = \int_{F_s} Z_s \, dP,$$

i.e.

$$E[Z_t | \mathcal{F}_s] = Z_s).$$

**Assumption** (from now on in force):  $Z$  has continuous sample paths  $P$ -a.s.. (This is enough in many applications. Note that by Ito’s representation theorem any martingale with respect to a filtration “generated” by a Brownian motion has this property!)

**Lemma 4.2.1.** *For every  $(\mathcal{F}_t)$ -stopping time  $T$  we have*

$$\tilde{P} = Z_T P \quad \text{on } \mathcal{F}_T \cap \{T < \infty\} \text{ (trace } \sigma\text{-field)}.$$

*In particular,  $\tilde{P} \ll P$  on  $\mathcal{F}_T$  if  $T$  is finite.*

*Proof.* Let  $A \in \mathcal{F}_T$ . Then, by a lemma in [Röc11] for all  $t \geq 0$

$$A \cap \{T \leq t\} \in \mathcal{F}_{T \wedge t} \subset \mathcal{F}_t,$$

Hence,

$$\tilde{P}[\underbrace{A \cap \{T \leq t\}}_{\in \mathcal{F}_t}] = \int_{A \cap \{T \leq t\}} Z_t \, dP \stackrel{A \cap \{T \leq t\} \in \mathcal{F}_{T \wedge t}}{=} \int_{A \cap \{T \leq t\}} Z_{T \wedge t} \, dP = \int_{A \cap \{T \leq t\}} Z_T \, dP.$$

Letting  $t \rightarrow \infty$  and applying monotone convergence the assertion follows.  $\square$

#### 4. Girsanov Transformation

The following lemma describes how martingales “behave” under change from  $P$  to  $\tilde{P}$ . Define

$$\xi(\omega) := \inf\{t \geq 0 \mid Z_t(\omega) = 0\} \quad (\text{Then } Z_\xi = 0).$$

$\xi(\omega)$  is a stopping time. Recall that then  $Z_t(\omega) = 0$  for all  $t \in [\xi(\omega), \infty[$ , since  $(Z_t)_{t \geq 0}$  is a right continuous positive (super)martingale (cf. [Röc11]).

**Lemma 4.2.2.** (i)  $\xi = \infty$   $\tilde{P}$ -a.s. (not necessarily  $P$ -a.s.).

(ii) For all  $s \leq t$ ,  $\varphi_t$   $\mathcal{F}_t$ -measurable and positive we have

$$E_{\tilde{P}}[\varphi_t | \mathcal{F}_s] = 1_{\{Z_s \neq 0\}} Z_s^{-1} E_P[\varphi_t Z_t | \mathcal{F}_s] \quad \tilde{P} - a.s..$$

(iii) Let  $\tilde{M} := (\tilde{M}_t)_{t \geq 0}$  be an  $(\mathcal{F}_t)$ -adapted continuous process. Then  $\tilde{M}$  is a local  $\tilde{P}$ -martingale (up to  $\infty$ ), if  $\tilde{M} \cdot Z$  is a local  $P$ -martingale (up to  $\xi$ ).

*Proof.* (i) By Lemma 4.2.1 with  $T = \xi \wedge t$ , we have

$$\tilde{P}[\underbrace{\xi \leq t}_{\in \mathcal{F}_{t \wedge \xi}}] = E_P[Z_{\xi \wedge t}, \xi \leq t] = E_P[\underbrace{Z_\xi}_{=0}, \xi \leq t] = 0.$$

Letting  $t \rightarrow \infty$ , the assertion follows.

(ii) For all  $\varphi_s$   $\mathcal{F}_s$ -measurable and positive,

$$\begin{aligned} E_{\tilde{P}}[\varphi_s \varphi_t] &= E_{\tilde{P}}[\varphi_s 1_{\{Z_s \neq 0\}} \varphi_t] \\ &= E_P[\varphi_s 1_{\{Z_s \neq 0\}} Z_t \varphi_t] \\ &= E_P[\varphi_s 1_{\{Z_s \neq 0\}} E_P[Z_t \varphi_t | \mathcal{F}_s]] \\ &= E_{\tilde{P}}[\varphi_s 1_{\{Z_s \neq 0\}} Z_s^{-1} E_P[\varphi_t Z_t | \mathcal{F}_s]] \end{aligned}$$

(iii) The assertion directly follows from (ii). But it is also a consequence of Lemma 4.2.1: Suppose  $T_1 \leq T_2 \leq \dots$  and that  $T_n < \xi$  on  $\{\xi > 0\}$  be a localizing sequence for the local  $P$ -martingale  $\tilde{M} \cdot Z$  (hence in particular  $\sup_n T_n = \xi$ ). Then for all bounded stopping times  $T$

$$E_{\tilde{P}}[\tilde{M}_{T \wedge T_n}] \stackrel{4.2.1}{=} E_P[\tilde{M}_{T \wedge T_n} Z_{T \wedge T_n}] = E_P[\tilde{M}_0 Z_0] = E_{\tilde{P}}[\tilde{M}_0].$$

Hence,  $(\tilde{M}_{T_n \wedge t})_{t \geq 0}$  is a  $\tilde{P}$ -martingale and the assertion follows, since  $\xi = \infty$   $\tilde{P}$ -a.s. by (i).  $\square$

**Proposition 4.2.3** (General Girsanov transform). *Let  $M$  be a  $P$ -a.s. continuous local  $P$ -martingale (up to  $\infty$ ). Then*

$$\tilde{M} := M - \left\langle M, \int_0^\cdot \frac{1}{Z_s} dZ_s \right\rangle$$

*is a continuous local  $\tilde{P}$ -martingale (up to  $\infty$ ).*

*Proof.* By Lemma 4.2.2(iii) we have to show that  $\tilde{M}Z$  is a local  $P$ -martingale up to  $\xi$ .

We have that  $\tilde{M}Z$  is  $P$ -a.s. continuous. Let  $\xi_n := \inf\{t \geq 0 \mid Z_t \leq \frac{1}{n}\}$  and  $\tilde{T}_n$ ,  $n \in \mathbb{N}$ ,  $\tilde{T}_n \nearrow \infty$ , be bounded stopping times forming a localizing sequence for the following three local  $P$ -martingales  $M$ ,  $MZ - \langle M, Z \rangle$  and

$$\int_0^\cdot \left( \int_0^r \frac{1}{Z_s} d\langle M, Z \rangle_s \right) dZ_r.$$

Define  $T_n := \xi_n \wedge \tilde{T}_n$ . Note that  $\int_0^r \frac{1}{Z_s} d\langle M, Z \rangle_s$  is predictable, since it is  $P$ -a.s. continuous in  $r$  and adapted. We have that  $0 \leq T_1 \leq T_2 \leq \dots$  and that  $T_n < \xi$  on  $\{\xi > 0\}$ , since  $\xi_n < \xi$  on  $\{0 < \xi < \infty\}$  because  $Z$  is  $P$ -a.s. continuous. Furthermore,  $\xi_n \nearrow \xi$ , because  $Z$  is  $P$ -a.s. continuous, hence  $T_n \nearrow \xi$ . Then by Itô's product rule for all  $t \geq 0$

$$\begin{aligned} (\tilde{M}Z)_{t \wedge T_n} &= (MZ)_{t \wedge T_n} - Z_{t \wedge T_n} \int_0^{t \wedge T_n} \frac{1}{Z_s} d\langle M, Z \rangle_s \\ &= (MZ)_{t \wedge T_n} - \int_0^{t \wedge T_n} Z_r d\left(\int_0^r \frac{1}{Z_s} d\langle M, Z_s \rangle\right) - \int_0^{t \wedge T_n} \int_0^r \frac{1}{Z_s} d\langle M, Z \rangle_s dZ_r \\ &= (MZ)_{t \wedge T_n} - \int_0^{t \wedge T_n} Z_r \frac{1}{Z_r} d\langle M, Z \rangle_r - \int_0^{t \wedge T_n} \int_0^r \frac{1}{Z_s} d\langle M, Z \rangle_s dZ_r \\ &= (MZ)_{t \wedge T_n} - \langle M, Z \rangle_{t \wedge T_n} - \int_0^{t \wedge T_n} \int_0^r \frac{1}{Z_s} d\langle M, Z \rangle_s dZ_r \end{aligned}$$

Here,  $\int_0^{t \wedge T_n} \int_0^r \frac{1}{Z_s} d\langle M, Z \rangle_s dZ_r$  is a  $P$ -martingale, and  $(MZ)_{t \wedge T_n} - \langle M, Z \rangle_{t \wedge T_n}$  is a  $P$ -martingale. Therefore,  $\tilde{M}Z$  is a local  $P$ -martingale up to  $\xi$ .  $\square$

**Remark 4.2.4** (“ $Z$  as an exponential martingale”). *We have by Itô formula for  $t < \xi$   $P$ -a.s.*

$$\log Z_t = \log Z_0 + \underbrace{\int_0^t \frac{1}{Z_s} dZ_s}_{=: Y_t} - \frac{1}{2} \int_0^t \frac{1}{Z_s^2} d\langle Z \rangle_s, \quad (\star)$$

where  $Y_t$  is a local  $P$ -martingale up to  $\xi$ , since  $Z$  is a  $P$ -martingale up to  $\infty$ , and for its pathwise quadratic variation (cf. 5.3.9) we have

$$\langle Y \rangle_t = \int_0^t \frac{1}{Z_s^2} d\langle Z \rangle_s, \quad t < \xi.$$

Exponentiating  $(\star)$  yields

$$Z_t = Z_0 \cdot e^{Y_t - \frac{1}{2}\langle Y \rangle_t} \quad (\text{exponential } P\text{-martingale}).$$

Then  $Z$  solves the following SDE for given  $Y$  up to  $\xi$ :

$$dZ = Z dY \Leftrightarrow Z_t = Z_0 + \int_0^t Z_s dY_s, \quad (Z(0) = Z_0)$$

*Proof.* By the 2-dimensional Itô's formula we have

$$\begin{aligned} Z_t &= Z_0 + Z_0 \int_0^t e^{Y_s - \frac{1}{2}\langle Y \rangle_s} dY_s - \frac{1}{2} Z_0 \int_0^t e^{Y_s - \frac{1}{2}\langle Y \rangle_s} d\langle Y \rangle_s + \frac{1}{2} Z_0 \int_0^t e^{Y_s - \frac{1}{2}\langle Y \rangle_s} d\langle Y \rangle_s \\ &= Z_0 + \int_0^t Z_s dY_s. \end{aligned}$$

Note that the remaining terms in Itô's formula vanish, since  $\langle Y \rangle$  is of bounded variation.  $\square$

**Corollary 4.2.5.** *For  $\tilde{M}$  as in Proposition 4.2.3 we have  $\tilde{M} = M - \langle M, Y \rangle$   $P$ -a.s. up to  $\xi$ , hence  $\tilde{P}$ -a.s. up to  $\infty$ .*

*Proof.* The assertion follows by the definition of  $\tilde{M}$  and  $Y$ .  $\square$

### 4.3. Girsanov Transform with Brownian Motion

Let  $(X_t)_{t \in [0,1]}$  be a Brownian motion on a probability space  $(\Omega, \mathcal{F}, P)$ , adapted to a right-continuous filtration  $(\mathcal{F}_t)_{t \in [0,1]}$ . (Take for example  $(\Omega, \mathcal{F}, P)$  to be the canonical Wiener space  $(C[0, 1])_0, \mathcal{F}, P$ ) with classical Wiener measure  $P$ .)

**Heuristics:** Let  $Y$  be a local continuous  $P$ -martingale and  $\tilde{P} \ll_{\text{loc}} P$  with density

$$Z_t := e^{Y_t - \frac{1}{2}\langle Y \rangle_t} \quad \text{on } \mathcal{F}_t.$$

Then we know that  $\tilde{M} = X - \langle X, Y \rangle$  is a local continuous  $\tilde{P}$ -martingale up to  $\infty$ , where  $X$  is a martingale with respect to  $P$  (as  $M$  above). Since  $\langle \tilde{M} \rangle_t = \langle X \rangle_t = t$ , it follows by Levy's characterization theorem of Brownian motion that  $\tilde{M}$  is a Brownian motion under  $\tilde{P}$ . We want to get that

$$d\langle X, Y \rangle = b_t dt.$$

We succeed by the following

**Ansatz:**

$$Y_t := \int_0^t b_s dX_s.$$

**Remark 4.3.6.** *In order to make this work, we need that*

(i)  $(b_t)_{t \in [0,1]}$  is progressively measurable,

(ii)

$$P \left[ \int_0^1 b_s^2 ds < \infty \right] = 1$$

(Then,  $Y$  is a continuous local  $P$ -martingale!) AND!

(iii)

$$\tilde{P} = e^{\int_0^1 b_s dX_s - \frac{1}{2} \int_0^1 b_s^2 ds} \cdot P$$

is a probability measure, i.e.

$$E_P \left[ e^{\int_0^1 b_s dX_s - \frac{1}{2} \int_0^1 b_s^2 ds} \right] = 1 \tag{4.3.3}$$

**Theorem 4.3.7** (Girsanov transform for  $M$  as a Brownian motion): Assume (i) - (iii) from Remark 4.3.6. Let

$$Z_t := \exp \left[ \int_0^t b_s dX_s - \frac{1}{2} \int_0^t b_s^2 ds \right], \quad t \in [0, 1],$$

and

$$\tilde{P} := Z_1 \cdot P.$$

Then

$$W_t := X_t - \int_0^t b_s ds, \quad t \in [0, 1],$$

is a Brownian motion under  $\tilde{P}$ .

*Proof. Claim:*  $(Z_t)_{t \in [0,1]}$  is a  $P$ -martingale up to 1.

**Step 1:**  $(Z_t)_{t \in [0,1]}$  is a (global)  $P$ -supermartingale.

*Proof.* It is clear, that  $(Z_t)$  is a local continuous  $P$ -martingale. Let  $0 \leq T_1 \leq \dots \leq T_n \dots < 1$  be a localizing sequence of stopping times for  $(Z_t)_{t \in [0,1]}$ . Then for  $0 \leq s < t \leq 1$

$$\begin{aligned} E_P[Z_t | \mathcal{F}_s] &= E_P[\lim_{n \rightarrow \infty} Z_{t \wedge T_n} | \mathcal{F}_s] \\ &\stackrel{Z_t \geq 0}{\leq} \liminf_{n \rightarrow \infty} E_P[Z_{t \wedge T_n} | \mathcal{F}_s] \\ &= \liminf_{n \rightarrow \infty} Z_{s \wedge T_n} = Z_s. \end{aligned}$$

□

**Step 2:**  $(Z_t)_{t \in [0,1]}$  is a  $P$ -martingale.

*Proof.* By (iii) for all  $s \in [0, 1]$

$$1 = E[Z_1] \leq E[Z_s] \leq E[\underbrace{Z_0}_{=1}] = 1.$$

In addition,

$$0 \leq Z_s - E_P[Z_1 | \mathcal{F}_s]$$

and

$$\int (Z_s - E_P[Z_1 | \mathcal{F}_s]) \, dP = 0.$$

So,  $Z_s = E_P[Z_1 | \mathcal{F}_s]$   $P$ -a.s.

□

We have

$$\left\langle X, \int_0^\cdot b_s \, dX_s \right\rangle_t = \int_0^t b_s \, ds.$$

Hence, by Corollary 4.2.5 it follows that  $W$  is a continuous local  $\tilde{P}$ -martingale. But  $\langle W \rangle_t = \langle X \rangle_t = t$   $P$ -a.s., hence  $\tilde{P}$ -a.s.. So, by Lévy (Proposition 1.5.34) the assertion follows. □

**Remark 4.3.8.** (i) *Special case:*

$$b_t(\omega) := b(X_t(\omega), t),$$

*i.e. depending only on the “present” time, where  $b : \mathbb{R} \times [0, 1] \rightarrow \mathbb{R}$  is  $\mathcal{B}(\mathbb{R}) \otimes \mathcal{B}([0, 1])$  measurable. Then 4.3.6(i) is fulfilled.*

*4.3.6(ii) is fulfilled, if e.g.*

$$\begin{aligned} E \left[ \int_0^1 b_t^2 \, dt \right] &= \int_0^1 E[b^2(X_t, t)] \, dt \\ &\stackrel{E[f(X_t)] = p_t f(0)}{=} \int_0^1 p_t(b^2(\cdot, t))(0) \, dt \\ &\stackrel{p_t(0, dx) = N(0, t)}{=} \int_0^1 \int_{\mathbb{R}} b^2(x, t) N(0, t)(dx) \, dt \\ &= \int_0^1 \int_{\mathbb{R}} b^2(x, t) \frac{1}{\sqrt{2\pi t}} e^{-\frac{x^2}{2t}} \, dx \, dt < \infty. \end{aligned}$$

*In particular, it is not necessary that  $b$  is bounded.*

*To satisfy 4.3.6(iii) we have to work harder (see Theorem 4.4.15 and Remark 4.4.16(i), as well as for a very special case Example 4.3.9 below)!*

#### 4. Girsanov Transformation

(ii) In Theorem 4.3.7  $(X_t)_{t \geq 0}$  solves the following SDE under  $\tilde{P}$

$$dX_t = dW_t + b_t dt.$$

Here,  $W_t$  is a Brownian motion under  $\tilde{P}$ , not under  $P$ .

**Example 4.3.9.** Consider  $b_t = \alpha \in \mathbb{R}$  fixed for all  $t$ . Then, clearly, 4.3.6(i),(ii) hold, but also:

**Claim:** (iii) holds.

*Proof.* Since  $X_1$  is normal distributed, we have

$$\begin{aligned} E \left[ e^{\alpha X_1 - \frac{1}{2}\alpha^2} \right] &= e^{-\frac{1}{2}\alpha^2} \int e^{\alpha x} N(0, 1)(dx) \\ &= e^{-\frac{1}{2}\alpha^2} \int_{\mathbb{R}} e^{\alpha x} \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} dx \\ &= e^{-\frac{1}{2}\alpha^2} \cdot e^{\frac{1}{2}\alpha^2} = 1. \end{aligned}$$

Here, we have used that for the Laplace-transform  $\mathcal{L}$  we have

$$\mathcal{L}(N(0, \sigma^2))(\xi) = \int e^{\xi x} \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{1}{2\sigma^2}x^2} dx = e^{\frac{1}{2}\xi^2\sigma^2}.$$

□

Hence, by Theorem 4.3.7 under  $\tilde{P} := Z_1 \cdot P$  the process  $W_t = X_t - \alpha t$ ,  $t \geq 0$ , is a Brownian motion.

**Theorem 4.3.10** (Cameron-Martin). Consider the canonical model for Brownian motion, i.e.  $\Omega = C([0, 1])$ ,  $P_0$  classical Wiener measure and  $X_t(\omega) := \omega(t)$ ,  $t \geq 0$ . Let

$$h \in E := C([0, 1])_0 := \{h \in C([0, 1]) | h(0) = 0\}$$

and

$$X_h : C([0, 1])_0 \rightarrow C([0, 1])_0$$

defined by

$$X_h(\omega) := X(\omega) + h.$$

Let  $P_h := P_0 \circ X_h^{-1}$  be the law of  $X_h$  on  $C([0, 1])$ . Then the following assertions are equivalent:

(i)  $P_h \approx P_0$ .

(ii)

$$h \in H := \left\{ h \in C([0, 1])_0 \mid h \text{ is absolutely continuous and } \dot{h} \in L^2([0, 1], ds) \right\}.$$

In this case

$$\frac{dP_h}{dP_0} = \exp \left( \int_0^1 \dot{h}(s) dX(s) - \frac{1}{2} \int_0^1 (\dot{h}(s))^2 ds \right).$$

$H$  is called *Cameron-Martin space*.  $H$  is a Hilbert space with inner product

$$\langle h, \tilde{h} \rangle_H := \int_0^1 \dot{h}(s) \dot{\tilde{h}}(s) ds, \quad h, \tilde{h} \in H.$$

Note that  $\|h\|_H^2 = \langle h, h \rangle_H = 0$  implies  $\dot{h}(s) = 0$   $ds$ -a.s.. Hence,

$$h(t) = h(0) + \int_0^t \dot{h}(s) ds = h(0) = 0 \quad \forall t \geq 0.$$

So,  $\|\cdot\|_H$  is not only a semi norm, but a norm. We have that  $H$  is dense in  $E := C([0, 1])_0$  with respect to  $\|\cdot\|_H$ . Then identifying  $H$  with its dual  $H'$  by the Riesz map  $R$  we get

$$E' \hookrightarrow H' \xrightarrow{R} H \subset E$$

continuously (see below). Before we prove Theorem 4.3.10, we want to characterize  $E'$ .

**Lemma 4.3.11.** *For*

$$M_0 := \left\{ \mu \mid \mu \text{ is a signed measure of finite total variation on } [0, 1], \text{ such that } \mu(1) := \int 1 \, d\mu = 0 \right\},$$

we have

$$E' = M_0.$$

**Remark 4.3.12.** *A signed measure  $\mu$  can be written as  $\mu = \mu_1 - \mu_2$  with positive measures  $\mu_1$  and  $\mu_2$  and the total variation of  $\mu$  is defined as*

$$\sup_{A \in \mathcal{B}([0,1])} \mu(A) = \|\mu\|_{Var} < \infty.$$

Furthermore, for a signed measure  $\mu$ , there exist positive measures  $\mu^+, \mu^-$  and  $S \in \mathcal{B}([0, 1])$  such that  $\mu = \mu^+ - \mu^-$  (“Hahn-Jordan decomposition”) and

$$\mu^+(S^c) = 0 \text{ and } \mu^-(S) = 0.$$

Then

$$\|\mu\|_{Var} = \mu^+([0, 1]) + \mu^-([0, 1]).$$

*Proof of 4.3.11.* “ $\supset$ ”: Defining  $f \mapsto \mu(f) := \int f \, d\mu$  for  $\mu \in M_0$  we see that this way any  $\mu \in M_0$  defines an element in  $E'$ . Furthermore, if  $\mu, \nu \in M_0$  such that  $\mu(f) = \nu(f) \, \forall f \in E$ , then for all  $f \in C([0, 1])$

$$\begin{aligned} \int f \, d\mu &= \int f \, d\mu - f(0) \int 1 \, d\mu = \int \underbrace{(f - f(0))}_{\in E} \, d\mu \\ &= \int (f - f(0)) \, d\nu = \int f \, d\nu - \underbrace{\int f(0) \, d\nu}_{=0} = \int f \, d\nu. \end{aligned}$$

Hence,  $\mu = \nu$  and, thus  $M_0 \hookrightarrow E'$ . (See also [Cho69a] and [Cho69b].)

“ $\subset$ ”: Let  $l \in E'$ . Then by the Hahn-Banach theorem there exists an  $\bar{l} \in (C[0, 1])'$  such that  $\bar{l} = l$  on  $E (\subset C([0, 1]))$ . Hence, by Riesz-Markov (cf. [Röc04]) there exists a signed measure  $\nu$  on  $[0, 1]$  of finite total variation such that

$$\bar{l}(f) = \int f \, d\nu \quad \forall f \in C([0, 1]).$$

Set  $\mu := \nu - \nu(1)\delta_0$ , where  $\delta_0$  is the Dirac measure at  $0 \in [0, 1]$ . Then  $\mu(1) = 0$ , so  $\mu \in M_0$  and for all  $f \in E$

$$\int f \, d\mu = \int f \, d\nu - \nu(1) \underbrace{f(0)}_{=0} = \int f \, d\nu = \bar{l}(f) = l(f).$$

□

#### 4. Girsanov Transformation

*Proof of Theorem 4.3.10.* (ii)  $\Rightarrow$  (i): (An alternative proof using Fourier transform can be found in [MR92, Chapter II].) We first check the conditions of Remark 4.3.6 (i)-(iii):

(i) and (ii) are obviously satisfied for

$$b_s(\omega) := \dot{h}(s) \quad \forall \omega \in \Omega, s \in [0, 1].$$

(iii): Define

$$Y_t := \int_0^t \dot{h}(s) dX_s \quad (\text{It\^o-Integral}).$$

Then  $Y_1$  is centered (i.e.  $E[Y_1] = 0$ ) and normally distributed as an  $L^2$ -limit of centered, normally distributed random variables (cf. [Röc11]). Furthermore,

$$E[Y_1^2] (= E[\langle Y \rangle_1]) = E\left(\int_0^1 (\dot{h}(s))^2 ds\right) = \|h\|_H^2,$$

i.e.

$$Y_1 \sim N(0, \|h\|_H^2).$$

Hence,

$$E\left[e^{Y_1 - \frac{1}{2}\|h\|_H^2}\right] = e^{-\frac{1}{2}\|h\|_H^2} \underbrace{E[e^{Y_1}]}_{=e^{\frac{1}{2}\|h\|_H^2}} = 1$$

So, 4.3.6(iii) holds. Therefore, applying Girsanov (Theorem 4.3.7) we obtain that under

$$\tilde{P} := Z_1 \cdot P := \exp\left[\int_0^1 \dot{h}(t) dX_t - \frac{1}{2}\int_0^1 \dot{h}(t)^2 dt\right] \cdot P$$

$W := X - h$  is a Brownian motion, i.e.  $\tilde{P} \circ W^{-1} = P_0$ , i.e., since  $W : \Omega \rightarrow \Omega$  is bijective,  $\tilde{P} = P_0 \circ (W^{-1})^{-1} = P_0 \circ X_h^{-1} = P_h$ . But, since  $E[Y_1^2] < \infty$ , we have  $Z_1 > 0$   $P$ -a.s., therefore,  $P_h \approx P$ .

(i)  $\Rightarrow$  (ii): We know that  $P_0(E) = 1$  with  $E := C([0, 1])_0$ . Clearly, we have that  $H \subset E$  dense with respect to the norm  $\|\cdot\|_\infty$  (cf. functional analysis). Let  $h \in H$ . Then (by definition of absolute continuity)

$$h(t) = \underbrace{h(0)}_{=0} + \int_0^t \dot{h}(s) ds \quad \forall t \in [0, 1],$$

thus,

$$|h(t)| \leq \int_0^t |\dot{h}(s)| ds \leq \int_0^1 |\dot{h}(s)| ds \leq \left(\int_0^1 |\dot{h}(s)|^2 ds\right)^{\frac{1}{2}} = \|h\|_H, \quad \forall t \in [0, 1],$$

hence,

$$\|h\|_\infty \leq \|h\|_H.$$

Hence  $H \subset E$  continuously w.r.t. the norms  $\|\cdot\|_H$  and  $\|\cdot\|_\infty$ . Let  $R : H' \rightarrow H$  be the Riesz isomorphism. Then

$$E' \subset H' \xrightarrow{R} H \subset E \quad (\text{continuously}).$$

(Cf. Lemma 4.5.20 below:  $R(\mu) = -\int_0^\cdot \mu([0, s]) ds$ ,  $\mu \in E'$ .)

**Claim 1:** Let  $\mu \in E'$  ( $= M_0$ ). The map  $C([0, 1])_0 \ni \omega \mapsto \mu(\omega) \in \mathbb{R}$  is Gaussian distributed under  $P_0$ , more precisely

$$P_0 \circ \mu^{-1} = N\left(0, \underbrace{E[\mu^2]}_{\int \mu(\omega)^2 P_0(d\omega)}\right).$$

*Proof.* Recall that  $\mu = \lim_{n \rightarrow \infty} \underbrace{\sum_{i=1}^{N_n} \alpha_i^{(n)} \delta_{t_i^{(n)}}}_{=: \mu_n}$ ,  $\alpha_i^{(n)} \in \mathbb{R}$ ,  $t_i^{(n)} \in [0, 1]$  weakly (cf. [Bau78]). But  $\mu_n : E \rightarrow \mathbb{R}$ ,  $n \in \mathbb{N}$ , are jointly (centred) Gaussian under  $P_0$ , because

$$\sum_{l=1}^N (\gamma_l \mu_{n_l})(\omega) = \sum_{l=1}^N \gamma_l \sum_{i=1}^{N_{n_l}} \alpha_i^{(n_l)} \delta_{t_i^{(n_l)}}(\omega) = \sum_{l=1}^N \gamma_l \sum_{i=1}^{N_{n_l}} \alpha_i^{(n_l)} \underbrace{\omega(t_i^{(n_l)})}_{=: X_{t_i^{(n_l)}}^{(n_l)}}(\omega) \quad \forall f \in E$$

i.e.

$$\sum_{l=1}^N \gamma_l \mu_{n_l} = \underbrace{\sum_{l=1}^N \gamma_l \sum_{i=1}^{N_{n_l}} \alpha_i^{(n_l)} X_{t_i^{(n_l)}}^{(n_l)}}_{(\mathbb{R}\text{-valued) centred Gaussian}} \sim N(0, E(\sum_{i=1}^N \gamma_l \mu_{n_l})^2)$$

Hence, since  $\mu_n \xrightarrow{n \rightarrow \infty} \mu$  weakly, i.e. pointwise on  $E$ , we have that  $\hat{\mu}_n \xrightarrow{n \rightarrow \infty} \hat{\mu}$  and  $\mu$  is centered Gaussian with variance  $E[\mu^2]$ .  $\square$

**Claim 2:** Let  $\mu \in M_0$  of the form  $\mu = \varrho \cdot dt$  and  $\varrho$  bounded and

$$\mu(1) = \int_0^1 \varrho dt = 0. \quad (\star)$$

Then

$$E[\mu^2] = \int_0^1 \left( \int_0^t \varrho(s) ds \right)^2 dt = \|R(\mu)\|_H^2. \quad (4.3.4)$$

*Proof.*

$$\begin{aligned} E[\mu^2] &\stackrel{X_t(\omega) \equiv \omega(t)}{=} E \left( \int_0^1 X_t \varrho(t) dt \int_0^1 X_{t'} \varrho(t') dt' \right) \\ &\stackrel{\text{Fubini}}{=} \int_0^1 \int_0^1 \varrho(t) \varrho(t') E[X_t X_{t'}] dt' dt \\ &= \int_0^1 \int_0^1 \varrho(t) \varrho(t') (t \wedge t') dt' dt = 2 \int_0^1 \int_0^1 t_{\{t < t'\}}^1 \varrho(t) \varrho(t') dt dt' \\ &= 2 \int_0^1 t \varrho(t) \int_t^1 \varrho(t') dt' dt \\ &= - \int_0^1 t \frac{d}{dt} \left( \left( \int_t^1 \varrho(t') dt' \right)^2 \right) dt \\ &\stackrel{\text{I.b.p.}}{=} \int_0^1 \left( \int_t^1 \varrho(t') dt' \right)^2 dt \\ &= \int_0^1 \left( \int_0^t -\varrho(s) ds \right)^2 dt \\ &= \int_0^1 \underbrace{\left( \frac{d}{dt} \int_0^t \left( \int_0^{t'} -\varrho(s) ds \right) dt' \right)^2}_{=: \tilde{R}(\mu)(t)} dt = \int_0^1 \left( \frac{d}{dt} \tilde{R}(\mu)(t) \right)^2 dt = \|\tilde{R}(\mu)\|_H^2. \end{aligned} \quad (4.3.5)$$

For Claim 2 it remains to show that  $\tilde{R}(\mu) = R(\mu)$ , i.e.

$$R(\mu) = - \int_0^{\cdot} \left( \int_0^{t'} \varrho(s) ds \right) dt'. \quad (4.3.6)$$

#### 4. Girsanov Transformation

To this end let  $\tilde{h} \in H$ . Then

$$\mu(\tilde{h}) = \int_0^1 \tilde{h}(t) \varrho(t) dt \stackrel{\text{I.b.p.}}{=} - \int_0^1 \frac{d}{dt} \tilde{h}(t) \int_0^t \varrho(s) ds dt = \left\langle \tilde{h}, - \int_0^{\cdot} \left( \int_0^{t'} \varrho(s) ds \right) dt' \right\rangle_H \quad (4.3.7)$$

and Claim 2 is proved.  $\square$

Since  $C_0^1(]0, 1[)$  is dense in  $L^2([0, 1], dt)$ , for all  $\tilde{h} \in H$  there exists a sequence  $(v_n)_{n \in \mathbb{N}}$  in  $C_0^1(]0, 1[)$  such that  $v_n \rightarrow \tilde{h}$  as  $n \rightarrow \infty$  in  $L^2([0, 1], dt)$ . Hence

$$u_n := \int_0^{\cdot} v_n dt \xrightarrow{n \rightarrow \infty} \tilde{h} \quad \text{in } H.$$

Since by (4.3.6)

$$u_n = R(\mu_n) \quad \text{for} \quad \mu_n := -\dot{v}_n dt \in E'$$

it follows that for

$$\tilde{M}_0 := \left\{ \varrho dt \mid \varrho \in C([0, 1]), \int_0^1 \varrho dt = 0 \right\}$$

that  $R(\tilde{M}_0)$  is dense in  $H$  with respect to  $\|\cdot\|_H$ .  $R(\tilde{M}_0)$  is also a linear subspace of  $H$ .

Let  $h \in E(= C([0, 1])_0)$  such that  $P_h \ll P$ .

**Claim:** Let  $\mu_n \in \tilde{M}_0$  such that  $R(\mu_n) \xrightarrow{n \rightarrow \infty} 0$  in  $H$ . Then  $\mu_n(h) \xrightarrow{n \rightarrow \infty} 0$ .

Suppose the claim is true. Then  $h \in H$ .

*Proof.* By the claim the map  $R(\tilde{M}_0) \ni R(\mu) \mapsto \mu(h) \in \mathbb{R}$  is a linear continuous functional on  $(R(\tilde{M}_0), \|\cdot\|_H)$ . Hence, by Riesz (note  $R(\tilde{M}_0)$  is dense in  $H$ , hence Riesz is applicable) there exists an unique  $h_0 \in H$  such that

$$\mu(h) = \langle R(\mu), h_0 \rangle_H \quad \forall \mu \in \tilde{M}_0.$$

But by 4.3.11 we also have  $\langle R(\mu), h_0 \rangle_H = \mu(h_0)$ . Therefore,

$$\mu(h) = \mu(h_0), \forall \mu \in \tilde{M}_0 \quad (4.3.8)$$

Hence, for all  $\varrho \in C([0, 1])$  and for

$$\tilde{h} := h - \int_0^1 h(t) dt, \quad \tilde{h}_0 := h_0 - \int_0^1 h_0(t) dt$$

we have

$$\begin{aligned} \int \tilde{h} \varrho dt &= \int \tilde{h} \underbrace{\left( \varrho - \int_0^1 \varrho ds \right)}_{\in \tilde{M}_0} dt = \int h \left( \varrho - \int_0^1 \varrho ds \right) dt \\ &\stackrel{(4.3.8)}{=} \int h_0 \left( \varrho - \int_0^1 \varrho ds \right) dt = \int \tilde{h}_0 \left( \varrho - \int_0^1 \varrho ds \right) dt = \int h_0 \varrho dt. \end{aligned}$$

Hence,  $\tilde{h} = \tilde{h}_0$ , therefore,  $h = h_0$ , because  $h(0) = h_0(0) = 0$ . Therefore, the assertion is true as long as the claim holds.  $\square$

*Proof of Claim.* Since  $R(\mu_n) \rightarrow 0$  in  $H$ , it follows by (4.3.4) that

$$E_{P_0}[\mu_n^2] = \|R(\mu_n)\|_H^2 \xrightarrow{n \rightarrow \infty} 0,$$

i.e.  $\mu_n \rightarrow 0$  in  $L^2(P)$ , hence, also in  $P_0$ -measure, therefore, because of  $P_h \ll P_0$ , also in  $P_h$ -measure. Since  $\{\mu_n, n \in \mathbb{N}\}$  is a Gaussian family under  $P_h$ , it follows by [Röc11] that

$$\mu_n \xrightarrow{n \rightarrow \infty} 0 \quad \text{in } L^p(P_h) \quad \forall p \geq 1. \quad (4.3.9)$$

Since we have

$$E_{P_h}[\mu_n] = \int \underbrace{\mu_n(\omega + h)}_{\mu_n(\omega) + \mu_n(h)} P(d\omega) = \underbrace{\int \mu_n(\omega) P(d\omega)}_{=0} + \mu_n(h), \quad (4.3.10)$$

applying (4.3.9) for  $p = 1$ , it follows that

$$\limsup_{n \rightarrow \infty} |\mu_n(h)| \stackrel{(4.3.10)}{\leq} \limsup_{n \rightarrow \infty} E_{P_h}[|\mu_n|] = 0.$$

So, the claim is proved.  $\square$

## 4.4. Novikov condition

Let  $(\Omega, \mathcal{F}, P)$  be a probability space together with a right-continuous complete filtration  $(\mathcal{F}_t)$  and let  $(Y_t)_{t \geq 0}$  be a  $P$ -a.s. continuous local martingale such that  $Y_0 = 0$ . Then by 1.4.33

$$Z_t := \exp\left(Y_t - \frac{1}{2}\langle Y \rangle_t\right), \quad t \geq 0,$$

is a ( $P$ -a.s.) continuous local martingale. We want to derive a condition for (4.3.3).

**Lemma 4.4.13.** *Let  $t \geq 0$  and  $\langle Y \rangle_t$  be bounded. Then  $Z_t \in \mathcal{L}^p$  for all  $p > 1$  and*

$$E(Z_t^p) \leq \exp\left(\frac{1}{2}p(p-1)\|\langle Y \rangle_t\|_\infty\right).$$

*Proof.*

$$\begin{aligned} E(Z_t^p) &= E\left[\exp\left(pY_t - \frac{1}{2}p^2\langle Y \rangle_t + \left(\frac{1}{2}p^2 - \frac{1}{2}p\right)\langle Y \rangle_t\right)\right] \\ &\leq \exp\left(\frac{1}{2}p(p-1)\|\langle Y \rangle_t\|_\infty\right) E\left[\exp\left(pY_t - \frac{1}{2}p^2\langle Y \rangle_t\right)\right]. \end{aligned}$$

Set  $\tilde{Y}_t := pY_t$ . Then  $\langle \tilde{Y} \rangle_t = p^2\langle Y \rangle_t$  and, therefore,

$$E(\exp(pY_t - \frac{1}{2}p^2\langle Y \rangle_t)) = E(\exp(\tilde{Y}_t - \frac{1}{2}\langle \tilde{Y} \rangle_t)) \leq E(\exp(\tilde{Y}_0 - \frac{1}{2}\langle \tilde{Y} \rangle_0)) = 1,$$

since  $Z$  is a supermartingale (by Fatou).  $\square$

Now we can prove a condition implying (4.3.3).

**Theorem 4.4.14** (Novikov). *If  $E(\exp(\frac{1}{2}\langle Y \rangle_t)) < \infty$  for all  $t$ , then  $E(Z_t) = 1$  for all  $t$ .*

**Remark 4.4.15.** (i) *In the examples above (see 4.3.8 and 4.3.9) we had  $Y_t = \int_0^t b_s dX_s$  and*

$$E\left(\exp\left(\frac{1}{2}\int_0^t b_s^2 ds\right)\right) < \infty$$

*was always satisfied.*

#### 4. Girsanov Transformation

(ii) For the proof of Theorem 4.4.14 we need that one can construct a Wiener process  $W$ , such that

$$Y_t = W_{\langle Y \rangle_t}$$

and (for every fixed  $t$ )  $\langle Y \rangle_t$  is a stopping time with respect to a suitable filtration for which  $W$  is adapted. This means that  $Y_t$  has the form  $W_T$ , with

$$\langle Y \rangle_t = T.$$

In other words, any  $P$ -a.s. continuous local martingale is the time change of a Brownian motion. The details are presented in Appendix A.

Now Theorem 4.4.14 follows from

**Theorem 4.4.16.** Let  $(W_t)_{t \geq 0}$  be a Wiener process on  $(\Omega, \mathcal{F}, P)$  and  $T$  be a stopping time. If

$$E(\exp(\frac{1}{2}T)) < \infty,$$

then the following "Wald identity" holds:

$$E(\exp(W_T - \frac{1}{2}T)) = 1.$$

**Remark 4.4.17.** Set  $M_t := \exp(W_t - \frac{1}{2}t)$ . Then  $(M_t)$  is a continuous martingale, since it is a continuous positive supermartingale and  $E[M_t] = 1$  (cf. Example 4.3.9). By the optional sampling theorem for unbounded stopping times (cf. [Röc11]), we have

$$E(M_T) \leq E(M_0) = 1.$$

Thus, it is clear that  $E(\exp(W_T - \frac{1}{2}T)) \leq 1$  in 4.4.16. For the proof of " $\geq$ " in 4.4.16 we will need two Lemmas.

**Lemma 4.4.18.** Let  $\tilde{P}$  be a probability measure on  $(\Omega, \mathcal{F}, P)$  with

$$\tilde{P}|_{\mathcal{F}_t} = \exp(W_t - \frac{1}{2}t) \cdot P|_{\mathcal{F}_t}, \quad \forall t \geq 0,$$

and  $T$  be a stopping time with  $P(T < \infty) = 1$ . Then

$$E(\exp(W_T - \frac{1}{2}T)) = \tilde{P}(T < \infty). \quad (4.4.11)$$

In particular, the above Wald identity (cf. 4.4.16) holds if and only if

$$\tilde{P}(T < \infty) = 1.$$

*Proof.* Since  $M_t := \exp(W_t - \frac{1}{2}t)$  is a martingale and  $\{T \leq t\} \in \mathcal{F}_{t \wedge T}$  we have

$$\tilde{P}[T \leq t] = E[1_{\{T \leq t\}} M_t] = E[1_{\{T \leq t\}} M_{t \wedge T}] = E[1_{\{T \leq t\}} M_T].$$

Letting  $t \rightarrow \infty$  we get (since  $P[T < \infty] = 1$ )

$$\tilde{P}[T < \infty] = E[M_T].$$

□

**Lemma 4.4.19.** Let  $c > 0$  and define the “passage time of  $(W_t - t)$ ” by

$$T_c := \inf\{t > 0 | W_t = t - c\}. \quad (\Rightarrow W_{T_c} = T_c - c)$$

Then  $T_c$  is a stopping time such that

$$\tilde{P}(T_c < \infty) = 1$$

and, thus, the above Wald identity holds for  $T_c$ . Furthermore,

$$E\left(\exp\left(\frac{1}{2}T_c\right)\right) = e^c.$$

*Proof.* By the law of iterated logarithm we have

$$P(T_c < \infty) = 1.$$

$\tilde{W}_t = W_t - t$  is a Brownian motion with respect to  $\tilde{P}$  (cf. Example 4.3.9 with  $\alpha = 1$ ), which means that  $T_c$  is a passage time of  $(\tilde{W}_t)$  with respect to  $\tilde{P}$ . Therefore, again by the law of iterated logarithm

$$\tilde{P}(T_c < \infty) = 1.$$

Thus, by 4.4.18

$$1 = E\left(\exp\left(W_{T_c} - \frac{1}{2}T_c\right)\right) = e^{-c}E\left(\exp\left(\frac{1}{2}T_c\right)\right).$$

□

*Proof of Theorem 4.4.16.* It remains to show that “ $\geq$ ” holds:

$M_t := \exp(W_t - \frac{1}{2}t)$ ,  $t \geq 0$ , is a positive continuous supermartingale. Hence,

$$1 \geq E(M_{T_c \wedge T}) \geq E(M_{T_c}) \stackrel{4.4.19}{=} 1.$$

But then for all  $c > 0$

$$\begin{aligned} 1 &= E(M_{T_c \wedge T}) \stackrel{W_{T_c} = T_c - c}{=} E\left(\exp\left(\frac{1}{2}T_c\right) \exp(-c), T_c \leq T\right) + E\left(\exp\left(W_T - \frac{1}{2}T\right), T_c > T\right) \\ &\leq e^{-c} \underbrace{E(e^{\frac{1}{2}T})}_{< \infty \text{ (Novikov)}} + E\left(\exp\left(W_T - \frac{1}{2}T\right)\right) \\ &\stackrel{c \rightarrow \infty}{\rightarrow} E\left(\exp\left(W_T - \frac{1}{2}T\right)\right) \end{aligned}$$

□

## 4.5. Integration by Parts on Wiener Space: A First Introduction to the Malliavin Calculus Following J.M. Bismut

Fix the following framework:

Let  $P := P_0$  be the Wiener measure on  $\Omega = C([0, 1])_0 =: E$ ,  $(X_t)$  the coordinate process,

$$H := \left\{ h \in C([0, 1])_0 \mid h \text{ is absolutely continuous and } \int_0^1 \dot{h}(s)^2 ds < \infty \right\}$$

the Cameron-Martin space.  $H$  is a Hilbert space with inner product

$$\langle h, g \rangle_H = \int_0^1 \dot{h}(s)\dot{g}(s) ds = \langle \dot{h}, \dot{g} \rangle_{L^2([0,1], dt)},$$

and  $F : \Omega \rightarrow \mathbb{R}$  be the Wiener functional. We already know (by Lemma 4.3.11) that

$$E' = \{ \mu \mid \mu \text{ is a signed measure on } [0, 1] \text{ of bounded variation such that } \mu(1) = 0 \}$$

and

$$E' \subset H' \xrightarrow{R} H \subset E$$

continuous and densely, where  $R$  denotes the Riesz map.

**Lemma 4.5.20.** *Let  $\mu \in E'$ . Then*

$$R(\mu) = \int_0^1 \mu(]s, 1]) ds \left( = - \int_0^1 \mu([0, s[) ds \right).$$

(Cf. (4.3.6) as a special case.)

*Proof.* Let  $h \in H$ . Then

$$\begin{aligned} \langle R(\mu), h \rangle_H = \mu(h) &= \int_0^1 h(t)\mu(dt) = \int_0^1 \int_0^t \dot{h}(s) ds \mu(dt) = \int_0^1 \int_0^1 1_{[0,t[}(s)\dot{h}(s) ds \mu(dt) \\ &\stackrel{\text{Fubini}}{=} \int_0^1 \int_0^1 1_{]s,1]}(t)\mu(dt)\dot{h}(s) ds = \int_0^1 \mu(]s, 1])\dot{h}(s) ds \\ &= \int_0^1 \frac{d}{ds} \left( \int_0^s \mu(]r, 1]) dr \right) \dot{h}(s) ds = \left\langle \int_0^1 \mu(]s, 1]) ds, h \right\rangle_H \end{aligned}$$

Hence  $R(\mu) = \int_0^1 \mu(]s, 1]) ds$ . □

We recall

**Definition 4.5.21.**  $F : C([0, 1])_0 \rightarrow \mathbb{R}$  is called Fréchet-differentiable in  $\omega \in C([0, 1])_0$ , if there exists  $F'(\omega) \in E'$  such that

$$F(\omega + \eta) = F(\omega) + F'(\omega)(\eta) + o(\|\eta\|), \quad \forall \eta \in C([0, 1])_0.$$

In this case

$$\nabla F(\omega) := R(F'(\omega)) \in H$$

is called gradient of  $F$  at  $\omega$ . Note that  $\nabla F(\omega)$  is in the "tangent space" of  $H$  in  $\omega$ .

**Remark 4.5.22.** Because of Lemma 4.5.20 we have for the measure

$$F'(\omega)(dt) =: F'(\omega, dt)$$

that

$$\nabla F(\omega)(\cdot) = R(F'(\omega, dt)) = \int_0^\cdot F'(\omega, ]s, 1]) ds. \quad (4.5.12)$$

Then by definition of the derivative

$$\lim_{\lambda \rightarrow 0} \frac{F(\omega + \lambda\eta) - F(\omega)}{\lambda} = \langle \nabla F(\omega), \eta \rangle_H \stackrel{(4.5.12)}{=} \int_0^1 F'(\omega, ]s, 1]) \dot{\eta}(s) ds \quad \forall \eta \in H(\subset E := C([0, 1])_0).$$

**Definition 4.5.23.**  $F \in \mathcal{L}^2(P)$  is called  $H$ -differentiable, if for all  $(\mathcal{F}_t)$ -adapted real processes  $(u_s)_{s \in [0, 1]}$ , product-measurable, bounded and

$$U_t(\omega) := \int_0^t u_s(\omega) ds, \quad t \in [0, 1] \quad (H - \text{vector field on } E)$$

(i.e.  $U(\omega) \in H$ ) there exists a  $\mathcal{F}/\mathcal{B}(H)$ -measurable map  $\nabla F : E \rightarrow H$  such that

$$E(\|\nabla F\|_H^2) < \infty \quad (\nabla F \in L^2(\Omega \rightarrow H, P))$$

and

$$\frac{F(\omega + \lambda U(\omega)) - F(\omega)}{\lambda} \xrightarrow{\lambda \rightarrow 0} \langle \nabla F(\omega), U(\omega) \rangle_H \quad \text{in } \mathcal{L}^2(P)$$

or equivalently

$$\nabla_U F := \lim_{\lambda \rightarrow 0} \frac{F(X + \lambda U) - F(X)}{\lambda} = \langle \nabla F, U \rangle_H \quad \text{in } \mathcal{L}^2(P).$$

Here,  $\nabla F$  is called the Malliavin gradient (cf. [Wat84]).

Define the Malliavin derivative

$$D_t F := (\nabla F(\omega))^\bullet(t).$$

In particular,  $(D_t F)_{0 \leq t \leq 1}$  is a process. (This process is product-measurable in  $(\omega, t)$ !)

Geometric interpretation of  $\nabla F : H$ -vector field on  $E$ .

Geometric interpretation of  $DF : L^2([0, 1], dt)$ -vector field on  $E$ .

**Remark 4.5.24.** Let  $u$  and  $U$  as in Definition 4.5.23.

(i) We have

$$\langle \nabla F(\omega), U(\omega) \rangle_H = \langle D_t F(\omega), u \rangle_{L^2([0, 1], dt)}$$

(ii) Let

$$Z_t^\lambda := \exp \left( \lambda \int_0^t u_s dX_s - \frac{1}{2} \lambda^2 \int_0^t u_s^2 ds \right).$$

Then Novikov's condition is fulfilled, since  $u$  is bounded. So, Girsanov's theorem implies that  $X^\lambda := X - \lambda U$  is a Wiener process under  $P^\lambda := Z_1^\lambda P$ . Hence,

$$E_P[F(X^\lambda) Z_1^\lambda] = \int F(X^\lambda) dP^\lambda \stackrel{P^\lambda \circ (X^\lambda)^{-1} = P}{=} \int F(X) dP = E_P[F(X)] \quad (4.5.13)$$

**Lemma 4.5.25.**

$$\lim_{\lambda \rightarrow 0} \frac{Z_1^\lambda - 1}{\lambda} = \int_0^1 u_s dX_s \quad \text{in } \mathcal{L}^2(P).$$

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In fact this limit exists even in  $\mathcal{L}^p(P)$  for all  $p \geq 1$  (exercise).

*Proof.* By Itô

$$Z_t^\lambda = 1 + \lambda \int_0^t Z_s^\lambda u_s \, dX_s. \quad (\star)$$

Hence, for  $t = 1$

$$\frac{Z_1^\lambda - 1}{\lambda} - \int_0^1 u_s \, dX_s = \int_0^1 (Z_s^\lambda - 1) u_s \, dX_s. \quad (4.5.14)$$

But by Ito-isometry

$$E \left[ \left( \int_0^1 (Z_s^\lambda - 1) u_s \, dX_s \right)^2 \right] = E \left[ \int_0^1 (Z_s^\lambda - 1)^2 u_s^2 \, ds \right] \leq \|u\|_\infty^2 E \left[ \int_0^1 (Z_s^\lambda - 1)^2 \, ds \right].$$

$Z_s^\lambda - 1$  is a martingale, since  $Z_s^\lambda$  is in  $\mathcal{L}^2(P)$  (see below), thus,  $(Z_s^\lambda - 1)^2$  is a submartingale. Therefore,

$$E \left[ \left( \int_0^1 (Z_s^\lambda - 1) u_s \, dX_s \right)^2 \right] \leq \|u\|_\infty^2 E \left[ (Z_1^\lambda - 1)^2 \right]$$

It remains to prove that  $(Z_1^\lambda - 1) \rightarrow 0 \in \mathcal{L}^2(P)$  as  $\lambda \rightarrow 0$ . Clearly,  $(Z_1^\lambda - 1) \rightarrow 0$   $P$ -a.s. But the set  $\{(Z_1^\lambda - 1)^2 | 0 \leq \lambda \leq 1\}$  is uniformly  $P$ -integrable, because

$$E[(Z^\lambda)^p] \stackrel{4.4.13}{\leq} \exp \left[ \frac{1}{2} \lambda^2 p(p-1) \left\| \int_0^1 u_s^2 \, ds \right\|_\infty \right] \leq \exp \left[ \frac{1}{2} p(p-1) \lambda^2 \|u\|_\infty^2 \right],$$

i.e.  $\{Z_1^\lambda | 0 \leq \lambda \leq 1\}$  is  $\mathcal{L}^p$  bounded for all  $p \geq 2$ . Hence, the assertion follows by Lebesgue's dominated convergence theorem.  $\square$

**Proposition 4.5.26** (Bismut's integration by parts formula on Wiener space). *Let  $u, U$  be as in Proposition 4.5.23 and let  $F : E \rightarrow \mathbb{R}$  be  $H$ -differentiable. Then*

$$(E [\langle D.F, u \rangle_{L^2([0,1], dt)}]) = E[\langle \nabla F, U \rangle_H] = E \left[ F \underbrace{\int_0^1 u_s \, dX_s}_{=-\operatorname{div} U = D^* u} \right]. \quad (\Rightarrow u \in \operatorname{Dom} D^*)$$

**Remark 4.5.27.** *Proposition 4.5.26 identifies a duality between  $D$  and  $\int \cdot \, dX$  (i.e. between the Malliavin derivative and the Ito-integral). This is the starting point for defining an extension of the Itô-integral, namely the Skorohod integral. This extension is simply defined as the adjoint  $D^*$  of  $D$ . Note that  $\operatorname{dom} D^*$  contains also non- $(\mathcal{F}_t)$ -adapted processes (cf. Lectures on Malliavin calculus!).  $\operatorname{dom} D$  is the set of all  $H$ -differentiable functions  $F : E \rightarrow L^2(E, P_0)$ ,*

$$D : \operatorname{dom} D \subset L^2(E, P) \rightarrow L^2(E \rightarrow L^2([0, 1], dt), P).$$

Hence,

$$D^* : \operatorname{dom} D^* \subset L^2(E \rightarrow L^2([0, 1], dt), P) \rightarrow L^2(E, P).$$

*Proof of Proposition 4.5.26.* (4.5.13) implies

$$E_{(P)} \left[ \frac{F(X^\lambda) Z_1^\lambda}{\lambda} \right] = E_{(P)} \left[ \frac{F(X)}{\lambda} \right].$$

Therefore, (recall  $X(\omega) = \omega$ )

$$E \left[ \frac{F(X^\lambda) - F(X)}{\lambda} Z_1^\lambda \right] = E \left[ -F(X) \frac{Z_1^\lambda - 1}{\lambda} \right] \xrightarrow{\lambda \rightarrow 0} E \left[ -F \int_0^1 u_s \, dX_s \right]$$

by Lemma 4.5.25. But the left hand side is equal to

$$E \left[ \left( \frac{1}{\lambda} (F(X^\lambda) - F(X)) + \langle \nabla F, U \rangle_H \right) Z_1^\lambda \right] - E \left[ \langle \nabla F, U \rangle_H Z_1^\lambda \right].$$

Since  $\langle \nabla F, U \rangle_H \in \mathcal{L}^2(E, P)$  and  $Z_1^\lambda \xrightarrow{\lambda \rightarrow 0} 1$  in  $\mathcal{L}^2(P)$  (see Proof of 4.5.25), the second summand converges to  $-E[\langle \nabla F, U \rangle_H]$  as  $\lambda \rightarrow 0$ . The first summand converges to 0 by Cauchy-Schwarz as  $\lambda \rightarrow 0$ , and the assertion follows.  $\square$

**First application:**

Identification of the integrand in Ito's representation theorem.

**Corollary 4.5.28** (Clark-Formula). *Let  $F \in \mathcal{L}^2(P)$  be  $H$ -differentiable. Then for  $\mathcal{F}_t := \mathcal{F}_{t+}^0$  with  $\mathcal{F}_t^0 := \sigma(X_s | s \leq t)$ ,  $t \in [0, 1]$ ,*

$$F = E[F] + \int_0^1 E[D_t F | \mathcal{F}_t] dX_t \quad P\text{-a.s.}$$

**Exercise:** Show that the process  $(D_t F)_{0 \leq t \leq 1}$  has a version which is a  $\mathcal{B}([0, 1]) \otimes \mathcal{F}$ -measurable function. Hint: First consider  $N_t(\omega) = 1_{[a, b](t)} 1_A(\omega)$  for  $A \in \mathcal{F}$  instead of  $D_t F$ . (Recall  $\bar{P}_x(dt, d\omega) = d\langle X \rangle_t(\omega) P(d\omega) = dt P(d\omega)$ ).

*Proof of 4.5.28.* Without loss of generality  $E[F] = 0$ . Let  $G \in \mathcal{L}^2(P)$ . Then by Corollary 2.5.44

$$G = E[G] + \int_0^1 u_t dX_t,$$

where  $u \in \mathcal{L}^2(\bar{\Omega}, \bar{\mathcal{F}}, \bar{P}_X)$  and  $u$  is  $(\mathcal{F}_t)$ -adapted. Then by Bismut's Integration by Parts-formula we have for

$$u^{(n)} := (u \wedge n) \vee (-n), \quad n \in \mathbb{N},$$

that

$$\begin{aligned} E[FG] &= \lim_{n \rightarrow \infty} E \left[ F \int_0^1 u_t^{(n)} dX_t \right] \stackrel{4.5.26}{=} \lim_{n \rightarrow \infty} E \left[ \int_0^1 D_t F u_t^{(n)} dt \right] \\ &\stackrel{\text{Fubini}}{=} \lim_{n \rightarrow \infty} \int_0^1 E[D_t F u_t^{(n)}] dt = \int_0^1 E[D_t F u_t] dt \\ &\stackrel{u \text{ adapted}}{=} E \left[ \int_0^1 E[D_t F | \mathcal{F}_t] u_t dt \right] \\ &\stackrel{2.3.22}{=} E \left[ \left( \int_0^1 E[D_t F | \mathcal{F}_t] dX_t \right) G \right]. \end{aligned}$$

and the assertion follows, since  $G \in \mathcal{L}^2(P)$  was arbitrary.  $\square$

**Example 4.5.29.** (i) Define

$$F := \int_0^1 X_t dt.$$

Note that  $E \ni \omega \mapsto F(\omega) = \int_0^1 X_t dt$  is linear and continuous on  $E$ , hence  $F \in E'$ . Then  $E' \ni F'(\omega, dt) = dt - \delta_0$ , hence by (4.5.12) for all  $t \in [0, 1]$

$$D_t F(\omega) = F'(\omega, ]t, 1]) = 1 - t.$$

So, by 4.5.28

$$\int_0^1 X_t dt = F = \int_0^1 (1 - t) dX_t.$$

Note that this can be also (in fact much more easily) proved by Itô's product rule (see 2.5.46).

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(ii)  $F := f(X_1)$ , where  $f \in \mathcal{C}^2(\mathbb{R})$ . Then by the chain rule (which also holds for Frechet-differentiable functions on Banach spaces)  $F$  is Frechet-differentiable and

$$F'(\omega, dt) = f'(X_1(\omega))\delta_1 - f'(X_1(\omega))\delta_0 \quad (\in E').$$

Thus,

$$D_t F(\omega) = F'(\omega, ]t, 1]) = f'(X_1(\omega)).$$

Hence, by Corollary 4.5.28

$$f(X_1) = F = E[f(X_1)] + \int_0^1 E[f'(X_1)|\mathcal{F}_t] dX_t. \quad (\star)$$

$F$  depends only on time  $t = 1$ , but is represented as an integral along paths. The corresponding integrands can be interpreted in terms of solutions  $h(X_t(\omega), t)$  to the following “final value problem”)

$$\begin{aligned} h(\cdot, 1) &= f(\cdot), \\ \frac{1}{2}h_{xx} + h_t &= 0. \end{aligned}$$

The solution has the form  $h(x, t) = p_{1-t}f(x)$  (as easy to see by elementary calculation, cf. Chapter III). Hence by Itô's formula (cf. 1.3.1(iii))

$$h(X_1, 1) = h(0, 0) + \int_0^1 h_x(X_s, s) dX_s, \quad (\star\star)$$

where  $h(X_1, 1) = f(X_1)$  and  $h(0, 0) = p_1 f(0) = E[f(X_1)]$ . By  $(\star)$  and  $(\star\star)$  and uniqueness in Itô's representation theorem we can conclude that

$$E[f'(X_1)|\mathcal{F}_s] = h_x(X_s, s) = (\partial_x p_{1-s} f)(X_s) = (\partial_x (E[f(X_{1-s} + x)]))_{x=X_s}.$$

To show that  $F \in \mathcal{L}^2(P)$  is  $H$ -differentiable is rather difficult in general. The following sufficient condition might be useful to check  $H$ -differentiability.

**Proposition 4.5.30.** *The following is a sufficient condition for the  $H$ -differentiability of  $F$  in  $\mathcal{L}^2(P)$ :*

*There exists a kernel  $F'(\omega, dt)$  from  $\Omega$  to  $\mathcal{B}([0, 1])$  such that for all  $U$  as in 4.5.23*

(i)

$$\frac{F(X + \lambda U) - F(X)}{\lambda}(\omega) \xrightarrow{\lambda \rightarrow 0} \int_0^1 F'(\omega, dt) U_t(\omega) \quad \text{for } P\text{-a.e. } \omega \in E.$$

(ii) For all  $c > 0$

$$|F(X + U) - F(X)| \leq c \|U\|_\infty \quad P\text{-a.s.}$$

In this case

$$(H \ni) \nabla F(\omega) = R(\underbrace{F'(\omega, dt) - F'(\omega, [0, 1]) \cdot \delta_0}_{=: \mu_\omega(dt) \in E'}) = \int_0^\cdot F'(\omega, ]s, 1]) ds.$$

*Proof.* By 4.5.20 we have

$$\int F'(\omega, dt) U_t(\omega) \stackrel{U_0=0}{=} \int \mu_\omega(dt) U_t(\omega) = \langle R(\mu_\omega), U(\omega) \rangle_H = \langle \nabla F(\omega), U(\omega) \rangle_H.$$

Hence, the assertion follows by Lebesgue's dominated convergence theorem.  $\square$

**Example 4.5.31.**  $F(\omega) := \max_{0 \leq t \leq 1} \omega(t)$  is not Frechet-differentiable, but it is  $H$ -differentiable. To see this, define

$$T := \inf\{t > 0 | X_t = F\}.$$

Then it follows by Proposition 4.5.30 that  $F$  is  $H$ -differentiable and

$$D_t F(\omega) = \delta_{T(\omega)}(\cdot | t, 1] = 1_{\{T > t\}}(\omega)$$

(cf. Exercises).

**Next step:** Identify  $E[D_t F | \mathcal{F}_t]$  in order to use the Clark formula.

Define

$$M_t := \max_{0 \leq s \leq t} X_s.$$

Then we have for  $P$ -a.e.  $\omega \in E$

$$\begin{aligned} E[D_t F | \mathcal{F}_t](\omega) &= P[T > t | \mathcal{F}_t](\omega) \\ &= P\left(\max_{t \leq s \leq 1} X_s > M_t | \mathcal{F}_t\right)(\omega) \\ &= P\left(\max_{0 \leq s \leq 1-t} X_{s+t} > M_t | \mathcal{F}_t\right)(\omega) \end{aligned}$$

Now, we use the superstrong Markov property, i.e. for any  $\mathcal{F} \otimes \mathcal{F}_t$ -measurable positive or bounded function  $G : E \times E \rightarrow \mathbb{R}$  we have

$$E_x[G(\vartheta_t, \cdot) | \mathcal{F}_t](\omega) = E_{X_t(\omega)}[H(\cdot, \omega)] \text{ for } P_x\text{-a.e. } \omega \in E,$$

where  $P_x$  is the law of Brownian motion started at  $x \in \mathbb{R}$  and  $\vartheta_t(\omega) := \omega(\cdot + t)$ .

Hence,

$$\begin{aligned} E[D_t F | \mathcal{F}_t](\omega) &= P_{X_t(\omega)}\left(\max_{0 \leq s \leq 1-t} X_s > M_t(\omega)\right) \\ &= P_{X_t(\omega)}\left(\max_{0 \leq s \leq 1-t} X_s - X_t(\omega) > M_t(\omega) - X_t(\omega)\right) = P_0\left(\max_{0 \leq s \leq 1-t} X_s > M_t(\omega) - X_t(\omega)\right) \end{aligned}$$

By the reflection principle, this is equal to

$$\begin{aligned} &2 \cdot P_0(X_{1-t} > M_t(\omega) - X_t(\omega)) \\ &= 2 \cdot N(0, (1-t))(\cdot | M_t(\omega) - X_t(\omega), \infty] \\ &= 2 \cdot N(0, 1)\left(\frac{M_t(\omega) - X_t(\omega)}{\sqrt{1-t}}, \infty\right] \\ &= 2 \int_{\frac{M_t(\omega) - X_t(\omega)}{\sqrt{1-t}}}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} dx \\ &= 2 \cdot \left(1 - \Phi\left(\frac{M_t(\omega) - X_t(\omega)}{\sqrt{1-t}}\right)\right). \end{aligned}$$

(Vgl. hierzu [Röc11, Beispiel 10.2.8].)



## 5. Brownian motion and potential theory

**Attention:** This chapter has not been proofread yet!

### 5.1. Brownian motion and Laplace Operator

Let  $(\Omega, \mathcal{F}, (X_t)_{t \geq 0}, (P_x)_{x \in \mathbb{R}^d})$  be the canonical Brownian Motion on  $\mathbb{R}^d$ , i.e.:

$$\Omega := C([0, \infty[, \mathbb{R}^d), \quad X_t(\omega) = \omega(t), \quad t \geq 0, \quad \mathcal{F} := \sigma(X_t | t \geq 0), \quad \mathcal{F}_t := \bigcap_{\varepsilon > 0} \sigma\{X_s | s \leq t + \varepsilon\},$$

$P_x :=$  Pushforward measure of the Wiener measure  $P_0$  under  $\omega \mapsto x + \omega$

Let  $(p_t)_{t > 0}$  be the corresponding semi-group of kernels, i.e. for  $f : \mathbb{R}^d \rightarrow \mathbb{R}_+$ ,  $\mathcal{B}(\mathbb{R}^d)$ -measurable,  $t > 0$ ,

$$\int f(X_t) dP_x = E_x[f(X_t)] = \int f(y) p_t(x, dy) = \int f(y) \frac{1}{(2\pi t)^{\frac{d}{2}}} e^{-\frac{\|x-y\|^2}{2t}} dy, \quad x \in \mathbb{R}^d.$$

Let  $D \subset \mathbb{R}^d$ , open.

$$T_D := \text{exit time from } D \quad (:= \inf\{t > 0 | X_t \notin D\} = \sigma_{D^c}).$$

$\forall x \in D$  the following holds for  $P_x$ -a.e.  $\omega$  and all  $t \in [0, T_D(\omega)[$  (cf. I.2.72):

$$f(X_t(\omega)) - \underbrace{f(X_0(\omega))}_{=f(x)} = \int_0^t \nabla f(X_s) dX_s(\omega) + \frac{1}{2} \int_0^t \Delta f(X_s(\omega)) ds, \quad \text{if } f \in C^2(D)$$

In particular

$$M_t^f(\omega) := f(X_t(\omega)) - f(X_0(\omega)) - \frac{1}{2} \int_0^t \Delta f(X_s(\omega)) ds$$

is Itô-Integral with respect to Brownian Motion with

$$\langle M^f \rangle(\omega) = \int_0^t \|\nabla f(X_s)\|^2 ds. \quad (*)$$

**Corollary 5.1.1.** (i)  $\forall f \in C^2(D)$   $(M_t^f)$  is local martingale (until  $T_D$ ) under  $P_x \forall x \in D$  (cf. I.3.4.), i.e. Brownian Motion is solution for the (so called) martingale problem for  $(\mathcal{L} := \frac{1}{2}\Delta, C^2(D))$  with initial condition  $x \in D$  (cf. [SV06]).

(ii) **Preliminary remark:**  $D_0$  open rel. kp with  $\bar{D}_0 \subset D \Rightarrow E_x[T_{D_0}] < \infty \quad \forall x \in D_0$  (cf. III.3.3.3). Hence, because of (\*),  $E_X[\langle M^f \rangle_{T_{D_0}}] < \infty \quad \forall x \in D_0$

$\xrightarrow{I.3.5.}$  Dynkin formula: Let  $T$  be stopping time,  $\leq T_{D_0} (< T_D \text{ as } \bar{D}_0 \subset D)$ , then  $\forall x \in D_0$

$$E_x[f(X_t)] - f(x) = \frac{1}{2} E_x \left[ \int_0^T \Delta f(X_s) ds \right]$$

5. Brownian motion and potential theory

**Application 5.1.2:** (i)  $T = T_\varepsilon =$  exit time from open  $\varepsilon$ -ball  $K_\varepsilon(x)$  around  $x$ . Then there exists for  $f \in C^2(D)$

$$\mathbf{a}f(x) := \lim_{\varepsilon \searrow 0} \frac{E_x[f(X_{T_\varepsilon})] - f(x)}{E_x[T_\varepsilon]} \quad \text{“characteristic (Dynkin) operator” (local),}$$

and we have

$$\mathbf{a}f(x) = \frac{1}{2}\Delta f(x), \quad \mathbf{a} = \frac{1}{2}\Delta \text{ on } C^2(D) \text{ for Brownian Motion (cf. “special script”)}$$

(ii)  $D = \mathbb{R}^d, f \in C_0^2(\mathbb{R}^d)$ :

$$Af(x) := \lim_{t \downarrow 0} \frac{p_t f(x) - f(x)}{t} = \lim_{t \downarrow 0} \frac{E_x[f(X_t)] - f(x)}{t} = \frac{1}{2}\Delta f(x)$$

the “infinitesimal operator or generator” of the semi-group.

$$A = \frac{1}{2}\Delta \text{ on } C_0^\infty(\mathbb{R}^d) \text{ for Brownian Motion}$$

Attention: In what way  $\frac{1}{2}\Delta$  really generates  $(p_t)_{t>0}$  and with it the Brownian Motion, hence in particular is uniquely determined, can not be seen from above (cf. “special script” or [MR92]).

(iii) It follows from the Dynkin formula with  $T_r =$  exit time from  $K_r(x_0)$  and  $f(x) = \|x - x_0\|^2$ , since  $\frac{1}{2}\Delta f = d, \forall x \in K_r(x_0)$  that

$$\begin{aligned} E_x[\underbrace{(X_{T_r} - x_0)^2}_{=r^2}] - \|x - x_0\|^2 &= dE_x[T_r] \\ \implies E_x[T_r] &= \frac{r^2 - \|x - x_0\|^2}{d}. \end{aligned}$$

## 5.2. Stochastic solution of the Dirichlet- resp. Poisson-Problem

$D \subset \mathbb{R}^d$ , open, rel. *kp*.

$T =$  exit time from  $D$ , then  $E_x[T] < \infty \quad \forall x \in D$ .

**Poisson problem**

Have:  $f \in C(\partial D), g \in C_b(D)$

Want:  $h \in C^2(D) \cap C(\bar{D})$  with

$$\begin{aligned} \frac{1}{2}\Delta h &= g \text{ on } D \\ h &= f \text{ on } \partial D \end{aligned}$$

If  $g \equiv 0$ :

**Dirichlet problem**

**Theorem 5.2.3** (“Representation Theorem”). *Let  $h$  be a solution of the Poisson problem, then*

$$\forall x \in D \quad h(x) = E_x[f(X_{T_D})] - E_x \left[ \int_0^{T_D} g(X_s) ds \right]$$

5.2. Stochastic solution of the Dirichlet- resp. Poisson-Problem

*Proof.* Let  $D_n \nearrow D$ ,  $\bar{D}_n \subsetneq D$ , i.e. in particular  $\bar{D}_n$  kp.

$$\begin{aligned} \xRightarrow{\text{Dynkin formula}} \forall x \in D_n \forall n \quad E_x \left[ \underbrace{h(X_{T_{D_n}})}_{\nearrow X_{T_D} \in \partial D} \right] - h(x) &= E_x \left[ \int_0^{T_{D_n}} g(X_s) ds \right]_{T_{D_n} \nearrow T_D} \\ \xrightarrow[n \rightarrow \infty]{\text{Lebesgue}} E_x [f(X_{T_D})] - h(x) &= E_x \left[ \int_0^{T_D} g(X_s) ds \right] \end{aligned}$$

□

We consider now the case  $g \equiv 0$ , i.e. a Dirichlet problem. For  $D_0$  open, rel. kp in  $D$  we define for  $x \in D_0$ .

$$\mu_x^{D_0}(A) := P_x \left[ X_{T_{D_0}} \in A \right], \quad A \in \mathcal{B}(\partial D_0)$$

Then we get

**Corollary 5.2.4** (“generalized mean value property”). *Let  $h$  be harmonic in  $D$ . Then  $\forall D_0$  open, rel. kp in  $D$*

$$h(x) = E_x \left[ h(X_{T_{D_0}}) \right] = \int_{\partial D_0} h \, d\mu_x^{D_0}, \quad x \in D_0.$$

**Definition 5.2.5.**  $\mu_x^{D_0}$  is called the harmonic measure for  $D_0$  in  $x$  ( $x \in D_0$ ). (=exit distribution  $P_x \circ X_{T_{D_0}}^{-1}$  of the Brownian motion from  $D_0$  with start in  $x$ )

**Remark 5.2.6.** *It follows from the definition of the Brownian motion, that it is translation- and rotation invariant. Thus,  $\mu_x^{K_r(x)}$  is also rotation invariant.*

$$\implies \mu_x^{K_r(x)} = \text{const} \cdot \underbrace{\sigma_{x,r}}_{\substack{\text{norm. surface} \\ \text{measure}}} \implies \mu_x^{K_r(x)}(\partial K_r(x))=1 \implies \mu_x^{K_r(x)} = \sigma_{x,r}.$$

More generally, we have  $\forall y \in K_r(x)$

$$\mu_y^{K_r(x)}(dz) = r^{n-2} \underbrace{\frac{r^2 - \|y-x\|^2}{\|y-z\|^n}}_{\text{Poisson kernel for } K_r(x)} d\sigma_{x,r}$$

**Theorem 5.2.7** (“Existence of a generalized solution of the Dirichlet problem”).  $D \subset \mathbb{R}^d$ , open,  $f$  bounded, measurable on  $\partial D$ . Then we have for

$$h(x) := E_x [f(X_{T_D})], \quad x \in D :$$

(i)  $h$  harmonic on  $D$

(ii)  $\lim_{t \nearrow T_D} h(X_t) = f(X_{T_D})$   $P_x$ -a.s  $\forall x \in D$ .

*Proof.* Set  $T := T_D$  and let  $D_n \nearrow D$ ,  $D_n$  open,  $\bar{D}_n \subset D_{n+1} \subset D$ . Set  $T_n := T_{D_n}$ .  $x \in D$  fixed. Without loss of generality :  $x \in D_1$ .



**Preparation for the proof**

Let  $A_{r,R} := \{x \in \mathbb{R}^d \mid r < \|x\| < R\}$  and  $f = \begin{cases} 1 & \text{in } \{x \mid \|x\| = r\} \\ 0 & \text{in } \{x \mid \|x\| = R\} \end{cases}$  be a solution to the corresponding Dirichlet problem

$$h_{r,R}(x) := E_x [f \circ X_{T_{A_{r,R}}}] = P_x [\|X_{T_{A_{r,R}}}\| = r].$$

One can easily show that analytically (write  $\Delta$  in spherical coordinates)

$$h_{r,R}(x) = \frac{\varphi_d(\|x\|) - \varphi_d(R)}{\varphi_d(r) - \varphi_d(R)} \quad \text{with } \varphi_d(\rho) = \begin{cases} \rho & , d = 1 \\ -\log \rho & , d = 2 \\ \rho^{2-d} & , d \geq 3 \end{cases} \quad (5.3.1)$$

(cf. Ruinproblem W.theorie II)

**Theorem 5.3.9.**  $\forall d \geq 2$  every point in  $\mathbb{R}^d$  is polar, i.e.:

$$P_y[\sigma_{\{x\}} < \infty] = 0 \quad \forall y \in \mathbb{R}^d.$$

(with  $\sigma_{\{x\}} := \inf\{t > 0 \mid X_t = x\}$ .)

*Proof.* Without loss of generality :  $x = 0$  (because of translation invariance)

(i)  $y \neq x = 0$  :

$$\begin{aligned} P_y[\sigma_{\{0\}} < \infty] &= \lim_{n \rightarrow \infty} P_y[\sigma_{\{0\}} < \sigma_{K_n(0)^c}] \\ &\overset{\nearrow}{\lim_{n \rightarrow \infty}} P_y - \text{a.s.} \\ &\text{( since for } P_y - \text{a.a. } \omega \{X_s(\omega) \mid 0 \leq s \leq t\} \text{ bounded } \forall t) \\ &\leq \overline{\lim}_{n \rightarrow \infty} \overline{\lim}_{\substack{r \rightarrow 0 \\ r < \|y\|}} P_y [\|X_{T_{A_{r,n}}}\| = r] \\ &= \overline{\lim}_{n \rightarrow \infty} \overline{\lim}_{\substack{r \rightarrow 0 \\ r < \|y\|}} h_{r,n}(y) = 0 \quad \text{because of (5.3.1), since } d \geq 2. \end{aligned}$$

(ii)  $y = x$  :

$$\begin{aligned} P_x[\sigma_{\{x\}} < \infty] &\leq \lim_{t \downarrow 0} P_x[\sigma_{\{x\}} \circ \theta_t < \infty] \\ &\stackrel{\text{(MP)}}{=} \lim_{t \downarrow 0} E_x \left[ \underbrace{P_{X_t}[\sigma_{\{x\}} < \infty]} \right] \\ &\stackrel{1)}{=} 0 \quad P_x\text{-a.s.,} \\ &\quad \text{since } X_t \neq x \quad P_x\text{-a.s.} \end{aligned}$$

□

**Remark 5.3.10.** For  $x \notin K_r(0)$

$$\begin{aligned} P_x[\sigma_{K_r(0)} < \infty] &= \lim_{n \rightarrow \infty} P_x \left[ \overbrace{\|X_{T_{A_{r,n}}}\| = r}^{\nearrow_n \nearrow P_x\text{-a.s.}} \right] \\ &\stackrel{\text{(5.3.1)}}{=} \begin{cases} 1, & d = 2 \\ \frac{\|x\|^{2-d}}{r^{2-d}}, & d \geq 3. \end{cases} \end{aligned}$$

## 5. Brownian motion and potential theory

Now:

*Proof of Theorem 5.3.8.* Let  $x_0 \in \mathbb{R}^d$ , Without loss of generality  $x_0 = 0$ .  $T_n :=$  exit time from  $K_n(0)$ , then  $\forall x \in \mathbb{R}^d$

$$\begin{aligned}
 P_x \left[ \underbrace{\lim_{t \uparrow \infty} \|X_t\| < r}_{\substack{= \bigcap_n [\sigma_{K_r(0)} \circ \theta_{T_n} < \infty] \\ \searrow n \nearrow}} \right] &= \lim_{n \rightarrow \infty} P_x [\sigma_{K_r(0)} \circ \theta_{T_n} < \infty] \\
 &\stackrel{\text{(SMP)}}{=} \lim_{n \rightarrow \infty} E_x [P_{X_{T_n}} [\sigma_{K_r(0)} < \infty]] \\
 &\stackrel{5.3.10}{=} \lim_{n \rightarrow \infty} \begin{cases} 1 & , d = 2 \\ \frac{n^{2-d}}{r^{2-d}} \xrightarrow{n \rightarrow \infty} & , d = 3. \end{cases}
 \end{aligned}$$

□

### 5.4. Regularity of boundary points

Let  $D \subset \mathbb{R}^d$ , open, rel. kp.,  $\partial D := \bar{D} \cap \overline{(\mathbb{R}^d \setminus D)}$

**Definition 5.4.11.**  $z \in \partial D$  is called regular, if  $P_z[\sigma_{D^c} = 0] = 1$ .

**Remark 5.4.12.** Since  $\{\sigma_{D^c} = 0\} \in \mathcal{F}_0 := \bigcap_{\varepsilon > 0} \mathcal{F}_\varepsilon$ , we have accordingly to Blumenthal's zero-one law a priori  $P_x[\sigma_{D^c} = 0] \in \{0, 1\}$ .

**Example 5.4.13.** Let  $D = U \setminus \{z\}$  with  $U$  rel. kp. open,  $z \in U$ . Then  $z \in \partial D$  and  $z$  is irregular, if  $d \geq 2$ , since  $z$  polar. We consider the behaviour of the stochastic solution of the Dirichlet problem for  $D$  with boundary condition  $f = 1_{\{z\}} + 0 \cdot 1_{(\partial D) \setminus \{z\}}$

$$\begin{aligned}
 h(x) &= E_x [f(X_{T_D})] \\
 &= 0 \cdot P_x [X_{T_D} \in \partial D \setminus \{z\}] + 1 \cdot \underbrace{P_x [X_{T_D} \in \{z\}]}_{\substack{=0, \\ \text{if } d \geq 2, \text{ since then } z \text{ polar.}}} \\
 &\equiv 0
 \end{aligned}$$

If we understand the above as an “interference” of the solution of the Dirichlet problem for  $D \cup \{z\}$ , we then obtain “stability”.

**Theorem 5.4.14.**  $z \in \partial D$  regular  $\Leftrightarrow \forall f$  bounded, measurable on  $\partial D$ , continuous in  $z$  we have:

$$h(x) := E_x [f(X_{T_D})] \xrightarrow[x \in D]{x \rightarrow z} f(z)$$

*Proof.* We only proof “ $\Rightarrow$ ”: Let in the following always  $x \in D$ .

**Claim:**  $\lim_{x \rightarrow z} P_x [X_{T_D} \notin K_\delta(z)] = 0 \quad \forall \delta > 0$ . Here  $K_\delta(z)$  open ball.

If this claim is true, then select  $\delta = \delta(\varepsilon)$ , such that  $|f(x) - f(z)| < \varepsilon \quad \forall x \in K_\delta(z) \cap \partial D$ . It follows that

$$|h(x) - f(z)| \leq E_x [|f(X_{T_D}) - f(z)|] \leq \varepsilon + 2\|f\|_\infty P_x [X_{T_D} \notin K_\delta(z)]$$

$$\Rightarrow \overline{\lim}_{x \rightarrow z} |h(x) - f(z)| \leq \varepsilon \quad \forall \varepsilon > 0, \text{ hence “}\Rightarrow\text{” proved.}$$

□

*Proof of the claim.* Let  $\delta > 0$  fixed. If  $X_0(\omega) = x \in K_{\frac{\delta}{2}}(z)$ , then  $\forall s > 0$ :

$$X_{T_D(\omega)} \notin K_\delta(z) \implies T_D(\omega) > s \text{ or } T_{K_{\frac{\delta}{2}}}(x) \leq s.$$

So we have  $\forall x \in K_{\frac{\delta}{2}}(z) \quad \forall s > 0$

$$P_x [X_{T_D} \notin K_\delta(z)] \leq P_x [T_D > s] + P_x \left[ T_{K_{\frac{\delta}{2}}}(x) \leq s \right]$$

Because of the translation invariance we get

$$P_x \left[ T_{K_{\frac{\delta}{2}}}(x) \leq s \right] = P_0 \left[ T_{K_{\frac{\delta}{2}}}(0) \leq s \right] \left( \begin{array}{c} \downarrow \\ s \downarrow 0 \end{array} P_0 \left[ T_{K_{\frac{\delta}{2}}}(0) = 0 \right] \right) \begin{array}{c} = \\ \text{because of right} \\ \text{continuous paths} \end{array} 0 \Bigg) < \varepsilon \quad \text{for } s \text{ small}$$

Let  $s > 0$ . To show:

$$y \mapsto \varphi_s(y) := P_y [T_D > s], \quad y \in \mathbb{R}^d, \text{ is not semi-continuous in } y := z$$

Because then  $0 \leq \overline{\lim}_{x \rightarrow z} \varphi_s(x) \leq \varphi_s(z) \stackrel{z \text{ regular}}{=} 0$ .

But  $\forall y \in \mathbb{R}^d$

$$\begin{aligned} \varphi_s(y) = P_y [T_D > s] &= \lim_{t \downarrow 0} P_y \left[ \underbrace{t + T_D \circ \theta_t}_{\text{first exit time after } t} > s \right] \\ &= P_y [T_D \circ \theta_t > s - t] \\ &\stackrel{MP}{=} E_y \left[ \underbrace{P_{X_t} [T_D > s - t]}_{\varphi_{s-t}(X_t)} \right] \\ &= \lim_{t \downarrow 0} \underbrace{p_t \varphi_{s-t}(y)}_{\substack{\text{cont. as } p_t \\ \text{strong Feller}}} \Bigg\} \text{n.o.h.cont.} \end{aligned}$$

□

Now an analytic criterion for regularity:

**Theorem 5.4.15** (“Zaremba’s cone condition”). *Let  $z \in \partial D$  such that there exists a cone  $C$  with apex in 0 and  $h > 0$  with  $z + C_h \subseteq D^c$ , where  $C_h := C \cap K_h(0)$ . Then  $z$  is regular.*

*Proof.*

$$\begin{aligned} P_z [T_D \leq t] &\geq P_z [X_t \in z + C_h] \quad \forall t \in ]0, 1[ \\ &= P_0 [X_t \in C_h] \\ &= P_0 \left[ \sqrt{t} X_{\frac{t}{(\sqrt{t})^2}} \in C_h \right], \text{ since } \left( c X_{\frac{t}{c^2}} \right)_{t \geq 0} \text{ again B.M.; cf. W.th.II, ch. III, Satz 5.1} \\ &= P_0 \left[ X_1 \in t^{-\frac{1}{2}} C_h \right] \\ &\stackrel{(t \leq 1)}{\geq} P_0 [X_1 \in C_h] > 0 \end{aligned}$$

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$$\begin{aligned} \implies P_z[T_D = 0] &= \lim_{t_n \searrow 0} P_z[T_D \leq t_n] \geq P_0[X_1 \in C_h] > 0 \quad \forall n \\ &\stackrel{\text{Blumentahl's}}{\implies} P_z[T_D = 0] = 1. \\ &\quad \text{0-1 law} \end{aligned}$$

□

Idea:  $D_c$  shouldn't be too "small" in  $z$ .

Exercise: for  $d = 2$   $C$ -line is enough.

In general  $(z + C_h) \cap H \subseteq D^c$  is enough, where  $H$  hypersurface (cf. [?], p. 276 or 350, analytical with Wiener theorem).

**Example 5.4.16** ("Lebesgue-sting"). *Let*

$$\begin{aligned} I_n &:= [2^{-n}, 2^{-(n-1)}] \\ k_n &:= \{(x_1, x_2, x_3) \in \mathbb{R}^3 \mid x_1 \in I_n, x_2 = x_3 = 0\} \\ F_n &:= \{(x_1, x_2, x_3) \in \mathbb{R}^3 \mid x_1 \in I_n, x_2^2 + x_3^2 \leq \varepsilon_n^2\} \\ D &= \text{cylinder} \setminus \bigcup_n F_n \\ &\quad \text{(height } z) \end{aligned}$$

We want to choose  $\varepsilon_n$  such that  $0 \in \partial D$  irregular.

$d = 3$ :  $\implies P_0[X_0 \text{ never hit } K_n] = 1$ , since

$$\begin{aligned} &P_0[X_0 \text{ never goes back to } \{x_2 = x_3 = 0\}] \quad (\leftarrow \text{no condition for } X_0^{(1)}) \\ &= P_0^{d=2}[B.M. \text{ in } \mathbb{R}^2 \text{ never goes back to } 0] = 1 \quad , 0 \text{ polar!} \end{aligned}$$

Since Brownian Motion in  $\mathbb{R}^3$  is transient  $\exists$  for  $P_0$ -a.a.  $\omega$  an  $s(\omega) \geq 0$  with  $\|X_t(\omega)\| \geq 2$   $\forall t \geq s(\omega)$ . It is obvious that

$$a_n(\omega) := \text{dist}(\{X_t(\omega) \mid 0 \leq t \leq s(\omega)\}, K_n) > 0$$

(as both sets kp. and disjoint). Hence  $\{a_n \geq \frac{1}{m}\} \nearrow_{m \uparrow \infty} \Omega$   $P$ -a.s., thus  $\exists \varepsilon_n$  with  $P_0[a_n \leq \varepsilon_n] \leq 3^{-n}$ .

Then

$$\begin{aligned} &P_0[\underbrace{X_t \in F_n \text{ for a } t}_{\{\sigma_{F_n} < \infty\}}] \leq P_0[a_n \leq \varepsilon_n] \leq 3^{-n} \\ \implies &P_0[T_D = 0] \leq \sum_n P_0[\underbrace{X_t \in F_n \text{ for a } t}_{\{\sigma_{F_n} < \infty\}}] \leq \sum 3^{-n} < 1 \\ \implies &0 \text{ irregular.} \end{aligned}$$

**Excursus to Brownian motion in  $\mathbb{R}^1$ .** Without loss of generality  $a \geq 0$

Define *passage time*  $T_a := \inf\{t > 0 \mid X_t > a\}$ .

Because  $X_{T_a} = a$  and continuity of paths, we have:

$$\begin{aligned} T_a &= \inf\{t > 0 \mid X_t \geq a\} = \inf\{t > 0 \mid X_t = a\} \\ P &:= P_0 = \text{Brownian Motion with start in } 0. \end{aligned}$$

**Theorem 5.4.17** (Reflection principle).

$$\frac{1}{2}P[T_a \leq t] = P[X_t \geq a] \quad \left( = \frac{1}{2}P[|X_t| \geq a] \right)$$

*Proof.*

$$\begin{aligned} P[X_t \geq a] &= P[T_a \leq t, \underbrace{X_{t-T_a}(\theta_{T_a})}_{=X_t \text{ since canonical model}} \geq a] \\ &= E[\Phi(\cdot, \theta_{T_a(\cdot)}(\cdot)); T_a \leq t], \end{aligned}$$

where  $\Phi(\omega, \eta) = 1_{[a, \infty[} \circ X_{t-T_a(\omega)}(\eta)$ ,  $(\omega, \eta) \in \Omega \times \Omega = C([0, \infty]) \times C([0, \infty])$ .  
 $\Phi$  is  $\mathcal{F}_{T_a} \otimes \mathcal{F}$ -measurable. We need:

**Lemma 5.4.18** (“(very) SMP”). *Let  $T$  be stopping time,  $\Phi : \Omega \times \Omega \rightarrow \mathbb{R}^+$   $\mathcal{F}_T \otimes \mathcal{F}$ -measurable. Then it follows for  $H(\omega) := \Phi(\omega, \theta_{T(\omega)}(\omega))$  and all  $x \in \mathbb{R}^d$*

$$E_x[H|\mathcal{F}_T](\omega) = E_{X_T(\omega)}[\Phi(\omega, \cdot)] \quad P_x\text{-a.a. } \omega \in \Omega.$$

*Proof.* Let  $F$   $\mathcal{F}_T$ -measurable,  $G$   $\mathcal{F}$ -measurable and  $\Phi(\omega, \eta) = F(\omega)G(\eta)$ . Then because of SMP for  $P_x$ -a.a.  $\omega \in \Omega$

$$E_x[F G \circ \theta_T | \mathcal{F}_T](\omega) = F(\omega) E_x[G \circ \theta_T | \mathcal{F}_t](\omega) = F(\omega) E_{X_T(\omega)}[G] = E_{X_T(\omega)}[F(\omega)G].$$

The rest follows with monotone classes. □

**Continuation of proof 5.4.17.**

According to the lemma: for  $P$ -a.a.  $\omega \in \Omega$

$$\begin{aligned} &E[\Phi(\cdot, \theta_{T_a(\cdot)}(\cdot)); T_a \leq t](\omega) \\ &= \int_{\{T_a \leq t\}} \underbrace{P_{X_{T_a(\omega)}}[X_{t-T_a(\omega)}(\cdot) \geq a]}_{=P_s(a, [a, \infty]) = \frac{1}{2}} P(d\omega) = \frac{1}{2} P[T_a \leq t]. \end{aligned}$$

□

**Corollary 5.4.19** (“Distribution of the passage time”).  $P \circ T_a^{-1} = \varphi_a \cdot ds$ , where

$$\varphi_a(s) = \begin{cases} \frac{|a|}{\sqrt{2\pi s^3}} e^{-\frac{a^2}{2s}} & , s > 0 \\ 0 & , s \leq 0. \end{cases}$$

(“stable with exponent  $a$ ”, since then  $E[e^{ixT_a}] = e^{-|x|^a}$ )

*Proof.*

$$\begin{aligned} P[T_a \leq t] &\stackrel{5.4.17}{=} 2P[X_t \geq a] \\ &= 2 \int_a^\infty \frac{1}{\sqrt{2\pi t}} e^{-\frac{x^2}{2t}} dx \\ &= \int_0^t \underbrace{s^{-\frac{3}{2}} \frac{a}{\sqrt{2\pi}} e^{-\frac{a^2}{2s}}}_{\varphi_a(s)} ds. \end{aligned}$$

since  $\frac{a^2}{2s} = \frac{x^2}{st} \implies x = a\sqrt{\frac{t}{s}}$  and  $\frac{dx}{ds} = a\sqrt{t}(-\frac{1}{2})s^{-\frac{3}{2}}$  □

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**Corollary 5.4.20.** (i)  $P[T_a > t]\sqrt{t} \xrightarrow[t \nearrow \infty]{} \sqrt{\frac{2}{\pi}}a. (a \geq 0)$

(ii)  $E[T_a] = +\infty$

*Proof.* (i) :

$$P[T_a > t]\sqrt{t} \stackrel{5.4.17}{=} \sqrt{t}P[|X_t| \leq a] = \frac{1}{\sqrt{2\pi}} \int_{-a}^a e^{-\frac{x^2}{2t}} dx \xrightarrow[t \nearrow \infty]{} \frac{2a}{\sqrt{2\pi}}.$$

(ii) :

$$\begin{aligned} E[T_a] &= \int_0^\infty P[T_a > t] dt \\ &\stackrel{(i)}{\geq} \int_{t_0}^\infty \left( \sqrt{\frac{2}{\pi}}a - \varepsilon \right) t^{-\frac{1}{2}} dt = \infty \end{aligned}$$

(i)  $\varepsilon$  small  
 $t_0 = t_0(\varepsilon)$  big

□

**Example 5.4.21** (Solution of the Dirichlet problem on the half-space.).

$D := \{x = (x_1, \dots, x_d) \in \mathbb{R}^d | x_d > 0\}$ . Consider  $D$  not rel. kp. as always assumed before! But an essential assumption for the existence of a solution is only  $P_x[T_D < \infty] = 1 \forall x \in D$ . (cf. 5.2.7 and 5.4.15). Since  $\partial D := \{z = (y, 0) | y \in \mathbb{R}^{d-1}\}$  no longer kp, we have to assume  $f \in C_b(\partial D)$ .

**Stochastic:**

Cone condition fulfilled  $\forall z \in \partial D!$

**Calculation of the harmonic measure**

Under  $P_x$  the exit time  $T := T_D = \inf\{t > 0 | X_t^d = 0\}$  (=passage time) has density  $\varphi_a$  from 5.4.19 with  $a = x_d$ . With notation  $\omega = (\omega_1, \dots, \omega_d) \in (C[0, \infty))^d$ ,  $X_T := (Y_T, 0)$  it follows  $\forall x \in D$

$$\begin{aligned} h(x) &\stackrel{2.5!}{=} E_x[f(X_T)] = E_{(x_1, \dots, x_{d-1})} \left[ \underbrace{E_{x_d}[f(Y_{T(\omega_d)}(\omega_1, \dots, \omega_{d-1}), 0)]}_{\stackrel{5.4.19}{=} \int_0^\infty f(Y_t(\omega_1, \dots, \omega_{d-1}), 0) \underbrace{\frac{x_d}{\sqrt{2\pi t^3}} e^{-\frac{x_d^2}{2t}} dt}_{\varphi_{x_d}(t)}} \right] \\ &\stackrel{\text{Fubini}}{=} \int_0^\infty \frac{x_d}{\sqrt{2\pi t^3}} e^{-\frac{x_d^2}{2t}} \left( \int_{\mathbb{R}^{d-1}} f(y, 0) \frac{1}{\sqrt{2\pi t}^{d-1}} \exp\left[-\frac{\|y - (x_1, \dots, x_{d-1})\|^2}{2t}\right] dy \right) dt \\ &\stackrel{\text{Fubini}}{=} \int_{\mathbb{R}^{d-1}} f(y, 0) \frac{x_d}{\sqrt{2\pi}^d} \underbrace{\int_0^\infty t^{-(\frac{d}{2}+1)} \exp\left[-\frac{\|(y, 0) - x\|^2}{2t}\right] dt}_{=\Gamma(\frac{d}{2})\|(y, 0) - x\|^{-d} 2^{\frac{d}{2}}} dy \end{aligned}$$

because:  $\int_0^\infty t^{-\beta} e^{-\frac{\alpha}{t}} dt \stackrel{s=\frac{\alpha}{t}}{=} \alpha^{1-\beta} \int_0^\infty s^{\beta-2} e^{-s} ds = \alpha^{1-\beta} \Gamma(\beta - 1)$

$$\implies \mu_x^D(dy) = \frac{\Gamma(\frac{d}{2})}{\pi^{\frac{d}{2}}} \frac{x_d}{\|x - (y, 0)\|^d} \underbrace{\lambda_{d-1}}_{\substack{\text{Leb.meas.} \\ \text{on } \mathbb{R}^{d-1}}} (dy)$$

**Remark 5.4.22.** The stochastic solution of the Dirichlet problem leads us to the harmonic measure  $\mu_x^D =$  exit distribution of the Brownian Motion. For the Poisson problem we get Green function = residence density of the BM in  $D$ . We want to show this in the next section.

## 5.5. Poisson equality and Green function

Assumption:  $P_x[T_D < \infty] = 1 \forall x \in D$ . ( $D$  not necessarily bounded)

Have  $f \in C_b(D)$ . Want  $u \in C^2(D) \cap C(\bar{D})$  with

$$\begin{cases} \frac{1}{2}\Delta u = -f & \text{on } D \\ u = 0 & \text{on } \partial D \end{cases} \quad \left\| \begin{array}{l} \text{Poisson problem with homog. boundary condition} \end{array} \right.$$

**Remark 5.5.23.** The general case  $u = g$  on  $\partial D$  can be traced back to the above by adding the solution of the Dirichlet problem with boundary condition  $g$ . We then know, because of 5.2.3, that

$$u(x) = E_x \left[ \int_0^T f(X_s) ds \right], \quad x \in D. \quad (5.5.2)$$

with  $T := T_D$  ( $(\mathcal{F}_t)$ -stopping time with  $\mathcal{F}_t := \bigcap_{\varepsilon > 0} \{\sigma(X_s) | s \leq t + \varepsilon\}$ ).

**Remark 5.5.24.**  $p_t^D(x, A) := P_x[X_t \in A, t < T]$ ,  $A \subseteq D$ , measurable, defines a measure on  $D$ . It is obvious that

$$p_t^D(x, \cdot) \leq \overbrace{p_t(x, \cdot)}^{\text{B.sgr.}} \ll dy \quad (= \text{Lebesgue-measure on } \mathbb{R}^d)$$

$$\implies \exists p_t^D(x, y) : p_t^D(x, dy) = p_t^D(x, y) dy.$$

**Definition 5.5.25.**

$$G^D(x, y) := \int_0^\infty p_t^D(x, y) dt \quad , \quad x, y \in D$$

is called the Green function of  $D$ .

Attention: only if  $d \geq 3$ ,  $G^D(x, y) < \infty \forall x \neq y$ . If  $d = 1, 2$  and  $D = \mathbb{R}^d$ ,  $G^D(x, y) = \infty \forall x, y$ .

**Aim:** Representation of  $p_t^D(x, y)$  and thus of  $G^D(x, y)$ . Because then for  $u$  as in (5.5.2)

$$\begin{aligned} u(x) &= E_x \left[ \int_0^\infty f(X_s) 1_{\{T > s\}} ds \right] \stackrel{\text{Fubini}}{=} \int_0^\infty \underbrace{E_x[f(X_s); T > s]}_{\int f(y) p_s^D(x, y) dy} ds \\ &= \int f(y) \int_0^\infty p_s^D(x, y) ds dy \end{aligned}$$

Thus

**Theorem 5.5.26.**  $u(x) = \int G^D(x, y) f(y) dy$  solves Poisson problem for  $D$  with boundary condition 0.

If  $G^D$  "explicit", then  $u$  explicit.

**Theorem 5.5.27.** (i)

$$p_t^D(x, y) = \overbrace{p_t(x, y)}^{\text{Density of the Br. sgr.}} - \underbrace{E_x \left[ \overbrace{p_{t-T}(X_T, y)}^{\text{cont. in } t}; \overbrace{T < t}^{\text{left cont. in } t} \right]}_{\text{left cont. (Lebesgue)}}, \quad x, y \in D$$

Note  $|y - X_T| \geq d(y, \partial D) > 0$  since  $y \in D$ .

(ii)  $p_t^D(x, y)$  is symmetric in  $x, y$ .

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*Proof.* (i)  $A \in \mathcal{B}(D)$

$$\begin{aligned}
 P_x[X_t \in A, t < T] &= P_x[X_t \in A] - \underbrace{P_x[X_t \in A, T \leq t]}_{=P_x[X_{t-T} \circ \theta_T \in A, T \leq t]} \\
 &\stackrel{5.4.18}{=} \int_{\{T < t\}} P_{X_{T(\omega)}}[X_{t-T(\omega)} \in A] dP_x(\omega) \\
 &= \int_A p_t(x, y) dy - \int_{\{T < t\}} \int_A p_{t-T(\omega)}(X_{T(\omega)}, y) dy dP_x(\omega) \\
 &= \int_A [p_t(x, y) - E_x[p_{t-T}(X_T, y), T < t]] dy
 \end{aligned}$$

$\implies$  (i)

(ii) By approximation: Let  $\overline{D_n} \subset D$ ,  $D_n \nearrow D$ ,  $\overline{D_n}$  kp.,  $T_n := T_{D_n}$ . Then  $T_n \nearrow T$  and hence  $\{T > t\} = \bigcup_n \{T_n > t\}$ . Let  $\tau_m = m$ -th dyadic decomposition of  $[0, \infty[$ .

$$\begin{aligned}
 \implies P_x[X_t \in A, T > t] &= \lim_n (\nearrow) P_x[X_t \in A, T_n > t] \\
 &= \lim_n (\nearrow) \lim_m (\searrow) \underbrace{P_x[X_t \in A, X_{t_i} \in D_n \forall t_i \in \tau_m, t_i \leq t]}_{=: q_{n,m}(x,y) \text{ is symm., since } p_s(x',y') = p_s(y',x')} \\
 &= \int_A \int_{D_n} \dots \int_{D_n} \underbrace{p_{\underbrace{t_1}_{=\frac{1}{2}m}}(x, x_1) p_{\underbrace{t_2-t_1}_{=\frac{1}{2}m}}(x_1, x_2) \dots p_{\underbrace{t-t_{M-1}}_{=\frac{1}{2}m}}(x_M, y)}_{=: q_{n,m}(x,y) \text{ is symm., since } p_s(x',y') = p_s(y',x')} dx_1 \dots dx_M dy \\
 &\stackrel{\text{Levi}}{=} \int_A \underbrace{\lim_n (\nearrow) \lim_m (\searrow) q_{n,m}(x, y)}_{=: p_t^D(x,y) \text{ symm. as limit of symm. ones}} dy
 \end{aligned}$$

□

**Corollary 5.5.28.** *Let  $d \geq 3$ .*

(i)  $G^D(x, y) = G(x, y) - E_x[G(X_T, y)]$ ,  
where

$$G(x, y) := 2^{1-\frac{d}{2}} \frac{\Gamma(\frac{d}{2} - 1)}{(2\pi)^{\frac{d}{2}}} \cdot \|x - y\|^{2-d} = \int_0^\infty p_t(x, y) dt. \quad \text{Green function on } \mathbb{R}^d$$

(ii)  $G^D$  is symm. and  $G^D(\cdot, y)$  is harmonic on  $D \setminus \{y\}$ .

*Proof.* (i) Clear, by 5.5.27 and calculation of  $\int_0^\infty p_t(x, y) dt$  (Exercise!)

(ii) Symm. clear. Also clear that  $x \mapsto E_x[G(X_T, y)]$  harmonic on  $D$  and  $\Delta G(\cdot, y) = 0$  on  $\mathbb{R}^d \setminus \{y\}$  (Exercise!)

□

## 5.6. Self intersection of the paths

$d \geq 2$ .  $t \geq 0$  fix.

$$\begin{aligned}
 P_x[\text{Returning to } X_t] &= P_x[X_s = S_t \text{ for one } s > t] & (5.6.3) \\
 &= \int 1_{\{\sigma_{X_t(\omega)} < \infty\}} \cdot \theta_t(\omega) dP_x(\omega) \\
 &\stackrel{5.4.18}{=} \int_{X_t \mathcal{F}\text{-meas.}} \underbrace{P_{X_t(\omega)}[\sigma_{X_t(\omega)} < \infty]}_{=0, \text{ since pt. polar, as } d \geq 2} dP_x(\omega) = 0
 \end{aligned}$$

Nevertheless we have  $P_x$ -a.s.

$d \leq 2$ :  $\forall n \exists n$ -times self intersection, i.e.  
 $\exists t_1(\omega) < \dots < t_n(\omega) : X_{t_1(\omega)}(\omega) = \dots = X_{t_n(\omega)}(\omega)$ .

$d \leq 3$ : the double points lie dense on path

$d = 3$ :  $\nexists$  triple points

$d \geq 4$ :  $\nexists$  double points.

We want to show now:

**Theorem 5.6.29.**  $d = 3 \implies \forall x \in \mathbb{R}^3$

$$P_x[\text{double points lie dense on path}] = 1$$

Therefore we need some preparations.

### Excursus to the capacity

$$d \geq 3 : G(x, y) := \frac{1}{\|x - y\|^{d-2}} \quad (\text{i.e. as before up to constants})$$

**Definition 5.6.30.** (i) Let  $\mu$  be positive measure on  $\mathbb{R}^d$ .

$$G\mu(x) := \int G(x, y)\mu(dy) \quad , \quad x \in \mathbb{R}^d, \quad (\in [0, \infty])$$

is called the potential generated by  $\mu$  and

$$e(\mu) := \int G\mu d\mu$$

the energy of  $\mu$ .

(ii) For  $K \subset \mathbb{R}^d$  kp.,

$$C(K) := (\inf\{e(\mu) \mid \mu \text{ p.measure on } K\})^{-1}$$

is the capacity of  $K$ .

**Remark 5.6.31.** (i)  $C(\{x\}) = 0 \forall x \in \mathbb{R}^d$ , since  $e(\varepsilon_x) = \infty$ .

(ii)  $C(K) > 0 \Leftrightarrow \exists \mu \text{ p.measure on } K : e(\mu) < \infty$

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(iii)  $C(K) = 0 \implies \lambda(K) = 0$  ( $\lambda := \text{Leb. measure}$ ),  
 ( $\neq$ ) because:

$$\lambda(K) > 0 \implies \int G \mu d\mu < \infty \text{ with } \mu := \lambda(K)^{-1} \mathbf{1}_K \cdot \lambda$$

$$= \frac{1}{\lambda(K)} \int \frac{1}{|x|^{d-2}} \underbrace{\mathbf{1}_K \times \mathbf{1}_K(x)}_{\text{kp. support}} \lambda(dx)$$

**Theorem 5.6.32** (“equilibrium principle”).  $C(K) > 0 \implies \exists$  finite measure  $\mu_K$  on  $K$  with:

$$G_{\mu_K} \leq 1, \quad = 1 \mu_K \text{ a.e. on } K$$

without proof.

In particular:  $e(\mu_K) = \|\mu_K\| := \mu_K(K)$  and for  $\tilde{\mu}_K := \frac{\mu_K}{\|\mu_K\|}$  (p.measure):

$$e(\tilde{\mu}_K) \frac{1}{\|\mu_K\|^2} e(\mu_K) = \frac{1}{\|\mu_K\|}$$

$$\implies C(K) \geq \|\mu_K\|.$$

**Theorem 5.6.33.**  $C(K) > 0 \implies h_K(x) := P_x[\overset{=\sigma_K}{T_{K^c}} < \infty] > 0$  on  $K^c$  (and in fact harmonic on  $K^c$ ).

*Proof.*  $u := G_{\mu_K}$  harmonic on  $K^c$ , since for  $x \in B_r(x_0) \subset K^c$

$$\int G_{\mu_K}(z) \mu_x^{B_r(x_0)}(dz) = \int_K \int \underbrace{G(z, y)}_{\substack{\text{harm. on} \\ \mathbb{R}^d \setminus \{y\}}} \mu_x^{B_r(x_0)}(dz) \mu_K(dy) = \int G(x, y) \mu_K(dy).$$

Clear:  $0 < u \leq 1$  and  $u(x) \leq \|\mu\|_K \sup_{y \in K} G(x, y) = \|\mu\|_K d(x, K)^{2-d} \rightarrow 0$  for  $\|x\| \rightarrow \infty$ .

Choose  $D_n$  rel. kp.  $\bar{D}_n \subseteq K^c$ ,  $D_n \nearrow K^c$ , thus  $T_n := T_{D_n} \nearrow T := T_{K^c}$ . Hence  $\forall x \in K^c$

$$\begin{aligned} 0 < u(x) &= \lim_{n \rightarrow \infty} E_x [u(X_{T_n})] \quad (u \text{ harmonice on } K^c) \\ &\leq E_x \left[ \overline{\lim}_n u(X_{T_n}) \right] \\ &= \underbrace{E_x \left[ \overline{\lim}_n u(X_{T_n}); \lim_n T_n = \infty \right]}_{\substack{=0, \text{ since on } \{\lim T_n = \infty\}: \\ \|X_{T_n}\| \rightarrow \infty \text{ because of transience}}} + \underbrace{E_x \left[ \overline{\lim}_n u(X_{T_n}); \sup_n T_n < \infty \right]}_{\substack{\leq P_x[\sup T_n < \infty] \\ = P_x[T_{K^c} < \infty]}} \end{aligned}$$

$$\implies P_x [T_{K^c} < \infty] \geq u(x) > 0. \quad \square$$

End of excursus.

**Theorem 5.6.34.** Let  $0 \leq a < b$ ,  $d = 3$ .  $\forall x \in \mathbb{R}^3$ , we have for  $P_x$ -a.a.  $\omega$ :

$$C(\underbrace{\{X_T(\omega) | a \leq t \leq b\}}_{=: K_\omega}) > 0$$

*Proof.*  $\mu_\omega :=$  Pushforward measure of  $\underbrace{\lambda|_{[a,b]}}_{\substack{1-\text{dim.} \\ \text{Leb. meas.}}}$  under  $s \mapsto X_s(\omega)$ .

(“occupation measure” for  $[a, b]$ )  
 $\mu_\omega$  measure on  $K_\omega$  and

$$e(\mu_\omega) = \int \int G(x, y) \mu_\omega(dx) \mu_\omega(dy) = \text{const.} \underbrace{\int_a^b \int_a^b \|X_s(\omega) - X_t(\omega)\|^{2-d} ds dt}_{(*)}$$

$< \infty$   $P_x$  - a.s., if  $d = 3$ , since:

$$\begin{aligned} E_x[(*)] &= \int_a^b \int_a^b E_0 [\|X_s - X_t\|^{2-d}] ds dt \\ &= \underbrace{E_0 [\|X_1\|^{2-d}]}_{< \infty} \underbrace{\int_a^b \int_a^b \frac{1}{|t-s|^{\frac{d-2}{2}}} ds dt}_{(**)} \\ &< \infty \end{aligned}$$

$$(**) = \begin{cases} < \infty & , d = 3 \\ = \infty & , d > 3 \end{cases}$$

since

$$\begin{aligned} (**) &= 2 \int_a^b \int_t^b \frac{1}{(s-t)^{\frac{d-2}{2}}} ds dt = 2 \int_a^b \underbrace{\int_0^{b-t} s^{1-\frac{d}{2}} ds}_{< \infty} dt \\ &\Leftrightarrow 1 - \frac{d}{2} > -1 \\ &\Leftrightarrow 4 > d \end{aligned}$$

and then bounded in  $t$ . □

*Proof of 5.6.29.* Let  $K_\omega := \{X_s(\omega) | 0 \leq s \leq 1\}$

$$(\omega, \eta) \mapsto T(\omega, \eta) := \sigma_{K_\omega}(\eta) \quad (:= \inf\{t > 0 | X_t(\eta) \in K_\omega\})$$

is  $\mathcal{F}_1 \otimes \mathcal{F}$ -measurable,

since  $\{T \leq t\} \stackrel{!}{=} \bigcap_n \bigcup_{z \in (2^{-n}\mathbb{Z})^3} \{\sigma_{K_{2^{-n-1}}}(z) \leq 1\} \times \{\sigma_{K_{2^{-n-1}}}(z) \leq t\} \in \mathcal{F}_1 \otimes \mathcal{F}_t$

where  $K_{2^{-n-1}}(z)$  is a bounded ball around  $z$  with radius  $2^{-n-1}$  in max. norm.

$$\begin{aligned} &\implies P_x [X_t = X_s \text{ for an } s \in [0, 1] \text{ and a } t \in ]2, \infty[ ] \\ &= P_x [\sigma_{K_0} \circ \theta_2 < \infty] \\ &= \int E_x [1_{\{\sigma_{K_0} < \infty\}} \circ \theta_2 | \mathcal{F}_2](\omega) P_x(d\omega) \\ &\stackrel{5.4.18}{=} \int \underbrace{P_{X_2(\omega)}[\sigma_{K_\omega} < \infty]}_{P_x\text{- a.s. } > 0, \text{ because:}} P_x(d\omega) > 0 \end{aligned}$$



*Proof.* Define  $Z_t := u(X_t) \cdot A_t$  with  $A_t := \exp(\int_0^t V(X_s) ds)$ . ( $C^1!$ )

$$\begin{aligned} \xRightarrow[\text{Ito's product rule}]{} Z_t &= Z_0 + \int_0^t u(X_s) \underbrace{dA_s}_{=V(X_s)A_s ds} + \int_0^t A_s du(X_s), \quad t < T \end{aligned}$$

with

$$\begin{aligned} \int_0^t A_s du(X_s) &= \lim_n \sum_{\substack{t_i \in \tau_n \\ t_i \leq t}} A_{t_i} (u(X_{t_{i+1}}) - u(X_{t_i})) \\ &\stackrel{\text{Ito}}{=} \lim_n \underbrace{\sum_{t_i \in \tau_n} A_{t_i} \left( \int_{t_i}^{t_{i+1}} (\nabla u, dX_s) \right)}_{=: M_t^n} + \frac{1}{2} \int_0^t A_s \Delta u(X_s) ds \end{aligned}$$

Let  $\bar{D}_m \subset D_{m+1}$ ,  $D_m$  open,  $D_m \nearrow D$ . Then  $(M_t^n)$  is local martingale until  $T$  with local sequence  $T_m := T_{D_m} \wedge m$ . Since for  $s \leq T_{D_m} \wedge m$ ,  $A_s \leq \exp(m\|V\|_{D_m})$ , it follows that

$$M_t := \lim_{n \rightarrow \infty} M_t^n,$$

is local martingale with local sequence  $(T_m)$  (cf. poof of I.3.2, II.1.4).

Furthermore

$$\begin{aligned} Z_t &= Z_0 + M_t + \underbrace{\int_0^t \left( \frac{1}{2} \Delta u + Vu \right) (X_s) A_s ds}_{-g(X_s)}, \quad t < T \\ \implies \underbrace{E_x[Z_{T_m \wedge t}]}_{\substack{m \rightarrow \infty \\ \text{then } t \rightarrow \infty \\ \rightarrow E_x[Z_T]}} &= \underbrace{E_x[Z_0]}_{u(x)} + \underbrace{E_x \left[ - \int_0^{t \wedge T_m} g(X_s) A_s ds \right]}_{\substack{m \rightarrow \infty \\ \text{then } t \rightarrow \infty \\ \rightarrow E_x[-\int_0^T g(X_s) A_s ds]}} \end{aligned}$$

since  $|Z_{T_m \wedge t}| \leq \|u\|_\infty \exp(\int_0^T v^+(X_s) ds) \in \mathcal{L}^1$ . □

**Example 5.7.36.**

$$\begin{aligned} \frac{1}{2} \Delta u + \alpha u &= -g \quad \text{on } D \\ u &\equiv 0 \quad \text{on } \partial D \end{aligned}$$

At least if  $\alpha < \varepsilon = \varepsilon(D)$ ,  $\varepsilon$  so small (cf. III 3.3 3)), such that  $E_x [e^{\varepsilon T}] < \infty$ ,  $x \in D$ . Then

$$u(x) = E_x \left[ \int_0^{T_D} e^{\alpha t} g(X_t) dt \right]$$

## 5.8. The heat equation

$D = \mathbb{R}^d \times ]0, T]$ .

Have:  $f \in C(\mathbb{R}^d)$  with  $\exists b, c > 0 : |f(x)| \leq be^{c\|x\|^2} \forall x \in \mathbb{R}^d$  (\*)

Want:  $v \in C^2(D)$ , continuous on  $\mathbb{R}^d \times [0, T]$

$$\begin{cases} \frac{\partial v}{\partial t} &= \frac{1}{2} \Delta v \text{ on } D \quad (\Delta \text{ on } \mathbb{R}^d) \\ v(\cdot, 0) &= f \text{ on } \mathbb{R}^d \end{cases} \quad (5.8.5)$$

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Set  $v(x, t) := \underset{\substack{\uparrow \\ \text{Br.sgr.}}}{p_t} f(x) = E_x[f(X_t)] = \frac{1}{\sqrt{2\pi t^d}} \underbrace{\int f(y) e^{-\frac{\|y-x\|^2}{2t}} dy}_{\text{Exists only if (*) is fulfilled}}, (x, t) \in D.$

Then  $v$  is a solution of (5.8.5) if  $T < \frac{1}{2c}$  (exercice).  $\implies$  existence

**Dual heat equation on  $\mathbb{R}^d \times [0, T]$ :**

$$\left\{ \begin{array}{l} \frac{\partial u}{\partial t} + \frac{1}{2}\Delta u = 0 \text{ on } \mathbb{R}^d \times [0, T[ \\ u(\cdot, T) = f \text{ on } \mathbb{R}^d \end{array} \right. \left( \begin{array}{l} u \in C^2(D), u \text{ cont. on } \mathbb{R}^d \times [0, T]; \\ f \in C(\mathbb{R}^d) \end{array} \right) \quad (5.8.6)$$

Solution via time reversal on  $T$ , i.e.: Let  $v$  be solution of (5.8.5), then

$$\begin{aligned} u(x, t) &:= v(x, T - t) \quad \text{solution of (5.8.6)} \\ &= E_x[f(X_{T-t})], \quad (x, t) \in \mathbb{R}^d \times [0, T] \\ &\text{if } T < \frac{1}{2c} \end{aligned}$$

$\implies$  existence for 5.8.6

**Theorem 5.8.37 (Uniqueness).** Let  $f \in C(\mathbb{R}^d)$  with (\*) and  $T < \frac{1}{2c}$ .

Then  $\exists!$  solution  $u \in C^2(D)$  of  $\frac{\partial u}{\partial t} + \frac{1}{2}\Delta u = 0$  on  $\mathbb{R}^d \times [0, T[$  with

(i)  $\lim_{\substack{t \nearrow T \\ y \rightarrow x}} u(y, t) = f(x)$

(ii)  $\sup_{0 \leq t < T} |U(x, t)| \leq K e^{a\|x\|^2}; K, a > 0.$

, and  $u(x, t) = E_x[f(X_{T-t})]$  is this solution.

*Proof.*  $u(x, t) := E_x[f(X_{T-t})] = p_{T-t}f(x)$  fulfills (i), (ii) (Exercise!)

(with  $K = b, a = \frac{c}{1-2cT}, |p_{T-t}f(x)| \leq \text{const } e^{\frac{c\|x\|^2}{1-2c(T-t)}}$ )

Uniqueness: Let  $u$  be solution of (5.8.6) with (i), (ii). Without loss of generality :  $f = 0$ .

Need to show:  $u(x, t) = 0$

$$u(X_x, t + s) = u(X_0, t) + \underbrace{\int_0^s \nabla u(X_r, r) dX_r}_{=: M_s \text{ (Def.)}} + \underbrace{\int_0^s \left( \frac{1}{2}\Delta + \frac{\partial}{\partial r} \right) u(X_r, r) dr}_{=0, \text{ since } r < s < T-t < T}$$

but: local martingale (cf. proof of 5.3.9)

Define

$$R_n := \inf\{s > 0 \mid |X_s| > n\}, n \in \mathbb{N},$$

and let  $t_k \nearrow T - t, t_k < T - t.$

$$\begin{aligned} \implies u(x, t) &= E_x[u(X_{R_n \wedge t_k}, t + R_n \wedge t_k)] \\ &\stackrel{\substack{\text{stopping theorem} \\ + \text{remark II.1.5}}}{=} E_x[u(X_{t_k}, t + t_k); R_n > t_k] + E_x[u(X_{R_n}, t + R_n); R_n \leq t_k] \\ &= \underbrace{E_x[u(X_{t_k}, t + t_k); R_n > t_k]}_{=: I_{k,n}} + \underbrace{E_x[u(X_{R_n}, t + R_n); R_n \leq t_k]}_{\substack{\leq K e^{a\|x_{R_n}\|^2} P_x[R_n \leq t_k] \\ \text{(ii)} \\ \xrightarrow{t_k \nearrow T-t} K e^{an^2} P_x[R_n < T-t] \\ \text{since } 1_{\{R_n \leq t_k\}} \nearrow 1_{\{R_n < T-t\}}} \end{aligned}$$

since  $I_{k,n} \xrightarrow{t_k \nearrow T-t} E_x[f(X_{T-t}) ; R_n \geq T-t] = 0 \forall n$  because of (i), (ii) and because of

$$\text{Let } Q_n := \left\{ x \in \mathbb{R}^d \mid \max_{1 \leq i \leq d} |x_i| \leq \frac{n}{\sqrt{d}} \right\} \subset K_n(0)$$

$$\begin{aligned} \implies P_x[R_n < T-t] &\leq \sum_{j=1}^d P_x \left[ \max_{0 \leq s \leq T-t} X_s^j \geq \frac{n}{\sqrt{d}} \right] + P_x \left[ \max_{0 \leq s \leq T-t} (-X_s^j) \geq \frac{n}{\sqrt{d}} \right] \\ &\stackrel{5.4.17}{=} \sum_{j=1}^d 2P_x \left[ X_{T-t}^j \geq \frac{n}{\sqrt{d}} \right] + 2P_x \left[ -X_{T-t}^j \geq \frac{n}{\sqrt{d}} \right] \\ &\leq \text{const.} \sum_{j=1}^d \left[ e^{-\left(\frac{n}{\sqrt{d}} - x_j\right)^2 / 2(T-t)} + e^{-\left(\frac{n}{\sqrt{d}} + x_j\right)^2 / 2(T-t)} \right] \end{aligned}$$

where the last inequality holds because of  $\int_a^\infty e^{-\frac{x^2}{2}} dx \leq a^{-1} e^{-\frac{a^2}{2}}$

$$\implies K e^{an^2} P_x[R_n < T-t] \xrightarrow{n \rightarrow \infty} 0, \text{ if } \frac{1}{2d(T-t)} > a.$$

$$\implies u(x, t) = 0, \text{ if } t > T - \frac{1}{2ad}.$$

If we apply the above to  $T_1 := T - \frac{1}{3ad}$ , we get

$$u(x, t) = 0, \quad \text{if } t > T_1 - \frac{1}{2ad} = T - \frac{1}{3ad} - \frac{1}{2ad}$$

It follows with iteration

$$u(x, t) = 0 \quad \forall t \in [0, T], \quad x \in \mathbb{R}^d.$$

□

**More general:**

$D \subset \mathbb{R}^d \times [0, \infty[$ . We want solution of

$$\begin{cases} \left(\frac{\partial}{\partial t} + \frac{1}{2}\Delta\right)u &= 0 & \text{on } D \\ u &= f & \text{on } \partial D \end{cases} \quad (5.8.7)$$

**Idea:** Interpretation as *Dirichlet problem for the "space-time process"*.

We define for  $r \geq 0$ ,  $\sigma_r : [0, \infty[ \rightarrow [0, \infty[$  by  $\sigma_r(s) := r + s$ .

$$\text{Let } \bar{\Omega} := \Omega \times \{\sigma_r \mid r \in [0, \infty[\}$$

$$\bar{\omega} = (\omega, \sigma_r) \quad (\text{sarting point } t)$$

$$\bar{X}_s(\bar{\omega}) = (X_s(\omega), r + s) \in \mathbb{R}^d \times [0, \infty[$$

$$P_{(x,t)} := P_x \otimes \delta_{\sigma_r}.$$

If we have again everything "enough" bounded, we get

$$u(x, t) = E_{(x,t)}[f(\bar{X}_{\bar{T}})]$$

with  $\bar{T}(\bar{\omega}) := \inf\{t > 0 \mid \bar{X}_t(\bar{\omega}) \notin D\}$ .

In the special case above:  $D := \mathbb{R}^d \times [0, T[$  with  $f : \partial D = \mathbb{R}^d \times \{T\} \rightarrow \mathbb{R}$  we have

$$\bar{T} = T - t \quad P_{(x,t)} - \text{ a.s., i.e.}$$

$$\begin{aligned} u(x, t) &= E_{(x,t)}[f(X_{T-t}, \sigma_0(T-t))] \\ &= E_x[f(X_{T-t}; T)] \end{aligned}$$

Hence as before.



## 6. Stochastic Differential Equations

### 6.1. Solution definitions, examples

We consider the following SDE:

$$\begin{cases} X_0 = \xi_0 & \text{(initial condition)} \\ dX_t = \sigma_t(X)dW_t + b_t(X)dt \end{cases} \quad (6.1.1)$$

For the sake of simplicity we will mostly consider only the Markov case:

$$\begin{cases} \sigma_t(x) = \sigma(X_t, t) \\ b_t(X) = b(X_t, t) \end{cases} \quad (6.1.2)$$

and only in  $\mathbb{R}^1$  (thus  $X_t \in \mathbb{R}$ ), where we assume that

$$\sigma = \sigma(x, t), \quad b = b(x, t) \quad \text{Borel measurable on } \mathbb{R} \times [0, \infty[. \quad (6.1.3)$$

Let the right continuous filtration  $(\mathcal{F}_t)_{t \geq 0}$  be as in the following definitions:

$$\mathcal{F}_t \supset \text{completion of } \bigcap_{\varepsilon > 0} \sigma(\xi_0, W_s; s \leq t + \varepsilon) \quad \text{with respect to } P \text{ in } \mathcal{F},$$

where  $\xi_0 : \Omega \rightarrow \mathbb{R}$ ,  $\mathcal{F}$ -measurable.

**Definition 6.1.1.** (i) A weak solution of (6.1.1) is a triple  $(X, W)$ ,  $(\Omega, \mathcal{F}, P)$ ,  $(\mathcal{F}_t)_{t \geq 0}$  with:  $W$   $(\mathcal{F}_t)$ -BM on  $(\Omega, \mathcal{F}, P)$ ,  $X$   $(\mathcal{F}_t)$ -adapted, such that  $P$ -a.s.

$$X_t = \xi_0 + \underbrace{\int_0^t \sigma_s(X) dW_s}_{\text{in part. } \exists} + \underbrace{\int_0^t b_s(X) ds}_{\text{in part. } \exists} \quad \forall t \geq 0 \text{ (resp. } \forall t < \zeta : \text{stopping time)}. \quad (6.1.4)$$

(ii) This solution is called weakly unique, if the distribution of  $\xi_0$  (on  $\mathbb{R}$ ) determines uniquely the distribution of  $X$ .

**Definition 6.1.2.** (i) (6.1.1) has a strong solution, if  $\exists F : \mathbb{R} \times C([0, \infty[)_0 \rightarrow C([0, \infty[)_0$ , such that  $F(x, \cdot)$  is  $\mathcal{F}_t^0 / \mathcal{F}_t^0$ -measurable  $\forall x, t$  and  $\forall (\Omega, \mathcal{F}, P)$ ,  $(\mathcal{F}_t)$ , and all  $(\mathcal{F}_t)$ -BM's  $W$ ,

$$X := F(\xi_0, W)$$

fulfills (6.1.4) (" $F$  is the solution"). Here  $\mathcal{F}_t^0 := \sigma(\pi_s | s \leq t)$  with  $\pi_s : C([0, \infty[)_0 \rightarrow \mathbb{R}$ ,  $\pi_s(\omega) = \omega(s)$ .

(ii) This solution is called strong unique, if  $\forall (X, W)$ ,  $(\Omega, \mathcal{F}, P)$ ,  $(\mathcal{F}_t)_{t \geq 0}$ ,  $\xi_0$ , which fulfill (6.1.4)

$$X = F(\xi_0, W)$$

("in other words  $F$  is uniquely determined") ( $\implies$  weak uniqueness)

6. Stochastic Differential Equations

**Remark 6.1.3.**  $\mathcal{F}_t/\mathcal{F}_t$ -measurability of  $F(x, \cdot)$  implies:  $X$  adapted

**Example 6.1.4.** (i) The Ornstein-Uhlenbeck process is a strong (unique) solution of  $dX = dW - \alpha X dt$  (cf. solution formula in Section 4.1)

(ii)  $dX = \alpha X dW + \beta X dt$  has as strong (unique) solution

$$X_t := X_0 \exp \left[ \alpha W_t + \left( \beta - \frac{1}{2} \alpha^2 \right) t \right]$$

(Heuristically: Itô  $\implies$

$$\begin{aligned} \ln X_t &= \ln X_0 + \int_0^t \frac{1}{X_t} dX_t - \frac{1}{2} \int_0^t \frac{1}{X_t^2} \overbrace{d\langle X \rangle_t}^{=\alpha^2 X_t^2 d\langle W \rangle_t} \\ &= \ln X_0 + \alpha W_t + \beta t - \frac{1}{2} \alpha^2 t \\ \implies X_t &= X_0 \exp \left[ \alpha W_t + \left( \beta - \frac{1}{2} \alpha^2 \right) t \right] \end{aligned}$$

except for “ $X_t$  could become 0”  $\approx$  proof of existence.)

*Proof.* strong solution: Check with Itô  $\forall X_0, W$ .

Uniqueness: Let  $(X, W, X_0)$  with (ii). Show with Itô:

$$X_t \exp \left[ -\alpha W_t - \left( \beta - \frac{1}{2} \alpha^2 \right) t \right] = X_0$$

□

(iii) Tanaka’s example (shows that one is possibly forced to look for a weak solution).

$$\begin{cases} X_0 = 0 \\ dX = \text{sign}(X) dW \end{cases}, \text{ where } \text{sign}(x) := \begin{cases} 1 & , x \geq 0 \\ -1 & , x < 0 \end{cases}$$

( $\text{sign}(x) \neq 0$ , according to our definition, excludes the trivial solution  $X \equiv 0$ !)

**Claim:**  $\exists!$  weak solution.

*Proof.* weak uniqueness: Let  $X$  be a solution  $\implies$

$$\begin{aligned} X_t &= \int_0^t \text{sign}(X_s) dW_s \text{ is martingale} \\ \langle X \rangle_t &= \int_0^t \underbrace{(\text{sign}(X_s))^2}_{=1} ds = t. \end{aligned}$$

$\implies$   $X$  is a Wiener process.  
Levy

Existence: Let  $X$  be a Wiener process

$$\begin{aligned} \implies W_t &= \int_0^t \text{sign}(X_s) dX_s \text{ is a Wiener process} \\ \implies X_t &= \int_0^t \underbrace{\text{sign}(X_s) \text{sign}(X_s)}_{=1} dX_s \\ &= \int_0^t \text{sign}(X_s) dW_s \end{aligned}$$

(by Cor. 2.4.34) is a weak solution.

One can show that  $\nexists$  strong solution (cf. [KS91]).

□

## 6.2. Construction of solutions via transformation of the state space

**Example 6.2.5.** (i)  $X_t := W_t^2$  fulfills (according to Itô)

$$dX_t = 2W_t dW_t + dt$$

Let

$$\tilde{W}_t := \int_0^t \text{sign} W_s dW_s \quad \text{Wiener process}$$

$$\implies 2W_t dW_t = 2|W_t| \text{sign} W_s dW_s = 2|W_t| d\tilde{W}_t = 2\sqrt{X_t} d\tilde{W}_t,$$

i.e.  $dX_t = 2\sqrt{X_t} d\tilde{W}_t + dt$  has as weak solution  $X_t := W_t^2$ .

$X$  large: strong diffusion! Although its drift is  $+1$ ,  $X$  will be going back to 0 again and again (since it is the square of BM and therefore the law of the iterated log holds.)

(ii) **Bessel process**  $\|W\|$

$$X := \|W\| = \sqrt{\sum_{i=1}^n (W^i)^2}, \quad W = (W^1, \dots, W^n), \quad \text{BM in } \mathbb{R}^n \text{ with start } W_0 = 0.$$

Let

$$Y := X^2 = \sum_{i=1}^n (W^i)^2 \quad (\implies X = \sqrt{Y})$$

$$\begin{array}{l} \implies \\ \text{multidim.} \\ \text{Itô} \end{array} \quad dY = \sum 2W^i dW^i + ndt$$

and

$$d\langle Y \rangle_t = \sum 4(W^i)^2 dt = 4Y dt.$$

Hence  $\tilde{W}_t := \int_0^t \frac{1}{2\sqrt{Y}} dY - \frac{nt}{2\sqrt{Y_t}} = \int_0^t \frac{1}{\sqrt{Y}} W^i dW^i$  is a local martingale with  $\langle \tilde{W} \rangle_t = \int_0^t \frac{1}{4Y_s} \underbrace{d\langle Y \rangle_s}_{4Y_s ds} = t$ , hence  $\tilde{W}$  BM, since  $\tilde{W}$  is a local martingale.

Apply Itô-formula to  $X = \sqrt{Y}$  (ok for  $n \geq 2$ , since  $0 \in \mathbb{R}^n$  polar, i.e.  $Y_t > 0 \forall t > 0$  P-a.s.) to get

$$\begin{aligned} dX &= \frac{1}{2\sqrt{Y}} dY + \frac{1}{2} \left( -\frac{1}{4} Y^{-\frac{3}{2}} d\langle Y \rangle \right) \\ &= d\tilde{W} + \frac{1}{2\sqrt{Y}} ndt - \frac{1}{8} Y^{-\frac{3}{2}} 4Y dt \\ &= d\tilde{W} + \frac{n-1}{2X} dt. \end{aligned}$$

That is, the Bessel process  $X := \|W\|$  is a weak solution of

$$dX = d\tilde{W} + \frac{n-1}{2X} dt \quad (\text{strongly drifting away from 0!})$$

**Theorem 6.2.6** (Zvonkin). Let  $b : \mathbb{R} \rightarrow \mathbb{R}$  be Borel measurable, bounded. Then

$$dX_t = dW_t + b(X_t) dt$$

has a strong solution.

6. Stochastic Differential Equations

*Proof. Idea:* Transform  $X$  into a local martingale  $Y := \varphi(X)$  (for some to be found transformation  $\varphi : \mathbb{R} \rightarrow \mathbb{R}$ ).

**Ansatz:** Let  $X$  be a solution of  $dX = dW + b(X) dt$ . Define

$$\begin{aligned} Y_t &:= \varphi(X_t) \\ \xRightarrow{\text{It\^o}} dY &= \varphi'(X)dX + \frac{1}{2}\varphi''(X)dt \\ &= \varphi'(X)dW + \{\varphi'(X)b(X) + \frac{1}{2}\varphi''(X)\}dt \end{aligned}$$

Choose  $\varphi$  such that  $\{\varphi'(X)b(X) + \frac{1}{2}\varphi''(X)\} = 0$ , so  $Y$  is a local martingale and we get rid of the (only) measurable  $b$ ! I.e.

$$\begin{aligned} \varphi'' &= -2b\varphi' \\ \implies \varphi'(y) &= C \cdot \exp\left[-\int_0^y 2b(z)dz\right] \quad (> 0) \\ \implies \varphi(x) &:= C \int_0^x \exp\left[-\int_0^y 2b(z)dz\right] dy + \tilde{C} \quad (\text{strongly increasing, } C^1) \\ &\quad (\implies \exists \varphi^{-1} : C^1) \end{aligned}$$

Show:  $\varphi(x) = \int_0^x \exp\left[-\int_0^y 2b(z)dz\right] dy$  is as desired.

Sketch: Then  $dY = \underbrace{\varphi'(\varphi^{-1}(Y))}_{=X} dW$ .

Let  $\sigma(y) := \varphi'(\varphi^{-1}(Y))$ , hence

$$dY = \sigma(Y)dW \tag{*}$$

Note that

$$\begin{aligned} \sigma(y) &= \exp\left[-\int_0^{\varphi^{-1}(y)} 2b(z)dz\right] \\ \implies \sigma'(y) &= -\sigma(y) 2b(\varphi^{-1}(y)) \underbrace{\varphi^{-1}(y)'}_{=(\varphi'(\varphi^{-1}(y)))^{-1}=\sigma(y)^{-1}} = -2b(\varphi^{-1}(y)) \end{aligned}$$

Thus  $\sigma'$  globally bounded, hence  $\sigma$  globally Lipschitz continuous:

$$|\sigma(x) - \sigma(y)| \leq 2\|b\|_\infty|x - y|$$

Thus in (\*) we have more regular coefficients than in the original equation. According to the results in Section 6.4 below, (\*) has a unique strong solution  $Y$  (as  $\sigma$  is Lipschitz).

$$\implies X = \varphi^{-1}(Y) \quad (\text{This even gives uniqueness!})$$

Problem:  $\varphi^{-1}$  only  $C^1$ , but not  $C^2$ ! Hence attention with It\^o, but  $\varphi'$  is still weakly  $C^1$ .

**Now rigorously:** Solve (\*) and set

$$X := \varphi^{-1}(Y)$$

Then check by It\^o that  $X$  solves  $dX_t = dW_t + b_t dt$ . □

Now more general:

$$dX = \sigma(X)dW + b(X)dt \tag{*}$$

### 6.3. Method of Lamperti-Doss-Sussmann

Lamperti ca. 1964

Doss-Sussmann end of the 70th

Surprisingly: One can solve (\*) **pathwise!**

For simplicity here only for  $\sigma$ ,  $b$  sufficiently smooth and for the Markoff case:

$$dX_t = \sigma(X_t)dW_t + b(X_t)dt. \quad (6.3.5)$$

We want to rewrite (6.3.5) with respect to the **Stratonovich integral**:  $\int \sigma(X) \circ dW$ .

**Definition 6.3.7.** Assume  $X, Y$  are (local) semimartingales. Then the Stratonovich integral of  $X$  against  $Y$  is defined as

$$\int_0^t X_s \circ dY_s := \int_0^t X_s dY_s + \frac{1}{2} \langle X, Y \rangle_t, \quad t \geq 0.$$

If we apply this to the semimartingales  $\sigma(X)$  and  $W$  we obtain

$$\begin{aligned} \int_0^t \sigma(X_s) \circ dW_s &= \int_0^t \sigma(X_s) dW_s + \frac{1}{2} \langle \sigma(X), W \rangle_s \\ &= \int_0^t \sigma(X_s) dW_s + \frac{1}{2} \int_0^t \sigma'(X_s) d\langle X, W \rangle_s \\ &\text{(since } d\sigma(X) = \sigma'(X)dX + \frac{1}{2}\sigma''(X)d\langle X \rangle) \\ &\stackrel{(6.3.5)}{=} \int_0^t \sigma(X_s) dW_s + \frac{1}{2} \int_0^t \sigma'(X_s) \sigma(X_s) ds. \end{aligned}$$

or in differential form

$$\sigma(X) \circ dW = \sigma(X)dW + \frac{1}{2}(\sigma'\sigma)(X)dt.$$

**Recall.** In Section 1.2 we looked at another special case of Definition 6.3.7, namely for the semimartingales  $f(X)$  and  $X$  (see p. 5) to get

$$\begin{aligned} \int_0^t f(X) \circ dX &= \int_0^t f(X) dX + \frac{1}{2} \langle f(X), X \rangle_t \\ &= \int_0^t f(X) dX + \frac{1}{2} \int_0^t f'(X) d\langle X \rangle \end{aligned}$$

(since  $df(X) = f'(X)dX + \frac{1}{2}f''(X)d\langle X \rangle$ .) Hence

$$dF(X) = F'(X) \circ dX.$$

Let  $b^*(X_t) := \{b(X_t) - \frac{1}{2}\sigma'(X_t)\sigma(X_t)\}$ , then (6.3.5) can be written as

$$dX_t = \sigma(X_t) \circ dW_t + b^*(X_t)dt \quad (6.3.6)$$

Let  $\Phi(x, t)$  be the **flow** corresponding to the ODE

$$\dot{x} = \sigma(x)$$

i.e.

$$\begin{aligned} \Phi(x, 0) &= x \\ \Phi_t &:= \frac{d}{dt} \Phi(x, t) = \sigma(\Phi(x, t)) \\ \implies \Phi_{tt} &:= \frac{d^2}{dt^2} \Phi(x, t) = \sigma'(\Phi(x, t)) \frac{d}{dt} \Phi(x, t) = (\sigma'\sigma)(\Phi(x, t)) \end{aligned}$$

6. Stochastic Differential Equations

**Ansatz:**  $X_t = \Phi(\xi_t, \eta_t)$  solves (6.3.6) with

$$\begin{aligned} \xi_0 &= X_0 \quad (= X_0(\omega)) \\ d\xi &= \dot{\xi} dt \quad (\text{in part. } t \mapsto \xi_t(\omega) \text{ is } C^1) \\ (*) \quad d\eta_t &= dW_t + \underbrace{\tilde{b}^{X_0}}_{\text{to be det.}}(\eta_t) dt, \quad \eta_0 = 0 \\ \xRightarrow[2 \text{ dim. It\^o}]{} dX_t &= \Phi_x(\xi_t, \eta_t) d\xi_t + \Phi_t(\xi_t, \eta_t) d\eta_t + \frac{1}{2} \Phi_{tt}(\xi_t, \eta_t) dt \\ &= \underbrace{\sigma(X_t) dW_t + \frac{1}{2} (\sigma' \sigma)(X_t) dt}_{\sigma(X_t) \circ dW_t} + \underbrace{\left\{ \Phi_x(\xi_t, \eta_t) \dot{\xi} + \sigma(\Phi(\xi_t, \eta_t)) \tilde{b}^{X_0}(\eta_t) \right\} dt}_{\stackrel{!}{=} b^*(\underbrace{\Phi(\xi_t, \eta_t)}_{X_t})} \end{aligned}$$

There are two possibilities to simplify :

$$\dot{\xi} = 0 \quad \text{or} \quad \tilde{b}^{X_0} = 0$$

**1st possibility: Lamperti**

Let  $\xi_t \equiv X_0$ , thus  $\dot{\xi} = 0$ . Hence it should hold that

$$\tilde{b}^{X_0}(y) = \frac{b^*}{\sigma}(\Phi(X_0, y)) \quad \forall y \in \mathbb{R}.$$

Then the problem is reduced to solve

$$\begin{aligned} d\eta_t &= dW_t + \frac{b^*}{\sigma}(\Phi(X_0, \eta_t)) dt \\ \eta_0 &= 0 \end{aligned} \quad (*)$$

(ok, if (e.g.)  $X_0 = x_0 \in \mathbb{R}$ , and  $\sigma, b$  are such that  $\frac{b^*}{\sigma}(\Phi(x_0, \cdot))$  globally Lipschitz; cf. Section 6.4 below.)

$$\implies X_t = \Phi(X_0, \eta_t) \text{ solves (6.3.6).}$$

(strong solution, if (\*) has strong solution)

**2nd possibility: Doss-Sussmann**

$$\tilde{b}^{X_0} \equiv 0, \quad \text{i.e.: } \eta_t = W_t$$

$$\implies \dot{\xi}_t = \Phi_x^{-1}(\xi_t, W_t(\omega)) b^*(\Phi(\xi_t, W_t(\omega))) \quad \forall \omega \text{ fixed, that is:}$$

$$\begin{aligned} \dot{\xi} &= F_\omega(\xi_t, t) \\ \xi_0(\omega) &= X_0(\omega) \quad , \quad \omega \text{ fixed} \end{aligned}$$

Hence  $\xi_t = \xi_t(\omega)$ , but “smooth” in  $t$  for  $\omega$  fixed. Thus the problem reduces to solve a family of certain differential equations. It follows:

$$X_t := \Phi(\xi_t, W_t) \text{ solves (6.3.6)}$$

Special case for both:

$$dX = \sigma(X) \circ dW$$

(i.e.  $b^* \equiv 0$ )

6.4. Construction of strong solutions on Lipschitz conditions (Picard-Lindelöf method)

1. Lamperti:  $d\eta = dW$ , hence  $X_t = \Phi(X_0, W_t)$
2. Doss-Sussmann:  $\dot{\xi}_t = 0$ ,  $\xi_t \equiv X_0$ , hence  $X_t = \Phi(X_0, W_t)$

**Example 6.3.8.**  $\sigma(x) = x^2$  (and  $b^* \equiv 0$ )

$$\implies \Phi(x, t) = \frac{1}{\frac{1}{x} - t}$$

According to the special case  $X_t = \Phi(X_0, W_t)$  solves

$$dX_t = \sigma(X_t) \circ dW_t \quad (\text{Stratonovich-DE!})$$

But explosion at time  $t_0 := \inf\{t > 0 | W_t = \frac{1}{x_0}\}$ .

Consider now for  $\sigma(x) = x^2$

$$\begin{aligned} dX_t &= \sigma(X)dW \quad (\text{Itô-DE!}) \\ &= \sigma(X) \circ dW - \underbrace{\frac{1}{2} \sigma'(X)}_{2 \cdot X^3} dt \end{aligned}$$

Then the strong drift  $X^3$  prevents an explosion, because due to Lamperti we get, with initial condition  $X_0 = x_0$ ,

$$X_t = \frac{1}{\frac{1}{x_0} - \eta_t},$$

where

$$\begin{aligned} d\eta_t &= dW_t + \underbrace{\frac{b^*}{\sigma}}_{-id} (\Phi(x_0, \eta_t)) dt \\ &= dW_t + \frac{1}{\eta_t - \frac{1}{x_0}} dt, \end{aligned}$$

and  $\eta_0 = 0$ .

Solution:  $\eta =$  Bessel process in  $\mathbb{R}^3$  at  $\frac{1}{x_0}$ .

( $\frac{1}{x_0}$  polar for BM in  $\mathbb{R}^3$ , hence also  $\eta$  doesn't hit  $\frac{1}{x_0}$ , thus  $(X_t)$  doesn't explode in finite time.)

## 6.4. Construction of strong solutions on Lipschitz conditions (Picard-Lindelöf method)

In this section let  $d \in \mathbb{N}$  arbitrarily,  $G : \Omega \times \mathbb{R}_+ \times \mathbb{R}^d \rightarrow \mathbb{R}^{d^2}$ ,  $b : \Omega \times \mathbb{R}_+ \times \mathbb{R}^d \rightarrow \mathbb{R}^d$  product measurable and so  $\mathcal{F}_t$ -adapted that  $\exists K > 0$  with

$$\begin{aligned} & \|G(\omega, t, x) - G(\omega, t, y)\| + \|b(\omega, t, x) - b(\omega, t, y)\| \leq K \|x - y\| \\ & \text{and } \|G(\omega, t, x)\| + \|b(\omega, t, x)\| \leq K(1 + \|x\|) \text{ for all } \omega \in \Omega, t \in \mathbb{R}_+, x, y \in \mathbb{R}^d. \end{aligned} \quad (6.4.7)$$

Let  $W$  be a given  $\mathbb{R}^d$ -valued Wiener process on  $(\Omega, \mathcal{F}, P)$ ,  $(\mathcal{F}_t)$ .

**Theorem 6.4.9.** *For an arbitrary initial condition  $\xi_0 \in \mathcal{L}^2(\Omega; \mathbb{R}^d)$ ,  $\mathcal{F}_0$ -measurable, there exists exactly one strong solution  $X$  of the stochastic differential equation (here:  $G(\cdot, t, x) =: G(t, x)$ ,  $b(\cdot, t, x) =: b(t, x)$ )*

$$dX_t = G(t, X_t)dW_t + b(t, X_t)dt, \quad X_0 = \xi_0, \quad (6.4.8)$$

## 6. Stochastic Differential Equations

i.e.,  $P$ -a.s.

$$X_t = \xi_0 + \int_0^t G(s, X_s) dW_s + \int_0^t b(s, X_s) ds \quad \forall t \geq 0$$

(in particular all integrals on the right hand side exist in the usual sense).

In particular,  $X$  has continuous sample paths.

Furthermore we have for every  $T > 0$

$$E \left[ \sup_{0 \leq t \leq T} \|X_t\|^2 \right] \leq C (1 + E [\|\xi_0\|^2]),$$

where the constant  $C$  only depends on  $K$  and  $T$ .

For the proof we need the following version of the Banach fixed-point theorem:

**Theorem 6.4.10.** *Let  $(E, d)$  be a complete metric space,  $\Lambda : E \rightarrow E$  and  $c_n \in [0, \infty[$ ,  $n \geq 1$ , with  $\sum_{n \geq 1} c_n < +\infty$  fixed, such that*

$$d(\Lambda^n x, \Lambda^n y) \leq c_n d(x, y) \quad \forall x, y \in E, \quad n \geq 1. \quad (6.4.9)$$

Then  $\Lambda$  has exactly one fixed-point  $x$  and we have for all  $y \in E$

$$d(x, y) \leq (1 + \sum_{n \geq 1} c_n) d(\Lambda y, y).$$

*Proof.* Let  $x_0 \in E$ ,  $x_n := \Lambda^n x_0$ . Then

$$\begin{aligned} d(x_{n+1}, x_n) &= d(\Lambda^n x_1, \Lambda^n x_0) \leq c_n \cdot d(x_1, x_0) \\ \implies_{n > m} d(x_n, x_m) &\leq \sum_{k=m}^{n-1} d(x_{k+1}, x_k) \leq \left( \sum_{k \geq m} c_k \right) d(x_1, x_0) \xrightarrow{m \nearrow \infty} 0. \\ \implies_{E \text{ complete}} \exists x : &= \lim_{n \rightarrow \infty} x_n \text{ and } \Lambda x = \lim_{n \rightarrow \infty} \Lambda^{n+1} x_0 = x. \end{aligned}$$

If  $\tilde{x} \in E$  is another fixed-point, we get

$$d(x, \tilde{x}) = d(\Lambda^n x, \Lambda^n \tilde{x}) \leq c_n d(x, \tilde{x}) \xrightarrow{n \rightarrow \infty} 0, \quad \text{i.e. } x = \tilde{x}.$$

If  $y \in E$ , we get

$$\begin{aligned} d(x, y) &= \lim_{n \rightarrow \infty} d(\Lambda^n y, y) \leq \overline{\lim}_{n \rightarrow \infty} \sum_{k=0}^{n-1} d(\Lambda^{k+1} y, \Lambda^k y) \\ &\leq \overline{\lim}_{n \rightarrow \infty} \left( 1 + \sum_{k=1}^{n-1} c_k \right) d(\Lambda y, y) = \left( 1 + \sum_{k \geq 1} c_k \right) d(\Lambda y, y). \end{aligned}$$

□

*Proof of 6.4.9.* Let  $E := L^2(\Omega; C([0, T]; \mathbb{R}^d))$  and  $|X|_T := E[\sup_{0 \leq t \leq T} \|X(t)\|^2]^{\frac{1}{2}}$ . Then  $(E, |\cdot|_T)$  is complete, and for  $X : \Omega \rightarrow C([0, T]; \mathbb{R}^d)$  define

$$(\Lambda X)(t) := \xi_0 + \underbrace{\int_0^t G(s, X(s)) dW_s}_{=: I_t} + \int_0^t b(s, X(s)) ds, \quad t \in [0, T].$$

6.4. Construction of strong solutions on Lipschitz conditions (Picard-Lindelöf method)

**1st step:**  $\Lambda E \subseteq E$ .

Since  $\Lambda X(\omega) \in C([0, T]; \mathbb{R}^d)$ ,  $E[\sup_{0 \leq t \leq T} I_t^2] \stackrel{\text{Doob}}{\leq} 4E[I_T^2] \stackrel{\text{isometry}}{=} 4E[\int_0^T \|G(s, X(s))\|^2 ds]$ ,  
hence

$$\begin{aligned} |\Lambda X|_T &\leq E[\|\xi_0\|^2]^{\frac{1}{2}} + 2E\left[\int_0^T \|G(s, X(s))\|^2 ds\right]^{\frac{1}{2}} + E\left[\int_0^T \|b(s, X(s))\|^2 ds\right]^{\frac{1}{2}} \\ &\leq E[\|\xi_0\|^2]^{\frac{1}{2}} + 3KE\left[\int_0^T (1 + \|X(s)\|)^2 ds\right]^{\frac{1}{2}} \\ &\leq E[\|\xi_0\|^2]^{\frac{1}{2}} + 3KE\left[\int_0^T \|X(s)\|^2 ds\right]^{\frac{1}{2}} + 3KT^{\frac{1}{2}} \\ &\leq E[\|\xi_0\|^2]^{\frac{1}{2}} + 3KT^{\frac{1}{2}}(|X|_T + 1) < \infty \end{aligned}$$

**2nd step:**  $\Lambda$  fulfills (6.4.9), because

$$\begin{aligned} X, Y \in E &\implies \\ |\Lambda^{n+1}X - \Lambda^{n+1}Y|_t &= E\left[\sup_{0 \leq r \leq t} \left\| \int_0^r G(s, \Lambda^n X(s)) - G(s, \Lambda^n Y(s)) dW_s + \int_0^r b(s, \Lambda^n X(s)) - b(s, \Lambda^n Y(s)) ds \right\|^2\right]^{\frac{1}{2}} \\ &\leq 2E\left[\int_0^t \|G(s, \Lambda^n X(s)) - G(s, \Lambda^n Y(s))\|^2 ds\right]^{\frac{1}{2}} + E\left[\int_0^t \|b(s, \Lambda^n X(s)) - b(s, \Lambda^n Y(s))\|^2 ds\right]^{\frac{1}{2}} \\ &\leq 3KE\left[\int_0^t \|\Lambda^n X(s) - \Lambda^n Y(s)\|^2 ds\right]^{\frac{1}{2}} \\ &\leq 3K\left(\int_0^t |\Lambda^n X - \Lambda^n Y|_s^2 ds\right)^{\frac{1}{2}}. \end{aligned}$$

By iteration we get

$$\begin{aligned} |\Lambda^n X - \Lambda^n Y|_T &\leq 3^n K^n \left( \int_0^T \int_0^{t_1} \dots \int_0^{t_{n-1}} |X - Y|_{t_n}^2 dt_n \dots dt_1 \right)^{\frac{1}{2}} \\ &\leq 3^n K^n \left( \frac{T^n}{n!} \right)^{\frac{1}{2}} |X - Y|_T. \end{aligned} \tag{6.4.10}$$

$\implies$   $\exists!$  fixed-point  $X \in E$ , furthermore  
6.4.10

$$\begin{aligned} |X|_T &\leq \left( 1 + \sum_{n \geq 1} 3^n K^n \frac{T^{\frac{n}{2}}}{(n!)^{\frac{1}{2}}} \right) \cdot \left( E[\|\xi_0\|^2]^{\frac{1}{2}} + 3KT^{\frac{1}{2}} \right) \\ &\leq C(1 + E[\|\xi_0\|^2])^{\frac{1}{2}} \quad \text{for a constant } C = C(K, T). \end{aligned}$$

(where the first inequality holds because of the last line of Theorem 6.4.10 with  $y \equiv 0$  and step 1.)

## 6. Stochastic Differential Equations

**3rd step:** Let  $Y$  be an arbitrary continuous adapted process which fulfills (6.4.8), then  $Y = X$ .  
Let  $S_n := \inf\{t \geq 0 \mid \|X_t\| \geq n \text{ or } \|Y_t\| \geq n\}$  ( $\rightarrow +\infty$ ,  $n \rightarrow +\infty$ )

$$\begin{aligned} \implies X_{t \wedge S_n} - Y_{t \wedge S_n} &= \int_0^{t \wedge S_n} (G(s, X_s) - G(s, Y_s)) dW_s + \int_0^{t \wedge S_n} (b(s, X_s) - b(s, Y_s)) ds \\ \implies E [\|X_{t \wedge S_n} - Y_{t \wedge S_n}\|^2] &\leq 2E \left[ \int_0^{t \wedge S_n} \|G(s, X_s) - G(s, Y_s)\|^2 ds \right] + 2E \left[ \int_0^{t \wedge S_n} \|b(s, X_s) - b(s, Y_s)\|^2 ds \right] \\ &\leq 4KE \left[ \int_0^{t \wedge S_n} \|X_s - Y_s\|^2 ds \right] \\ &\leq 4K \int_0^t E [\|X_{s \wedge S_n} - Y_{s \wedge S_n}\|^2] ds \end{aligned} \quad (6.4.11)$$

$$\begin{aligned} \implies E [\|X_{t \wedge S_n} - Y_{t \wedge S_n}\|^2] &= 0 \quad \forall t \forall n & \implies X_t = Y_t \quad \forall t \leq S_n \text{ } P\text{-a.s. } \forall n, \\ \text{Gronwall's Lemma} & & \begin{array}{l} n \nearrow \infty \\ \text{paths} \\ \text{are continuous} \end{array} \end{aligned}$$

hence  $X = Y$ . □

**Lemma 6.4.11** (Gronwall). *Let  $f : [0, T] \rightarrow \mathbb{R}$  continuous. If there exist constants  $A \in \mathbb{R}$ ,  $b \in ]0, \infty[$  with*

$$f(t) \leq A + \int_0^t bf(s) ds, \quad t \in [0, T], \quad (6.4.12)$$

then  $f(t) \leq Ae^{bt}$ ,  $t \in [0, T]$ .

*Proof.* Without loss of generality  $A = 0$ , otherwise consider  $f(t) - Ae^{bt}$ , which fulfills (6.4.12) because

$$f(t) - Ae^{bt} \leq A \underbrace{(1 - e^{bt})}_{=-\int_0^t be^{bs} ds} + \int_0^t bf(s) ds = \int_0^t b(f(s) - Ae^{bs}) ds.$$

But if  $A = 0$ , we have due to (6.4.12) for  $g(t) := e^{-bt} \int_0^t f(s) ds$

$$g'(t) = -bg(t) + e^{-bt} f(t) \leq -bg(t) + bg(t) = 0, \quad g(0) = 0$$

$\implies g(t) \leq 0 \quad \forall t \in [0, T]$ . Hence by (6.4.12)  $f(t) \leq 0 \quad \forall t \in [0, T]$ . □

**Lemma 6.4.12** (generalized Gronwall). *Let  $f, b \in L^1([0, T])$ , such that  $b$  is nonnegative,  $f \cdot b \in L^1([0, T])$  and let  $A : [0, T] \rightarrow \mathbb{R}$  be an increasing function, such that*

$$f(t) \leq A(t) + \int_0^t b(s) f(s) ds \quad \text{for } dt - \text{a.e. } t \in [0, T]. \quad (6.4.13)$$

Then

$$\text{ess sup}_{s \in [0, t]} f(s) \leq A(t) e^{\int_0^t b(s) ds} \quad \text{for } dt - \text{a.e. } t \in [0, T].$$

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*Proof.*

**Step 1** Assume  $A$  is constant.

Then without loss of generality  $A = 0$ ; otherwise consider  $f - Ae^{\int_0^\bullet b(s)ds}$ , which fulfills (6.4.13) because for  $dt$ -a.e.  $t \in [0, T]$

$$\begin{aligned} f(t) - Ae^{\int_0^t b(s)ds} &\leq A \underbrace{\left(1 - e^{\int_0^t b(s)ds}\right)}_{- \int_0^t b(s)e^{\int_0^s b(r)dr} ds} + \int_0^t b(s)f(s)ds \\ &= \int_0^t b(s) \left(f(s) - Ae^{\int_0^s b(r)dr}\right) ds. \end{aligned}$$

But if  $A = 0$ , we have for  $g(t) := e^{-\int_0^t b(s)ds} \int_0^t b(s)f(s)ds$  and  $dt$ -a.e.  $t \in [0, T]$

$$\begin{aligned} g'(t) &= -b(t)g(t) + e^{-\int_0^t b(s)ds} \underbrace{b(t) f(t)}_{\leq \int_0^t b(s)f(s)ds} \\ &\leq 0. \end{aligned}$$

Since  $g(0) = 0$ , it follows that  $g \leq 0$   $dt$ -a.e., hence by (6.4.13) (with  $A = 0$ )  $f \leq 0$   $dt$ -a.e.

**Step 2**  $A$  increasing (not necessarily constant).

Let  $t \in [0, T]$ . Then by (6.4.13)

$$f(s) \leq A(t) + \int_0^s b(r)f(r)dr \text{ for } ds - \text{a.e. } s \in [0, t].$$

So, by Step 1, for  $ds$ -a.e.  $s \in [0, t]$

$$\begin{aligned} f(s) &\leq A(t)e^{\int_0^s b(r)dr} \\ &\leq A(t)e^{\int_0^t b(r)dr}. \end{aligned}$$

□

**Remark 6.4.13.** (i) For the proof of uniqueness (more precisely in (6.4.11)) we only used that  $G, b$  are locally Lipschitz, i.e. Lipschitz continuous on compact sets (e.g.  $\overline{B_n(0)}$ ). In other words: If  $G$  and  $b$  are locally Lipschitz, then the stochastic differential equation (6.4.8) has **at most one** strong solution.

(ii) If  $G \equiv 1$ , then (6.4.8) can be solved directly pathwise deterministic. (Use transformation  $x_t \rightarrow x_t - W_t$ , but “new”  $b$  depends explicitly on time and  $\omega$ ) If  $d = 1$ , then (6.4.8) can be reduced via Lamperti to the case  $G = 1$  if  $G, b$  are enough regular.

**Theorem 6.4.14.** Let  $b : \mathbb{R}^d \rightarrow \mathbb{R}^d$  be locally Lipschitz. Then for  $x_0 \in \mathbb{R}^d$

$$\exists! \text{ solution of } X(t) = x_0 + W(t) + \int_0^t b(X(s))ds$$

on  $[0, \xi[$ , where  $\xi =$  “explosion time”, i.e.  $\lim_{t \nearrow \xi(\omega)} |X(t)(\omega)| = +\infty$  if  $\xi(\omega) < +\infty$ . We have that  $\xi \equiv +\infty$  if  $b$  is globally Lipschitz.

(Without proof).

(Idea: transformation  $X_t \rightarrow X_t - W_t$ , then argue pathwise. Then  $b$  depends explicitly on time and  $\omega$ )



## 7. The martingale problem

Literature: [Str87]

### 7.1. Formulation, motivation

$\Omega := C([0, \infty); \mathbb{R}^N)$  (polish!),  $\mathcal{M} := \mathcal{B}_\Omega$ ,  $\mathcal{M}_t := \sigma(x(s) | 0 \leq s \leq t)$ ,  $t \geq 0$ ,  $x(s)(\omega) := \omega(s)$ ,

$M_1(\Omega) :=$  all probability-measures on  $(\Omega, \mathcal{M})$ .

$S^+(\mathbb{R}^N) :=$  all non-negative definite symmetric  $N \times N$  matrices.

Let  $a : [0, \infty[ \times \mathbb{R}^N \rightarrow S^+(\mathbb{R}^N)$ ,  $b : [0, \infty[ \times \mathbb{R}^N \rightarrow \mathbb{R}^N$  bounded, measurable. Define operator-valued map  $t \in [0, \infty[ \mapsto L_t$  by

$$L_t := \frac{1}{2} \sum_{i,j=1}^N a^{ij}(t, x) \partial_{x_i} \partial_{x_j} + \sum_{i=1}^N b^i(t, x) \partial_{x_i} \quad (7.1.1)$$

with  $\partial_{x_i} := \frac{\delta}{\delta x_i}$ ,  $1 \leq i \leq N$ .

**Definition 7.1.1.**  $P \in M_1(\Omega)$  is called solution of the martingale problem for  $(L_t)$  (with test function space  $C_0^\infty(\mathbb{R}^N)$ ) with start in  $(s, x) \in [0, \infty[ \times \mathbb{R}^N$ , denoted by:  $P \in M.P.((s, x); (L_t))$ , if

(i)  $P[x(0) = x] = 1$  (thus process  $(s + t, x(t))_{t \geq 0}$  starts in  $(s, \underbrace{x(0)}_{=x \text{ P-a.s.}})$ ),

(ii)  $\varphi(x(t)) - \int_0^t [L_{s+u}\varphi](x(u))du$ ,  $t \geq 0$ , is an  $(\mathcal{M}_t)$ -martingale under  $P \forall \varphi \in C_0^\infty(\mathbb{R}^N)$ .

For  $\varphi \in C^2(\mathbb{R}^N)$  set

$$X_{s,\varphi}(t) := \varphi(x(t)) - \int_0^t [L_{s+u}\varphi](x(u))du, \quad t \geq 0. \quad (7.1.2)$$

**Example 7.1.2.** Suppose there exists a probability space  $(\Omega, \mathcal{F}, P)$  with filtration  $(\mathcal{F}_t)$  and an  $(\mathcal{F}_t)$ -BM  $(W(t))_{t \geq 0}$  and an  $(\mathcal{F}_t)$ -adapted continuous process  $X = X(t)$ ,  $t \geq 0$ , such that for some  $x \in \mathbb{R}^N$

$$X(t) = x + \int_0^t \sigma(s+u, X(u))dW(u) + \int_0^t b(s+u, X(u))du, \quad t \geq 0, \quad P\text{-a.s..}$$

Let  $\varphi \in C_0^2(\mathbb{R}^N)$ . Then by the  $d$ -dimensional Itô-formula (see Proposition 1.3.21) with  $(\cdot, \cdot)_{\mathbb{R}^N} :=$  Euclidean inner product in  $\mathbb{R}^N$  and  $X = (X^1, \dots, X^N)$  for  $t \geq 0$

$$\begin{aligned} \varphi(X(t)) &= \varphi(x) + \int_0^t (\nabla \varphi(X(u)), dX(u))_{\mathbb{R}^N} \\ &\quad + \frac{1}{2} \int_0^t \sum_{i,j=1}^N \frac{\partial^2 \varphi}{\partial x_i \partial x_j}(X(u)) d\langle X^i, X^j \rangle_u. \end{aligned}$$

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But since

$$\begin{aligned} \langle X^i, X^j \rangle_t &= \left\langle \sum_{\ell=1}^N \int_0^\bullet \sigma^{i\ell}(s+u, X(u)) dW^\ell(u), \sum_{\ell'=1}^N \int_0^\bullet \sigma^{j\ell'}(s+u, X(u)) dW^{\ell'}(u) \right\rangle_t \\ &= \int_0^t \sum_{\ell, \ell'=1}^N \sigma^{i\ell}(s+u, X(u)) \sigma^{j\ell'}(s+u, X(u)) d \underbrace{\langle W^\ell, W^{\ell'} \rangle_u}_{\delta_{\ell, \ell'} du}, \end{aligned}$$

we obtain for  $(a^{ij}) := \sigma \sigma^T$  and  $t \geq 0$

$$\begin{aligned} \varphi(X(t)) &= \varphi(x) + \int_0^t (\nabla \varphi(X(u)), \sigma(s+u, X(u)) dW(u))_{\mathbb{R}^N} \\ &\quad + \int_0^t (\nabla \varphi(X(u)), b(s+u, X(u)))_{\mathbb{R}^N} du \\ &\quad + \frac{1}{2} \int_0^t \sum_{i,j=1}^N a^{ij}(s+u, X(u)) \frac{\partial^2 \varphi}{\partial x_i \partial x_j}(X(u)) du \\ &= \int_0^t [L_{s+u} \varphi](X(u)) du \\ &\quad + \underbrace{\varphi(x) + \int_0^t (\nabla \varphi(X(u)), \sigma(s+u, X(u)) dW(u))_{\mathbb{R}^N}}_{a \text{ martingale!}} \end{aligned}$$

Hence  $P \circ X^{-1} \in M.P.((s, x), (L_t))$ .

Let  $P \in M.P.((s, x); (L_t))$  be fixed and  $\varphi \in C_0^\infty(\mathbb{R}^N)$ . Then  $X_{s,\varphi} \in \text{Mart}_c^2(P)$  ( $:=$  all  $(\mathcal{M}_t)$ -martingales under  $P$  in  $L^2$  with P-a.s. continuous paths).

**Lemma 7.1.3.**  $\forall \varphi, \psi \in C_0^\infty(\mathbb{R}^N)$

$$\langle X_{s,\varphi}, X_{s,\psi} \rangle_t = \int_0^t (\nabla \varphi, a \nabla \psi)_{\mathbb{R}^N}(s+u, x(u)) du, \quad t \geq 0, \quad P - a.s. \quad (7.1.3)$$

*Proof.* By polarization without loss of generality :  $\varphi = \psi$ . We have

$$\begin{aligned}
 X_{s,\varphi}(t)^2 & \stackrel{(7.1.2)}{=} \varphi(x(t))^2 - 2\varphi(x(t)) \int_0^t [L_{s+u}\varphi](x(u))du + \left[ \int_0^t [L_{s+u}\varphi](x(u))du \right]^2 \\
 & \stackrel{(7.1.2)}{=} \varphi^2(x(t)) - 2X_{s,\varphi}(t) \int_0^t [L_{s+u}\varphi](x(u))du - \left[ \int_0^t [L_{s+u}\varphi](x(u))du \right]^2 \\
 & \stackrel{(7.1.2)+}{=} \underbrace{X_{s,\varphi^2}(t)}_{\text{mart.!\}} + \int_0^t [L_{s+u}\varphi^2](x(u))du - \left[ \int_0^t [L_{s+u}\varphi](x(u))du \right]^2 \\
 & \quad \text{It\^o's product formula} \\
 & \quad - 2 \underbrace{\int_0^t \int_0^u [L_{s+u'}\varphi](x(u'))du' dX_{s,\varphi}(u)}_{\text{mart.}} \\
 & \quad - 2 \int_0^t \underbrace{\left( \underbrace{X_{s,\varphi}(u)}_{\stackrel{(7.1.2)}{=} \varphi(x(u))} - \int_0^u [L_{s+u'}\varphi](x(u'))du' \right)}_{\substack{= 2 \int_0^t [\varphi L_{s+u}\varphi](x(u))du - [\int_0^t [L_{s+u}\varphi](x(u))du]^2 \\ \text{(It\^o's) product formula}}} d \left( \int_0^\bullet [L_{s+u'}\varphi](x(u'))du' \right) (u) \\
 & = \int_0^t \underbrace{[L_{s+u}\varphi^2 - 2\varphi L_{s+u}\varphi]}_{\stackrel{(*)}{=} \sum a^{ij}(s+u, \cdot) \partial_{x_i}\varphi \partial_{x_j}\varphi} (x(u))du + \text{martingale} \\
 & = \int_0^t (\nabla\varphi, a\nabla\varphi)(s+u, x(u))du + \text{martingale}
 \end{aligned}$$

where (\*) holds because  $\partial_{x_i}\partial_{x_j}\varphi^2 = 2\partial_{x_i}(\varphi\partial_{x_j}\varphi) = 2\varphi\partial_{x_i}\partial_{x_j}\varphi + 2(\partial_{x_i}\varphi)(\partial_{x_j}\varphi)$  and  $\partial_{x_i}\varphi^2 = 2\varphi\partial_{x_i}\varphi$ .  $\square$

**Lemma 7.1.4.** (i)  $\forall \varphi \in C^2(\mathbb{R}^N)$ ,  $X_{s,\varphi} \in \text{Mart}_c^{\text{loc}}(P)$  ( $:=$  all continuous loc.  $(\mathcal{M}_t)$ -martingales until  $+\infty$ ) under  $P$  with  $P$ -a.s. continuous paths), and (7.1.3) holds  $\forall \varphi, \psi \in C^2(\mathbb{R}^N)$ .

(ii) If  $\bar{x}(t) := x(t) - \int_0^t b(s+u, x(u))du$ ,  $t \geq 0$ , then  $\bar{x}_i \in \text{Mart}_c^2(P)$ ,  $1 \leq i \leq N$  (even  $\bar{x}_i \in L^q(P) \forall q \in ]0, \infty[$ ) and

$$(\langle \bar{x}_i, \bar{x}_j \rangle(t))_{1 \leq i, j \leq N} = \int_0^t a(s+u, x(u))du, \quad t \geq 0 \quad P - \text{a.s.} \quad (7.1.4)$$

(iii) If  $f \in C^{1,2}([0, \infty[ \times \mathbb{R}^N)$  (i.e.  $C^1$  in  $t$ ,  $C^2$  in  $x$ ), then

$$\begin{aligned}
 & f(t, x(t)) - f(0, x) - \int_0^t [(\partial_u + L_{s+u})f](u, x(u))du \\
 & = \underbrace{\sum_{i=1}^N \int_0^t \partial_{x_i} f(u, x(u)) d\bar{x}_i(u)}_{\text{loc. mart.}} =: \int_0^t [\nabla_x f](u, x(u)) \cdot d\bar{x}(u), \quad t \geq 0, \quad P - \text{a.s.}
 \end{aligned}$$

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*Proof.* (i) : Clear (with trivial approximation)  $X_{s,\varphi} \in \text{Mart}_c^2(P) \forall \varphi \in C_0^2(\mathbb{R}^N)$ . Now let  $\varphi \in C^2(\mathbb{R}^N)$  and  $n \in \mathbb{N}$ . Define

$$\sigma_n := \inf\{t \geq 0 \mid |x(t)| \geq n\},$$

and take  $\eta_n \in C_0^\infty(\mathbb{R}^N)$  with  $\eta_n = 1$  on  $B_{n+1}(0)$ . Then  $\sigma_n \uparrow \infty$  and  $X_{s,\varphi}(\cdot \wedge \sigma_n) = X_{s,\eta_n \cdot \varphi}(\cdot \wedge \sigma_n) \in \text{Mart}_c^2(P)$ , hence  $X_{s,\varphi} \in \text{Mart}_c^{\text{loc}}(P)$  and (7.1.3) is still true.

(ii) : (i) with  $\varphi(x) = x_i$ , implies that  $\bar{x}_i \in \text{Mart}_c^{\text{loc}}(P)$ ,  $1 \leq i \leq N$ , and (7.1.3) holds with  $\psi(x) = x_j$  which implies (7.1.4). By Cor. 1.4.32 it follows that  $\bar{x}_i$  is a square integrable martingale  $\forall 1 \leq i \leq N$ , since by (7.1.4)  $\langle \bar{x}_i \rangle_t$  is bounded for every  $t \geq 0$ .  
Now we need

**Lemma 7.1.5.** *Let  $X \in \text{Mart}_c^{\text{loc}}(P)$  and  $\sigma \leq \tau$  stopping times with  $\langle X \rangle(\tau) - \langle X \rangle(\sigma) \leq A$  for some  $A \in ]0, \infty[$ . Then*

$$P\left[\sup_{\sigma \leq t \leq \tau} |X(t) - X(\sigma)| \geq R\right] \leq 2 \exp(-R^2/2A) \quad \forall R \geq 0.$$

In particular  $\forall q \in ]0, \infty[ \exists C_q \in [1, \infty[$  with

$$E_p \left[ \sup_{\sigma \leq t \leq \tau} |X(t) - X(\sigma)|^q \right] \leq C_q A^{\frac{q}{2}}.$$

*Proof.* [Str87, p.59, Lemma (3.10)] □

Continuing with proof of 7.1.4 (ii): 7.1.5  $\implies \forall 0 \leq s < T$

$$P \left[ \sup_{s \leq t \leq T} \|\bar{x}(t) - \bar{x}(s)\| \geq R \right] \leq 2 \cdot N \exp \left[ -\frac{R^2}{2AN(T-s)} \right] \quad (7.1.5)$$

with  $A := \sup_{t,y} \|a(t,y)\|_{\text{oper. norm}}$ . In particular  $\forall q \in ]0, \infty[ \exists C(q, N) \in [1, \infty[$  with

$$\left| \sup_{s \leq t \leq T} \|\bar{x}(t) - \bar{x}(s)\| \right|_{L^q(P)} \leq C(q, N)(A(T-s))^{\frac{1}{2}} \quad (7.1.6)$$

In particular  $\bar{x}_i \in L^q(P) \forall 1 \leq i \leq N$ .

(iii) Follows directly with (ii) from Itô's formula (for the time-dependent case). □

**Remark 7.1.6.**  $a \equiv 0 \xrightarrow[\substack{(7.1.4) \\ \& 2.2.8(iii)}]{\implies} \bar{x}(t) = \bar{x}(0) = x(0) = x$ . Thus

$$x(t) = x + \int_0^t b(s+u, x(u)) du, \quad t \geq 0 \quad P - a.s.$$

Hence ( $P$ -a.s.) deterministic case included.

**Theorem 7.1.7.** *Let  $a : [0, \infty[ \times \mathbb{R}^N \rightarrow S^+(\mathbb{R}^N)$ ,  $b : [0, \infty[ \times \mathbb{R}^N \rightarrow \mathbb{R}^N$  bounded, measurable and  $P \in M.P.((s, x), L_t)$  for an  $(s, x) \in [0, \infty[ \times \mathbb{R}^N$ . Supposing  $\exists \sigma : [0, \infty[ \times \mathbb{R}^N \rightarrow \text{Hom}(\mathbb{R}^d; \mathbb{R}^N)$  ( $= d \times N$  matrices over  $\mathbb{R}$ ) bounded, measurable with  $a = \sigma \sigma^T$ .*

Then  $\exists$   $d$ -dim. BM  $((\beta(t))_{t \geq 0}, (\mathcal{F}_t)_{t \geq 0}, Q)$  on a probability space  $(\tilde{\Omega}, \mathcal{F}, Q)$  and continuous,  $(\mathcal{F}_t)_{t \geq 0}$ -progressiv measurable map  $X : [0, \infty[ \times \tilde{\Omega} \rightarrow \mathbb{R}^N$ , such that

$$X(t) = x + \int_0^t \sigma(s+u, X(u)) d\beta(u) + \int_0^t b(s+u, X(u)) du, \quad t \geq 0, \quad Q - a.s.$$

and

$$P = Q \circ X^{-1}.$$

(i.e. briefly:  $dX(t) = \sigma(s+t, X(t))d\beta(t) + b(s+t, X(t))dt$ ,  $X(0) = x$ ). If particularly  $a(t, y) > 0$   $\forall (t, y) \in [0, \infty[ \times \mathbb{R}^N$  and  $\sigma := a^{\frac{1}{2}}$  (as pos. s.a. operator uniquely determined!), then one can take:  $\tilde{\Omega} := \Omega$ ,  $Q := P$ ,  $X(\cdot) = x(\cdot)$  and

$$\beta(t) := \int_0^t \sigma^{-1}(s+u, x(u)) d\bar{x}(u), \quad t \geq 0, \quad \mathcal{F}_t := \mathcal{M}_t, \quad \mathcal{F} := \mathcal{M},$$

with  $\bar{x}(t) := x(t) - \int_0^t b(s+u, x(u)) du$ ,  $t \geq 0$ . (martingale according to 7.1.4(ii)!)

*Proof.* [Str87, Theorem(2.6), p.91 ff.] □

**Exercise 7.1.8:** (i) Show:  $P \in M.P.((s, x), (L_t)) \Leftrightarrow P[x(0) = x] = 1$  and  $\bar{x}_i \in Mart_c^{loc}(P)$   $\forall 1 \leq i \leq N$  and (7.1.4) holds.

(ii) Show: What we have done so far, can be generalized to the following situation: Let  $a : [0, \infty[ \times \mathbb{R}^N \times \Omega \rightarrow S^+(\mathbb{R}^N)$ ,  $b : [0, \infty[ \times \mathbb{R}^N \times \Omega \rightarrow \mathbb{R}^N$  bounded,  $(\mathcal{M}_t)$ -progressiv measurable and  $(L_t)$  as in (7.1.1) with these  $a, b$ . Define  $M.P.(x; (L_t))$  as the set of all  $P \in M_1(\Omega)$  with  $P[x(0) = x] = 1$  and

$$\varphi(x(t)) - \int_0^t [L_u \varphi](x(u)) du, \quad t \geq 0,$$

is an  $(\mathcal{M}_t)$ -martingale under  $P \forall \varphi \in C_0^\infty(\mathbb{R}^N)$ .

Define  $\bar{x}(t) := x(t) - \int_0^t b(u) du$ . Show:

(a)  $P \in M.P.(x; (L_t)) \Leftrightarrow P[x(0) = x] = 1$  and  $\bar{x}_i \in Mart_c^{loc}(P) \forall 1 \leq i \leq N$  and  $(\langle \bar{x}_i, \bar{x}_j \rangle)_{i,j} = \int_0^\bullet a(u) du$   $P$ -a.s.

(b)  $P \in M.P.(x; (L_t)) \implies$  7.1.4(iii), (7.1.5), (7.1.6) hold.

(iii) If  $a, b$  in (7.1.1) are independent of  $t$ , then the martingale problem in 7.1.1 is time independent. Every time dependent martingale problem becomes time independent via the following trick (“space time homogenization”): Set  $\tilde{\Omega} := C([0, \infty[; \mathbb{R}^{N+1}) \equiv C([0, \infty); \mathbb{R}^1) \times \Omega$ . Show:  $P \in M.P.((s, x), (L_t)) \Leftrightarrow \tilde{P} \in M.P.(\tilde{x}, \tilde{L})$ , where  $\tilde{x} := (s, x) (\in [0, \infty) \times \mathbb{R}^N)$ ,  $\tilde{L} := \partial_t + L_t$ ,  $\tilde{P} := \sigma_{s+} \otimes P$  and  $\sigma_{s+} \in M_1(C([0, \infty[; \mathbb{R}^1))$  is Dirac measure on path  $t \mapsto s+t$ .

**Definition 7.1.9.** The martingale problem for  $(L_t)$  in (7.1.1) (with test function space  $C_0^\infty(\mathbb{R}^N)$ ) is called “well-posed”, if  $\forall (s, x) \in [0, \infty[ \times \mathbb{R}^N \# M.P.((s, x), (L_t)) = 1$ .

**Theorem 7.1.10.** Suppose the martingale problem for  $(L_t)$  in (7.1.1) (with  $C_0^\infty(\mathbb{R}^N)$ ) is “well-posed”. Let  $\{P_{s,x} | (s, x) \in [0, \infty[ \times \mathbb{R}^N\}$  be the corresponding family of solutions. Then:

(i)  $(s, x) \mapsto P_{s,x}$  is measurable from  $[0, \infty[ \times \mathbb{R}^N$  to  $M_1(\Omega)$  (with weak topology).

(ii) Suppose we have

$$x \mapsto \int_0^T a(t, x) dt, \quad x \mapsto \int_0^T b(t, x) dt \text{ continuous } \forall T > 0. \quad (7.1.7)$$

Then  $(s, x) \mapsto P_{s,x}$  continuous from  $[0, \infty[ \times \mathbb{R}^N$  to  $M_1(\Omega)$ .

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*Proof.* [Str87, p.85 ff.] □

**Theorem 7.1.11** (Existence). *Let  $(L_t)$  be as in (7.1.1), and (7.1.7) holds. Then*

$$\#M.P.((s, x), (L_t)) \geq 1 \quad \forall (s, x) \in [0, \infty[ \times \mathbb{R}^N.$$

*Proof.* [Str87, p. 84 (1.20) & (1.21)] □

**Theorem 7.1.12** (Uniqueness). *Let  $(L_t)$  be as in (7.1.1). Suppose the following condition holds:  $\forall (s, x) \in [0, \infty[ \times \mathbb{R}^N$  we have*

$$P, Q \in M.P.((s, x); (L_t)) \implies P \circ x(t)^{-1} = Q \circ s(t)^{-1} \quad \forall t \geq 0. \quad (7.1.8)$$

*Then*

$$\#M.P.((s, x); (L_t)) \leq 1 \quad \forall (s, x) \in [0, \infty[ \times \mathbb{R}^N.$$

7.1.11 is useful for applications, 7.1.12 not. To get a sufficient condition for (7.1.8) we need some preparations:

**Definition 7.1.13.** *Let  $\mathcal{F}$  be a set of bounded, measurable functions on a measurable space  $(E, \mathcal{B})$ .  $\mathcal{F}$  is called determining set, if  $\forall \mu, \nu \in M_1(E)$*

$$\int \varphi d\mu = \int \varphi d\nu \quad \forall \varphi \in \mathcal{F} \implies \mu = \nu.$$

**Theorem 7.1.14.** *Suppose there exists a determining set  $\mathcal{F} \subset C_b(\mathbb{R}^N)$ , such that one of the following conditions holds:*

(i)  $\forall T > 0, \varphi \in \mathcal{F} \exists u_{T,\varphi} \in C_b^{1,2}([0, T[ \times \mathbb{R}^N)$  with

$$(\partial_t + L_t)u_{T,\varphi} = 0 \text{ on } [0, T[ \times \mathbb{R}^N$$

and

$$\lim_{t \uparrow T} u_{T,\varphi}(t, x) = \varphi(x) \quad \forall x \in \mathbb{R}^N$$

(automatically uniform in  $x$  because of (\*)).

(ii)  $\forall T > 0, \varphi \in \mathcal{F} \exists \tilde{u}_{T,\varphi} \in C_b^{1,2}([0, T[ \times \mathbb{R}^N)$  with

$$(\partial_t + L_t)\tilde{u}_{T,\varphi} = \varphi \text{ on } [0, T[ \times \mathbb{R}^N$$

and

$$\lim_{t \uparrow T} \tilde{u}_{T,\varphi}(t, x) = 0 \quad \forall x \in \mathbb{R}^N$$

(automatically uniform in  $x$ ).

*Then  $\#M.P.((s, x); (L_t)) \leq 1 \quad \forall (s, x) \in [0, \infty[ \times \mathbb{R}^N$ .*

*More precisely, in case (i):  $\forall (s, x) \in [0, \infty[ \times \mathbb{R}^N$  and  $\forall P \in M.P.((s, x); (L_t))$*

$$E_P[\varphi(x(T-s))] = u_{T,\varphi}(s, x) \quad \forall T > s \quad (7.1.9)$$

*Thus (7.1.8) is fulfilled.*

*More precisely, in case (ii):  $\forall (s, x) \in [0, \infty[ \times \mathbb{R}^N$  and  $\forall P \in M.P.((s, x); (L_t))$*

$$E_P \left[ \int_0^{T-s} \varphi(x(u)) du \right] = -\tilde{u}_{T,\varphi}(s, x) \quad \forall T > s \quad (7.1.10)$$

*Thus (7.1.8) is fulfilled.*

*Proof.* Suppose (i) holds: Let  $T > s$ . Then according to 7.1.4(iii) with  $f := u_{T,\varphi}(\cdot + s, \cdot)$   $\forall t \in [0, T - s[$

$$E_P [u_{T,\varphi}(t + s, x(t))] = u_{T,\varphi}(s, x).$$

With  $t \uparrow T - s$  (7.1.9) follows, hence (7.1.8) holds and the assertion is implied by 7.1.12. Suppose (ii) holds: Let  $T > s$ . Then according to 7.1.4(iii) with  $f := \tilde{u}_{T,\varphi}(\cdot + s, \cdot)$   $\forall t \in [0, T - s[$

$$E_P [\tilde{u}_{T,\varphi}(t + s, x(t))] - \tilde{u}_{T,\varphi}(s, x) = E_P \left[ \int_0^t \varphi(x(u)) du \right].$$

With  $t \uparrow T - s$  (7.1.10) follows, thus  $\forall a, b \in \mathbb{R}_+, a < b$

$$\int_a^b E_P[\varphi(x(u))du] = -\tilde{u}_{b+s,\varphi}(s, x) + \tilde{u}_{a+s,\varphi}(s, x)$$

Hence by continuity  $u \mapsto E_P[\varphi(x(u))]$  is uniquely determined on  $\mathbb{R}_+$ . Thus (7.1.8) holds and 7.1.12 implies the assertion.  $\square$

Finally we want to prove 7.1.12. Therefore we need:

**Theorem 7.1.15.** *Let  $P \in M.P.((s, x), (L_t))$  and  $\tau$  a stopping time. Then:*

(i)  $\mathcal{M}_\tau = \sigma(\{x(t \wedge \tau) | t \geq 0\})$ . *In particular  $\mathcal{M}_\tau$  is countably generated.*

(ii) *Let  $\mathcal{A}$  be a countably generated sub- $\sigma$ -algebra of  $\mathcal{M}_\tau$ , such that*

$$\omega \mapsto x(\tau(\omega), \omega) \quad \mathcal{A} - \text{measurable.}$$

*Let  $\omega \mapsto P_\omega$  be a regular conditional probability distribution of  $P$  given  $\mathcal{A}$  (i.e.:  $(\omega, A) \mapsto P_\omega[A], \omega \in \Omega, A \in \mathcal{M}$ , is stochastic kernel on  $(\Omega, \mathcal{M})$  such that:*

a)  $\forall A \in \mathcal{M} : P[A|\mathcal{A}](\omega) = P_\omega[A]$  for  $P$ -a.a.  $\omega \in \Omega$ .

b)  $\forall A \in \mathcal{M} : \omega \mapsto P_\omega(A)$  is  $\mathcal{A}$ -measurable and ("regular")

$$P_\omega[A] = 1_A(\omega) \text{ for all } \omega \in \Omega, \text{ if } A \in \mathcal{A}.$$

*(exists, since  $\mathcal{A}$  is countably generated) i.e.:  $P_{\omega|\mathcal{A}} = \epsilon_\omega$ ).*

*Then  $\exists P$ -null set  $\Lambda \in \mathcal{A}$  such that:  $\forall \omega \in \Lambda^c$*

$$P_\omega \circ \theta_{\tau(\omega)}^{-1} \in M.P.((s + \tau(\omega), x(\tau(\omega), \omega)); (L_t))$$

*(where for  $t \geq 0 : \theta_t : \Omega \rightarrow \Omega, \theta_t(\omega) = \omega(\cdot + t)$ .)*

*Proof.* cf. [Str87, III.(1.13), p.78, and III.(1.14), p. 79].  $\square$

*Proof of 7.1.12.* Let  $P, Q \in M.P.((s, x), (L_t))$ .

To show: (\*)  $P \circ (x(t_1), \dots, x(t_n))^{-1} = Q \circ (x(t_1), \dots, x(t_n))^{-1} \forall 0 \leq t_1 < \dots < t_n$  and **all**  $n \in \mathbb{N}$ .

Induction over  $n$ :

$n = 1$ : Ok, according to assumption.

$n \mapsto n + 1$ : Suppose (\*) holds for  $n \in \mathbb{N}$  and all  $0 \leq t_1 < \dots < t_n$ . Let  $0 \leq t_1 < \dots < t_n < t_{n+1}$ .

Let  $\mathcal{A} =: \sigma(\{x(t_1), \dots, x(t_n)\})$  and  $\omega \mapsto P_\omega$  resp.  $\omega \mapsto Q_\omega$  regular conditional probability distributions of  $P$  resp.  $Q$  given  $\mathcal{A}$ .

According to 7.1.14(ii)  $\exists$  a  $P$ -null set  $\Lambda_1 \in \mathcal{A}$  resp. a  $Q$ -null set  $\Lambda_2 \in \mathcal{A}$  such that

$$P_\omega \circ \theta_{t_n}^{-1}, Q_\omega \circ \theta_{t_n}^{-1} \in M.P.((s + t_n, x(t_n, \omega)); (L_t))$$

$\forall \omega \in \Lambda_1^c \cap \Lambda_2^c$ .

## 7. The martingale problem

Hence for  $\Lambda := \Lambda_1 \cup \Lambda_2$  we have, according to assumption (7.1.8),  $\forall \omega \in \Lambda^c$

$$\begin{aligned}
 P_\omega \circ x(t_{n+1})^{-1} &= (P_\omega \circ \theta_{t_n}^{-1}) \circ x(t_{n+1} - t_n)^{-1} \\
 &\stackrel{(7.1.8)}{=} (Q_\omega \circ \theta_{t_n}^{-1}) \circ x(t_{n+1} - t_n)^{-1} \\
 &= Q_\omega \circ x(t_{n+1})^{-1}. \tag{**}
 \end{aligned}$$

In addition, due to the induction hypothesis  $P = Q$  on  $\mathcal{A}$ , hence  $P[\Lambda] = Q[\Lambda] = 0$ , and thus  $\forall B_1, \dots, B_{n+1} \in \mathcal{B}(\mathbb{R}^N)$

$$\begin{aligned}
 &P[x(t_1) \in B_1, \dots, x(t_{n+1}) \in B_{n+1}] \\
 &= E_P[\underbrace{1_{B_1}(x(t_1)) \dots 1_{B_n}(x(t_n))}_{\in \mathcal{A}} \underbrace{E_P[1_{B_{n+1}}(x(t_{n+1}))]}_{\mathcal{A}\text{-measurable}})] \\
 &= E_Q[1_{\Lambda^c} 1_{B_1}(x(t_1)) \dots 1_{B_n}(x(t_n)) \underbrace{E_P[1_{B_{n+1}}(x(t_{n+1}))]}_{\stackrel{(**)}{=} E_Q[1_{B_{n+1}}(x(t_{n+1}))}})] \\
 &\stackrel{\text{analogously}}{=} Q[x(t_1) \in B_1, \dots, x(t_{n+1}) \in B_{n+1}].
 \end{aligned}$$

□

## A. Time Transformation

Let  $(\Omega, \mathcal{F}, P)$  be a probability space,  $(\mathcal{F}_t)_{t \geq 0}$  a right-continuous, complete filtration and  $Y$  a  $P$ -a.s. continuous locale martingale up to  $\infty$  (with respect to  $(\mathcal{F}_t)$ ) such that  $Y_0 \equiv 0$ . For simplicity let

$$\langle Y \rangle_\infty (:= \lim_{t \rightarrow \infty} \langle Y \rangle_t) = \infty \quad P\text{-a.s.}$$

Define the "inverse" of  $\langle Y \rangle_t$  by

$$C_t := \inf \underbrace{\{s > 0 \mid \langle Y \rangle_s > t\}}_{\neq \emptyset}.$$

**Theorem A.0.1.** *Let  $\langle Y \rangle_\infty = \infty$   $P$ -a.s. and  $W_t := Y_{C_t}$ ,  $t \geq 0$ . Then,  $W$  is a  $(\mathcal{G}_t)_{t \geq 0} := \mathcal{F}_{C_t}$ -Brownian motion (cf. 1.5.34) and  $Y_t = W_{\langle Y \rangle_t}$  for all  $t \geq 0$ .*

For the proof we need the following two lemmas.

**Lemma A.0.2.** (i) *The map  $t \mapsto C_t$  is increasing and right-continuous.*

(ii)  *$C_t$  is an  $(\mathcal{F}_s)$ -stopping time for all  $t$ .*

(iii)

$$\langle Y \rangle_{C_t} = t \quad P\text{-a.s. } \forall t.$$

(iv)  *$t \leq C_{\langle Y \rangle_t}$ . (Note that in general case  $t \neq C_{\langle Y \rangle_t}$ !)*

*Proof.* (iii): Without loss of generality  $\langle Y \rangle$  is continuous everywhere. Then  $\langle Y \rangle_{C_t} \geq t$ , since  $t \mapsto \langle Y \rangle_t$  is (right-)continuous.

Assumption:  $\langle Y \rangle_{C_t} \geq t + \varepsilon$  for  $\varepsilon > 0$ .

Then, since  $t \mapsto \langle Y \rangle_t$  is continuous, there exists a  $\delta > 0$  such that  $\langle Y \rangle_{C_t - \delta} \geq t + \frac{\varepsilon}{2} > t$ . Hence,  $C_t \leq C_t - \delta$ . Since this is impossible it follows that  $\langle Y \rangle_{C_t} < t + \varepsilon$  for all  $\varepsilon > 0$ .

(i): Obviously  $t \mapsto C_t$  is increasing. It remains to show that

$$\lim_{u \searrow t} C_u \leq C_t.$$

(" $\geq$ " is clear since  $C_k$  is increasing.)

Let  $\varepsilon > 0$ . Then  $\langle Y \rangle_{C_t + \varepsilon} > t$ . Hence, there exists a  $\delta > 0$  such that

$$\langle Y \rangle_{C_t + \varepsilon} > u \quad \forall u \in [t, t + \delta].$$

Thus,

$$C_u \leq C_t + \varepsilon \quad \forall u \in [t, t + \delta].$$

(ii): We have

$$\{C_t < u\} = \{\langle Y \rangle_u > t\} \quad \forall u, t, \tag{A.0.1}$$

because " $\subset$ " is clear and, if  $\langle Y \rangle_u > t$ , then there exists an  $\varepsilon > 0$  such that  $\langle Y \rangle_{u - \varepsilon} > t$  and, therefore,  $C_t \leq u - \varepsilon < u$ . But

$$\{\langle Y \rangle_u > t\} \in \mathcal{F}_u.$$

Hence, by (A.0.1)

$$\{C_t \leq u\} \in \mathcal{F}_{u+} = \mathcal{F}_u.$$

(iv): If  $C_{\langle Y \rangle_t} < t$ , then by (A.0.1) we would get  $\langle Y \rangle_{C_{\langle Y \rangle_t}} > \langle Y \rangle_t$ . Therefore,  $C_{\langle Y \rangle_t} \geq t$ .  $\square$

### A. Time Transformation

Note that  $\mathcal{G}_t := \mathcal{F}_{C_t}$ ,  $t \geq 0$  (filtration, since  $C_t$  is increasing!).

**Lemma A.0.3.** (i)  $(\mathcal{G}_t)_{t \geq 0}$  is right-continuous and complete.

(ii)  $\langle Y \rangle_t$  is a stopping time with respect to  $(\mathcal{G}_s)_{s \geq 0}$  for all  $t$ .

*Proof.* (ii) is clear by (A.0.1).

(i): Right-continuity: Let  $A \in \bigcap_{\varepsilon > 0} \mathcal{F}_{C_{t+\varepsilon}}$  (in particular,  $A \in \mathcal{F}_{C_{t+\frac{1}{n}}} \forall n$ ). Hence, for all  $s$

$$A \cap \{C_{t+\frac{1}{n}} < s\} \in \mathcal{F}_s \quad \forall n, s,$$

therefore,

$$A \cap \underbrace{\bigcup_n \{C_{t+\frac{1}{n}} < s\}}_{=\{C_t < s\}} \in \mathcal{F}_s \quad \forall s,$$

thus  $A \cap \{C_t \leq s\} \in \mathcal{F}_{s+} = \mathcal{F}_s \forall s$ , and  $A \in \mathcal{F}_{C_t}$ .

Completeness: Let  $A_0 \in \mathcal{F}_{C_t}$  with  $P[A_0] = 0$  and  $A \subset A_0$ . Then for all  $s$

$$A \cap \{C_t \leq s\} \in \mathcal{F}_s$$

as a subset of an  $\mathcal{F}_s$ -measurable  $P$ -zero set  $A_0 \cap \{C_t \leq s\}$ , since  $\mathcal{F}_s$  is complete. Hence,  $A \in \mathcal{F}_{C_t}$ .  $\square$

*Proof of A.0.1.* Since  $Y$  is  $P$ -a.s. continuous and  $(C_t)$  is right-continuous,  $(W_t)$  is  $P$ -a.s. right-continuous.

**Step 1.**  $t \mapsto W_t = Y_{C_t}$  is  $P$ -a.s. continuous:

It suffice to show that (cf. Exercises)

$$P[Y_u = Y_t \quad \text{for } t \leq u \leq \sigma_t \quad \forall t \geq 0] = 1, \quad (\text{A.0.2})$$

where

$$\sigma_t := \inf\{s > t \mid \langle Y \rangle_s > \langle Y \rangle_t\},$$

hence,

$$\langle Y \rangle_u = \langle Y \rangle_t \quad \text{for } u \in [t, \sigma_t].$$

(A.0.2) is sufficient, since then  $Y$  is constant on the interval, where  $\langle Y \rangle$  is constant and by definition this is the case on  $[C_{t-}, C_t]$ . Therefore,  $Y_{C_t} = Y_{C_{t-}}$ . For (A.0.2) it remains to show that for all  $r \in \mathbb{Q}^+$

$$P[Y_u = Y_r \quad \text{for } r \leq u \leq \sigma_r \quad \forall r \in \mathbb{Q}^+] = 1, \quad (\text{A.0.3})$$

because, if  $t \geq 0$  with  $t < \sigma_t$ , then for all  $r \in [t, \sigma_t] \cap \mathbb{Q}^+$   $\sigma_r = \sigma_t$  and (A.0.3) implies (A.0.2), since  $Y$  is  $P$ -a.s. (right-)continuous.

Let  $(T'_n)_{n \in \mathbb{N}}$  be a localizing sequence for  $Y$ . Then

$$T_n := \inf\{t > 0 \mid |Y_t| > n\} \wedge T'_n, \quad n \in \mathbb{N},$$

is again a localizing sequence. Fix  $r \in \mathbb{Q}^+$ . For  $n \in \mathbb{N}$  set

$$N_t^{(n)} := Y_{(r+t) \wedge \sigma_r \wedge T_n} - Y_{r \wedge T_n}, \quad t \geq 0,$$

$$\tilde{\mathcal{F}}_t := \mathcal{F}_{t+r}, \quad t \geq 0.$$

Then  $N^{(n)}$  is a continuous bounded martingale with respect to  $(\tilde{\mathcal{F}}_t)$ , since by the stopping theorem  $\forall s \leq t$

$$E[N_t^{(n)} \mid \tilde{\mathcal{F}}_s] = E[Y_{(r+t) \wedge \sigma_r \wedge T_n} - Y_{r \wedge T_n} \mid \mathcal{F}_{r+s}] = Y_{(r+s) \wedge \sigma_r \wedge T_n} - Y_{r \wedge T_n} = N_s^{(n)},$$

since  $\sigma_r$  is a stopping time with respect to  $(\tilde{\mathcal{F}}_t)$ . Additionally,

$$\langle N^{(n)} \rangle_t = \langle Y \rangle_{(r+t) \wedge \sigma_r \wedge T_n} - \langle Y \rangle_{r \wedge T_n} = 0 \quad \forall t \geq 0.$$

Hence,

$$E[(N_t^{(n)})^2] = E[\langle N^{(n)} \rangle_t] = 0 \quad \forall t \geq 0,$$

thus

$$N_t^{(n)} = 0 \quad P\text{-a.s.}$$

Letting  $n \rightarrow \infty$  implies

$$Y_{(r+t) \wedge \sigma_r} - Y_r = 0 \quad P\text{-a.s.} \quad \forall t \geq 0,$$

which yields (A.0.3).

**Step 2.**  $W$  is a local martingale (up to  $\infty$ ) with respect to  $(\mathcal{G}_t)$ :

By A.0.2(ii)  $C_t$  is an  $(\mathcal{F}_s)$ -stopping time. For  $t \geq 0$

$$E[\langle Y^{C_t} \rangle_s] = E[\langle Y \rangle_{C_t \wedge s}] \leq E[\langle Y \rangle_{C_t}] = t.$$

Therefore, by Corollary 1.4.32  $Y^{C_t}$  is a martingale and

$$E[(Y_s^{C_t})^2] \leq \liminf_n E[Y_{s \wedge C_t \wedge T_n}^2] = \liminf_n E[\langle Y \rangle_{s \wedge C_t \wedge T_n}] \leq t.$$

Hence,  $(Y_s^{C_t})_{s \geq 0}$  and  $((Y_s^{C_t})^2)_{s \geq 0}$  are uniformly integrable.

Moreover,  $C_{\langle Y \rangle_{T_n}}$  is a  $(\mathcal{F}_t)$ -stopping time, because:

$$\begin{aligned} \{C_{\langle Y \rangle_{T_n}} < u\} &\stackrel{(A.0.1)}{=} \{\langle Y \rangle_u > \langle Y \rangle_{T_n}\} \\ &= \bigcup_{r \in \mathbb{Q}} \{\langle Y \rangle_u > r\} \cap \{r \geq \langle Y \rangle_{T_n}\} \\ &\stackrel{(A.0.1)}{=} \bigcup_{r \in \mathbb{Q}} \{C_r < u\} \cap \{C_r \geq T_n\} \\ &= \bigcup_{r \in \mathbb{Q}} \underbrace{\{C_r < u\}}_{\in \mathcal{F}_u} \cap \underbrace{\{C_r \wedge u \geq T_n\}}_{\in \mathcal{F}_{C_r \wedge u \wedge T_n} \subset \mathcal{F}_u} \end{aligned}$$

Thus,

$$\{C_{\langle Y \rangle_{T_n}} \leq u\} \in \mathcal{F}_{u+} = \mathcal{F}_u.$$

Furthermore,  $\langle Y \rangle_{T_n}$  is a stopping time with respect to  $(\mathcal{G}_t)$ , since

$$\{\langle Y \rangle_{T_n} \leq u\} \stackrel{(A.0.1)}{=} \{C_u \geq T_n\} \in \mathcal{F}_{T_n \wedge C_u} \subset \mathcal{F}_{C_u} = \mathcal{G}_k.$$

Hence, for all  $t > s$  (since  $C_{t \wedge \langle Y \rangle_{T_n}} = C_t \wedge C_{\langle Y \rangle_{T_n}}$ , because  $(C_t)$  is increasing)

$$\begin{aligned} E[W_{t \wedge \langle Y \rangle_{T_n}} | \mathcal{G}_s] &= E[W_{t \wedge \langle Y \rangle_{T_n}} | \mathcal{G}_s] = E[Y_{C_{t \wedge \langle Y \rangle_{T_n}}} | \mathcal{G}_s] \\ &= E[Y_{C_t \wedge C_{\langle Y \rangle_{T_n}}} | \mathcal{F}_{C_s}] = E[Y_{C_t \wedge C_{\langle Y \rangle_{T_n}}}^{C_t} | \mathcal{F}_{C_s}] \end{aligned}$$

Since  $(Y_s^{C_t})^{\wedge C_{\langle Y \rangle_{T_n}}}$  is an uniformly integrable martingale (note that  $C_s$  is not necessarily bounded), it follows that

$$= Y_{C_s \wedge C_{\langle Y \rangle_{T_n}}}^{C_t} \stackrel{s \leq t}{=} Y_{C_s \wedge C_{\langle Y \rangle_{T_n}}} = W_{s \wedge \langle Y \rangle_{T_n}}.$$

### A. Time Transformation

Hence,  $W$  is a local martingale with localising sequence  $(\langle Y \rangle_{T_n})_{n \in \mathbb{N}}$  (up to  $\infty$ ).

**Step 3.**  $\langle W \rangle_t = t \forall t$ . In particular,  $W$  is a Brownian motion:

Since by  $L^2$ -martingale convergence theorem  $((Y_s^{C_t})^2)_{s \geq 0}$  is uniformly integrable and  $\langle Y^{C_t} \rangle_s = \langle Y \rangle_{C_t \wedge s} \leq \langle Y \rangle_{C_t} = t \quad \forall s$ , it follows that  $(Y^{C_t})^2 - \langle Y^{C_t} \rangle$  is an uniformly integrable martingale. Hence, (though  $C_t, C_s$  are not bounded) by the stopping theorem we get that for all  $t > s$

$$\begin{aligned} E[W_t^2 - t | \mathcal{G}_s] &= E[Y_{C_t}^2 - \langle Y \rangle_{C_t} | \mathcal{F}_{C_s}] \\ &= E[(Y_{C_t}^{C_t})^2 - \langle Y^{C_t} \rangle_{C_t} | \mathcal{F}_{C_s}] = (Y_{C_s}^{C_t})^2 - \langle Y^{C_t} \rangle_{C_s} \\ &= Y_{C_s}^2 - \langle Y \rangle_{C_s} = W_s^2 - s. \end{aligned}$$

Therefore,  $(W_t^2 - t)_{t \geq 0}$  is a continuous martingale with respect to  $(\mathcal{G})_{t \geq 0}$ . By the Doob-Meyer decomposition it follows that  $\langle W \rangle_t = t$  for all  $t$  and by 1.4.32 that  $(W_t)_{t \geq 0}$  is a martingale. By Lévy's characterization theorem 1.5.34  $W_t$  is a Brownian motion.

**Step 4.**  $Y_t = W_{\langle Y \rangle_t}$  for all  $t \geq 0$ :

We have

$$W_{\langle Y \rangle_t} = Y_{C_{\langle Y \rangle_t}}$$

and since  $s \mapsto \langle Y \rangle_s$  is increasing and continuous

$$C_{\langle Y \rangle_t} = \inf\{s > 0 | \langle Y \rangle_s > \langle Y \rangle_t\} = \inf\{s > t | \langle Y \rangle_s > \langle Y \rangle_t\} = \sigma_t.$$

Therefore,  $\langle Y \rangle$  is constant on  $[t, C_{\langle Y \rangle_t}]$ . By (A.0.2) we get that  $Y$  is constant on  $[t, C_{\langle Y \rangle_t}]$ , thus,  $W_{\langle Y \rangle_t} = Y_{C_{\langle Y \rangle_t}} = Y_t$ .

**Remark A.0.4.** We have supposed that  $P[\langle Y \rangle_\infty = \infty] = 1$ . (This has been necessary as the counter example

$$\Omega := \{\omega\}, Y \equiv 0$$

shows.) Basically, the theorem also holds for the case, where  $P[\langle Y \rangle_\infty < \infty] > 0$ . But, one possibly has to enlarge  $\Omega$ .

**Construction of the enlarged Wiener space:**

Let  $(W'_t)_{t \geq 0}$  be a Wiener process on  $(\Omega', \mathcal{F}', P')$  with respect to  $(\mathcal{F}'_t)$ . Set

$$\begin{aligned} \tilde{\Omega} &:= \Omega \times \Omega', \\ \tilde{\mathcal{F}} &:= \mathcal{F} \otimes \mathcal{F}', \\ \tilde{P} &:= P \times P', \\ C_t &:= \begin{cases} \inf\{s | \langle Y \rangle_s > t\} & \text{on } \{\langle Y \rangle_\infty > t\}, \\ \infty & \text{on } \{\langle Y \rangle_\infty \leq t\}. \end{cases} \end{aligned}$$

Let

$$\begin{aligned} \hat{\mathcal{F}}_t &:= \sigma(\mathcal{F}_{C_t \wedge s} | s \geq 0), \\ \mathcal{G}_t &:= \hat{\mathcal{F}}_t \times \mathcal{F}'_t, \\ W_t &:= \begin{cases} Y_{C_t} & \text{on } \{\langle Y \rangle_\infty > t\}, \\ W'_t - W'_{\langle Y \rangle_\infty} + \langle Y \rangle_\infty & \text{on } \{\langle Y \rangle_\infty \leq t\}. \end{cases} \end{aligned}$$

Then  $W$  is a Wiener process on  $(\tilde{\Omega}, \tilde{\mathcal{F}}, \tilde{P})$  with respect to  $(\mathcal{G}_t)_{t \geq 0}$ ,  $\langle Y \rangle_t$  is a  $(\mathcal{G}_t)$ -stopping time and we have

$$Y_t = W_{\langle Y \rangle_t}.$$

For the proof cf. [IW89, Chapter II, Theorem 7,21].

□

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