

## Stochastic Calculus Notes

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### 1 Itô's Formula.

In this section we state and prove the "fundamental theorem of calculus" for stochastic integrals. Recall that if  $x(t)$  is continuously differentiable, then

$$f(x(t)) = f(x(0)) + \int_0^t f(x(s))x'(s)ds$$

However, for Brownian motion we have

$$f(B_t) = f(B_0) + \int_0^t f(B_s)dB_s + \frac{1}{2} \int_0^t f''(B_s)ds$$

This is a special case of Itô's formula. If the process we are integrating possesses jumps, we will need to add more terms to the right-hand side.

In the proofs in this section will not be needed the right-continuity of the filtration  $\{\mathcal{F}_t\}$ . However, this may be necessary to construct the stochastic integral. In particular, this assures that the local martingale in the semimartingale decomposition whose local martingale is a local  $L^2$ -martingale. The problem only can arise when the local martingale has unbounded jumps (recall Lemma 1.26 from part 2). In particular, when the martingale is continuous or its jumps have a uniform bound, it can be localized into an  $L^2$ -martingale.

#### 1.1 Itô's formula: proofs and special cases

We prove first Itô's formula for real-valued semimartingales and then consider the following special cases: continuous semimartingales, FV processes and Brownian motion. Next we consider vector-valued semimartingales.

If  $F \subset \mathbb{R}, C^2(D)$  is the space of functions  $f : D \rightarrow \mathbb{R}$  such that  $f', f''$  exist on  $D$  and are continuous. For a real or vector-valued cadlag process  $X$ , the jump at time  $s$  is denoted by  $\Delta X_s = X_s - X_{s-}$ . Recall also the closure of the path over a time interval for a cadlag process, given by

$$\overline{X[0, t]} = \{X(s) : 0 \leq s \leq t\} \cup \{X(s-) : 0 < s \leq t\}$$

Recall that a cadlag has at most countable many discontinuities, so in the theorem below the sum is over countable many terms.

**Theorem 1.1.** *Fix  $0 < T < \infty$ . Let  $D$  be a open subset of  $\mathbb{R}$  and  $f \in C^2(D)$ . Let  $Y$  be a cadlag semimartingale with quadratic variation process  $[Y]$ . Assume that for all  $\omega$  outside some event of probability zero,  $\overline{Y}[0, T] \subset D$ . Then*

$$\begin{aligned} f(Y_t) = f(Y_0) &+ \int_{(0, t]} f'(Y_{s-}) dY_s + \frac{1}{2} \int_{(0, t]} f''(Y_{s-}) d[Y]_s \\ &+ \sum_{s \in (0, t]} \left( f(Y_s) - f(Y_{s-}) - f'(Y_{s-}) \Delta Y_s - \frac{1}{2} f''(Y_{s-}) (\Delta Y_s)^2 \right) \end{aligned} \quad (1.1)$$

*Part of the conclusion is that the last sum over  $s \in (0, t]$  converges absolutely for almost every  $\omega$ . Both sides of the equality above are cadlag processes, and the equality above means indistinguishability on  $[0, T]$ , that is, the equality (1.1) holds almost surely for  $0 \leq t \leq T$ .*

*Proof.* Define the function  $\gamma$  on  $D \times D$  by  $\gamma(x, x) = 0$  for  $x \in D$ , and for  $x \neq y$ ,

$$\gamma(x, y) = \frac{1}{(y - x)^2} \left( f(y) - f(x) - f'(y - x) - \frac{1}{2} f''(x)(y - x)^2 \right) \quad (1.2)$$

Clearly,  $\gamma$  is continuous on  $\{(x, y) \in D \times D : x \neq y\}$ . We prove that  $\gamma$  is also continuous at diagonal points. To do this we use Taylor's theorem (Theorem A.1). Given  $z \in D$ , pick  $r > 0$  small enough so that  $(z - r, z + r) \subset D$ . Then for  $x, y \in D, x \neq y$ , there exists  $\theta_{x,y}$  between  $x$  and  $y$  such that

$$f(y) = f(x) + f'(x)(y - x) + \frac{1}{2} f''(\theta_{x,y})(y - x)^2$$

So for these  $(x, y)$

$$\gamma(x, y) = \frac{1}{2} (f''(\theta_{x,y}) - f''(x))$$

As  $(x, y) \rightarrow (z, z), \theta_{x,y} \rightarrow z$ , so by the continuity of  $f''$ ,

$$\gamma(x, y) \rightarrow 0 = \gamma(z, z).$$

This proves the continuity of  $\gamma$  on  $D \times D$ .

Now we write

$$f(y) - f(x) = f'(x)(y - x) + \frac{1}{2}f''(x)(y - x)^2 + \gamma(x, y)(y - x)^2$$

Given a partition  $\pi = \{t_i\}$  of  $[0, \infty)$ , we apply the identity above to obtain

$$f(Y_t) - f(Y_0) = \sum_i f'(Y_{t \wedge t_i})(Y_{t_{i+1} \wedge t} - Y_{t_i \wedge t}) \quad (1.3)$$

$$+ \frac{1}{2} \sum_i f''(Y_{t \wedge t_i})(Y_{t \wedge t_i} - Y_{t_i \wedge t})^2 \quad (1.4)$$

$$+ \sum_i \gamma(Y_{t_i}, Y_{t_{i+1}})(Y_{t \wedge t_i} - Y_{t_i \wedge t})^2 \quad (1.5)$$

By Proposition 1.37 and 1.59 from part 4 we can fix a sequence of partitions  $\pi^l$  such that  $\text{mesh}(\pi^l) \rightarrow 0$ , and so the following limits happen almost surely, uniformly on  $t \in [0, T]$ , as  $l \rightarrow \infty$ .

- i) The sum (1.3) converges to  $\int_{(0,t]} f'(Y_{s-}) dY_s$ .
- ii) The sum (1.4) converges to  $\frac{1}{2} f''(Y_{s-}) d[Y]_s$ .
- iii)  $\sum_i (Y_{t \wedge t_{i+1}} - Y_{t \wedge t_i})^2 \rightarrow [Y]_t$ .

Fix  $\omega$  such that the limits in i) – iii) hold. We apply Lemma A.2 in a simplified form so that the functions  $\phi, \gamma$  in (A.3) have no time variables. We use this lemma with  $d = 1$ , the cadlag  $s \rightarrow Y_s(\omega)$  on  $[0, t]$  and the sequence of partitions  $\pi^l$  chosen above. The closed set we take is  $K = \overline{Y[0, T]}$ . The continuous function is  $\phi(x, y) = \gamma(x, y)(y - x)^2$ . By hypothesis,  $K \subset D$ . Consequently, the function

$$\gamma(x, y) = \begin{cases} (x - y)^{-2} \phi(x, y), & x \neq y \\ 0, & x = y \end{cases}$$

is continuous on  $K \times K$ . Assumption (A.4) holds by iii). Thus the hypotheses of Lemma A.2 are satisfied. The conclusion is that for this fixed  $\omega$  and each  $t \in [0, T]$ , the sum on line (1.5)

converges to

$$\begin{aligned}\sum_{s \in (0, t]} \phi(Y_{s-}, Y_s) &= \sum_{s \in (0, t]} \gamma(Y_{s-}, Y_s) (Y_s - Y_{s-})^2 \\ &= \sum_{s \in (0, t]} \left( f(Y_s) - f(Y_{s-}) - f'(Y_{s-}) \Delta Y_s - \frac{1}{2} f''(Y_{s-}) (\Delta Y_s)^2 \right)\end{aligned}$$

And this sum is absolutely convergent by [Lemma A.2](#).

We have proved that given  $0 < T < \infty$ , for almost every  $\omega$  [\(1.1\)](#) holds for all  $0 \leq t \leq T$ .  $\square$

**Corollary 1.2.** *Under the conditions of [Theorem 1.1](#),  $f(Y)$  is a semimartingale.*

*Proof.* Using the formula [\(1.1\)](#), we obtain  $f(Y) = f(Y_0) + M + V$ , with  $M = \int f'(Y_-) dY$ , and  $V$  the remaining sums. We prove that this is a FV process. Clearly  $\int_{(0, t]} f''(Y_-) d[Y]$  is a FV process. Now we check that the sum over the jumps is also a FV process. Fix  $\omega$  such that  $Y_s(\omega)$  is a cadlag function and [\(1.1\)](#) holds. Let  $\{s_i\}$  denote the (at most countable many) jumps of  $s \rightarrow Y_s(\omega)$  in  $[0, T]$ . The theorem gives the absolute convergence

$$\sum_{s \in (0, t]} |f(Y_s) - f(Y_{s-}) - f'(Y_{s-}) \Delta Y_s - \frac{1}{2} f''(Y_{s-}) (\Delta Y_s)^2| < \infty$$

Consequently for this fixed  $\omega$  the sum in [\(1.1\)](#) defines a function in  $BV[0, T]$ .  $\square$

We state some simplifications of Itô's formula.

**Corollary 1.3.** *under the hypotheses of [Theorem 1.1](#) we have the following special cases.*

a) *If  $Y$  is continuous on  $[0, T]$ , then*

$$f(Y_t) = f(Y_0) + \int_0^t f'(Y_s) dY_s + \frac{1}{2} \int_0^t f''(Y_s) d[Y]_s \quad (1.6)$$

b) *If  $Y$  has bounded variation on  $[0, T]$ , then*

$$f(Y_t) = f(Y_0) + \int_{(0, t]} f'(Y_{s-}) dY_s + \sum_{s \in (0, t]} (f(Y_s) - f(Y_{s-}) - f'(Y_{s-}) \Delta Y_s) \quad (1.7)$$

c) *if  $Y_t = Y_0 + B_t$ , where  $B$  is a standard Brownian motion independent of  $Y_0$ , then*

$$f(Y_t) = f(Y_0) + \int_0^t f'(Y_0 + B_s) dB_s + \frac{1}{2} \int_0^t f''(Y_0 + B_s) ds \quad (1.8)$$

*Proof.* a) If  $Y$  is continuous, the jumps are zero, so the sums over jumps disappear in (1.1).

b) By lemma 1.36 in part 2 applied to the cadlag BV path  $Y$  we have

$$[Y]_t = \sum_{s \in (0, t]} (\Delta Y_s)^2$$

Consequently

$$\frac{1}{2} f''(Y_{s-}) d[Y]_s = \sum_{s \in (0, t]} \frac{1}{2} f''(Y_{s-}) (\Delta Y_s)^2$$

and we get (1.7).

c) This is a special case of a), where  $[B]_t = t$ .

□

**Remark 1.4.** The hypothesis  $\overline{Y[0, T]} \subset D$  is crucial in Theorem 1.1. The next example shows that  $Y[0, T] \subset D$  is not enough.

**Example 1.5.** Let  $D = (-\infty, 1) \cup (\frac{3}{2}, \infty)$ , and define

$$f(x) = \begin{cases} \sqrt{1-x}, & x < 1 \\ 0, & x > \frac{3}{2} \end{cases}$$

Then  $f \in C^2(D)$ . Define the deterministic process

$$Y_t = \begin{cases} t, & 0 \leq t < 1 \\ 1+t, & t \geq 1. \end{cases}$$

Then  $Y_t \in D$ , but if  $t < 1$ ,  $\overline{Y[0, t]}$  is not contained in  $D$  because  $1 \in \overline{Y[0, T]}$ . In that case,

$$\begin{aligned} \int_{(0, t]} f'(Y_{s-}) dY_s &= \int_{(0, 1)} f'(s) ds + f'(Y_{1-}) + \int_{(1, t]} f'(s) ds \\ &= -1 + (-\infty) + 0 \end{aligned}$$

So the integral is not finite.

Now we proceed to prove Itô's formula for vector-valued semimartingales. For purposes of matrix multiplications we think of points  $\mathbf{x} \in \mathbb{R}^d$  as columns vectors, so

$$\mathbf{x} = [x_1, x_2, \dots, x_d]^T$$

Let  $Y_1(t), Y_2(t), \dots, Y_d(t)$  be cadlag semimartingales with respect to a common filtration  $\{\mathcal{F}_t\}$ . We write  $Y(t) = [Y_1(t), \dots, Y_d(t)]^T$  for the columns vector, and call  $Y$  an  $\mathbb{R}^d$ -valued semimartingale. Its jump is the vector of jumps in the coordinates,

$$\Delta Y(t) = [\Delta Y_1(t), \dots, \Delta Y_d(t)]^T$$

For  $0 < T < \infty$  and an open subset  $D \subset \mathbb{R}^d$ ,  $C^{1,2}([0, T], D)$  is the space of continuous functions  $f : [0, T] \times D \rightarrow \mathbb{R}$  whose partial derivatives  $f_t, f_x$ , and  $f_{x_i, x_j}$  exists and are continuous in  $(0, T) \times D$ , and extend as continuous functions to  $[0, T] \times D$ . For  $f \in C^{1,2}([0, T] \times D)$  and  $(t, \mathbf{x}) \in [0, T] \times D$ , the spatial gradient is

$$Df(t, \mathbf{x}) = [f_{x_1}(t, \mathbf{x}), f_{x_2}(t, \mathbf{x}), \dots, f_{x_d}(t, \mathbf{x})]^T$$

and the Hessian matrix  $D^2 f(t, \mathbf{x})$  is the  $d \times d$  matrix of second-order spatial partial derivatives:

$$D^2 f(t, \mathbf{x}) = \begin{bmatrix} f_{x_1, x_1}(t, \mathbf{x}) & f_{x_1, x_2}(t, \mathbf{x}) & \cdots & f_{x_1, x_d}(t, \mathbf{x}) \\ f_{x_2, x_1}(t, \mathbf{x}) & f_{x_2, x_2}(t, \mathbf{x}) & \cdots & f_{x_2, x_d}(t, \mathbf{x}) \\ \vdots & \vdots & \ddots & \vdots \\ f_{x_d, x_1}(t, \mathbf{x}) & f_{x_d, x_2}(t, \mathbf{x}) & \cdots & f_{x_d, x_d}(t, \mathbf{x}) \end{bmatrix}$$

**Theorem 1.6.** Fix  $d \geq 1$ , and  $0 < T < \infty$ . Let  $D$  be an open subset of  $\mathbb{R}^d$  and  $f \in C^{1,2}([0, T] \times D)$ . Let  $Y$  be an  $\mathbb{R}^d$ -valued cadlag semimartingale such that outside some event of probability zero,  $\overline{Y[0, T]} \subset D$ . Then

$$\begin{aligned} & f(t, Y(t)) \\ &= f(0, Y(0)) + \int f_t(s, Y(s)) ds + \sum_{j=1}^d \int_{(0, t]} f_{x_j}(s, Y(s-)) dY_j(s) \\ &+ \frac{1}{2} \sum_{1 \leq j, k \leq d} \int_{(0, t]} f_{x_j, x_k}(s, Y(s-)) d[Y_j, Y_k](s) \\ &+ \sum_{s \in (0, t]} \left( f(s, Y(s)) - f(s, Y(s-)) - Df(s, Y(s-))^T \Delta Y(s) - \frac{1}{2} \Delta Y(s)^T D^2 f(s, Y(s-)) \Delta Y(s) \right) \end{aligned} \tag{1.9}$$

*Proof.* The idea is similar as in the scalar case. We write  $Y_t^k = Y_k(t)$ . Define a function  $\phi$  on

$[0, T]^2 \times D^2$  by the equality

$$\begin{aligned} f(t, \mathbf{y}) - f(s, \mathbf{x}) &= f_t(s, \mathbf{x})(t - s) + Df(s, \mathbf{x})^T(\mathbf{y} - \mathbf{x}) \\ &\quad + \frac{1}{2}(\mathbf{y} - \mathbf{x})^T D^2 f(s, \mathbf{x})(\mathbf{y} - \mathbf{x}) + \phi(s, t, \mathbf{x}, \mathbf{y}) \end{aligned} \quad (1.10)$$

Apply this to the partition intervals to write

$$\begin{aligned} f(t, Y_t) - f(0, Y_0) &= \sum_i f_t(t \wedge t_i, Y_{t \wedge t_i})((t \wedge t_{i+1}) - (t \wedge t_i)) \end{aligned} \quad (1.11)$$

$$+ \sum_{k=1}^d \sum_i f_{x_k}(t \wedge t_i, Y_{t \wedge t_i})(Y_{t \wedge t_{i+1}}^k - Y_{t \wedge t_i}^k) \quad (1.12)$$

$$+ \frac{1}{2} \sum_{1 \leq j, k \leq d} \sum_i f_{x_j, x_k}(t \wedge t_i, Y_{t \wedge t_i})(Y_{t \wedge t_{i+1}}^j - Y_{t \wedge t_i}^j)(Y_{t \wedge t_{i+1}}^k - Y_{t \wedge t_i}^k) \quad (1.13)$$

$$+ \sum_i \phi(t \wedge t_i, t \wedge t_{i+1}, Y_{t \wedge t_i}, Y_{t \wedge t_{i+1}}). \quad (1.14)$$

By Propositions 1.37 and 1.59 in part 4, we can fix a sequence of partitions  $\pi^l$  such that  $\text{mesh}(\pi^l) \rightarrow 0$  and so that the following limits happen almost surely, uniformly on  $[0, T]$ , as  $l \rightarrow \infty$ .

- i) (1.11) converges to  $\int_{(0, t]} f_t(s, Y_s) ds$ .
- ii) (1.12) converges to  $\sum_{k=1}^d \int_{(0, t]} f_{x_k}(s, Y_{s-}) dY_s^k$
- iii) (1.13) converges to  $\frac{1}{2} \sum_{1 \leq j, k \leq d} \int_{(0, t]} f_{x_j, x_k}(s, Y_{s-}) d[Y^j, Y^k]_s$
- iv)  $\sum_i (Y_{t \wedge t_{i+1}}^k - Y_{t \wedge t_i}^k)^2 \rightarrow [Y^k]_t$  for  $1 \leq k \leq d$ .

in i) – iii) the integrand is the left-limit process. For example in ii)

$$\lim_{r \searrow s} f_{x_k}(r, Y_r) = f_{x_k}(\lim_{r \searrow s} (r, Y_r)) = f_{x_k}(s, Y_{s-})$$

However in i) the  $ds$  integral does not distinguish between  $Y_{s-}$  and  $Y_s$  because a cadlag path has at most countable many jumps.

Fix  $\omega$  such that  $\overline{Y[0, T]} \subset D$  and the limits in items i) – iv) hold. These conditions hold for almost every  $\omega$ . We need to study the sum (1.14). To do this, we apply Lemma A.2 to the  $\mathbb{R}^d$ -valued cadlag function  $s \rightarrow Y_s(\omega)$  on  $[0, T]$ , with the function  $\phi$  defined by (1.10), the closed

set  $K = \overline{Y[0, T]}$ , and the sequence of partitions  $\pi^l$  chosen above. We need to check that  $\phi$  and the set  $K$  satisfy the hypotheses of [Lemma A.2](#). Continuity of  $\phi$  follows from [\(1.10\)](#). Next we argue that if  $(s_n, t_n, \mathbf{x}_n, \mathbf{y}_n) \rightarrow (u, u, \mathbf{z}, \mathbf{z})$  in  $[0, T]^2 \times K^2$  while for each  $n$ , either  $s_n \neq t_n$  or  $\mathbf{x}_n \neq \mathbf{y}_n$ , then

$$\frac{\phi(s_n, t_n, \mathbf{x}_n, \mathbf{y}_n)}{|t_n - s_n| + |\mathbf{y}_n - \mathbf{x}_n|^2} \rightarrow 0 \quad (1.15)$$

Given  $\epsilon > 0$ , let  $U$  be an interval around  $u$  in  $[0, T]$  and let  $B$  be an open ball centered at  $\mathbf{z}$  and contained in  $D$  such that

$$|f_t(v, \mathbf{w}) - f_t(u, \mathbf{z})| + |D^2 f(v, \mathbf{w}) - D^2 f(u, \mathbf{z})| \leq \epsilon \quad (1.16)$$

for all  $v \in I, \mathbf{w} \in B$ . Such an interval  $I$  and ball  $B$  exists by the openness of  $D$  and by the assumption of continuity of derivatives of  $f$  in  $[0, T] \times D$ . For large enough  $n$ , we have  $s_n, t_n \in I, \mathbf{x}_n, \mathbf{y}_n \in B$ . Since a ball is convex, by Taylor's formula [\(A.2\)](#) applied to  $f(t_n, y_n)$  and the definition of  $\phi$ , we can write

$$\begin{aligned} \phi(s_n, t_n, \mathbf{x}_n, \mathbf{y}_n) &= ((f_t(\tau_n, \mathbf{y}_n) - f_t(s_n, \mathbf{x}_n))(t_n - s_n) \\ &\quad + \frac{1}{2}(\mathbf{y}_n - \mathbf{x}_n)^T (D^2 f(s_n, \xi_n) - D^2 f(s_n, x_n))(\mathbf{y}_n - \mathbf{x}_n) \end{aligned}$$

where  $\tau_n$  lies between  $s_n$  and  $t_n$ , and  $\xi_n$  is a point on the line segment connecting  $\mathbf{x}_n$  and  $\mathbf{y}_n$ . In particular,  $\tau_n \in I$  and  $\xi_n \in B$ . By the Schwarz inequality for vectors  $|\mathbf{x}^t A \mathbf{y}| \leq |\mathbf{x}| |A| |\mathbf{y}|$  (Recall that if  $A = (a_{i,j})$ , then  $|A| = (\sum_{i,j} a_{i,j}^2)^{1/2}$ ) we get

$$\begin{aligned} |\phi(s_n, t_n, \mathbf{x}_n, \mathbf{y}_n)| &\leq |(f_t(\tau_n, \mathbf{y}_n) - f_t(s_n, \mathbf{x}_n))| \cdot |t_n - s_n| \\ &\quad + |D^2 f(s_n, \xi_n) - D^2 f(s_n, x_n)| \cdot |\mathbf{y}_n - \mathbf{x}_n|^2 \\ &\stackrel{(1.16)}{\leq} 2\epsilon(|t_n - s_n| + |\mathbf{y}_n - \mathbf{x}_n|^2) \end{aligned}$$

Thus for  $n$  large enough,

$$\left| \frac{\phi(s_n, t_n, \mathbf{x}_n, \mathbf{y}_n)}{|t_n - s_n| + |\mathbf{y}_n - \mathbf{x}_n|^2} \right| \leq 2\epsilon$$

This proves [\(1.15\)](#). Clearly, the function

$$(s, t, \mathbf{x}, \mathbf{y}) \rightarrow \frac{\phi(s, t, \mathbf{x}, \mathbf{y})}{|t - s| + |\mathbf{y} - \mathbf{x}|^2}$$



is continuous at points where either  $s \neq t$  or  $\mathbf{x} \neq \mathbf{y}$ , because is the quotient of two continuous functions. This proves the continuity of  $\gamma$  defined in [Lemma A.2](#) on  $[0, T]^2 \times K^2$ . The hypothesis in [\(A.4\)](#) is consequence of *iv*). This proves that the sum in [\(1.14\)](#) converges to

$$\begin{aligned} & \sum_{s \in (0, t]} \phi(s, s, Y_{s-}, Y_s) \\ &= \sum_{s \in (0, t]} \left( f(s, Y_s) - f(s, Y_{s-}) - Df(s, Y_{s-})^T \Delta Y_s - \frac{1}{2} \Delta Y_s^T D^2 f(s, Y_{s-}) \Delta Y_s \right) \end{aligned}$$

This completes the proof.  $\square$

**Remark 1.7.** If  $Y$  is a continuous  $\mathbb{R}^d$ -valued semimartingale, [\(1.6\)](#) can be expressed in differential notation as

$$df(t, Y(t)) = f_t(t, Y(t))dt + \sum_{j=1}^d f_{x_j}(t, Y(t))dY_j(t) + \frac{1}{2} \sum_{1 \leq j, k \leq d} f_{x_j, x_k}(t, Y(t))d[Y_j, Y_k](t) \quad (1.17)$$

We state the Brownian motion case as a corollary.

**Corollary 1.8.** Let  $B(t) = (B_1(t), \dots, B_d(t))$  be a Brownian motion in  $\mathbb{R}^d$ , with random initial point  $B(0)$ , and  $f \in C^2(\mathbb{R}^d)$ . Then

$$f(B(t)) = f(B(0)) + \int_0^t Df(B(s))^T dB(s) + \frac{1}{2} \int_0^t \Delta f(B(s))ds \quad (1.18)$$

Suppose  $f$  is harmonic (i.e.  $\Delta f = 0$ ) in an open set  $D \subset \mathbb{R}^d$ . Let  $D_1$  be an open subset of  $D$  such that  $\text{dist}(D_1, D^c) > 0$ . Assume initially  $B(0) = \mathbf{z}$  for some point  $\mathbf{z} \in D_1$ , and let

$$\tau = \inf\{t \geq 0 : B(t) \in D_1^c\} \quad (1.19)$$

Then  $f(B^\tau(t))$  is a local  $L^2$ -martingale.

*Proof.* Equality [\(1.18\)](#) comes directly from [\(1.6\)](#), noting that  $[B_j, B_k]_t = t\delta_{i,j}$ .

The process  $B^\tau$  is a vector valued  $L^2$ -martingale that satisfies  $\overline{B^\tau[0, T]} \subset D$  for all  $T < \infty$ . Thus Itô's formula applies. Note that  $[B_i^\tau, B_j^\tau]_t = [B_i, B_j]_t^\tau = \delta_{i,j}(t \wedge \tau)$ .  $\Delta f = 0$  implies that the second integral in [\(1.18\)](#) vanishes, so

$$f(B^\tau(t)) = f(\mathbf{z}) + \int_0^t Df(B^\tau(s))^T dB^\tau(s)$$

this shows that  $f(B^\tau)$  is a local  $L^2$ -martingale.  $\square$

## 1.2 Applications of Itô's formula.

**Example 1.9.** If we want to obtain  $\int_0^t B_s^k dB_s$  for a standard Brownian motion and  $k \geq 1$ , consider  $f(x) = \frac{1}{k+1}x^{k+1}$ , so that  $f'(x) = x^k$ ,  $f''(x) = kx^{k-1}$ . Now we apply Itô's formula to get

$$\int_0^t B_s^k dB_s = \frac{1}{k+1}B_t^{k+1} - \frac{k}{2} \int_0^t B_s^{k-1} ds$$

the integral on the right is the Riemann integral of the (almost surely) continuous function  $s \rightarrow B_s^{k-1}$ .

The next lemma shows how find martingales using Brownian motion and the heat equation in one dimension.

**Lemma 1.10.** Suppose  $f \in C^{1,2}(\mathbb{R}_+ \times \mathbb{R})$  and  $f_t + \frac{1}{2}f_{xx} = 0$ . Let  $B_t$  be a one-dimensional standard Brownian motion. Then  $f(t, B_t)$  is a local  $L^2$ -martingale. If

$$\int_0^T E[f_x(t, B_t)^2] dt < \infty \quad (1.20)$$

then  $f(t, B_t)$  is an  $L^2$ -martingale on  $[0, T]$ .

*Proof.* Recall that  $[B]_t = t$ , so we apply (1.6) to get

$$\begin{aligned} f(t, B_t) &= f(0, 0) + \int_0^t f_t(s, B_s) ds + \int_0^t f_x(s, B_s) dB_s + \frac{1}{2} \int_0^t f_{xx}(s, B_s) ds \\ &= f(0, 0) + \int_0^t f_x(s, B_s) dB_s + \int_0^t \left( f_t(s, B_s) + \frac{1}{2} f_{xx}(s, B_s) \right) ds \\ &= f(0, 0) + \int_0^t f_x(s, B_s) dB_s \end{aligned}$$

where the last line is a local  $L^2$ -martingale, because  $f_x(s, B_s)$  is a continuous process, hence predictable, and is bounded on compact time intervals, so condition *i*) in Proposition 1.28 from part 4 is satisfied.

The integrability condition (1.20) guarantees that  $f_x(s, B_s)$  lies in the space  $\mathcal{L}_2(B, \mathcal{P})$  of integrands on the interval  $[0, T]$ . Recall that our definition of stochastic integrals involves the time interval  $[0, \infty)$ . However, we can extend  $f_x(s, B_s)$  by declaring it identically zero on  $(T, \infty)$ . This does not change the integral on  $[0, T]$ .  $\square$

**Example 1.11.** Let  $\mu \in \mathbb{R}, \sigma \neq 0$  be constants. Let  $a < 0 < b$ , and let  $B_t$  be one-dimensional standard Brownian motion. Define  $X_t := \mu t + \sigma B_t$ , a Brownian motion with drift. We want to answer the following question: What is the probability that  $X_t$  exists the interval  $(a, b)$  through the point  $b$ ?

To solve this, define the stopping time

$$\tau := \inf\{t > 0 : X_t = a \text{ or } X_t = b\}$$

We check first that  $\tau < \infty$  almost surely. Consider the events

$$A_n := \{\sigma(B_{n+1} - B_n) > b - a + |\mu| + 1\}, n \in \mathbb{N}$$

Those events are independent, and have the same probability (because  $B_{n+1} - B_n \sim N(0, 1)$ ). Let's say that  $P(A_n) = z \in (0, 1)$ . Note that

$$P\left(\bigcap_{n=1}^N A_n^c\right) = \prod_{n=1}^N P(A_n^c) = (1 - z)^N$$

As  $z \in (0, 1)$ ,  $(1 - z)^N \rightarrow 0$  when  $N \rightarrow \infty$ . So  $P(\cap_{n \in \mathbb{N}} A_n^c) = 0$ . This is equivalent to say that  $P(\cup_{n \in \mathbb{N}} A_n) = 1$ . So almost surely some of the  $A'_n$ 's must happen. In that case, note that

$$A_n \subset \{X_{n+1} - X_n > b - a + 1\}$$

This says that if  $X_n \in (a, b)$ , then  $X_{n+1} \notin (a, b)$ . In particular,  $X_n$  cannot remain in  $(a, b)$  for all  $n$ . We have checked that  $\tau < \infty$  almost surely.

Next we are looking for a function  $h$  such that  $h(X_t)$  is a martingale. In that case, if we could justify  $Eh(X_\tau) = h(0)$ , we could compute  $P(X_\tau = b)$  from

$$h(0) = Eh(X_\tau) = h(a)P(X_\tau = a) + h(b)P(X_\tau = b).$$

Let  $f(t, x) = h(\mu t + \sigma x)$ . The condition from the last lemma  $f_t + f_{xx} = 0$  is equivalent to

$$\mu h' + \frac{1}{2}\sigma^2 h'' = 0$$

Suppose  $\mu \neq 0$ . The general solution of this ode is

$$h(x) = C_1 e^{-2\mu\sigma^{-2}x} + C_2, \quad C_1, C_2 \text{ are constants} \quad (1.21)$$

To verify (1.20), using (1.21) we obtain  $f_x(t, x) = \sigma h'(\mu t + \sigma x) = -2C_1\mu\sigma^{-1}e^{-2\mu\sigma^{-2}(\mu t + \sigma x)}$ . Therefore

$$\begin{aligned} \int_0^T E[f_x(t, B_t)^2]dt &= -2C_1\mu\sigma^{-1} \int_0^T E\left(e^{-4\mu\sigma^{-1}(\mu t + \sigma B_t)}\right) dt = K(T) \int_0^T \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi t}} e^{\sigma x - x^2/2t} dx dt \\ &= K(T) \int_0^T e^{\sigma^2 t/2} dt < \infty \end{aligned}$$

For all  $T < \infty$ , and then (1.20) is satisfied. Then  $M_t = f(t, B_t) = h(\mu t + \sigma B_t) = C_1 e^{-2(\frac{\mu}{\sigma})B_t - 2(\frac{\mu}{\sigma})^2 t} + C_2$  is a martingale. By optional stopping,  $M_{t \wedge \tau}$  is also a martingale, and so  $EM_{t \wedge \tau} = EM_0 = h(0)$ . By path continuity and  $\tau < \infty$  almost surely,  $M_{t \wedge \tau} \rightarrow M_\tau$  almost surely as  $t \rightarrow \infty$ . Furthermore the process  $M_{t \wedge \tau}$  is bounded, because up to time  $\tau$  the process  $X_t$  remains in  $[a, b]$ , and so  $|M_{t \wedge \tau}| \leq C := \sup_{a \leq x \leq b} |h(x)|$ . Dominated convergence gives  $EM_\tau = h(0)$ .

Finally, we can choose constants  $C_1, C_2$  so that  $h(b) = 1, h(a) = 0$ . In that case, we obtain

$$P(X_\tau = b) = h(0) = \frac{e^{-2\mu a/\sigma^2} - 1}{e^{-2\mu a/\sigma^2} - e^{-2\mu b/\sigma^2}} \quad (1.22)$$

The case  $\mu = 0$  is much simpler. In that case,  $h'' = 0$ , so  $h(x) = C_1 x + C_2$  for some constants  $C_1, C_2$ . Then clearly  $f(t, B_t) = C_1(\sigma B_t) + C_2$  is a martingale and using the same arguments as before, if we want  $h(a) = 0, h(b) = 1$ , then  $C_1 = \frac{1}{b-a}, C_2 = \frac{-a}{b-a}$ . So  $P(X_\tau = b) = \frac{-a}{b-a}$ .

The martingale found before is a special case of the following:  $M_t = e^{X_t - \frac{1}{2}[X]_t}$  is a continuous local  $L^2$ -martingale whenever  $X$  is. To check this, we apply Itô's formula to the semimartingale  $Y = X - \frac{1}{2}[X]$ ,  $f(x) = e^x$  so that  $f'(x) = f''(x) = f(x)$ , and then

$$dM = M dY + \frac{1}{2} M d[Y, Y] = M dX - \frac{1}{2} M d[X] + \frac{1}{2} M d[X - [X]/2] = M dX$$

To obtain the last equality, we write  $[X - [X]/2] = [X] + \frac{1}{4}[[X]] - 2[X, [X]/2] = [X]$  due to continuity and bounded variation (see Lemma 1.36 in part 2). We conclude that  $M_t = M_0 + \int_0^t M_s dX_s$ , a continuous local  $L^2$ -martingale by the construction of the stochastic integral.

Next we investigate transience and recurrence in Brownian motion.

**Proposition 1.12.** *Let  $B_t$  be a Brownian motion on  $\mathbb{R}$ . Then  $\limsup_{t \rightarrow \infty} B_t = \infty$  and  $\liminf_{t \rightarrow \infty} B_t = -\infty$  almost surely. Consequently, almost every Brownian motion path visits every point infinitely often.*

*Proof.* Let  $\tau_0 = 0$  and  $\tau_{k+1} := \inf\{t > \tau_k : |B_t - B_{\tau_k}| = 4^{k+1}\}$ . By the strong Markov property, for each  $k$ ,  $\{B_{t+\tau_k} - B_{\tau_k} : t \geq 0\}$  is a standard Brownian motion, independent of

$\mathcal{F}_{\tau_k}$ . By symmetry,

$$P[B_{\tau_{k+1}} - B_{\tau_k} = 4^{k+1}] = P[B_{\tau_{k+1}} - B_{\tau_k} = -4^{k+1}] = \frac{1}{2}$$

And the random variables  $\{B_{\tau_{k+1}} - B_{\tau_k} : k \geq 0\}$  are independent. Thus, for any  $n \in \mathbb{N}$

$$\begin{aligned} P\left(\bigcap_{m \geq n} B_{\tau_{m+1}} - B_{\tau_m} = 4^{m+1}\right) &= 0 \\ P\left(\bigcap_{m \geq n} B_{\tau_{m+1}} - B_{\tau_m} = -4^{m+1}\right) &= 0 \end{aligned}$$

This says that almost surely, for any  $n \in \mathbb{N}$  there exists some indexes  $j, k$  such that  $B_{\tau_{j+1}} - B_{\tau_j} = 4^{j+1}, B_{\tau_{k+1}} - B_{\tau_k} = -4^{k+1}$ . But then since

$$|B_{\tau_j}| \leq \sum_{i=1}^j |B_{\tau_i} - B_{\tau_{i-1}}| \leq \sum_{i=1}^j 4^i = \frac{4^{j+1} - 1}{4 - 1} \leq \frac{4^{j+1}}{2}$$

$B_{\tau_{j+1}} \geq 4^j$ . By the same argument,  $B_{\tau_{k+1}} \leq -4^k$ . Thus  $\limsup_{t \rightarrow \infty} B_t = \infty$  and  $\liminf_{t \rightarrow \infty} B_t = -\infty$  almost surely. Finally, note that as there are sequences  $\{t_n\}, \{s_n\}$  with  $t_n, s_n \nearrow \infty$  and  $B_{t_n} \rightarrow \infty, B_{s_n} \rightarrow -\infty$ , for any  $x \in \mathbb{R}$  there are infinitely many  $n$  such that  $B_{t_n} \geq x, B_{s_n} \leq x$  and by the continuity of Brownian motion, it has to cross the point  $x$  each opportunity, so it visits every point infinitely often.  $\square$

**Proposition 1.13.** *Let  $B_t$  be a Brownian motion in  $\mathbb{R}^d$ , and let  $P^z$  be the probability measure when the process  $B_t$  started at point  $z \in \mathbb{R}^d$ . Let  $\tau_r := \inf\{t \geq 0 : |B_t| \leq r\}$  be the first time Brownian motion hits the ball of radius  $r$  around the origin*

a) *If  $d = 2, P^z(\tau_r < \infty) = 1$  for all  $r > 0, z \in \mathbb{R}^d$ .*

b) *If  $d \geq 3$ , then for  $z$  outside the ball of radius  $r$ ,*

$$P^z(\tau_r < \infty) = \left(\frac{r}{|z|}\right)^{d-2}$$

*There will be an almost surely finite time  $T$  such that  $|B_t| > r$  for all  $t \geq T$ .*

c) *For  $d \geq 2$  and any  $z, y \in \mathbb{R}^d$ ,*

$$P^z[B_t \neq y \text{ for all } 0 < t < \infty] = 1$$

*The case  $y = z$  is allowed (that's why the time  $t = 0$  is excluded).*

*Proof.* Note first that in c) it's enough to consider  $\mathbf{y} = 0$ , because

$$P^{\mathbf{z}}[B_t \neq \mathbf{y} \text{ for all } 0 < t < \infty] = P^{\mathbf{z}-\mathbf{y}}[B_t \neq 0 \text{ for all } 0 < t < \infty]$$

Also, it's suffices to consider  $\mathbf{z} \neq 0$ , because if we know the result for  $\mathbf{z} \neq 0$ , then the case  $\mathbf{z} = 0$  comes from the Markov property:

$$\begin{aligned} P^0[B_t \neq 0 \text{ for all } 0 < t < \infty] &= \lim_{s \searrow 0} P^0[B_t \neq 0 \text{ for all } s < t < \infty] \\ &= \lim_{s \searrow 0} P^{B(s)}[B_t \neq 0 \text{ for all } 0 < t < \infty] = 1 \end{aligned}$$

because  $P(B(s) = 0) = \lim_{\epsilon \rightarrow 0} P(B_s \in (-\epsilon, \epsilon)) = 0$ , thus  $B_s \neq 0$  with probability one.

We can assume  $\mathbf{z} \neq 0$  and  $r < |\mathbf{z}|$ .

First we consider the case  $d = 2$ . The function  $g(\mathbf{x}) = \log(|\mathbf{x}|)$  is harmonic in  $D = \mathbb{R}^2 \setminus \{0\}$ . Let  $\sigma_R := \inf\{t \geq 0 : |B_t| \geq R\}$ . Pick  $r < |\mathbf{z}| < R$ , and define the annulus  $A = \{\mathbf{x} : r < |\mathbf{x}| < R\}$ . The time to exit the annulus is  $\zeta = \tau_r \wedge \sigma_R$ . As we proved before, each coordinate of  $B_t$  has  $\limsup \infty$  almost surely, the exit time of  $\sigma_R$  is finite, and hence  $\zeta$ . We apply (1.8) to the harmonic function

$$f(\mathbf{x}) = \frac{\log R - \log|\mathbf{x}|}{\log R - \log r}$$

and the annulus  $A$ . We get that  $f(B_{\zeta \wedge t})$  is a local  $L^2$ -martingale, and since  $f$  is bounded on the closure of  $A$ , by dominated convergence,  $f(B_{\zeta \wedge t})$  is an  $L^2$ -martingale. If we repeat the optional stopping argument from the last example, we obtain that

$$E^{\mathbf{z}} f(B_{\zeta \wedge t}) = E^{\mathbf{z}}(f(B_0)) = f(\mathbf{z})$$

Taking  $t \rightarrow \infty$  and using that  $f(r) = 1, f(R) = 0$ ,

$$f(\mathbf{z}) = E^{\mathbf{z}} f(B_{\zeta}) = P^{\mathbf{z}}(|B_{\zeta}| = r) f(r) + P^{\mathbf{z}}(|B_{\zeta}| = R) f(R) = P^{\mathbf{z}}(|B|_{\zeta} = r) = P^{\mathbf{z}}(\tau_r < \sigma_R)$$

We conclude that

$$P^{\mathbf{z}}(\tau_r < \sigma_R) = \frac{\log R - \log|\mathbf{z}|}{\log R - \log r} \quad (1.23)$$

From this we can obtain a), c) for  $d = 2$ . To check a), note that  $\sigma_R \nearrow \infty$  as  $R \nearrow \infty$ , because for a fixed path Brownian motion is bounded on bounded time intervals. We deduce that

$$P^{\mathbf{z}}(\tau_r < \infty) = \lim_{R \rightarrow \infty} P^{\mathbf{z}}(\tau_r < \sigma_R) = 1$$

To prove c) in case  $d = 2$ , consider  $r = r(k) = (1/k)^k$  and  $R = R(k) = k$ . Then we get

$$P^{\mathbf{z}}(\tau_{r(k)} < \sigma_{R(k)}) = \frac{\log k - \log |z|}{(k+1)\log k} \xrightarrow{k \rightarrow \infty} 0$$

Let  $\tau = \inf\{t \geq 0 : B_t = 0\}$ . For  $0 < r < |z|$ ,  $\tau_r \leq \tau$  because it has to enter to the ball of radius  $r$  before hitting 0. Since  $\sigma_{R(k)} \nearrow \infty$  as  $k \nearrow \infty$ ,

$$P^{\mathbf{z}}(\tau < \infty) = \lim_{k \rightarrow \infty} P^{\mathbf{z}}(\tau < \sigma_{R(k)}) \leq \lim_{k \rightarrow \infty} P^{\mathbf{z}}(\tau_{r(k)} < \sigma_{R(k)}) = 0$$

Now we proceed to the case  $d \geq 3$ . Consider the harmonic function  $g(\mathbf{z}) = |\mathbf{z}|^{2-d}$ , and apply Itô's formula to the function

$$f(\mathbf{x}) = \frac{R^{2-d} - |\mathbf{z}|^{2-d}}{R^{2-d} - r^{2-d}}$$

The annulus  $A$  and the stopping times  $\sigma_R, \zeta$  are defined as before. Applying the same reasoning as above the result is

$$P^{\mathbf{z}}(\tau_r < \sigma_R) = \frac{R^{2-d} - |z|^{2-d}}{R^{2-d} - r^{2-d}} \quad (1.24)$$

Taking  $R \rightarrow \infty$ ,

$$P^{\mathbf{z}}(\tau_r < \infty) = \frac{|z|^{2-d}}{r^{2-d}} = \left( \frac{r}{|z|} \right)^{d-2}$$

this shows part b) for  $d \geq 3$ . Part c) follows because the quantify above converges to zero when  $r \rightarrow 0$ .

It's remains to show that there will be an almost surely finite time  $T$  such that  $|B_t| \geq r$  for all  $t \geq T$ . Let  $r < R$ . Define  $\sigma_R^1 = \sigma_R$  and for  $n \geq 2$ ,

$$\tau_r^n := \inf\{t > \sigma_R^{n-1} : |B_t| \leq r\}$$

$$\sigma_R^n := \inf\{t > \tau_r^n : |B_t| \geq R\}$$

So that  $\sigma_R^1 < \tau_r^2 < \sigma_R^2 < \tau_r^3 < \dots$  are the successive visits to radius  $R$  and back to radius  $r$ . Let  $\alpha = (r/R)^{d-2} < 1$ . We claim that for  $n \geq 2$ ,  $P^{\mathbf{z}}(\tau_r^n < \infty) = \alpha^{n-1}$ . In that case,

$$\sum_n P^{\mathbf{z}}(\tau_r^n < \infty) < \infty$$

and Borel-Cantelli tell us that  $\tau_r^n < \infty$  only can happen finitely many times. To prove the claim, the case  $n = 2$  followed from the strong Markov property,

$$P^{\mathbf{z}}(\tau_r^2 < \infty) = P^{\mathbf{z}}(\sigma_R^1 < \tau_r^2 < \infty) = P^{B(\sigma_R^1)}(\tau_r < \infty) = \left( \frac{r}{|B_{\sigma_R^1}|} \right)^{d-2} = \alpha$$

Now we procede by induction

$$\begin{aligned} P^{\mathbf{z}}(\tau_r^n < \infty) &= P^{\mathbf{z}}(\tau_r^{n-1} < \sigma_R^{n-1} < \tau_r^n < \infty) = P^{\mathbf{z}}(\tau_r^{n-1} < \sigma_R^{n-1} < \infty, \sigma_R^{n-1} < \tau_r^n < \infty) \\ &\stackrel{\text{indep}}{=} P^{\mathbf{z}}(\tau_r^{n-1} < \sigma_R^{n-1} < \infty) P^{\mathbf{z}}(\sigma_R^{n-1} < \tau_r^n < \infty) = P^{\mathbf{z}}(\tau_r^{n-1} < \infty) \cdot \alpha = \alpha^{n-1} \end{aligned}$$

In the last step, we used that if  $\tau_r^{n-1} < \infty$ , then  $\sigma_R^{n-1}$  has to be finite, because each coordinate of  $B_t$  has  $\limsup \infty$ . This concludes the claim, and also the proof.  $\square$

Now comes an important theorem of characterization of Brownian motion. We will need this lemma.

**Lemma 1.14.** *Let  $X$  be a random  $d$ -vector and  $\mathcal{A}$  a sub- $\sigma$ -algebra on  $(\Omega, \mathcal{F}, P)$ . Let*

$$\phi(\theta) := \int_{\mathbb{R}^d} e^{i\theta^T \mathbf{x}} \mu(d\mathbf{x}) \quad (\theta \in \mathbb{R}^d)$$

*be the characteristic function of a probability distribution  $\mu$  on  $\mathbb{R}^d$ . Assume*

$$E\left(e^{i\theta^T X} \mathbb{1}_A\right) = \phi(\theta) P(A)$$

*for all  $\theta \in \mathbb{R}^d$ ,  $A \in \mathcal{A}$ . Then  $X$  has distribution  $\mu$  and is independent of  $\mathcal{A}$ .*

*Proof.* Taking  $A = \Omega$ , we get the first claim. Fix  $A$  such that  $P(A) > 0$ , and define the probability measure  $\nu_A$  on  $\mathbb{R}^d$  by

$$\nu_A(B) := \frac{E(\mathbb{1}_B(X) \mathbb{1}_A)}{P(A)}, \quad B \in \mathcal{B}(\mathbb{R}^d)$$

By definition, we have

$$\int_{\mathbb{R}^d} \mathbb{1}_B(x) \nu(dx) = \frac{1}{P(A)} \int_{\Omega} \mathbb{1}_A \mathbb{1}_B(X) dP$$

Thus for any  $f : \mathbb{R}^d \rightarrow \mathbb{C}$  continuous and bounded,

$$\int_{\mathbb{R}^d} f(x) \nu(dx) = \frac{1}{P(A)} \int_{\Omega} \mathbb{1}_A f(X) dP$$

Taking  $f = e^{i\theta^T x}$  for some fixed  $\theta$ ,

$$\int_{\mathbb{R}^d} e^{i\theta^T x} \nu(dx) = \frac{1}{P(A)} \int_{\Omega} \mathbb{1}_A e^{i\theta^T X} dP = \phi(\theta)$$

So  $\phi$  is the characteristic function of  $\nu$ , and so  $\nu = \mu$ . In particular,

$$P(\{X \in B\} \cap A) = \nu_A(B) P(A) = \mu(B) P(A) = P(X \in B) P(A)$$

As  $A \in \mathcal{A}$  is arbitrary, the independence of  $X, \mathcal{A}$  follows.  $\square$



**Theorem 1.15** (Lévy's Characterization of Brownian Motion.). *Let  $M = [M_1, \dots, M_d]^T$  be a continuous  $\mathbb{R}^d$ -valued local martingale and  $X(t) = M(t) - M(0)$ . Then  $X$  is a standard Brownian motion relative to  $\{\mathcal{F}_t\}$  iff  $[X_i, X_j]_t = \delta_{i,j}t$ . In particular, in this case the process  $X$  is independent of  $\mathcal{F}_0$ .*

*Proof.* We have already seen that  $d$ -dimensional standard Brownian motion satisfies  $[B_i, B_j]_t = \delta_{i,j}t$ . Now we proceed to prove the converse. First observe that a continuous local martingale is also a local  $L^2$ -martingale. Fix a vector  $\theta = (\theta_1, \dots, \theta_d)^T \in \mathbb{R}^d$ , and define

$$f(t, x) := e^{i\theta^T x + \frac{1}{2}|\theta|^2 t}$$

Let  $Z_t := f(t, X(t))$ , and apply Itô's formula (it also applies to complex-valued functions) to this function,

$$\begin{aligned} Z_t &= 1 + \frac{1}{2}|\theta|^2 \int_0^t Z_s ds + \sum_{j=1}^d i\theta_j \int_0^t Z_s dX_j(s) - \frac{1}{2} \int_0^t Z_s ds \underbrace{\sum_{j=1}^d \theta_j^2}_{=|\theta|^2} \\ &= 1 + i \sum_{j=1}^d \theta_j \int_0^t Z_s dX_j(s) \end{aligned}$$

This shows that  $Z$  is a local  $L^2$ -martingale. However, on any bounded time interval,  $Z$  is bounded, so that  $Z$  is an  $L^2$ -martingale. The martingale property  $E[Z_t | \mathcal{F}_s] = Z_s$  for  $s < t$  can be rewritten as

$$E \left( e^{i\theta^T (X(t) - X(s))} | \mathcal{F}_s \right) = e^{-\frac{1}{2}|\theta|^2(t-s)} \quad (1.25)$$

The right-hand side is  $\phi(\theta)$ , where  $\phi$  is the characteristic function of a  $N(0, t-s)$  random variable. Thus  $X(t) - X(s) \sim N(0, t-s)$  by the previous lemma and  $X(t) - X(s)$  is independent of  $\mathcal{F}_s$ . Therefore  $X$  satisfies all the properties of a standard Brownian motion relative to  $\{\mathcal{F}_t\}$ .  $\square$

Here is an application of Lévy's criterion.

**Example 1.16.** (Bessel processes.) Let  $d \geq 2$  and  $B(t) = [B_1(t), \dots, B_d(t)]^T$  a  $d$ -dimensional Brownian motion. Set

$$R_t := |B(t)| = (B_1(t)^2 + \dots + B_d(t)^2)^{1/2} \quad (1.26)$$

We will find the semimartingale decomposition of  $R_t$ .

Suppose that  $B(0) = \mathbf{z} \neq 0$  so that  $R_0 = |\mathbf{z}| > 0$ . Let  $D = \mathbb{R}^d \setminus \{0\}$  and  $f(\mathbf{x}) = |\mathbf{x}|$ . Then  $f \in C^2(D)$ ,  $f_{x_i} = x_i |\mathbf{x}|^{-1}$ ,  $\Delta f = (d-1) |\mathbf{x}|^{-1}$ . By [Proposition 1.13 c\)](#), with probability 1 the path  $B[0, T]$  is a closed subset of  $D$ , for any  $T < \infty$ . We can apply Itô's formula and gives

$$R_t = |z| + \sum_{j=1}^d \int_0^t \frac{B_j(s)}{|B_s|} dB_j(s) + \frac{d-1}{2} \int_0^t \frac{ds}{R_s} \quad (1.27)$$

We prove that the stochastic integral is a 1-dimensional Brownian motion using the Lévy's criterion. Let

$$W_t := \sum_{j=1}^d \int_0^t \frac{B_j(s)}{|B_s|} dB_j(s)$$

and compute their quadratic variation

$$\begin{aligned} [W]_t &= \left[ \sum_{j=1}^d \int_0^t \frac{B_j(s)}{|B_s|} dB_j(s), \sum_{j=1}^d \int_0^t \frac{B_j(s)}{|B_s|} dB_j(s) \right]_t \\ &= \sum_{i,j} \left[ \int_0^t \frac{B_i(s)}{|B_s|} dB_i(s), \int_0^t \frac{B_j(s)}{|B_s|} dB_j(s) \right] \\ &= \sum_{i,j} \int_0^t \frac{B_i(s) B_j(s)}{|B_s|^2} d[B_i, B_j]_s \\ &= \sum_i \int_0^t \frac{B_i(s)^2}{|B_s|^2} ds = \int_0^t ds = t \end{aligned}$$

Thus  $W$  is a standard Brownian motion. We rewrite (1.27) as

$$R_t = |z| + \frac{d-1}{2} \int_0^t R_s^{-1} ds + W_t$$

Process  $R_t$  is the *Bessel process* with dimension  $d$ , or with parameter  $\frac{d-1}{2}$ .

As a final application, we prove one case of the Burkholder-Davis-Gundy inequalities. Recall the notation  $M_t^* = \sup_{0 \leq s \leq t} |M_s|$ .

**Proposition 1.17.** *Let  $p \in [2, \infty)$ , and  $C_p = (p(p-1)e)^{p/2}$ . Then for all continuous local martingales  $M$  with  $M_0 = 0$  and all  $0 < t < \infty$ ,*

$$E[(M_t^*)^p] \leq C_p E([M_t]^{p/2}) \quad (1.28)$$

*Proof.* Let  $f(x) = |x|^p$ . We have  $f \in C^2(\mathbb{R})$ , with  $f'(x) = \text{sign}(x)p|x|^{p-1}$ ,  $f''(x) = p(p-1)|x|^{p-2}$ , where  $\text{sign}(x) = \frac{x}{|x|}$  for  $x \neq 0$  (and the convention at  $x = 0$  is immaterial). We apply Itô's formula for continuous local martingales to get

$$|M_t|^p = p \int_0^t \text{sign}(M_s) |M_s|^{p-1} dM_s + \frac{p(p-1)}{2} \int_0^t |M_s|^{p-2} d[M]_s$$

Suppose first  $M$  is bounded. Then  $M$  is an  $L^2$ -martingale and  $M \in \mathcal{L}_2(M, \mathcal{P})$ . Consequently the term  $p \int_0^t \text{sign}(M_s) |M_s|^{p-1} dM_s$  is a mean zero  $L^2$ -martingale. Take expectations and apply Holder inequality with exponents  $\frac{p-2}{p}, \frac{p}{2}$ :

$$\begin{aligned} E(|M_t|^p) &= \frac{p(p-1)}{2} E \int_0^t |M_s|^{p-2} d[M]_s \leq \frac{p(p-1)}{2} E((M_t^*)^{p-2} [M]_t) \\ &\leq \frac{p(p-1)}{2} (E((M_t^*)^p))^{1-2/p} E([M]_t^{p/2})^{2/p} \end{aligned}$$

On the other hand, by Doob's inequality and the fact that  $(\frac{p}{p-1})^p \leq 2e$  for  $p \geq 2$ ,

$$E[(M_t^*)^p] \leq \left( \frac{p}{p-1} \right)^p E(|M_t|^p) \leq ep(p-1) (E((M_t^*)^p))^{1-2/p} E([M]_t^{p/2})^{2/p}$$

Rearranging the above inequality gives the conclusion for a bounded martingale.

The general case comes from localization. Let  $\tau_k = \inf\{t \geq 0 : |M|_t \geq k\}$ . By continuity,  $M^{\tau_k}$  is bounded, so we can apply the previous result to  $M^{\tau_k}$  and obtain

$$E[(M_{\tau_k \wedge t})^p] \leq C_p E([M]_{\tau_k \wedge t}^{p/2})$$

Finally, take  $k \nearrow \infty$  and apply monotone convergence and deduce (1.28).  $\square$

## A Some auxilliary results.

We state the Taylor's theorem for real and vector-valued functions. This is used in the proof of Itô's formula.

**Theorem A.1.** *a) Let  $(a, b)$  be an open interval in  $\mathbb{R}$ ,  $f \in C^{1,2}([0, T] \times (a, b))$ ,  $s, t \in [0, T]$  and  $x, y \in (a, b)$ . Then there exists a point  $\tau$  between  $s$  and  $t$ ,  $\theta$  between  $x$  and  $y$  such that*

$$f(t, y) = f(s, x) + f_x(s, x)(y - x) + f_t(\tau, y)(t - s) + \frac{1}{2} f_{xx}(s, \theta)(y - x)^2 \quad (\text{A.1})$$

b) Let  $G$  be an open convex set in  $\mathbb{R}^d$ ,  $f \in C^{1,2}([0, T] \times G)$ ,  $s, t \in [0, T]$  and  $\mathbf{x}, \mathbf{y} \in G$ . Then there exists a point  $\tau$  between  $s$  and  $t$ ,  $\theta \in [0, 1]$  such that, with  $\xi = \theta\mathbf{x} + (1 - \theta)\mathbf{y}$ ,

$$f(t, \mathbf{y}) = f(s, \mathbf{x}) + Df(s, \mathbf{x})^T(\mathbf{y} - \mathbf{x}) + f_t(\tau, \mathbf{y})(t - s) + \frac{1}{2}(\mathbf{y} - \mathbf{x})^T D^2 f(s, \xi)(\mathbf{y} - \mathbf{x}) \quad (\text{A.2})$$

The following technical lemma is also used in the proof of Itô's formula.

**Lemma A.2.** Let  $g_1, \dots, g_d$  be cadlag functions on  $[0, T]$ , and form the  $\mathbb{R}^d$ -valued function  $\mathbf{g} = (g_1, \dots, g_d)^T$ . A cadlag function on a bounded set is bounded, so there exists a closed set  $K \subset \mathbb{R}^d$  such that  $\mathbf{g}(s) \in K$  for all  $s \in [0, T]$ .

Let  $\phi$  be a continuous function on  $[0, T]^2 \times K^2$  such that the function

$$\gamma(s, t, \mathbf{x}, \mathbf{y}) = \begin{cases} \frac{\phi(s, t, \mathbf{x}, \mathbf{y})}{|t - s| + |\mathbf{y} - \mathbf{x}|^2}, & s \neq t \text{ or } \mathbf{x} \neq \mathbf{y} \\ 0, & s = t \text{ and } \mathbf{x} = \mathbf{y} \end{cases} \quad (\text{A.3})$$

is also continuous on  $[0, T]^2 \times K^2$ . Let  $\pi^l = \{0 = t_0^l < t_1^l < \dots < t_{m(l)}^l = T\}$  be a sequence of partitions on  $[0, T]$  such that  $\text{mesh}(\pi^l) \rightarrow 0$  as  $l \rightarrow \infty$ , and

$$C_0 = \sup_l \sum_{i=0}^{m(l)-1} |\mathbf{g}(t_{i+1}^l) - \mathbf{g}(t_i^l)|^2 < \infty \quad (\text{A.4})$$

Then

$$\lim_{l \rightarrow \infty} \sum_{i=0}^{m(l)-1} \phi(t_i^l, t_{i+1}^l, \mathbf{g}(t_i^l), \mathbf{g}(t_{i+1}^l)) = \sum_{s \in (0, T]} \phi(s, s, \mathbf{g}(s-), \mathbf{g}(s)) \quad (\text{A.5})$$

The limit on the right is a finite, absolutely convergent sum.