

## 4 Brownian martingales

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### 4a Heat equation appears

Recall 3a5:  $\mathbb{E} B^2(T) = \mathbb{E} T$  for every stopping time  $T$  such that (say)  $T \leq 1$  a.s. This is a manifestation of the martingale property of the process

$$M(t) = B^2(t) - t,$$

as explained below. Here is a solution of 3a5 (hopefully, not new to you). Using 2f8 we have<sup>1</sup>

$$\begin{aligned} 0 &= \mathbb{E} (B^2(1) - 1) = \iint (Y^2(1)(\omega_1, \omega_2) - 1) P(d\omega_1)P(d\omega_2) = \\ &= \int P(d\omega_1) \int P(d\omega_2) \left( (B(T(\omega_1))(\omega_1) + B(1 - T(\omega_1))(\omega_2))^2 - 1 \right) = \\ &= \int P(d\omega_1) f(T(\omega_1), B(T(\omega_1))(\omega_1)) = \mathbb{E} g(T, B(T)) = \mathbb{E} (B^2(T) - T), \end{aligned}$$

where

$$g(t, x) = \int P(d\omega_2) ((x + B(1-t)(\omega_2))^2 - 1) = \mathbb{E} ((x + B(1-t))^2 - 1) = x^2 - t.$$

The relevant property of the function  $f(t, x) = x^2 - t$  is  $\mathbb{E} f(1, x + B(1-t)) = f(t, x)$ . More generally,<sup>2</sup>

$$(4a1) \quad \mathbb{E} f(s+t, x + B(t)) = f(s, x), \quad \text{that is,} \\ \int f(s+t, x+y) p_t(y) dy = f(s, x).$$

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<sup>1</sup>Compare it with the proof of (2a7).

<sup>2</sup>As before,  $p_t(x) = (2\pi t)^{-1/2} \exp(-\frac{x^2}{2t})$ .

Three examples of such functions:

$$(4a2) \quad \begin{aligned} f(t, x) &= x, \\ f(t, x) &= x^2 - t, \\ f(t, x) &= x^3 - 3tx \end{aligned}$$

(check it). We define new functions  $f_{+t}$  for  $t \in [0, \infty)$  by<sup>1</sup>

$$(4a3) \quad f_{+t}(s, x) = \mathbb{E} f(s+t, x+B(t)) = \int f(s+t, x+y)p_t(y) dy$$

and note that  $(f_{+t})_{+u} = f_{+(t+u)}$  (think, why). Now, the idea is simple and natural. We have a dynamics in (some) space of functions, and (4a1) means that  $f$  is a fixed point,

$$f_{+t} = f \quad \text{for all } t \geq 0,$$

that is, the speed vanishes at  $f$ ,

$$\frac{1}{\varepsilon}(f_{+\varepsilon} - f) \rightarrow 0 \quad \text{as } \varepsilon \rightarrow 0+.$$

Denoting for convenience  $\frac{\partial^{i+j}}{\partial s^i \partial x^j} f(s, x)$  by  $f_{i,j}(s, x)$  we have for small  $t, y$

$$\begin{aligned} f(s+t, x+y) &\approx f(s, x) + f_{1,0}(s, x)t + f_{0,1}(s, x)y + \frac{1}{2}f_{0,2}(s, x)y^2; \\ f_{+\varepsilon}(s, x) &= \mathbb{E} f(s+\varepsilon, x+B(\varepsilon)) \approx f(s, x) + f_{1,0}(s, x)\varepsilon + \frac{1}{2}f_{0,2}(s, x)\underbrace{\mathbb{E} B^2(\varepsilon)}_{=\varepsilon}; \\ \frac{1}{\varepsilon}(f_{+\varepsilon} - f) &\rightarrow f_{1,0} + \frac{1}{2}f_{0,2}. \end{aligned}$$

No one of the higher terms contributes (think, why). Thus, we guess that (4a1) is equivalent to a partial differential equation (PDE) well-known as the heat equation:<sup>2</sup>

$$(4a4) \quad \left( \frac{\partial}{\partial t} + \frac{1}{2} \frac{\partial^2}{\partial x^2} \right) f(t, x) = 0.$$

The question is, how to prove it, and what to require of  $f$ .

<sup>1</sup>Assuming integrability. Of course,  $p_t$  does not work for  $t = 0$ .

<sup>2</sup>Or rather, time reversed heat equation with coefficient 1/2; the standard heat equation contains  $\frac{\partial}{\partial t} - \frac{\partial^2}{\partial x^2}$ .

**4a5 Lemma.** Let  $f : (0, \infty) \times \mathbb{R} \rightarrow \mathbb{R}$  be a continuous function such that the derivatives  $f_{i,j}$  exist and are continuous for  $(i, j) \in \{(1, 0), (0, 2)\}$ . Assume that<sup>1</sup>

$$(4a6) \quad \frac{1}{x^2} \ln^+ |f_{i,j}(t, x)| \rightarrow 0 \quad \text{as } x \rightarrow \pm\infty$$

for every  $t \in (0, \infty)$ ,  $(i, j) \in \{(0, 0), (1, 0), (0, 2)\}$ , and moreover, it holds uniformly in  $t \in [a, b]$  whenever  $0 < a < b < \infty$ . Then  $f_{+t}$  is well-defined (by (4a3)) for all  $t \in (0, \infty)$ , and

$$\frac{d}{dt} f_{+t}(s, x) = \int \left( f_{1,0}(s+t, x+y) + \frac{1}{2} f_{0,2}(s+t, x+y) \right) p_t(y) dy$$

for all  $t \in (0, \infty)$  and  $(s, x) \in (0, \infty) \times \mathbb{R}$ . Both sides are claimed to be well-defined (the derivative in the left-hand side and the integral in the right-hand side).

**4a7 Exercise.** For every twice continuously differentiable function  $g : [x - \varepsilon, x + \varepsilon] \rightarrow \mathbb{R}$ ,

$$\min_{[x-\varepsilon, x+\varepsilon]} g''(\cdot) \leq \frac{g(x-\varepsilon) - 2g(x) + g(x+\varepsilon)}{\varepsilon^2} \leq \max_{[x-\varepsilon, x+\varepsilon]} g''(\cdot).$$

Prove it.

*Proof of 4a5.* First,  $f_{+t}$  is well-defined due to (4a6) for  $(i, j) = (0, 0)$ .

Second, without loss of generality we assume that  $s = 0$ ,  $x = 0$  (since the shifted function  $(s_1, x_1) \mapsto f(s + s_1, x + x_1)$  satisfies all the conditions imposed on  $f$ ).

Right derivative is considered below; left derivative, treated similarly, is left to the reader.

We have for every  $\varepsilon > 0$

$$f_{+t}(0, 0) = \int f(t, y) p_t(y) dy;$$

$$\begin{aligned} f_{+(t+\varepsilon)}(0, 0) &= \int f(t + \varepsilon, y) p_{t+\varepsilon}(y) dy = \\ &= \int dy p_t(y) \int dz p_\varepsilon(z) f(t + \varepsilon, y + z) = \int dy p_t(y) \int dz p_1(z) f(t + \varepsilon, y + z\sqrt{\varepsilon}) = \\ &= \int dy p_t(y) \int dz p_1(z) \frac{f(t + \varepsilon, y - z\sqrt{\varepsilon}) + f(t + \varepsilon, y + z\sqrt{\varepsilon})}{2}; \end{aligned}$$

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<sup>1</sup>Here  $\ln^+ a = \max(0, \ln a)$ .

$$\begin{aligned} & \frac{f_{+(t+\varepsilon)}(0,0) - f_{+t}(0,0)}{\varepsilon} = \\ & = \int dy p_t(y) \int dz p_1(z) \frac{f(t+\varepsilon, y - z\sqrt{\varepsilon}) - 2f(t+\varepsilon, y) + f(t+\varepsilon, y + z\sqrt{\varepsilon})}{2\varepsilon}. \end{aligned}$$

By 4a7 and continuity of  $f_{0,2}$ ,

$$\frac{f(t+\varepsilon, y - z\sqrt{\varepsilon}) - 2f(t+\varepsilon, y) + f(t+\varepsilon, y + z\sqrt{\varepsilon})}{2\varepsilon} \rightarrow \frac{z^2}{2} f_{0,2}(t, y) \quad \text{as } \varepsilon \rightarrow 0+.$$

Taking into account that

$$\frac{f(t+\varepsilon, y) - f(t, y)}{\varepsilon} \rightarrow f_{1,0}(t, y)$$

we get

$$\frac{f(t+\varepsilon, y - z\sqrt{\varepsilon}) - 2f(t+\varepsilon, y) + f(t+\varepsilon, y + z\sqrt{\varepsilon})}{2\varepsilon} \rightarrow f_{1,0}(t, y) + \frac{z^2}{2} f_{0,2}(t, y)$$

as  $\varepsilon \rightarrow 0+$ . Now we need an integrable majorant. Using 4a7 again,

$$\left| \frac{f(t+\varepsilon, y - z\sqrt{\varepsilon}) - 2f(t+\varepsilon, y) + f(t+\varepsilon, y + z\sqrt{\varepsilon})}{2\varepsilon} \right| \leq \max_{|y-z|\sqrt{\varepsilon}, y+z|\sqrt{\varepsilon}} |f_{0,2}(t+\varepsilon, \cdot)| \leq C(\delta) \exp(\delta(|y|+|z|\sqrt{\varepsilon})^2) \leq C(\delta) \exp(2\delta(y^2+z^2\varepsilon))$$

by (4a6) for  $f_{0,2}$  (locally uniform in  $t \dots$ ); any  $\delta > 0$  may be chosen. Also,

$$\left| \frac{f(t+\varepsilon, y) - f(t, y)}{\varepsilon} \right| \leq \max_{[t, t+\varepsilon]} |f_{1,0}(\cdot, y)| \leq C(\delta) \exp(\delta y^2)$$

by (4a6) for  $f_{1,0}$  (locally uniform in  $t$ ). We have a majorant

$$C(\delta) \exp(2\delta(y^2 + z^2\varepsilon)) p_t(y) p_1(z),$$

integrable if  $\delta$  is small enough (namely,  $2\delta < \frac{1}{2t}$  and  $2\delta\varepsilon < 1/2$ ). By the dominated convergence theorem (applied to  $\iint dy dz \dots$ ),

$$\begin{aligned} \frac{f_{+(t+\varepsilon)}(0,0) - f_{+t}(0,0)}{\varepsilon} & \rightarrow \int dy p_t(y) \int dz p_1(z) \left( f_{1,0}(t, y) + \frac{z^2}{2} f_{0,2}(t, y) \right) = \\ & = \int dy p_t(y) \left( f_{1,0}(t, y) + \frac{1}{2} f_{0,2}(t, y) \right) \quad \text{as } \varepsilon \rightarrow 0+. \end{aligned}$$

□

**4a8 Exercise.** Consider in detail the other case: left derivative.

See also [1], Sect. 7.5, Exercise 5.5.

**4a9 Proposition.** Condition (4a1) is equivalent to the PDE (4a4) for every function satisfying the conditions of Lemma 4a5.

*Proof.* Follows from 4a5, since  $f_{+\varepsilon} \rightarrow f$  as  $\varepsilon \rightarrow 0+$ . □

**4a10 Proposition.**  $\mathbb{E} f(T, B(T)) = f(0, 0)$  for every function  $f$  satisfying the conditions of Lemma 4a5 and the PDE (4a4), and every stopping time  $T$  such that  $\exists t \mathbb{P}(T \leq t) = 1$ .

The proof given for  $f(t, x) = x^2 - t$  in the beginning of 4a generalizes immediately.

**4a11 Exercise.** For every polynomial  $P$  on  $\mathbb{R}$  the following polynomial  $f$  on  $\mathbb{R}^2$  satisfies (4a1):

$$f(t, x) = \sum_k (-1)^k \frac{2^k}{(2k)!} P^{(k)}(t) x^{2k}.$$

Prove it.

Now we can continue (4a2) a little:

$$\begin{aligned} f(t, x) &= x^2 - t, & P(t) &= -t, \\ f(t, x) &= x^4 - 6tx^2 + 3t^2, & P(t) &= 3t^2. \end{aligned}$$

**4a12 Exercise.** <sup>1</sup>  $\text{Var} T = 2/3$  for  $T = \min\{t : |B(t)| = 1\}$ .

Prove it. (Warning: be careful with  $t \rightarrow \infty$ .)

An astonishing counterexample was found by Tychonoff<sup>2,3</sup>: let

$$P(t) = \begin{cases} \exp(-(1-t)^{-2}) & \text{for } t \in [0, 1), \\ 0 & \text{for } t \in [1, \infty) \end{cases}$$

(not a polynomial, of course, but a non-analytic infinitely differentiable function), then the formula given in 4a11 produces a power series convergent for

<sup>1</sup>See [1], Sect. 7.5, Theorem (5.9).

<sup>2</sup>A.N. Tychonoff, *Matem. Sbornik* **32** (1935), 199–216.

<sup>3</sup>See (1.18)–(1.24) on page 212 in: F. John, “Partial differential equations”, Springer (fourth edition).

all  $x$  (and all  $t$ ) to an infinitely differentiable function that satisfies the PDE (4a4) but violates (4a1).

Trying  $f(t, x) = \exp(at + bx)$  we get  $f_{1,0} = af$  and  $f_{0,2} = b^2f$ , thus, (4a1) is satisfied if and only if  $a + 0.5b^2 = 0$ ;

$$f(t, x) = e^{\lambda x} e^{-\lambda^2 t/2}.$$

Also functions

$$\begin{aligned} \frac{f(t, x) + f(t, -x)}{2} &= e^{-\lambda^2 t/2} \cosh \lambda x, \\ \frac{f(t, x) - f(t, -x)}{2} &= e^{-\lambda^2 t/2} \sinh \lambda x \end{aligned}$$

satisfy (4a1). Replacing  $\lambda$  with  $i\lambda$  we get functions

$$\begin{aligned} f(t, x) &= e^{\lambda^2 t/2} \cos \lambda x, \\ f(t, x) &= e^{\lambda^2 t/2} \sin \lambda x \end{aligned}$$

satisfying (4a1).

**4a13 Exercise.** <sup>1</sup> Let  $T = \min\{t : |B(t)| = 1\}$ , then

$$\mathbb{E} e^{\lambda T} = \begin{cases} \frac{1}{\cosh \sqrt{2|\lambda|}} & \text{for } -\infty < \lambda \leq 0, \\ \frac{1}{\cos \sqrt{2\lambda}} & \text{for } 0 \leq \lambda < \pi^2/8, \\ \infty & \text{for } \lambda \geq \pi^2/8. \end{cases}$$

Prove it. Also, give a new proof of 3a7.

## 4b Conditioning and martingales

Conditioning is simple in two frameworks: discrete probability, and densities. However, conditioning of a Brownian motion on its past goes far beyond these two frameworks. The clue is, the ‘restart’ introduced in Sect. 2:

$$X(t)(\omega_1, \omega_2) = \begin{cases} B(t)(\omega_1) & \text{if } t \leq T(\omega_1), \\ B(T(\omega_1))(\omega_1) + B(t - T(\omega_1))(\omega_2) & \text{if } t \geq T(\omega_1). \end{cases}$$

Here  $T$  is a stopping time (as defined by 2f5) and of course,  $T(\omega)$  means  $T(B(\cdot)(\omega))$ . Let us write  $X$  in a shorter form:

$$(4b1) \quad X(\cdot)(\omega_1, \omega_2) = B(\cdot)(\omega_1) \bigg|_{\bigg|}^{T(\omega_1)} B(\cdot)(\omega_2),$$

<sup>1</sup>See also [1], Sect. 7.5, Th. (5.7).

where  $f \overset{a}{\sqcup} g$  is defined for  $f, g \in C[0, \infty)$  and  $a \in [0, \infty)$  by

$$f \overset{a}{\sqcup} g = h \in C[0, \infty),$$

$$h(t) = \begin{cases} f(t) & \text{if } t \leq a, \\ f(a) + g(t - a) - g(0) & \text{if } t \geq a. \end{cases}$$

Not only the map  $(f, g) \mapsto f \overset{a}{\sqcup} g$  is Borel measurable for each  $a$ , but also

$$(f, a, g) \mapsto f \overset{a}{\sqcup} g$$

is a Borel measurable map  $C[0, \infty) \times \mathbb{R} \times C[0, \infty) \rightarrow C[0, \infty)$ , and therefore

$(f, g) \mapsto f \overset{T(f)}{\sqcup} g$  is a Borel measurable map  $C[0, \infty) \times C[0, \infty) \rightarrow C[0, \infty)$ .<sup>1</sup>

**4b2 Definition.** The conditional distribution of  $B(\cdot)$  given  $B(\cdot)|_{[0, a]} = f|_{[0, a]}$  is the distribution of  $f \overset{a}{\sqcup} B$ .

Some Borel functions  $\varphi : C[0, \infty) \rightarrow \mathbb{R}$  are such that  $\varphi(f \overset{T(f)}{\sqcup} g)$  depends on  $f$  only (not  $g$ ). Some of these functions are indicators,  $\varphi : C[0, \infty) \rightarrow \{0, 1\}$ . The corresponding sets are, by definition, the sub- $\sigma$ -field  $\mathcal{B}_T \subset \mathcal{B}_\infty$ .

A Borel function  $\varphi : C[0, \infty) \rightarrow \mathbb{R}$  is  $\mathcal{B}_T$ -measurable if and only if  $\varphi(f \overset{T(f)}{\sqcup} g)$  depends on  $f$  only (think, why).

Events of the form  $\{B(\cdot) \in G\}$  for  $G \in \mathcal{B}_T$  are, by definition, the sub- $\sigma$ -field  $\mathcal{F}_T$  on  $\Omega$ . Especially, the sub- $\sigma$ -field  $\mathcal{F}_\infty$  on  $\Omega$  generated by the Brownian motion consists of the events of the form  $\{B(\cdot) \in G\}$  for  $G \in \mathcal{B}_\infty$ .<sup>2</sup>

Conditional probability of an event of  $\mathcal{F}_\infty$  given  $\mathcal{F}_T$  is (by definition) the random variable

$$(4b3) \quad \mathbb{P}(B(\cdot) \in G | \mathcal{F}_T)(\omega_1) = \mathbb{P}\left(\{\omega_2 : B(\cdot)(\omega_1) \overset{T(\omega_1)}{\sqcup} B(\cdot)(\omega_2) \in G\}\right).$$

In other words, it is the probability according to the conditional distribution of  $B(\cdot)$  given  $B(\cdot)|_{[0, T]}$ . If  $G \in \mathcal{B}_T$  then  $\mathbb{P}(B(\cdot) \in G | \mathcal{F}_T) = \mathbb{1}_G$  (as it should be).

By the Fubini theorem,

$$\mathbb{E}(\mathbb{P}(B(\cdot) \in G | \mathcal{F}_T)) = \mathbb{P}(X(\cdot) \in G),$$

<sup>1</sup>Recall Lemma B in Correction to 2f.

<sup>2</sup>Often  $\mathcal{F}_\infty = \mathcal{F}$ ; always  $\mathcal{F}_\infty \subset \mathcal{F}$ ; and sometimes  $\mathcal{F}_\infty \neq \mathcal{F}$ , recall 3c ( $\Omega \times \{0, 1, \dots, n\}^\infty$ ).

$X$  being defined by (4b1). Thus, the strong Markov property turns into the total probability formula

$$\mathbb{P}(B(\cdot) \in G) = \mathbb{E}(\mathbb{P}(B(\cdot) \in G | \mathcal{F}_T)) \quad \text{for } G \in \mathcal{B}_\infty.$$

In other words,

$$\mathbb{P}(A) = \mathbb{E}(\mathbb{P}(A | \mathcal{F}_T)) \quad \text{for } A \in \mathcal{F}_\infty.$$

You may be astonished if you are acquainted with the general theory of conditioning. In that framework the total probability formula holds for every sub- $\sigma$ -field, irrespective of any Markov property!

In that framework, however, conditional probabilities are defined as to satisfy this formula, for each sub- $\sigma$ -field separately.<sup>1</sup> In contrast, we define them constructively by (4b3). In our definitions, the conditional distribution of  $B(\cdot)$  given  $B(\cdot)|_{[0,T]}$  at  $\omega$  depends on  $T(\omega)$  and the path on  $[0, T(\omega)]$ , not on the choice of  $T$  (as far as  $T(\omega)$  is fixed).

The space  $L_1(\mathcal{F}_\infty) = L_1(\Omega, \mathcal{F}_\infty, P)$  consists of (equivalence classes of) random variables of the form  $\varphi(B(\cdot))$  where  $\varphi : C[0, \infty) \rightarrow \mathbb{R}$  is a Borel function such that  $\mathbb{E}|\varphi(B(\cdot))| < \infty$ .

Conditional expectation of a random variable of  $L_1(\mathcal{F}_\infty)$  given  $\mathcal{F}_T$  is (by definition) the random variable

$$(4b4) \quad \mathbb{E}(\varphi(B(\cdot)) | \mathcal{F}_T)(\omega_1) = \int \varphi\left(B(\cdot)(\omega_1) \bigsqcup_{\omega_2}^{T(\omega_1)} B(\cdot)(\omega_2)\right) P(d\omega_2).$$

In other words, it is the expectation according to the conditional distribution of  $B(\cdot)$  given  $B(\cdot)|_{[0,T]}$ .

**4b5 Exercise.** Prove the total expectation formula:

$$\mathbb{E} \varphi(B(\cdot)) = \mathbb{E}(\mathbb{E}(\varphi(B(\cdot)) | \mathcal{F}_T))$$

for  $\varphi(B(\cdot)) \in L_1(\mathcal{F}_\infty)$ .

In other words:

$$\mathbb{E} X = \mathbb{E}(\mathbb{E}(X | \mathcal{F}_T)) \quad \text{for } X \in L_1(\mathcal{F}_\infty).$$

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<sup>1</sup>An example in the framework of densities: the conditional density of  $X$  given  $Y = 0$  is proportional to  $p_{X,Y}(x, 0)$ , but the conditional density of  $X$  given  $Y/X = 0$  is proportional to  $|x|p_{X,Y}(x, 0)$ .



**4b6 Exercise.** For all  $X \in L_\infty(\mathcal{F}_T)$ ,  $Y \in L_1(\mathcal{F}_\infty)$ ,

$$\mathbb{E}(XY | \mathcal{F}_T) = X\mathbb{E}(Y | \mathcal{F}_T).$$

Prove it.

**4b7 Definition.** A *Brownian martingale*<sup>1</sup> is a family  $(M_t)_{t \in [0, \infty)}$  of  $M_t \in L_1(\mathcal{F}_t)$  such that

$$\mathbb{E}(M_t | \mathcal{F}_s) = M_s \quad \text{a.s. for } 0 \leq s \leq t < \infty.$$

(I write  $M_t$  and  $M(t)$  interchangeably.)

Examples of Brownian martingales:

$$\begin{aligned} M(t) &= B(t), & M(t) &= B^2(t) - t, \\ M(t) &= B^3(t) - 3tB(t), & M(t) &= B^4(t) - 6tB^2(t) + 3t^2, \\ M(t) &= e^{-\lambda^2 t/2} \exp \lambda B(t), \\ M(t) &= e^{-\lambda^2 t/2} \cosh \lambda B(t), & M(t) &= e^{-\lambda^2 t/2} \sinh \lambda B(t), \\ M(t) &= e^{\lambda^2 t/2} \cos \lambda B(t), & M(t) &= e^{\lambda^2 t/2} \sin \lambda B(t) \end{aligned}$$

and, more generally,

$$(4b8) \quad M(t) = f(t, B(t))$$

where  $f$  satisfies the conditions of Lemma 4a5 and the PDE (4a4). And, of course, the process

$$M_t = \mathbb{E}(X | \mathcal{F}_t)$$

is a Brownian martingale for every  $X \in L_1(\mathcal{F}_\infty)$ .

**4b9 Exercise.** Prove that the following process is a Brownian martingale:

$$M(t) = \begin{cases} \int_0^t B(s) ds + (1-t)B(t) & \text{if } t \in [0, 1], \\ \int_0^1 B(s) ds & \text{if } t \in [1, \infty). \end{cases}$$

**4b10 Theorem.** If  $f$  satisfies the conditions of Lemma 4a5 then the following process is a Brownian martingale:

$$M(t) = f(t, B(t)) - \int_0^t g(s, B(s)) ds,$$

where  $g = f_{1,0} + \frac{1}{2}f_{0,2}$ , that is,

$$g(t, x) = \left( \frac{\partial}{\partial t} + \frac{1}{2} \frac{\partial^2}{\partial x^2} \right) f(t, x).$$

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<sup>1</sup>A Brownian martingale is a martingale w.r.t. the Brownian filtration  $(\mathcal{F}_t)_t$ . Generally, a martingale w.r.t. a given filtration is defined similarly.

**4b11 Exercise.** Prove that

$$\frac{d}{dt} \mathbb{E} \int_0^t g(s, B(s)) ds = \mathbb{E} g(t, B(t)) \quad \text{for } t > 0.$$

**4b12 Exercise.** Prove that

$$\frac{d}{dt} \mathbb{E} M(t) = 0 \quad \text{for } t > 0.$$

**4b13 Exercise.** Prove Theorem 4b10.

## 4c Nothing happens suddenly to Brownian motion

**4c1 Theorem.** Every Brownian martingale can be upgraded<sup>1,2</sup> to a random continuous function.

Similarly to the Brownian motion itself, we always upgrade  $(M_t)_t$  this way.

Sample functions of a Brownian martingale are continuous.

**4c2 Corollary.** For every stopping time  $T$  such that  $T > 0$  a.s. there exist stopping times  $T_1, T_2, \dots$  such that almost surely

$$T_n \uparrow T, \quad T_n < T.$$

Nothing happens suddenly to Brownian motion.

A striking contrast to jumping processes! I mean the Poisson process, the special Levy process, the Cauchy process.

*Proof of Corollary 4c2.* The process<sup>3</sup>

$$M_t = \mathbb{E} \left( \frac{T}{T+1} \mid \mathcal{F}_t \right)$$

is a Brownian martingale. By Theorem 4c1 it is continuous a.s. Clearly,  $M_T = \frac{T}{T+1}$  and moreover,

$$T = \inf \left\{ t : M_t - \frac{t}{t+1} = 0 \right\}$$

<sup>1</sup>In the sense discussed in 1e (after 1e6).

<sup>2</sup>Generally, a martingale (w.r.t. any filtration) can be upgraded to a random r.c.l.l function.

<sup>3</sup>The idea is, to consider the expected remaining time,  $\mathbb{E}(T-t \mid \mathcal{F}_t)$ . However,  $T$  need not be integrable.

(think, why). We take

$$T_n = \min\{t \in [0, \infty) : M_t - \frac{t}{t+1} \leq \frac{1}{n}\}.$$

□

Now we have to prove Theorem 4c1.

We consider  $(M_t)_{t \in [0,1]}$  (larger intervals are treated similarly); thus,

$$M(t) = \mathbb{E}(X | \mathcal{F}_t)$$

for some  $X \in L_1(\mathcal{F}_1)$  (namely,  $X = M_1$ ). Our first goal is to show that it is sufficient to ensure a.s. continuity of  $M(\cdot)$  for a dense set of  $X \in L_1(\mathcal{F}_1)$ .

**4c3 Proposition.** For every  $X \in L_1(\mathcal{F}_1)$  there exist  $X_n \in L_1(\mathcal{F}_1)$  such that

- (a)  $\mathbb{E}|X_n - X| \rightarrow 0$  as  $n \rightarrow \infty$ ,
- (b) each martingale

$$M_n(t) = \mathbb{E}(X_n | \mathcal{F}_t)$$

can be upgraded to a random continuous function,

- (c)  $M_n(\cdot)$  converge in  $C[0,1]$  almost surely; that is,

$$\max_{t \in [0,1]} |M_n(t) - M_\infty(t)| \rightarrow 0 \quad \text{as } n \rightarrow \infty$$

for some random continuous function  $M_\infty(\cdot)$ .

*Proof of Th. 4c1, given Prop. 4c3.* The equivalence class  $M(t)$  contains  $M_\infty(t)$ , since

$$\begin{aligned} \mathbb{E}|M(t) - M_n(t)| &= \mathbb{E}|\mathbb{E}(X - X_n | \mathcal{F}_t)| \leq \\ &\leq \mathbb{E}(\mathbb{E}(|X - X_n| | \mathcal{F}_t)) = \mathbb{E}|X - X_n| \rightarrow 0 \end{aligned}$$

and therefore  $M_{n_k}(t) \rightarrow M(t)$  a.s. (for an appropriate subsequence). □

**4c4 Exercise.** Let  $(M_t)_t$  be a Brownian martingale, and  $T \subset [0,1]$  a countable set. Then

$$\mathbb{P}\left(\sup_{t \in T} |M(t)| \geq c\right) \leq \frac{\mathbb{E}|M(1)|}{c} \quad \text{for } c \in (0, \infty).$$

Prove it.

**4c5 Exercise.** Prove Item (c) of Prop. 4c3, assuming Items (a) and (b).

It remains to ensure a.s. continuity (in  $t$ ) of  $M(t) = \mathbb{E}(X | \mathcal{F}_t)$  for all  $X$  of a dense set of  $\mathcal{X} \subset L_1(\mathcal{F}_1)$ . Moreover, it is enough if linear combinations of these  $X$  are dense. There are several reasonable choices of such  $\mathcal{X}$ . You may try bounded uniformly continuous functions  $C[0, \infty) \rightarrow \mathbb{R}$ , or  $X = f(B(t_1), \dots, B(t_n))$  for bounded continuous  $f : \mathbb{R}^n \rightarrow \mathbb{R}$ , or  $X = \int_0^1 \exp(if(t)B(t)) dt$  etc. I prefer indicators  $X = \mathbb{1}_A$  of events of the form

$$(4c6) \quad A = \{a_1 \leq B(t_1) \leq b_1, \dots, a_n \leq B(t_n) \leq b_n\}.$$

**4c7 Exercise.** Let  $0 < t_1 < \dots < t_n < 1$ , and  $-\infty < a_k < b_k < \infty$  for  $k = 1, \dots, n$ . Then the random function

$$M(t) = \mathbb{E} \left( \prod_{k=1}^n \mathbb{1}_{[a_k, b_k]}(B(t_k)) \middle| \mathcal{F}_t \right)$$

is continuous a.s.

Prove it.

Linear combinations of (arbitrary) indicators are dense in  $L_1(\mathcal{F}_1)$ . It remains to prove that sets of the form (4c6) and their disjoint unions are dense in  $\mathcal{F}_1 \pmod{0}$  according to the metric  $\text{dist}(A, B) = \mathbb{P}(A \setminus B) + \mathbb{P}(B \setminus A)$ . This claim follows easily from the next general lemma.

**4c8 Lemma.** Let  $(\Omega, \mathcal{F}, P)$  be a probability space, and  $\mathcal{A} \subset \mathcal{F}$  an algebra of sets that generates  $\mathcal{F} \pmod{0}$ . Then:

(a) For every  $C \in \mathcal{F}$  and  $\varepsilon > 0$  there exist  $A_1, A_2, \dots \in \mathcal{A}$  such that  $C \subset A_1 \cup A_2 \cup \dots$  and  $P(A_1 \cup A_2 \cup \dots) \leq P(C) + \varepsilon$ . In other words, there exists  $A \in \mathcal{A}_\sigma$  such that  $C \subset A$  and  $P(A) \leq P(C) + \varepsilon$ .

(b) For every  $C \in \mathcal{F}$  there exist  $B_1, B_2, \dots \in \mathcal{A}_\sigma$  such that  $B_1 \supset B_2 \supset \dots$ ,  $C \subset B_1 \cap B_2 \dots$ , and  $P(C) = P(B_1 \cap B_2 \dots)$ . In other words, there exists  $B \in \mathcal{A}_{\sigma\delta}$  such that  $C \subset B$  and  $P(B \setminus C) = 0$ .

(c) For every  $C \in \mathcal{F}$  and  $\varepsilon > 0$  there exists  $A \in \mathcal{A}$  such that  $P(A \setminus C) + P(C \setminus A) \leq \varepsilon$ .

See [1], Appendix A.3.

The proof of Theorem 4c1 is thus finished.

On the other hand, our approach to conditioning gives us another way of upgrading a martingale to a random function, at least when  $X = M(1) = \varphi(B(\cdot))$  for a bounded Borel function  $\varphi : C[0, 1] \rightarrow \mathbb{R}$ .<sup>1</sup> Namely,

$$(4c9) \quad M(t)(\omega_1) = \mathbb{E}(X | \mathcal{F}_t)(\omega_1) = \int \varphi \left( B(\cdot)(\omega_1) \bigsqcup^t B(\cdot)(\omega_2) \right) P(d\omega_2).$$

<sup>1</sup>I often write  $\varphi(B(\cdot))$  instead of  $\varphi(B(\cdot)|_{[0,1]})$ .

Several questions appear naturally. What happens if we change  $\varphi$  on a negligible (w.r.t. the Wiener measure) set? What happens if  $\varphi$  is not bounded? And, above all: is it a random *continuous* function?

**4c10 Exercise.** Let  $\varphi, \varphi_1, \varphi_2, \dots : C[0, 1] \rightarrow [-1, 1]$  be Borel functions such that  $\varphi_n(x) \uparrow \varphi(x)$  (as  $n \rightarrow \infty$ ) for every  $x \in C[0, 1]$ , and  $M, M_1, M_2, \dots$  random functions corresponding to  $\varphi, \varphi_1, \varphi_2, \dots$  according to (4c9). If each  $M_k$  is a.s. continuous, then  $M$  is a.s. continuous.

Prove it.

Due to 4c7 we have an algebra  $\mathcal{A} \subset \mathcal{B}_1$  such that

(a)  $\mathcal{A}$  generates  $\mathcal{B}_1$ ,

(b) for every  $A \in \mathcal{A}$  the random function  $M_A(t) = \mathbb{P}(A | \mathcal{F}_t)$  is a.s. continuous.

By 4c10,  $M_B(\cdot)$  is a.s. continuous for all  $B \in \mathcal{A}_\sigma$  and moreover, all  $B \in \mathcal{A}_{\sigma\delta}$ .

**4c11 Exercise.** If  $C \in \mathcal{B}_1$  is negligible in the sense that  $\mathbb{P}(B(\cdot) \in C) = 0$ , then  $\mathbb{P}(\forall t M_C(t) = 0) = 1$ .

Prove it.

Combining 4c11 and 4c8(b) we get the following.

**4c12 Theorem.** For every Borel set  $G \subset C[0, 1]$ , the following random function is a.s. continuous:

$$M_G(t) = \mathbb{P}(B(\cdot) \in G | \mathcal{F}_t),$$

that is,

$$M_G(t)(\omega_1) = \mathbb{P}\left(\left\{\omega_2 : B(\cdot)(\omega_1) \bigsqcup_{\cdot}^t B(\cdot)(\omega_2) \in G\right\}\right).$$

**4c13 Exercise.**  $\forall G \quad \mathbb{P}(M_G(\cdot) \text{ is continuous}) = 1$ , however,  $\mathbb{P}(\forall G M_G(\cdot) \text{ is continuous}) = 0$ ; here  $G$  runs over all Borel subsets of  $C[0, 1]$ .

Prove it.

**4c14 Exercise.** Let  $A \in \mathcal{F}_t$  for all  $t > 0$ ; then either  $\mathbb{P}(A) = 0$  or  $\mathbb{P}(A) = 1$ . ('Blumenthal's 0 – 1 law')

Prove it.

In other words,  $\mathcal{F}_{0+} = \mathcal{F}_0 \pmod{0}$ . See also [1], Sect. 7.2, (2.5).

**4c15 Exercise.** Give another proof to 2b6, using 4c14 (and not the distribution of  $T_x$ ).

**4c16 Exercise.** For every  $f : (0, \infty) \rightarrow (0, \infty)$  the random variable

$$\limsup_{t \rightarrow 0^+} \frac{B(t)}{f(t)} \in [0, \infty]$$

is degenerate (that is, equal a.s. to a constant).

Prove it.

**4c17 Exercise.** Generalize 4c14 for  $A \in \cap_{\varepsilon > 0} \mathcal{F}_{T+\varepsilon}$ , where  $T$  is a stopping time. That is, show that  $\mathcal{F}_{T+} = \mathcal{F}_T \pmod{0}$ .

**4c18 Exercise.** Generalize 4c10 to the case when  $\varphi_k : C[0, 1] \rightarrow [0, \infty)$  are bounded, while  $\varphi : C[0, 1] \rightarrow [0, \infty)$  need not be bounded, but  $\mathbb{E} \varphi(B(\cdot)) < \infty$ .

**4c19 Exercise.** Let  $\varphi : C[0, 1] \rightarrow \mathbb{R}$  be a Borel function such that  $\mathbb{E} \varphi(B(\cdot)) < \infty$ . Then the random function

$$M(t) = \mathbb{E}(\varphi(B(\cdot)) | \mathcal{F}_t)$$

is a.s. continuous.

Prove it.

## 4d Hints to exercises

4a12:  $\mathbb{E}(B^4(T \wedge n) - 6(T \wedge n)B^2(T \wedge n)) = -3(T \wedge n)^2$ ; use  $T$  as an integrable majorant in the left-hand side. . .

4a13:  $\mathbb{E}(e^{\lambda^2 T/2} \mathbb{1}_{[0, t]}(T)) \leq \frac{1}{\cos \lambda}$  for all  $\lambda \in [0, \pi/2)$  and  $t \in (0, \infty)$ .

4c4: recall 3c11.

4c5:  $\sum_n \mathbb{P}(\max_{[0, 1]} |M_n - M_{n+1}| \geq 1/n^2) \leq \sum_n n^2 \mathbb{E} |X_n - X_{n+1}| < \infty$ .

4c7: Consider the integral of the density; note that  $B(t_k) \neq a_k, B(t_k) \neq b_k$  a.s.

4c10: By (4c9) and the bounded convergence theorem,  $\mathbb{P}(\forall t \in [0, 1] M_n(t) \uparrow M(t)) = 1$ . On the other hand, similarly to 4c5,  $M_n(\cdot)$  converge in  $C[0, 1]$  to some  $M_\infty(\cdot)$  a.s., if  $\varphi_n \rightarrow \varphi$  fast enough (take a subsequence). Therefore  $M(\cdot) = M_\infty(\cdot)$  a.s.

4c11: 4c8(b) gives a negligible  $B \in \mathcal{A}_{\sigma\delta}$  such that  $C \subset B$ ; thus,  $M_C(\cdot) \leq M_B(\cdot)$ ,  $M_B(\cdot)$  is continuous, and  $\mathbb{E} M_B(t) = \mathbb{E} M_B(1) = 0$ .

4c13: Consider  $G = G_x = \{f : f(1) \leq x\}$  for all  $x \in \mathbb{R}$ .

4c14: consider  $\mathbb{P}(A | \mathcal{F}_t)$ .

4c19:  $\varphi = \varphi^+ - \varphi^-$ ; use 4c18, each  $\varphi_k$  being a linear combination of indicators.

## References

- [1] R. Durrett, *Probability: theory and examples*, 1996.

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