# Notes 29 : Brownian motion: martingale property

Math 733-734: Theory of Probability

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References:[Dur10, Section 8.5, 8.6, 8.8], [MP10, Section 2.4, 5.1, 5.3]. Recall:

**DEF 29.1 (Brownian motion)** The continuous-time stochastic process  $\{X(t)\}_{t\geq 0}$  is a standard Brownian motion if it has almost surely continuous paths and stationary independent increments such that X(s + t) - X(s) is Gaussian with mean 0 and variance t.

**DEF 29.2 (Filtration)** A filtration is a family  $\{\mathcal{F}(t) : t \ge 0\}$  of sub- $\sigma$ -fields such that  $\mathcal{F}(s) \subseteq \mathcal{F}(t)$  for all  $s \le t$ .

**DEF 29.3** Let  $\{B(t)\}$  be a BM. Then we denote

$$\mathcal{F}^0(t) = \sigma(B(s) : 0 \le s \le t).$$

Recall this is the smallest  $\sigma$ -field such that each  $B(s, \cdot)$ ,  $0 \le s \le t$ , is measurable. Moreover, we let

$$\mathcal{F}^+(t) = \bigcap_{s>t} \mathcal{F}^0(s).$$

Clearly  $\mathcal{F}^0(t) \subseteq \mathcal{F}^+(t)$ . The latter has the advantage of being right-continuous, that is,

$$\bigcap_{\varepsilon>0} \mathcal{F}^+(t+\varepsilon) = \mathcal{F}^+(t).$$

**DEF 29.4 (Germ field)** The germ  $\sigma$ -field is  $\mathcal{F}^+(0)$ .

**DEF 29.5 (Stopping time)** A RVT with values in  $[0, +\infty]$  is a stopping time with respect to the filtration  $\{\mathcal{F}(t)\}_{t\geq 0}$  if for all  $t\geq 0$ ,

$$\{T \le t\} \in \mathcal{F}(t).$$

**THM 29.6** If the filtration  $\{\mathcal{F}(t)\}_{t\geq 0}$  is right-continuous in the previous definition, then an equivalent definition is obtained by using a strict inequality. **DEF 29.7** Let T be a stopping time with respect to  $\{\mathcal{F}^+(t)\}_{t>0}$ . Then we let

$$\mathcal{F}^+(T) = \{A : A \cap \{T \le t\} \in \mathcal{F}^+(t), \forall t \ge 0\}.$$

**THM 29.8 (Strong Markov property)** Let  $\{B(t)\}_{t\geq 0}$  be a BM and T, an almost surely finite stopping time. Then the process

$$\{B(T+t) - B(T) : t \ge 0\},\$$

is a BM started at 0 independent of  $\mathcal{F}^+(T)$ .

### **1** Martingales

We first generalize MGs to continuous time.

**DEF 29.9 (Continuous-time martingale)** A real-valued SP  $\{X(t)\}_{t\geq 0}$  is a martingale with respect to a filtration  $\{\mathcal{F}(t)\}$  if it is adapted, that is,  $X(t) \in \mathcal{F}(t)$  for all  $t \geq 0$ , if  $E|X(t)| < +\infty$  for all  $t \geq 0$ , and if

$$\mathbb{E}[X(t) \,|\, \mathcal{F}(s)] = X(s),$$

almost surely, for all  $0 \le s \le t$ .

**EX 29.10** Let  $\{B(t)\}$  be a standard BM. Then

$$\mathbb{E}[B(t) | \mathcal{F}^+(s)] = \mathbb{E}[B(t) - B(s) | \mathcal{F}^+(s)] + B(s)$$
$$= \mathbb{E}[B(t) - B(s)] + B(s)$$
$$= B(s),$$

by the Markov property. Hence BM is a MG.

#### **2** Optional stopping theorem

We obtain a version of the optional stopping theorem for BMs.

**THM 29.11 (Optional stopping theorem)** Suppose  $\{X(t)\}_{t\geq 0}$  is a continuous MG, and  $0 \leq S \leq T$  are stopping times. If the process  $\{X(T \land t)\}_{t\geq 0}$  is dominated by an integrable RV X, then

$$\mathbb{E}[X(T) \,|\, \mathcal{F}(S)] = X(S),$$

almost surely.

**Proof:** We proceed by discretization. Fix N and consider the discrete-time MG

$$X_n = X(T \wedge n2^{-N})$$

and the stopping times

$$S'_N = \lfloor 2^N S \rfloor + 1$$

and

$$T'_N = \lfloor 2^N T \rfloor + 1$$

with respect to the filtration

$$\mathcal{G}_n = \mathcal{F}(n2^{-N}).$$

The discrete-time optional stopping theorem gives

$$\mathbb{E}[X_{T'_N} \mid \mathcal{G}_{S'_N}] = X_{S'_N},$$

which is equivalent to

$$\mathbb{E}[X(T \wedge 2^{-N}T'_N) | \mathcal{F}(2^{-N}S'_N)] = \mathbb{E}[X(T) | \mathcal{F}(2^{-N}S'_N)] = X(T \wedge 2^{-N}S'_N).$$

For  $A \in \mathcal{F}(S) \subseteq \mathcal{F}(2^{-N}S'_N)$ , by the definition of the conditional expectation and the dominated convergence theorem,

$$\mathbb{E}[X(T); A] = \lim_{N} \mathbb{E}[\mathbb{E}[X(T) \mid \mathcal{F}(2^{-N}S'_{N})]; A]$$
$$= \mathbb{E}[\lim_{N} X(T \land 2^{-N}S'_{N}); A]$$
$$= \mathbb{E}[X(S); A],$$

where we used continuity. The first line above is true for each N and therefore for the limit.

# **3** Applications

A typical application is Wald's lemma.

**THM 29.12 (Wald's lemma for BM)** Let  $\{B(t)\}$  be a standard BM and T a stopping time with respect to  $\{\mathcal{F}^+(t)\}$  such that either:

- 1.  $\mathbb{E}[T] < +\infty$ , or
- 2.  $\{B(t \wedge T)\}$  is dominated by an integrable RV.

Then  $\mathbb{E}[B(T)] = 0.$ 

**Proof:** The result under the second condition follows immediately from the optional stopping theorem with S = 0. We show that the first condition implies the second one.

Assume  $\mathbb{E}[T] < +\infty$ . Define

$$M_k = \max_{0 \le t \le 1} |B(t+k) - B(k)|,$$

and

$$M = \sum_{k=1}^{\lceil T \rceil} M_k,$$

and note that  $|B(t \wedge T)| \leq M$ .

Then

$$E[M] = \sum_{k} \mathbb{E}[\mathbb{1}\{T > k - 1\}M_{k}]$$
$$= \sum_{k} \mathbb{P}[T > k - 1]\mathbb{E}[M_{k}]$$
$$= \mathbb{E}[M_{0}]\mathbb{E}[T + 1] < +\infty$$

by our result on the maximum from the previous lecture.

We state without proof:

**THM 29.13 (Wald's second lemma)** Let  $\{B(t)\}$  be a standard BM and T a stopping time with respect to  $\{\mathcal{F}^+(t)\}$  such that  $\mathbb{E}[T] < +\infty$ . Then

$$\mathbb{E}[B(T)^2] = E[T].$$

**Proof:** The proof is based on the fact that  $B(t)^2 - t$  is a MG. Consider

$$T_n = \inf\{t \ge 0 : |B(t)| = n\},\$$

and take an appropriate limit. See [MP10] for details.

An immediate application of Wald's lemma gives:

**THM 29.14** *Let*  $\{B(t)\}$  *be a standard BM. For* a < 0 < b *let* 

$$T = \inf\{t \ge 0 : B(t) \in \{a, b\}\}.$$

Then

$$\mathbb{P}[B(T) = a] = \frac{b}{|a| + b}.$$

Moreover,

$$\mathbb{E}[T] = |a|b.$$

Note that the stopped process in bounded.

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### 4 Skorokhod embedding

BM can be used to embed any square-integrable RV.

**THM 29.15 (Skorokhod embedding)** Suppose  $\{B(t)\}_t$  is a standard BM and that X is a RV with  $\mathbb{E}[X] = 0$  and  $\mathbb{E}[X^2] < +\infty$ . Then there exists a stopping time T with respect to  $\{\mathcal{F}^+(t)\}_t$  such that B(T) has the law of X and  $\mathbb{E}[T] = \mathbb{E}[X^2]$ .

The proof uses a binary splitting MG:

**DEF 29.16** A  $\{X_n\}_n$  is binary splitting *if, whenever the event* 

$$A(x_0, \dots, x_n) = \{X_0 = x_0, \dots, X_n = x_n\},\$$

for some  $x_0, \ldots, x_n$ , has positive probability, then the RV  $X_{n+1}$  conditioned on  $A(x_0, \ldots, x_n)$  is supported on at most two values.

**LEM 29.17** Let X be a RV with  $\mathbb{E}[X^2] < +\infty$ . Then there is a binary splitting  $MG\{X_n\}_n$  such that  $X_n \to X$  almost surely and in  $\mathcal{L}^2$ .

Proof: The MG is defined recursively. Let

$$\mathcal{G}_0 = \{\emptyset, \Omega\},\$$

and

$$X_0 = \mathbb{E}[X].$$

For n > 0, we let

$$\xi_n = \begin{cases} 1, & \text{if } X \ge X_n \\ -1, & \text{if } X < X_n \end{cases}$$

and

$$\mathcal{G}_n = \sigma(\xi_0, \ldots, \xi_{n-1}),$$

and

$$X_n = \mathbb{E}[X \,|\, \mathcal{G}_n].$$

Then  $\{X_n\}_n$  is a binary splitting MG. It remains to prove the convergence claim. By (cJENSEN)

$$\mathbb{E}[X_n^2] \le \mathbb{E}[X^2],$$

so  $\{X_n\}_n$  is bounded in  $\mathcal{L}^2$  and we have by Lévy's upward theorem

$$X_n \to X_\infty = \mathbb{E}[X \,|\, \mathcal{G}_\infty],$$

almost surely and in  $\mathcal{L}^2$ , where

$$\mathcal{G}_{\infty} = \sigma\left(\bigcup_{i} \mathcal{G}_{i}\right).$$

We need to show that  $X = X_{\infty}$ .

CLAIM 29.18 Almost surely,

$$\lim_{n} \xi_n (X - X_{n+1}) = |X - X_{\infty}|.$$

We first finish the proof of the lemma. Note that

$$\mathbb{E}[\xi_n(X - X_{n+1})] = \mathbb{E}[\xi_n \mathbb{E}[X - X_{n+1} | \mathcal{G}_{n+1}]] = 0.$$

Since  $\{\xi_n(X - X_{n+1})\}_n$  is bounded in  $\mathcal{L}^2$ , the expectations converge and

$$\mathbb{E}|X - X_{\infty}| = 0.$$

Finally we prove the claim. If  $X = X_{\infty}$ , both sides are 0. If  $X < X_{\infty}$ , then for *n* large enough,  $X < X_n$  and  $\xi_n = -1$  by construction and the result holds. Similarly for the other case.

**Proof:**(of Theorem) Take a binary splitting MG as in the previous lemma. Since  $X_n$  conditioned on  $A(x_0, \ldots, x_{n-1})$  is supported on two values, we can use the stopping time from last time (for BM started at  $x_{n-1}$ ) and we get a sequence of stopping times

$$T_0 \le T_1 \le \dots \le T_n \le \dots \uparrow T$$

for some T such that

$$B(T_n) \sim X_n,$$

and

$$\mathbb{E}[T_n] = \mathbb{E}[B(T_n)^2].$$

By (MON) and  $\mathcal{L}^2$  convergence

$$\mathbb{E}[T] = \lim_{n} \mathbb{E}[T_n] = \lim_{n} \mathbb{E}[X_n^2] = \mathbb{E}[X].$$

By continuity of paths,

$$B(T_n) \to B(T),$$
 a.s.

and

$$B(T) \sim X.$$

# References

- [Dur10] Rick Durrett. *Probability: theory and examples*. Cambridge Series in Statistical and Probabilistic Mathematics. Cambridge University Press, Cambridge, 2010.
- [MP10] Peter Mörters and Yuval Peres. *Brownian motion*. Cambridge Series in Statistical and Probabilistic Mathematics. Cambridge University Press, Cambridge, 2010. With an appendix by Oded Schramm and Wendelin Werner.