

# Day 8 : The Strong Markov Property of Brownian Motions

Dain Kim

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# Today's Reading

[L] Chapter 2.4

# Review: Stopping Time

A *stopping time* with respect to a filtration  $\{\mathcal{F}_t, t \in [0, \infty)\}$  is a mapping  $T: \Omega \rightarrow [0, \infty]$  such that  $\{T \leq t\} \in \mathcal{F}_t$  for every  $t \in [0, \infty)$ .

# Review: Basic Properties of Brownian Motions

## Proposition

- 1)  $-B_t$  is also a Brownian motion
- 2) For every  $\lambda > 0$ ,  $B_t^\lambda = \frac{1}{\lambda} B_{\lambda^2 t}$  is a Brownian motion
- 3) For every  $s \geq 0$ ,  $B_t^{(s)} = B_{s+t} - B_s$  is a Brownian motion and is independent of  $\sigma(B_r, r \leq s)$ .

# The $\sigma$ -algebras of stopping times

Let  $(B_t)_{t \geq 0}$  be a Brownian motion. We use the notation

$$\mathcal{F}_t = \sigma\{B_s : s \leq t\}, \quad \mathcal{F}_\infty = \sigma\{B_s : s \geq 0\}.$$

Let  $T$  be a stopping time. The  *$\sigma$ -algebra of the past before  $T$*  is

$$\mathcal{F}_T = \{A \in \mathcal{F}_\infty : \forall t \geq 0, A \cap \{T \leq t\} \in \mathcal{F}_t\}.$$

Define the real random variable  $\mathbb{1}_{\{T < \infty\}} B_T(\omega)$  by

$$\mathbb{1}_{\{T < \infty\}} B_T(\omega) = \begin{cases} B_{T(\omega)}(\omega), & T(\omega) < \infty \\ 0, & T(\omega) = \infty \end{cases}.$$

# The $\sigma$ -algebras of stopping times

The random variable  $\mathbb{1}_{\{T < \infty\}} B_T(\omega)$  is  $\mathcal{F}_T$ -measurable:

$$\begin{aligned}\mathbb{1}_{\{T < \infty\}} B_T &= \lim_{n \rightarrow \infty} \sum_{i=0}^{\infty} \mathbb{1}_{\{i2^{-n} \leq T \leq (i+1)2^{-n}\}} B_{i2^{-n}} \\ &= \lim_{n \rightarrow \infty} \sum_{i=0}^{\infty} \mathbb{1}_{\{T \leq (i+1)2^{-n}\}} \mathbb{1}_{\{i2^{-n} \leq T\}} B_{i2^{-n}}\end{aligned}$$

and for any interval  $A$  not containing 0,

$$\{B_s \mathbb{1}_{\{s \leq T\}} \in A\} \cap \{T \leq t\} = \begin{cases} \emptyset, & t < s \\ \{B_s \in A\} \cap \{s \leq T \leq t\}, & t \geq s. \end{cases}$$

# Strong Markov Property

## Theorem (Strong Markov Property)

Let  $T$  be a stopping time such that  $\mathbf{P}(T < \infty) > 0$ . Set, for  $t \geq 0$ ,

$$B_t^{(T)} = \mathbb{1}_{\{T < \infty\}}(B_{T+t} - B_T).$$

Then under the probability measure  $\mathbf{P}(\cdot | T < \infty)$ , the process  $(B_t^{(T)})$ ,  $t \geq 0$ , is a Brownian motion independent of  $\mathcal{F}_T$ .

*Proof.* We prove for the case  $T < \infty$ , a.s. Fix  $A \in \mathcal{F}_T$  and  $0 \leq t_1 \leq \dots \leq t_p$ . Let  $F: \mathbb{R}^p \rightarrow \mathbb{R}^+$  be a bounded continuous function. We show that

$$\mathbf{E}[\mathbb{1}_A F(B_{t_1}^{(T)}, \dots, B_{t_p}^{(T)})] = P(A) \mathbf{E}[F(B_{t_1}, \dots, B_{t_p})].$$

## Strong Markov Property

Denote by  $[t]_n$  the smallest real of the form  $k2^{-n}$  at least  $t$ . By convention, we let  $[\infty]_n = \infty$ . By the continuity of  $F$ ,

$$F(B_{t_1}^{(T)}, \dots, B_{t_p}^{(T)}) = \lim_{n \rightarrow \infty} F(B_{t_1}^{([T]_n)}, \dots, B_{t_p}^{([T]_n)}),$$

so by Lebesgue Dominated Convergence Theorem and simple Markov property,

$$\begin{aligned} \mathbf{E}[\mathbb{1}_A F(B_{t_1}^{(T)}, \dots, B_{t_p}^{(T)})] &= \lim_{n \rightarrow \infty} \mathbf{E}[\mathbb{1}_A F(B_{t_1}^{([T]_n)}, \dots, B_{t_p}^{([T]_n)})] \\ &= \lim_{n \rightarrow \infty} \sum_{k=0}^{\infty} \mathbf{E}[\mathbb{1}_A \mathbb{1}_{\{(k-1)2^{-n} \leq T \leq k2^{-n}\}} \\ &\quad \cdot F(B_{k2^{-n}+t_1} - B_{k2^{-n}}, \dots, B_{k2^{-n}+t_p} - B_{k2^{-n}})] \\ &= \lim_{n \rightarrow \infty} \sum_{k=0}^{\infty} \mathbf{P}(A \cap \{(k-1)2^{-n} < T \leq k2^{-n}\}) \mathbf{E}[F(B_{t_1}, \dots, B_{t_p})] \\ &= \mathbf{P}(A) \mathbf{E}[F(B_{t_1}, \dots, B_{t_p})]. \end{aligned}$$



# Strong Markov Property

If  $\mathbf{P}(T = \infty) > 0$ , we may substitute  $A$  by  $A \cap \{T < \infty\}$  and the desired result follows.

Once we have

$$\mathbf{E}[\mathbb{1}_A F(B_{t_1}^{(T)}, \dots, B_{t_p}^{(T)})] = P(A) \mathbf{E}[F(B_{t_1}, \dots, B_{t_p})],$$

for  $A = \Omega$ , since  $B^{(T)}$  has the same finite-dimensional marginal distributions as  $B$ , it is also a Brownian motion. Then the above equation shows that  $B^{(T)}$  is independent of  $\mathcal{F}_T$ . □

# Reflection Principle

## Theorem

For every  $t > 0$ , let  $S_t = \sup_{s \leq t} B_s$ . Then if  $a \geq 0$  and  $b \in (-\infty, a]$ , we have

$$\mathbf{P}(S_t \geq a, B_t \leq b) = \mathbf{P}(B_t \geq 2a - b).$$

In particular,  $S_t$  has the same distribution as  $|B_t|$ .

*Proof.* Define a stopping time

$$T_a = \inf\{t \geq 0 : B_t = a\}.$$

Note that  $T_a < \infty$ . We have

$$\mathbf{P}(S_t \geq a, B_t \leq b) = \mathbf{P}(T_a \leq t, B_t \leq b) = \mathbf{P}(T_a \leq t, B_{t-T_a}^{(T_a)} \leq b - a).$$

## Reflection Principle

Write  $B' = B^{(T_a)}$  and note that  $B'$  is a Brownian motion independent of  $\mathcal{F}_{T_a}$ . Note also that  $-B'$  is a Brownian motion as well, and that  $(T_a, B')$  and  $(T_a, -B')$  have the same distribution. Define

$$H = \{(s, w) \in \mathbb{R}^+ \times C(\mathbb{R}^+, \mathbb{R}) : s \leq t, w(t-s) \leq b-a\}.$$

Then

$$\begin{aligned}\mathbf{P}((T_a, B') \in H) &= \mathbf{P}((T_a, -B') \in H) \\ &= \mathbf{P}(T_a \leq t, -B_{t-T_a}^{(T_a)} \leq b-a) \\ &= \mathbf{P}(T_a \leq t, B_t \geq 2a-b) \\ &= \mathbf{P}(B_t \geq 2a-b).\end{aligned}$$

Finally,

$$\begin{aligned}\mathbf{P}(S_t \geq a) &= \mathbf{P}(S_t \geq a, B_t \geq a) + \mathbf{P}(S_t \geq a, B_t \leq a) \\ &= 2\mathbf{P}(B_t \geq a) = \mathbf{P}(|B_t| \geq a).\end{aligned}$$

So  $S_t$  has the same distribution as  $|B_t|$ .



# Distribution of $T_a$

## Corollary

For every  $a > 0$ ,  $T_a$  has the same distribution as  $a^2/B_1^2$  and has density

$$f(t) = \frac{a}{\sqrt{2\pi t^3}} \exp\left(-\frac{a^2}{2t}\right) \mathbb{1}_{\{t>0\}}.$$

*Proof.* For every  $t \geq 0$ ,

$$\begin{aligned} \mathbf{P}(T_a \leq t) &= \mathbf{P}(S_t \geq a) \\ &= \mathbf{P}(|B_t| \geq a) \\ &= \mathbf{P}(B_t^2 \geq a^2) \\ &= \mathbf{P}(tB_1^2 \geq a^2) \\ &= \mathbf{P}\left(\frac{a^2}{B_1^2} \leq t\right). \end{aligned}$$

# Real Brownian Motions

If  $Z$  is a real random variable, a stochastic process  $(X_t)$  is a *real Brownian motion* if

$$X_t = Z + B_t$$

where  $B$  is a real Brownian motion started from 0 and independent of  $Z$ .

A stochastic process  $B_t = (B_t^1, \dots, B_t^d)$  with values in  $\mathbb{R}^d$  is a *d-dimensional Brownian motion* started from 0 if its components  $B^1, \dots, B^d$  are independent real Brownian motions started from 0.