## Math 280C, Spring 2005

## Strong Markov Property

Notation is that introduced in class.

**Theorem.** (Strong Markov Property of Brownian Motion). If T is an  $(\mathcal{F}_{t+})$ -stopping time and  $\mu$  is any initial distribution, then

(1) 
$$\mathbf{E}_{\mu}[F \circ \theta_T | \mathcal{F}_{T+}] = \mathbf{E}_{B(T)}(F), \qquad \mathbf{P}_{\mu}\text{-a.s. on } \{T < \infty\}$$

for each  $F \in b\mathcal{F}$ . [Here B(T) is alternate notation for  $B_T$ .]

*Proof.* The right side of (1) is the composition of the  $\mathcal{F}_{T+}$ -measurable map  $\omega \mapsto B_{T(\omega)}(\omega)$  with the Borel measurable map  $x \mapsto \mathbf{E}_x[F]$ , and is therefore  $\mathcal{F}_{T+}$ -measurable. It remains to show that

(2) 
$$\mathbf{E}_{\mu}[H \cdot F \circ \theta_{T}] = \mathbf{E}_{\mu}\left[H \cdot \mathbf{E}_{B(T)}[F]\right]$$

for every  $H \in b\mathcal{F}_{T+}$ . In view of the monotone class theorem for functions, it suffices to prove (2) for F of the special form

$$F(\omega) = \prod_{k=1}^{n} f_k(B_{t_k}(\omega))$$

where n is a positive integer,  $0 < t_1 < t_2 < \cdots < t_n$ , and each  $f_k : \mathbf{R} \to \mathbf{R}$  is a bounded continuous function. It has been proved in class that for such an F, the function  $x \mapsto \mathbf{E}_x[F]$ is (bounded and) continuous. We proceed by approximating the stopping time T as follows:

$$T_n(\omega) = \begin{cases} (k+1)2^{-n}, & \text{if } k2^{-n} \le T(\omega) < (k+1)2^{-n}, \ k = 0, 1, 2, \dots; \\ +\infty, & \text{if } T(\omega) = +\infty. \end{cases}$$

Then, for t > 0,

$$\{T_n \le t\} = \bigcup_{k:(k+1)2^{-n} \le t} \{k2^{-n} \le T < (k+1)2^{-n}\} \in \mathcal{F}_t,$$

since  $\{T < (k+1)2^{-n}\} \in \mathcal{F}_{(k+1)2^{-n}+}$  because T is an  $(\mathcal{F}_{t+})$ -stopping time. Thus  $T_n$  is a stopping time, and it is clear that  $T_n(\omega)$  decreases to  $T(\omega)$  for each  $\omega$ . Consequently, because  $t \mapsto B_t(\omega)$  is continuous,  $\lim_{n\to\infty} B_{T_n}(\omega) = B_T(\omega)$ , and so

$$\lim_{n \to \infty} \mathbf{E}_{B_{T_n(\omega)}(\omega)}[F] = \mathbf{E}_{B_{T(\omega)}(\omega)}[F].$$

Therefore, noting that  $\{T < \infty\} = \{T_n < \infty\},\$ 

(3)  
$$\mathbf{E}_{\mu} \left[ H \cdot \mathbf{E}_{B(T)}[F]; T < \infty \right] = \mathbf{E}_{\mu} \left[ H \cdot \lim_{n} \mathbf{E}_{B(T_{n})}[F]; T_{n} < \infty \right]$$
$$= \lim_{n} \mathbf{E}_{\mu} \left[ H \cdot \mathbf{E}_{B(T_{n})}[F]; T_{n} < \infty \right]$$

But, by the law of total probability and Fubini's theorem,

(4)  

$$\mathbf{E}_{\mu} \left[ H \cdot \mathbf{E}_{B(T_n)}[F]; T_n < \infty \right] = \sum_{k=0}^{\infty} \mathbf{E}_{\mu} \left[ H \cdot \mathbf{E}_{B((k+1)2^{-n})}[F]; T_n = (k+1)2^{-n} \right]$$

$$= \sum_{k=0}^{\infty} \mathbf{E}_{\mu} \left[ H \cdot F \circ \theta_{(k+1)2^{-n}}; T_n = (k+1)2^{-n} \right]$$

$$= \mathbf{E}_{\mu} \left[ H \cdot F \circ \theta_{T_n}; T_n < \infty \right].$$

The second equality above results from the simple Markov property because (i)

$$\{T_n = (k+1)2^{-n}\} = \{k2^{-n} \le T < (k+1)2^{-n}\} \in \mathcal{F}_{(k+1)2^{-n}}\}$$

and (ii)  $H \cdot 1_{\{k2^{-n} \leq T < (k+1)2^{-n}\}} \in \mathcal{F}_{(k+1)2^{-n}}$  as H is  $\mathcal{F}_{T+}$ -measurable. On the other hand, because of the special choice of F, the map  $t \mapsto F(\theta_t \omega)$  is continuous, so  $\lim_n F(\theta_{T_n(\omega)}\omega) = F(\theta_{T(\omega)}\omega)$ , at least for  $\omega$  in  $\{T < \infty\}$ . It now follows from the dominated convergence theorem that

(5) 
$$\lim_{n} \mathbf{E}_{\mu}[H \cdot F \circ \theta_{T_{n}}; T_{n} < \infty] = \mathbf{E}_{\mu}[H \cdot F \circ \theta_{T}; T < \infty].$$

Combining (3), (4), and (5), we obtain (2).

The following reformulation of the strong Markov property for Brownian motion is often convenient. For simplicity I assume that T is finite.

**Corollary.** If T is an  $(\mathcal{F}_{t+})$ -stopping time and  $\mu$  is any initial distribution such that  $\mathbf{P}_{\mu}[T < \infty] = 1$ , then the process  $B^{(T)}$  defined by

$$B_t^{(T)} := B_{T+t} - B_T, \quad t \ge 0,$$

has the same distribution as Brownian motion started at 0, and is independent of  $\mathcal{F}_{T+}$ .

*Proof.* For this proof only let  $\tau_x : \Omega \to \Omega$  be defined by  $(\tau_x \omega)(t) := \omega(t) - x$ . Then  $B^{(T)}(\omega)$  is equal to  $\theta_{T(\omega)} = \tau_x [\theta_{T(\omega)} \omega]$  when x is set equal to  $B_T(\omega)$ . Consequently, if  $H \in b\mathcal{F}_{T+}$  and  $F \in b\mathcal{F}$ ,

$$\begin{aligned} \mathbf{E}_{\mu}[H \cdot F(B^{(T)})] &= \mathbf{E}_{\mu}[H \cdot F(\tau_{B(T)}\theta_{T})] \\ &= \mathbf{E}_{\mu}[H \cdot \mathbf{E}_{\mu}[F(\tau_{B(T)}\theta_{T})|\mathcal{F}_{T+}] = \mathbf{E}_{\mu}[H \cdot f(B_{T})], \end{aligned}$$

where  $f(x) = \mathbf{E}_x[F(\tau_x)]$ . But Brownian motion started at x is equal in distribution to Brownian motion started at 0 and then translated by x. This translation is cancelled by  $\tau_x$ , so  $f(x) = \mathbf{E}_x[F(\tau_x)] = \mathbf{E}_0[F]$ . Thus,

$$\mathbf{E}_{\mu}[H \cdot F(B^{(T)})] = \mathbf{E}_{\mu}[H \cdot \mathbf{E}_0[F]] = \mathbf{E}_{\mu}[H] \cdot \mathbf{E}_0[F],$$

which proves both assertions of the theorem.  $\Box$ 

**Example.** The Brownian transition density  $p_t(x, y)$  induces an operator  $P_t$  on functions as follows: If  $f : \mathbf{R} \to \mathbf{R}$  is Borel measurable and bounded (or positive) then we define  $P_t f$  as the function

$$P_t f(x) := \int_{\mathbf{R}} p_t(x, y) f(y) \, dy, \qquad x \in \mathbf{R}.$$

Evidently,  $||P_t f||_{\infty} \leq ||f||_{\infty}$ , and (Scheffe's lemma again!)  $x \mapsto P_t f(x)$  is a continuous function if f is bounded.

Let T be a stopping time, and let us compute the conditional expectation  $\mathbf{E}_{\mu}[f(B_t)|\mathcal{F}_{T+}]$ , where t > 0 is fixed. Observe that if  $T(\omega) < t$  then

$$f(B_t(\omega)) = g(\theta_{T(\omega)}, T(\omega)),$$

where  $g(\omega', s) := f(w'(t-s))$ . Since T is  $\mathcal{F}_{T+}$ -measurable, we can use (1) to compute

$$\mathbf{E}_{\mu}[f(B_t)|\mathcal{F}_{T+}] = \mathbf{1}_{\{T \ge t\}}f(B_t) + \mathbf{1}_{\{T < t\}}h(B_T, T),$$

where, for  $0 \leq s < t$ ,

$$h(x,s) := \mathbf{E}_x[g(\cdot,s)] = \mathbf{E}_x[f(B_{t-s})] = P_{t-s}f(x).$$

Thus,

$$\mathbf{E}_{\mu}[f(B_t)|\mathcal{F}_{T+}] = \mathbf{1}_{\{T \ge t\}}f(B_t) + \mathbf{1}_{\{T < t\}}P_{t-T}f(B_T).$$