

Lecture 22

Wed Apr 10/2024

Last time

▷ Existence

Today

▷ The Markov Property
▷ Consequences

The Markov Property

Now that we are convinced of the existence of B_t , we can study its properties. We didn't cover it, but it is relatively easy to show that B_t is nowhere differentiable (Theorem 7.16 Durrett).

Today, we study the Markov Property, which intuitively reads: "For $s \geq 0$ $B(s+t) - B(s)$ is a Brownian motion starting at $B(s)$ and ind. of what happened in the past."

We consider two slightly different filtrations:

$$\mathcal{F}_t^0 = \sigma(B_s : s \leq t)$$

$$\mathcal{F}_t^+ = \bigcap_{u \geq t} \mathcal{F}_u^0 \quad (\text{germ field})$$

Notice that for $t < s$, $\mathcal{F}_t^0 \subseteq \mathcal{F}_t^+ \leq \mathcal{F}_s$.

Indeed \mathcal{F}_t^+ is right continuous, i.e.

$$\bigcap_{s>t} \mathcal{F}_s^+ = \bigcap_{s>t} \bigcap_{u>s} \mathcal{F}_u^0 = \bigcap_{u>t} \mathcal{F}_u^0 = \mathcal{F}_t^+.$$

Intuitively, \mathcal{F}_s^+ sees infinitesimally into the future.

Example: The random variable

$$\limsup_{s \downarrow t} \frac{B_s - B_t}{s - t}$$

is measurable w.r.t. \mathcal{F}_t^+ , but not w.r.t. \mathcal{F}_t^0 . \square

We shall see that the distinction between the two is rather negligible.

For any $x \in \mathbb{R}$, we let P_x be the prob. dist. of Brownian Motion with $B_0 = x$.

Recall that we constructed BM via

$$B_t(w) = w(t).$$

Def: For any $s \geq 0$, define the shift operator $\theta_s : \Omega \rightarrow \Omega$ by

$$\theta_s(w)(t) = w(s+t).$$

\square

Theorem (Markov property): If $s \geq 0$ is

bounded and \mathcal{F} measurable, then for all $x \in \mathbb{R}$

$$\mathbb{E}_x(Y \circ \theta_s | \mathcal{F}_s^+) = \mathbb{E}_{B_s} Y$$

where the right-hand side is the function $\varphi(x) = \mathbb{E}_x Y$ evaluated at B_s . \dashv

The proof is very similar to our proof of the Markov Property in Lecture 15, where we established the result for simple functions and leveraged the $T\!-\lambda$ Theorem and the Monotone class Theorem to extend. We defer the interested student to Theorem 7.2.1 in Durrett).

Instead we focus on consequences.

Consequences

First we show a couple of intermediate results.

Theorem: If Z is measurable w.r.t. \mathcal{Z} and bounded then $\forall s \geq 0$ and $x \in \mathbb{R}^d$

$$\mathbb{E}_x(Z | \mathcal{F}_s^+) = \mathbb{E}_x(Z | \mathcal{F}_s^\circ).$$

Proof: Thanks to the Monotone Class Theorem it suffices to show this for

$$Z = \prod_{m=1}^n f_m(B_{t_m})$$

with f_m measurable and $0 < t_1 < \dots < t_n$.
 Notice that we can split Z into two: let $\Delta = \{m \leq n : t_m \leq s\}$

$$Z = \left(\prod_{m \in \Delta} f_m(B_{t_m}) \right) \left(\prod_{m \in \Delta^c} f_m(B_{t_m}) \right)$$

$x \in \mathcal{F}_s^\circ$ w

further we can write $w = Y \circ \theta_s$
 (simply take $Y = \prod_{m \in \Delta^c} f_m(B_{t_{m-s}})$). Thus,

$$\begin{aligned} \mathbb{E}[Z | \mathcal{F}_s^+] &= \mathbb{E}[X(Y \circ \theta_s) | \mathcal{F}_s^+] \\ &= X \mathbb{E}[Y \circ \theta_s | \mathcal{F}_s^+] \end{aligned}$$

Markov $\rightarrow = X \mathbb{E}_{B_s} Y = (\star)$

Further notice that $\mathbb{E}_{B_s} Y \in \mathcal{F}_s^\circ$.

Recall that if $\mathcal{F} \subseteq G$ and $\mathbb{E}[\alpha | G] \in \mathcal{F}$
 $\Rightarrow \mathbb{E}[\alpha | G] = \mathbb{E}[\alpha | \mathcal{F}]$ (Lecture 8)

Thus

$$\begin{aligned} (\star) &= X \mathbb{E}[Y \circ \theta_s | \mathcal{F}_s^+] \\ &= \mathbb{E}[Z | \mathcal{F}_s^+] \end{aligned}$$

This theorem implies that if $Z \in \mathcal{F}_s^+$, then \square

$Z = \mathbb{E}_x[Z | \mathcal{F}_s^0] \in \mathcal{F}_s^0$, so \mathcal{F}_s^+ and \mathcal{F}_s^0 are the same up to zero measure sets. This has beautiful consequences.

Theorem (Blumenthal's 0-1 law) If $A \in \mathcal{F}_0^+$, then $\forall x \in \mathbb{R}$,

$$\mathbb{P}_x(A) \in \{0, 1\}.$$

Proof: By the previous theorem

$$\mathbb{1}_A = \mathbb{E}_x[\mathbb{1}_A | \mathcal{F}_0^+] = \mathbb{E}_x[\mathbb{1}_A | \mathcal{F}_0^0] = \mathbb{P}_x(A)$$

almost surely. \square

This is particularly useful to understand the behavior of BM locally.

Corollary: Consider $\tau_+ = \inf \{t > 0 : B_t > 0\}$, then $\mathbb{P}_0(\tau_+ = 0) = 1$.

Proof: Since the normal dist is symmetric around 0

$$\mathbb{P}_0(\tau_+ \leq t) \geq \mathbb{P}_0(B_t > 0) = 1/2.$$

Moreover, letting $t \downarrow 0$

$$\mathbb{P}_0(\tau_+ = 0) = \lim_{t \downarrow 0} \mathbb{P}(\tau_+ \leq t) \geq 1/2.$$

So by Blumenthal's 0-1 law, $\mathbb{P}_0(\tau_+ = 0) = 1$. \square

Notice that the same conclusion holds for $\tau_- = \inf \{t > 0 : B_t < 0\}$. Further B_t is continuous so we can easily derive:

Corollary: Consider $\tau_0 = \inf \{t > 0 : B_t = 0\}$, then $P_0(\tau_0 = 0) = 1$.

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Blumenthal's 0-1 law allows us to understand questions as $t \rightarrow 0$, but how about when $t \uparrow \infty$? In turn, we can reverse BM and understand asymptopia via $t \downarrow 0$.

Theorem: If B_t is a BM starting at 0, then so is the process defined by $X_0 = 0$ and $X_t = t B_{1/t}$ as $t \rightarrow 0$.

Next class we will prove this result, but for now let us cover a relevant consequence. Let

$$\mathcal{F}_t^1 = \sigma(B_s : s > t)$$

The future at t .

$$\mathcal{T} = \bigcap_{t \geq 0} \mathcal{F}_t^1$$

tail σ -algebra.

Theorem: Let $A \in \mathcal{I}$, then either $\Pr_x(A) = 0 \ \forall x$
or $\Pr_x(A) = 1 \ \forall x$.

TO BE CONTINUED . . .