

# Math 275D Lecture 7 Notes

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## 1 Equality of $\sigma$ -Fields and Brownian Inversion

### 1.1 $\mathcal{F}_s^0$ and $\mathcal{F}_s^+$ are almost the same

Last time, we showed the Markov property for Brownian motion:

$$\mathbb{E}_x[Y \circ \theta_s \mid \mathcal{F}_s^+] = \mathbb{E}_{B(s)}[Y].$$

This is actually a bit stronger than a Markov property, since it uses  $\mathcal{F}_s^+$ , not  $\mathcal{F}_s^0$ .

**Proposition 1.1.**  $\mathcal{F}_s^+ = \mathcal{F}_s^0$  modulo null sets.

*Proof.* We claim that  $\mathbb{E}_x[Y \circ \theta_s \mid \mathcal{F}_s^0] = \mathbb{E}_{B(s)}[Y]$ . The right hand side is  $\mathcal{F}_s^0$ -measurable, and the Markov property shows that it satisfies the definition of the conditional expectation. Then for any  $\mathcal{F}$ -measurable  $Z$ ,

$$\mathbb{E}[Z \mid \mathcal{F}_s^+] = \mathbb{E}[Z \mid \mathcal{F}_s^0].$$

This follows from the monotone class argument, which tells us we only need to show it for  $Z = \prod_{i=1}^k f(B(t_i))$ . We can assume that  $t_1 < t_2 < \dots < t_m \leq s$  and  $t_{m+1} > \dots > t_k > s$ . Then  $Z = X \cdot (Y \circ \theta_s)$ , where  $X = \prod_{i=1}^m f(B(t_i))$  and  $Y = \prod_{j=1}^{k-m} f(B(t_j - s))$ . Then  $X$  is  $\mathcal{F}_s^0$ -measurable, so

$$\begin{aligned} \mathbb{E}[Z \mid \mathcal{F}_s^0] &= \mathbb{E}[X(Y \circ \theta_s) \mid \mathcal{F}_s^0] \\ &= X \mathbb{E}[Y \circ \theta_s \mid \mathcal{F}_s^0] \\ &= X \mathbb{E}[Y \circ \theta_s \mid \mathcal{F}_s^+] \\ &= \mathbb{E}[X(Y \circ \theta_s) \mid \mathcal{F}_s^+] \\ &= \mathbb{E}[Z \mid \mathcal{F}_s^+]. \end{aligned}$$

□

## 1.2 $tB(1/t)$ is a Brownian motion

Last time, we mentioned the following property.

**Proposition 1.2.** *Let  $Y(t) = tB(1/t)$ . Then  $Y(t)$  is a Brownian motion.*

*Proof.*  $(Y(t_1), \dots, Y(t_n))$  is a Gaussian random vector. So to prove that  $Y(t_2) - Y(t_1) \perp Y(t_4) - Y(t_3)$ , for example, we only need to prove that they are uncorrelated. It now remains to show that we can define  $Y(0) = 0$ .  $\square$

We need to know that  $\lim_{t \rightarrow \infty} \frac{B(t)}{t} = 0$  a.s.

**Proposition 1.3.**  $\lim_{n \rightarrow \infty} \frac{B(n)}{n} = 0$ . a.s.

*Proof.*  $B(n) = \sum_{i=1}^n X_i$ , where  $X_i = B(i) - B(i-1)$ . The  $X_i$  are iid with  $N(0,1)$  distribution, so the strong law of large numbers gives the result.  $\square$

What if we want to find the following probability:

$$\mathbb{P} \left( \max_{m \in [n, n+1]} \frac{|B(m) - B(n)|}{n^{2/3}} \geq 1 \right).$$

We can try looking at the following:

$$\mathbb{P} \left( \max_{m \in n + \mathbb{Q}_{[0,1]}^{(k)}} \frac{|B(m) - B(n)|}{n^{2/3}} \geq 1 \right),$$

where  $\mathbb{Q}_{[0,1]}^{(k)} = \{\ell/k \in [0, 1] : k, \ell \in \mathbb{Z}\}$

You could try a union bound:

$$\leq \sum_{\ell=1}^k \mathbb{P} \left( \left| \frac{B(n + \ell/k) - B(n)}{n^{2/3}} \right| \geq 1 \right).$$

However, this probability does not decay with  $k$ , and we have to add together  $k$  of them. So this will not work.

Let  $X_\ell = B(n + \ell/k) - B(n + (\ell-1)/k)$ , and let  $Y_\ell = \sum_{\ell' < \ell} X_{\ell'}$ . The  $X_\ell$ s are iid, so  $Y_\ell$  is a Markov chain and a Martingale. We have the general inequality:

$$\mathbb{P} \left( \max_{1 \leq \ell \leq k} |Y_\ell| \geq a \right) \leq \frac{\mathbb{E}[Y_k^2]}{a^2}$$

This gives us

$$\mathbb{P} \left( \max_{1 \leq \ell \leq k} |Y_\ell| \geq n^{2/3} \right) \leq \frac{1}{n^{4/3}}.$$

Let  $k = 2^{\tilde{k}}$ , and define the event  $A_{\tilde{k}} = \{\max_{m \in n + \mathbb{Q}_k} |B(m) - B(n)| \leq n^{2/3}\}$ . Then  $A_{\tilde{k}} \supseteq A_{\tilde{k}+1}$ . We also have that

$$\mathbb{P}(A_{\tilde{k}+1}) \geq 1 - n^{-4/3}.$$

So

$$\mathbb{P}\left(\bigcup_{\tilde{k}} A_{\tilde{k}}\right) \geq 1 - n^{-4/3}.$$

This gives us

$$\mathbb{P}\left(\max_{m \in n + [0,1]} \left| \frac{B(m) - B(n)}{n^{2/3}} \right| \geq 1\right) \leq n^{-4/3}.$$

If we call this event  $C_n$ , we get that

$$\mathbb{P}(C_n \text{ i.o.}) = 0$$

by the first Borel-Cantelli lemma.

Together with the fact that  $\lim_n \frac{B(n)}{n} \rightarrow 0$ , we get:

**Proposition 1.4.** *With probability 1,*

$$\lim_{t \rightarrow \infty} \frac{B(t)}{t} \rightarrow 0.$$

**Corollary 1.1.** *The tail  $\sigma$ -field of Brownian motion is trivial.*

*Proof.* This follows from the fact that  $\mathcal{F}_0^0 \equiv \mathcal{F}_0^+$ , while  $\mathcal{F}_0^0$  is trivial. □