# Statistics 212: Lecture 7 (Feb 19, 2025)

# Construction of Brownian Motion

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# 1 Definition

**Definition 1.1.** A Brownian Motion on  $t \in [0, 1]$ , is a random continuous function  $B: [0, 1] \to \mathbb{R}$  such that:

- (a)  $B_t B_s \sim \mathcal{N}(0, t s)$  if  $t \ge s$
- (b) if  $t_1 \le t_2 \le ... \le t_k$ , then  $(B_{t_2} B_{t_1}, B_{t_3} B_{t_2}, ..., B_{t_k} B_{t_{k-1}})$  are independent.

### 2 Questions Surrounding Definition

- (a) **Existence:** Does such a function in Definition 1.1 exist as a  $\mathscr{C}([0, 1])$ -valued random variable?
- (b) **Uniqueness:** Is such a random function *B* unique?
- (c) **Uncountable set:** How do we handle the uncountability of [0, 1]?

To state these questions formally, we should have some probability measure  $\mu$  on  $\mathscr{C}([0,1])$  such that with  $\varphi_t : \mathscr{C}([0,1]) \to \mathbb{R}$  given by  $\varphi_t(f) = f(t)$ , we should have  $\text{Law}(\varphi_t(B) - \varphi_s(B)) \sim \mathscr{N}(0, t - s)$ , etc, where  $\varphi_t(B) = B_t$  and  $\varphi_s(B) = B_s$ .

**Initial Attempt:** One natural approach is to construct Brownian Motion from finite-dimensional distributions.

The followings are two thoughts that we may have when attempting to construct a Brownian Motion.

- Given  $t_1, t_2, ..., t_n$ , the defining property 2 of Definition 1.1 tells us the joint law of  $(B_{t_1}, ..., B_{t_k})$
- We should check if these distributions are consistent, i.e., if forgetting  $t_j$ , we can still recover correct law on  $(B_{t_1}, ..., B_{t_{j-1}}, B_{t_{j+1}}, ..., B_{t_k})$

**Theorem 2.1** (Kolmogorov Extension (or Consistency) Theorem). *There always exists a probability measure*  $\tilde{\mu}$  on  $\mathbb{R}^{[0,1]} \equiv \operatorname{func}([0,1] \to \mathbb{R}; i.e., the set of all functions from the interval <math>[0,1]$  to  $\mathbb{R}$ ) which has all these finitedimensional laws in the defining property 2 in Definition 1.1, given that these distributions are consistent. ⇒ However,  $\tilde{\mu}$  is not unique. For example, we can choose  $u \sim \text{Unif}(0, 1)$ . We can start with  $\tilde{\mu}$  but force  $B_u = 100$ . The stochastic process still obeys these defining properties 2 and the consistency property.

**Problem**:  $\sigma$ -algebra on  $\mathbb{R}^{[0,1]}$  is generated by the evaluation mapping  $\varphi_t$ . In other words sets of the form  $\{f \in \mathbb{R}^{[0,1]} | f(t) \in (a, b)\}$  are measurable, and the  $\sigma$ -algebra is the one generated by these. Continuity of f is not even a measurable property.

## **3** Construction of Brownian Motion

#### 3.1 Constructing a Sequence

To construct a Brownian Motion, we construct a sequence of piecewise linear interpolation. Specifically, we split the [0, 1] interval *k* times into k + 1 equal intervals. For the trivial case where k = 0, we have

$$B_t^0 = \begin{cases} 0 & t = 0, \\ z_0 & t = 1, \\ \text{linear interpolation} & \text{otherwise} \end{cases}$$

where  $z_0 \sim \mathcal{N}(0, 1)$ .



Figure 1: Case of K = 0 (Left) and K = 1 (Right)

Formally, we define  $B_t^k$  by

$$B_{j/2^{k}}^{k+1} = B_{j/2^{k}}^{k}, \quad \forall j \in \mathbb{Z}$$
  
If  $j$  is odd :  
$$B_{j/2^{k+1}}^{(k+1)} = \left(\frac{B_{(j-1)/2^{k+1}}^{(k)} + B_{(j+1)/2^{k+1}}^{(k)}}{2}\right) + \frac{Z_{k+1,j}}{\sqrt{2^{k+2}}}$$

where all  $Z_j$ 's are IID. (Note that the first line exactly covers the *j* even case of the second line.) We claim the following proposition:

**Proposition 3.1.** Defining properties of Brownian Motion hold for  $B^{(k)}$  at times  $t_1, ..., t_i \in 2^{-k} \cdot \mathbb{Z}$ 

*Proof.* The point is to induct on *k*. As Mark did in the class, we check the variance of new points, assuming things work so far (so we are doing a representative part of a full induction, some remaining parts are left to homework). That is,

$$\mathbb{E}\left[\left(B_{j/2^{k+1}}^{(k+1)}\right)^2\right]$$

$$\mathbb{E}\left[\left(\frac{B_{(j-1)/2^{k+1}}^{(k)} + B_{(j+1)/2^{k+1}}^{(k)}}{2}\right)^2\right] + \mathbb{E}\left[\left(\frac{Z_{k+1,j}}{\sqrt{2^{k+2}}}\right)^2\right]$$
$$= \frac{1}{4}\left(\mathbb{E}\left[\underbrace{\left(B_{(j-1)/2^{k+1}}^{(k)}\right)^2}_{\left(\frac{j-1}{2^{k+1}}\right)^2} + \underbrace{2B_{(j-1)/2^{k+1}}^{(k)}B_{(j+1)/2^{k+1}}^{(k)}}_{\frac{j-1}{2^{k+1}}} + \underbrace{\left(B_{(j+1)/2^{k+1}}^{(k)}\right)^2}_{\frac{j-1}{2^{k+1}}}\right]\right) + \frac{1}{2^{k+2}}$$

and the last term is canceled out.

Next, we will show  $\{B^{(k)}\}\$  is an almost surely Cauchy sequence with respect to  $d_{sup}$ . Hence, it has a limit *B*. To do this, we prove and use the following lemma:

**Lemma 3.2.** 
$$\sum_{k=0}^{\infty} \mathbb{E}[d_{sup}(B^k, B^{k+1})] < \infty$$

Given this claim, we have  $\forall \epsilon, \exists N(\epsilon, \omega), \sum_{k=N}^{\infty} d_{sup}(B^k, B^{k+1}) \leq \epsilon$ . Consequently, we have  $d_{sup}(B^M, B^L) \leq \epsilon$ ,  $\forall M, L \geq N$ . We also prove Lemma 3.2:

*Proof of Lemma 3.2.* Up to scale, it suffices to prove that  $\mathbb{E}[\max_{i=1}^{n} |Z_i|] \leq O(\sqrt{\log(n)})$ , where  $\{Z_i\}$  are i.i.d. random variables following standard Gaussian distribution. Fix  $\lambda$ , and by Jensen's inequality,

$$e^{\lambda \mathbb{E}[\max_{i=1}^{n} |Z_i|]} \le \mathbb{E}[e^{\lambda \max|Z_i|}] \le \mathbb{E}[\sum_{i=1}^{n} e^{\lambda Z_i} + e^{-\lambda Z_i}] = 2ne^{\lambda^2/2}$$
$$\Rightarrow \mathbb{E}[\max|Z_i|] \le \inf_{\lambda} \frac{1}{\lambda} \left(\frac{\lambda^2}{2} + \log(2n)\right).$$

Choosing  $\lambda = \sqrt{\log(n)}$  gives the desired bound  $\mathbb{E}[\max_{i=1}^{n} |Z_i|] \le O(\sqrt{\log(n)}).$ 

Hence, 
$$\mathbb{E}[d_{sup}(B^{(k)}, B^{(k+1)})] = \frac{\max_{j} |Z_{k+1,j}|}{\sqrt{2^{k+2}}} \le O(\sqrt{k} \cdot 2^{-k/2}).$$

*Remark.* In fact, the limiting function  $B_t$  is  $(\frac{1}{2} - \epsilon)$  Hölder  $\forall \epsilon > 0$ , which means that  $\sup_{t,s \in [0,1]} \frac{|B_t - B_s|}{|t-s|^{\frac{1}{2}-\epsilon}} < \infty \forall \epsilon > 0$ .

*Proof.* Here, we only give the outline of the overall proof. The direction is analogous to the previous one. Define  $||f||_{C^{\frac{1}{2}-\epsilon}} = \sup_{t} |f(t)| + \sup_{t,s} \frac{|f(t)-f(s)|}{|t-s|^{\frac{1}{2}-\epsilon}}$ , which is a complete metric space (but not separable).  $B^{k}$  is still Cauchy and is  $\frac{\mathbb{E}[\max|Z_{k+1,j}|]}{2^{k/2}} \times 2^{k(\frac{1}{2}-\epsilon)} \approx \sqrt{k}2^{-k\epsilon}$ , which is still summable.

However, this metric space is not separable.

#### 3.2 Desired Properties

**Question (measurability)**: Why does this yield a probability measure on *C*([0, 1])?

**Proposition 3.3.** For each t,  $B_t = \lim_{k \to \infty} B_t^{(k)}$  is measurable with respect to the sequence of IID Gaussians  $(Z_{k,j})$ .

*Proof.*  $B_t$  is an infinite weighted sum of  $(Z_{k,j})$ .

**Proposition 3.4.** Borel  $\sigma$ -algebra on C([0,1]) is exactly the one generated by evaluation functions  $\varphi(t)$ . In other words, the smallest  $\sigma$ -field on C([0,1]) such that all maps  $\varphi(t)[B] = B_t$  are measurable is exactly the Borel  $\sigma$ -algebra.

Specifically, letting *F* denote the "construction of Brownian motion" above (which results in a function  $B = B_{[0,1]}$  from  $[0,1] \rightarrow \mathbb{R}$ ) and  $\phi_t$  the evaluation at time *t*, we have:

- (a)  $(Z_{k,j}) \xrightarrow{F} B \xrightarrow{\varphi_t} B_t \in \mathbb{R}$ , where  $Z_{k,j}$  lies in probability space  $(\Omega, \mathcal{F}, \nu)$ .
- (b)  $\varphi_t \circ F$  is measurable  $\forall t$  if and only if F is measurable wrt the Borel  $\sigma$ -algebra.
- (c) As a consequence, letting v be the product measure on our countably infinite family of Gaussians  $Z_{k,j}$ , the pushforward  $\mu = F \circ v$  is well defined, and so we have constructed a genuine probability measure for Brownian motion on C([0, 1]).

and  $A = \{S \subseteq C([0,1]) : F^{-1}(S) \in \mathcal{F}\}$  is a  $\sigma$ -field and  $A \supseteq \varphi_t^{-1}((a,b)), \forall t, a, b. \Rightarrow A \supseteq \text{Borel}(C([0,1])).$ 

*Proof.* Each  $\varphi_t$  is continuous with respect to  $d_{sup}$ , hence it is measurable with respect to Borel  $\sigma$ -algebra  $\Rightarrow \sigma(\varphi_t)_{t \in [0,1]} \subseteq \text{Borel}(C([0,1]))).$ 

In the other direction, we claim that  $\sigma(\varphi_t)_{t \in [0,1]}$  contains open balls  $\{f : d_{sup}(f,g) < \epsilon\} = B_{\epsilon}(g)$ . Indeed, we can write

$$B_{\varepsilon}(g) = \bigcup_{n \ge 1} \bigcap_{q \in \mathbb{Q}} \left\{ f : |f(q) - g(q)| < \varepsilon - \frac{1}{n} \right\}.$$

(Here the 1/n terms are needed in case e.g.  $|f(x) - g(x)| = \epsilon$  holds at exactly one value of *x* which is irrational.)