

---

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

## Construction of Brownian Motion

---

Instructor: Mark Sellke

Scribe: Zhiyu Li, Kentaro Nakamura

### 1 Definition

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

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

### 2 Questions Surrounding Definition

- (a) **Existence:** Does such a function in Definition 1.1 exist as a  $\mathcal{C}([0, 1])$ -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  $\mathcal{C}([0, 1])$  such that with  $\varphi_t : \mathcal{C}([0, 1]) \rightarrow \mathbb{R}$  given by  $\varphi_t(f) = f(t)$ , we should have  $\text{Law}(\varphi_t(B) - \varphi_s(B)) \sim \mathcal{N}(0, t - s)$ , 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  $t_1, t_2, \dots, t_n$ , the defining property 2 of Definition 1.1 tells us the joint law of  $(B_{t_1}, \dots, B_{t_k})$
- We should check if these distributions are consistent, i.e., if forgetting  $t_j$ , we can still recover correct law on  $(B_{t_1}, \dots, B_{t_{j-1}}, B_{t_{j+1}}, \dots, B_{t_k})$

**Theorem 2.1** (Kolmogorov Extension (or Consistency) Theorem). *There always exists a probability measure  $\tilde{\mu}$  on  $\mathbb{R}^{[0,1]}$  ( $\equiv \text{func}([0, 1] \rightarrow \mathbb{R}$ ; i.e., the set of all functions from the interval  $[0, 1]$  to  $\mathbb{R}$ ) which has all these finite-dimensional laws in the defining property 2 in Definition 1.1, given that these distributions are consistent.*

$\Rightarrow$  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  $\{f \in \mathbb{R}^{[0,1]} \mid f(t) \in (a, b)\}$  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  $z_0 \sim \mathcal{N}(0, 1)$ .

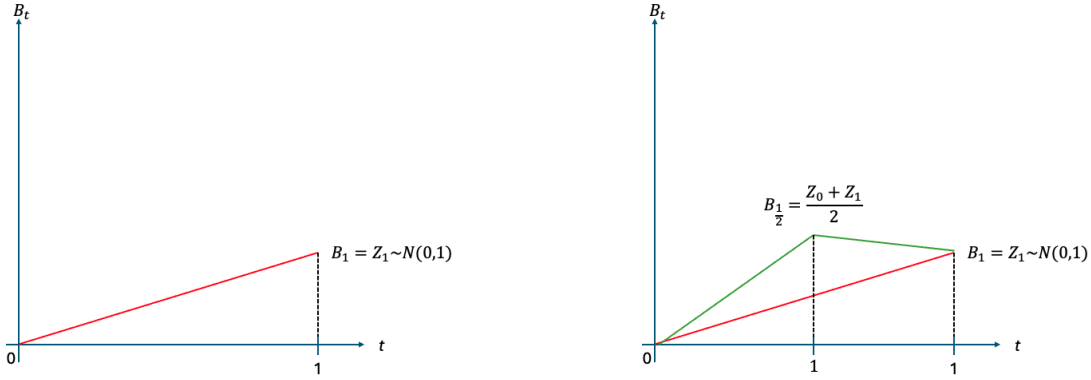


Figure 1: Case of  $K = 0$  (Left) and  $K = 1$  (Right)

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  $t_1, \dots, t_i \in 2^{-k} \cdot \mathbb{Z}$*

*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,

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

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

and the last term is canceled out.  $\square$

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,

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

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

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

*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  $\frac{\mathbb{E}[\max_j |Z_{k+1,j}|]}{2^{k/2}} \times 2^{k(\frac{1}{2}-\epsilon)} \approx \sqrt{k} 2^{-k\epsilon}$ , which is still summable.  $\square$

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

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

**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)  $(Z_{k,j}) \xrightarrow{F} B \xrightarrow{\phi_t} B_t \in \mathbb{R}$ , where  $Z_{k,j}$  lies in probability space  $(\Omega, \mathcal{F}, \nu)$ .
- (b)  $\phi_t \circ F$  is measurable  $\forall t$  if and only if  $F$  is measurable wrt the Borel  $\sigma$ -algebra.
- (c) As a consequence, letting  $\nu$  be the product measure on our countably infinite family of Gaussians  $Z_{k,j}$ , the pushforward  $\mu = F \circ \nu$  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

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

(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.) □