
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 $B : [0, 1] \rightarrow \mathbb{R}$ 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 μ 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: σ -algebra on $\mathbb{R}^{[0,1]}$ is generated by the evaluation mapping φ_t . $\{f \in \mathbb{R}^{[0,1]}, f(t) \in (a, b)\}$ is a measurable set (the sets defined by requiring the value at a particular time t to lie in some open interval (a, b) are members of our σ -algebra).

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

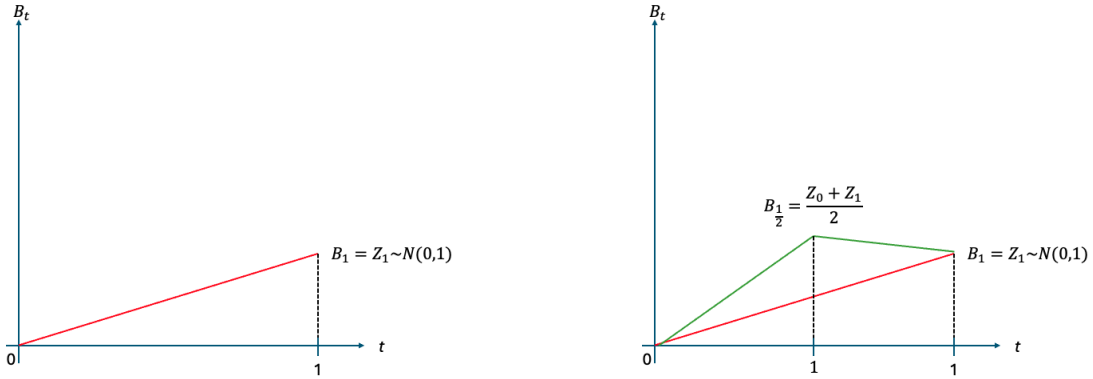


Figure 1: Case of $K = 0$ (Left) and $K = 1$ (Right)

Formally, we approximate 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 i.i.d.

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. Here, as Mark did in the class, we check the variance of new points. That is,

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

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

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, we need to understand $\mathbb{E}[\max_{i=1}^n |Z_i|] \leq O(\sqrt{\log(n)})$, where $\{Z_i\}$ are i.i.d. random variables following standard Gaussian distribution. Fix λ , and by Jensen's inequality,

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

where $\lambda = \sqrt{\log(n)}$

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 $(\frac{1}{2} - \epsilon)$ holder $\forall \epsilon > 0$, i.e., $\sup_{t,s} \frac{|B_t - B_s|}{|t-s|^{\frac{1}{2}-\epsilon}} < \infty \forall \epsilon > 0$.

Proof. Here, we only give the outline of the overall proof. The direction is analogous to the previous one. Define $\|f\|_{C^{\frac{1}{2}-\epsilon}} = d_{sup}(f, 0) + \sup_{t,s} \frac{|f(t)-f(s)|}{|t-s|^{\frac{1}{2}-\epsilon}}$, 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 , $B_t = \lim_{t \rightarrow \infty} B_t^{(k)}$ is measurable with respect to $(Z_{k,j})$.

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

Proposition 3.4. Borel σ -algebra on $C([0, 1])$ is exactly the one generated by evaluation function $\varphi(t)$. Specifically,

(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 $\forall t \Rightarrow F$ is measurable and so $\mu = F \circ \nu$ is well defined.

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

Proof. Each φ_t is continuous with respect to d_{sup} , hence it is measurable with respect to Borel σ -algebra $\Rightarrow \sigma(\varphi_t \leq \text{Borel}(C([0, 1])))$. We also need $\sigma(\varphi_t)$ contains open balls $\{f : d_{sup}(f, g) < \epsilon\} = B_\epsilon(g) = \bigcup_{n \geq 1} \bigcap_{q \in \mathbb{Q}} \{f : |f(q) - g(q)| < \epsilon - \frac{1}{n}\}$ □