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# Chapter 1

## Basics

variation, quadratic variation  
filtration

## 1.1 CONTINUITY

**Definition 1.** A function  $f : I \rightarrow \mathbb{R}$  is  $\gamma$ -**Hölder continuous** if there is a  $C < \infty$  such that

$$|f(t) - f(s)| \leq C |t - s|^\gamma$$

for all  $s, t \in I$ . Functions with  $\gamma = 1$  are **Lipschitz continuous**.

**Theorem 1** (Kolmogorov Continuity Theorem). Let  $\{X_t\}$  be a stochastic process on  $[0, 1]$ . If there are  $\alpha, \beta, C > 0$  such that

$$\mathbb{E}(|X_t - X_s|^\alpha) \leq C |t - s|^{1+\beta},$$

then there is a version  $\tilde{X}_t$  of  $X_t$  with sample paths that are almost surely  $\gamma$ -Hölder continuous for  $\gamma \in (0, \beta/\alpha)$ .

**version means**  $\mathbb{P}(\tilde{X}_t = X_t) = 1$  **for all**  $t$ .

## Chapter 2

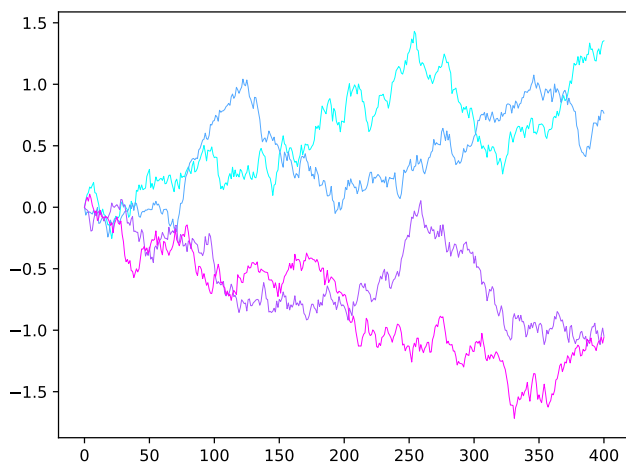
# Brownian Motion

**Definition 2.** A standard **Brownian motion**  $B(t, \omega)$  is a continuous time  $\mathbb{R}$ -valued stochastic process over some  $(\Omega, \mathcal{F}, \mathbb{P})$  such that

1.  $B_t - B_s \sim \mathcal{N}(0, t - s)$ ;
2. Disjoint increments are independent;
3. The sample path  $t \mapsto B_t(\omega)$  is continuous with probability 1.

At all times, a Brownian motion receives an infinitesimal Gaussian kick. The intuition here is that “ $dB$ ” is then a Gaussian random variable. Of course,  $dB$  is meaningless right now since  $B$  is nowhere differentiable with probability 1, but we will give it meaning later in terms of Itô integrals, and the interpretation will be the same.

A useful fact for proving that disjoint intervals are independent: two Gaussians are independent  $\iff$  they have 0 covariance.



**Proposition 1.** If  $B_t$  is a Brownian motion, then so are the following two processes:

- $X_t := \frac{1}{\sqrt{\alpha}} B_{\alpha t}$  for fixed  $\alpha > 0$ ;
- $Y_t := B_{s+t} - B_s$  for fixed  $s > 0$ ;
- $Z_t := tB_{1/t}$ .

**Proposition 2.** If  $B_t$  is a Brownian motion, then  $\text{Cov}(B_t, B_s) = \min(t, s)$ .

**Construct a BM using Wiener measure and  $B_t(\omega) = \omega_t$ .**

**Is the following the finite dimensional distribution stuff?**

Let  $A := \{\omega \mid B_{t_k}(\omega) \in (a_k, b_k) \text{ for } k = 1, \dots, N\}$ . If

$$\phi(s, y) := \frac{\exp(-y^2/(2s))}{\sqrt{2\pi s}},$$

then the probability of  $A$  is

$$\mathbb{P}(A) = \int_{a_1}^{b_1} \cdots \int_{a_n}^{b_n} \phi(t_1, x_1) \prod_{i=2}^N \phi(t_i - t_{i-1}, x_i - x_{i-1}) dx_1 \cdots dx_n.$$

The idea here is that  $\phi(t_i - t_{i-1}, x_i - x_{i-1})$  is the conditional density for  $B_{t_k}$  given  $B_{t_{k-1}} = x_{k-1}$ .

**Proposition 3.** The sample paths of Brownian motion are almost surely  $\gamma$ -Hölder continuous for  $\gamma \in (0, 1/2)$ .

**Proposition 4.** If  $B$  is a Brownian motion on  $[0, T]$ , then with probability 1,

- $V^p(B, [0, T]) < \infty$  for  $p > 2$ ;
- $V^p(B, [0, T]) = \infty$  for  $p < 2$ .

The quadratic variation of  $B$  is  $[B, B](t) = t$ .

## Chapter 3

# Integration

### 3.1 INTEGRATION OF SIMPLE PROCESSES

Suppose  $B_t$  is a Brownian motion adapted to  $\{\mathcal{F}_t\}$ . Then  $\mathcal{L}_A^2([0, T] \times \Omega)$  is the space of all processes  $X(t, \omega)$  adapted to  $\{\mathcal{F}_t\}$  such that

$$\mathbb{E} \left( \int_0^T X^2 ds \right) < \infty.$$

This space is Banach space (complete normed vector space) with norm

$$\|X\|_{\mathcal{L}_A^2} = \sqrt{\mathbb{E} \left( \int_0^T X^2 ds \right)}.$$

The subspace  $\mathcal{L}_{A,0}^2 \subset \mathcal{L}_A^2$  of *simple* adapted processes is dense in  $\mathcal{L}_A^2$ : for any  $X \in \mathcal{L}_A^2$ , there is a sequence  $\{X_n\} \subset \mathcal{L}_{A,0}^2$  converging to  $X$  in the  $\mathcal{L}^2$  sense, i.e.

$$\lim_{n \rightarrow \infty} \|X_n - X\|_{\mathcal{L}_A^2} = \lim_{n \rightarrow \infty} \sqrt{\mathbb{E} \left( \int_0^T (X_n - X)^2 ds \right)} = 0.$$

We'll define the Itô integral for simple adapted processes, then extend it to general adapted processes in the next section.

**finish**

**Proposition 5.** The quadratic variation of  $X(t) = \int_0^t \sigma dB$  is

$$[X, X](t) = \int_0^t \sigma^2 ds.$$

Note that if  $\sigma$  depends on  $\omega$ , then  $[X, X](t)$  is still a random variable.

### 3.2 EXTENDING THE ITÔ INTEGRAL

For  $X \in \mathcal{L}_A^2$ , we know there's a sequence  $\{X_n\}$  converging to  $X$  in the  $\mathcal{L}^2$  sense. Then by the Itô isometry, the sequence  $\{I_n\}$  given by

$$I_n := \int_0^T X_n dB$$

is a Cauchy sequence. Thus there is a random variable  $I \in L^2$  such that  $I_n \rightarrow I$  in the  $L^2$  sense, i.e.

$$\lim_{n \rightarrow \infty} \|I_n - I\|_{L^2} = \lim_{n \rightarrow \infty} \mathbb{E}(|I_n - I|^2) = 0.$$

**Definition 3.** For  $X \in \mathcal{L}_A^2$ ,

$$\int_0^T X dB$$

is the unique limit of the sequence given by  $I_n := \int_0^T X_n dB$ , where  $X_n \rightarrow X$ .

This integral has all the same properties as the one for simple processes. **Further extension where martingale property becomes \*local\* martingale property.**

### 3.3 ITÔ PROCESSES

do this.



### 3.4 ITÔ'S FORMULA

**do this.**