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Chapter 1

Basics

variation, quadratic variation
filtration

1.1 CONTINUITY

Definition 1. A function $f : I \rightarrow \mathbb{R}$ is γ -**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 γ -Hölder continuous for $\gamma \in (0, \beta/\alpha)$.

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

Chapter 2

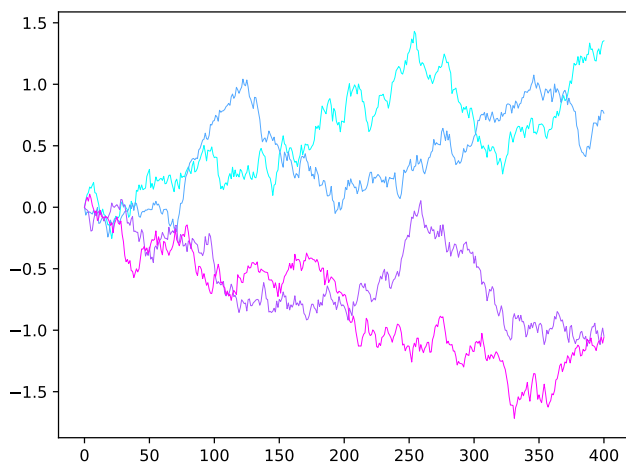
Brownian Motion

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 γ -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. Let $\sigma \in \mathcal{L}_A^2$, then the quadratic variation of $X(t) = \int_0^t \sigma dB$ is

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

Note that if σ depends on ω , 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

Theorem 2. Let Z be an Itô process satisfying

$$dZ = \mu \, dt + \sigma \, dB.$$

Let $f(t, x) \in C^2$, then

$$df(t, Z_t) = \left[\frac{\partial f}{\partial x} \right] dZ + \left[\frac{\partial f}{\partial t} + \frac{1}{2} \frac{\partial^2 f}{\partial x^2} \sigma^2 \right] dt,$$

where all partial derivatives are evaluated at (t, Z_t) .