

# Girsanov's Theorem

2022年7月13日 20:03

$B_t$  is a BM on  $(\Omega, \mathcal{F}, \mathbb{P})$

$\gamma$  is a constant (可依赖于 time-dependent & random, i.e. stochastic process)

$$Z_t = e^{\int_0^t \gamma dB_s - \frac{1}{2} \int_0^t \gamma^2 ds}$$

$$\tilde{\mathbb{P}}(dz) = e^{\mu z - \frac{1}{2} \mu^2} \mathbb{P}(dz) \sim N(\mu, 1)$$

$$\mathbb{R} \times \tilde{\mathbb{P}} \text{ on } \tilde{\mathcal{F}}_T : \mathbb{E}(\mathbb{1}_A Z_T) = \mathbb{E} \mathbb{1}_A = \mathbb{P}(A)$$

$$\tilde{B}_t = B_t - \gamma t \text{ 为 BM } (\Omega, \tilde{\mathcal{F}}, \tilde{\mathbb{P}})$$

So assume that  $B_t, t \geq 0$  defined on  $(\Omega, \mathcal{F}, \mathbb{P})$  is a Brownian motion starting from zero. Now define for a given constant  $\gamma$

$$Z_t = e^{\int_0^t \gamma dB_s - \frac{1}{2} \int_0^t \gamma^2 ds} = e^{\gamma B_t - \frac{1}{2} \gamma^2 t} \quad (3)$$

and for each  $T > 0$  define a new probability measure  $\tilde{\mathbb{P}}_T$  on  $\mathcal{F}_T$  (i.e. on the  $\sigma$ -algebra of the events known at time  $T$ ) by

$$\tilde{\mathbb{P}}_T = \mathbb{E}(\mathbb{1}_A Z_T) \text{ for } A \in \mathcal{F}_T. \quad (4)$$

**Theorem 1** (Girsanov, Cameron-Martin). Define a process  $\tilde{B}_t = B_t - \gamma t$ . Then for each fixed  $T < \infty$  the process  $\tilde{B}_t$  is a Brownian motion on  $(\Omega, \mathcal{F}_T, \tilde{\mathbb{P}}_T)$ .

$$\text{trivial case: } \gamma=0 \Rightarrow Z_t=1 \Rightarrow \tilde{\mathbb{P}}(A)=\mathbb{P}(A)$$

pf: ①  $Z_t$  为  $\tilde{\mathbb{P}}$  下 martingale. (证明: Itô's Lemma)

②  $\tilde{\mathbb{P}}_T$  测度 are consistent. i.e., if  $t \in T, A \in \tilde{\mathcal{F}}_t, \tilde{\mathbb{P}}_t(A) = \tilde{\mathbb{P}}_T(A)$

**Lemma 3.** Measures  $\tilde{\mathbb{P}}_T$  are consistent, that is for  $t < T$  and  $A \in \mathcal{F}_t$

$$\tilde{\mathbb{P}}_T(A) = \tilde{\mathbb{P}}_t(A). \quad (5)$$

*Proof of Lemma 3.* Indeed, using the tower property of conditional expectation

$$\begin{aligned} \tilde{\mathbb{P}}_T(A) &= \mathbb{E}_T(\mathbb{1}_A) = \mathbb{E}(Z_T \mathbb{1}_A) = \mathbb{E}(\mathbb{E}(Z_T \mathbb{1}_A | \mathcal{F}_t)) \\ &= \mathbb{E}(\mathbb{1}_A \mathbb{E}(Z_T | \mathcal{F}_t)) = \mathbb{E}(\mathbb{1}_A Z_t) = \tilde{\mathbb{P}}_t(A). \end{aligned} \quad (6)$$

□

Fix a sequence  $t_1 \leq t_2 \leq \dots \leq t_n$  and consider Brownian increments  $B(t_1), B(t_2) - B(t_1), \dots, B(t_n) - B(t_{n-1})$ . By the definition of  $\tilde{\mathbb{P}}_T$

$$\begin{aligned} \tilde{\mathbb{P}}_T[B(t_1) \in dz_1, B(t_2) - B(t_1) \in dz_2, \dots, B(t_n) - B(t_{n-1}) \in dz_n] \\ &= e^{\gamma(B(t_n) - \gamma^2 t_n / 2)} \mathbb{P}[B(t_1) \in dz_1, B(t_2) - B(t_1) \in dz_2, \dots, B(t_n) - B(t_{n-1}) \in dz_n] \\ &= e^{\gamma(t_1 + \dots + t_n) - \gamma^2 t_n / 2} \frac{1}{(\sqrt{2\pi})^n} e^{-\frac{1}{2}(dz_1^2 + \dots + dz_n^2)} \\ &= \frac{1}{(\sqrt{2\pi})^n} e^{\gamma(t_2 - \gamma^2 t_1 / 2 - z_1^2 / (2t_1)) + \dots + (t_n - \gamma(t_n - t_{n-1}) / 2 - z_n^2 / (2(t_n - t_{n-1})))} \\ &= \frac{1}{(\sqrt{2\pi})^n} e^{-(t_1 - t_1)^2 / (2t_1) - \dots - (t_n - \gamma(t_n - t_{n-1}))^2 / (2(t_n - t_{n-1}))}. \end{aligned} \quad (7)$$

Thus under  $\tilde{\mathbb{P}}$  random variables  $B(t_1) - \gamma t_1, B(t_2) - B(t_1) - \gamma(t_2 - t_1), \dots, B(t_n) - B(t_{n-1}) - \gamma(t_n - t_{n-1})$  are independent normal random variables with expectation zero and variances  $t_1, t_2 - t_1, \dots, t_n - t_{n-1}$ . So we proved that the finite dimensional distributions of  $\tilde{B}_t = B_t - \gamma t$  under  $\tilde{\mathbb{P}}$  coincide with the finite dimensional distributions of Brownian motion. □

Define  $A = \{B_{t_1} \in [z_1, z_1 + dz_1], B_{t_2} - B_{t_1} \in [z_2, z_2 + dz_2], \dots, B_{t_n} - B_{t_{n-1}} \in [z_n, z_n + dz_n]\}$

$$\tilde{\mathbb{P}}(A) = \mathbb{P}(B_{t_1} \in dz_1) \cdot \mathbb{P}(B_{t_2} - B_{t_1} \in dz_2) \cdot \dots \cdot \mathbb{P}(B_{t_n} - B_{t_{n-1}} \in dz_n)$$

$$\int_{z_1}^{\infty} \int_{z_2}^{\infty} f(z_1, z_2) dz_1 dz_2 = f(z_1) \cdot dz_1 \cdot dz_2$$

$$\Rightarrow \tilde{\mathbb{P}}(A) = \frac{1}{\sqrt{2\pi t_1}} \exp\left(-\frac{z_1^2}{2t_1}\right) dz_1 \times \frac{1}{\sqrt{2\pi(t_2-t_1)}} \exp\left(-\frac{z_2^2}{2(t_2-t_1)}\right) dz_2 \dots$$

$$\begin{aligned} \text{By Lemma 3} &\Rightarrow \tilde{\mathbb{P}}_T(A) = \tilde{\mathbb{P}}_{t_n}(A) \\ \text{By def} &\Rightarrow \tilde{\mathbb{P}}_T = \mathbb{E}(Z_n \cdot \mathbb{1}_A) \end{aligned}$$

$$\int_{z_1}^{\infty} \int_{z_2}^{\infty} Z_n \exp\left(\gamma(B_{t_n} - \frac{1}{2} \gamma^2 t_n)\right) dz_1 \dots dz_n$$

$$\therefore Z_{t_n} = \exp\left(\gamma(Z_{t_n} - \frac{1}{2} \gamma^2 t_n)\right) = \mathbb{E}(Z_{t_n} \mathbb{1}_A) = e^{\gamma(t_n - \gamma^2 t_n / 2) - \frac{\gamma^2}{2}(t_1 + t_2 + \dots + t_{n-1})}$$

$$\int_{z_1}^{\infty} \int_{z_2}^{\infty} \exp\left(\gamma(z_1 + \dots + z_n)\right) = \frac{1}{\sqrt{2\pi t_1}} e^{\frac{\gamma^2}{2} t_1} \cdot e^{\gamma z_1 - \frac{1}{2} \gamma^2 t_1}$$

$$\int_{z_2}^{\infty} \exp\left(\gamma z_2 - \frac{1}{2} \gamma^2 (t_2 - t_1) + \frac{z_2^2}{2t_1}\right) \text{ 为完全平方}$$

$$Z_{t_n} = \frac{1}{\sqrt{2\pi t_1}} e^{-\frac{(z_1 - \gamma t_1)^2}{2t_1}} dz_1 \times \frac{1}{\sqrt{2\pi(t_2-t_1)}} \exp\left(-\frac{(z_2 - \gamma(t_2-t_1))^2}{2(t_2-t_1)}\right) \times \dots \times \frac{1}{\sqrt{2\pi(t_n-t_{n-1})}} \exp\left(-\frac{(z_n - \gamma(t_n-t_{n-1}))^2}{2(t_n-t_{n-1})}\right)$$

$$\mathbb{P}(z > b) = \tilde{\mathbb{P}} = N(a, b)$$