

Financial Stochastic Analysis

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Chapter 1

Brownian Motion

Brownian motion at time t is limit of infinite fast random walk W_t^n , it can be equivalently characterized by

1. Any increment $W_{t_1} - W_{t_2}$ is normal distributed with mean 0 and variance $t_1 - t_2$. Disjoint increment are independency.
2. For any time t_1, t_2, \dots, t_m , $\mathbf{W} = (W_{t_1}, W_{t_2}, \dots, W_{t_m})$ is normal distributed with zero mean and covariance

$$\Sigma = \begin{bmatrix} t_1 & t_1 & \dots & t_1 \\ t_1 & t_2 & \dots & t_2 \\ \vdots & \vdots & & \vdots \\ t_1 & t_2 & \dots & t_m \end{bmatrix}$$

3. \mathbf{W} has MGF

$$\varphi(\mathbf{t}) = \exp \left\{ \sum_{i=1}^m \frac{1}{2} \left(\sum_{j=i}^m t_j \right)^2 (t_i - t_{i-1}) \right\}$$

where $t_0 = 0$.

Brownian motion is a Markov martingale with variation:

1. $dW_t = \infty$
2. $dW_t dW_t = dt$
3. $dW_t dt = dt^2 = 0$

1.1 Markov Property

Lemma 1.1 (Independence Lemma). *Suppose $X \in \mathcal{A}$, $Y \perp \mathcal{A}$, then*

$$\mathbb{E}_{\mathcal{A}} f(X, Y) = \mathbb{E} f(x, Y)|_{x=X}$$

Proof. When $f = g \times h$ for some g, h , then

$$\mathbb{E}_{\mathcal{A}} f(X, Y) = \int K(X, dy) f(X, y) = \int \mu(dy) f(X, y) = \mathbb{E} f(x, Y)|_{x=X}$$

since product σ algebra is generated by measurable rectangles, monotone class theorem completes the proof. \square

Preceding lemma implies

$$\begin{aligned} \mathbb{E}_s f(W_t) &= \mathbb{E}_s f(W_t - W_s + W_s) \\ &= \mathbb{E} f(W_t - W_s + x)|_{x=W_s} \\ &= \frac{1}{\sqrt{2\pi(t-s)}} \int_{\mathbb{R}} f(w+x) \exp\left\{-\frac{w^2}{2(t-s)}\right\} dw|_{x=W_s} \\ &\stackrel{\tau=t-s, y=w+x}{=} \frac{1}{\sqrt{2\pi\tau}} \int_{\mathbb{R}} f(y) \exp\left\{-\frac{w^2}{2\tau}\right\} dw|_{x=W_s} \\ &= \int_{\mathbb{R}} f(y) p(\tau, W_s, y) dy \end{aligned}$$

where $p(\tau, W_s, y)$ is pdf of $\mathcal{N}(W_s, \tau)$.

1.2 Exponential Martingale

Proposition 1.1. *Suppose W_t is a Brownian Motion with filtration \mathbb{F} , then process*

$$Z_t = \exp\left\{\sigma W_t - \frac{1}{2}\sigma^2 t\right\}$$

is a martingale.

Define the first passage time to m as

$$\tau_m = \min\{t \geq 0, W_t = m\}$$

recall the stopped martingale, we have

$$1 = \mathbb{E} Z_0 = \mathbb{E} Z_{t \wedge \tau_m} = \mathbb{E} \exp \left\{ \sigma W_{t \wedge \tau_m} - \frac{1}{2} \sigma^2 (t \wedge m) \right\}$$

Taking limit inside expectations:

$$\lim_{t \rightarrow \infty} \exp \left\{ \sigma W_{t \wedge \tau_m} - \frac{1}{2} \sigma^2 (t \wedge m) \right\} = \mathbf{1}_{\tau_m < \infty} \exp \left\{ \sigma m - \frac{1}{2} \sigma^2 \tau_m \right\}$$

that implies

$$\mathbb{E} \mathbf{1}_{\tau_m < \infty} \exp \left\{ -\frac{1}{2} \sigma^2 \tau_m \right\} = \exp \{-\sigma m\}$$

take $\sigma \searrow 0$, we have τ_m is finite *a.s.*

And the characteristic function of τ_m is given by taking $t = \frac{1}{2} \sigma^2$:

$$\mathbb{E} \exp \{-t \tau_m\} = \exp \left\{ -|m| \sqrt{2t} \right\}$$

1.3 Reflection

By the symmetry of Brownian motion, we have

$$\mathbb{P} \{\tau_m \leq t, W_t \leq w\} = \mathbb{P} \{W_t \geq 2m - w\}$$

when $0 < m \leq w$. On the other hand:

$$\mathbb{P} \{\tau_m \leq t, W_t \geq w\} = \mathbb{P} \{W_t \geq w\}$$

take $m = w$ and adding these two:

$$\mathbb{P} \{\tau_m \leq t\} = 2 \mathbb{P} \{W_t \geq m\}$$

1.3.1 Joint Distribution of Brownian Motion and its maximum

Define maximum process:

$$M_t = \max_{0 \leq s \leq t} W_s$$

clearly, $M_t \geq m$ iff $\tau_m \leq t$, thus

$$\mathbb{P} \{M_t \geq m, W_t \leq w\} = \mathbb{P} \{W_t \geq 2m - w\}$$

from which we have:

$$f_{M,W}(m,w)=\frac{2(2m-w)}{t\sqrt{2\pi t}}e^{-\frac{(2m-\omega)^2}{2t}}$$

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