## Stochastic Financial Models 21

Michael Tehranchi

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## 1 Properties of Brownian motion

Theorem (Wiener 1923). Brownian motion exists.

Remark. A Brownian motion is called a Wiener process in the US.

**Theorem.** Brownian motion is a martingale in its filtration  $\mathcal{F}_t = \sigma(W_s : 0 \le s \le t)$ .

*Proof.* Brownian motion is integrable, adapted and

$$\mathbb{E}[W_t - W_s | \mathcal{F}_s] = \mathbb{E}[W_t - W_s] = 0$$

for  $0 \le s \le t$  by the independence of  $W_t - W_s$  and  $\mathcal{F}_s$ .

**Theorem.** Brownian motion is a Markov process.

*Proof.* Since  $W_s$  is  $\mathcal{F}_s$  measurable and  $W_t - W_s$  is independent of  $\mathcal{F}_s$  for  $0 \le s \le t$ , we have

$$\mathbb{E}[g(W_t)|\mathcal{F}_s] = \mathbb{E}[g(W_t - W_s + W_s)|\mathcal{F}_s]$$
$$= \mathbb{E}[g(W_t - W_s + x)]\big|_{x = W_s}$$
$$= \mathbb{E}[g(W_t)|W_s]$$

**Definition.** A process  $(X_t)_{t\geq 0}$  is Gaussian iff the random variables  $X_{t_1}, \ldots, X_{t_n}$  are jointly normal for all  $0 \leq t_1 \leq \ldots \leq t_n$ , i.e. the random variable  $\sum_{i=1}^n a_i X_{t_i}$  is normally distributed for all constants  $a_1, \ldots, a_n$ .

**Theorem.** The following are equivalent

- 1.  $(W_t)_{t\geq 0}$  is a Brownian motion
- 2.  $(W_t)_{t\geq 0}$  is a Gaussian process such that
  - $t \mapsto W_t$  is continuous

- $\mathbb{E}[W_t] = 0$  for all  $t \ge 0$
- $\mathbb{E}[W_s W_t] = s \text{ for all } 0 \le s \le t$

*Proof.* Suppose  $(W_t)_{t\geq 0}$  is a Brownian motion. Fix  $0=t_0\leq t_1\leq \ldots \leq t_n$  and  $a_1,\ldots,a_n$ . Note

$$\sum_{i=1}^{n} a_i W_{t_i} = \sum_{i=1}^{n} b_i (W_{t_i} - W_{t_{i-1}})$$

where  $b_k = \sum_{i=k}^n a_i$ . Since  $W_{t_1} - W_{t_0}, \dots, W_{t_n} - W_{t_{n-1}}$  are independent normals, and the linear combination of independent normals is normal, we have that  $(W_t)_{t\geq 0}$  is Gaussian with  $\mathbb{E}[W_t] = \mathbb{E}[W_0] = 0$  and

$$\mathbb{E}[W_s W_t] = \mathbb{E}[W_s^2] + \mathbb{E}[W_s (W_t - W_s)]$$

$$= \operatorname{Var}(W_s) + \mathbb{E}(W_s) \mathbb{E}(W_t - W_s)$$

$$= 0.$$

for  $0 \le s \le t$ , since  $W_s$  and  $W_t - W_s$  are independent.

Conversely, suppose  $(W_t)_{t\geq 0}$  is a continuous Gaussian process such that  $\mathbb{E}[W_t]=0$  and  $\mathbb{E}[W_sW_t]=s$  for all  $0\leq s\leq t$ . Then for  $0\leq u\leq s\leq t$  we have

$$Cov(W_u, W_t - W_s) = \mathbb{E}[W_u W_t] - \mathbb{E}[W_u W_s]$$
$$= u - u = 0$$

By normality, the increment  $W_t - W_s$  is independent of  $W_u$ . By Gaussianity, the increment is independent of  $(W_u)_{0 \le u \le s}$ .

**Theorem.** Let  $(W_t)_{t\geq 0}$  be a Brownian motion. Then each of the following processes are also Brownian motions.

- 1.  $\tilde{W}_t = cW_{t/c^2}$ , for any constant  $c \neq 0$ .
- 2.  $\tilde{W}_t = W_{t+T} W_T$  for any constant  $T \ge 0$ .
- 3.  $\tilde{W}_0 = 0$  and  $\tilde{W}_t = tW_{1/t}$  for t > 0.

*Proof.* Check that each process is a continuous mean-zero Gaussian process with the correct covariance. [For 3, we technically need the Brownian law of large number  $\frac{W_s}{s} \to 0$  as  $s \to \infty$  to prove continuity of  $\tilde{W}$  at t = 0.]

## 2 Reflection principle

**Theorem.** Let  $(W_t)_{t\geq 0}$  be a Brownian motion, and  $T_a = \inf\{t \geq 0 : W_t = a\}$ . Then  $T_a < \infty$  almost surely.

*Proof.* Consider the case a > 0. (The case a < 0 is similar.) We must show

$$\sup_{t>0} W_t > a \text{ almost surely}$$

By Brownian scaling, for any c > 0 and 0 < a < b, we have

$$\mathbb{P}(a < \sup_{t \ge 0} W_t < b) = \mathbb{P}(a < \sup_{t \ge 0} cW_{t/c^2} < b)$$
$$= \mathbb{P}(a/c < \sup_{s \ge 0} W_s < b/c) \quad \text{letting } t/c^2 =$$
$$\to 0$$

by sending  $c \uparrow \infty$ . Since  $Z = \sup_{t \geq 0} W_t \geq W_0 = 0$ , we have shown that  $Z \in \{0, +\infty\}$  almost surely.

Let  $\hat{Z} = \sup_{t \geq 1} (W_t - W_1)$ . Note Z and  $\hat{Z}$  have the same distribution, so  $\hat{Z} \in \{0, +\infty\}$  almost surely.

Note that  $\{\hat{Z} = \infty\} = \{Z = \infty\}$  since  $\sup_{0 \le t \le 1} W_t$  is finite by the continuity of Brownian motion. Hence

$$\begin{split} p &= \mathbb{P}(Z=0) = \mathbb{P}(Z=0, \hat{Z}=0) \\ &\leq \mathbb{P}(W_1 \leq 0, \hat{Z}=0) \ \text{ as Z = 0 implies W\_t <= 0 for all t} \\ &= \frac{1}{2}\mathbb{P}(\hat{Z}=0) = \frac{1}{2}p \end{split}$$

so p = 0. Hence  $\sup_{t > 0} W_t = \infty$  almost surely.

**Theorem.** Let  $(W_t)_{t\geq 0}$  be a Brownian motion and T a finite stopping time. The process  $W_{t+T} - W_T$  is also a Brownian motion independent of  $(W_t)_{0\leq t\leq T}$ 

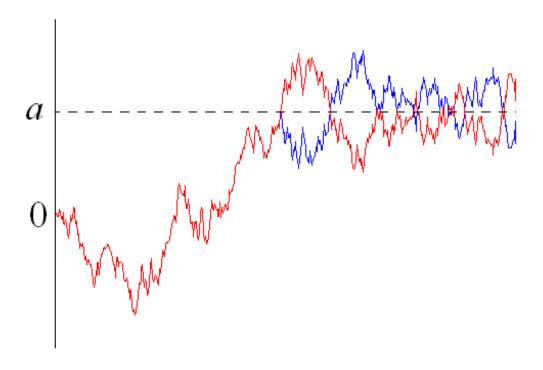
*Proof.* Omitted. The idea is Brownian motion is a strong Markov process.  $\Box$ 

Applying this with the finite stopping time  $T_a$  together with the symmetry of Brownian motion, we have

**Theorem** (Reflection principle). Let  $(W_t)_{t\geq 0}$  be a Brownian motion and let

$$\tilde{W}_t = \begin{cases} W_t & \text{if } 0 \le t < T_a \\ 2a - W_t & \text{if } t \ge T_a \end{cases}$$

Then  $(\tilde{W}_t)_{t\geq 0}$  is a Brownian motion.



Reflection principle: Key formula

$$\boxed{\mathbb{P}(\max_{0 \le s \le t} W_s \ge a, W_t \le b) = \mathbb{P}(W_t \ge 2a - b) \text{ for } a \ge 0, b \le a}$$

*Proof.* We have

$$\mathbb{P}(\max_{0 \le s \le t} W_s \ge a, W_t \le b) = \mathbb{P}(\tilde{W}_t \ge 2a - b)$$
$$= \mathbb{P}(W_t \ge 2a - b)$$