

**Brownian motion:**

- 1)  $B(0) = 0$ .
- 2) If  $0 \leq s_1 \leq t_1 \leq s_2 \leq t_2$ ,  $B(t_1) - B(s_1)$  and  $B(t_2) - B(s_2)$  are independent.
- 3)  $B(t) - B(s)$  has a Normal distribution with mean 0 and variance  $\sigma^2(t-s)$ ,  $0 \leq s \leq t$ .

$(B(t) | B(s) = b)$  has a Normal distribution with mean with mean  $b$   
and variance  $\sigma^2(t-s)$ ,  $0 \leq s \leq t$ .

$(B(s) | B(t) = b)$  has a Normal distribution with mean with mean  $\frac{s}{t} b$   
and variance  $\frac{s(t-s)}{t} \sigma^2$ ,  $0 \leq s \leq t$ .

$$r_B(s, t) = \text{Cov}(B(s), B(t)) = \sigma^2 \min(s, t), \quad s, t \geq 0.$$

**1.** Let  $B(t)$  be a Brownian motion with parameter  $\sigma = 5$ .

- a) Find the probability  $P(B(10) < 18)$ .

$$P(B(10) < 18) = P\left(Z < \frac{(18-0)-0}{\sqrt{5^2(10-0)}}\right) \approx P(Z < 1.14) = \mathbf{0.8729}.$$

- b) Find the probability  $P(B(10) < 18 | B(6) = 8)$ .

$$P(B(10) < 18 | B(6) = 8) = P\left(Z < \frac{18-8}{\sqrt{5^2(10-6)}}\right) = P(Z < 1.00) = \mathbf{0.8413}.$$

c) Find the probability  $P(B(6) < 8 \mid B(10) = 18)$ .

$(B(s) \mid B(t) = b)$  has a Normal distribution with mean with mean  $\frac{s}{t} b$

and variance  $\frac{s(t-s)}{t} \sigma^2$ ,  $0 \leq s \leq t$ .

$$P(B(6) < 8 \mid B(10) = 18) = P\left(Z < \frac{8 - \frac{6}{10} \cdot 18}{\sqrt{\frac{6 \cdot (10-6)}{10} \cdot 5^2}}\right) = P\left(Z < \frac{8 - 10.8}{\sqrt{60}}\right)$$

$$\approx P(Z < -0.36) = \mathbf{0.3594}.$$

d) Find the probability  $P(B(10) - B(6) < 18)$ .

$$P(B(10) - B(6) < 18) = P\left(Z < \frac{18 - 0}{\sqrt{5^2 (10-6)}}\right) = P(Z < 1.80) = \mathbf{0.9641}.$$

e) Find the probability  $P(B(10) + B(6) < 18)$ .

$$E(B(10) + B(6)) = 0.$$

$$\begin{aligned} \text{Var}(B(10) + B(6)) &= \text{Var}(B(10)) + 2 \text{Cov}(B(10), B(6)) + \text{Var}(B(6)) \\ &= 10 \sigma^2 + 2 \times 6 \sigma^2 + 6 \sigma^2 = 28 \sigma^2 = 700. \end{aligned}$$

$$P(B(10) + B(6) < 18) = P\left(Z < \frac{18 - 0}{\sqrt{700}}\right) \approx P(Z < 0.68) = \mathbf{0.7517}.$$

Let  $B(t)$  be a Brownian motion with parameter  $\sigma$ .

Let  $T_a$  denote the first time the Brownian motion process hits  $a > 0$ .

$$P(B(t) \geq a) = P(B(t) \geq a \mid T_a \leq t) \cdot P(T_a \leq t) + P(B(t) \geq a \mid T_a > t) \cdot P(T_a > t).$$

By symmetry,  $P(B(t) \geq a \mid T_a \leq t) = \frac{1}{2}.$

By continuity,  $P(B(t) \geq a \mid T_a > t) = 0.$

Thus  $F_{T_a}(t) = P(T_a \leq t) = 2 P(B(t) \geq a) = 2 P(Z \geq \frac{a}{\sigma \sqrt{t}})$

$$= \int_{\frac{a}{\sigma \sqrt{t}}}^{\infty} \frac{2}{\sqrt{2\pi}} e^{-z^2/2} dz, \quad t > 0.$$

$$\begin{aligned} f_{T_a}(t) = F'_{T_a}(t) &= \frac{2}{\sqrt{2\pi}} e^{-a^2/(2\sigma^2 t)} \cdot \left( \frac{a}{2\sigma \sqrt{t^3}} \right) \\ &= \frac{a}{\sqrt{2\pi} \sigma t^{3/2}} e^{-a^2/(2\sigma^2 t)}, \quad t > 0. \end{aligned}$$

If  $W$  has a Normal distribution with mean 0 and variance  $\frac{\sigma^2}{a^2}$ ,

then  $T_a$  has the same distribution as  $\frac{1}{W^2}$ .

$$P(T_a > t) = 1 - F_{T_a}(t) = P\left(-\frac{a}{\sigma \sqrt{t}} < Z < \frac{a}{\sigma \sqrt{t}}\right) \rightarrow 0 \quad \text{as } t \rightarrow \infty.$$

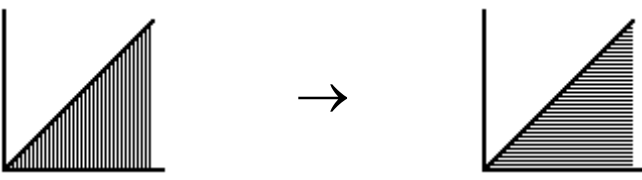
$$P(T_a = \infty) = \lim_{t \rightarrow \infty} P(T_a > t) = 0. \quad \Rightarrow \quad P(T_a < \infty) = 1.$$

Fact: Let  $X$  be a nonnegative continuous random variable with p.d.f.  $f(x)$  and c.d.f.  $F(x)$ . Then

$$E(X) = \int_0^{\infty} (1 - F(x)) dx.$$

Proof:

$$E(X) = \int_0^{\infty} x f(x) dx = \int_0^{\infty} \left( \int_0^x dy \right) f(x) dx = \int_0^{\infty} \left( \int_0^x f(x) dy \right) dx$$



$$\int_0^{\infty} \left( \int_0^x f(x) dy \right) dx = \int_0^{\infty} \left( \int_y^{\infty} f(x) dx \right) dy$$

$$\Rightarrow E(X) = \int_0^{\infty} \left( \int_y^{\infty} f(x) dx \right) dy = \int_0^{\infty} P(X > y) dy = \int_0^{\infty} (1 - F(y)) dy.$$

$$\begin{aligned} E(T_a) &= \int_0^{\infty} (1 - F_{T_a}(t)) dt = \int_0^{\infty} \int_0^{\frac{a}{\sigma\sqrt{t}}} \frac{2}{\sqrt{2\pi}} e^{-z^2/2} dz dt \\ &= \int_0^{\infty} \int_0^{\frac{a^2}{\sigma^2 z^2}} \frac{2}{\sqrt{2\pi}} e^{-z^2/2} dt dz = \int_0^{\infty} \frac{2}{\sqrt{2\pi}} \frac{a^2}{\sigma^2 z^2} e^{-z^2/2} dz = \infty. \end{aligned}$$

This is not surprising, even though  $P(T_a < \infty) = 1$ , since a Brownian motion is the limit of a (null recurrent) symmetric random walk.

Let  $M(t) = \max_{0 \leq s \leq t} B(s)$ ,  $t > 0$ .

By continuity,  $P(M(t) \geq a) = P(\max_{0 \leq s \leq t} B(s) \geq a) = P(T_a \leq t)$ ,  $a > 0$ ,  $t \geq 0$ .

$$F_{M(t)}(a) = 1 - P(\max_{0 \leq s \leq t} B(s) \geq a) = 1 - P(T_a \leq t) = 1 - \int_{\frac{a}{\sigma\sqrt{t}}}^{\infty} \frac{2}{\sqrt{2\pi}} e^{-z^2/2} dz, \quad a > 0.$$

$$f_{M(t)}(a) = \frac{2}{\sqrt{2\pi}} e^{-a^2/(2\sigma^2 t)} \cdot \left( \frac{1}{\sigma\sqrt{t}} \right), \quad a > 0, \quad t \geq 0.$$

$\max_{0 \leq s \leq t} B(s)$  has the same probability distribution as  $|B(t)|$ .

Example: Let  $B(t)$  be a Brownian motion with parameter  $\sigma^2 = 1$ . Then the process  $X(t) = |B(t)|$ ,  $t \geq 0$ , is called Brownian motion reflected at the origin.

a) Find the density function of  $X(t)$ ,  $t \geq 0$ .

$$F_{X(t)}(x) = P(X(t) \leq x) = P(-x \leq B(t) \leq x) = F_{B(t)}(x) - F_{B(t)}(-x). \quad x > 0.$$

$$\begin{aligned} f_{X(t)}(x) &= F'_{X(t)}(x) = f_{B(t)}(x) + f_{B(t)}(-x) \\ &= \frac{1}{\sqrt{2\pi t}} e^{-x^2/2t} + \frac{1}{\sqrt{2\pi t}} e^{-(-x)^2/2t} \\ &= \frac{2}{\sqrt{2\pi t}} e^{-x^2/2t}, \quad x > 0. \end{aligned}$$

b) Find  $E[X(t)]$  and  $\text{Var}[X(t)]$ .

Hint: Compute  $E[X(t)]$  directly after  $u = x^2$  and  $E[(X(t))^2]$  by comparing the integral with the integral representing the variance of a random variable that is  $N(0, t)$ .

$$E[X(t)] = \int_0^{\infty} x \frac{2}{\sqrt{2\pi t}} e^{-x^2/2t} dx = \dots$$

$$u = \frac{x^2}{2t} \quad du = \frac{1}{t} x dx$$

$$\dots = \int_0^{\infty} \frac{2\sqrt{t}}{\sqrt{2\pi}} e^{-u} du = \frac{2\sqrt{t}}{\sqrt{2\pi}} = \frac{\sqrt{2t}}{\sqrt{\pi}}.$$

$$\begin{aligned} E[(X(t))^2] &= \int_0^{\infty} x^2 \frac{2}{\sqrt{2\pi t}} e^{-x^2/2t} dx = \int_{-\infty}^{\infty} x^2 \frac{1}{\sqrt{2\pi t}} e^{-x^2/2t} dx \\ &= E(W^2) = t, \quad \text{where } W \sim N(0, t). \end{aligned}$$

OR

$$E[(X(t))^2] = \int_0^{\infty} x^2 \frac{2}{\sqrt{2\pi t}} e^{-x^2/2t} dx = \dots$$

$$u = \frac{x^2}{2t} \quad du = \frac{1}{t} x dx$$

$$\begin{aligned} \dots &= \int_0^{\infty} \frac{2t}{\sqrt{\pi}} \sqrt{u} e^{-u} du = \frac{2t}{\sqrt{\pi}} \int_0^{\infty} u^{1/2} e^{-u} du = \frac{2t}{\sqrt{\pi}} \Gamma\left(\frac{3}{2}\right) \\ &= \frac{2t}{\sqrt{\pi}} \cdot \frac{1}{2} \Gamma\left(\frac{1}{2}\right) = t. \end{aligned}$$

$$\text{Var}[X(t)] = E[(X(t))^2] - [E(X(t))]^2 = t - \frac{2t}{\pi} = t \left(1 - \frac{2}{\pi}\right).$$

**2.** Let  $B(t)$  be a Brownian motion.

a)  $P(T_2 < T_{-1})$ .

$$P(T_2 < T_{-1}) = P(\text{up } A=2 \text{ before down } B=1) = \frac{B}{A+B} = \frac{1}{2+1} = \frac{\mathbf{1}}{\mathbf{3}}.$$

b)  $P(T_2 < T_{-1} < T_3)$ .

$$\begin{aligned} P(T_2 < T_{-1} < T_3) &= P(\text{first, up 2 before down 1,} \\ &\quad \text{then (from 2) down 3 (to } -1) \text{ before up 1 (to 3)}) \\ &= \frac{1}{2+1} \times \frac{1}{1+3} = \frac{\mathbf{1}}{\mathbf{12}}. \end{aligned}$$