

Variations and Brownian Motion with drift

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Reflected Brownian Motion

$$R(t) = \begin{cases} B(t) & \text{if } B(t) \geq 0 \\ -B(t) & \text{if } B(t) < 0 \end{cases}$$

$$\mathbb{E}(R(t)) = \sqrt{2t/\pi}$$

$$\text{Var}(R(t)) = \left(1 - \frac{2}{\pi}\right) t$$

$$\mathbb{P}\{R(t) \leq y | R(0) = x\} = \int_{-y}^y \phi_t(z - x) dx$$

$$p(y, t|x) = \phi_t(y - x) + \phi_t(y + x)$$



Today:

- ▶ Various variations of Brownian motion, reflected, absorbed, Brownian bridge, with drift, geometric

Next week

- ▶ General course overview



Absorbed Brownian Motion

The movement ceases once the level 0 is reached.

$$G_t(x, y) = \mathbb{P}\{A(t) > y | A(0) = x\}$$

$$= \mathbb{P}\{B(t) > y, \min\{B(u) > 0; 0 \leq u \leq t\} | B(0) = x\}$$

We first observe

$$\mathbb{P}\{B(t) > y | B(0) = x\} = G_t(x, y)$$

$$+ \mathbb{P}\{B(t) > y, \min\{B(u) \leq 0; 0 \leq u \leq t\} | B(0) = x\}$$

Due to reflection the latter term is also

$$\mathbb{P}\{B(t) > y, \min\{B(u) \leq 0; 0 \leq u \leq t\} | B(0) = x\}$$

$$= \mathbb{P}\{B(t) < -y, \min\{B(u) \leq 0; 0 \leq u \leq t\} | B(0) = x\}$$

$$= \mathbb{P}\{B(t) < -y | B(0) = x\}$$



Absorbed Brownian Motion

Summarizing we get

$$\begin{aligned} \mathbb{P}\{A(t) > y | A(0) = x\} &= G_t(x, y) \\ &= 1 - \Phi_t(y - x) - \Phi_t(-y - x) \\ &= \Phi_t(y + x) - \Phi_t(y - x) \end{aligned}$$

We have already seen that

$$\mathbb{P}\{A(t) = 0 | A(0) = x\} = 2[1 - \Phi_t(x)]$$



Brownian Bridge

Distribution of $B(t); 0 \leq t \leq 1$ conditioned on $\{B(0) = 0, B(1) = 0\}$.

$X = B(t)$ and $B(1) - B(t)$ are independent normal variables.

$Y = B(1) - B(t) + B(t)$ and X are bivariate normal.

$$\begin{aligned} \text{Var}(X) &= t \\ \text{Cov}(X, Y) &= \mathbb{E}(B(t)B(t) + B(t)(1 - B(t))) = t = \rho\sqrt{t}\sqrt{1} \\ \mathbb{E}(X | Y = y = 0) &= \mathbb{E}(X) + \rho \frac{\sigma_X}{\sigma_Y} \cdot (y - \mathbb{E}(Y)) = 0 \\ \text{Var}(X | Y) &= t(1 - \rho^2) = t(1 - t) \end{aligned}$$

$$\mathbb{E}\{B(t) | B(0) = 0, B(1) = 0\} = 0$$

$$\text{Var}\{B(t) | B(0) = 0, B(1) = 0\} = t(1 - t)$$

$$\text{Cov}\{B(s), B(t) | B(0) = 0, B(1) = 0\} = s(1 - t)$$

The process is Gaussian



Brownian Motion with Drift

$$X(t) = \mu t + \sigma B(t)$$

▶ A continuous sample path process with independent increments

▶ $X(t + s) - X(s) \sim N(\mu t, \sigma^2 t)$

$$\mathbb{P}\{X(t) \leq y | X(0) = x\} = \Phi\left(\frac{y - \mu t - x}{\sigma\sqrt{t}}\right)$$



Absorption Probabilities

$$T_{ab} = \min\{t \geq 0; X(t) = a \text{ or } X(t) = b\}$$

$$u(x) = \mathbb{P}\{X(T_{ab}) = b | X(0) = x\}$$

We assume that absorption has not occurred at $(\Delta t, x + \Delta X)$

$$u(x) = \mathbb{E}(u(x + \Delta X))$$

Taylor series expansion of right hand side

$$\begin{aligned} u(x) &= \mathbb{E}\left[u(x) + u'(x)\Delta(X) + \frac{1}{2}u''(x)(\Delta X)^2 + o((\Delta X)^2)\right] \\ &= u(x) + u'(x)\mathbb{E}[\Delta(X)] + \frac{1}{2}u''(x)\mathbb{E}[(\Delta X)^2] + \mathbb{E}\left[o((\Delta X)^2)\right] \end{aligned}$$

$$\mathbb{E}[\Delta(X)] = \mu\Delta t, \quad \mathbb{E}[(\Delta X)^2] = \sigma^2\Delta t + (\mu\Delta t)^2$$

$$\mu u'(x) + \frac{1}{2}\sigma^2 u''(x) = 0, \quad u(x) = A e^{-\frac{2\mu x}{\sigma^2}} + B$$



$$\mu u'(x) + \frac{1}{2}\sigma^2 u''(x) = 0, \quad u(x) = Ae^{-\frac{2\mu x}{\sigma^2}} + B$$

$$u(x) = Ae^{-\frac{2\mu x}{\sigma^2}} + B, \quad u(a) = 0, u(b) = 1$$

$$u(x) = \frac{e^{-\frac{2\mu x}{\sigma^2}} - e^{-\frac{2\mu a}{\sigma^2}}}{e^{-\frac{2\mu b}{\sigma^2}} - e^{-\frac{2\mu a}{\sigma^2}}}$$



$$T_{ab} = \min\{t \geq 0; X(t) = a \text{ or } X(t) = b\}$$

$$v(x) = \mathbb{E}\{T_{ab} | X(0) = x\}$$

$$v(x) = \Delta t + \mathbb{E}(v(x + \Delta X))$$

$$1 + \mu v'(x) + \frac{1}{2}\sigma^2 v''(x) = 0$$

$$v(a) = v(b) = 0$$

$$v(x) = \frac{1}{\mu} [u(x)(b - a) - (x - a)]$$



Maximum of Brownian Motion with Negative Drift

We assume $\mu < 0$, using Theorem 8.2.1

$$u(x) = \frac{e^{-\frac{2\mu x}{\sigma^2}} - e^{-\frac{2\mu a}{\sigma^2}}}{e^{-\frac{2\mu b}{\sigma^2}} - e^{-\frac{2\mu a}{\sigma^2}}}$$

we get

$$\mathbb{P}\{\max_{0 \leq t} X(t) > y\} = \lim_{a \rightarrow -\infty} \frac{e^{-\frac{2\mu \cdot 0}{\sigma^2}} - e^{-\frac{2\mu a}{\sigma^2}}}{e^{-\frac{2\mu y}{\sigma^2}} - e^{-\frac{2\mu a}{\sigma^2}}} = e^{-\frac{2|\mu|}{\sigma^2} y}$$



Geometric Brownian Motion

$$Z(t) = ze^{(\alpha - \frac{1}{2}\sigma^2)t + \sigma B(t)} \quad (Z(0) = z)$$

$$X(t) = \log(z) + \log(Z(t)), \quad X(t) \sim N\left(\alpha - \frac{1}{2}\sigma^2, \sigma^2\right)$$

$$\mathbb{E}(Z(t)) = ze^{(\alpha - \frac{1}{2}\sigma^2)t} \mathbb{E}\left[e^{\sigma B(t)}\right] = ze^{(\alpha - \frac{1}{2}\sigma^2)t} e^{\frac{1}{2}\sigma^2 t} = ze^{\alpha t}$$

For $\alpha < \frac{1}{2}\sigma^2$ we have that $Z(t) \rightarrow 0$ with probability 1

$$\text{Var}(Z(t)) = z^2 e^{2\alpha t} (e^{\sigma^2 t} - 1)$$

