## Solution of Geometric Brownian Motion SDE

December 14, 2023

## 1 The SDE

Geometric Brownian motion SDE:

$$\frac{dS(t)}{S(t)} = \mu dt + \sigma \sqrt{dt}z$$

$$z \sim \mathcal{N}(0, 1)$$
(1)

Raise to square:

$$dS(t)^{2} = (\mu S(t)dt + \sigma S(t)\sqrt{dt}z)^{2}$$

$$= \mu^{2}S(t)^{2}dt^{2} + 2\mu\sigma S(t)^{2}dt\sqrt{dt} + \sigma^{2}S(t)^{2}(\sqrt{dt})^{2}z^{2}$$
(2)

When  $dt \to 0$  we have  $dt^2 = 0$  and  $dt \sqrt{dt} = 0$ . Also since  $z \sim \mathcal{N}(0, 1)$  one can verify that on average  $z^2 = 1$ .

Therefore:

$$dS(t)^2 = \sigma^2 S(t)^2 dt \tag{3}$$

## 2 Stochastic Integrals

Take the following integral:

$$I = \int_0^t dW(t)$$

$$= \int_0^t \sigma \sqrt{dt}z$$

$$z \sim \mathcal{N}(0, 1)$$
(4)

Dividing the interval [0,t] into  $n\to\infty$  equal subintervals of equal size dt we have:

$$I = \sum_{i=0}^{n} \left( \sigma \sqrt{dt} z_i \right)$$

$$= \sigma \sqrt{dt} \sum_{i=0}^{n} z_i$$
(5)

Since we are adding n normal distributed random variables of mean zero and variance  $\sigma^2 dt$ , their sum will be still a normal distributed random variable of mean zero and variance  $n\sigma^2 dt$ . But ndt = t so finally we get the variance  $\sigma^t t$  so the deviation  $\sigma \sqrt{t}$ . Therefore:

$$I = W(t)$$

$$\int_{0}^{t} \sigma \sqrt{dt} z = \sigma \sqrt{t} z$$

$$z \sim \mathcal{N}(0, 1)$$
(6)

## 3 The solution

Take the function  $f(S,t) = \ln S(t)$ . We have it's Taylor series expansion:

$$df(S,t) \approx \frac{\partial f}{\partial t}dt + \frac{\partial f}{\partial S}dS + \frac{1}{2}\frac{\partial^2 f}{\partial t^2}dt^2 + \frac{\partial^2 f}{\partial t \partial S}dtdS + \frac{1}{2}\frac{\partial^2 f}{\partial S^2}dS^2$$
 (7)

We then have the partial derivatives:

$$\frac{\partial \ln S(t)}{\partial t} = 0$$

$$\frac{\partial \ln S(t)}{\partial S} = \frac{1}{S(t)}$$

$$\frac{\partial^2 \ln S(t)}{\partial t^2} = 0$$

$$\frac{\partial^2 \ln S(t)}{\partial S \partial t} = 0$$

$$\frac{\partial^2 \ln S(t)}{\partial S^2} = -\frac{1}{S(t)^2}$$
(8)

Replacing in the We Taylor series expansion we get:

$$d\left(\ln S(t)\right) \approx \frac{1}{S(t)} dS(t) - \frac{1}{S(t)^2} dS(t)^2 \tag{9}$$

By equation (3) we then have:

$$d(\ln S(t)) \approx \frac{dS(t)}{S(t)} - \frac{1}{2} \frac{\sigma^2 S(t)^2}{S(t)^2} dt \Leftrightarrow$$

$$d(\ln S(t)) = \frac{dS(t)}{S(t)} - \frac{\sigma^2}{2} dt$$
(10)

Replacing from (1)  $\frac{dS(t)}{S(t)}$  we get:

$$d(\ln S(t)) = \mu dt + \sigma \sqrt{dt}z - \frac{\sigma^2}{2}dt \Leftrightarrow$$

$$d(\ln S(t)) = \left(\mu - \frac{\sigma^2}{2}\right)dt + \sigma \sqrt{dt}z \qquad (11)$$

$$d(\ln S(t)) = \left(\mu - \frac{\sigma^2}{2}\right)dt + \sigma dW(t)$$

Where we have denoted the differential of the Weiner process  $dW(t) \approx \sqrt{dt}z$ . Integrating:

$$\int_{0}^{t} d\left(\ln S(t)\right) = \int_{0}^{t} \left(\mu - \frac{\sigma^{2}}{2}\right) dt + \int_{0}^{t} \sigma dW(t) \Rightarrow$$

$$\left.\ln S(t)\right|_{0}^{t} = \left(\mu - \frac{\sigma^{2}}{2}\right) t\Big|_{0}^{t} + \sigma W(t)\Big|_{0}^{t} \Leftrightarrow$$

$$\ln S(t) - \ln S(0) = \left(\mu - \frac{\sigma^{2}}{2}\right) t + \sigma W(t) \Leftrightarrow$$

$$\ln \frac{S(t)}{S(0)} = \left(\mu - \frac{\sigma^{2}}{2}\right) t + \sigma \sqrt{t}z$$

$$z \sim \mathcal{N}(0, 1)$$

$$(12)$$

Taking the exponential:

$$e^{\ln \frac{S(t)}{S(0)}} = e^{\left(\mu - \frac{\sigma^2}{2}\right)t + \sigma\sqrt{t}z} \Rightarrow$$

$$\frac{S(t)}{S(0)} = e^{\left(\mu - \frac{\sigma^2}{2}\right)t + \sigma\sqrt{t}z}$$
(13)

Which finally gives the solution:

$$S(t) = S(0)e^{\left(\mu - \frac{\sigma^2}{2}\right)t + \sigma\sqrt{t}z}$$

$$z \sim \mathcal{N}(0, 1)$$
(14)