# Black-Scholes Formula, a physicist's perspective

#### 1 (a)

Rewritten Using Brownian Motion:

$$ds(t) = \phi s(t)dt + \sigma s(t)dW(t). \tag{1}$$

where W(t) is a standard brownian motion.

To illustrate the relation between **Gaussian Noise** and **Brownian Motion**, consider when using R(t), we're actually suggesting  $s(t+\epsilon) = s(t) + \phi s(t)\epsilon + \sigma R\epsilon$ . In this case,  $R \sim \mathcal{N}(0,\frac{1}{\epsilon})$ , therefore  $R\epsilon \sim \mathcal{N}(0,\epsilon)$ , which can be characterized as  $W(t+\epsilon) - W(t)$ . As  $\epsilon \to 0$ ,  $W(t+\epsilon) - W(t) \to dW(t)$ . (It's really clearer to use *brownian motion* notation.) Brownian motion has the property that dW(t)dW(t) = dt, dW(t)dt = 0.

The original statement can be rewritten as:

$$df(t,s(t)) = f_t dt + \frac{1}{2}\sigma^2 s^2 f_{ss} dt + \phi s f_s dt + \sigma s dW(t).$$
 (2)

According to Taylor expansion formula, we can write

$$df = f_t dt + f_s ds + \frac{1}{2} \{ f_{tt} dt^2 + (f_{ts} + f_{st}) dt ds(t) + f_{ss} ds(t) ds(t) \} + o(dt^2)$$
 (3)

$$\Longrightarrow df = f_t dt + f_s ds + \frac{1}{2} f_{ss} ds(t) ds(t) \tag{4}$$

Considering  $ds(t) = \phi s(t) dt + \sigma s(t) dW(t), dW(t) dW(t) = dt, dW(t) dt = 0$ , we have

$$df = f_t dt + \frac{1}{2}\sigma^2 s^2 f_{ss} dt + \phi s f_s dt + \sigma s dW(t)$$
 (5)

Using the relationship between W(t) and R(t), we can change dW(t) and derive:

$$\frac{df}{dt} = \frac{\partial f}{\partial t} + \frac{1}{2}\sigma^2 s^2 \frac{\partial^2 f}{\partial s^2} + \frac{\partial f}{\partial s}(\phi s + \sigma s R). \tag{6}$$

2 (b)

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We've already known that c = c(t, s(t)). Consider a portfolio  $\Pi = c - \frac{\partial c}{\partial s}s$ . To calculate the derivative of  $\Pi$  we can write down the differential of  $\Pi$ :

$$d\Pi = c_t dt + c_s ds + \frac{1}{2} c_{ss} ds ds - c_s ds. \tag{7}$$

Because we've got  $dS^2 = \sigma^2 s^2 dt$ , we can derive:

$$\frac{d\Pi}{dt} = c_t + \frac{1}{2}c_{ss}\sigma^2 s^2. \tag{8}$$

## 3 (c)

r means the short-term risk-free interest rate. The following equation must be true:

$$\frac{d\Pi}{dt} = r\Pi. \tag{9}$$

Otherwise arbitrage will exist, which contradicts our assumption. Therefore we can write  $d\Pi = r(c-c_s s)dt$ . It's equivalent to the previous result  $c_t dt + \frac{1}{2}c_{ss}\sigma^2 s^2 dt$ .

At last we have the **Black-Scholes formula**:

$$c_t + \frac{1}{2}\sigma^2 s^2 c_{ss} + rsc_s - rc = 0. (10)$$

#### **Another Way of Obtaining B-S Formula**

According to non-arbitrage postulate, if at time 0,  $c(0, S(0)) = c_0$ , then we should should be able to construct a portfolio X(t) (with  $X(0) = c_0$ ) to replicate exactly this option. We should use the underlying stock S(t) and the money market with interest rate r. Suppose we hold  $\Delta(t)$  share of stock at time t, then it follows:

$$dX(t) = \Delta(t)dS(t) + r(X(t) - \Delta(t)S(t))dt = D_t dt + D_w dW(t)$$
(11)

where  $D_t = \Delta(t)S(t)(\phi - r) + rX(t)$ ,  $D_w = \Delta(t)S(t)\sigma$ .

At the same time,

$$dc(t, S(t)) = c_t dt + c_s dS(t) + \frac{1}{2} c_{ss} dS(t) dS(t) = D'_t dt + D'_w dW(t)$$
 (12)

where  $D'_t = c_t + c_s \phi S(t) + \frac{1}{2} \sigma^2 S(t)^2 c_{ss}, D'_w = c_s \sigma S(t)$ .

Therefore, it should follows that  $D_t = D'_t, D_w = D'_w$ . The second relation yields instantly that  $\Delta(t) = c_s$ , while the first would amount to the Black-Scholes formula:

$$c_t + \frac{1}{2}\sigma^2 s^2 c_{ss} + rsc_s - rc = 0$$
 (13)

4 (d)

#### 4 (d)

Change variable  $s = e^x$ , we have:

$$c_x = e^{-x}c_s. (14)$$

$$c_t = rc - rsc_s - \frac{1}{2}\sigma^2 s^2 c_{ss} = \left(r - \left(r - \frac{1}{2}\sigma^2\right)\frac{\partial}{\partial x} - \frac{1}{2}\sigma^2\frac{\partial^2}{\partial x^2}\right)c. \tag{15}$$

Therefore we can easily prove that

$$H_{BS} = \left(1 - \frac{\partial}{\partial x}\right)\left(r + \frac{1}{2}\sigma^2\frac{\partial}{\partial x}\right) = -\frac{\sigma^2}{2}\frac{\partial^2 c}{\partial x^2} + \left(\frac{\sigma^2}{2} - r\right)\frac{\partial c}{\partial x} + rc. \tag{16}$$

which is the Hamiltonian for Black-Scholes model.

#### 5 (e)

$$p_{BS}(x,\tau;x') = \langle x|e^{-\tau H}|x'\rangle = \int_{-\infty}^{\infty} \frac{dp}{2\pi} \langle x|e^{-\tau H}|p\rangle \langle p|x'\rangle$$
 (17)

Taking  $p = i \frac{\partial}{\partial x}$ , using  $\langle x | p \rangle = e^{ipx}$ :

$$p_{BS}(x,\tau;x') = e^{-r\tau} \int_{\infty}^{\infty} \frac{dp}{2\pi} exp\{-\frac{1}{2}\sigma^2 p^2 \tau + ip(x-x') + ip\tau(r-\frac{\sigma^2}{2})\}$$
(18)

Finally, perform the Gaussian integration:

$$p_{BS}(x,\tau;x') = e^{-r\tau} \frac{1}{\sqrt{2\pi\tau\sigma^2}} exp\{-\frac{1}{2\sigma^2\tau} (x - x' + \tau(r - \frac{\sigma^2}{2}))^2\}$$
(19)

## 6 (f)

We've known that P(x,T-t,x') is the conditional probability density that, given security price x at time t, it will have a value of x' at time T.

The expectation of x' at time  $\tau$  can therefore be calculated (it's a simple convolution):

$$\langle x(\tau) \rangle = \int_{-\infty}^{+\infty} x' P_{BS}(x, \tau; x') dx' = \frac{e^{-r\tau}}{\sqrt{2\pi\tau\sigma^2}} \int_{-\infty}^{+\infty} x' e^{-\frac{1}{2\tau\sigma^2} [x - x' + \tau(r - \frac{\sigma^2}{2})]^2} dx'.$$
(20)

Finally we can derive:

$$\langle x(\tau) \rangle = [x + \tau(r - \frac{\sigma^2}{2})]e^{-r\tau}.$$
 (21)

The result tells us the evolution of x( which equals ln(S)) over time.

7 (i)

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We know that r(t) satisfied Langevin equation:

$$\frac{dr}{dt} = a(r,t) + \sigma(r,t)R(t). \tag{22}$$

Here R(t) is still Gaussian noise.

Define the propagator  $P(r, t; r_0)$ : if  $r(t_0) = r_0$ , the probability of r(t)=r equals  $P(r, t; r_0)$ .

From the Langevin equation we have:

$$r(t+\varepsilon) = r(t) + \varepsilon[a + \sigma R(t)]. \tag{23}$$

Change it into the following formula:

$$r = r' + \varepsilon [a(r') + \sigma(r')R(t)]. \tag{24}$$

Thus we can calculate the propagator:

$$P(r,t+\varepsilon,r_0) = P(r',t;r_0)|_{r'=r-\varepsilon[a(r')+\sigma(r')R(t)]} = \int P(r',t;r_0)\delta(r-r'-\varepsilon[a(r')+\sigma(r')R(t)])dr' \simeq \int P(r',t;r_0)\delta(r-r'-\varepsilon[a(r')+\sigma(r')R(t)]$$

Because  $\langle R^2(t) \rangle = \frac{1}{\varepsilon}$  and  $\langle R(t) \rangle = 0$ , the previous formula can be rewritten like this:

$$P(r,t+\varepsilon,r_0) = P(r',t;r_0) + \int dr' P(r',t;r_0) \left\{ \frac{\partial \delta(r-r')}{\partial r'} \varepsilon a(r') + \frac{1}{2} \frac{\partial^2 \delta(r-r')}{\partial r'^2} \varepsilon^2 \sigma^2(r') \frac{1}{\varepsilon} \right\} = P(r',t;r_0) - \frac{1}{2} \frac{\partial^2 \delta(r-r')}{\partial r'^2} \left\{ \frac{\partial \delta(r-r')}{\partial r'} \varepsilon a(r') + \frac{1}{2} \frac{\partial^2 \delta(r-r')}{\partial r'^2} \varepsilon^2 \sigma^2(r') \frac{1}{\varepsilon} \right\} = P(r',t;r_0) - \frac{1}{2} \frac{\partial^2 \delta(r-r')}{\partial r'} \left\{ \frac{\partial \delta(r-r')}{\partial r'} \varepsilon a(r') + \frac{1}{2} \frac{\partial^2 \delta(r-r')}{\partial r'^2} \varepsilon^2 \sigma^2(r') \frac{1}{\varepsilon} \right\} = P(r',t;r_0) - \frac{1}{2} \frac{\partial^2 \delta(r-r')}{\partial r'} \left\{ \frac{\partial \delta(r-r')}{\partial r'} \varepsilon a(r') + \frac{1}{2} \frac{\partial^2 \delta(r-r')}{\partial r'^2} \varepsilon^2 \sigma^2(r') \frac{1}{\varepsilon} \right\} = P(r',t;r_0) - \frac{1}{2} \frac{\partial^2 \delta(r-r')}{\partial r'} \left\{ \frac{\partial \delta(r-r')}{\partial r'} \varepsilon a(r') + \frac{1}{2} \frac{\partial^2 \delta(r-r')}{\partial r'^2} \varepsilon^2 \sigma^2(r') \frac{1}{\varepsilon} \right\} = P(r',t;r_0) - \frac{1}{2} \frac{\partial^2 \delta(r-r')}{\partial r'} \left\{ \frac{\partial \delta(r-r')}{\partial r'} \varepsilon a(r') + \frac{1}{2} \frac{\partial^2 \delta(r-r')$$

If a variable is  $o(\varepsilon)$ , it is automatically neglected.

Thus, from the definition of derivative, we have:

$$\frac{\partial P(r,t;r_0)}{\partial t} = \left[\frac{1}{2}\frac{\partial^2}{\partial r^2}\sigma^2(r) - \frac{\partial}{\partial r}a(r)\right]P(r,t;r_0). \tag{27}$$

We've already known that:

$$\frac{\partial P(r,t;r_0)}{\partial t} = -H_F P(r,t;r_0). \tag{28}$$

Therefore we can prove:

$$H_F = -\frac{1}{2}\frac{\partial^2}{\partial r^2}\sigma^2(r) + \frac{\partial}{\partial r}a(r) = -\frac{1}{2}\frac{\partial^2}{\partial r^2}\sigma^2(r) + a(r)\frac{\partial}{\partial r} + \frac{\partial a(r)}{\partial r}.$$
 (29)

8 (k)

From a different perspective, we define  $P_B(R, t; r)$  as the back propagator. Similarly we have (since the time flows backwards this time):

$$\frac{\partial P_B(R,t;r)}{\partial t} = +H_B P_B(R,t;r). \tag{30}$$

Finally we can calculate:

$$H_B = -\frac{1}{2}\sigma^2(r)\frac{\partial^2}{\partial r^2} - a(r)\frac{\partial}{\partial r}.$$
 (31)

 $H_B = H_F^{\dagger}$  is obvious.

## 8 (k)

The Vasicek model can be described using the following equation:

$$\frac{dr}{dt} = a(b-r) + \sigma R(t). \tag{32}$$

Note that if we set  $\begin{cases} a(r,t) = a(b-r). \\ \sigma(r,t) = \sigma. \end{cases}$  in Langevin equation  $(a,b,\sigma)$  are all constants, we have the Vasicek model naturally.

Thus we get:

$$L_V = -\frac{\left[\frac{dr}{dt} - a(b-r)\right]^2}{2\sigma^2}.$$
 (33)

$$S_V = -\frac{1}{2\sigma^2} \int_{t_0}^{T} \left[ \frac{dr}{dt} - a(b-r) \right]^2 dt.$$
 (34)

In the next section we'll use  $S_V$  to calculate the propagator of Vasicek model.