Black-Scholes Formula, a physicist's perspective

1 (a)

Rewritten Using Brownian Motion:

 $ds(t) = \phi s(t)dt + \sigma s(t)dW(t)$, 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:

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

According to Taylor expansion formula, we can write

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

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

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

$$df = f_t dt + \frac{1}{2}\sigma^2 s^2 f_{ss} dt + \phi s f_s dt + \sigma s dW(t)$$
(3)

2 (b)

$$c = c(t, s(t)), d\Pi = c_t dt + c_s ds + \frac{1}{2}c_{ss} ds ds - c_s ds.$$

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

3 (c)

3 (c)

$$d\Pi = r(c - c_s s)dt = c_t dt + \frac{1}{2}c_{ss}\sigma^2 s^2 dt$$

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

Another Way of Obtaining BS 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:

$$dX(t) = \Delta(t)dS(t) + r(X(t) - \Delta(t)S(t))dt = D_t dt + D_w dW(t)$$
(4)

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

At the same time,

$$dc(t, S(t)) = c_t dt + c_s dS(t) + \frac{1}{2} c_{ss} dS(t) dS(t) = D'_t dt + D'_w dW(t)$$
 (5)

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:

$$c_t + \frac{1}{2}\sigma^2 s^2 c_{ss} + rsc_s - rc = 0$$
(6)

4 (d)

Change variable
$$s=e^x$$
, we have $c_x=e^{-x}c_s$. $c_t=rc-rsc_s-\frac{1}{2}\sigma^2s^2c_{ss}=(r-(r-\frac{1}{2}\sigma^2)\frac{\partial}{\partial x}-\frac{1}{2}\sigma^2\frac{\partial^2}{\partial x^2})c$. Therefore $H_{BS}=(1-\frac{\partial}{\partial x})(r+\frac{1}{2}\sigma^2\frac{\partial}{\partial x})$.

5 (e)

$$p_{BS}(x,\tau;x') = \langle x|e^{-\tau H}|x'\rangle = \int_{\infty}^{\infty} \frac{dp}{2\pi} \langle x|e^{-\tau H}|p\rangle \langle p|x'\rangle \tag{7}$$

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

$$p_{BS}(x,\tau;x') = e^{-r\tau} \int_{-\infty}^{\infty} \frac{dp}{2\pi} exp\{-\frac{1}{2}\sigma^2 p^2 \tau + ip(x-x') + ip\tau(r-\frac{\sigma^2}{2})\}$$
 (8)

6 (g)

Finally, perform the Gaussian integration:

$$p_{BS}(x,\tau;x') = e^{-r\tau} \frac{1}{\sqrt{2\pi\tau\sigma^2}} exp\{-\frac{1}{2\sigma^2\tau} (x - x' + \tau(r - \frac{\sigma^2}{2}))^2\}$$
(9)

section(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 τ can therefore be calculated (it's a simple convolution):

$$\langle x(\tau) \rangle = \int_{-\infty}^{+\infty} x' P_{BS}(x, \tau; x') dx' = \frac{e^{-r\tau}}{\sqrt{2\pi\tau\sigma^2}} \int_{-\infty}^{+\infty} x' e^{-\frac{1}{2\tau\sigma^2} [x - x' + \tau(r - \frac{\sigma^2}{2})]^2} dx'.$$
(10)

Finally we can derive:

$$\langle x(\tau) \rangle = [x + \tau(r - \frac{\sigma^2}{2})]e^{-r\tau}.$$
 (11)

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

6 (g)

Consider a down-and-out barrier European option. If $s \leq e^B(x \leq B)$ it will become worthless, which means c = 0.

For an arbitrary potential v(x), we have: $\frac{\partial c}{\partial t} = Hc$, where H can be written as:

$$H = -\frac{\sigma^2}{2} \frac{\partial^2}{\partial x^2} + (\frac{\sigma^2}{2} - r) \frac{\partial}{\partial x} + V(x). \tag{12}$$

Obviously, if we set $V(x) = \infty$ when $x \leq B$, then c = 0 is automatically satisfied. However, when x > B, the changing of c still conform to the B-S formula. comparing with H_{BS} , we have: $V(x) = \begin{cases} +\infty, x \leq B. \\ r, x > B. \end{cases}$

7 (i)

We know that r(t) satisfied Langevin equation:

$$\frac{dr}{dt} = a(r,t) + \sigma(r,t)R(t). \tag{13}$$

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

7 (i) 4

 $P(r,t;r_0)$.

From the Langevin equation we have:

$$r(t+\varepsilon) = r(t) + \varepsilon[a + \sigma R(t)]. \tag{14}$$

Change it into the following formula:

$$r = r' + \varepsilon [a(r') + \sigma(r')R(t)]. \tag{15}$$

Thus we can calculate the propagator:

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

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

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

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

Thus, from the definition of derivative, we have:

$$\frac{\partial P(r,t;r_0)}{\partial t} = \left[\frac{1}{2}\frac{\partial^2}{\partial r^2}\sigma^2(r) - \frac{\partial}{\partial r}a(r)\right]P(r,t;r_0). \tag{18}$$

We've already known that:

$$\frac{\partial P(r,t;r_0)}{\partial t} = -H_F P(r,t;r_0). \tag{19}$$

Therefore we can prove:

$$H_F = -\frac{1}{2}\frac{\partial^2}{\partial r^2}\sigma^2(r) + \frac{\partial}{\partial r}a(r) = -\frac{1}{2}\frac{\partial^2}{\partial r^2}\sigma^2(r) + a(r)\frac{\partial}{\partial r} + \frac{\partial a(r)}{\partial r}.$$
 (20)

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):

$$\frac{\partial P_B(R,t;r)}{\partial t} = +H_B P_B(R,t;r). \tag{21}$$

Finally we can calculate:

$$H_B = -\frac{1}{2}\sigma^2(r)\frac{\partial^2}{\partial r^2} - a(r)\frac{\partial}{\partial r}.$$
 (22)

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

8 (j) 5

(j) 8

In this part we'll focus on the so-called "stochastic Quantization". The Langevin equation:

$$\frac{dr}{dt} = a(r,t) + \sigma(r,t)R(t) \tag{23}$$

is satisfied at any time between t_0 and T. We must consider that both r and R are stochastic variables. Therefore when calculating z_B (can be compared to partition function in statistical mechanics), we must integrate over all possible paths

$$Z_B = \int DRDr \prod_{t=t_0}^{T} \delta[\frac{dr}{dc} - a(r,t) - \sigma(r,t)R(t)]e^{-\frac{1}{2}} \int_{t_0}^{T} R^2(t)dt$$
 (24)

Dr means integrating over all possible $r(t): \int Dr = \int_{-\infty}^{+\infty} \prod_{t=t_0}^{T} dr(t)$.

We hope to write Z_B int the form of $z_B = \int Dre^{s_B}$. Because of this, we calculate the integral over R first:

$$Z_{B} = \int Drexp(-\frac{1}{2} \int_{t_{0}}^{T} \frac{\left[\frac{dr}{dt} - a(r,t)\right]^{2}}{\sigma^{2}(r,t)} dt).$$
 (25)

Because the $Dirac - \delta$ function makes sure we only keep R(t) that satisfies the Langevin equation: $R(t) = \frac{\frac{dr}{dt} - a(r,t)}{\sigma(r,t)}$. It's easy to calculate S_B :

$$S_B = -\frac{1}{2} \int_{t_0}^T \frac{\left[\frac{dr}{dt} - a(r,t)\right]^2}{\sigma^2(r,t)} dt.$$
 (26)

From the definition of $S_B: S_B = \int_{t_0}^T L dt$, it's clear that:

$$L = -\frac{\left[\frac{dr}{dt} - a(r,t)\right]^2}{2\sigma^2(r,t)}dt.$$
 (27)

9 (k)

The Vasicek model can be described using the following equation:

$$\frac{dr}{dt} = a(b-r) + \sigma R(t). \tag{28}$$

9 (k)

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

Thus we get:

$$L_V = -\frac{\left[\frac{dr}{dt} - a(b-r)\right]^2}{2\sigma^2}.$$
 (29)

$$S_V = -\frac{1}{2\sigma^2} \int_{t_0}^{T} \left[\frac{dr}{dt} - a(b-r) \right]^2 dt.$$
 (30)

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