Homework #5 Stochastic Calculus Solutions Black-Scholes and Feyman-Kac

Note: The calculations in the solutions have not been thoroughly checked and may contain errors. Use the solutions as an extended hint on solving the problems.

Requirement: For the following problems, you need to provide the dynamic replication/hedging argument to derive the pricing partial equation, invoke the Feyman-Kac to obtain the solution, and then evaluate the expection to obtain the final answer.

Problem 1 [Pricing the log Contract]. Suppose we have the assumptions of the Black-Scholes Model. Find the pricing formula for the European Style derivative whose payoff is given by function $log(S_T)$, where S_T is the stock price in the BS model on date T.

Answer: In the BS model assumes the the risk-free bond which follows

$$dB_t = rB_t dt,$$

where r is the *continuously compounded* interest rate. And a stock whose price follows Geometric BM [Note that the drift is μ]:

$$dS_t = \mu S_t dt + \sigma S_t dW_t.$$

Note that our log contract's payoff is $\log(S_T)$. Our goal is to derive the pricing formula for this contract. We achieve this by assuming the formula is given by a function C(x,t) (which gives the price of the log contract to be $C(S_t,t)$ when the stock price is S_t at time t), and then find what conditions C(x,t) has to satisfy.

With this assumed pricing formula $C(S_t, t)$, we know the dynamics of the price of this log contract is given by Ito's formula as follows

$$dC(S_{t},t) = C_{t}(S_{t},t) dt + C_{x}(S_{t},t) dS_{t} + \frac{1}{2}C_{xx}(S_{t},t) \sigma^{2}S_{t}^{2}dt$$

$$= C_{t}(S_{t},t) dt + C_{x}(S_{t},t) [\mu S_{t}dt + \sigma S_{t}dW_{t}] + \frac{1}{2}C_{xx}(S_{t},t) \sigma^{2}S_{t}^{2}dt$$

$$= \left[C_{t}(S_{t},t) + C_{x}(S_{t},t) \mu S_{t} + \frac{1}{2}C_{xx}(S_{t},t) \sigma^{2}S_{t}^{2}\right] dt + C_{x}(S_{t},t) \sigma S_{t}dW_{t}.$$

Now we creat a riskless portfolio by going long one of this log contract and shorting $C_x(S_t, t)$ share of stock. The dynamics of this portfolio is given by

$$dC(S_{t},t) - C_{x}(S_{t},t) dS_{t} = \left[C_{t}(S_{t},t) + C_{x}(S_{t},t) \mu S_{t} + \frac{1}{2} C_{xx}(S_{t},t) \sigma^{2} S_{t}^{2} \right] dt + C_{x}(S_{t},t) \sigma S_{t} dW_{t}$$

$$-C_{x}(S_{t},t) \mu S_{t} dt - C_{x}(S_{t},t) \sigma S_{t} dW_{t}$$

$$= \left[C_{t}(S_{t},t) + \frac{1}{2} C_{xx}(S_{t},t) \sigma^{2} S_{t}^{2} \right] dt$$

Note that the dW_t term vanished, and hence this portfolio is instantaneously riskless. Hence it should earn risk-free rate if we assume no-arbitrage or the Law of One Price:

$$\frac{\left[C_{t}\left(S_{t},t\right)+\frac{1}{2}C_{xx}\left(S_{t},t\right)\sigma^{2}S_{t}^{2}\right]}{C\left(S_{t},t\right)-C_{x}\left(S_{t},t\right)S_{t}}=r$$

which yields the Black-Scholes PDE for derivatives pricing: [Note that this equation is the same for any derivatives.]

$$C_t(S_t, t) + rS_tC_x(S_t, t) + \frac{1}{2}\sigma^2 S_t^2 C_{xx}(S_t, t) = rC(S_t, t).$$
 (1)

Of course, on the maturity date T, $C(S_t,t)$ is equal to the payoff function which is a boundary condition

$$C(S_T, t) = \log(S_T). \tag{2}$$

The above equations (1) and (2) are written in terms of S_t . When S_t runs through all sample paths, the (1) and (2) is equivalent to

$$C_t(x,t) + rxC_x(x,t) + \frac{1}{2}\sigma^2 x^2 C_{xx}(x,t) = rC(x,t).$$
 (3)

with the boundary condition

$$C(x,t) = \log(x). \tag{4}$$

The PDE in (3) and (4) can be solved in Feynman-Kac Theorem. The Feynman-Kac Theorem says that the solutions to (3) and (4) is given by the following expectation

$$C(x,t) = E^{x,t} \left[e^{-r(T-t)} \log(X_t) \right]$$
(5)

where the process X_t solves the following SDE [How are the drift and diffusion coefficients obtained? Note that the drift term is r now from (3) and not μ !]

$$dX_t = rX_t dt + \sigma X_t dW_t.$$

But such X_t is a geometric BM what can be easily solved:

$$\log(X_T) - \log X_t = (r - \frac{1}{2}\sigma^2)(T - t) + \sigma [W_T - W_t]$$

This means that $\log(X_T)$ has normal distribution with mean $\log x + (r - \frac{1}{2}\sigma^2)(T - t)$, and variance of $\sigma^2(T - t)$ conditional on $X_t = x$:

$$\log(X_T) \sim \log x + (r - \frac{1}{2}\sigma^2)(T - t) - \sigma\sqrt{T - t}Z.$$

[here Z is the standard normal random variable] It follows that

$$C(x,t) = E^{x,t} \left[e^{-r(T-t)} \log(X_t) \right]$$

$$= e^{-r(T-t)} E^{x,t} \left[\log(X_T) \right]$$

$$= e^{-r(T-t)} \int \left[\log x + (r - \frac{1}{2}\sigma^2)(T-t) - \sigma\sqrt{T-t}z \right] \frac{1}{\sqrt{2\pi}} e^{-\frac{z^2}{2}} dz$$

$$= e^{-r(T-t)} \left[\log x + (r - \frac{1}{2}\sigma^2)(T-t) \right].$$

So the log contract's price at time t is given by

$$C(S_t, t) = e^{-r(T-t)} \left[\log S_t + (r - \frac{1}{2}\sigma^2)(T-t) \right],$$

where S_t is the underlying stock price at time t.

Problem 2 [Bonus][Pricing the Variance Contract]. Suppose we have the assumptions of the Black-Scholes Model. Find the pricing formula for the European Style derivative whose payoff is given by function $[\max(S_T - K, 0]^2]$, where S_T is the stock price in the BS model on date T.

Answer: Repeating the argument in Problem 1, we are lead to that the price function C(x,t) for this variance contract satisfies

$$C_{t}\left(x,t\right)+rxC_{x}\left(x,t\right)+\frac{1}{2}\sigma^{2}x^{2}C_{xx}\left(x,t\right)=rC\left(x,t\right).$$

with the boundary condition

$$C(x,t) = [\max(x - K, 0)]^2$$
.

The above PDE can be solved by Feyman-Kac Theorem. The Feyman-Kac Theorem says that the solution is given by the following expectation

$$C(x,t) = E^{x,t} \left[e^{-r(T-t)} \left[\max(X_T - K, 0) \right]^2 \right]$$

where the process X_t solves the following SDE [How are the drift and diffusion coefficients obtained? Note that the drift term is r and not μ !]

$$dX_t = rX_t dt + \sigma X_t dW_t$$

with the initial condition $X_t = x$. But such an X_t is a geometric BM what can be easily solved (c.f. lecture notes and prior HW):

$$\log(X_T) - \log X_t = (r - \frac{1}{2}\sigma^2)(T - t) + \sigma [W_T - W_t]$$

This means that $\log(X_T)$ has normal distribution with mean $\log x + (r - \frac{1}{2}\sigma^2)(T - t)$, and variance of $\sigma^2(T - t)$ conditional on $X_t = x$ [here Z is the standard normal random variable]:

$$\log(X_T) \sim \log x + (r - \frac{1}{2}\sigma^2)(T - t) - \sigma\sqrt{T - t}Z$$

It follows that

$$\begin{split} C(x,t) &= E^{x,t} \left[e^{-r(T-t)} \left[\max(X_T - K, 0) \right]^2 \right] \\ &= e^{-r(T-t)} \int \left[\max(X_T - K, 0) \right]^2 \frac{1}{\sqrt{2\pi}} e^{-\frac{z^2}{2}} dz \\ &= e^{-r(T-t)} E^{x,t} \left[\max\left(e^{\log x + (r - \frac{1}{2}\sigma^2)(T-t) - \sigma\sqrt{T-t}Z} - K, 0 \right) \right]^2 \\ &= e^{-r(T-t)} \int_{z < d^-} \left(e^{\left[\log x + (r - \frac{1}{2}\sigma^2)(T-t) \right] - \sigma\sqrt{T-t}Z} - K \right)^2 \frac{1}{\sqrt{2\pi}} e^{-\frac{z^2}{2}} dz \\ &= e^{-r(T-t)} \int_{z < d^-} e^{2\left[\log x + (r - \frac{1}{2}\sigma^2)(T-t) \right] - 2\sigma\sqrt{T-t}Z} \frac{1}{\sqrt{2\pi}} e^{-\frac{z^2}{2}} dz \\ &- 2Ke^{-r(T-t)} \int_{z < d^-} e^{\left[\log x + (r - \frac{1}{2}\sigma^2)(T-t) \right] - \sigma\sqrt{T-t}Z} \frac{1}{\sqrt{2\pi}} e^{-\frac{z^2}{2}} dz \\ &+ K^2 e^{-r(T-t)} \int_{z < d^-} \frac{1}{\sqrt{2\pi}} e^{-\frac{z^2}{2}} dz \\ &\equiv I + II + III. \end{split}$$

The three terms can be evaluated easily using properties of normal distribution.

$$\begin{split} I &= e^{-r(T-t)} \int_{z < d^{-}} e^{2\left[\log x + (r - \frac{1}{2}\sigma^{2})(T-t)\right] - 2\sigma\sqrt{T-t}z} \frac{1}{\sqrt{2\pi}} e^{-\frac{z^{2}}{2}} dz \\ &= e^{-r(T-t)} \int_{z < d^{-}} e^{2\left[\log x + (r - \frac{1}{2}\sigma^{2})(T-t)\right] + 2\sigma^{2}(T-t)} \frac{1}{\sqrt{2\pi}} e^{-\frac{(z+2\sigma\sqrt{T-t})^{2}}{2}} dz \\ &= e^{-r(T-t)} \cdot e^{2\left[\log x + (r - \frac{1}{2}\sigma^{2})(T-t)\right] + 2\sigma^{2}(T-t)} \int_{z < d^{-}} \frac{1}{\sqrt{2\pi}} e^{-\frac{(z+2\sigma\sqrt{T-t})^{2}}{2}} dz \\ &= x^{2} e^{\sigma^{2}(T-t)} \int_{y < d^{-} + 2\sigma\sqrt{T-t}} \frac{1}{\sqrt{2\pi}} e^{-\frac{y^{2}}{2}} dy \\ &= x^{2} e^{\sigma^{2}(T-t)} N(d^{-} + 2\sigma\sqrt{T-t}), \end{split}$$

where d^- is given by

$$d^{-} = \frac{\log(x/K) + (r - \frac{1}{2}\sigma^{2}(T - t))}{\sigma\sqrt{(T - t)}}.$$

$$\begin{split} II &= -2Ke^{-r(T-t)} \int_{z < d^{-}} e^{\left[\log x + (r - \frac{1}{2}\sigma^{2})(T-t)\right] - \sigma\sqrt{T-t}z} \frac{1}{\sqrt{2\pi}} e^{-\frac{z^{2}}{2}} dz \\ &= -2Ke^{-r(T-t)} \int_{z < d^{-}} e^{\left[\log x + (r - \frac{1}{2}\sigma^{2})(T-t)\right] + \frac{1}{2}\sigma^{2}(T-t)} \frac{1}{\sqrt{2\pi}} e^{-\frac{(z + \sigma\sqrt{T-t})^{2}}{2}} dz \\ &= -2Ke^{-r(T-t)} \cdot e^{\left[\log x + (r - \frac{1}{2}\sigma^{2})(T-t)\right] + \frac{1}{2}\sigma^{2}(T-t)} \int_{z < d^{-}} \frac{1}{\sqrt{2\pi}} e^{-\frac{(z + \sigma\sqrt{T-t})^{2}}{2}} dz \\ &= -2Kx \int_{y < d^{-} + \sigma\sqrt{T-t}} \frac{1}{\sqrt{2\pi}} e^{-\frac{y^{2}}{2}} dy \\ &= -2KxN(d^{-} + \sigma\sqrt{T-t}). \end{split}$$

$$III = K^{2}e^{-r(T-t)} \int_{z< d^{-}} \frac{1}{\sqrt{2\pi}} e^{-\frac{z^{2}}{2}} dz$$
$$= K^{2}e^{-r(T-t)} N(d^{-}).$$

Hence the pricing function is given by

$$C(x,t) = x^{2}e^{\sigma^{2}(T-t)}N(d^{-} + 2\sigma\sqrt{T-t}) - 2KxN(d^{-} + \sigma\sqrt{T-t}) + K^{2}e^{-r(T-t)}N(d^{-}),$$

where d^- is defined in the Black-Scholes option pricing formula

$$d^{-} = \frac{\log(x/K) + (r - \frac{1}{2}\sigma^{2}(T - t))}{\sigma\sqrt{(T - t)}}.$$

The derivative's price at time t when the underlying stock price is S_t is just $C(S_t, t)$.