Kapitel 8: Stochastic Calculus und Payoff Replication im Black-Scholes Modell, Teil1

In the last chapter we derived the Black-Scholes equation by considering the recursion relations of the replicating portfolio in the approximating Binomial model and then we took the continuous time limit. In this chapter we ask the following question: Is it possible to derive the Black-Scholes equation directly from the Black-Scholes model

$$dS_t/S_t = \mu dt + \sigma dx_t \tag{1}$$

in continuous time, without using the approximating Binomial model? The answer is yes. In the following, we will use the Ito-formula to make the appropriate calculations. For simplicity, we start with zero rates, r = 0.

In the first chapter we saw that the portfolio value V_{t_k} of a selffinancing strategy, which holds $\delta_{t_{k-1}}$ stocks 'at the end of time t_{k-1} ' or 'at the beginning of time t_k ' and readjusts this to δ_{t_k} stocks 'at the end of time t_k after the asset price has switched from $S_{t_{k-1}}$ to S_{t_k} , is given by

$$V_{t_k} = V_0 + \sum_{j=1}^k \delta_{t_{j-1}} \cdot (S_{t_j} - S_{t_{j-1}}) = V_{t_{k-1}} + \delta_{t_{k-1}} \cdot (S_{t_k} - S_{t_{k-1}})$$
 (2)

In continuous time with 'continuous trading' this may be rewritten as a stochastic integral, as an Ito-integral

$$V_t = V_0 + \int_0^t \delta_\tau \, dS_\tau \tag{3}$$

or in differential form, if we subtract the $V_{t_{k-1}}$ -term on the right hand side (2),

$$dV = \delta dS \tag{4}$$

where dV is the limit of

$$V_t(S_t) - V_{t-\Delta t}(S_{t-\Delta t}) = V(S_t, t) - V(S_{t-\Delta t}, t - \Delta t) \stackrel{\Delta t \to 0}{\to} dV$$
 (5)

Since we have $V = V(S_t, t)$ and S_t is a stochastic quantity, we have to use the Ito-Formula, the differential version of the Ito-Formula, to calculate the dV. Let's start by recalling the calculation rules for the Brownian motion,

$$(dx_t)^2 = dt$$

$$dx_t dt = 0$$

$$(dt)^2 = 0$$
(6)

As a consequence of these rules, we obtained the Ito-Formula in Chapter 4. There was a differential version and an integral version. Let's summarize both versions in the following

Theorem 8.1 (Ito-Formula for Functions of a Brownian Motion): Let

$$F = F(x) : \mathbb{R} \to \mathbb{R}$$

be an arbitrary two-times differentiable function of one variable and let $\{x_t\}_{0 \le t \le T}$ be a Brownian motion. Then we have the following identities:

a) Differential Version: Let $dF(x_t) := F(x_t) - F(x_{t-dt})$. Then

$$dF(x_t) = F'(x_t) dx_t + \frac{1}{2} F''(x_t) (dx_t)^2$$

= $F'(x_t) dx_t + \frac{1}{2} F''(x_t) dt$

b) Integral Version: We have

$$F(x_T) - F(x_0) = \int_0^T F'(x_t) dx_t + \frac{1}{2} \int_0^T F''(x_t) dt$$

where the stochastic dx_t -integral above is to be defined as an Ito-integral according to

$$\int_0^T f(x_t) \, dx_t = \lim_{\Delta t \to 0} \sum_{k=1}^N f(x_{t_{k-1}}) \, \Delta x_{t_k} = \lim_{\Delta t \to 0} \sum_{k=1}^N f(x_{t_{k-1}}) \, \sqrt{\Delta t} \, \phi_k$$

and the Brownain motion $x_{t_{k-1}}$ at time $t_{k-1} = (k-1)\Delta t$ given by

$$x_{t_{k-1}} = \sqrt{\Delta t} \sum_{j=1}^{k-1} \phi_j$$
.

A slightly generalized version of this is the following

Theorem 8.2 (Ito-Formula for Functions of a Brownian Motion and Time): Let

$$F = F(x,t) : \mathbb{R}^2 \to \mathbb{R}$$

be an arbitrary two-times differentiable function of two variables and let $\{x_t\}_{0 \le t \le T}$ be a Brownian motion. Then we have the following identities:

a) Differential Version: Let $dF(x_t,t) := F(x_t,t) - F(x_{t-dt},t-dt)$. Then

$$dF = \frac{\partial F}{\partial x} dx_t + \frac{1}{2} \frac{\partial^2 F}{\partial x^2} (dx_t)^2 + \frac{\partial F}{\partial t} dt$$
$$= \frac{\partial F}{\partial x} dx_t + \left\{ \frac{1}{2} \frac{\partial^2 F}{\partial x^2} + \frac{\partial F}{\partial t} \right\} dt$$

b) Integral Version: We have

$$F(x_T, T) - F(x_0, 0) = \int_0^T \frac{\partial F}{\partial x} dx_t + \int_0^T \left\{ \frac{1}{2} \frac{\partial^2 F}{\partial x^2} + \frac{\partial F}{\partial t} \right\} dt$$

where the stochastic dx_t -integral above again is to be defined as an Ito-integral.

Since we want to calculate quantities like

$$V_t(S_t) - V_{t-\Delta t}(S_{t-\Delta t}) = V(S_t, t) - V(S_{t-\Delta t}, t - \Delta t) \stackrel{\Delta t \to 0}{\to} dV$$

where S is given by

$$S = S(x_t, t) = S_0 e^{\sigma x_t + (\mu - \sigma^2/2)t}$$

we need actually a more general version than the two theorems above. Namely, the F above in the theorems is now the V, the portfolio value. However, we want to consider the V as a function of S_t , not of x_t . That is, we plug in stochastic objects, but not directly the Brownian motion, but functions of it. To specify the class of stochastic objects we can plug into the V or some $F = F(S_t, t)$, we need the following

Definition 8.3: An Ito diffusion is a stochastic process X_t given by the SDE

$$dX_t = a(X_t, t) dt + b(X_t, t) dx_t$$

with x_t being a Brownian motion.

Example: The Black-Scholes model given by the geometric Brownain motion

$$S_t = S(x_t, t) = S_0 e^{\sigma x_t + (\mu - \sigma^2/2)t}$$

is an Ito-diffusion since with Theorem 8.2

$$dS_t = \frac{\partial S}{\partial x} dx_t + \left\{ \frac{1}{2} \frac{\partial^2 S}{\partial x^2} + \frac{\partial S}{\partial t} \right\} dt$$
$$= \sigma S_t dx_t + \left\{ \frac{\sigma^2}{2} S_t + (\mu - \sigma^2/2) S_t \right\} dt$$
$$= \sigma S_t dx_t + \mu S_t dt$$

which is of course equivalent to the SDE we derived already in Chapter 4,

$$dS_t/S_t = \mu dt + \sigma dx_t$$

Thus we have

$$a(S_t, t) = \mu S_t$$

$$b(S_t, t) = \sigma S_t$$

in Definition 8.3 and S_t is an Ito-diffusion. Now we can state a third theorem which summarizes the formulae we will actually use:

Theorem 8.4 (Ito-Formula for Functions of an Ito-Diffusion and Time): Let

$$F = F(x,t) : \mathbb{R}^2 \to \mathbb{R}$$

be an arbitrary two-times differentiable function of two variables and let $\{x_t\}_{0 \le t \le T}$ be a Brownian motion. Let X_t be an Ito-diffusion given by the SDE

$$dX_t = a(X_t, t) dt + b(X_t, t) dx_t$$

We plug X_t into the first argument of F and consider the function $F = F(X_t, t)$. Then we have the following identities:

a) Differential Version: Let $dF(X_t, t) := F(X_t, t) - F(X_{t-dt}, t - dt)$ with X_t being the Ito-diffusion from above. Then

$$dF = \frac{\partial F}{\partial x} dX_t + \frac{1}{2} \frac{\partial^2 F}{\partial x^2} (dX_t)^2 + \frac{\partial F}{\partial t} dt$$

$$= \frac{\partial F}{\partial x} (a dt + b dx_t) + \frac{1}{2} \frac{\partial^2 F}{\partial x^2} (a dt + b dx_t)^2 + \frac{\partial F}{\partial t} dt$$

$$= \frac{\partial F}{\partial x} (a dt + b dx_t) + \frac{b^2}{2} \frac{\partial^2 F}{\partial x^2} dt + \frac{\partial F}{\partial t} dt$$

$$= \left\{ a \frac{\partial F}{\partial x} + \frac{b^2}{2} \frac{\partial^2 F}{\partial x^2} + \frac{\partial F}{\partial t} \right\} dt + b \frac{\partial F}{\partial x} dx_t$$

b) Integral Version: We have

$$F(X_T,T) - F(X_0,0) = \int_0^T \left\{ a \frac{\partial F}{\partial x} + \frac{b^2}{2} \frac{\partial^2 F}{\partial x^2} + \frac{\partial F}{\partial t} \right\} dt + \int_0^T b \frac{\partial F}{\partial x} dx_t$$

where the stochastic dx_t -integral above again is to be defined as an Ito-integral.

Now we are in a position to calculate dV, the change of the value of the replicating portfolio in continuous time. With the Ito-Formula, we get

$$dV = \frac{\partial V}{\partial S} dS + \frac{1}{2} \frac{\partial^2 V}{\partial S^2} (dS)^2 + \frac{\partial V}{\partial t} dt$$
$$= \frac{\partial V}{\partial S} dS + \left\{ \frac{1}{2} \frac{\partial^2 V}{\partial S^2} \sigma^2 S^2 + \frac{\partial V}{\partial t} \right\} dt \tag{7}$$

Thus, if this change should be given by trading δ stocks of the underlying, that is, if this should be equal to δdS ,

$$dV = \frac{\partial V}{\partial S} dS + \left\{ \frac{1}{2} \frac{\partial^2 V}{\partial S^2} \sigma^2 S^2 + \frac{\partial V}{\partial t} \right\} dt \stackrel{!}{=} \delta dS$$
 (8)

we have to have the equations

$$\delta = \frac{\partial V}{\partial S} \tag{9}$$

which coincides with the definition of the previous chapter and

$$\frac{1}{2} \frac{\partial^2 V}{\partial S^2} \sigma^2 S^2 + \frac{\partial V}{\partial t} = 0 \tag{10}$$

which is the Black-Scholes equation for zero interest rates. Thus, if (9) and (10) are fulfilled, we can use the integral version of Theorem 8.4 with $X_t = S_t$ and $F(X_t, t) = V(S_t, t)$ and

$$(dS_t)^2 = S_t^2 (\mu dt + \sigma dx_t)^2 \stackrel{\text{Rechenregeln BB}}{=} S_t^2 \sigma^2 dt$$
 (11)

to obtain

$$V(S_T, T) - V(S_0, 0) = \int_0^T \frac{\partial V}{\partial S} dS_t + \int_0^T \underbrace{\left\{ \frac{1}{2} \frac{\partial^2 V}{\partial S^2} \sigma^2 S^2 + \frac{\partial V}{\partial t} \right\}}_{= 0} dt$$
$$= \int_0^T \delta(S_t, t) dS_t$$
(12)

Thus, some payoff $H = H(S_T)$ can be exactly replicated in continuous time if we impose the final condition

$$V(S_T, T) = H(S_T) (13)$$

in addition to (10).

When interest rates are present, a similar derivation can be done. Since this is an important calculation and an important result, in the continuous time Black-Scholes model exact payoff replication is still possible, we state this in a separate theorem which we will formulate and prove tomorrow.