

Chapter 8

Option Hedging in Continuous Time

Here we review some applications to mathematical finance of the tools introduced in the previous chapters. We construct a market model with jumps in which exponential normal martingales are used to model random prices. We obtain pricing and hedging formulas for contingent claims, extending the classical Black-Scholes theory to other complete markets with jumps.

8.1 Market Model

Let $(M_t)_{t \in \mathbb{R}_+}$ be a martingale having the chaos representation property of Definition 2.8.1 and angle bracket given by $d\langle M, M \rangle_t = \alpha_t^2 dt$. By a modification of Proposition 2.10.2, $(M_t)_{t \in [0, T]}$ satisfies the structure equation

$$d[M, M]_t = \alpha_t^2 dt + \phi_t dM_t.$$

When $(\phi_t)_{t \in [0, T]}$ is deterministic, $(M_t)_{t \in [0, T]}$ is alternatively a Brownian motion or a compensated Poisson martingale, depending on the vanishing of $(\phi_t)_{t \in [0, T]}$.

Let $r : \mathbb{R}_+ \rightarrow \mathbb{R}$ and $\sigma : \mathbb{R}_+ \rightarrow (0, \infty)$ be deterministic non negative bounded functions. We assume that $1 + \sigma_t \phi_t > 0$, $t \in [0, T]$. Let $(A_t)_{t \in \mathbb{R}_+}$ denote the price of the riskless asset, given by

$$\frac{dA_t}{A_t} = r_t dt, \quad A_0 = 1, \quad t \in \mathbb{R}_+, \quad (8.1.1)$$

i.e.

$$A_t = A_0 \exp \left(\int_0^t r_s ds \right), \quad t \in \mathbb{R}_+.$$

For $t > 0$, let $(S_{t,u}^x)_{u \in [t, T]}$ be the price process with risk-neutral dynamics given by

$$dS_{t,u}^x = r_t S_{t,u}^x du + \sigma_u S_{t,u}^x dM_u, \quad u \in [t, T], \quad S_{t,t}^x = x,$$

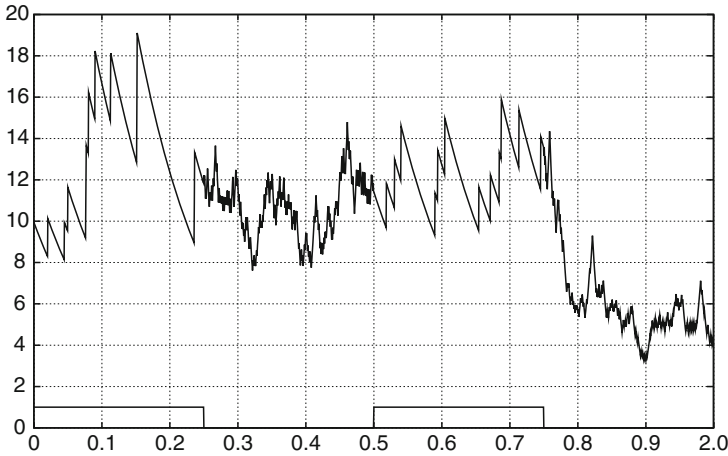


Fig. 8.1 Sample trajectory of $(S_t)_{t \in [0, T]}$

cf. Relation 2.13.5. Recall that when $(\phi_t)_{t \in \mathbb{R}_+}$ is deterministic we have

$$\begin{aligned}
 S_{t, T}^x &= x \exp \left(\int_t^T \sigma_u \alpha_u i_u dB_u + \int_t^T (r_u - \phi_u \lambda_u \sigma_u - \frac{1}{2} i_u \sigma_u^2 \alpha_u^2) du \right) \\
 &\quad \times \prod_{k=1+N_t}^{k=N_T} (1 + \sigma_{T_k} \phi_{T_k}), \tag{8.1.2}
 \end{aligned}$$

$0 \leq t \leq T$, with $S_t = S_{0,t}^1$, $t \in [0, T]$. Figure 8.1 shows a sample path of $(S_t)_{t \in [0, T]}$ when the function $(i_t)_{t \in [0, T]}$ takes values in $\{0, 1\}$, with $S_0 = 10$, $\sigma_t = 10$, and $\alpha_t = 1$, $t \in [0, T]$.

Let η_t and ζ_t be the numbers of units invested at time t , respectively in the assets $(S_t)_{t \in \mathbb{R}_+}$ and $(A_t)_{t \in \mathbb{R}_+}$. The value of the portfolio V_t at time t is given by

$$V_t = \zeta_t A_t + \eta_t S_t, \quad t \in \mathbb{R}_+. \tag{8.1.3}$$

Definition 8.1.1. *The portfolio V_t is said to be self-financing if*

$$dV_t = \zeta_t dA_t + \eta_t dS_t. \tag{8.1.4}$$

The self-financing condition can be written as

$$A_t d\zeta_t + S_t d\eta_t = 0, \quad 0 \leq t \leq T$$

under the approximation $d\langle S_t, \eta_t \rangle \simeq 0$.

Let also

$$\tilde{V}_t = V_t \exp\left(-\int_0^t r_s ds\right) \quad \text{and} \quad \tilde{S}_t = S_t \exp\left(-\int_0^t r_s ds\right)$$

denote respectively the discounted portfolio price and underlying asset price.

Lemma 8.1.2. *The following statements are equivalent:*

- i) the portfolio V_t is self-financing,
- ii) we have

$$\tilde{V}_t = \tilde{V}_0 + \int_0^t \sigma_u \eta_u \tilde{S}_u dM_u, \quad t \in \mathbb{R}_+, \quad (8.1.5)$$

- iii) we have

$$V_t = V_0 \exp\left(\int_0^t r_u du\right) + \int_0^t \sigma_u \eta_u \exp\left(\int_u^t r_u du\right) S_u dM_u, \quad (8.1.6)$$

$t \in \mathbb{R}_+$.

Proof. First, note that (8.1.5) is clearly equivalent to (8.1.6). Next, the self-financing condition (8.1.4) shows that

$$\begin{aligned} dV_t &= \zeta_t dA_t + \eta_t dS_t \\ &= \zeta_t A_t r_t dt + \eta_t r_t S_t dt + \sigma_t \eta_t S_t dM_t \\ &= r_t V_t dt + \sigma_t \eta_t S_t dM_t, \end{aligned}$$

$t \in \mathbb{R}_+$, hence

$$\begin{aligned} d\tilde{V}_t &= d\left(\exp\left(-\int_0^t r_s ds\right) V_t\right) \\ &= -r_t \exp\left(-\int_0^t r_s ds\right) V_t dt + \exp\left(-\int_0^t r_s ds\right) dV_t \\ &= \exp\left(-\int_0^t r_s ds\right) \sigma_t \eta_t S_t dM_t, \quad t \in \mathbb{R}_+, \end{aligned}$$

i.e. (8.1.5) holds. Conversely, if (8.1.5) is satisfied we have

$$\begin{aligned} dV_t &= d(A_t \tilde{V}_t) \\ &= \tilde{V}_t dA_t + A_t d\tilde{V}_t \\ &= \tilde{V}_t A_t r_t dt + \sigma_t \eta_t S_t dM_t \\ &= V_t r_t dt + \sigma_t \eta_t S_t dM_t \\ &= \zeta_t A_t r_t dt + \eta_t S_t r_t dt + \sigma_t \eta_t S_t dM_t \\ &= \zeta_t dA_t + \eta_t dS_t, \end{aligned}$$

hence the portfolio is self-financing. \square

8.2 Hedging by the Clark Formula

In the next proposition we compute a self-financing hedging strategy leading to an arbitrary square-integrable random variable F , using the Clark formula Proposition 4.2.3.

Proposition 8.2.1. *Given $F \in L^2(\Omega)$, let*

$$\eta_t = \frac{\exp\left(-\int_t^T r_s ds\right)}{\sigma_t S_t} \mathbb{E}[D_t F | \mathcal{F}_t], \quad (8.2.1)$$

$$\zeta_t = \frac{\exp\left(-\int_t^T r_u du\right) \mathbb{E}[F | \mathcal{F}_t] - \eta_t S_t}{A_t}, \quad t \in [0, T]. \quad (8.2.2)$$

Then the portfolio $(\eta_t, \zeta_t)_{t \in [0, T]}$ is self-financing and yields a hedging strategy leading to F , i.e. letting

$$V_t = \zeta_t A_t + \eta_t S_t, \quad 0 \leq t \leq T,$$

we have

$$V_t = \exp\left(-\int_t^T r_u du\right) \mathbb{E}[F | \mathcal{F}_t], \quad (8.2.3)$$

$0 \leq t \leq T$. In particular we have $V_T = F$ and

$$V_0 = \exp\left(-\int_0^T r_u du\right) \mathbb{E}[F].$$

Proof. Applying (8.2.2) at $t = 0$ we get

$$\mathbb{E}[F] \exp\left(-\int_0^T r_u du\right) = V_0,$$

hence from (8.2.2), the definition (8.2.1) of η_t and the Clark formula we obtain

$$\begin{aligned} V_t &= \zeta_t A_t + \eta_t S_t \\ &= \exp\left(-\int_t^T r_u du\right) \mathbb{E}[F | \mathcal{F}_t] \\ &= \exp\left(-\int_t^T r_u du\right) \left(\mathbb{E}[F] + \int_0^t \mathbb{E}[D_u F | \mathcal{F}_u] dM_u\right) \\ &= V_0 \exp\left(\int_0^t r_u du\right) + \exp\left(-\int_t^T r_u du\right) \int_0^t \mathbb{E}[D_u F | \mathcal{F}_u] dM_u \\ &= V_0 \exp\left(\int_0^t r_u du\right) + \int_0^t \eta_u \sigma_u S_u \exp\left(\int_u^t r_s ds\right) dM_u, \quad 0 \leq t \leq T, \end{aligned}$$

and from Lemma 8.1.2 this also implies that the portfolio $(\eta_t, \zeta_t)_{t \in [0, T]}$ is self-financing. \square

The above proposition shows that there always exists a hedging strategy starting from

$$V_0 = \mathbb{E}[F] \exp \left(- \int_0^T r_u du \right).$$

Conversely, since there exists a hedging strategy leading to

$$\tilde{V}_T = F \exp \left(- \int_0^T r_u du \right),$$

then by (8.1.5), $(\tilde{V}_t)_{t \in [0, T]}$ is necessarily a martingale with initial value

$$\tilde{V}_0 = \mathbb{E}[\tilde{V}_T] = \mathbb{E}[F] \exp \left(- \int_0^T r_u du \right).$$

We now consider the hedging of European call option with payoff $F = (S_T - K)^+$ using the Clark formula in the setting of deterministic structure equations. In this case the next proposition allows us to compute the hedging strategy appearing in (8.2.1).

Proposition 8.2.2. *Assume that $\phi_t \geq 0$, $t \in [0, T]$. Then for $0 \leq t \leq T$ we have*

$$\begin{aligned} \mathbb{E}[D_t(S_T - K)^+ | \mathcal{F}_t] &= \mathbb{E} \left[i_t \sigma_t S_{t, T}^x \mathbf{1}_{[K, \infty)}(S_{t, T}^x) \right. \\ &\quad \left. + \frac{j_t}{\phi_t} (\sigma_t \phi_t S_{t, T}^x - (K - S_{t, T}^x)^+) \mathbf{1}_{[\frac{K}{1+\sigma_t}, \infty)}(S_{t, T}^x) \right]_{x=S_t}. \end{aligned}$$

Proof. By Lemma 4.6.2, using Definition 4.6.1 and Relation (4.6.4) we have, for any $F \in \mathcal{S}$,

$$D_t F = D_t^B F + \frac{j_t}{\phi_t} (T_t^\phi F - F), \quad t \in [0, T]. \tag{8.2.4}$$

We have $T_t^\phi S_T = (1 + \sigma_t \phi_t) S_T$, $t \in [0, T]$, and the chain rule $D^B f(F) = f'(F) D^B F$, cf. Relation (5.2.1), holds for $F \in \mathcal{S}$ and $f \in \mathcal{C}_b^2(\mathbb{R})$. Since \mathcal{S} is an algebra for deterministic $(\phi_t)_{t \in [0, T]}$, we may approach $x \mapsto (x - K)^+$ by polynomials on compact intervals and proceed e.g. as in [97], p. 5-13. By dominated convergence, $F = (S_T - K)^+ \in \text{Dom}(D)$ and (8.2.4) becomes

$$D_t(S_T - K)^+ = i_t \sigma_t S_T \mathbf{1}_{[K, \infty)}(S_T) + \frac{j_t}{\phi_t} ((1 + \sigma_t \phi_t) S_T - K)^+ - (S_T - K)^+,$$

$0 \leq t \leq T$. The Markov property of $(S_t)_{t \in [0, T]}$ implies

$$\mathbb{E} [D_t^B (S_T - K)^+ | \mathcal{F}_t] = i_t \sigma_t \mathbb{E} [S_{t, T}^x \mathbf{1}_{[K, \infty)}(S_{t, T}^x)]_{x=S_t},$$

and

$$\begin{aligned}
& \frac{\dot{j}_t}{\phi_t} \mathbb{E}[(T_t^\phi S_T - K)^+ - (S_T - K)^+ | \mathcal{F}_t] \\
&= \frac{\dot{j}_t}{\phi_t} \mathbb{E} \left[((1 + \sigma_t \phi_t) S_{t,T}^x - K)^+ - (S_{t,T}^x - K)^+ \right]_{x=S_t} \\
&= \frac{\dot{j}_t}{\phi_t} \mathbb{E} \left[((1 + \sigma_t \phi_t) S_{t,T}^x - K) \mathbf{1}_{[\frac{K}{1+\sigma_t \phi_t}, \infty)}(S_{t,T}^x) \right]_{x=S_t} \\
&\quad - \frac{\dot{j}_t}{\phi_t} \mathbb{E} \left[(S_{t,T}^x - K)^+ \mathbf{1}_{[K, \infty)}(S_{t,T}^x) \right]_{x=S_t} \\
&= \frac{\dot{j}_t}{\phi_t} \mathbb{E} \left[\sigma_t \phi_t S_{t,T}^x \mathbf{1}_{[\frac{K}{1+\sigma_t \phi_t}, \infty)}(S_{t,T}^x) + (S_{t,T}^x - K) \mathbf{1}_{[\frac{K}{1+\sigma_t \phi_t}, K]}(S_{t,T}^x) \right]_{x=S_t} \\
&= \frac{\dot{j}_t}{\phi_t} \mathbb{E} \left[\sigma_t \phi_t S_{t,T}^x \mathbf{1}_{[\frac{K}{1+\sigma_t \phi_t}, \infty)}(S_{t,T}^x) - (K - S_{t,T}^x)^+ \mathbf{1}_{[\frac{K}{1+\sigma_t \phi_t}, \infty)}(S_{t,T}^x) \right]_{x=S_t} \\
&= \frac{\dot{j}_t}{\phi_t} \mathbb{E} \left[(\sigma_t \phi_t S_{t,T}^x - (K - S_{t,T}^x)^+) \mathbf{1}_{[\frac{K}{1+\sigma_t \phi_t}, \infty)}(S_{t,T}^x) \right]_{x=S_t}.
\end{aligned}$$

□

If $(\phi_t)_{t \in [0, T]}$ is not constrained to be positive then

$$\begin{aligned}
\mathbb{E}[D_t(S_T - K)^+ | \mathcal{F}_t] &= i_t \sigma_t \mathbb{E} \left[S_{t,T}^x \mathbf{1}_{[K, \infty)}(S_{t,T}^x) \right]_{x=S_t} \\
&\quad + \frac{\dot{j}_t}{\phi_t} \mathbb{E} \left[\sigma_t \phi_t S_{t,T}^x \mathbf{1}_{[\frac{K}{1+\sigma_t \phi_t}, \infty)}(S_{t,T}^x) + (S_{t,T}^x - K) \mathbf{1}_{[\frac{K}{1+\sigma_t \phi_t}, K]}(S_{t,T}^x) \right]_{x=S_t},
\end{aligned}$$

with the convention $\mathbf{1}_{[b, a]} = -\mathbf{1}_{[a, b]}$, $0 \leq a < b \leq T$. Proposition 8.2.2 can also be proved using Lemma 3.7.2 and the Itô formula (2.12.4).

In the sequel we assume that $(\phi_t)_{t \in \mathbb{R}_+}$ is deterministic and

$$dM_t = i_t dB_t + \phi_t (dN_t - \lambda_t dt), \quad t \in \mathbb{R}_+, \quad M_0 = 0,$$

as in Relation (2.10.4).

Next we compute

$$\exp \left(- \int_0^T r_s ds \right) \mathbb{E} [(S_T - K)^+]$$

in terms of the Black-Scholes function

$$\text{BS}(x, T; r, \sigma^2; K) = e^{-rT} \mathbb{E}[(x e^{rT - \sigma^2 T/2 + \sigma W_T} - K)^+],$$

where W_T is a centered Gaussian random variable with variance T .

Proposition 8.2.3. *The expectation*

$$\exp\left(-\int_0^T r_s ds\right) \mathbb{E}[(S_T - K)^+]$$

can be computed as

$$\begin{aligned} & \exp\left(-\int_0^T r_s ds\right) \mathbb{E}[(S_T - K)^+] \\ &= \exp(-\Gamma_0(T)) \sum_{k=0}^{\infty} \frac{1}{k!} \int_0^T \cdots \int_0^T \\ & \quad BS\left(S_0 \exp\left(-\int_0^T \phi_s \gamma_s \sigma_s ds\right) \prod_{i=1}^{i=k} (1 + \sigma_{t_i} \phi_{t_i}), T; R_T, \frac{\Gamma_0(T)}{T}; K\right) \\ & \quad \gamma_{t_1} \cdots \gamma_{t_k} dt_1 \cdots dt_k. \end{aligned}$$

Proof. Similarly to Proposition 3.7.3 we have

$$\mathbb{E}[e^{-TR_T}(S_T - K)^+] = \sum_{k=0}^{\infty} \mathbb{E}[e^{-TR_T}(S_T - K)^+ | N_T = k] \mathbb{P}(N_T = k),$$

with

$$\mathbb{P}(N_T = k) = \exp(-\Gamma_0(T)) \frac{(\Gamma_0(T))^k}{k!}, \quad k \in \mathbb{N}.$$

Conditionally to $\{N_T = k\}$, the jump times (T_1, \dots, T_k) have the law

$$\frac{k!}{(\Gamma_0(T))^k} \mathbf{1}_{\{0 < t_1 < \cdots < t_k < T\}} \gamma_{t_1} \cdots \gamma_{t_k} dt_1 \cdots dt_k,$$

since the process $(N_{\Gamma_0^{-1}(T)_t})_{t \in \mathbb{R}_+}$ is a standard Poisson process. Hence, conditionally to

$$\{N(\Gamma_0^{-1}(\Gamma_0(T))) = k\} = \{N_T = k\},$$

its jump times $(\Gamma_0(T_1), \dots, \Gamma_0(T_k))$ have a uniform law on $[0, \Gamma_0(T)]^k$. We then use the fact that $(\tilde{B}_t)_{t \in \mathbb{R}_+}$ and $(N_t)_{t \in \mathbb{R}_+}$ are also independent under \mathbb{P} since $(r_t)_{t \in \mathbb{R}_+}$ is deterministic, and the identity in law

$$S_T \stackrel{\text{law}}{=} S_0 X_T \exp\left(-\int_0^T \phi_s \lambda_s \sigma_s ds\right) \prod_{k=1}^{k=N_T} (1 + \sigma_{T_k} \phi_{T_k}),$$

where

$$X_T = \exp\left(TRR_T - I_0(T)/2 + \left(\frac{I_0(T)}{T}\right)^{1/2} W_T\right),$$

and W_T is independent of N . □

8.3 Black-Scholes PDE

As in the standard Black-Scholes model, it is possible to determine the hedging strategy in terms of the Delta of the price in the case $(r_t)_{t \in \mathbb{R}_+}$ is deterministic.

Let the function $C(t, x)$ be defined by

$$\begin{aligned} C(t, S_t) &= V_t \\ &= \exp\left(-\int_t^T r_u du\right) \mathbb{E}[(S_T - K)^+ | \mathcal{F}_t] \\ &= \exp\left(-\int_t^T r_u du\right) \mathbb{E}[(S_T - K)^+ | S_t], \quad t \in \mathbb{R}_+. \end{aligned}$$

cf. (8.2.3). An application of the Itô formula leads to

$$\begin{aligned} dC(t, S_t) &= \left(\frac{\partial C}{\partial t} + r_t S_t \frac{\partial C}{\partial x} + \frac{1}{2} \frac{\partial^2 C}{\partial x^2} i_t \alpha_t^2 S_t^2 \sigma_t^2 + \lambda_t \Theta C\right)(t, S_t) dt \\ &\quad + S_t \sigma_t \frac{\partial C}{\partial x}(t, S_t) dM_t + (C(t, S_t(1 + \sigma_t \phi_t)) - C(t, S_t))(dN_t - \lambda_t dt) \end{aligned} \tag{8.3.1}$$

where

$$\Theta C(t, S_t) = C(t, S_t(1 + \sigma_t \phi_t)) - C(t, S_t) - \frac{\partial C}{\partial x}(t, S_t) S_t \sigma_t \phi_t.$$

The process

$$\begin{aligned} \tilde{C}_t &:= C(t, S_t) \exp\left(-\int_0^t r_s ds\right) \\ &= \exp\left(-\int_0^T r_u du\right) \mathbb{E}[(S_T - K)^+ | \mathcal{F}_t] \\ &= \tilde{V}_t \end{aligned}$$

is a martingale from Propositions 2.5.8 and 8.2.1-(ii), with

$$\begin{aligned} d\tilde{C}_t &= \eta_t d\tilde{S}_t \\ &= \sigma_t \eta_t S_t dM_t \\ &= i_t \sigma_t \eta_t S_t dB_t + \sigma_t \phi_t \eta_t S_t (dN_t - \lambda_t dt), \end{aligned} \tag{8.3.2}$$

from Lemma 8.1.2. Therefore, by identification of (8.3.1) and (8.3.2),

$$\begin{cases} r_t C(t, S_t) = \left(\frac{\partial C}{\partial t} + r_t S_t \frac{\partial C}{\partial x} + \frac{1}{2} i_t \alpha_t^2 S_t^2 \sigma_t^2 \frac{\partial^2 C}{\partial x^2} + \lambda_t \Theta C \right) (t, S_t), \\ \eta_t \sigma_t S_t dM_t = S_t \sigma_t \frac{\partial C}{\partial x} (t, S_t) dM_t + (C(t, S_t(1 + \sigma_t \phi_t)) - C(t, S_t)) (dN_t - \lambda_t dt). \end{cases}$$

Therefore, by identification of the Brownian and Poisson parts,

$$\begin{cases} i_t \eta_t S_t \sigma_t = i_t S_t \sigma_t \frac{\partial C}{\partial x} (t, S_t) \\ j_t \eta_t S_t \sigma_t \phi_t = C(t, S_t(1 + \sigma_t \phi_t)) - C(t, S_t). \end{cases} \tag{8.3.3}$$

The term $\Theta C(t, S_t)$ vanishes on the set

$$\{t \in \mathbb{R}_+ : \phi_t = 0\} = \{t : i(t) = 1\}.$$

Therefore, (8.3.3) reduces to

$$\eta_t = \frac{\partial C}{\partial x} (t, S_t),$$

i.e. the process $(\eta_t)_{t \in \mathbb{R}_+}$ is equal to the usual Delta (8.3) on $\{t \in \mathbb{R}_+ : i_t = 1\}$, and to

$$\eta_t = \frac{C(t, S_t(1 + \phi_t \sigma_t)) - C(t, S_t)}{S_t \phi_t \sigma_t}$$

on the set $\{t \in \mathbb{R}_+ : i_t = 0\}$.

Proposition 8.3.1. *The Black-Scholes PDE for the price of a European call option is written as*

$$\frac{\partial C}{\partial t} (t, x) + r_t x \frac{\partial C}{\partial x} (t, x) + \frac{1}{2} \alpha_t^2 x^2 \sigma_t^2 \frac{\partial^2 C}{\partial x^2} (t, x) = r_t C(t, x),$$

on $\{t : \phi_t = 0\}$, and as

$$\frac{\partial C}{\partial t} (t, x) + r_t x \frac{\partial C}{\partial x} (t, x) + \lambda_t \Theta C(t, x) = r_t C(t, x),$$

on the set $\{t \in \mathbb{R}_+ : \phi_t \neq 0\}$, under the terminal condition $C(T, x) = (x - K)^+$.

8.4 Asian Options and Deterministic Structure

The price at time t of an Asian option is defined as

$$\mathbb{E} \left[e^{-\int_t^T r_s ds} \left(\frac{1}{T} \int_0^T S_u du - K \right)^+ \middle| \mathcal{F}_t \right].$$

The next proposition provides a replicating hedging strategy for Asian options in the case of a deterministic structure equation. Following [74], page 91, and [13], we define the auxiliary process

$$Y_t = \frac{1}{S_t} \left(\frac{1}{T} \int_0^t S_u du - K \right), \quad t \in [0, T]. \quad (8.4.1)$$

Proposition 8.4.1. *There exists a measurable function \tilde{C} on $\mathbb{R}_+ \times \mathbb{R}$ such that $\tilde{C}(t, \cdot)$ is \mathcal{C}^1 for all $t \in \mathbb{R}_+$, and*

$$S_t \tilde{C}(t, Y_t) = \mathbb{E} \left[\left(\frac{1}{T} \int_0^T S_u du - K \right)^+ \middle| \mathcal{F}_t \right].$$

Moreover, the replicating portfolio for an Asian option with payoff

$$\left(\frac{1}{T} \int_0^T S_u du - K \right)^+$$

is given by (8.1.3) and

$$\begin{aligned} \eta_t &= \frac{1}{\sigma_t} e^{-\int_t^T r_s ds} \left(\tilde{C}(t, Y_t) \sigma_t \right. \\ &\quad \left. + (1 + \sigma_t \phi_t) \left(\frac{\dot{Y}_t}{\phi_t} \left(\tilde{C} \left(t, \frac{Y_t}{1 + \sigma_t \phi_t} \right) - \tilde{C}(t, Y_t) \right) - i_t \sigma_t Y_t \partial_2 \tilde{C}(t, Y_t) \right) \right). \end{aligned} \quad (8.4.2)$$

Proof. With the above notation, the price at time t of the Asian option becomes

$$\mathbb{E} \left[e^{-\int_t^T r_s ds} S_T (Y_T)^+ \middle| \mathcal{F}_t \right].$$

For $0 \leq s \leq t \leq T$, we have

$$d(S_t Y_t) = \frac{1}{T} d \left(\int_0^t S_u du - K \right) = \frac{S_t}{T} dt,$$

hence

$$\frac{S_t Y_t}{S_s} = Y_s + \frac{1}{T} \int_s^t \frac{S_u}{S_s} du.$$

Since S_u/S_t is independent of S_t by (8.1.2), we have, for any sufficiently integrable payoff function H ,

$$\begin{aligned} \mathbb{E} \left[H(S_T Y_T) \mid \mathcal{F}_t \right] &= \mathbb{E} \left[H \left(S_t Y_t + \frac{1}{T} \int_t^T S_u du \right) \mid \mathcal{F}_t \right] \\ &= \mathbb{E} \left[H \left(xy + \frac{x}{T} \int_t^T \frac{S_u}{S_t} du \right) \right]_{y=Y_t, x=S_t}. \end{aligned}$$

Let $C \in \mathcal{C}_b^2(\mathbb{R}_+ \times \mathbb{R}^2)$ be defined as

$$C(t, x, y) = \mathbb{E} \left[H \left(xy + \frac{x}{T} \int_t^T \frac{S_u}{S_t} du \right) \right],$$

i.e.

$$C(t, S_t, Y_t) = \mathbb{E} \left[H(S_T Y_T) \mid \mathcal{F}_t \right].$$

When $H(x) = \max(x, 0)$, since for any $t \in [0, T]$, S_t is positive and \mathcal{F}_t -measurable, and S_u/S_t is independent of \mathcal{F}_t , $u \geq t$, we have:

$$\begin{aligned} \mathbb{E} \left[H(S_T Y_T) \mid \mathcal{F}_t \right] &= \mathbb{E} \left[S_T (Y_T)^+ \mid \mathcal{F}_t \right] \\ &= S_t \mathbb{E} \left[\left(Y_T \frac{S_T}{S_t} \right)^+ \mid \mathcal{F}_t \right] \\ &= S_t \mathbb{E} \left[\left(Y_t + \frac{1}{T} \int_t^T \frac{S_u}{S_t} du \right)^+ \mid \mathcal{F}_t \right] \\ &= S_t \mathbb{E} \left[\left(y + \frac{1}{T} \int_t^T \frac{S_u}{S_t} du \right)^+ \right]_{y=Y_t} \\ &= S_t \tilde{C}(t, Y_t), \end{aligned}$$

with

$$\tilde{C}(t, y) = \mathbb{E} \left[\left(y + \frac{1}{T} \int_t^T \frac{S_u}{S_t} du \right)^+ \right].$$

We now proceed as in [13], which deals with the sum of a Brownian motion and a Poisson process. From the expression of $1/S_t$ given by (8.1.2) we have

$$d\left(\frac{1}{S_t}\right) = \frac{1}{S_{t-}} \left(\left(-r_t + \frac{\alpha_t^2 \sigma_t^2}{1 + \sigma_t \phi_t} \right) dt - \frac{\sigma_t}{1 + \sigma_t \phi_t} dM_t \right),$$

hence by (2.12.4), Itô's formula and the definition (8.4.1) of Y_t , we have

$$dY_t = Y_t \left(-r_t + \frac{\alpha_t^2 \sigma_t^2}{1 + \sigma_t \phi_t} \right) dt + \frac{1}{T} dt - \frac{Y_{t-} \sigma_t}{1 + \sigma_t \phi_t} dM_t.$$

Assuming that $H \in \mathcal{C}_b^2(\mathbb{R})$ and applying Lemma 3.7.2 we get

$$\begin{aligned} \mathbb{E} \left[D_t H(S_T Y_T) \middle| \mathcal{F}_t \right] &= L_t C(t, S_t, Y_t) \\ &= i_t \left(\sigma_t S_{t-} \partial_2 C(t, S_t, Y_t) - \frac{Y_t \sigma_t}{1 + \sigma_t \phi_t} \partial_3 C(t, S_t, Y_t) \right) \\ &\quad + \frac{j_t}{\phi_t} \left(C \left(t, S_{t-} + \sigma_t S_{t-}, Y_{t-} - \frac{Y_t \sigma_t}{1 + \sigma_t \phi_t} \right) - C(t, S_{t-}, Y_{t-}) \right), \end{aligned} \quad (8.4.3)$$

where L_t is given by (2.12.5). Next, given a family $(H_n)_{n \in \mathbb{N}}$ of \mathcal{C}_b^2 functions, such that $|H_n(x)| \leq x^+$ and $|H'_n(x)| \leq 2$, $x \in \mathbb{R}$, $n \in \mathbb{N}$, and converging pointwise to $x \rightarrow x^+$, by dominated convergence (8.4.3) holds for $C(t, x, y) = x \tilde{C}(t, y)$ and we obtain:

$$\begin{aligned} \mathbb{E} \left[D_t \left(\frac{1}{T} \int_0^T S_u du - K \right)^+ \middle| \mathcal{F}_t \right] &= i_t \tilde{C}(t, Y_t) \sigma_t S_t \\ &\quad + S_t \left(\frac{j_t}{\phi_t} \left(\tilde{C} \left(t, \frac{Y_t}{1 + \sigma_t \phi_t} \right) - \tilde{C}(t, Y_t) \right) - i_t \sigma_t Y_t \partial_2 \tilde{C}(t, Y_t) \right) \\ &\quad + S_t \sigma_t \phi_t \left(\frac{j_t}{\phi_t} \left(\tilde{C} \left(t, \frac{Y_t}{1 + \sigma_t \phi_t} \right) - \tilde{C}(t, Y_t) \right) - i_t \sigma_t Y_t \partial_2 \tilde{C}(t, Y_t) \right). \end{aligned}$$

□

As a particular case we consider the Brownian motion model, i.e. $\phi_t = 0$, for all $t \in [0, T]$, so $i_t = 1$, $j_t = 0$ for all $t \in [0, T]$, and we are in the Brownian motion model. In this case we have

$$\begin{aligned} \eta_t &= e^{-\int_t^T r_s ds} \left(-Y_t \partial_2 \tilde{C}(t, Y_t) + \tilde{C}(t, Y_t) \right) \\ &= e^{-\int_t^T r_s ds} \left(S_t \frac{\partial}{\partial x} \tilde{C} \left(t, \frac{1}{x} \left(\frac{1}{T} \int_0^t S_u du - K \right) \right) \Big|_{x=S_t} + \tilde{C}(t, Y_t) \right) \\ &= \frac{\partial}{\partial x} \left(x e^{-\int_t^T r_s ds} \tilde{C} \left(t, \frac{1}{x} \left(\frac{1}{T} \int_0^t S_u du - K \right) \right) \right) \Big|_{x=S_t}, \quad t \in [0, T], \end{aligned}$$

which can be denoted informally as a partial derivative with respect to S_t .

8.5 Notes and References

See e.g. [74] and [135] for standard references on stochastic finance, and [97] for a presentation of the Malliavin calculus applied to continuous markets. The use of normal martingales in financial modelling has been first considered in [35]. The material on Asian options is based on [70] and [12]. Hedging strategies for Lookback options have been computed in [15] using the Clark-Ocone formula.