

Master 2 of Quantitative Finance, Numerical Finance

October 20, 2020

Dynamic Hedging Strategy in the Black and Scholes framework

Due for November 12th 2020.

Let us consider a (complete) financial market with two assets: a risky one $(S_t)_{t \geq 0}$ and a non risky one $(S_t^0)_{t \geq 0}$. It is assumed that the interest rate is constant and denoted by r . Thus, the dynamics of the non risky asset (or bond) writes:

$$dS_t^0 = rS_t^0 dt. \quad (\text{B})$$

It is also assumed that the risky asset follows a Black and Scholes dynamics, i.e.

$$dS_t = S_t(\mu dt + \sigma dW_t), \quad (\text{B\&S})$$

where μ is the *market trend* of the asset, σ the *volatility* and W stands for a standard Brownian motion, associated with a filtration $(\mathcal{F}_t)_{t \geq 0}$, under the *market probability* \mathbb{P} .

The goal of this exercise is twofold. First, to give the fair price of an option on S . Second, from the fair price, to implement an investment strategy that will guarantee the option is indeed replicated at maturity.

Let $T > 0$ be a fixed final time horizon and let Φ be a non-negative pay-off function which has polynomial growth, i.e. there exists $p \geq 1$ and $C \geq 1$ such that for all $x \in \mathbb{R}_+$, $0 \leq \Phi(x) \leq C(1 + |x|^p)$.

1. Justify that for all given $T > 0$, the quantity $\Phi(S_T) \in L^1(\mathbb{P})$, i.e. that $\mathbb{E}^\mathbb{P}[\Phi(S_T)] < +\infty$, where $\mathbb{E}^\mathbb{P}$ denotes the expectation under \mathbb{P} .
2. Establish that the fair price at time $t \in [0, T]$ of the option on S with pay-off Φ and maturity T writes:

$$V_t := \exp(-r(T-t))\mathbb{E}^\mathbb{Q}[\Phi(S_T)|\mathcal{F}_t] = \exp(-r(T-t))\mathbb{E}^\mathbb{Q}[\Phi(S_T)|S_t], \quad (\text{P})$$

where \mathbb{Q} stands for the *risk-neutral* probability, under which the dynamics of S rewrites for $s \geq t$ as

$$dS_s = S_s(r ds + \sigma dW_s^\mathbb{Q}), \quad (\text{B\&S}_\mathbb{Q})$$

where $W^\mathbb{Q}$ is a \mathbb{Q} Brownian motion.

3. Justify from (P) that $V_t = v(t, S_t)$ for some function v to be specified. Assuming now that v is *smooth*, namely that $v \in C^{1,2}([0, T] \times \mathbb{R}_+^*, \mathbb{R}) \cap C^0([0, T] \times \mathbb{R}_+^*, \mathbb{R})$, with $\mathbb{R}_+^* := (0, +\infty)$, prove that

$$\begin{cases} \partial_t v(t, x) + rx\partial_x v(t, x) + \frac{1}{2}\sigma^2 x^2 \partial_x^2 v(t, x) - rv(t, x) = 0, (t, x) \in [0, T) \times \mathbb{R}_+^*, \\ v(T, x) = \Phi(x), x \in \mathbb{R}_+^*. \end{cases} \quad (\text{PDE})$$

The Partial Differential Equation (PDE) is called the *valuation PDE* and provides an alternative to the estimation of the expectation in (P) for the numerical approximation of the fair price.

Bonus: prove that if Φ is also continuous, then v indeed satisfies the indicated smoothness conditions.

4. The point is now to implement a dynamic strategy to replicate from the initial wealth V_0 the final pay-off $\Phi(S_T)$ at time T . The idea is to invest in continuous time partly in the risky asset, partly in the non risky one and to construct a *self financing* portfolio $(P_t)_{t \in [0, T]}$ achieving this goal. Thus from $P_0 = V_0$, initial amount of money received by the manager corresponding to the *fair price* of the option, the point is to find $(\delta_t)_{t \geq 0}, (\delta_t^0)_{t \geq 0}$, corresponding respectively to the quantities of risky and non-risky asset at time $t \in [0, T]$, such that $P_T = V_T = \Phi(S_T)$. The value at time t of the portfolio is

$$P_t = \delta_t S_t + \delta_t^0 S_t^0, \quad (\text{P})$$

and the *self financing* constraint writes:

$$dP_t = \delta_t dS_t + \delta_t^0 dS_t^0. \quad (\text{SF})$$

We carefully specify that the dynamics in (SF) does not come from an Itô type expansion but simply expresses that, once the strategy is chosen at some time t , then the portfolio feels the infinitesimal variations of the asset S and the bond S^0 .

A natural idea to replicate the option is to construct a portfolio P such that its value at time t corresponds to the option price at the same time. In other words, it is tempting to have $P_t = V_t$. The remaining point consists in exhibiting an investment strategy such that (SF) is satisfied and $P_t = V_t$. To this end, we will use the valuation PDE (PDE). Fix $0 \leq u < t < T$. Expand then v with Itô's formula, recall that $v \in C^{1,2}([u, t] \times \mathbb{R}_+^*, \mathbb{R})$:

$$\begin{aligned} v(t, S_t) &= v(u, S_u) + \int_u^t \partial_s v(s, S_s) ds + \int_u^t \partial_x v(s, S_s) dS_s + \frac{1}{2} \int_u^t \partial_x^2 v(s, S_s) d\langle S \rangle_s \\ &= v(u, S_u) + \int_u^t \partial_s v(s, S_s) ds + \int_u^t \partial_x v(s, S_s) dS_s + \frac{1}{2} \int_u^t \sigma^2 S_s^2 \partial_x^2 v(s, S_s) ds \\ &= v(u, S_u) + \int_u^t \partial_x v(s, S_s) dS_s + \int_u^t \left(\partial_s v(s, S_s) + r S_s \partial_x v(s, S_s) + \sigma^2 S_s^2 \partial_x^2 v(s, S_s) \right) ds \\ &\quad - \int_u^t r S_s \partial_x v(s, S_s) ds \\ &= v(u, S_u) + \int_u^t \partial_x v(s, S_s) dS_s + \int_u^t \left(r v(s, S_s) - r S_s \partial_x v(s, S_s) \right) ds \\ &= v(u, S_u) + \int_u^t \partial_x v(s, S_s) dS_s + \int_u^t \left(v(s, S_s) - S_s \partial_x v(s, S_s) \right) r ds, \end{aligned}$$

where we have used (PDE) for the last but one equality. This equivalently reads in differential form:

$$dv(t, S_t) = \partial_x v(t, S_t) dS_t + \left(v(t, S_t) - S_t \partial_x v(t, S_t) \right) r dt. \quad (1)$$

This equality precisely allows to identify P_t and V_t taking $\delta_t = \partial_x v(s, S_s)$ observing as well that for this choice, from (P), $\delta_t^0 S_t^0 = V_t - \delta_t S_t$ so that from (1) we indeed get:

$$dV_t = dv(t, S_t) = \partial_x v(t, S_t) dS_t + \delta_t^0 S_t^0 r dt = \delta_t dS_t + \delta_t^0 dS_t^0 = dP_t. \quad (\text{I})$$

This thus means that, if one could update in practice the investment strategy in continuous time, then it would be precisely possible to replicate the pay-off following the dynamics (I).

In practice, we are going to discretize the dynamics on a time grid.

5. Let $N \in \mathbb{N}$ be fixed and consider the grid $\Lambda_N := \{(t_i)_{i \in \llbracket 0, N \rrbracket}, t_i := ih, h := \frac{T}{N}\}$. Starting from $V_0^h := v(0, S_0) = \exp(-rT) \mathbb{E}^\mathbb{Q}[\Phi(S_T) | S_0]$ the discretization of (I) reads for all $i \in \llbracket 0, N-1 \rrbracket$:

$$\begin{aligned} V_{t_{i+1}}^h - V_{t_i}^h &= \delta_{t_i}(S_{t_{i+1}} - S_{t_i}) + \delta_{t_i}^0(S_{t_{i+1}}^0 - S_{t_i}^0) = \delta_{t_i}(S_{t_{i+1}} - S_{t_i}) + \delta_{t_i}^0 S_{t_i}^0 (\exp(rh) - 1) \\ &= \delta_{t_i}(S_{t_{i+1}} - S_{t_i}) + (V_{t_i}^h - \delta_{t_i} S_{t_i})(\exp(rh) - 1). \end{aligned} \quad (\text{I}_h)$$

Implement now for $\Phi(x) = (x - K)^+$, $S_0 = 100$, $\sigma = 0.2$, $\mu = 0.03$, $r = 0.015$, $K = 97$, $T = 1$, $N \in \{10, 100, 500, 1000, 5000, 10000\}$ the strategy (I_h) . Pay attention that this requires:

- i. To simulate $(S_{t_i})_{i \in \llbracket 1, N \rrbracket}$ under the market probability \mathbb{P} . Namely, given a collection of standard i.i.d. Gaussian variables $(N_i)_{i \in \mathbb{N}}$ s.t. $N_1 \stackrel{(\text{law})}{=} \mathcal{N}(0, 1)$, this can be done for all $i \in \llbracket 0, N - 1 \rrbracket$ from the recursion:

$$S_{t_{i+1}} = S_{t_i} \exp(\sigma \sqrt{h} N_{i+1} + (\mu - \frac{\sigma^2}{2})h).$$

This represents the pathwise evolution of the asset under \mathbb{P} .

- ii. To update at any discretization time $\delta_{t_i} := \partial_x v(t, S_{t_i})$ which in turn implies to differentiate for the current example the Black and Scholes formula (price under the risk neutral probability measure \mathbb{Q}).

You should eventually observe

$$V_T^h - V_T = V_T^h - \Phi(S_T) \xrightarrow[h \rightarrow 0]{a.s.} 0. \quad (\text{C}_H)$$

Namely, the discretized strategy converges *almost surely* to the final pay-off at maturity when the rebalancing step tends to 0. To illustrate (C_H) you could for instance run, for fixed h , a number M of independent simulations $((V_T^h - V_T)^i)_{i \in \llbracket 1, M \rrbracket}$ and observe that the associated standard deviation goes to zero with M .

6. Bonus: Prove that (C_H) holds. You might use a regularization of the pay-off or an estimate of the probability that S_T belongs to a neighborhood where the pay-off is not *smooth*.