Abstract. This paper investigates a problem arising in asset-liability management in life insurance. As shown by other authors, an insurance company can guarantee its solvency by purchasing a Margrabe option enabling it to exchange its assets for a certain portfolio replicating its insurance liabilities. The objective of the paper is to investigate numerically a valuation technique for such an option in a situation when the insurance company is a "large" investor, implying that its trading decisions can affect asset prices. This setting contradicts the assumptions underlying traditional financial models and requires alternative pricing techniques. One existing approach to dealing with such problems relies on the use of forward-backward stochastic differential equations (FBSDEs). We use this framework to formulate a pricing equation and solve the latter numerically to obtain the price of the option.

1 Introduction

Usual continuous-time financial models assume that individual investor’s behaviour does not affect stock prices. This idea is known under the name "small investor hypothesis", implying that each investor is assumed to be unable to move market prices by his trading decisions and acts as a price...
taker. In contrast to that hypothesis, many authors investigated a setting in which market prices depended on the size of the transaction.

This research direction can be classified into two broad categories. The first one deals with liquidity risk and investigates its impact on asset prices and hedging strategies. In particular, the authors in [2] and [4] assume that market prices result from interaction of small price taking investors with an exogenously given asset supply curve. The latter determines the price for a given transaction size. This research category also includes [3], [14], as well as [15], among others.

The second type of literature in this research direction focuses on feedback effects of hedging strategies on underlying asset prices. It is assumed that a large investor can - directly or indirectly - affect market prices. In particular, it is shown in [13] that large investor’s hedging decisions can lead to increased volatility of the underlying asset. The authors in [10] also present a model in which the existence of a large trader results in higher market volatility. These authors, as well as [11], [9], and [12] model stock prices as being directly dependent on large investor’s stock holdings through a certain reaction function. The authors in [1] use a similar reaction function setting to the one employed in [10] to specify the dynamics of the asset price as a semi-martingale parametrized by the large investor’s position in the stock. In contrast to their approach, [6], as well as [5], assume that parameters in the diffusion driving the evolution of the stock price, rather than the price itself, depend on large investor’s trading decisions. The authors in [7] develop a general equilibrium model that justifies the models used by Cvitanic and Ma, and Cuoco and Cvitanic.

The present paper belongs to the second group. We adopt the framework used in [6] to study the problem faced by an insurance company that wants to achieve solvency. This can be done by means of a Margrabe option that gives the company the right to exchange its asset portfolio for a valuation portfolio. The latter is a replicating portfolio for the insurance company’s liabilities. One can obtain the price of a Margrabe option within the Black-Scholes setting by representing it as a standard European Put option written on a certain artificial asset. However, if the company is a large investor whose hedging strategies can have a feedback effect on the price of the underlying asset, one needs to use a different pricing methodology. In this paper it is assumed that market volatility depends on the value of the replicating portfolio held by the large investor and the amount of money he invests in the asset. Our objective is to study the hedging problem faced by the large investor. Namely, we intend to find the price of and the hedging strategy for the Margrabe option in a situation where the insurance company is no longer
a price taker. In mathematical terms, the problem translates into a forward-backward stochastic differential equation. The solution of its backward part gives us the quantities we are looking for.

The paper is organised as follows. Section two presents the model, explains the derivation of the associated FBSDE and PDE and ends with an explanation of a numerical scheme that will be used to obtain the solution. Section three discusses the application at hand. It gives a brief overview of the relevant material in [16] and puts it in the context of the large investor problem. Section four gives details of numerical simulations and presents the findings.

2 Model

We study the hedging problem for a large investor on a finite time horizon \([0, T]\), given the initial stock price \(S(0)\) and the terminal wealth \(g(S(T))\) referring to the pay-off of the option to be hedged. The objective of the hedger is to find a portfolio process and the minimal initial wealth \(x = X(0)\), such that at the option’s expiration date \(X(T) = g(S(T))\).

We assume that there are two investment possibilities on the market:

- Bond with the following dynamics:

\[
    dB(t) = B(t) \cdot r \cdot dt \quad (1)
\]

\[B(0) = 1,\]

where \(r\) is a riskless interest rate.

- Stock that follows the dynamics:

\[
    dS(t) = S(t) \cdot b \cdot dt + S(t) \cdot \gamma(t, S(t), X(t), \pi(t)) dW \quad (2)
\]

\[= \mu \cdot dt + \sigma dW\]

\[S(0) = s,\]

where \(b\) is a constant drift, \(\gamma\) is an appropriate volatility function, \(X(t)\) is the the value of the replicating portfolio held by large investor and \(\pi(t)\) is the amount of money he invests in the stock. The main thing to be noticed here is that stock’s volatility is assumed to depend on large investor’s strategies. \(W(t)\) is a standard Brownian motion defined on a complete probability space \((\Omega, \mathcal{F}, \{\mathcal{F}_t\}_{0 \leq t \leq T}, \mathbb{P})\) with filtration generated by \(W(t)\).
The investor begins with a certain endowment $x > 0$ and allocates his wealth in the stock and the bond according to a certain strategy. The portfolio process $\pi = \pi(t), t \in [0, T]$ is assumed to be a real-valued, progressively measurable, square-integrable stochastic process.

At each point of time the value of the replicating portfolio equals:

$$X(t) = \frac{\pi(t)}{S(t)} \cdot S(t) + \frac{(X(t) - \pi(t))}{B(t)} \cdot B(t)$$  \hspace{1cm} (3)

and has the following dynamics:

$$dX(t) = \frac{\pi(t)}{S(t)} \cdot dS(t) + \frac{(X(t) - \pi(t))}{B(t)} \cdot dB(t) = \hspace{1cm} (4)$$

$$= (\pi(t)b + (X(t) - \pi(t)) \cdot r)dt + \pi(t) \cdot \gamma(t, S(t), X(t), \pi(t))dW(t)$$

Following [6], we introduce the following assumptions.

• The drift and volatility functions in the stock price dynamics, $\mu$, and $\sigma$, are twice continuously differentiable. The functions $b$ and $\gamma$ together with their first order partial derivatives are bounded uniformly in $(t,s,x,\pi)$. Further, it is assumed that partial derivatives of $b$ and $\gamma$ in $s$ satisfy

$$sup_{(t,s,x,\pi)} \left( |s \cdot \frac{\partial b}{\partial s}|, |s \cdot \frac{\partial \gamma}{\partial s}| \right) < \infty$$  \hspace{1cm} (5)

• There exists a positive constant $k$, such that $\gamma^2 > k$ for all $(t,s,x,\pi)$.

According to Lemma 2.3 in [6] these assumptions guarantee that the stock price remains almost surely positive.

• Also, we assume that option’s pay-off function $g$ is non-negative and $\lim_{s \to \infty} g(s) = \infty$. Moreover, $g$ has bounded, continuous partial derivatives up to third order and there exist constants $K > 0$ and $M > 0$, such that

$$|s \cdot \frac{dg}{ds}| \leq K \cdot (1 + g(s))$$  \hspace{1cm} (6)

and

$$sup_{s>0} |s^2 \cdot \frac{d^2 g}{ds^2}| = M < \infty$$  \hspace{1cm} (7)
In order to satisfy these assumptions we use the following smoothed pay-off function (suggested in [9]):

\[ g(x) = \frac{1}{2} \left( K - x + \sqrt{(K - x)^2 + \alpha} \right), \] (8)

where \( \alpha \) is a smoothing parameter.

- Finally, we assume that partial derivatives of \( \gamma \) in \( x \) and \( \pi \) satisfy the following condition.

\[ \sup_{t,s,x,\pi} \left( \left| x \cdot \frac{d\gamma}{dx} \right| + \left| \pi \cdot \frac{d\gamma}{d\pi} \right| \right) < \infty \] (9)

These assumptions are satisfied by the constant drift coefficient \( b \) in SDE (2) and the following volatility function (suggested in [6])

\[ \gamma(t, S(t), X(t), \pi(t)) = \sigma + \frac{1}{4} \cdot \arctan(X(t)^2 + \pi(t)^2) \] (10)

**Definition 2.1.** ([6]) For any given initial wealth \( x > 0 \) the portfolio process \( \pi \) is called admissible (with respect to \( x \)) if for any \( s > 0 \) the corresponding price process \( S(\cdot) \) and wealth process \( X(\cdot) \) satisfy \( S(t) > 0 \) and \( X(t) \geq 0, \forall t \in [0,T] \). For each initial wealth \( x \), we denote the set of admissible portfolio strategies by \( \mathcal{A} \).

We are now going to formulate the hedging problem in terms of a forward-backward stochastic differential equation. Consider a FBSDE given by the stock price SDE - the forward part - and the BSDE describing the evolution of the portfolio wealth - the backward part.

\[
\begin{align*}
\frac{dS(t)}{S(t)} &= S(t) \cdot b \cdot dt + S(t) \cdot \gamma(t, S(t), X(t), \pi(t)) \, dW \\
\frac{dX(t)}{X(t)} &= \left( \pi(t) b + (X(t) - \pi(t)) \cdot r \right) dt + \pi(t) \cdot \gamma(t, S(t), X(t), \pi(t)) \, dW(t) \\
S(0) &= s \\
X(T) &= g(S(T))
\end{align*}
\] (11)

**Definition 2.2.** ([6]) A triple \((S, X, \pi)\) is called an adapted solution of FBSDE (11) if \( S, X \) and \( \pi \) are \( \{F_t\} \)-adapted, square-integrable stochastic processes satisfying (11) a.s.

The assumptions in the previous section guarantee that the FBSDE (11) has a unique adapted solution, as shown in [6].
We now intend to obtain a partial differential equation characterising the FBSDE (11). Denote $S^x_y := (S(t)|S(\tau) = y)$ and $X^x_y := V(t, S^x_y)$. Then by the Itô’s formula the price of the option written on the stock has the following dynamics:

\[
dV(t, S^x_y) = \frac{\partial V}{\partial t} dt + \frac{\partial V}{\partial S} dS + \frac{1}{2} \frac{\partial^2 V}{\partial S^2} (dS)^2
\]

\[
= \left( \frac{\partial V}{\partial t} + \frac{\partial V}{\partial S} Sb + \frac{1}{2} \frac{\partial^2 V}{\partial S^2} [\sigma + \arctan(X(t)^2 + \pi(t)^2)]^2 \right) dt
\]

\[
+ \frac{\partial V}{\partial S} S \cdot [\sigma + \arctan(X(t)^2 + \pi(t)^2)] dW(t),
\]

where $V(t, S^x_y)$ denotes the option price. At the same time, as can be seen from SDE (4):

\[
dX(t) = (\pi(t)b + (X(t) - \pi(t)) \cdot r) dt + \pi(t) \cdot [\sigma + \arctan(X(t)^2 + \pi(t)^2)] dW(t)
\]

Matching the $dt$ and $dW$ terms we find the following:

\[
\pi(t) = \frac{\partial V}{\partial S}
\]

\[
\frac{\partial V}{\partial t} + \frac{1}{2} \frac{\partial^2 V}{\partial S^2} S^2 \left[ \sigma + \arctan(X(t)^2 + \pi(t)^2) \right]^2
\]

\[
= \left( V - \frac{\partial V}{\partial S} S \right) \cdot r
\]

Condition (14) gives us the hedging strategy right away. Combining it with equation (15) produces the following PDE:

\[
\frac{\partial V}{\partial t} + \frac{1}{2} \frac{\partial^2 V}{\partial S^2} S^2 \left[ \sigma + \arctan(X(t)^2 + \pi(t)^2) \right]^2 - \left( V - \frac{\partial V}{\partial S} S \right) \cdot r = 0
\]

This is a quasi-liner parabolic partial differential equation that needs to be solved in order to obtain function $V(t, S(t))$, which represents the option price, as well as its first derivative $\frac{\partial V}{\partial S}$, which is needed to compute the hedging strategy.
Obtaining an analytical solution for PDE (16) is rather formidable. Instead, it can be solved by means of the Finite Differences method. In particular, we are going to employ the Predictor-Corrector method, which is successfully used to solve non-linear PDEs and described in [8].

The idea of the method is the following. First, we split the time interval \([0,T]\) and the space interval \((0,S_{\text{max}})\) into \(N\) and \(M\) subintervals, correspondingly, and denote by \(V_{t,n} := V(t \cdot \Delta t, n \cdot \Delta S)\), where \(\Delta t := \frac{T}{N}\) and \(\Delta S := \frac{S_{\text{max}}}{M}\). Then, we write down an explicit finite difference scheme for PDE (16). We use the backward finite difference in time and the central finite difference in space in order to approximate the first order partial derivatives of function \(V(t, S(t))\) with respect to \(t\) and \(S\) respectively.

\[
V_{t-1,n} = V_{t,n} + \Delta t \cdot \left[ \frac{1}{2} \cdot \frac{V_{t,n+1} - 2V_{t,n} + V_{t,n-1}}{(\Delta S)^2} \cdot \left[ \sigma + \arctan \left( V_{t,n}^2 + \left( \frac{V_{t,n+1} - V_{t,n-1}}{2\Delta S} \right)^2 \right) \right]^2 - \left( V_{t,n} - \frac{V_{t,n+1} - V_{t,n-1}}{2\Delta S} \right) \cdot r \right]
\]

(17)

The value \(V_{t-1,n}\) obtained this way is a preliminary estimate, which is to be "corrected" through the use of the implicit scheme, which looks as follows:

\[
V_{t-1,n} = V_{t,n} + \Delta t \cdot \left[ \frac{1}{2} \cdot \frac{V_{t-1,n+1} - 2V_{t-1,n} + V_{t-1,n-1}}{(\Delta S)^2} \cdot \left[ \sigma + \arctan \left( V_{t-1,n}^2 + \left( \frac{V_{t-1,n+1} - V_{t-1,n-1}}{2\Delta S} \right)^2 \right) \right]^2 - \left( V_{t-1,n} - \frac{V_{t-1,n+1} - V_{t-1,n-1}}{2\Delta S} \right) \cdot r \right]
\]

(18)

In the traditional case, the use of the implicit scheme (18) would involve a solution of a non-linear system of equations, which would require numerical methods for itself. Instead, the predictor-corrector method suggests using the preliminary estimate of \(V_{t-1,n}\) obtained in (17) to substitute for the corresponding values on the right hand side of (18). Thus, we obtain a
The algorithm is exited after a predefined number of iterations, or when a certain measure of distance between the consecutive solutions falls below a pre-specified value.

The initial condition in this numerical scheme comes from the option’s pay-off function, while boundary conditions are obtained by using the put-call parity.

3 Application: Asset-Liability management in life insurance

This section relies on material in [16]. The central problem in insurance business is solvency. In simple words, a company is called solvent if it is able to meet all its financial obligations. Formally,

**Definition 3.1. [16]** A company is solvent at time $t$ if

$$\mathcal{E}_t[S] \geq \mathcal{E}_t[VaPo], \quad \forall t \geq 0,$$

where $VaPo$ is a valuation portfolio, which can be viewed as a replicating portfolio for the insurance liabilities in terms of financial instruments; $S$ is the existing asset portfolio of the insurance company; $\mathcal{E}$ is the accounting principle used to value assets and liabilities. A usual choice is the economic accounting principle, which corresponds to the valuation at current market prices.

The goal of asset-liability management is to maximize returns on company’s assets under the solvency constraint (20). An obvious way to guarantee solvency is to invest in the valuation portfolio, i.e. company’s assets consist of $VaPo$ and a certain excess capital $F$, which must always be non-negative. However, a mismatch between the actual company’s assets and
*VaPo* is often preferred due to company’s desire to maximize its returns. In this case solvency is guaranteed through the use of a Margrabe option. The latter is the right to exchange one asset for another at a pre-specified time moment. We consider a time period \([t, t+1]\). At \(t\) we decompose company’s assets as follows:

\[
S = \tilde{S} + M + F,
\]

where \(\tilde{S}\) is an asset portfolio satisfying the solvency condition (20); \(F\) denotes free reserves or excess capital; and \(M\) denotes the value of the Margrabe option, giving the holder the right to switch from \(\tilde{S}\) to *VaPo* at time \(t+1\), if needed.

The objective now is to price such an option in case company’s trading actions can affect market prices. This situation is not unrealistic, as insurance companies normally manage large asset portfolios and their market trades can also be quite substantial. To formalize the problem we consider two stochastic processes

\[
Y_t = \mathcal{E}_t[\tilde{S}],
\]

and

\[
V_t = \mathcal{E}_t[VaPo],
\]

Then by the risk-neutral valuation the price of the Margrabe option is given by

\[
P(t,t+1) \cdot \mathbb{E}^*[V_{t+1} - Y_{t+1}|\mathcal{F}_t],
\]

where \(P(t,t+1)\) is a zero coupon bond maturing at \(t+1\) and \(\mathbb{E}^*\) is an expectation under the corresponding risk-neutral measure. Changing the numeraire from \(P(t,t+1)\) to \(V_t\) and introducing a new process \(\bar{Y}_t := \frac{Y_t}{V_t}\), (24) can be rewritten as follows

\[
V_t \cdot \mathbb{E}^{**}[1 - \bar{Y}_{t+1}|\mathcal{F}_t],
\]

where \(\mathbb{E}^{**}\) is an expectation with respect to the risk-neutral measure under a new numeraire. Thus, the price of the Margrabe option is equivalent to the price of a European put written on the artificial asset \(\bar{Y}\), having strike \(K = 1\) and maturing at \(t+1\).
4 Numerical simulations

We assume that $\hat{Y}$ has the following dynamics

$$
\begin{align*}
\begin{cases}
    d\hat{Y}(t) &= \hat{Y}(t) \cdot \gamma(X(t), \pi(t)) \, dW \\
    \hat{Y}(0) &= 100,
\end{cases}
\end{align*}
$$

(26)

where the volatility function $\gamma(X(t), \pi(t))$ is the same as in (10). For the sake of computational stability we assume the initial value of $\hat{Y}$ equal 100 instead of 1. The replicating portfolio has the following dynamics

$$
\begin{align*}
\begin{cases}
    dX(t) &= (X(t) - \pi(t)) r \, dt + \pi(t) \cdot \gamma(X(t), \pi(t)) \, dW(t) \\
    X(T) &= g(\hat{Y}(T)),
\end{cases}
\end{align*}
$$

(27)

where the terminal condition is given by the smoothed European Put pay-off function with strike equal to 100, as in (8)

$$
g(x) = \frac{1}{2} \cdot \left( 100 - x + \sqrt{(x - 100)^2 + \alpha} \right),
$$

(28)

where we take $\alpha = 10$. The PDE to solve is the same as (16), with boundary conditions coming from the well known Put-Call parity condition:

$$
\begin{align*}
\begin{cases}
    V(t, \hat{Y}_{max}) \approx 0 \\
    V(t, 0) \approx e^{-r(T-t)},
\end{cases}
\end{align*}
$$

(29)

where $T$ is the time of maturity of the option, assumed equal 1. The space interval, represented by possible values of $\hat{Y}$ is assumed to be given by $(0, \hat{Y}_{max}) = (0, 200)$. The time interval spans from 0 to 1. To apply the Finite Difference approach we split the time and space intervals into $N = 2000$ and $M = 20$ subintervals, correspondingly. Also, we use 3 iteration cycles in the predictor-corrector scheme. The results of the numerical simulations are shown below.

Figure 1 shows the dependence of the price of the European Put option for a large investor on time and the price of the underlying asset. The same graph for a small investor is shown in Figure 2. A comparison graph in Figure 3 shows the difference between option prices for the large and small investors. It also contains a graph of a theoretical option price computed according to the Black-Scholes formula.

As could have been expected, the price for the large investor is higher than that for a small investor, highlighting the fact that large investor’s attempts to hedge the option would result in higher volatility of the underlying. It is also seen at the graphs that the derivative $\frac{\partial V}{\partial S}$ used in computing
Figure 1: The value of the European Put option for the "large" investor.

Figure 2: The value of the European Put option for a "small" investor.
Figure 3: Price of a European Put option for the "large" and "small" investors.

Figure 4: Price of a European Put option for the "large" and "small" investors. Black-Scholes volatility has been increased from 0.2 to 0.4.
the hedging strategy, is always positive, implying that the investor must always go long in the underlying asset to hedge the option.

The qualitative situation does not change if we alter the underlying Black-Scholes volatility. Figure 4 shows comparison results for increased Black-Scholes volatility. We still observe that the large investor’s price is considerably higher than the Black-Scholes one. Our computations confirm that a large investor suffers from adverse conditions when it comes to pricing and hedging options. The necessity to take into account increased volatility of the underlying makes large investor’s hedging strategy more expensive, thus making the option itself more valuable. As for the practical situation under our investigation, one can conclude that buying a Margrabe option from an investor who uses the Black-Scholes formula is a good deal for the large investor, since the replicating alternative would cost him more. Large investor’s direct market involvement in an attempt to construct a replicating portfolio for the Margrabe option would mean a long position in the underlying, thus resulting in higher asset volatility. This situation, in its turn, would compel the large investor to expand his long position even further, leading to higher replicating costs. Our main recommendation to an insurance company in this situation would be to avoid option replication and instead try to buy the Margrabe option from the market.

References


