

ANALYSIS OF THE EXCHANGE RATE AND PRICING FOREIGN CURRENCY OPRIONS ON THE CROATIAN MARKET: THE NGARCH MODEL AS AN ALTERNATIVE TO THE BLACK-SCHOLES MODEL

Petra POSEDEL, MSc
Faculty of Economic and Business, Zagreb

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Abstract

The interest of professional investors in financial derivatives on the Croatian market is steadily increasing and trading is expected to start after the establishment of the legal framework. The quantification of the fair price of such financial instruments is therefore becoming increasingly important. Once the derivatives market is formed, the use of the Black-Scholes option pricing model is also expected. However, contrary to the assumptions of the Black-Scholes model, research in the field of option markets worldwide suggests that the volatility of the time-series returns is not constant over time. The present study analyzes the implications of volatility that changes over time for option pricing. The nonlinear-in-mean asymmetric GARCH model that reflects asymmetry in the distribution of returns and the correlation between returns and variance is recommended. For the purpose of illustration, we use the NGARCH model for the pricing of foreign currency options. Possible prices for such options having different strikes and maturities are then determined using Monte Carlo simulations. The improvement provided by the NGARCH model is that the option price is a function of the risk premium embedded in the underlying asset. This contrasts with the standard preference-free option pricing result that is obtained in the Black-Scholes model.

Key words: Black-Scholes model, NGARCH model, heteroscedasticity, volatility, risk premium, risk-neutral measure, no arbitrage, Monte Carlo simulations.

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1 Introduction

An explosive increase in trading with different financial instruments on the Croatian market inspires and motivates us to model prices of various assets, foreign currency and exchange rates and their volatilities. So far, trading with derivatives on the Croatian market is only of over-the-counter (OTC) type. The Croatian market for nonlinear derivatives-options does not exist yet, but the possibility of its being formed attracts the interest of many investors on the Croatian capital market. It is, thus, of increasing importance to quantify the *fair* price of such objects. We can not focus our research on data from options markets because these do not exist yet. Since the debate on whether to introduce a derivatives market is still open, this work tries to make another contribution on possible prices of European call options for assets. First of all, such a goal requires an appropriate econometric modelling of price dynamics of assets and their volatilities since option pricing formulas are functions of some or all parameters present in the model describing the price dynamics of assets. One of the great successes of modern financial economics is option pricing. Based on the well-known Law of One Price or no-arbitrage condition, the option pricing models of Black and Scholes (1973) and Merton (1973) gained an almost immediate acceptance among academics and investment professionals of a kind unparalleled in the history of economic science.¹

However, contrary to the assumptions of the Black-Scholes model, research in the field of option markets worldwide (e. g. Mandelbrot (1963) and Fama (1965) in their works base their empirical studies on log-returns of several US stocks) suggest that the volatility of time series returns is not constant over time. In fact, during *stress periods* on the market (political changes or disorders, economic crises, but also not so drastic changes like announcements of macroeconomics data) prices of financial assets fluctuate very much and the volatility changes over time. In this case we say that the process of interest is heteroscedastic. The first plausible success of econometrical modelling of the heteroscedasticity of time series is attributed to Engle (1982) who introduced the ARCH model. Various extensions of the ARCH model were subsequently made by many researchers. Bollerslev (1986) and Taylor (1986), independently of each other, suggested the generalized ARCH, i.e. the GARCH model, and Nelson (1991) the exponential GARCH, i.e. the EGARCH model. The leverage effect and other GARCH extensions are described in the works of Ding, Granger and Engle (1993), Glosten, Jagannathan and Runkle (1993) and Hentschel (1995). Bollerslev, Chou and Kroner (1992) and the work of Duan (1997) encompassed the existing GARCH models into a common system known as the augmented GARCH (p,q) process.

On the Croatian market, Šestović and Latković applied the GARCH (1,1) model for analyzing stocks of PLI-AA, ZAB-O and of the CROBEX index. The authors conclude that using models from the GARCH family improves predictions of market risk.

Models for option pricing that incorporate heteroscedasticity include, among others, the model of Merton (1976), Geske (1979), Rubinstein (1983) and the GARCH model of Duan (1995). In their works, the authors conclude that models that account

¹ See Bernstein (1992), Chapter 11, for a lively account of the intellectual history of the Black-Scholes-Merton option pricing formula.

heteroscedasticity can reflect changes in the conditional variance of assets. Furthermore, the numeric analysis in these works suggests that the proposed models can potentially explain some well-documented systematic biases associated with the Black-Scholes model, like underpricing of out-of-the-money options.

We ask if the situation regarding pricing foreign currency options is somehow similar to pricing equity options? Cooper et al. (1986) estimate the parameters of the stochastic process describing changes in volatilities and compare the predicted prices to Black-Scholes prices afterwards. Their results are similar in nature to those found in studies on stock options.

Prices in option markets are commonly quoted in terms of the Black-Scholes implied volatility. In Rebonato's terms (1999), the implied volatility is a "wrong number which, plugged into the wrong formula, gives the right answer!" This does not mean that market participants believe in the hypotheses of the Black-Scholes model – they do not: the Black-Scholes formula is not used as a pricing model for vanilla options but as a tool for *translating* market prices into a representation in terms of implied volatility.

The interest of participants on the Croatian market have been lately attracted by interventions of the Croatian National Bank on the domestic market resulting in changing the market value of the domestic currency. Seasonality trends typical for domestic currency and the possibility of changing the domestic currency in the future motivate us to explore different kinds of foreign currency options, with different strikes and maturities. The goal of this work is to analyze the empirical distribution of the EUR/HRK currency time series and the consequences that such analysis has for pricing foreign currency options. The primary goal of this study is to estimate the parameters appearing in the dynamics of the currency time series using the nonlinear-in-mean, asymmetric GARCH model. The obtained parameters are then used for simulating prices of foreign currency options implementing the NGARCH option pricing model. Furthermore, we compare the obtained prices with the Black-Scholes ones. The following addressed issue is to explore if the heteroscedasticity assumption for changes in the time series present in the NGARCH model reflects visible differences in option prices. Doing this we are trying to connect the very popular econometric GARCH model and the literature all over about contingent claims.

The article is organized as follows. In the second part of the work we introduce the non linear asymmetric GARCH model for describing the EUR/HRK currency dynamics for short time horizons in the period 2001-2005, we describe the properties of the model and estimate the parameters. Option pricing in the NGARCH framework is presented in the third part. Option prices for foreign currency implementing the NGARCH model and the analysis among different strikes and maturities are described in the fourth part and compared to prices derived from the constant volatility model of Black and Scholes. In the conclusion we address future issues and research.

2 Modelling the exchange rate with the NGARCH model

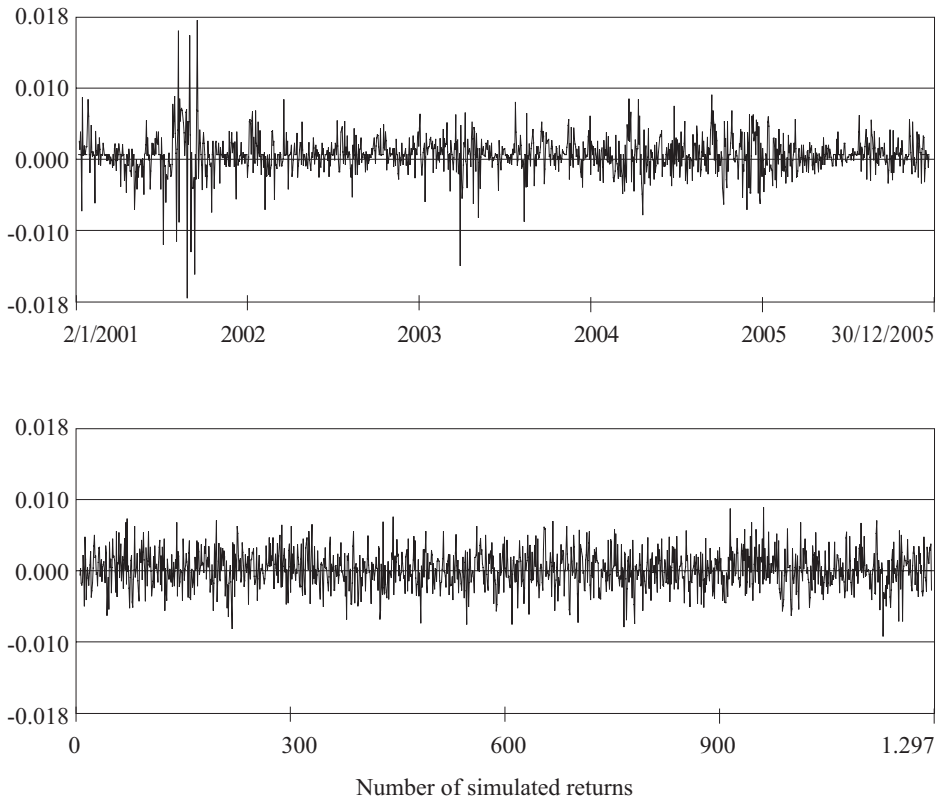
Models for time series of assets have a very important role in risk management that makes decisions on the basis of the observed assets prices C_t in discrete time. Relevant

values for investors are log-returns,² i.e. the differences in the closing prices during a one day period,

$$P_{t+1} = \ln C_{t+1} - \ln C_t. \quad (1)$$

Figure 1 compares increments of a Brownian motion with the returns on the EUR/HRK exchange rate with the same average volatility. While both return series have the same variance, the Brownian model achieves it by generating returns which always have roughly the same amplitude and manifest frequent large peaks representing some unexpected change on the market.

Figure 1 Daily log-returns for the EUR/HRK exchange rate



Note: (Top) compared to the log-returns of the Black-Scholes model (bottom) with same annualized mean and variance

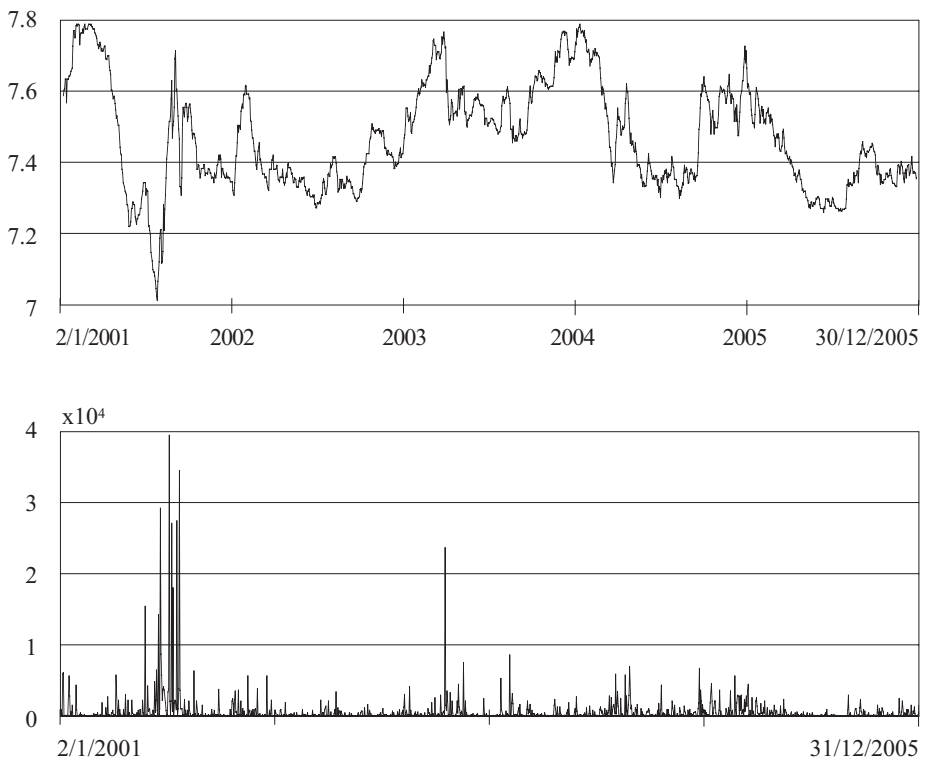
² In the remaining text we will call log-returns simply returns.

A lot of empirical research indicates the increasing need for a more realistic model for financial assets than the Brownian one. Experience has shown that returns volatility for most financial instruments is not constant over time. Rather, days of high (low) volatilities are followed by days of high (low) volatilities, a property known as *clustering*. For the purpose of estimating the variability of returns, the variable representing the squared innovation return is of crucial importance,

$$S_t^2 = (P_t - \mu)^2, \quad \mu = E(P_t). \quad (2)$$

Figure 2 describes the changes in the EUR/HRK exchange rate prices in the period from 2001 to 2005. It can be immediately noticed that some periods have very different volatilities, which can be also seen from the graph of the squared innovation returns.

Figure 2 The dynamics of the EUR/HRK exchange rate and squared innovations in the period 2001-2005



Serial dependence in data which is described by correlations among returns for different lags is shown in figure 3. The solid line indicates that autocorrelations of daily returns have small values and do not indicate any systematic patterns. On the other hand,

autocorrelations of squared daily returns have positive values for small lags and then go exponentially to zero when the number of lags increases. This shows that squared returns values do not forget the recent past and that they are positively correlated which can be seen from the dotted line in figure 3.

At first glance it may perhaps seem that the problem of pricing call options is solved by the use of the standard Black-Scholes formula, where we use domestic rate r_d as the short rate of interest, and the stock price C is replaced by the exchange rate. This line of argument is, however, incorrect and the reason is as follows. When we buy a stock (without dividends) we, in principle, buy a piece of paper which we keep until we sell it. On the other hand, when we buy a foreign currency we will, on the contrary, not just keep the physical currency bills until we sell them again. Instead we will typically put the currency into an account where they will grow at a certain rate of interest. The obvious implication of this fact is that a foreign currency plays very much the same role as a domestic stock with a continuous dividend. Thus, modelling foreign currency requires the introduction of the foreign interest rate in standard models for stocks that will play an important role in pricing currency options.

2.1 Description of the model

Suppose we are observing the discrete economy and denote with C_t the EUR/HRK exchange rate price in time t , defined as the number of Croatian kunas required to purchase one euro. The dynamics of the time series of return P_t is described with a nonlinear-in-mean, asymmetric GARCH (1,1)³ model (Engle and Ng, 1993):

$$P_{t+1} \equiv \ln\left(\frac{C_{t+1}}{C_t}\right) = r_d - r_s + \lambda\sigma_{t+1} - \frac{1}{2}\sigma_{t+1}^2 + \sigma_{t+1}Z_{t+1}, \quad (3)$$

$$\sigma_{t+1}^2 = \omega + \alpha(\sigma_t Z_t - \rho\sigma_t)^2 + \beta\sigma_t^2, \quad (4)$$

where Z_t are i.i.d. standard normal random variables $N(0,1)$ and

$$\omega > 0, \quad \alpha \geq 0, \quad \beta \geq 0 \quad \text{and} \quad \alpha(1 + \rho^2) + \beta < 1 \quad (5)$$

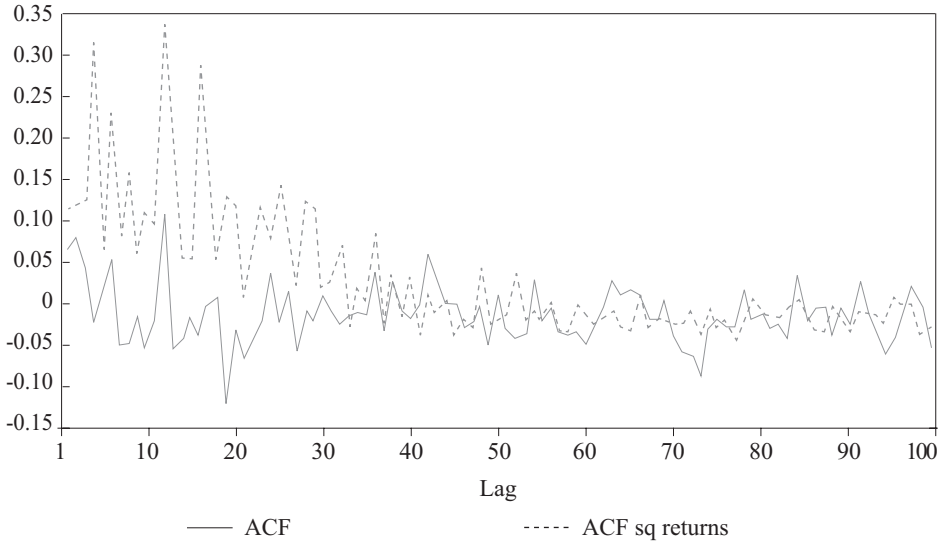
in order to insure the non negativity and stationarity of the variance process σ_t^2 . Variables r_d and r_s denote, respectively, the constant one period riskless domestic and foreign interest rate,⁴ while λ is the constant risk premium, i.e. the reward for investing in the foreign currency. The value of the premium will influence the conditional variance of the process

³ It is worthwhile to mention that very often in literature the asymmetric GARCH model is called NGARCH.

⁴ In this work, for illustration purposes, r_d and r_s are taken as constants of known values. It is possible to consider r_d and r_s as time series and in that case we estimate them from national treasury bills.

when doing option pricing.⁵ Special attention to the model is given by the parameter of asymmetry ρ which describes the correlation between returns and variance.⁶

Figure 3 Correlations of daily returns and squared daily returns



One of the biggest advantages of the GARCH model for risk management is the possibility of one-day-ahead forecasting of various quantities relevant for investors. From definition it immediately follows that σ_{t+1}^2 i.e. tomorrow's variance, is known at the end of today's t . Let us denote with $E_t[P_{t+1}]$ the expected⁷ return in time t . From expression it follows that the expected return and variance of the return P_{t+1} based on the information available until time t are given by

$$E_t[P_{t+1}] = r_d - r_s + \lambda \sigma_{t+1} - \frac{1}{2} \sigma_{t+1}^2 \quad \text{i} \quad \text{Var}_t[P_{t+1}] = \sigma_{t+1}^2. \quad (6)$$

Conditions ensure the stationarity of the variance process (σ_t^2) so we can define the unconditional variance as

⁵ Let us immediately notice that a negative value of λ decreases the mean value EUR/HRK exchange return which indicates the appreciation of the domestic currency. Furthermore, if we are observing the HRK/EUR exchange rate a positive value of the parameter λ will explain the appreciation of the domestic currency.

⁶ In the case of analyzing stock returns, the positive value of ρ reflects the empirically well known *leverage effect* indicating that a downward movement in the price of a stock causes more of an increase in variance more than a same-value downward movement in the price of a stock, meaning that returns and variance are negatively correlated.

⁷ $E_t[X]$ and $\text{Var}_t[X]$ are called conditional mean and variance of the variable X since we are conditioning the calculation on the set of available information until time t .

$$\sigma^2 \equiv E[\sigma_{t+1}^2] = \frac{\omega}{1 - \alpha(1 + \rho^2) - \beta}. \quad (7)$$

From relation (6) it follows that the forecast of the variance is directly given by the model with σ_{t+1}^2 . If we observe the forecast of daily returns variance for k periods ahead, using the recursive specification of the asymmetric GARCH model (4) it follows that

$$E_t[\sigma_{t+k}^2] = \sigma^2 + [\alpha(1 + \rho^2) + \beta]^{k-1}(\sigma_{t+1}^2 - \sigma^2), \quad (8)$$

where σ^2 is defined with $E_t[\sigma_{t+k}^2]$ represents the expected value of the future variance for horizon k . The expression $\alpha(1 + \rho^2) + \beta$ is called the persistence of the model. From expression it follows that if the value of $\alpha(1 + \rho^2) + \beta$ is near 1 then shocks in the market persist through a long time ($k \rightarrow \infty$). In that case we will say that the time series has a long memory. On the other hand, small values for $\alpha(1 + \rho^2) + \beta$ indicate that shocks in returns die out more quickly in time.

2.2 The maximum likelihood method

Since the conditional variance σ_{t+1}^2 is a non-observable variable, it has to be estimated along the other parameters of the model $\omega, \alpha, \beta, \lambda, \rho$. For the econometric analysis we use $T = 1.297$ daily returns of the exchange rate EUR/HRK in the period from January 1, 2001 to December 30, 2005.⁸ Data represent the inter-banking closing offers for purchasing foreign currency. For the daily domestic interest rate r_d and the foreign interest rate r_s we take the values 0.0131% and 0.0115% respectively.⁹ By assumption, (Z_t) is an i.i.d. $N(0, 1)$ sequence of random variables so the log-likelihood function has the form

$$L_T = \frac{1}{T} \sum_{t=1}^T \left[-\frac{1}{2} \ln(2\pi) - \frac{1}{2} \ln(\sigma_t^2) - \frac{1}{2} \frac{(P_t - (r_d - r_s + \lambda \sigma_t - \frac{1}{2} \sigma_t^2))^2}{\sigma_t^2} \right], \quad (9)$$

where T denotes the number of observed data.

Let us denote with $\theta = (\omega, \alpha, \beta, \rho, \lambda)$ the set of unknown parameters. It is necessary to find that vector of parameters θ for which the function L_T achieves the maximum value under conditions given by (5). The maximization of the function L_T by the parameters of the model is done using the numerical algorithm for finding the maximum of the function under certain constraints on the parameters. The values of the estimated parameters are shown in Table 1.¹⁰

⁸ The author thanks Privredna banka Zagreb for the given data.

⁹ On the annual level the respective rates are approximately 3.3% and 2.9%. The analysis of course can be done for any other choice of these values. Since the frequency of data is at a daily level, we transform the annual interest rate into a daily interest rate by dividing by 252 the average number of business days in one year.

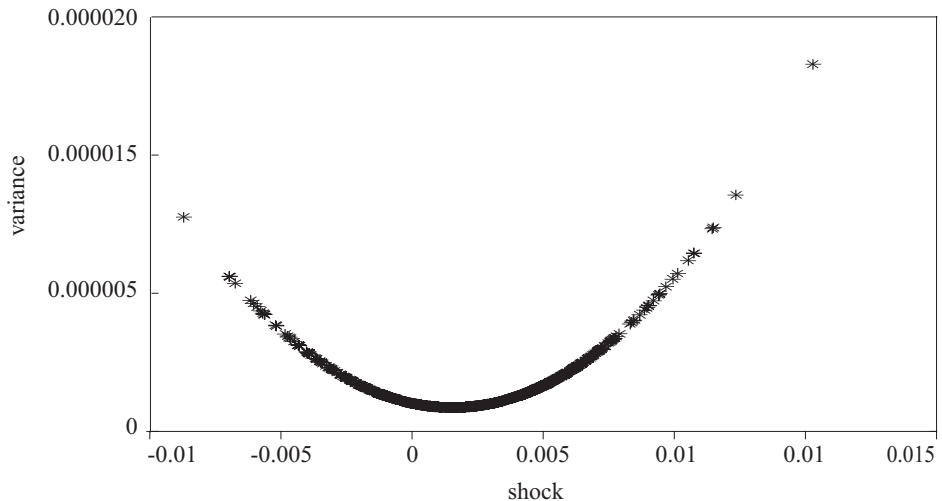
¹⁰ Parameters are obtained using the program packages Gauss and Excel independently.

Table 1 Parameter estimates

Parameter	Value	Sampling standard error
$\hat{\omega}$	1.7339×10^{-7}	2.92×10^{-8}
$\hat{\lambda}$	-0.0301153	0.11311
$\hat{\alpha}$	0.095345	0.012028
$\hat{\beta}$	0.86840994	0.014289
$\hat{\rho}$	-0.1707379	0.074885
$\hat{\alpha}(1+\hat{\rho}^2) + \hat{\beta}$	0.96653484	—

The GARCH asymmetry parameter ρ is negative indicating that an increase in the volatility of the exchange rate will be greater with appreciation of the foreign currency, or alternatively, with the depreciation of the domestic currency. We can immediately notice that the parameter ρ would be of the opposite sign if we consider the time series HRK/EUR. The news impact curve is shown in figure 4 in order to illustrate the level of asymmetry. Furthermore, the estimated risk premium is not statistically significant.¹¹ From the econometric analysis we also notice that the value of $\alpha(1+\rho^2) + \beta$ is very near to 1 indicating the long memory effect in the series. In other words, a shock today will have an influence on the future. In any case, in making interpretations for longer horizons one should be careful to check if there exists any tendency to keep the price around a certain level, which is not such a strange happening in exchange rates practice.

Figure 4 News impact curve



¹¹ By significance we refer to the significance at the 95% confidence level.

On the other hand, from the econometric analysis of the Brownian model, returns (P_1, \dots, P_T) are i.i.d. random variables defined by

$$P_t = \mu + \sigma W, \quad \mu \in R, \sigma > 0 \tag{10}$$

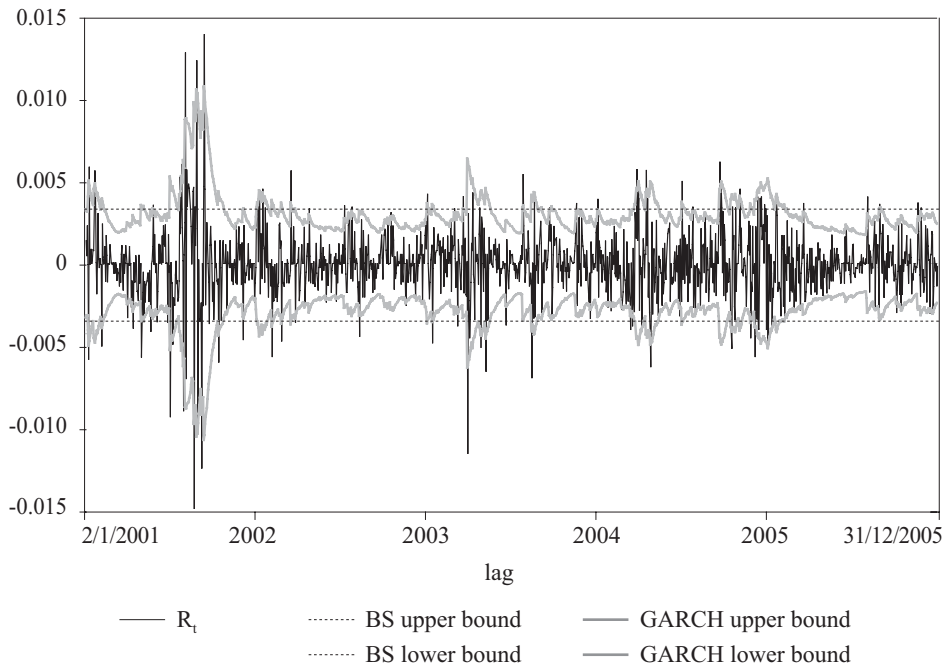
where $W \sim N(0,1)$ is the standard normal random variable. The parameter μ denotes the average log-return and σ is the volatility of returns. The average expected value and the variance for the given time series can be consistently estimated with

$$\hat{\mu} = \frac{1}{T} \sum_{t=1}^T P_t, \quad \hat{\sigma}^2 = \frac{1}{T} \sum_{t=1}^T [P_t - \hat{\mu}]^2, \tag{11}$$

with asymptotic variances σ^2/T and $2\sigma^4/T$ respectively. From the data it follows that

$$\hat{\mu} = -1.96838 \times 10^{-5} \quad (6.60515 \times 10^{-5}) \quad \hat{\sigma}^2 = 5.65855 \times 10^{-6} \quad (2.22203 \times 10^{-7}). \tag{12}$$

Figure 5 Returns and risk estimations



So the estimated volatility constant value is $\hat{\sigma} = \sqrt{\hat{\sigma}^2} = 0,0023788$. The corresponding standard errors are given in brackets. The parameter μ is not statistically significant (t statistic -0.298), so it is excluded from further analysis. Figure 5 shows the returns of exchange rates and estimated risk in two ways. One approach uses the constant estimated volatility $\hat{\sigma} = 0.0023788$. In the other approach we use the estimation deriving from the NGARCH model given by (4) using the parameters from Table 1. In the graph we plotted the interval of $1,65 \sigma$ in order to obtain the interval inside which we expect the absolute value of the next day return under confidence of 95%. This value is often called VaR, i. e. value at risk. The dotted line represents the interval obtained under the implementation of the Brownian model where on the other hand the solid line denotes the interval obtained from the NGARCH specification. It can be seen that in calm periods in the market the nonlinear asymmetric GARCH model gives lower and more appropriate volatility predictions than the Brownian model does. The constant volatility deriving from the Brownian model suffers from the intensive change periods on the market represented by high volatilities (e.g. 2001) so it gives higher values for calm periods on the market.

For further analysis which has as a goal option pricing, we fix the estimated values of parameters and consider them constants that dominate the model.

3 Option pricing

An option derives its value from an underlying asset but its payoff is not a linear function of the underlying asset price, and so the option price is not a linear function of the underlying price either. This nonlinearity results in complications for pricing and risk management.

3.1 The European call option on a foreign currency under the NGARCH model

We will focus the analysis of option pricing on European options that can be executed only on the maturity date.¹² In this section we present the model for pricing foreign currency options using results from Duan (1995).

The European call (put) option on foreign currency¹³ gives to the holder of the option the right to buy (sell) foreign currency on maturity T at the strike price K . The maturity date can be expressed in the number of days $\tau > 0$ remaining to maturity.

We will denote with co the price of the European call option. Days to maturity are counted in calendar days (365 in a year) and not in business days. We say that the option at time t is *in the money* if $C_t > K$ *at the money* if $C_t = K$, and *out of the money* if $C_t < K$. In order to give the *fair* price of the option it is necessary to introduce some risk neutral criterion. This does not mean that we assume that market participants are risk neutral but that for option pricing purposes we use the *equilibrium price measure*. From the economic point of view this is a very logical assumption. In fact, under no arbitrage conditions

¹² American options can be executed at any date up to and including maturity.

¹³ The call-put parity (20) enable us to focus the analysis on European call options and the price of the put option can be determined using that relation.

in the risk neutral world should the present value of the option be the expected value of all discounted future payoffs deriving from the asset. We say that an equilibrium price measure for the domestic market satisfies the *local risk neutral valuation relationship*, LRNVR,¹⁴ if for any asset value X_t which is measured in domestic currency the following conditions are satisfied:

- X_{t+1} / X_t is conditionally log-normal distributed with respect to the equilibrium measure

$$E_t^* [X_{t+1} / X_t] = e^{rd}, \quad (13)$$

$$Var_t^* [\ln(X_{t+1} / X_t)] = Var_t [\ln(X_{t+1} / X_t)], \quad (14)$$

where E_t^* , Var_t^* denote, respectively, the conditional expectation and variance with respect to the equilibrium measure. In order to calculate those conditional expectations and variance under the previous conditions it is necessary to specify the process under a new measure.

The process defined by

$$P_{t+1} \equiv \ln(C_{t+1}) - \ln(C_t) = r_d - r_s - \frac{1}{2} \sigma_{t+1}^2 + \sigma_{t+1} Z_{t+1}^*, \quad (15)$$

and

$$\sigma_{t+1}^2 = \omega + \alpha [\sigma_t Z_t^* - (\lambda + \rho) \sigma_t]^2 + \beta \sigma_t^2, \quad (16)$$

where $Z_{t+1}^* = Z_{t+1} + \lambda \sim N(0,1)$ i.e. variables Z_t^* are independent and identically distributed with respect to the measure that is locally risk neutral, satisfies properties 1, 2 and 3 from the previous definition.¹⁵

Relation (15) enables pricing foreign currency options. Furthermore, from relation (16) it follows that the risk premium λ has a global influence on the conditional variance process even if the risk was locally neutralized with respect to the equilibrium measure that satisfies the LRNVR criterion. In fact, in the expectation specification (15) it was substituted by the risk free interest rate and the conditional variance remains unchanged by property 3. This leads to the fact that the option price given by the GARCH model will be a function of the risk premium.

The *fair* price of the European call option, in the risk neutral world, in time t with strike K , and maturity date $t + \tau$, $\tau > 0$ is given by the discounted expected payoff value with respect to the locally risk neutral measure given all the information available up to time t ,

¹⁴ The definition holds also if we observe a foreign asset as long it is measured in domestic currency. It is clear that foreign currency is a special case of foreign asset measured in domestic currency. In that case, because of the rational investing behaviour it holds that $X_t = C_t \exp(r_s t)$.

¹⁵ The proof is given in Appendix 1.

$$co_t = \exp(-r_d \tau) E_t^* [\max(C_{t+\tau} - K, 0)]. \quad (17)$$

Since the distribution of the time aggregated price $C_{t+\tau}$ is not known in analytical form, the conditional expectation E_t^* cannot be computed explicitly. Simulations are therefore needed for computing the average future payoff and the obtained value is then used for estimating the expected value $E_t^*[\cdot]$. Approximations obtained by simulations are called Monte Carlo estimates. When the number of simulations MC is big enough ($MC \rightarrow \infty$), the average value converges on the expectation. Simulations are computed using relations (15) and (16) and are presented in the following section.

It is worthwhile to mention that, contrary to the relation (17), the value of which we cannot compute directly, from the Brownian model (10), under some conditions, the price of the European call option is given by (Campbell, Lo and MacKinlay (1997), Chapter 9)

$$co_t = C_t \exp(-r_s \tau) \Phi(d) - \exp(-r_d \tau) K \Phi(d - \sigma \sqrt{\tau}), \quad (18)$$

where $\Phi(z)$ denotes the standard normal distribution function and d is given by

$$d = \frac{\ln(C_t / K) + \tau(r_d - r_s + \sigma^2 / 2)}{\sigma \sqrt{\tau}}. \quad (19)$$

Formula (18) is known as the Black-Scholes-Merton formula.¹⁶ In the case of pricing foreign currency options, $\Phi(d) \exp(-r_s \tau)$ measures the sensitivity of the option price with respect to the changes in the exchange rate price. This value is called the *delta* of the option and we denote it Δ . Even though we focus our analysis on the European call option, the just obtained formulas can be used for pricing European put options po_t using the so called call-put parity

$$C_t \exp(-r_s) + po_t = co_t + K \exp(-r_d \tau). \quad (20)$$

In the sequel the option price obtained under NGARCH specification will be denoted by co^{GH} , and the one under the Black-Scholes model by co^{BS} . In the sequel we will compare by numerical analysis the NGARCH model for option pricing with the Black-Scholes model. First, let us stress that the GARCH model in its specification also subsumes the Black-Scholes model because the homoskedastic asset return process is a special case of the GARCH model. (for $\alpha = 0$ and $\beta = 0$).

¹⁶ Under no arbitrage conditions and no market imperfections like taxes and transaction costs, Black and Scholes (1973) obtained the formula (18) assuming CAPM to hold. On the other hand, Merton obtained the same formula looking at the self financing portfolio consisting of stocks, options and bonds.

4 Analysis of foreign currency option prices under NGARCH

In order to standardize the option prices that correspond to different maturities and strikes, in option analysis it is a common procedure to use formula (18) in order to calculate the Black-Scholes implied volatility. Since the Black-Scholes option price is an increasing function¹⁷ it is possible to find the unique value of the volatility $\tilde{\sigma}(T, K)$ such that

$$co_t^{BS}(C_t, K, T, \tilde{\sigma}) = C_t^*(T, K), \quad (21)$$

where $C_t^*(T, K)$ represents the observed market option price in time t with maturity T and strike K . For fixed t , the implied volatility $\tilde{\sigma}(T, K)$ depends on option characteristics like maturity date T and strike K . The function $\tilde{\sigma}(T, K)$ is called the implied volatility surface in time t . Option prices on the market are usually quoted in terms of the Black-Scholes implied volatility.¹⁸ For any fixed t we analyze

$$\tilde{\sigma} = co_t^{BS}(C_t, K, \tau, C_t^*)^{-1}. \quad (22)$$

Because of the constant volatility hypothesis, the Black-Scholes model predicts a flat implied volatility surface, or a constant $\sigma = \tilde{\sigma}(T, K)$. There are many documented empirical studies saying that implied volatility is not constant as a function of the strike or as a function of maturity (Cont and de Fonseca (2002), for example, Rubinstein (1985)). When pricing options on stocks, Johnson and Shanno (1987), Scott (1987) and Wiggins (1987) in their works analyzed the effect of stochastic volatility and showed that the predicted Black-Scholes option prices of the European type are overpriced for at the money options. If the option is in the money, the predicted Black-Scholes prices are usually underpriced. For options out of the money, the results they obtained are sensitive to parameters present in the stochastic process that describes the dynamics of the volatility and on correlation between the changes in volatility and stock prices.

In order to analyze if the NGARCH framework can describe the empirical observations on option markets worldwide, in the numerical analysis the option prices are calculated¹⁹ first under the NGARCH specification using the parameters from Table 1, and then, for different days to maturity τ , the graphs $(m, \tilde{\sigma}(m))$ are plotted where m denotes the moneyness defined as $\tilde{\sigma}$ is defined with (22), substituting co^{GH} instead of C^* . As a starting value for the volatility, the estimated unconditional volatility is used, and is computed from parameters in Table 1 by (7). This value is 0.036128 at the annual level assuming 252 days in one year and is plotted with a dotted line in figures 6, 7 and 8. Since it is not significant, we do not consider the parameter λ in further analysis.²⁰

¹⁷ It holds $\sigma \in (0, \infty)$ for every $\frac{\partial co^{BS}}{\partial \sigma} > 0$.

¹⁸ It is worthwhile to mention that from (21) it follows that the inverse function of co^{BS} exists.

¹⁹ All option prices are simulated under 50000 repetitions. The program language MATLAB is used for simulations.

²⁰ From relation (16) it follows that the expression $(\lambda + \rho)$ enters in the option pricing as one parameter so without loss of generality we can consider the case $\lambda = 0$ as long as we analyze the effect of the parameter ρ .

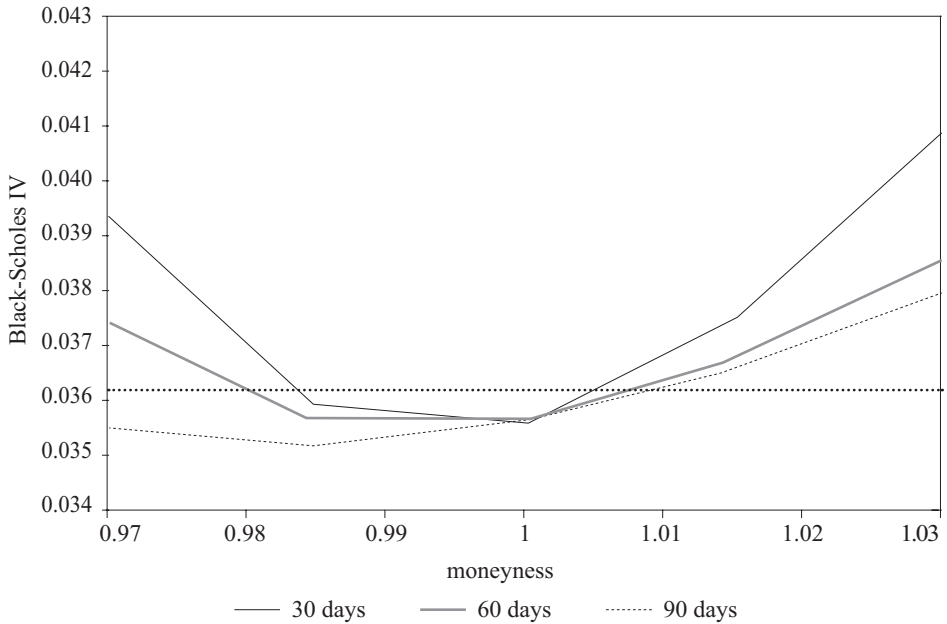
Furthermore, option prices are expressed in terms of days to maturity. Option prices are calculated for different days to maturity ($\tau = 30, 60$ and 90 days) and different moneyness ($m = 0.97, 0.985, 1.0, 1.015$ and 1.03 which corresponds, respectively, to different strikes $K = 7.11495, 7.224975, 7.335, 7.445, 7.555$) for currency spot price $C_t = 7.335$. Using relation (17), the option price obtained by simulations is given by

$$c^{GH} \approx \exp(-r_d \tau) \frac{1}{MC} \sum_{i=1}^{MC} \max\{C_{i,t+\tau} - K, 0\}, \quad (23)$$

where $MC = 50,000$, and for the i simulation we have

$$C_{i,t+\tau} = C_t \exp\left(\sum_{j=1}^{\tau} P_{i,t+j}\right), \quad i = 1, 2, \dots, MC. \quad (24)$$

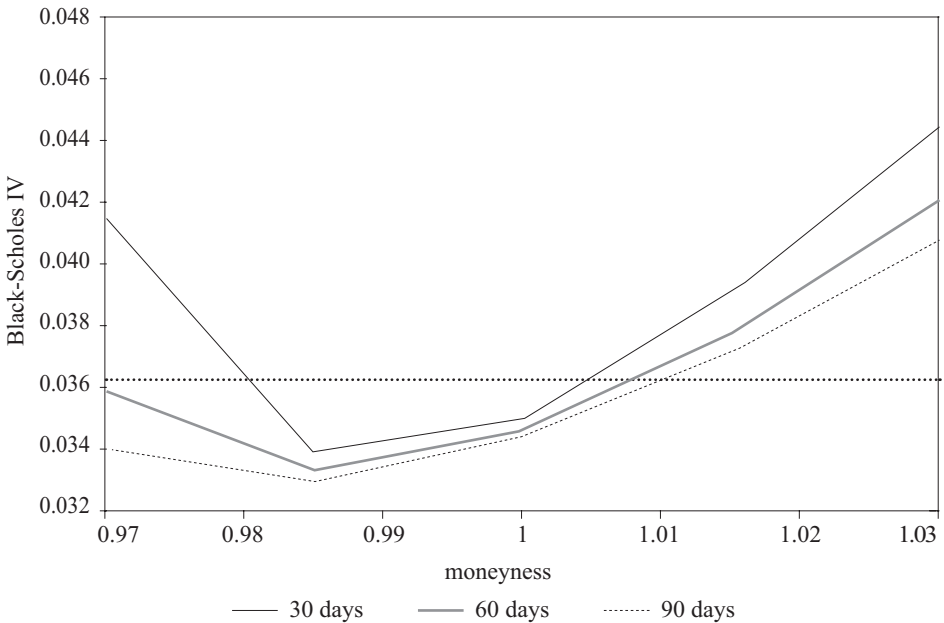
Figure 6 Implied volatility as a function of moneyness for the foreign currency European call option, $\rho = -0.17074$



All obtained values refer to the current option price which is denoted by $t = 0$ and we suppress it from further indexing. The obtained values are then compared to their Black-Scholes counterparts with the average annual volatility 0.036128 . The obtained results are given in Appendix 2. Figure 6 is based on the parameters from Table 1. Mild

negative asymmetry ($\rho = -0.17074$) affects the fact that options out of the money ($K/C > 1$) are underpriced in the Black-Scholes model with constant volatility $\sigma = 0.036128$, while options at the money are overpriced. Since the option price is an increasing function of the volatility, volatilities of higher values than the stationary one indicate a higher option price than the Black-Scholes model would predict for $\sigma = -0.036128$ and the same terminal conditions.

Figure 7 Implied volatility as a function of moneyness for the foreign currency European call option, $\rho = -0.461$.

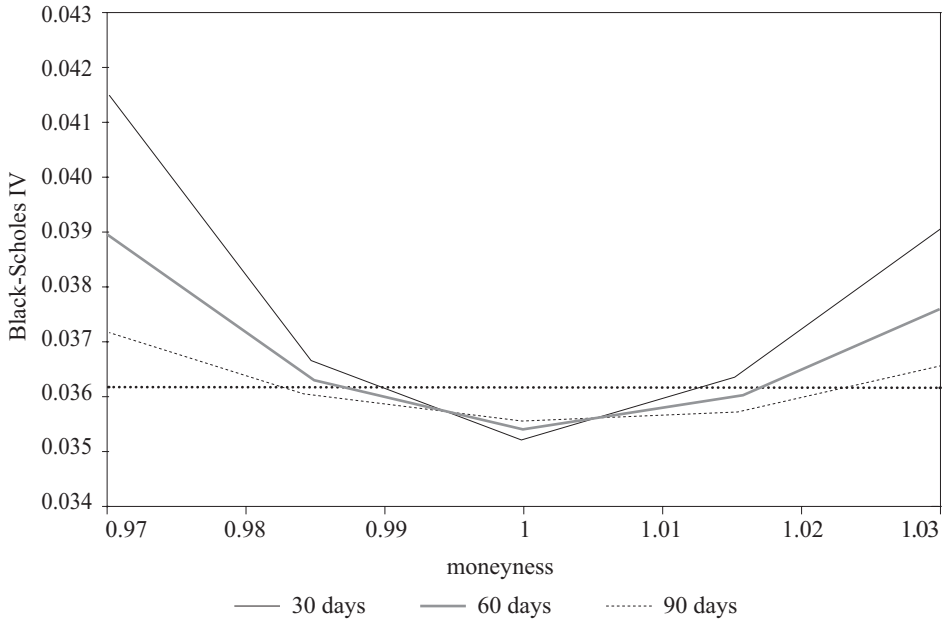


In order to analyze the effect of the asymmetry parameter ρ , we repeat the simulation procedure²¹ for $\rho = -0.461$ and $\rho = 0$ respectively, which is shown in figures 7 and 8. The parameter value ω is changed in order to have the same stationary volatility. In the case of moderate asymmetry $\rho = -0.467$ options out of the money are underpriced in the constant volatility model while some options ($\tau = 30$) in the money are also underpriced in the same model. The overpricing effect in the constant volatility model is now more pronounced for options in the money. In the case of $\rho = 0$ the graph of implied volatility becomes almost symmetric (centred in $K/C = 1$) since the asymmetry is completely absent. From figure 8 it can be seen that options deeply in the money and deeply out of the money are underpriced in the constant volatility model. The obtained values are consistent with the findings of Taylor and Xu (1994) and Duan (1999). For options out of the money, independently of the asymmetry parameter in question, the implied volatility is

²¹ The obtained results are given in Appendix A and B.

a decreasing function of the maturity. For options near the money, ($K/C = 1$), if asymmetry is absent or very mild ($\rho = 0$ or $\rho = -0.17074$), the implied volatility is an increasing function of maturity. This is consistent with findings of Shastri and Wethyavivorn (1987) and Duan (1999).

Figure 8 Implied volatility as a function of moneyness for the foreign currency European call option, $\rho = 0$.



5 Conclusion

The interest of professional investors in financial derivatives on the Croatian market is steadily increasing. No market for nonlinear derivatives in Croatia yet exists, but one is expected to be founded in the near future. In that case, as an alternative to the Black-Scholes model, this work suggests the NGARCH model could be used for option pricing. As an illustration, the consequences that the model has for pricing foreign currency options are presented. The locally risk-neutral pricing measure for the domestic economy is identified. Furthermore, the dynamic of foreign currency prices with respect to the locally risk-neutral pricing measure is presented. The analysis shows that introducing heteroscedasticity results in better fitting of the empirical distribution of foreign currency than the Brownian model. Simulations show that in the NGARCH framework it is possible to described the empirically observed facts from foreign currency option markets. Unlike the preference free Black-Scholes model, the NGARCH model for option pricing incorporates the risk premium in its formulation. It is not a priori obvious what should be the risk premium for the volatility. In this work, we have found it not significant in the

parameter estimation procedure using the time series of the underlying process X . But, once the data from the options market are at our disposal, it will be possible to conduct the analysis and the parameter estimation using data from options market only. In that case, it would be interesting to analyze whether the predictions for option prices obtained using the NGARCH model differ from those observed on the market. Also it would be interesting to measure to what extent it results in the improvement of the estimating procedure for prediction purposes. Since in the non-linear asymmetric GARCH framework it is possible to model the correlation of the conditional variance and lag returns, the model can be especially useful even for modelling stocks, and thus options on stocks, because of the well known leverage effect phenomenon (e. g. Christie, 1982). This indicates the possibility of applying the NGARCH model for option pricing on every asset of interest on the Croatian stock exchange. Furthermore, once the data from the options market are accessible, it will be interesting to explore how the predicted option prices can be improved using models with jump processes since such models can explain unpredictable market movements. Finally, once the options market is founded, there will be a variety of models trying to model different phenomena.

Appendix 1

Proof for the local risk neutrality for the GARCH model

Let us suppose that process specifications under the new measure are given by (15) and (16). Supposing rational investor behaviour (in the sense that prefers more money to less), the foreign currency is a special case of the foreign asset given by the relation

$$X_t = C_t \exp(r_s t). \quad (25)$$

Since from the model assumptions σ_{t+1}^2 is the value known at time t , and Z^* is a standard normal random variable, from specification (15) it follows that $\ln(X_{t+1}/X_t) = r_s + \ln(C_{t+1}/C_t)$ is conditionally normal distributed random variable thus X_{t+1}/X_t is conditionally log-normal.

Furthermore, calculating the conditional expectation of the relation (15) it follows that

$$\begin{aligned} E_t^*[X_{t+1} / X_t] &= E_t^*[C_{t+1} \exp(r_s(t+1)) / (C_t \exp(r_s t))] \\ &= \exp(r_s) E_t^*[\exp(r_d - r_s - \frac{1}{2} \sigma_{t+1}^2 + \sigma_{t+1} Z_{t+1}^*)] \\ &= \exp(r_d - \frac{1}{2} \sigma_{t+1}^2) E_t^*[\exp(\sigma_{t+1} Z_{t+1}^*)]. \end{aligned} \quad (26)$$

But, $\sigma_{t+1}^2 Z_{t+1}^*$ is conditionally normal distributed random variable of zero expectation and variance σ_{t+1}^2 , so the expression $E_t^*[\exp(\sigma_{t+1} Z_{t+1}^*)]$ represents the moment generating function for the normal random variable calculated in 1 which is exactly $\exp(\frac{1}{2} \sigma_{t+1}^2)$. So, from equality (26) it follows that

$$E_t^* [X_{t+1} / X_t] = \exp(r_d), \quad (27)$$

which fulfils the second condition.

Finally, taking into consideration that the variance of tomorrow's return, σ_{t+1}^2 , which is known at the end of today's day, i. e. in time t , and using relations (15) and (16) it follows that

$$\begin{aligned} \text{Var}_t^* [\ln(X_{t+1} / X_t)] &= E_t^* [\omega + \alpha(\sigma_t Z_t^* - \lambda\sigma_t - \rho\sigma_t)^2 + \beta\sigma_t^2] \\ &= E_t^* [\omega + \alpha(P_t - r_d + r_s + \frac{1}{2}\sigma_t^2 - \lambda\sigma_t - \rho\sigma_t)^2 + \beta\sigma_t^2] \\ &= E_t [\omega + \alpha(\sigma_t Z_t - \rho\sigma_t)^2 + \beta\sigma_t^2] \\ &= \text{Var}_t [\ln(X_{t+1} / X_t)], \end{aligned} \quad (28)$$

where the last equality follows from relation (3). This fulfils the third condition.

Appendix 2

Foreign currency option prices under the GARCH model and the Black-Scholes model for different maturities and strikes with volatility $\sigma_0 = 0.036128$ currency spot price $C = 7.335$ and parameters from Table 1²²

Days to maturity τ	m	Strike	co^{GH}	co^{BS}
30	0.97	7.12	0.22308690122941	0.22288483536895
	1.00	7.22	0.11741870260476	0.11766584897751
	1.00	7.34	0.03721059194186	0.03812329096746
	1.02	7.45	0.00624557117497	0.00563250518454
	1.03	7.56	0.00080196491670	0.00030742586018
60	0.97	7.12	0.22756291771718	0.22720379640897
	1.00	7.22	0.12794447130915	0.12867323461609
	1.00	7.34	0.05362443201262	0.05477715335458
	1.02	7.45	0.01624776243006	0.01581733927356
	1.03	7.56	0.00392655032981	0.00287215360745
90	0.97	7.12	0.23210542851231	0.23267520409955
	1.00	7.22	0.13733693276207	0.13898858113885
	1.00	7.34	0.06638849391364	0.06784528392649
	1.02	7.45	0.02594685840856	0.02563837476337
	1.03	7.56	0.00857959351282	0.00720066693796

²² In the table are presented results which have the estimated unconditional value σ_0^2 as the initial value of the GARCH variance. Nevertheless, as initial values for the NGARCH variance, the values equal to the unconditional one increased by 15% and 20% respectively are also considered. The graphs of the Black-Scholes implied volatilities with respect to the changed initial values of the variance are omitted since they give not rise to new conclusions that are not the logical ones, e. g. the constant volatility model underprice the options if the initial value of the GARCH variance is higher.

Appendix 3

Foreign currency option prices under the GARCH and the Black-Scholes model for different maturities and strikes with volatility $\sigma_0 = 0.036128$, currency spot price $C = 7.335$ and $\rho = -0.461$

Days to maturity τ	m	Strike	c_{GH}	c_{BS}
30	0.97	7.12	0.22329682837468	0.22288483525001
	1.00	7.22	0.11663240889513	0.11766584778476
	1.00	7.34	0.03681841852915	0.03812328834133
	1.02	7.45	0.00738232725972	0.00563250382702
	1.03	7.56	0.00140489838562	0.00030742568463
60	0.97	7.12	0.22711154073123	0.22720379566394
	1.00	7.22	0.12581817310359	0.12867323219096
	1.00	7.34	0.05224104494910	0.05477714965717
	1.02	7.45	0.01770478546070	0.01581733654321
	1.03	7.56	0.00587851654204	0.00287215259940
90	0.97	7.12	0.23137836550745	0.23267520260550
	1.00	7.22	0.13472117261886	0.13898857779652
	1.00	7.34	0.06468078932117	0.06784527941821
	1.02	7.45	0.02735599257122	0.02563837101399
	1.03	7.56	0.01151890310116	0.00720066497420

Appendix 4

Foreign currency option prices under the GARCH and the Black-Scholes model for different maturities and strikes with volatility $\sigma_0 = 0.036128$, currency spot price $C = 7.335$ and $\rho = 0$.

Days to maturity τ	m	Strike	c_{GH}	c_{BS}
30	0.97	7.12	0.22325783421416	0.22288483525001
	1.00	7.22	0.11790141075531	0.11766584778476
	1.00	7.34	0.03725499961867	0.03812328834133
	1.02	7.45	0.00575698868902	0.00563250382702
	1.03	7.56	0.00054626948211	0.00030742568463
60	0.97	7.12	0.22811189670277	0.22720379566394
	1.00	7.22	0.12880164404218	0.12867323219096
	1.00	7.34	0.05376611877815	0.05477714965717
	1.02	7.45	0.01565558552787	0.01581733654321
	1.03	7.56	0.00345550578795	0.00287215259940
90	0.97	7.12	0.23329522296873	0.23267520260550
	1.00	7.22	0.13885967166104	0.13898857779652
	1.00	7.34	0.06687554640161	0.06784527941821
	1.02	7.45	0.02507359991214	0.02563837101399
	1.03	7.56	0.00752584342935	0.00720066497420

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