On cross-currency models with stochastic volatility and correlated interest rates

Lech Grzelak and Kees Oosterlee

Delft University of Technology, CWI

1. June 2010

Online at https://mpra.ub.uni-muenchen.de/23020/
MPRA Paper No. 23020, posted 4. June 2010 10:22 UTC
ON CROSS-CURRENCY MODELS WITH
STOCHASTIC VOLATILITY AND CORRELATED
INTEREST RATES

Lech A. Grzelak,* Cornelis W. Oosterlee,†

June 1, 2010

Abstract

We construct multi-currency models with stochastic volatility and correlated
stochastic interest rates with a full matrix of correlations. We first deal with a
foreign exchange (FX) model of Heston-type, in which the domestic and foreign
interest rates are generated by the short-rate process of Hull-White [HW96]. We
then extend the framework by modeling the interest rate by a stochastic volatility
displaced-diffusion Libor Market Model [AA02], which can model an interest
rate smile. We provide semi-closed form approximations which lead to efficient
calibration of the multi-currency models. Finally, we add a correlated stock to the
framework and discuss the construction, model calibration and pricing of equity-
FX-interest rate hybrid payoffs.

Key words: Foreign-exchange (FX); stochastic volatility; Heston model; stochastic interest rates;
interest rate smile; forward characteristic function; hybrids; affine diffusion; efficient calibration.

1 Introduction

Since the financial crisis, investors tend to look for products with a long time horizon,
that are less sensitive to short-term market fluctuations. When pricing these exotic
contracts it is desirable to incorporate in a mathematical model the patterns present in
the market that are relevant to the product.

Due to the existence of complex FX products, like the Power-Reverse Dual-
Currency [SO02], the Equity-CMS Chameleon or the Equity-Linked Range Accrual
TRAN swaps [Cap07], that all have a long lifetime and are sensitive to smiles or skews
in the market, improved models with stochastic interest rates need to be developed.

The literature on modeling foreign exchange (FX) rates is rich and many stochastic
models are available. An industrial standard is a model from [SO02], where log-normally
distributed FX dynamics are assumed and Gaussian, one-factor, interest rates are used.
This model gives analytic expressions for the prices of basic products for at-the-money
options. Extensions on the interest rate side were presented in [Sch02b; Mik01], where
the short-rate model was replaced by a Libor Market Model framework.

A Gaussian interest rate model was also used in [Pit06], in which a local volatility
model was applied for generating the skews present in the FX market. In another
paper, [KJ07], a displaced-diffusion model for FX was combined with the interest rate
Libor Market Model.

Stochastic volatility FX models have also been investigated. For example, in [HP09]
the Schöbel-Zhu model was applied for pricing FX in combination with short-rate
processes. This model leads to a semi-closed form for the characteristic function.
However, for a normally distributed volatility process it is difficult to outperform the
Heston model with independent stochastic interest rates [HP09].

---

*Delft University of Technology, Delft Institute of Applied Mathematics, Delft, The Netherlands,
and Rabobank International, Utrecht. E-mail address: L.A.Grzelak@tudelft.nl.
†CWI – Centrum Wiskunde & Informatica, Amsterdam, the Netherlands, and Delft University of
Technology, Delft Institute of Applied Mathematics. E-mail address: C.W.Oosterlee@cwi.nl.
Research on the Heston dynamics in combination with correlated interest rates has led to some interesting models. In [And07] and [Gie04] an indirectly imposed correlation structure between Gaussian short-rates and FX was presented. The model is intuitively appealing, but it may give rise to very large model parameters [AAA08]. An alternative model was presented in [AM06; AAA08], in which calibration formulas were developed by means of Markov projection techniques.

In this article we present an FX Heston-type model in which the interest rates are stochastic processes, correlated with the governing FX processes. We first discuss the Heston FX model with Gaussian interest rate (Hull-White model [HW96]) short-rate processes. In this model a full matrix of correlations is used.

This model, denoted by FX-HHW here, is a generalization of our work in [GO09], where we dealt with the problem of finding an affine approximation of the Heston equity model with a correlated stochastic interest rate. In this paper, we apply this technique in the world of foreign exchange.

Secondly, we extend the framework by modeling the interest rates by a market model, i.e., by the stochastic volatility displaced-diffusion Libor Market Model [AA02; Pit05]. In this hybrid model, called FX-HLMM here, we incorporate a non-zero correlation between the FX and the interest rates and between the rates from different currencies. Because it is not possible to obtain closed-form formulas for the associated characteristic function, we use a linearization approximation, developed earlier, in [GO10].

For both models we provide details on how to include a foreign stock in the multi-currency pricing framework.

Fast model evaluation is highly desirable for FX options in practice, especially during the calibration of the hybrid model. This is the main motivation for the generalization of the linearization techniques in [GO09; GO10] to the world of foreign exchange. We will see that the resulting approximations can be used very well in the FX context.

The present article is organized as follows. In Section 2 we discuss the extension of the Heston model by stochastic interest rates, described by short-rate processes. We start the analysis with the specification of the underlying interest rate processes, rates

\[
dr_d(t) = \lambda_d (\theta_d(t) - r_d(t))dt + \eta_d W_{d}^{Q}(t),
\]

\[
dr_f(t) = \lambda_f (\theta_f(t) - r_f(t))dt + \eta_f W_{f}^{Q}(t),
\]

where \(W_{d}^{Q}(t)\) and \(W_{f}^{Q}(t)\) are Brownian motions under \(Q\) and \(Z\), respectively. Parameters \(\lambda_d, \lambda_f\) determine the speed of mean reversion to the time-dependent term structure functions \(\theta_d(t), \theta_f(t)\), and parameters \(\eta_d, \eta_f\) are the volatility coefficients.

These processes, under the appropriate measures, are linear in their state variables, so that for a given maturity \(T\) (\(0 < t < T\)) the zero-coupon bonds (ZCB) are known to be of the following form:

\[
P_d(t,T) = \exp (A_d(t,T) + B_d(t,T) r_d(t)),
\]

\[
P_f(t,T) = \exp (A_f(t,T) + B_f(t,T) r_f(t)),
\]

Secondly, we extend the framework by modeling the interest rates by a market model, i.e., by the stochastic volatility displaced-diffusion Libor Market Model [AA02; Pit05]. In this hybrid model, called FX-HLMM here, we incorporate a non-zero correlation between the FX and the interest rates and between the rates from different currencies. Because it is not possible to obtain closed-form formulas for the associated characteristic function, we use a linearization approximation, developed earlier, in [GO10].

For both models we provide details on how to include a foreign stock in the multi-currency pricing framework.

Fast model evaluation is highly desirable for FX options in practice, especially during the calibration of the hybrid model. This is the main motivation for the generalization of the linearization techniques in [GO09; GO10] to the world of foreign exchange. We will see that the resulting approximations can be used very well in the FX context.

The present article is organized as follows. In Section 2 we discuss the extension of the Heston model by stochastic interest rates, described by short-rate processes. We first discuss the Heston FX model with Gaussian interest rate (Hull-White model [HW96] one-factor model:

\[
 dr_d(t) = \lambda_d (\theta_d(t) - r_d(t))dt + \eta_d W_{d}^{Q}(t), \quad (2.1)
\]

\[
 dr_f(t) = \lambda_f (\theta_f(t) - r_f(t))dt + \eta_f W_{f}^{Q}(t), \quad (2.2)
\]

where \(W_{d}^{Q}(t)\) and \(W_{f}^{Q}(t)\) are Brownian motions under \(Q\) and \(Z\), respectively. Parameters \(\lambda_d, \lambda_f\) determine the speed of mean reversion to the time-dependent term structure functions \(\theta_d(t), \theta_f(t)\), and parameters \(\eta_d, \eta_f\) are the volatility coefficients.

These processes, under the appropriate measures, are linear in their state variables, so that for a given maturity \(T\) (\(0 < t < T\)) the zero-coupon bonds (ZCB) are known to be of the following form:

\[
P_d(t,T) = \exp (A_d(t,T) + B_d(t,T) r_d(t)),
\]

\[
P_f(t,T) = \exp (A_f(t,T) + B_f(t,T) r_f(t)), \quad (2.3)
\]
with \( A_d(t, T), A_f(t, T) \) and \( B_d(t, T), B_f(t, T) \) analytically known quantities (see for example [BM07]). In the model the money market accounts are given by:

\[
dM_d(t) = r_d(t)M_d(t)dt, \quad dM_f(t) = r_f(t)M_f(t)dt. \tag{2.4}
\]

By using the Heath-Jarrow-Morton arbitrage-free argument, [HJM92], the dynamics for the ZCBs, under their own measures generated by the money savings accounts, are known and given by the following result:

**Result 2.1** (ZCB dynamics under the risk-free measure). The risk-free dynamics of the zero-coupon bonds, \( P_d(t, T) \) and \( P_f(t, T) \), with maturity \( T \) are given by:

\[
\frac{dP_d(t, T)}{P_d(t, T)} = r_d(t)dt - \left( \int_t^T \Gamma_d(t, s)ds \right) dW_d^Q(t), \tag{2.5}
\]

\[
\frac{dP_f(t, T)}{P_f(t, T)} = r_f(t)dt - \left( \int_t^T \Gamma_f(t, s)ds \right) dW_f^Q(t), \tag{2.6}
\]

where \( \Gamma_d(t, T), \Gamma_f(t, T) \) are the volatility functions of the instantaneous forward rates \( f_d(t, T), f_f(t, T) \), respectively, that are given by:

\[
df_d(t, T) = \Gamma_d(t, T) \int_t^T \Gamma_d(t, s)ds + \Gamma_d(t, T)dW_d^Q(t), \tag{2.7}
\]

\[
df_f(t, T) = \Gamma_f(t, T) \int_t^T \Gamma_f(t, s)ds + \Gamma_f(t, T)dW_f^Q(t). \tag{2.8}
\]

**Proof.** For the proof see [MR97]. \( \square \)

The spot-rates at time \( t \) are defined by \( r_d(t) \equiv f_d(t, t), \quad r_f(t) \equiv f_f(t, t) \).

By means of the volatility structures, \( \Gamma_d(t, T), \Gamma_f(t, T) \), one can define a number of short-rate processes. In our framework the volatility functions are chosen to be \( \Gamma_d(t, T) = \eta_d \exp \left( -\lambda_d(T-t) \right) \) and \( \Gamma_f(t, T) = \eta_f \exp \left( -\lambda_f(T-t) \right) \). The Hull-White short-rate processes, \( r_d(t) \) and \( r_f(t) \) as in (2.1), (2.2), are then obtained and the term structures, \( \theta_d(t), \theta_f(t) \), expressed in terms of instantaneous forward rates, are also known. The choice of specific volatility determines the dynamics of the ZCBs:

\[
\frac{dP_d(t, T)}{P_d(t, T)} = r_d(t)dt + \eta_dB_d(t, T)dW_d^Q(t), \tag{2.9}
\]

\[
\frac{dP_f(t, T)}{P_f(t, T)} = r_f(t)dt + \eta_fB_f(t, T)dW_f^Q(t),
\]

with \( B_d(t, T) \) and \( B_f(t, T) \) as in (2.3), given by:

\[
B_d(t, T) = \frac{1}{\lambda_d} \left( e^{-\lambda_d(T-t)} - 1 \right), \quad B_f(t, T) = \frac{1}{\lambda_f} \left( e^{-\lambda_f(T-t)} - 1 \right). \tag{2.10}
\]

For a detailed discussion on short-rate processes, we refer to the analysis of Musiela and Rutkowski in [MR97]. In the next subsection we define the FX hybrid model.

### 2.1 The Model with Correlated, Gaussian Interest Rates

The FX-HHW model, with all processes defined under the domestic risk-neutral measure, \( \mathbb{Q} \), is of the following form:

\[
d\xi(t)/\xi(t) = \left( r_d(t) - r_f(t) \right) dt + \sqrt{\sigma(t)}dW^Q_\xi(t), \quad \xi(0) > 0,
\]

\[
d\sigma(t) = \kappa(\bar{\sigma} - \sigma(t))dt + \gamma\sqrt{\sigma(t)}dW^Q_\sigma(t), \quad \sigma(0) > 0,
\]

\[
dr_d(t) = \lambda_d(\theta_d(t) - r_d(t))dt + \eta_d dW^Q_d(t), \quad r_d(0) > 0,
\]

\[
dr_f(t) = \left( \lambda_f(\theta_f(t) - r_f(t)) - \eta_f \rho_{\xi,\sigma} \sqrt{\sigma(t)} \right) dt + \eta_f dW^Q_f(t), \quad r_f(0) > 0.
\]

\[
(2.11)
\]
Now, we consider a forward price, $\Pi(t)$, with price basic options on the FX rate, $V(t)$ under the domestic risk-neutral measure, $Q$. The drift in the short-rate process, $r_f(t)$, gives rise to an additional term, $-\eta_f \rho_{\xi,f} \sqrt{\sigma(t)}$. This term ensures the existence of martingales, under the domestic spot measure, for the following prices (for more discussion, see [Shr04]):

$$\chi_1(t) := \xi(t) \frac{M_f(t)}{M_d(t)} \text{ and } \chi_2(t) := \xi(t) \frac{P_f(t,T)}{M_d(t)},$$

where $P_f(t,T)$ is the price foreign zero-coupon bond (2.9), respectively, and the money savings accounts $M_d(t)$ and $M_f(t)$ are from (2.4). To see that the processes $\chi_1(t)$ and $\chi_2(t)$ are martingales, one can apply the Itô product rule, which gives:

$$\frac{d\chi_1(t)}{\chi_1(t)} = \sqrt{\sigma(t)} dW_\xi^Q(t),$$

$$\frac{d\chi_2(t)}{\chi_2(t)} = \sqrt{\sigma(t)} dW_\xi^Q(t) + \eta_f B_f(t,T) dW_f^Q(t).$$

The change of dynamics of the underlying processes, from the foreign-spot to the domestic-spot measure, also influences the dynamics for the associated bonds, which, under the domestic risk-neutral measure, $Q$, with the money savings account considered as a numéraire, have the following representations

$$\frac{dP_d(t,T)}{P_d(t,T)} = r_d(t) dt + \eta_d B_d(t,T) dW_d^Q(t),$$

$$\frac{dP_f(t,T)}{P_f(t,T)} = \left(r_f(t) - \rho_{\xi,f} \eta_f B_f(t,T) \sqrt{\sigma(t)}\right) dt + \eta_f B_f(t,T) dW_f^Q(t),$$

with $B_d(t,T)$ and $B_f(t,T)$ as in (2.10).

### 2.2 Pricing of FX Options

In order to perform efficient calibration of the model we need to be able to price basic options on the FX rate, $V(t,X(t))$, for a given state vector, $X(t) = [\xi(t), \sigma(t), r_d(t), r_f(t)]^T$:

$$V(t,X(t)) = \mathbb{E}^Q \left( \frac{M_d(t)}{M_d(T)} \max(\xi(T) - K, 0) | \mathcal{F}_t \right),$$

with $M_d(t) = \exp \left( \int_0^t r_d(s) ds \right)$.

Now, we consider a forward price, $\Pi(t)$, such that:

$$\mathbb{E}^Q \left( \frac{\max(\xi(T) - K, 0)}{M_d(T)} | \mathcal{F}_t \right) = \frac{V(t,X(t))}{M_d(t)} =: \Pi(t).$$
By Itô’s lemma we have:

$$d\Pi(t) = \frac{1}{M_d(t)} dV(t) - r_d(t) \frac{V(t)}{M_d(t)} dt,$$

with $V(t) := V(t, X(t))$. We know that $\Pi(t)$ must be a martingale, i.e.: $E(d\Pi(t)) = 0$. Including this in (2.17) gives the following *Fokker-Planck forward equation* for $V$:

$$r_dV = \frac{1}{2}\eta^2 \frac{\partial^2 V}{\partial r_f^2} + \rho_{d,f}\eta \frac{\partial^2 V}{\partial r_d \partial r_f} + \frac{1}{2}\rho^2 \frac{\partial^2 V}{\partial r_d^2} + \rho_{d,f} \rho \gamma \frac{\partial^2 V}{\partial \sigma \partial r_f}$$

$$+ \rho_{d,f} \rho \gamma \frac{\partial^2 V}{\partial r_d \partial \sigma} + \frac{1}{2} \gamma^2 \frac{\partial^2 V}{\partial \sigma^2} + \rho_{d,f} \rho \xi \sqrt{\sigma} \frac{\partial^2 V}{\partial \xi \partial r_f} + \rho_{d,a} \rho \xi \sqrt{\sigma} \frac{\partial^2 V}{\partial \xi \partial r_d}$$

$$+ \rho_{d,a} \rho \xi \sqrt{\sigma} \frac{\partial^2 V}{\partial r_d \partial \xi} + \frac{1}{2} \xi^2 \sigma \frac{\partial^2 V}{\partial \xi^2} + (\lambda_f(t) - \gamma) \frac{\partial V}{\partial r_f}$$

$$+ \lambda_d(t) - \gamma (r_d - r_f) \frac{\partial V}{\partial \sigma} + \sigma (\sigma - \gamma) \frac{\partial V}{\partial \sigma} + (r_d - r_f) \xi \frac{\partial V}{\partial \xi} + \frac{\partial V}{\partial t}. $$

This 4D PDE contains non-affine terms, like square-roots and products. It is therefore difficult to solve it analytically and a numerical PDE discretization, like finite differences, needs to be employed. Finding a numerical solution for this PDE is therefore rather expensive and not easily applicable for model calibration. In the next subsection we propose an *approximation* of the model, which is useful for calibration.

### 2.2.1 The FX Model under the Forward Domestic Measure

To reduce the complexity of the pricing problem, we move from the spot measure, generated by the money savings account in the domestic market, $M_d(t)$, to the *forward FX measure* where the *numéraire* is the domestic zero-coupon bond, $P_d(t, T)$. As indicated in [MR97; Pit06], the forward is given by:

$$FX^T(t) = \xi(t) \frac{P_f(t, T)}{P_d(t, T)},$$

where $FX^T(t)$ represents the forward exchange rate under the $T$-forward measure, and $\xi(t)$ stands for foreign exchange rate under the domestic spot measure. The superscript should not be confused with the transpose notation used at other places in the text.

By switching from the domestic risk-neutral measure, $Q$, to the domestic $T$-forward measure, $Q^T$, the discounting will be decoupled from taking the expectation, i.e.: 

$$\Pi(t) = P_d(t, T)E^T \left( \max \left( FX^T(T) - K, 0 \right) | F_t \right).$$

In order to derive the dynamics for $FX^T(t)$ in (2.18), we apply Itô’s formula:

$$dFX^T(t) = \frac{P_f(t, T)}{P_d(t, T)} d\xi(t) + \frac{\xi(t)}{P_d(t, T)} dP_f(t, T) - \xi(t) \frac{P_f(t, T)}{P_d^2(t, T)} dP_d(t, T)$$

$$+ \xi(t) \frac{P_f(t, T)}{P_d(t, T)} (dP_d(t, T))^2 + \frac{1}{P_d(t, T)} (d\xi(t) dP_f(t, T))$$

$$- \frac{P_f(t, T)}{P_d^2(t, T)} (dP_d(t, T) d\xi(t)) - \frac{\xi(t)}{P_d^2(t, T)} dP_d(t, T) dP_f(t, T).$$

After substitution of SDEs (2.11), (2.15) and (2.16) into (2.20), we arrive at the following FX forward dynamics:

$$dFX^T(t) = \eta_d B_d(t, T) \left( \eta_d B_d(t, T) - \rho_{d,a} \sqrt{\sigma(t)} - \rho_{d,d} \eta_f B_f(t, T) \right) dt$$

$$+ \sqrt{\sigma(t)} dW^G_d(t) - \eta_d B_d(t, T) dW^G_d(t) + \eta_f B_f(t, T) dW^G_f(t).$$

(2.21)
Since $FX^T(t)$ is a martingale under the $T$-forward domestic measure, i.e.,
$P_d(t,T)E^T(FX^T(T)|F_t) = P_d(t,T)FX^T(t) =: P_f(t,T)\xi(t)$, the appropriate Brownian motions under the $T$-forward domestic measure, $dW^T_\xi(t)$, $dW^T_\sigma(t)$, $dW^T_\eta(t)$ and $dW^T_f(t)$, need to be determined.

A change of measure from domestic-spot to domestic $T$-forward measure requires a change of numéraire from money savings account, $M_d(t)$, to zero-coupon bond $P_d(t,T)$. In the model we incorporate a full matrix of correlations, which implies that all processes will change their dynamics by changing the measure from spot to forward. Lemma 2.2 provides the model dynamics under the domestic $T$-forward measure, $Q^T$.

**Lemma 2.2** (The FX-HHW model dynamics under the $Q^T$ measure). Under the $T$-forward domestic measure, the model in (2.11) is governed by the following dynamics:

$$\frac{dFX^T(t)}{FX^T(t)} = \sqrt{\sigma(t)}dW^T_\xi(t) - \eta_d B_d(t,T) dW^T_\sigma(t) + \eta_f B_f(t,T) dW^T_f(t),$$  \hspace{1cm} (2.22)

where

$$d\sigma(t) = \left( \kappa(\bar{\sigma} - \sigma(t)) + \gamma \rho_{\sigma,d}\eta_d B_d(t,T) \sqrt{\sigma(t)} \right) dt + \gamma \sqrt{\sigma(t)} dW^T_\sigma(t),$$  \hspace{1cm} (2.23)

$$dr_d(t) = \left( \lambda_d(\theta_d(t) - r_d(t)) + \nu_d^2 B_d(t,T) \right) dt + \eta_d dW^T_d(t),$$  \hspace{1cm} (2.24)

$$dr_f(t) = \left( \lambda_f(\theta_f(t) - r_f(t)) - \eta_f \rho_{\xi,f} \sqrt{\sigma(t)} + \eta_d \eta_f \rho_{d,f} B_d(t,T) \right) dt + \eta_f dW^T_f(t).$$  \hspace{1cm} (2.25)

with a full matrix of correlations given in (2.12), and with $B_d(t,T), B_f(t,T)$ given by (2.10).

The proof can be found in Appendix A.

From the system in Lemma 2.2 we see that after moving from the domestic-spot $Q$-measure to the domestic $T$-forward $Q^T$ measure, the forward exchange rate $FX^T(t)$ does not depend explicitly on the short-rate processes $r_d(t)$ or $r_f(t)$. It does not contain a drift term and only depends on $dW^T_d(t)$, $dW^T_f(t)$, see (2.22).

**Remark.** Since the sum of three correlated, normally distributed random variables, $Q = X + Y + Z$, remains normal with the mean equal to the sum of the individual means and the variance equal to

$$\sigma_Q^2 = \sigma_X^2 + \sigma_Y^2 + \sigma_Z^2 + 2\rho_{X,Y}\sigma_X\sigma_Y + 2\rho_{X,Z}\sigma_X\sigma_Z + 2\rho_{Y,Z}\sigma_Y\sigma_Z,$$

we can represent the forward (2.22) as:

$$dFX^T/\sqrt{FX^T} = \left( \sigma + \eta_d^2 B_d^2 + \eta_f^2 B_f^2 - 2\rho_{\xi,d}\eta_d B_d \sqrt{\sigma} \right.$$

$$\left. + 2\rho_{\xi,f}\eta_f B_f \sqrt{\sigma} - 2\rho_{d,f}\eta_d \eta_f B_d B_f \right) \frac{1}{2} dW^T_P.$$

Although the representation in (2.26) reduces the number of Brownian motions in the dynamics for the FX, one still needs to find the appropriate cross-terms, like $dW^T_\xi(t)dW^T_\sigma(t)$, in order to obtain the covariance terms. For clarity we therefore prefer to stay with the standard notation.

**Remark.** The dynamics of the forwards, $FX^T(t)$ in (2.22) or in (2.26), do not depend explicitly on the interest rate processes, $r_d(t)$ and $r_f(t)$, and are completely described by the appropriate diffusion coefficients. This suggests that the short-rate variables will not enter the pricing PDE. Note, that this is only the case for models in which the diffusion coefficient for the interest rate does not depend on the level of the interest rate.

In next section we derive the corresponding pricing PDE and provide model approximations.
2.3 Approximations and the Forward Characteristic Function

As the dynamics of the forward foreign exchange, FX\(^T(t)\), under the domestic forward measure involve only the interest rate diffusions \(dW_d^T(t)\) and \(dW_f^T(t)\), a significant reduction of the pricing problem is achieved.

In order to find the forward ChF we take, as usual, the log-transform of the forward rate FX\(^T(t)\), i.e.: \(x^T(t) := \log FX^T(t)\), for which we obtain the following dynamics:

\[
dx^T(t) = \left(\frac{\partial}{\partial t} + \frac{1}{2}\sigma(t)^2\right)dt + \sigma(t)dB_t^T(t),
\]

where we used the notation \(B_d := B_d(t, T)\) and \(B_f := B_f(t, T)\), and \(\sqrt{\sigma(t)}\).

By applying the Feynman-Kac theorem we can obtain the characteristic function of the forward FX rate dynamics. The forward characteristic function:

\[
\phi^T(u, x(t), t, T) = E^T(e^{u\cdot x^T(T)}|\mathcal{F}_t),
\]

with final condition, \(\phi^T(u, x(T), t, T) = e^{iux^T(T)}\), is the solution of the following Kolmogorov backward partial differential equation:

\[
-\frac{\partial \phi^T}{\partial t} = \left(\frac{\partial^2 \phi^T}{\partial x^2} - \frac{\partial \phi^T}{\partial x}\right) + \left(\frac{\partial \phi^T}{\partial \sigma} \right) \left(\frac{\partial^2 \phi^T}{\partial x^2} - \frac{\partial \phi^T}{\partial x}\right) + \left(\frac{\gamma^2}{\sigma^2}\right) \left(\frac{\partial^2 \phi^T}{\partial x^2}\right).
\]

This PDE contains however non-affine \(\sqrt{\sigma}\)-terms so that it is nontrivial to find the solution. Recently, in [GO09], we have proposed two methods for linearization of these non-affine\(^1\) square-roots of the square root process [CIR85]. The first method is to project the non-affine square-root terms on their first moments. This is also the approach followed here.

The approximation of the non-affine terms in the corresponding PDE is then done as follows. We assume:

\[
\sqrt{\sigma(t)} \approx \mathbb{E} \left(\sqrt{\sigma(t)}\right) =: \varphi(t),
\]

with the expectation of the square root of \(\sigma(t)\) given by:

\[
\mathbb{E} \left(\sqrt{\sigma(t)}\right) = \sqrt{2c(t)} e^{-\frac{c(t)}{2}} \sum_{k=0}^{\infty} \frac{1}{k!} \left(\frac{\sigma(t)}{2}\right)^{2k} e^{-\sigma(t)},
\]

and

\[
c(t) = \frac{1}{4\kappa}\gamma^2 (1 - e^{-\kappa t}), \quad \ell = \frac{4\kappa\bar{\sigma}}{\gamma^2}, \quad c(t) = \frac{4\kappa\bar{\sigma}(0)e^{-\kappa t}}{\gamma^2(1 - e^{-\kappa t})}.
\]

\(^1\)According to [DPS00] the \(n\)-dimensional system of SDEs:

\[
dX(t) = \mu(X(t))dt + \sigma(X(t))dW(t),
\]

is of the affine form if:

\[
\mu(X(t)) = a_0 + a_1X(t), \quad \text{for any } (a_0, a_1) \in \mathbb{R}^n \times \mathbb{R}^{n \times n},
\]

\[
\sigma(X(t)) = (c_0)_{ij} + (c_1)_{ij}X(t), \quad \text{for arbitrary } (c_0, c_1) \in \mathbb{R}^{n \times n} \times \mathbb{R}^{n \times n \times n},
\]

\[
r(X(t)) = r_0 + r_1X(t), \quad \text{for } (r_0, r_1) \in \mathbb{R}^n,
\]

for \(i, j = 1, \ldots, n\), with \(r(X(t))\) being an interest rate component.
$\Gamma(k)$ is the gamma function defined by:

$$\Gamma(k) = \int_0^\infty t^{k-1}e^{-t}dt.$$  

Although the expectation in (2.30) is a closed form expression, its evaluation is rather expensive. One may prefer to use a proxy, for example,

$$\mathbb{E}(\sqrt{\sigma(t)}) \approx \beta_1 + \beta_2 e^{-\beta_3 t},$$  \hspace{1cm} (2.32)

in which the constant coefficients $\beta_1$, $\beta_2$ and $\beta_3$ can be determined by asymptotic equality with (2.30) (see [GO09] for details).

Projection of the non-affine terms on their first moments allows us to derive the corresponding forward characteristic function, $\phi^T$, which is then of the following form:

$$\phi^T(u, X(t), t, T) = \exp(A(u, \tau) + B(u, \tau)\mu^T(t) + C(u, \tau)\sigma(t)),$$

where $\tau = T-t$, and the functions $A(\tau) := A(u, \tau)$, $B(\tau) := B(u, \tau)$ and $C(\tau) := C(u, \tau)$ are given by:

- $B'(\tau) = 0$,
- $C'(\tau) = -\kappa C(\tau) + (B^2(\tau) - B(\tau))/2 + \rho_{\sigma, \epsilon}\gamma B(\tau)C(\tau) + \gamma^2 C^2(\tau)/2$,
- $A'(\tau) = \kappa \sigma C(\tau) + \rho_{\sigma, \epsilon} \gamma d \varphi(\tau)B_d(\tau)C(\tau) - \zeta(\tau, \varphi(\tau)) (B^2(\tau) - B(\tau))$
  $$+ (-\rho_{\sigma, \eta} \mu \varphi(\tau)B_d(\tau) + \rho_{\sigma, f} \gamma f \varphi(\tau)B_f(\tau)) B(\tau) C(\tau),$$

with $\varphi(t) = \mathbb{E}(\sqrt{\sigma(t)})$, and $B(\tau) = \lambda^{-1}(e^{-\lambda \tau} - 1)$ for $i = \{d, f\}$. The initial conditions are: $B(0) = iu$, $C(0) = 0$ and $A(0) = 0$.

With $B(\tau) = iu$, the complex-valued function $C(\tau)$ is of the Heston-type, [Hes93], and its solution reads:

$$C(\tau) = \frac{1 - e^{-d\tau}}{\gamma^2 (1 - g e^{-d\tau})} (\kappa - \rho_{\sigma, \epsilon} \gamma i u - d),$$  \hspace{1cm} (2.33)

with $d = \sqrt{(\rho_{\sigma, \epsilon} \gamma i u - \kappa)^2 - \gamma^2 i u (i u - 1)}$, $g = \frac{\kappa - \gamma \rho_{\sigma, \epsilon} i u - d}{\kappa - \gamma \rho_{\sigma, \epsilon} i u + d}$.

The parameters $\kappa, \gamma, \rho_{\sigma, \epsilon}$ are given in (2.11).

Function $A(\tau)$ is given by:

$$A(\tau) = \int_0^\tau \left( \kappa \sigma + \rho_{\sigma, d} \gamma d \varphi(s)B_d(s) - \rho_{\sigma, \epsilon} \gamma \varphi(s)B_d(s) i u \right.$$

$$\left. + \rho_{\sigma, f} \gamma f \varphi(s)B_f(s) i u \right) C(s) ds + (i u^2 + i u) \int_0^\tau \zeta(s, \varphi(s)) ds,$$  \hspace{1cm} (2.34)

with $C(s)$ in (2.33). It is most convenient to solve $A(\tau)$ numerically with, for example, Simpson’s quadrature rule. With correlations $\rho_{\sigma, d}$, $\rho_{\sigma, f}$ equal to zero, a closed-form expression for $A(\tau)$ would be available [GO09].

We denote the approximation, by means of linearization, of the full-scale FX-HHW model by FX-HHW1. It is clear that efficient pricing with Fourier-based methods can be done with FX-HHW1, and not with FX-HHW.

By the projection of $\sqrt{\sigma(t)}$ on its first moment in (2.29) the corresponding PDE is affine in its coefficients, and reads:

$$- \frac{\partial \phi^T}{\partial t} = (\kappa (\bar{\sigma} - \sigma) + \Psi_1) \frac{\partial \phi^T}{\partial \bar{\sigma}} + \left( \frac{1}{2} \sigma - \zeta(t, \varphi(t)) \right) \left( \frac{\partial^2 \phi^T}{\partial x^2} - \frac{\partial \phi^T}{\partial x} \right)$$

$$+ (\rho_{\sigma, \epsilon} \gamma \sigma - \Psi_2) \frac{\partial^2 \phi^T}{\partial \sigma \partial x} + \frac{1}{2} \sigma^2 \frac{\partial^2 \phi^T}{\partial \sigma^2},$$

with:

$$\phi^T(u, X(T), T, T) = \mathbb{E}^T \left( e^{i u x^T(T)} | F_T \right) = e^{i u x^T(T)},$$

and $\zeta(t, \varphi(t)) = \Psi_3 + \rho_{d, f} \eta d \varphi_B d(t, T) B_f(t, T) - \frac{1}{2} \left( \eta_d^2 B^2_d(t, T) + \eta_f^2 B^2_f(t, T) \right).$
The three terms, $\Psi_1$, $\Psi_2$, and $\Psi_3$, in the PDE (2.35) contain the function $\varphi(t)$:

\[
\begin{align*}
\Psi_1 & := \rho_s d \gamma d B_d(t, T) \varphi(t), \\
\Psi_2 & := (\rho_s d \gamma d B_d(t, T) - \rho_s f \gamma f B_f(t, T)) \varphi(t), \\
\Psi_3 & := (\rho_s d \gamma d B_d(t, T) - \rho_s f \gamma f B_f(t, T)) \varphi(t).
\end{align*}
\]

When solving the pricing PDE for $t \to T$, the terms $B_d(t, T)$ and $B_f(t, T)$ tend to zero, and all terms that contain the approximation vanish. The case $t \to 0$ is furthermore trivial, since $\sqrt{\sigma(t)} \to 0$.

Under the $T$-forward domestic FX measure, the projection of the non-affine terms on their first moments is expected to provide high accuracy. In Section 2.5 we perform a numerical experiment to validate this.

It is worth mentioning that also an alternative approximation for the non-affine terms $\sqrt{\sigma(t)}$ is available, see [GO09]. This alternative approach guarantees that the first two moments are exact. In this article we stay, however, with the first representation.

### 2.4 Pricing a Foreign Stock in the FX-HHW Model

Here, we focus our attention on pricing a foreign stock, $S_f(t)$, in a domestic market. With this extension we can in principle price equity-FX-interest rate hybrid products.

With an equity smile/skew present in the market, we assume that $S_f(t)$ is given by the Heston stochastic volatility model:

\[
\begin{align*}
dS_f(t)/S_f(t) &= r_f(t)dt + \sqrt{\omega(t)}dZ_f(t), \\
d\omega(t) &= \kappa_f(\bar{\omega} - \omega(t))dt + \gamma_f \sqrt{\omega(t)}dW_f(t), \\
dr_f(t) &= \lambda_f(\theta_f(t) - r_f(t))dt + \eta_f dW_f(t),
\end{align*}
\]

where $Z$ indicates the foreign-spot measure and the model parameters, $\kappa_f$, $\gamma_f$, $\lambda_f$, $\theta_f(t)$ and $\eta_f$, are as before.

Before deriving the stock dynamics in domestic currency, the model has to be calibrated in the foreign market to plain vanilla options. This can be efficiently done with the help of a fast pricing formula.

With the foreign short-rate process, $r_f(t)$, established in (2.11) we need to determine the drifts for $S_f(t)$ and its variance process, $\omega(t)$, under the domestic spot measure. The foreign stock, $S_f(t)$, can be expressed in domestic currency by multiplication with the FX, $\xi(t)$, and by discounting with the domestic money savings account, $M_d(t)$. Such a stock is a tradable asset, so the price $\xi(t)S_f(t)/M_d(t)$ (with $\xi(t)$ in (2.11), $S_f(t)$ from (2.36) and the domestic money-saving account $M_d(t)$ in (2.4)) needs to be a martingale.

By applying Itô’s lemma to $\xi(t)S_f(t)/M_d(t)$, we find

\[
\frac{d}{\xi(t)S_f(t)/M_d(t)} = \rho_x S_f \sqrt{\sigma(t)} \sqrt{\omega(t)}dt + \sqrt{\omega(t)}dW_f(t) + \sqrt{\sigma(t)}dW_Z(t),
\]

where $Q$ and $Z$ indicate the domestic-spot and foreign-spot measures, respectively. To make process $\xi(t)S_f(t)/M_d(t)$ a martingale we set:

\[
dW_Z(t) = dW_Q(t) - \rho_x S_f \sqrt{\sigma(t)}dt,
\]

where $\sigma(t)$ is the variance process of FX defined in (2.11).

Under the change of measure, from foreign to domestic-spot, $S_f(t)$ has the following dynamics:

\[
\begin{align*}
dS_f(t)/S_f(t) &= r_f(t)dt + \sqrt{\omega(t)}dW_P(t) \\
&= (r_f(t) - \rho_x S_f \sqrt{\sigma(t)} \sqrt{\omega(t)}) dt + \sqrt{\omega(t)}dW_P(t). \quad (2.38)
\end{align*}
\]
2.5 Numerical Experiment for the FX-HHW Model

In this section we check the errors resulting from the various approximations of the FX-HHW1 model. We use the set-up from [Pit06], which means that the interest rate curves are modeled by ZCBs defined by $P_d(t=0,T) = \exp(-0.02T)$ and $P_f(t=0,T) = \exp(-0.05T)$. Furthermore, $\eta_d = 0.7\%$, $\eta_f = 1.2\%$, $\lambda_d = 1\%$, $\lambda_f = 5\%$

We choose\(^2\):

$$\kappa = 0.5, \ \gamma = 0.3, \ \bar{\sigma} = 0.1, \ \sigma(0) = 0.1.$$  (2.40)

The correlation structure, defined in (2.12), is given by:

$$
\begin{pmatrix}
1 & \rho_{\xi,\sigma} & \rho_{\xi,d} & \rho_{\xi,f} \\
\rho_{\xi,\sigma} & 1 & \rho_{\xi,d} & \rho_{\xi,f} \\
\rho_{\xi,d} & \rho_{\xi,d} & 1 & \rho_{d,f} \\
\rho_{\xi,f} & \rho_{\xi,d} & \rho_{d,f} & 1
\end{pmatrix}
= 
\begin{pmatrix}
100\% & -40\% & -15\% & -15\% \\
-40\% & 100\% & 30\% & 30\% \\
-15\% & 30\% & 100\% & 25\% \\
-15\% & 30\% & 25\% & 100\%
\end{pmatrix}.  \quad (2.41)
$$

The initial spot FX rate (Dollar, $, per Euro, €) is set to 1.35. For the FX-HHW model we compute a number of FX option prices with many expiries and strikes, using two different pricing methods.

The first method is the plain Monte Carlo method, with 50.000 paths and 20$T_i$ steps, for the full-scale FX-HHW model, without any approximations.

For the second pricing method, we have used the ChF, based on the approximations in the FX-HHW1 model in Section 2.3. Efficient pricing of plain vanilla products is then done by means of the COS method [FO08], based on a Fourier cosine series expansion of the probability density function, which is recovered by the ChF with 500 Fourier cosine terms.

We also define the experiments as in [Pit06], with expiries given by $T_1, \ldots, T_{10}$, and the strikes are computed by the formula:

$$K_n(T_i) = FX^{T_i}(0) \exp\left(0.1\delta_n \sqrt{T_i}\right), \quad \text{with} \quad \delta_n = \{-1.5, -1.0, -0.5, 0, 0.5, 1.0, 1.5\},$$  (2.42)

and $FX^{T_i}(0)$ as in (2.18) with $\xi(0) = 1.35$. This formula for the strikes is convenient, since for $n = 4$, strikes $K_4(T_i)$ with $i = 1, \ldots, 10$ are equal to the forward FX rates for time $T_i$. The strikes and maturities are presented in Table B.1 in Appendix B.

The option prices resulting from both models are expressed in terms of the implied Black volatilities. The differences between the volatilities are tabulated in Table 2.1. The approximation FX-HHW1 appears to be highly accurate for the parameters considered. We report a maximum error of about 0.1% volatility for at-the-money options with a maturity of 30 years and less than 0.07% for the other options.

In the next subsection the calibration results to FX market data are presented.

\(^2\)The model parameters do not satisfy the Feller condition, $\gamma^2 > 2\kappa\bar{\sigma}$.
<table>
<thead>
<tr>
<th>$T_i$</th>
<th>$K_1(T_i)$</th>
<th>$K_2(T_i)$</th>
<th>$K_3(T_i)$</th>
<th>$K_4(T_i)$</th>
<th>$K_5(T_i)$</th>
<th>$K_6(T_i)$</th>
<th>$K_7(T_i)$</th>
</tr>
</thead>
<tbody>
<tr>
<td>6m</td>
<td>-0.0003</td>
<td>-0.0002</td>
<td>0.0000</td>
<td>0.0002</td>
<td>0.0003</td>
<td>0.0004</td>
<td>0.0005</td>
</tr>
<tr>
<td>1y</td>
<td>-0.0001</td>
<td>-0.0001</td>
<td>-0.0001</td>
<td>-0.0001</td>
<td>-0.0001</td>
<td>-0.0001</td>
<td>-0.0001</td>
</tr>
<tr>
<td>3y</td>
<td>0.0005</td>
<td>0.0004</td>
<td>0.0002</td>
<td>-0.0001</td>
<td>-0.0003</td>
<td>-0.0006</td>
<td>-0.0009</td>
</tr>
<tr>
<td>5y</td>
<td>0.0006</td>
<td>0.0004</td>
<td>0.0002</td>
<td>0.0000</td>
<td>0.0003</td>
<td>0.0007</td>
<td>-0.0010</td>
</tr>
<tr>
<td>7y</td>
<td>0.0008</td>
<td>0.0006</td>
<td>0.0004</td>
<td>0.0003</td>
<td>0.0001</td>
<td>-0.0001</td>
<td>-0.0003</td>
</tr>
<tr>
<td>10y</td>
<td>-0.0002</td>
<td>-0.0003</td>
<td>-0.0003</td>
<td>-0.0005</td>
<td>-0.0007</td>
<td>-0.0009</td>
<td>-0.0012</td>
</tr>
<tr>
<td>15y</td>
<td>-0.0012</td>
<td>-0.0010</td>
<td>-0.0009</td>
<td>-0.0009</td>
<td>-0.0009</td>
<td>-0.0009</td>
<td>-0.0010</td>
</tr>
<tr>
<td>20y</td>
<td>0.0009</td>
<td>0.0009</td>
<td>0.0009</td>
<td>0.0008</td>
<td>0.0008</td>
<td>0.0007</td>
<td>0.0006</td>
</tr>
<tr>
<td>25y</td>
<td>-0.0015</td>
<td>-0.0011</td>
<td>-0.0008</td>
<td>-0.0006</td>
<td>-0.0005</td>
<td>-0.0004</td>
<td>-0.0004</td>
</tr>
<tr>
<td>30y</td>
<td>0.0010</td>
<td>0.0011</td>
<td>0.0012</td>
<td>0.0012</td>
<td>0.0012</td>
<td>0.0012</td>
<td>0.0012</td>
</tr>
</tbody>
</table>

Table 2.1: Differences, in implied volatilities, between the FX-HHW and FX-HHW1 models. The corresponding FX option prices and the standard deviations are tabulated in Table B.4. Strike $K_4(T_i)$ is the at-the-money strike.

### 2.5.1 Calibration to Market Data

We discuss the calibration of the FX-HHW model to FX market data. In the simulation the reference market implied volatilities are taken from [Pit06] and are presented in Table B.2 in Appendix B. In the calibration routine the approximate model FX-HHW1 was applied. The correlation structure is as in (2.41). In Figure 2.1 some of the calibration results are presented.

![Figure 2.1: Comparison of implied volatilities from the market and the FX-HHW1 model for FX European call options for maturities of 1, 10 and 20 years. The strikes are provided in Table B.1 in Appendix B. $\xi(0) = 1.35$.](image)

Our experiments show that the model can be well calibrated to the market data. For long maturities and for deep-in-the-money options some discrepancy is present. This is however typical when dealing with the Heston model (not related to our approximation), since the skew/smile pattern in FX does not flatten for long maturities. This was sometimes improved by adding jumps to the model (Bates’ model). In Appendix B in Table B.3 the detailed calibration results are tabulated.

Short-rate interest rate models can typically provide a satisfactory fit to at-the-money interest rate products. They are therefore not used for pricing derivatives that are sensitive to the interest rate skew. This is a drawback of the short-rate interest rate models. In the next section an extension of the framework, so that interest rate smiles and skews can be modeled as well, is presented.

### 3 Multi-Currency Model with Interest Rate Smile

In this section we discuss a second extension of the multi-currency model, in which an interest rate smile is incorporated. This hybrid model models two types of smiles, the smile for the FX rate and the smiles in the domestic and foreign fixed income markets.
We abbreviate the model by FX-HLMM. It is especially interesting for FX products that are exposed to interest rate smiles. A description of such FX hybrid products can be found in the handbook by Hunter [Hun05].

A first attempt to model the FX by stochastic volatility and interest rates driven by a market model was proposed in [TT08], assuming independence between log-normal-Libor rates and FX. In our approach we define a model with non-zero correlation between FX and interest rate processes.

As in the previous sections, the stochastic volatility FX is of the Heston type, which under domestic risk-neutral measure, \( \mathbb{Q} \), follows the following dynamics:

\[
\begin{align*}
\frac{d\xi(t)}{\xi(t)} &= (\ldots)dt + \sqrt{\sigma(t)}dW^\xi(t), \quad S(0) > 0, \\
\frac{d\sigma(t)}{\sigma(t)} &= \kappa(\bar{\sigma} - \sigma(t))dt + \gamma\sqrt{\sigma(t)}dW^\sigma(t), \quad \sigma(0) > 0,
\end{align*}
\]

with the parameters as in (2.11). Since we consider the model under the forward measure the drift in the first SDE does not need to be specified (the dynamics of domestic-forward FX \( \xi(t)P_f(t,T)/P_d(t,T) \) do not contain a drift term).

In the model we assume that the domestic and foreign currencies are independently calibrated to interest rate products available in their own markets. For simplicity, we also assume that the tenor structure for both currencies is the same, i.e., \( T_d \equiv T_f = \{T_0, T_1, \ldots, T_N \equiv T\} \) and \( \tau_k = T_k - T_{k-1} \) for \( k = 1 \ldots N \). For \( t < T_{k-1} \) we define the forward Libor rates \( L_{d,k}(t) := L_d(t, T_{k-1}, T_k) \) and \( L_{f,k}(t) := L_f(t, T_{k-1}, T_k) \) as

\[
L_{d,k}(t) := \frac{1}{\tau_k} \left( \frac{P_{d}(t, T_{k-1})}{P_{d}(t, T_{k})} - 1 \right), \quad L_{f,k}(t) := \frac{1}{\tau_k} \left( \frac{P_{f}(t, T_{k-1})}{P_{f}(t, T_{k})} - 1 \right).
\]

(3.2)

For each currency we choose the DD-SV Libor Market Model from [AA02] for the interest rates, under the \( T \)-forward measure generated by the numéraires \( P_d(t, T) \) and \( P_f(t, T) \), given by:

\[
\begin{align*}
\frac{dL_{d,k}(t)}{\tau_k} &= \sigma_{d,k} \frac{\phi_{d,k}(t)}{\eta_d(t)} \sqrt{\nu_d(t)} \left( \mu_d(t) \nu_d(t) dt + dW_{d,k}^{d,T}(t) \right), \\
\frac{d\nu_d(t)}{\nu_d(t)} &= \lambda_d (\nu_d(0) - \nu_d(t)) dt + \eta_d \sqrt{\nu_d(t)} dW_{\nu_d}^{d,T}(t), \\
\end{align*}
\]

(3.3)

and

\[
\begin{align*}
\frac{dL_{f,k}(t)}{\tau_k} &= \sigma_{f,k} \frac{\phi_{f,k}(t)}{\eta_f(t)} \sqrt{\nu_f(t)} \left( \mu_f(t) \nu_f(t) dt + dW_{f,k}^{f,T}(t) \right), \\
\frac{d\nu_f(t)}{\nu_f(t)} &= \lambda_f (\nu_f(0) - \nu_f(t)) dt + \eta_f \sqrt{\nu_f(t)} dW_{\nu_f}^{f,T}(t),
\end{align*}
\]

(3.4)

with

\[
\mu_d(t) = -\sum_{j=k+1}^{N} \frac{\tau_j \phi_{d,j}(t) \sigma_{d,j} \rho_{d,j}^{d,k}}{1 + \tau_j L_{d,j}(t)} \rho_{d,j}^{d,k}, \quad \mu_f(t) = -\sum_{j=k+1}^{N} \frac{\tau_j \phi_{f,j}(t) \sigma_{f,j} \rho_{f,j}^{f,k}}{1 + \tau_j L_{f,j}(t)} \rho_{f,j}^{f,k},
\]

(3.5)

where

\[
\phi_{d,k} = \beta_{d,k} L_{d,k}(t) + (1 - \beta_{d,k}) L_{d,k}(0), \\
\phi_{f,k} = \beta_{f,k} L_{f,k}(t) + (1 - \beta_{f,k}) L_{f,k}(0).
\]

The Brownian motion, \( dW_{d,k}^{d,T} \), corresponds to the \( k \)-th domestic Libor rate, \( L_{d,k}(t) \), under the \( T \)-forward domestic measure, and the Brownian motion, \( dW_{f,k}^{f,T} \) relates to the \( k \)-th foreign market Libor rate, \( L_{f,k}(t) \) under the terminal foreign measure \( T \).

In the model \( \sigma_{d,k}(t) \) and \( \sigma_{f,k}(t) \) determine the level of the interest rate volatility smile, the parameters \( \beta_{d,k}(t) \) and \( \beta_{f,k}(t) \) control the slope of the volatility smile, and \( \lambda_d \), \( \lambda_f \) determine the speed of mean-reversion for the variance and influence the speed at which the interest rate volatility smile flattens as the swaption expiry increases [Pit05]. Parameters \( \eta_d \), \( \eta_f \) determine the curvature of the interest rate smile.
The following correlation structure is imposed, between

FX and its variance process, \( \sigma(t) \):
\[
dW^T_\xi(t)dW^T_\sigma(t) = \rho_{\xi,\sigma}dt,
\]

FX and domestic Libors, \( L_{d,j}(t) \):
\[
dW^T_\xi(t)dW^{dT}_j(t) = \rho_{\xi,j}dt,
\]

FX and foreign Libors, \( L_{f,j}(t) \):
\[
dW^T_\xi(t)d\hat{W}^{dT}_j(t) = \rho_{\xi,j}dt,
\]

Libors in domestic market:
\[
dW^{dT}_k(t)dW^{dT}_j(t) = \rho_{k,j}dt,
\]

Libors in foreign market:
\[
d\hat{W}^{dT}_k(t)d\hat{W}^{dT}_j(t) = \rho_{k,j}dt,
\]

Libors in domestic and foreign markets:
\[
dW^{dT}_k(t)d\hat{W}^{dT}_j(t) = \rho^{d,f}_{k,j}dt.
\]

We prescribe a zero correlation between the remaining processes, i.e., between

Libors and their variance process,
\[
dW^{dT}_k(t)dW^{dT}_\sigma(t) = 0,
\]

Libors and the FX variance process,
\[
dW^{dT}_k(t)dW^T_\sigma(t) = 0,
\]

all variance processes,
\[
dW^T_\xi(t)dW^{dT}_\sigma(t) = 0,
\]
\[
dW^T_\sigma(t)dW^{dT}_\sigma(t) = 0,
\]
\[
dW^T_\xi(t)d\hat{W}^{dT}_\sigma(t) = 0,
\]
\[
dW^T_\sigma(t)d\hat{W}^{dT}_\sigma(t) = 0,
\]

FX and the Libor variance processes,
\[
dW^T_\xi(t)dW^{dT}_\sigma(t) = 0,
\]
\[
dW^T_\xi(t)d\hat{W}^{dT}_\sigma(t) = 0.
\]

The correlation structure is graphically displayed in Figure 3.1.

![Figure 3.1: The correlation structure for the FX-HLMM model. Arrows indicate non-zero correlations. SV is Stochastic Volatility.](image)

Throughout this article we assume that the DD-SV model in (3.3) and (3.4) is already in the effective parameter framework as developed in [Pit05]. This means that approximate time-homogeneous parameters are used instead of the time-dependent parameters, i.e., \( \beta_k(t) \equiv \beta_k \) and \( \sigma_k(t) \equiv \sigma_k \).

With this correlation structure, we derive the dynamics for the forward FX, given by:
\[
FX^T(t) = \xi(t)\frac{P_f(t,T)}{P_d(t,T)},
\]
(see also (2.18)) with \( \xi(t) \) the spot exchange rate and \( P_d(t,T) \) and \( P_f(t,T) \) zero-coupon bonds. Note that the bonds are not yet specified.
When deriving the dynamics for (3.7), we need expressions for the zero-coupon bonds, $P_d(t, T)$ and $P_f(t, T)$. With Equation (3.2) the following expression for the final bond can be obtained:

$$
\frac{1}{P_i(t, T)} = \frac{1}{P_i(t, T_m(t))} \prod_{j=m(t)+1}^{N} (1 + \tau_j L_{i,j}(t)), \quad \text{for } i = \{d, f\},
$$

(3.8)

with $T = T_N$ and $m(t) = \min(k : t \leq T_k)$ (empty products in (3.8) are defined to be equal to 1). The bond $P_i(t, T_N)$ in (3.8) is fully determined by the Libor rates $L_{i,k}(t), \quad k=1, \ldots, N$ and the bond $P_i(t, T_{m(t)})$. Whereas the Libors $L_{i,k}(t)$ are defined by System (3.3) and (3.4), the bond $P_i(t, T_{m(t)})$ is not yet well-defined in the current framework.

To define continuous time dynamics for a zero-coupon bond, interpolation techniques are available (see, for example, [Sch02a; Pit04; DMP09; BJ09]). We consider here the linear interpolation scheme, proposed in [Sch02a], which reads:

$$
\frac{1}{P_i(t, T_{m(t)})} = 1 + (T_{m(t)} - t)L_{i,m(t)}(T_{m(t)-1}), \quad \text{for } T_{m(t)-1} < t < T_{m(t)}.
$$

(3.9)

In our previous work, [GO10], this basic interpolation technique was very satisfactory for the calibration. By combining (3.9) with (3.8), we find for the domestic and foreign bonds:

$$
\frac{1}{P_d(t, T)} = \left(1 + (T_{m(t)} - t)L_{d,m(t)}(T_{m(t)-1})\right) \prod_{j=m(t)+1}^{N} \left(1 + \tau_j L_{d,j}(t, T_{j-1}, T_j)\right),
$$

$$
\frac{1}{P_f(t, T)} = \left(1 + (T_{m(t)} - t)L_{f,m(t)}(T_{m(t)-1})\right) \prod_{j=m(t)+1}^{N} \left(1 + \tau_j L_{f,j}(t, T_{j-1}, T_j)\right).
$$

When deriving the dynamics for $FX^T(t)$ in (3.7) we will not encounter any $dt$-terms (as $FX^T(t)$ has to be a martingale under the numéraire $P_d(t, T)$).

For each zero-coupon bond, $P_d(t, T)$ or $P_f(t, T)$, the dynamics are determined under the appropriate T-forward measures (for $P_d(t, T)$ the domestic T-forward measure, and for $P_f(t, T)$ the foreign T-forward measure). The dynamics for the zero-coupon bonds, driven by the Libor dynamics in (3.3) and (3.4), are given by:

$$
\frac{dP_d(t, T)}{P_d(t, T)} = (\ldots)dt - \sqrt{v_d(t)} \sum_{j=m(t)+1}^{N} \frac{\tau_j \sigma_{d,j} \phi_{d,j}(t)}{1 + \tau_j L_{d,j}(t)} dW^d_{d,T}(t),
$$

(3.10)

$$
\frac{dP_f(t, T)}{P_f(t, T)} = (\ldots)dt - \sqrt{v_f(t)} \sum_{j=m(t)+1}^{N} \frac{\tau_j \sigma_{f,j} \phi_{f,j}(t)}{1 + \tau_j L_{f,j}(t)} dW^f_{f,T}(t),
$$

(3.11)

and the coefficients were defined in (3.3) and (3.4).

By changing the numéraire from $P_f(t, T)$ to $P_d(t, T)$ for the foreign bond, only the drift terms will change. Since $FX^T(t)$ in (3.7) is a martingale under the $P_d(t, T)$ measure, it is not necessary to determine the appropriate drift correction.

By taking Equation (2.20) for the general dynamics of (3.7) and neglecting all the $dt$-terms we get

$$
\frac{dFX^T(t)}{FX^T(t)} = \sqrt{\sigma(t)} dW^T_{\xi}(t) + \sqrt{v_d(t)} \sum_{j=m(t)+1}^{N} \frac{\tau_j \sigma_{d,j} \phi_{d,j}(t)}{1 + \tau_j L_{d,j}(t)} dW^d_{j,T}(t)
$$

$$
- \sqrt{v_f(t)} \sum_{j=m(t)+1}^{N} \frac{\tau_j \sigma_{f,j} \phi_{f,j}(t)}{1 + \tau_j L_{f,j}(t)} dW^f_{j,T}(t).
$$

(3.12)

Note that the hat in $\hat{W}$, disappeared from the Brownian motion $dW^f_{j,T}(t)$ in (3.12) which is an indication for the change of measure from the foreign to the domestic measure for the foreign Libors.
Since the stochastic volatility process, \( \sigma(t) \), for FX is independent of the domestic and foreign Libors, \( L_{d,k}(t) \) and \( L_{f,k}(t) \), the dynamics under the \( P_d(t, T) \)-measure do not change\(^3\) and are given by:

\[
de\sigma(t) = \kappa(\bar{\sigma} - \sigma(t))dt + \gamma\sqrt{\sigma(t)}dW^T(t). \tag{3.13}
\]

The model given in (3.12) with the stochastic variance in (3.13) and the correlations between the main underlying processes is not affine. In the next section we discuss a linearization.

### 3.1 Linearization and Forward Characteristic Function

The model in (3.12) is not of the affine form, as it contains terms like \( \phi_{i,j}(t)/(1 + \tau_{i,j}L_{i,j}(t)) \) with \( \phi_{i,j} = \beta_{i,j}L_{i,j}(t) + (1 - \beta_{i,j})L_{i,j}(0) \) for \( i = \{d, f\} \). In order to derive a characteristic function, we freeze the Libor rates, which is standard practice (see for example [GZ99; HW00; JR00]), i.e.:

\[
L_{d,j}(t) \approx L_{d,j}(0) \quad \Rightarrow \quad \phi_{d,j} \equiv L_{d,j}(0),
\]

\[
L_{f,j}(t) \approx L_{f,j}(0) \quad \Rightarrow \quad \phi_{f,j} \equiv L_{f,j}(0). \tag{3.14}
\]

This approximation gives the following FX\(^T\)-dynamics:

\[
\begin{align*}
\frac{dFX^T(t)}{FX^T(t)} & \approx \sqrt{\sigma(t)}dW^T(t) + \sqrt{v_d(t)} \sum_{j \in A} \psi_{d,j} dW^d_j(t) - \sqrt{v_f(t)} \sum_{j \in A} \psi_{f,j} dW^f_j(t), \\
\text{d} \sigma(t) &= \kappa(\bar{\sigma} - \sigma(t))dt + \gamma\sqrt{\sigma(t)}dW^T(t), \\
\text{d}v_i(t) &= \lambda_i(\bar{v}_i(t) - v_i(t))dt + \eta_i\sqrt{v_i(t)}dW^v_i(t),
\end{align*}
\]

with \( i = \{d, f\}, A = \{m(t) + 1, \ldots, N\} \), the correlations are given in (3.6) and

\[
\psi_{d,j} := \frac{\tau_j \sigma_{d,j} L_{d,j}(0)}{1 + \tau_j L_{d,j}(0)}, \quad \psi_{f,j} := \frac{\tau_j \sigma_{f,j} L_{f,j}(0)}{1 + \tau_j L_{f,j}(0)}. \tag{3.15}
\]

We derive the dynamics for the logarithmic transformation of \( FX^T(t) \), \( x^T(t) = \log FX^T(t) \), for which we need to calculate the square of the diffusion coefficients\(^4\).

With the notation,

\[
a := \sqrt{\sigma(t)}dW^T(t), \quad b := \sqrt{v_d(t)} \sum_{j \in A} \psi_{d,j} dW^d_j(t), \quad c := \sqrt{v_f(t)} \sum_{j \in A} \psi_{f,j} dW^f_j(t),
\]

we find, for the square diffusion coefficient \((a + b + c)^2 = a^2 + b^2 + c^2 + 2ab - 2ac - 2bc\).

So, the dynamics for the log-forward, \( x^T(t) = \log FX^T(t) \), can be expressed as:

\[
\text{d}x^T(t) \approx \left(-\frac{1}{2}(a + b + c)^2 + \sqrt{\sigma(t)}dW^T(t) + \sqrt{v_d(t)} \sum_{j \in A} \psi_{d,j} dW^d_j(t)
\right)
\]

\[
-\sqrt{v_f(t)} \sum_{j \in A} \psi_{f,j} dW^f_j(t), \tag{3.17}
\]

with the coefficients \( a, b \) and \( c \) given in (3.16). Since

\[
\left(\sum_{j=1}^{N} x_j\right)^2 = \sum_{j=1}^{N} x_j^2 + \sum_{i,j=1, i \neq j}^{N} x_i x_j, \quad \text{for} \ N > 0,
\]

\(^3\)In [GO10] the proof for this statement is given when a single yield curve is considered.

\(^4\)As in the standard Black-Scholes analysis for \( dS(t) = \sigma S(t)dW(t) \), the log-transform gives \( d\log S(t) = -\frac{1}{2}\sigma^2 dt + \sigma dW(t) \).
we find:
\[
\begin{align*}
    a^2 &= \sigma(t)dt, \\
    b^2 &= \nu_d(t) \left( \sum_{j \in A} \psi^2_{d,j} + \sum_{i,j \in A, i \neq j} \psi_{d,i} \psi_{d,j} \rho_{i,j} \right) dt =: \nu_d(t) A_d(t) dt, \quad (3.18) \\
    c^2 &= \nu_f(t) \left( \sum_{j \in A} \psi^2_{f,j} + \sum_{i,j \in A, i \neq j} \psi_{f,i} \psi_{f,j} \rho_{i,j} \right) dt =: \nu_f(t) A_f(t) dt, \quad (3.19) \\
    ab &= \sqrt{\sigma(t)} \sqrt{\nu_d(t)} \sum_{j \in A} \psi_{d,j} \rho_{d,j,x} dt, \\
    ac &= \sqrt{\sigma(t)} \sqrt{\nu_f(t)} \sum_{j \in A} \psi_{f,j} \rho_{f,j,x} dt, \\
    bc &= \sqrt{\nu_d(t)} \sqrt{\nu_f(t)} \sum_{j \in A} \psi_{d,j} \sum_{k \in A} \psi_{f,k} \rho_{d,f,j,k} dt,
\end{align*}
\]

with $\rho_{d,x}$, $\rho_{f,x}$ the correlation between the FX and $j$-th domestic and foreign Libor, respectively. The correlation between the $k$-th domestic and $j$-th foreign Libor is $\rho_{d,f}$.

By setting $f(t, \sqrt{\sigma(t)}, \sqrt{\nu_d(t)}, \sqrt{\nu_f(t)}) := (2ab - 2ac - 2bc)/dt$, we can express the dynamics for $dx^T(t)$ in (3.17) by:
\[
dx^T(t) \approx \frac{1}{2} \left( \sigma(t) + A_d(t) \nu_d(t) + A_f(t) \nu_f(t) + f(t, \sqrt{\sigma(t)}, \sqrt{\nu_d(t)}, \sqrt{\nu_f(t)}) \right) dt \\
+ \sqrt{\sigma(t)} dW^T(t) + \sqrt{\nu_d(t)} \sum_{j \in A} \psi_{d,j} dW_{d,j}^T(t) - \sqrt{\nu_f(t)} \sum_{j \in A} \psi_{f,j} dW_{f,j}^T(t).
\]

The coefficients $\psi_{d,j}$, $\psi_{f,j}$, $A_d$ and $A_f$ in (3.15), (3.18), and (3.19) are deterministic and piecewise constant.

In order to make the model affine, we linearize the non-affine terms in the drift in $f(t, \sqrt{\sigma(t)}, \sqrt{\nu_d(t)}, \sqrt{\nu_f(t)})$ by a projection on the first moments, i.e.,
\[
f(t, \sqrt{\sigma(t)}, \sqrt{\nu_d(t)}, \sqrt{\nu_f(t)}) \approx f(t, \mathbb{E}(\sqrt{\sigma(t)}), \mathbb{E}(\sqrt{\nu_d(t)}), \mathbb{E}(\sqrt{\nu_f(t)})) := f(t). \quad (3.20)
\]

The variance processes $\sigma(t)$, $\nu_d(t)$ and $\nu_f(t)$ are independent CIR-type processes [CIR85], so the expectation of their products equals the product of the expectations. Function $f(t)$ can be determined with the help of the formula in (2.30).

The approximation in (3.20) linearizes all non-affine terms in the corresponding PDE. As before, the forward characteristic function, $\phi^T := \phi^T(u, X(t), t, T)$, is defined as the solution of the following backward PDE:
\[
\begin{align*}
0 &= \frac{\partial \phi^T}{\partial t} + \frac{1}{2} \left( \sigma + A_d(t) \nu_d + A_f(t) \nu_f + f(t) \right) \left( \frac{\partial^2 \phi^T}{\partial x^2} - \frac{\partial \phi^T}{\partial x} \right) \\
&+ \lambda_d(\nu_d(0) - \nu_d) \frac{\partial \phi^T}{\partial \nu_d} + \lambda_f(\nu_f(0) - \nu_f) \frac{\partial \phi^T}{\partial \nu_f} + \kappa(\bar{\sigma} - \sigma) \frac{\partial \phi^T}{\partial \sigma} \\
&+ \frac{1}{2} \nu_d^2 \frac{\partial^2 \phi^T}{\partial \nu_d^2} + \frac{1}{2} \nu_f^2 \frac{\partial^2 \phi^T}{\partial \nu_f^2} + \frac{1}{2} \gamma^2 \sigma \frac{\partial^2 \phi^T}{\partial \sigma^2} + \rho_{x,\sigma} \gamma \sigma \frac{\partial^2 \phi^T}{\partial x \partial \sigma}, \quad (3.21)
\end{align*}
\]

with the final condition $\phi^T(u, X(T), T) = e^{iu \tau^T(T)}$. Since all coefficients in this PDE are linear, the solution is of the following form:
\[
\begin{align*}
\phi^T(u, X(t), t, T) &= \exp \left( A(u, \tau) + B(u, \tau) x^T(t) + C(u, \tau) \sigma(t) \\
&+ D_d(u, \tau) \nu_d(t) + D_f(u, \tau) \nu_f(t) \right), \quad (3.22)
\end{align*}
\]
with \( \tau := T - t \). Substitution of (3.22) in (3.21) gives us the following system of ODEs for the functions \( A(u, \tau) = A(u, \tau), B(u, \tau), C(u, \tau), D_d(u, \tau) \) and \( D_f(u, \tau) := D_f(u, \tau) \):

\[
A'(\tau) = f(t)(B^2(\tau) - B(\tau))/2 + \lambda_d v_d(0) D_1(\tau) + \lambda_f v_f(0) D_2(\tau) + \kappa \bar{\sigma} C(\tau),
B'(\tau) = 0,
C'(\tau) = (B^2(\tau) - B(\tau))/2 + (\rho_x \sigma \gamma B(\tau) - \kappa) C(\tau) + \gamma^2 C^2(\tau)/2,
D'_d(\tau) = A_d(t)(B^2(\tau) - B(\tau))/2 - \lambda_d D_d(\tau) + \eta_k^2 D_d^2(\tau)/2,
D'_f(\tau) = A_f(t)(B^2(\tau) - B(\tau))/2 - \lambda_f D_f(\tau) + \eta_f^2 D_f^2(\tau)/2,
\]

with initial conditions \( A(0) = 0, B(0) = iu, C(0) = 0, D_d(0) = 0, D_f(0) = 0 \) with \( A_d(t) \) and \( A_f(t) \) from (3.18), (3.19), respectively, and \( f(t) \) as in (3.20).

With \( B(\tau) = iu \), the solution for \( C(\tau) \) is analogous to the solution for the ODE for the FX-HHW1 model in Equation (2.33). As the remaining ODEs involve the piecewise constant functions \( A_d(t), A_f(t) \) the solution must be determined iteratively, like for the pure Heston model with piecewise constant parameters in [AA00]. For a given grid \( 0 = \tau_0 < \tau_1 < \cdots < \tau_N = \tau \), the functions \( D_d(u, \tau), D_f(u, \tau) \) and \( A(u, \tau) \) can be expressed as:

\[
D_d(u, \tau_j) = D_d(u, \tau_{j-1}) + \chi_d(u, \tau_j),
D_f(u, \tau_j) = D_f(u, \tau_{j-1}) + \chi_f(u, \tau_j),
\]

for \( j = 1, \ldots, N \), and

\[
A(u, \tau_j) = A(u, \tau_{j-1}) + \chi_A(u, \tau_j) - \frac{1}{2} (u^2 + u) \int_{\tau_{j-1}}^{\tau_j} f(s) ds,
\]

with \( f(s) \) in (3.20) and analytically known functions \( \chi_k(u, \tau_j) \), for \( k = \{d, f\} \) and \( \chi_A(u, \tau_j) \):

\[
\chi_k(u, \tau_j) := (\lambda_k - \delta_{k,j} - \eta_k^2 D_k(u, \tau_{j-1})) (1 - e^{-\delta_{k,j} \tau_j})/(\eta_k^2 (1 - \ell_{k,j} e^{-\delta_{k,j} \tau_j})),
\]

and

\[
\chi_A(u, \tau_j) = \frac{\kappa \bar{\sigma}}{\tau} \left( (\kappa - \rho_x \sigma \gamma iu - d_j) s_j - 2 \log \left( (1 - g_j e^{-d_j \tau_j})/(1 - g_j) \right) \right)
+ v_d(0) \frac{\lambda_d}{\eta_d} \left( (\lambda_d - \delta_{d,j}) s_j - 2 \log \left( (1 - \ell_{d,j} e^{-\delta_{d,j} \tau_j})/(1 - \ell_{d,j}) \right) \right),
\]

where

\[
d_j = \sqrt{(\rho_x \sigma \gamma iu - \kappa)^2 + \gamma^2 (iu + u^2)}, \quad g_j = \frac{(\kappa - \rho_x \sigma \gamma iu) - d_j - \gamma^2 C(u, \tau_{j-1})}{(\kappa - \rho_x \sigma \gamma iu) + d_j - \gamma^2 C(u, \tau_{j-1})},
\]

\[
\delta_{k,j} = \sqrt{\lambda_k^2 + 2 \lambda_k A_k(t)/(u^2 + iu)}, \quad \ell_{k,j} = \frac{\lambda_k - \delta_{k,j} - \eta_k^2 D_k(u, \tau_{j-1})}{\lambda_k + \delta_{k,j} - \eta_k^2 D_k(u, \tau_{j-1})},
\]

with \( s_j = \tau_j - \tau_{j-1}, \ j = 1, \ldots, N \), \( A_d(t) \) and \( A_f(t) \) are from (3.18) and (3.19).

The resulting approximation of the full-scale FX-HLMM model is called FX-LMMI here.

### 3.2 Foreign Stock in the FX-HLMM Framework

We also consider a foreign stock, \( S_f(t) \), driven by the Heston stochastic volatility model, with the interest rates driven by the market model. The stochastic processes
of the stock model are assumed to be of the same form as the FX (with one, foreign, interest rate curve) with the dynamics, under the forward foreign measure, given by:

\[
\frac{dS_f^T(t)}{S_f^T(t)} = \sqrt{\omega(t)}dW_{S_f}^{f,T}(t) + \sqrt{\nu_f(t)} \sum_{j=m(t)+1}^{N} \frac{\tau_f\sigma_f\phi_{f,j}(t)}{1 + \tau_fL_{f,j}(t)}dW_{j}^{f,T}(t),
\]

\[
d\omega(t) = \kappa_f(\bar{\omega} - \omega(t))dt + \gamma_f\sqrt{\omega(t)}dW_{f}^{f,T}(t).
\] (3.23)

Variance process, \(\omega(t)\), is correlated with forward stock \(S^T(t)\).

We move to the domestic-forward measure. The forward stock, \(S_f^T\), and forward foreign exchange rate, \(FX^T(t)\), are defined by

\[
S_f^T(t) = \frac{S_f(t)}{P_f(t,T)}, \quad FX^T(t) = \xi(t)\frac{P_f(t,T)}{P_d(t,T)}.
\] (3.24)

The quantity

\[
S_f^T(t)FX^T(t) = S_f(t)\frac{P_f(t,T)}{P_d(t,T)}\xi(t) = S_f(t)\frac{P_f(t,T)}{P_d(t,T)}\xi(t),
\] (3.25)

is therefore a tradable asset. So, foreign stock exchanged by a foreign exchange rate and denominated in the domestic zero-coupon bond is a tradable quantity, which implies that \(S_f^T(t)FX^T(t)\) is a martingale. By Itô’s lemma, one finds:

\[
\frac{d}{S_f^T(t)FX^T(t)} = \frac{dFX^T(t)}{FX^T(t)} + \frac{dS_f^T(t)}{S_f^T(t)} + \left(\frac{dFX^T(t)}{FX^T(t)}\right)\left(\frac{dS_f^T(t)}{S_f^T(t)}\right).
\] (3.26)

The two first terms at the RHS of (3.26) do not contribute to the drift. The last term involves all \(dt\)-terms, that, by a change of measure, will enter the drift of the variance process \(d\omega(t)\) in (3.23).

### 3.3 Numerical Experiments with the FX-HLMM Model

We here focus on the FX-HLMM model covered in Section 3 and consider the errors generated by the various approximations that led to the model FX-HLMM. We have performed basically two linearization steps to define FX-HLMM: We have frozen the Libors at their initial values and projected the non-affine covariance terms on a deterministic function. We check, by a numerical experiment, the size of the errors of these approximations.

We have chosen the following interest rate curves \(P_d(t = 0,T) = \exp(-0.02T)\), \(P_f(t = 0,T) = \exp(-0.05T)\), and, as before, for the FX stochastic volatility model we set:

\[
\kappa = 0.5, \quad \gamma = 0.3, \quad \sigma = 0.1, \quad \sigma(0) = 0.1.
\] (3.27)

In the simulation we have chosen the following parameters for the domestic and foreign markets:

\[
\beta_{d,k} = 95\%, \quad \sigma_{d,k} = 15\%, \quad \lambda_d = 100\%, \quad \eta_d = 10\%,
\]

\[
\beta_{f,k} = 50\%, \quad \sigma_{f,k} = 25\%, \quad \lambda_f = 70\%, \quad \eta_f = 20\%.
\] (3.28)

In the correlation matrix a number of correlations need to be specified. For the correlations between the Libor rates in each market, we prescribe large positive values, as frequently observed in fixed income markets (see for example [BM07]), \(\rho_{i,j}^{d} = 90\%\), \(\rho_{i,j}^{f} = 70\%\), for \(i, j = 1, \ldots, N\) \((i \neq j)\). In order to generate skew for FX, we prescribe a negative correlation between FX\(^T(t)\) and its stochastic volatility process, \(\sigma(t)\), i.e., \(\rho_{\xi,\sigma} = -40\%\). The correlation between the FX and the domestic Libors is set as \(\rho_{\xi,k}^{d} = -15\%\), for \(k = 1, \ldots, N\), and the correlation between FX and the foreign
Libors is \( \rho_{k}^{i} = -15\% \). The correlation between the domestic and foreign Libors is \( \rho_{i,j}^{d,f} = 25\% \) for \( i, j = 1, \ldots, N \) \( (i \neq j) \). The following block correlation matrix results:

\[
C = \begin{bmatrix}
C_{d} & C_{d,f} & C_{\xi,d} \\
C_{d,f}^T & C_{f} & C_{\xi,f} \\
C_{\xi,d}^T & C_{\xi,f} & 1
\end{bmatrix},
\]

(3.30)

with the domestic Libor correlations given by

\[
C_{d} = \begin{bmatrix}
1 & \rho_{1,2}^{d} & \ldots & \rho_{1,N}^{d} \\
\rho_{1,2}^{d} & 1 & \ldots & \rho_{2,N}^{d} \\
\vdots & \vdots & \ddots & \vdots \\
\rho_{1,N}^{d} & \rho_{2,N}^{d} & \ldots & 1
\end{bmatrix} = \begin{bmatrix}
1 & 90\% & \ldots & 90\% \\
90\% & 1 & \ldots & 90\% \\
\vdots & \vdots & \ddots & \vdots \\
90\% & 90\% & \ldots & 1
\end{bmatrix}_{N \times N},
\]

(3.31)

the foreign Libors correlations given by:

\[
C_{f} = \begin{bmatrix}
1 & \rho_{1,2}^{f} & \ldots & \rho_{1,N}^{f} \\
\rho_{1,2}^{f} & 1 & \ldots & \rho_{2,N}^{f} \\
\vdots & \vdots & \ddots & \vdots \\
\rho_{1,N}^{f} & \rho_{2,N}^{f} & \ldots & 1
\end{bmatrix} = \begin{bmatrix}
1 & 70\% & \ldots & 70\% \\
70\% & 1 & \ldots & 70\% \\
\vdots & \vdots & \ddots & \vdots \\
70\% & 70\% & \ldots & 1
\end{bmatrix}_{N \times N},
\]

(3.32)

the correlation between Libors from the domestic and foreign markets given by:

\[
C_{d,f} = \begin{bmatrix}
\rho_{1,2}^{d,f} & \ldots & \rho_{1,N}^{d,f} \\
\rho_{2,1}^{d,f} & \ldots & \rho_{2,N}^{d,f} \\
\vdots & \ddots & \vdots \\
\rho_{N,1}^{d,f} & \ldots & \rho_{N,N}^{d,f}
\end{bmatrix} = \begin{bmatrix}
1 & 25\% & \ldots & 25\% \\
25\% & 1 & \ldots & 25\% \\
\vdots & \vdots & \ddots & \vdots \\
25\% & 25\% & \ldots & 1
\end{bmatrix}_{N \times N},
\]

(3.33)

and the vectors \( C_{\xi,d} \) and \( C_{\xi,f} \), given by:

\[
C_{\xi,d} = \begin{bmatrix}
\rho_{1}^{\xi} \\
\rho_{2}^{\xi} \\
\vdots \\
\rho_{N}^{\xi}
\end{bmatrix}
\]

(3.34)

\[
C_{\xi,f} = \begin{bmatrix}
\rho_{1}^{\xi} \\
\rho_{2}^{\xi} \\
\vdots \\
\rho_{N}^{\xi}
\end{bmatrix}
\]

Since in both markets the Libor rates are assumed to be independent of their variance processes, we can neglect these correlations here.

Now we find the prices of plain vanilla options on FX in (3.7). The simulation is performed in the same spirit as in Section 2.5 where the FX-HHW model was considered. In Table 3.1 we present the differences, in terms of the implied volatilities between the models FX-HLMM and FX-HLMM1. While the prices for the FX-HLMM were obtained by Monte-Carlo simulation (20.000 paths and 20 intermediate points between the dates \( T_{i-1} \) and \( T_{i} \) for \( i = 1, \ldots, N \)), the prices for FX-HLMM1 were obtained by the Fourier-based COS method [FO08] with 500 Fourier series terms.

<table>
<thead>
<tr>
<th>( T_{i} )</th>
<th>( K_{1}(T_{i}) )</th>
<th>( K_{2}(T_{i}) )</th>
<th>( K_{3}(T_{i}) )</th>
<th>( K_{4}(T_{i}) )</th>
<th>( K_{5}(T_{i}) )</th>
<th>( K_{6}(T_{i}) )</th>
<th>( K_{7}(T_{i}) )</th>
</tr>
</thead>
<tbody>
<tr>
<td>2y</td>
<td>0.0019</td>
<td>0.0014</td>
<td>0.0009</td>
<td>0.0005</td>
<td>0.0000</td>
<td>-0.0005</td>
<td>-0.0010</td>
</tr>
<tr>
<td>3y</td>
<td>0.0029</td>
<td>0.0025</td>
<td>0.0021</td>
<td>0.0016</td>
<td>0.0011</td>
<td>0.0006</td>
<td>0.0002</td>
</tr>
<tr>
<td>5y</td>
<td>0.0032</td>
<td>0.0028</td>
<td>0.0023</td>
<td>0.0017</td>
<td>0.0010</td>
<td>0.0005</td>
<td>0.0000</td>
</tr>
<tr>
<td>7y</td>
<td>0.0030</td>
<td>0.0028</td>
<td>0.0025</td>
<td>0.0021</td>
<td>0.0018</td>
<td>0.0014</td>
<td>0.0010</td>
</tr>
<tr>
<td>10y</td>
<td>0.0032</td>
<td>0.0032</td>
<td>0.0025</td>
<td>0.0018</td>
<td>0.0012</td>
<td>0.0005</td>
<td>-0.0003</td>
</tr>
<tr>
<td>15y</td>
<td>0.0038</td>
<td>0.0029</td>
<td>0.0021</td>
<td>0.0013</td>
<td>0.0005</td>
<td>-0.0004</td>
<td>-0.0014</td>
</tr>
<tr>
<td>20y</td>
<td>0.0002</td>
<td>-0.0009</td>
<td>-0.0018</td>
<td>-0.0027</td>
<td>-0.0034</td>
<td>-0.0040</td>
<td>-0.0044</td>
</tr>
<tr>
<td>25y</td>
<td>0.0008</td>
<td>0.0004</td>
<td>-0.0014</td>
<td>-0.0025</td>
<td>-0.0034</td>
<td>-0.0040</td>
<td>-0.0046</td>
</tr>
<tr>
<td>30y</td>
<td>0.0011</td>
<td>0.0007</td>
<td>0.0000</td>
<td>-0.0009</td>
<td>-0.0018</td>
<td>-0.0021</td>
<td>-0.0024</td>
</tr>
</tbody>
</table>

Table 3.1: Differences, in implied Black volatilities, between the FX-HLMM and FX-LMM1 models. The corresponding strikes \( K_{1}(T_{i}), \ldots, K_{7}(T_{i}) \) are tabulated in Table B.1. The prices and associated standard deviations are presented in Table B.5.

The FX-HLMM1 model performs very well, as the maximum difference in terms of implied volatilities is about 0.2\% - 0.5\%.
3.3.1 Sensitivity to the Interest Rate Skew

Approximation FX-HLMM1 was based on freezing the Libor rates. By freezing the Libors, i.e.:

\[ L_{d,k}(t) \equiv L_{d,k}(0) \]
\[ L_{f,k}(t) \equiv L_{f,k}(0) \]

we have

\[ \phi_{d,k}(t) = \beta_{d,k} L_{d,k}(t) + (1 - \beta_{d,k}) L_{d,k}(0) = L_{d,k}(0), \quad (3.35) \]
\[ \phi_{f,k}(t) = \beta_{f,k} L_{f,k}(t) + (1 - \beta_{f,k}) L_{f,k}(0) = L_{f,k}(0). \quad (3.36) \]

In the DD-SV models for the Libor rates \( L_{d,k}(t) \) and \( L_{f,k}(t) \) for any \( k \), the parameters \( \beta_{d,k}, \beta_{f,k} \) control the slope of the interest rate volatility smiles. Freezing the Libors to \( L_{d,k}(0) \) and \( L_{f,k}(0) \) is equivalent to setting \( \beta_{d,k} = 0 \) and \( \beta_{f,k} = 0 \) in (3.35) and (3.36) in the approximation FX-HLMM1.

By a Monte Carlo simulation, we obtain the FX implied volatilities from the full scale FX-HLMM model for different values of \( \beta \) and by comparing them to those from FX-HLMM1 with \( \beta = 0 \) we check the influence of the parameters \( \beta_{d,k} \) and \( \beta_{f,k} \) on the FX. In Table 3.2 the implied volatilities for the FX European call options for FX-HLMM and FX-HLMM1 are presented. The experiments are performed for different combinations of the interest rate skew parameters, \( \beta_d \) and \( \beta_f \).

<table>
<thead>
<tr>
<th>strike</th>
<th>( \beta_d = 0 )</th>
<th>( \beta_d = 0.5 )</th>
<th>( \beta_d = 1 )</th>
<th>( \beta_f = 0 )</th>
<th>( \beta_f = 0.5 )</th>
<th>( \beta_f = 1 )</th>
</tr>
</thead>
<tbody>
<tr>
<td>0.6224</td>
<td>0.3198 (0.0020)</td>
<td>0.3191 (0.0017)</td>
<td>0.3198 (0.0017)</td>
<td>0.3199 (0.0015)</td>
<td>0.3196 (0.0018)</td>
<td>0.3156 (0.0015)</td>
</tr>
<tr>
<td>0.7900</td>
<td>0.3149 (0.0021)</td>
<td>0.3143 (0.0016)</td>
<td>0.3148 (0.0016)</td>
<td>0.3151 (0.0015)</td>
<td>0.3146 (0.0018)</td>
<td>0.3112 (0.0018)</td>
</tr>
<tr>
<td>0.8356</td>
<td>0.3102 (0.0021)</td>
<td>0.3096 (0.0017)</td>
<td>0.3101 (0.0020)</td>
<td>0.3104 (0.0015)</td>
<td>0.3097 (0.0018)</td>
<td>0.3069 (0.0018)</td>
</tr>
<tr>
<td>1.0001</td>
<td>0.3058 (0.0021)</td>
<td>0.3053 (0.0017)</td>
<td>0.3056 (0.0022)</td>
<td>0.3061 (0.0015)</td>
<td>0.3052 (0.0017)</td>
<td>0.3030 (0.0017)</td>
</tr>
<tr>
<td>1.1714</td>
<td>0.3016 (0.0020)</td>
<td>0.3011 (0.0017)</td>
<td>0.3015 (0.0024)</td>
<td>0.3020 (0.0015)</td>
<td>0.3008 (0.0016)</td>
<td>0.2993 (0.0016)</td>
</tr>
<tr>
<td>1.3721</td>
<td>0.2977 (0.0022)</td>
<td>0.2973 (0.0016)</td>
<td>0.2977 (0.0026)</td>
<td>0.2982 (0.0016)</td>
<td>0.2968 (0.0017)</td>
<td>0.2960 (0.0017)</td>
</tr>
<tr>
<td>1.6071</td>
<td>0.2941 (0.0024)</td>
<td>0.2938 (0.0017)</td>
<td>0.2943 (0.0028)</td>
<td>0.2948 (0.0017)</td>
<td>0.2931 (0.0018)</td>
<td>0.2930 (0.0018)</td>
</tr>
</tbody>
</table>

Table 3.2: Implied volatilities of the FX options from the FX-HLMM and FX-HLMM1 models, \( T = 10 \) and parameters were as in Section 3.3. The numbers in parentheses correspond to the standard deviations (the experiment was performed 20 times with 20T time steps).

The experiment indicates that there is only a small impact of the different \( \beta_{d,k} \)− and \( \beta_{f,k} \)−values on the FX implied volatilities, implying that the approximate model, FX-HLMM1 with \( \beta_{d,k} = \beta_{f,k} = 0 \), is useful for the interest rate modelling, for the parameters studied. With \( \beta_{d,k} \neq 0 \) and \( \beta_{f,k} \neq 0 \) the implied volatilities obtained by the FX-HLMM model appear to be somewhat higher than those obtained by FX-HLMM1, a difference of approximately 0.1% − 0.15%, which is considered highly satisfactory.

4 Conclusion

In this article we have presented two FX models with stochastic volatility and correlated stochastic interest rates. Both FX models were based on the Heston FX model and differ with respect to the interest rate processes.

In the first model we considered a model in which the domestic and foreign interest rates were driven by single factor Hull-White short-rate processes. This model enables pricing of FX-interest rate hybrid products that are not exposed to the smile in the fixed income markets.

For hybrid products sensitive to the interest rate skew a second model was presented in which the interest rates were driven by the stochastic volatility Libor Market Model.
For both hybrid models we have developed approximate models for the pricing of European options on the FX. These pricing formulas form the basis for highly efficient model calibration strategies.

The approximate models are based on the linearization of the non-affine terms in the corresponding pricing PDE, in a very similar way as in our previous article [GO09] on equity-interest rate options. The approximate models perform very well in the world of foreign exchange.

These models can also be used to obtain an initial guess when the full-scale models are used.

Acknowledgments

The authors would like to thank to Sacha van Weeren and Natalia Borovykh from Rabobank International for fruitful discussions and helpful comments.

References

A Proof of Lemma 2.2

Since the domestic short rate process, \( r_d(t) \), is driven by one source of uncertainty (only one Brownian motion \( d\tilde{W}_d(t) \)), it is convenient to change the order of the state variables, from \( d\mathbf{X}(t) = [d\tilde{X}^T(t),d\tilde{f}(t),d\sigma(t),d\eta(t),dr_d(t),dr_f(t)]^T \) to \( d\mathbf{X}^*(t) = [dr_d(t),dr_f(t),d\sigma(t),d\tilde{X}^T(t)/FX^T(t)]^T \) and express the model in terms of the independent Brownian motions \( d\tilde{W}^Q(t) = [d\tilde{W}_d(t),d\tilde{W}_f(t),d\tilde{W}_\sigma(t),d\tilde{W}_\xi(t)]^T \), i.e.:

\[
\begin{bmatrix}
    dr_d \\
    dr_f \\
    d\sigma \\
    d\tilde{X}^T/FX^T
\end{bmatrix}
= \mu(\mathbf{X}^*)dt +
\begin{bmatrix}
    \eta_d \\
    \eta_f \\
    0 \\
    -\eta_d B_d & \eta_f B_f & 0 & \gamma \sqrt{\sigma}
\end{bmatrix}
\begin{bmatrix}
    d\tilde{W}_d^Q \\
    d\tilde{W}_f^Q \\
    d\tilde{W}_\sigma^Q \\
    d\tilde{W}_\xi^Q
\end{bmatrix},
\]

(A.1)

which, equivalently, can be written as:

\[
d\mathbf{X}^*(t) = \mu(\mathbf{X}^*)dt + \mathbf{A} \mathbf{d}\tilde{W}^Q(t),
\]

(A.2)

where \( \mu(\mathbf{X}^*) \) represents the drift for system \( d\mathbf{X}^*(t) \) and \( \mathbf{H} \) is the Cholesky lower-triangular matrix of the following form:

\[
\mathbf{H} =
\begin{bmatrix}
    1 & 0 & 0 & 0 \\
    H_{2,1} & H_{2,2} & 0 & 0 \\
    H_{3,1} & H_{3,2} & H_{3,3} & 0 \\
    H_{4,1} & H_{4,2} & H_{4,3} & H_{4,4}
\end{bmatrix}
= \begin{bmatrix}
    1 & 0 & 0 & 0 \\
    \rho_{f,d} & H_{2,2} & 0 & 0 \\
    \rho_{r,d} & H_{3,2} & H_{3,3} & 0 \\
    \rho_{\eta,d} & H_{4,2} & H_{4,3} & H_{4,4}
\end{bmatrix}.
\]

(A.3)

The representation presented above seems to be favorable, since the short-rate process \( r_d(t) \) can be considered independently of the other processes.

The matrix model representation in terms of orthogonal Brownian motions results in the following dynamics for the domestic short rate \( r_d(t) \) under measure \( Q \):

\[
dr_d(t) = \lambda_d(\theta_d(t) - r_d(t))dt + \zeta_1(t)d\tilde{W}^Q(t),
\]

and for the domestic ZCB:

\[
\frac{dP_d(t,T)}{P_d(t,T)} = r_d(t)dt + B_d(t,T)\zeta_1(t)q^Q(t),
\]

with \( \zeta_k(t) \) being the \( k \)th row vector resulting from multiplying the matrices \( \mathbf{A} \) and \( \mathbf{H} \). Note, that for the 1D Hull-White short rate processes \( \zeta_1(t) = [\eta_d,0,0,0] \).

Now, we derive the Radon-Nikodým derivative [GKR96], \( \Lambda^Q(t) \):

\[
\Lambda^Q(t) = \frac{dQ}{dQ} = \frac{P_d(t,T)}{P_d(0,T)M_d(t)}.
\]

(A.4)

By calculating the Itô derivative of Equation (A.4) we get:

\[
\frac{d\Lambda^T}{\Lambda^Q} = B_d(t,T)\zeta_1(t)d\tilde{W}^Q(t),
\]

(A.5)

which implies that the Girsanov kernel for the transition from \( Q \) to \( Q^T \) is given by \( B_d(t,T)\zeta_1(t) \) which is the \( T \)-bond volatility given by \( \eta_d B_d(t,T) \), i.e.:

\[
\Lambda^T_Q = \exp \left( \frac{1}{2} \int_0^T B_d^2(s,T)\zeta_1^2(s)ds + \int_0^T B_d(s,T)\zeta_1(s)d\tilde{W}^Q(s) \right).
\]

(A.6)

So,

\[
d\tilde{W}^T(t) = -B_d(t,T)\zeta_1^T(t)dt + d\tilde{W}^Q(t).
\]

Since the vector \( \zeta_1^T(t) \) is of scalar form, the Brownian motion under the \( T \)-forward measure is given by:

\[
d\tilde{W}^Q(t) = \left[ d\tilde{W}_d^Q(t) + \eta_d B_d(t,T)dt, d\tilde{W}_f^T(t), d\tilde{W}_\sigma^T(t), d\tilde{W}_\xi^T(t) \right]^T.
\]
Now, from the vector representation (A.2) we get that:

\[
H_d \tilde{W}^Q = \begin{bmatrix}
\eta_d B_d + \\
\rho_{d,f} \eta_d B_d dt + \\
\rho_{d,a} \eta_d B_d dt + \\
\rho_{d,\sigma} \eta_d \tilde{W}_d^{\sigma} + \\
\rho_{\sigma,d} \eta_d \tilde{W}_d^{\sigma} + \\
\rho_{\sigma,\sigma} \eta_d \tilde{W}_d^{\sigma} + \\
\rho_{\xi,a} \eta_d \tilde{W}_d^{\sigma} + \\
\rho_{\xi,\sigma} \eta_d \tilde{W}_d^{\sigma} + \mathbf{H}_{2,2} \tilde{W}_d^f + \\
\rho_{\sigma,\xi} \eta_d \tilde{W}_d^{\sigma} + \mathbf{H}_{3,2} \tilde{W}_d^f + \\
\rho_{\sigma,\xi} \eta_d \tilde{W}_d^{\sigma} + \mathbf{H}_{4,2} \tilde{W}_d^f + \\
\rho_{\sigma,\sigma} \eta_d \tilde{W}_d^{\sigma} + \mathbf{H}_{4,4} \tilde{W}_d^f 
\end{bmatrix}. \tag{A.7}
\]

Returning to the dependent Brownian motions under the T-forward measure, gives us:

\[
\frac{dFX^T(t)}{FX^T(t)} = \sqrt{\sigma(t)} dW_T^\xi(t) - \eta_d B_d(t, T) dW_d^T(t) + \eta_f B_f(t, T) dW_f^T(t),
\]

\[
d\sigma(t) = \left(\kappa(\bar{\sigma} - \sigma(t)) + \gamma \rho_{\sigma,a} \eta_d B_d(t, T) \sqrt{\sigma(t)}\right) dt + \gamma \sqrt{\sigma(t)} dW_T^\sigma(t),
\]

\[
dr_d(t) = \left(\lambda_d(d\theta_d(t) - r_d(t)) + \eta_d^2 B_d(t, T)\right) dt + \eta_d dW_d^T(t),
\]

\[
dr_f(t) = \left(\lambda_f(d\theta_f(t) - r_f(t)) - \eta_f \rho_{\xi,f} \sqrt{\sigma(t)} + \eta_d \eta_f \rho_{d,f} B_d(t, T)\right) dt + \eta_f dW_f^T(t),
\]

with full matrix of correlations given in (2.12).

## B Tables

In this appendix we present tables with details for the numerical experiments.

<table>
<thead>
<tr>
<th>$T_i$</th>
<th>$K_1(T_i)$</th>
<th>$K_2(T_i)$</th>
<th>$K_3(T_i)$</th>
<th>$K_4(T_i)$</th>
<th>$K_5(T_i)$</th>
<th>$K_6(T_i)$</th>
<th>$K_7(T_i)$</th>
</tr>
</thead>
<tbody>
<tr>
<td>6m</td>
<td>1.1961</td>
<td>1.2391</td>
<td>1.2837</td>
<td>1.3299</td>
<td>1.3778</td>
<td>1.4273</td>
<td>1.4787</td>
</tr>
<tr>
<td>1y</td>
<td>1.1276</td>
<td>1.1854</td>
<td>1.2462</td>
<td>1.3101</td>
<td>1.3773</td>
<td>1.4479</td>
<td>1.5221</td>
</tr>
<tr>
<td>3y</td>
<td>0.9515</td>
<td>1.0376</td>
<td>1.1315</td>
<td>1.2338</td>
<td>1.3454</td>
<td>1.4671</td>
<td>1.5999</td>
</tr>
<tr>
<td>5y</td>
<td>0.8290</td>
<td>0.9291</td>
<td>1.0390</td>
<td>1.1620</td>
<td>1.2904</td>
<td>1.4531</td>
<td>1.6250</td>
</tr>
<tr>
<td>7y</td>
<td>0.7458</td>
<td>0.8399</td>
<td>0.9587</td>
<td>1.0943</td>
<td>1.2491</td>
<td>1.4251</td>
<td>1.6274</td>
</tr>
<tr>
<td>10y</td>
<td>0.6224</td>
<td>0.7290</td>
<td>0.8538</td>
<td>1.0001</td>
<td>1.1714</td>
<td>1.3721</td>
<td>1.6071</td>
</tr>
<tr>
<td>15y</td>
<td>0.4815</td>
<td>0.5844</td>
<td>0.7093</td>
<td>0.8608</td>
<td>1.0447</td>
<td>1.2680</td>
<td>1.5389</td>
</tr>
<tr>
<td>20y</td>
<td>0.3788</td>
<td>0.4737</td>
<td>0.5924</td>
<td>0.7409</td>
<td>0.9265</td>
<td>1.1987</td>
<td>1.4401</td>
</tr>
<tr>
<td>25y</td>
<td>0.3012</td>
<td>0.3868</td>
<td>0.4966</td>
<td>0.6377</td>
<td>0.8188</td>
<td>1.0514</td>
<td>1.3500</td>
</tr>
<tr>
<td>30y</td>
<td>0.2414</td>
<td>0.3174</td>
<td>0.4174</td>
<td>0.5489</td>
<td>0.7218</td>
<td>0.9492</td>
<td>1.2482</td>
</tr>
</tbody>
</table>

Table B.1: Expiries and strikes of FX options used in the FX-HHW model. Strikes $K_n(T_i)$ were calculated as given in (2.42) with $\xi(0) = 1.35$.

<table>
<thead>
<tr>
<th>$T_i$</th>
<th>$K_1(T_i)$</th>
<th>$K_2(T_i)$</th>
<th>$K_3(T_i)$</th>
<th>$K_4(T_i)$</th>
<th>$K_5(T_i)$</th>
<th>$K_6(T_i)$</th>
<th>$K_7(T_i)$</th>
</tr>
</thead>
<tbody>
<tr>
<td>6m</td>
<td>0.1141</td>
<td>0.1049</td>
<td>0.0966</td>
<td>0.0902</td>
<td>0.0872</td>
<td>0.0866</td>
<td>0.0868</td>
</tr>
<tr>
<td>1y</td>
<td>0.1223</td>
<td>0.1098</td>
<td>0.0982</td>
<td>0.0895</td>
<td>0.0859</td>
<td>0.0859</td>
<td>0.0865</td>
</tr>
<tr>
<td>3y</td>
<td>0.1294</td>
<td>0.1135</td>
<td>0.0989</td>
<td>0.0878</td>
<td>0.0834</td>
<td>0.0836</td>
<td>0.0846</td>
</tr>
<tr>
<td>5y</td>
<td>0.1344</td>
<td>0.1184</td>
<td>0.1038</td>
<td>0.0927</td>
<td>0.0876</td>
<td>0.0871</td>
<td>0.0883</td>
</tr>
<tr>
<td>7y</td>
<td>0.1429</td>
<td>0.1268</td>
<td>0.1123</td>
<td>0.1012</td>
<td>0.0952</td>
<td>0.0937</td>
<td>0.0943</td>
</tr>
<tr>
<td>10y</td>
<td>0.1643</td>
<td>0.1479</td>
<td>0.1334</td>
<td>0.1218</td>
<td>0.1143</td>
<td>0.1107</td>
<td>0.1099</td>
</tr>
<tr>
<td>15y</td>
<td>0.2093</td>
<td>0.1913</td>
<td>0.1756</td>
<td>0.1627</td>
<td>0.1529</td>
<td>0.1465</td>
<td>0.1429</td>
</tr>
<tr>
<td>20y</td>
<td>0.2296</td>
<td>0.2119</td>
<td>0.1968</td>
<td>0.1844</td>
<td>0.1750</td>
<td>0.1684</td>
<td>0.1646</td>
</tr>
<tr>
<td>25y</td>
<td>0.2397</td>
<td>0.2231</td>
<td>0.2092</td>
<td>0.1980</td>
<td>0.1895</td>
<td>0.1837</td>
<td>0.1802</td>
</tr>
<tr>
<td>30y</td>
<td>0.2569</td>
<td>0.2438</td>
<td>0.2217</td>
<td>0.2113</td>
<td>0.2035</td>
<td>0.1981</td>
<td>0.1948</td>
</tr>
</tbody>
</table>

Table B.2: Market implied Black volatilities for FX options as given in [Pit06]. The strikes $K_n(T_i)$ were tabulated in Table B.1.
### Table B.3: The calibration results for the FX-HHW model, in terms of the differences between the market (given in Table B.2) and FX-HHW model implied volatilities. Strikes $K_n(T_i)$ are given in Table B.1.

<table>
<thead>
<tr>
<th>$T_i$</th>
<th>$K_1(T_i)$</th>
<th>$K_2(T_i)$</th>
<th>$K_3(T_i)$</th>
<th>$K_4(T_i)$</th>
<th>$K_5(T_i)$</th>
<th>$K_6(T_i)$</th>
<th>$K_7(T_i)$</th>
</tr>
</thead>
<tbody>
<tr>
<td>6m</td>
<td>0.0012</td>
<td>-0.0012</td>
<td>-0.0025</td>
<td>-0.0023</td>
<td>-0.0001</td>
<td>0.0020</td>
<td>0.0022</td>
</tr>
<tr>
<td>1y</td>
<td>0.0013</td>
<td>-0.0008</td>
<td>-0.0018</td>
<td>-0.0009</td>
<td>0.0014</td>
<td>0.0016</td>
<td>-0.0014</td>
</tr>
<tr>
<td>3y</td>
<td>0.0016</td>
<td>-0.0007</td>
<td>-0.0017</td>
<td>-0.0008</td>
<td>0.0018</td>
<td>0.0022</td>
<td>-0.0014</td>
</tr>
<tr>
<td>5y</td>
<td>0.0011</td>
<td>-0.0006</td>
<td>-0.0012</td>
<td>-0.0007</td>
<td>0.0010</td>
<td>0.0013</td>
<td>-0.0014</td>
</tr>
<tr>
<td>7y</td>
<td>0.0007</td>
<td>-0.0003</td>
<td>-0.0006</td>
<td>-0.0003</td>
<td>0.0006</td>
<td>0.0010</td>
<td>-0.0008</td>
</tr>
<tr>
<td>10y</td>
<td>0.0004</td>
<td>-0.0001</td>
<td>-0.0001</td>
<td>-0.0002</td>
<td>0.0002</td>
<td>0.0005</td>
<td>-0.0002</td>
</tr>
<tr>
<td>15y</td>
<td>0.0011</td>
<td>-0.0005</td>
<td>-0.0009</td>
<td>-0.0004</td>
<td>0.0003</td>
<td>0.0009</td>
<td>-0.0005</td>
</tr>
<tr>
<td>20y</td>
<td>0.0009</td>
<td>0.0002</td>
<td>0.0002</td>
<td>-0.0019</td>
<td>-0.0024</td>
<td>-0.0016</td>
<td>0.0002</td>
</tr>
<tr>
<td>25y</td>
<td>0.0143</td>
<td>0.0059</td>
<td>-0.0002</td>
<td>-0.0043</td>
<td>-0.0063</td>
<td>-0.0064</td>
<td>-0.0054</td>
</tr>
<tr>
<td>30y</td>
<td>0.0165</td>
<td>0.0070</td>
<td>0.0000</td>
<td>-0.0048</td>
<td>-0.0074</td>
<td>-0.0082</td>
<td>-0.0074</td>
</tr>
</tbody>
</table>

### Table B.4: Average FX call option prices obtained by the FX-HHW model with 20 Monte-Carlo simulations, 50,000 paths and $20 \times T_i$ steps; $MC$ stands for Monte Carlo and $COS$ for Fourier Cosine expansion technique ([FO08]) for the FX-HHW1 model with 500 expansion terms. The strikes $K_n(T_i)$ are tabulated in Table B.1.
Table B.5: Average FX call option prices obtained by the FX-HLMM model with 20 Monte-Carlo simulations, 50,000 paths and 20 × $T_i$ steps; $MC$ stands for Monte Carlo and $COS$ for the Fourier Cosine expansion technique ([FO08]) for the FX-HLMM1 model with 500 expansion terms. Values of the strikes $K_n(T_i)$ are tabulated in Table B.1.

<table>
<thead>
<tr>
<th>$T_i$</th>
<th>method</th>
<th>$K_1(T_i)$</th>
<th>$K_2(T_i)$</th>
<th>$K_3(T_i)$</th>
<th>$K_4(T_i)$</th>
<th>$K_5(T_i)$</th>
<th>$K_6(T_i)$</th>
<th>$K_7(T_i)$</th>
</tr>
</thead>
<tbody>
<tr>
<td>2y</td>
<td>MC</td>
<td>0.3336</td>
<td>0.2889</td>
<td>0.2456</td>
<td>0.2046</td>
<td>0.1667</td>
<td>0.1327</td>
<td>0.1030</td>
</tr>
<tr>
<td></td>
<td>std dev</td>
<td>0.0008</td>
<td>0.0009</td>
<td>0.0010</td>
<td>0.0010</td>
<td>0.0011</td>
<td>0.0011</td>
<td>0.0012</td>
</tr>
<tr>
<td></td>
<td>COS</td>
<td>0.3326</td>
<td>0.2880</td>
<td>0.2450</td>
<td>0.2043</td>
<td>0.1667</td>
<td>0.1340</td>
<td>0.1031</td>
</tr>
<tr>
<td>3y</td>
<td>MC</td>
<td>0.3786</td>
<td>0.3299</td>
<td>0.2823</td>
<td>0.2366</td>
<td>0.1939</td>
<td>0.1553</td>
<td>0.1213</td>
</tr>
<tr>
<td></td>
<td>std dev</td>
<td>0.0006</td>
<td>0.0007</td>
<td>0.0008</td>
<td>0.0009</td>
<td>0.0011</td>
<td>0.0012</td>
<td>0.0013</td>
</tr>
<tr>
<td></td>
<td>COS</td>
<td>0.3768</td>
<td>0.3282</td>
<td>0.2808</td>
<td>0.2354</td>
<td>0.1931</td>
<td>0.1548</td>
<td>0.1212</td>
</tr>
<tr>
<td>5y</td>
<td>MC</td>
<td>0.4243</td>
<td>0.3738</td>
<td>0.3234</td>
<td>0.2743</td>
<td>0.2274</td>
<td>0.1843</td>
<td>0.1435</td>
</tr>
<tr>
<td></td>
<td>std dev</td>
<td>0.0012</td>
<td>0.0013</td>
<td>0.0014</td>
<td>0.0015</td>
<td>0.0016</td>
<td>0.0016</td>
<td>0.0016</td>
</tr>
<tr>
<td></td>
<td>COS</td>
<td>0.4222</td>
<td>0.3711</td>
<td>0.3215</td>
<td>0.2721</td>
<td>0.2260</td>
<td>0.1835</td>
<td>0.1435</td>
</tr>
<tr>
<td>7y</td>
<td>MC</td>
<td>0.4399</td>
<td>0.3914</td>
<td>0.3424</td>
<td>0.2908</td>
<td>0.2470</td>
<td>0.2031</td>
<td>0.1631</td>
</tr>
<tr>
<td></td>
<td>std dev</td>
<td>0.0013</td>
<td>0.0014</td>
<td>0.0015</td>
<td>0.0016</td>
<td>0.0018</td>
<td>0.0019</td>
<td>0.0021</td>
</tr>
<tr>
<td></td>
<td>COS</td>
<td>0.4379</td>
<td>0.3894</td>
<td>0.3402</td>
<td>0.2878</td>
<td>0.2438</td>
<td>0.2017</td>
<td>0.1621</td>
</tr>
<tr>
<td>10y</td>
<td>MC</td>
<td>0.4383</td>
<td>0.3928</td>
<td>0.3482</td>
<td>0.3031</td>
<td>0.2582</td>
<td>0.2162</td>
<td>0.1764</td>
</tr>
<tr>
<td></td>
<td>std dev</td>
<td>0.0012</td>
<td>0.0016</td>
<td>0.0019</td>
<td>0.0023</td>
<td>0.0026</td>
<td>0.0027</td>
<td>0.0028</td>
</tr>
<tr>
<td></td>
<td>COS</td>
<td>0.4338</td>
<td>0.3905</td>
<td>0.3461</td>
<td>0.3014</td>
<td>0.2576</td>
<td>0.2157</td>
<td>0.1767</td>
</tr>
<tr>
<td>15y</td>
<td>MC</td>
<td>0.3964</td>
<td>0.3632</td>
<td>0.3280</td>
<td>0.2917</td>
<td>0.2550</td>
<td>0.2186</td>
<td>0.1834</td>
</tr>
<tr>
<td></td>
<td>std dev</td>
<td>0.0008</td>
<td>0.0010</td>
<td>0.0012</td>
<td>0.0014</td>
<td>0.0016</td>
<td>0.0019</td>
<td>0.0023</td>
</tr>
<tr>
<td></td>
<td>COS</td>
<td>0.3944</td>
<td>0.3613</td>
<td>0.3265</td>
<td>0.2907</td>
<td>0.2545</td>
<td>0.2190</td>
<td>0.1848</td>
</tr>
<tr>
<td>20y</td>
<td>MC</td>
<td>0.3417</td>
<td>0.3171</td>
<td>0.2907</td>
<td>0.2629</td>
<td>0.2342</td>
<td>0.2052</td>
<td>0.1768</td>
</tr>
<tr>
<td></td>
<td>std dev</td>
<td>0.0010</td>
<td>0.0013</td>
<td>0.0015</td>
<td>0.0018</td>
<td>0.0021</td>
<td>0.0025</td>
<td>0.0030</td>
</tr>
<tr>
<td></td>
<td>COS</td>
<td>0.3416</td>
<td>0.3166</td>
<td>0.2918</td>
<td>0.2647</td>
<td>0.2367</td>
<td>0.2085</td>
<td>0.1806</td>
</tr>
<tr>
<td>25y</td>
<td>MC</td>
<td>0.2886</td>
<td>0.2715</td>
<td>0.2525</td>
<td>0.2321</td>
<td>0.2070</td>
<td>0.1867</td>
<td>0.1664</td>
</tr>
<tr>
<td></td>
<td>std dev</td>
<td>0.0011</td>
<td>0.0014</td>
<td>0.0016</td>
<td>0.0019</td>
<td>0.0023</td>
<td>0.0027</td>
<td>0.0033</td>
</tr>
<tr>
<td></td>
<td>COS</td>
<td>0.2883</td>
<td>0.2712</td>
<td>0.2532</td>
<td>0.2335</td>
<td>0.2127</td>
<td>0.1913</td>
<td>0.1697</td>
</tr>
<tr>
<td>30y</td>
<td>MC</td>
<td>0.2396</td>
<td>0.2281</td>
<td>0.2152</td>
<td>0.2011</td>
<td>0.1858</td>
<td>0.1669</td>
<td>0.1534</td>
</tr>
<tr>
<td></td>
<td>std dev</td>
<td>0.0012</td>
<td>0.0015</td>
<td>0.0018</td>
<td>0.0021</td>
<td>0.0024</td>
<td>0.0029</td>
<td>0.0035</td>
</tr>
<tr>
<td></td>
<td>COS</td>
<td>0.2393</td>
<td>0.2279</td>
<td>0.2152</td>
<td>0.2014</td>
<td>0.1866</td>
<td>0.1710</td>
<td>0.1548</td>
</tr>
</tbody>
</table>