



# Efficient Pricing of CMS Spread Options in a Stochastic Volatility Libor Market Model

Matthias Lutz

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## Outline

### Part 1: CMS Spread Option Pricing in a SV-LMM

- General Problem
- Spread option formulas
- Numerical Results

### Part 2: Extracting Correlations from the Market

- Correlation Parameterizations
- Calibration Example

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## Motivation

- ▶ Standard approach: Calibration of volatilities to caps/swaptions. Forward rate correlations are usually estimated from historical data.
- ▶ Swaptions exhibit only weak dependence on correlations.
- ▶ Even historical estimation of forward rate correlation matrices not trivial (choice of yield curve model, time horizon, backward-looking)
- ▶ Goal: Augment set of calibration instruments with CMS spread options.  
⇒ Consistent pricing of spread-related exotics.

## Model framework

### SV-LMM with time-dep. skews (Piterbarg, 2005)

$$dL_i(t) = \dots dt + [\beta_i(t)L_i(t) + (1 - \beta_i(t))L_i(0)]\sqrt{V(t)}\sigma_i(t)dW_i(t),$$

$$dV(t) = \kappa(1 - V(t))dt + \xi\sqrt{V(t)}dZ(t)$$

$$\langle dW_n(t), dW_m(t) \rangle = \rho(t, T_n, T_m)dt$$

$$\langle dW_n(t), dZ(t) \rangle = 0$$

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### Swap-rate dynamics

For the swap rates

$$S_j = \sum w_i L_i$$

we have approximately

$$dS_j(t) = \dots dt + [\beta_j(t)S_j(t) + (1 - \beta_j(t))S_j(0)]\sqrt{V(t)}\sigma_j(t)dW_j(t),$$

i.e., they are again of type 'displaced Heston'.

## Problem to be solved

For the calibration to CMS spread options we must be able to rapidly calculate expectations of the form

$$\mathbb{E}[(S_1(T) - S_2(T) - K)^+],$$

where  $S_1$  and  $S_2$  are for example a 10Y and a 2Y swap rate, respectively.

⇒ No Monte-Carlo !

## Method I: Antonov & Arneguy (2009), Hurd & Zhou (2009)

Derive the 2-dim Fourier transform  $\hat{F}$  of the payoff function

$$F(x_1, x_2) = (e^{x_1} - e^{x_2} - 1)^+$$

via complex  $\Gamma$ -function.

⇒ Price of spread option given as 2-dim inverse Fourier transform

$$\text{SprOpt}(0) = \frac{1}{2\pi} \iint_{\mathbb{R}^2 + i\varepsilon} e^{iuX'_0} \hat{F}(u_1, u_2) \Phi(u_1, u_2) d(u_1, u_2),$$

where  $\Phi$  denotes char. function of  $(\log S_1(T), \log S_2(T))$ .

However: Efficient 2-dim Fourier inversion not trivial ! (Choice of  $\varepsilon$ , highly oscillatory integrand for short maturities)



## Method II: Kiesel & Lutz (2010)

Brownian motion driving stochastic volatility is independent of the swap rate Brownian motions !

Define the integrated variance

$$\bar{V}_T := \int_0^T V_t dt.$$

$\Rightarrow (S_1(T), S_2(T)) | \bar{V}_T$  jointly log-normal (if skews = 1).

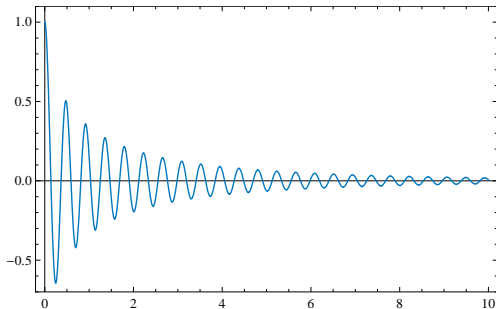
$$\begin{aligned} \Rightarrow \mathbb{E}[(S_1(T) - S_2(T) - K)^+] &= \mathbb{E}\left[\mathbb{E}[(S_1(T) - S_2(T) - K)^+ | \bar{V}_T]\right] \\ &= \int_0^\infty \int_{-\infty}^\infty g(u, v) \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}u^2} du f(v) dv, \end{aligned}$$

where  $g(u, v)$  is some 'nice' real function and  $f$  the density of the integrated variance  $\bar{V}_T$ .

## Inverse transform

Problem: Density  $f$  of  $\bar{V}_T$  only given via Fourier inversion  $\rightsquigarrow$  a further integral !

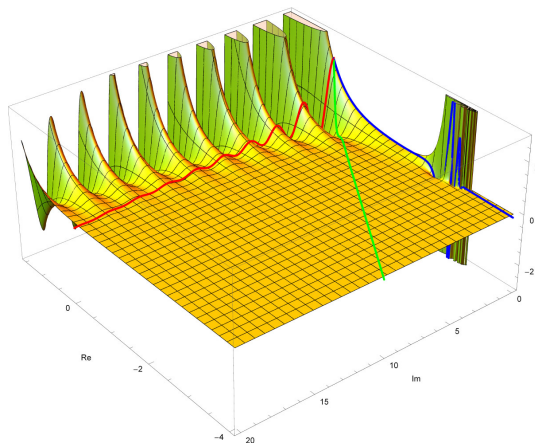
$$f(x) = \frac{1}{2\pi} \int_{-\infty}^{\infty} e^{x(a+iu)} \hat{f}(a+iu) du.$$



$\Rightarrow$  For each  $x$  at which the density is to be evaluated we need to calculate an oscillatory integral.

## Inverse transform II

Real part of the integrand in the complex plane

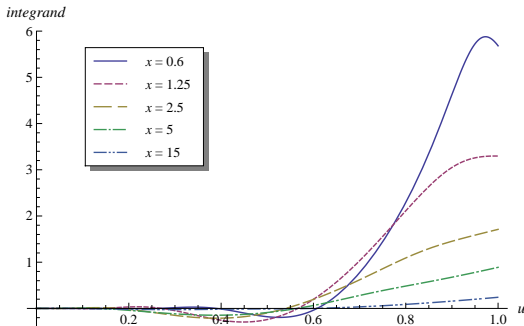


Green line: 'optimal' linear contour.

## Inverse transform III

With optimal linear contour and after transformation to finite integration interval we have

$$f(x) = -\frac{1}{\pi} \int_0^1 \operatorname{Im} \left( e^{x\tilde{s}(u)} \hat{f}(\tilde{s}(u)) \tilde{s}'(u) \right) du$$



⇒ With an adaptive integration scheme usually 20-40 sampling points are sufficient for most practical applications.

## Remarks

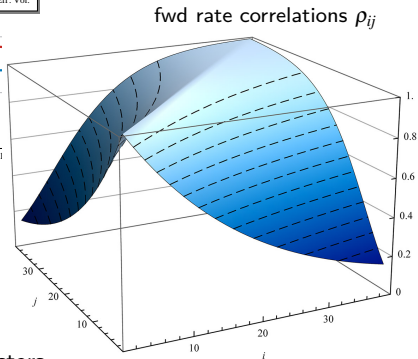
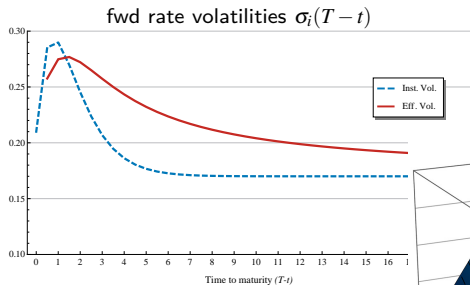
We can now rapidly calculate spread option prices for two correlated displaced Heston processes.

For application within LMM we further need

- ▶ appropriate approximations for swap rate drift terms (convexity adjustments)
- ▶ replace time-dep. swap-rate parameters with constants via parameter averaging (see Piterbarg (2005)).

→ see Kiesel & Lutz 2010.

## Numerical Results



- 20Y LMM based on 6M rates, 5 factors
- Typical upward sloping initial yield curve (from 3% to 5%)
- Skew parameters: linearly decreasing from 90% to 40%
- SV parameters:  $\kappa = 15\%$ ,  $\xi = 130\%$

## Numerical Results

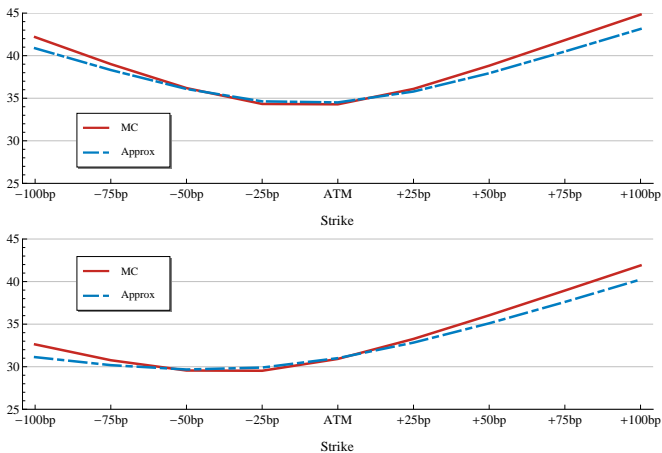


Figure: Implied normal spread volatilities in basis points for 10Y-2Y CMS spread options with 5Y maturity (top) and 10Y maturity (bottom). Computing time  $\sim 40$  ms.

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## Correlation parameterizations

- ▶ **Schoenmakers-Coffey (2003) 2-parametric form:**

$$\rho_{ij} = \exp\left(-\frac{|i-j|}{N-1}(-\log \rho_{\infty} + \eta h(i,j))\right),$$

$$\rho_{\infty} \in (0, 1), \quad \eta \in [0, -\log \rho_{\infty}]$$

- ▶ **Rebonato (2004) 3-parametric form (not necessarily pos. def.)**

$$\rho_{ij} = \rho_{\infty} + (1 - \rho_{\infty}) \exp(-\beta |i-j| \exp\{-\alpha \min(i,j)\}),$$

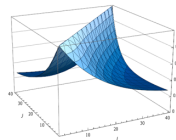
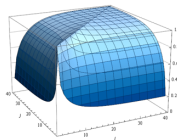
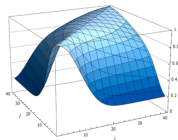
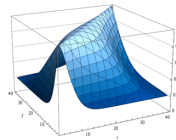
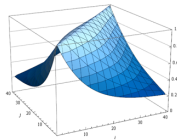
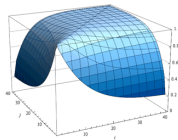
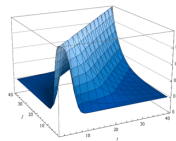
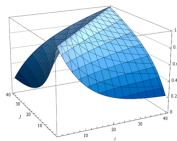
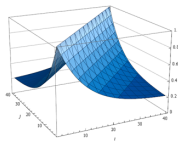
$$\beta > 0, \quad \alpha \in \mathbb{R}, \quad \rho_{\infty} \in [0, 1)$$

- ▶ **Lutz (2010) 5-parametric form:**

$$\rho_{ij} = \rho_{\infty} + (1 - \rho_{\infty}) \left[ \exp(-\beta(i^{\alpha} + j^{\alpha})) + \vartheta_{ij}(\delta, \gamma) \sqrt{(1 - \exp\{-2\beta i^{\alpha}\})(1 - \exp\{-2\beta j^{\alpha}\})} \right]$$

$$\alpha, \beta > 0, \quad \gamma, \delta \in \mathbb{R}, \quad \rho_{\infty} \in [0, 1).$$

## Lutz (2010) 5-parametric form



## Calibration example

- ▶ 30Y semi-annual LMM, 10 factors.
- ▶ Piecewise const. forward rate volatility and skew functions.
- ▶ EUR market data as of 01/14/2008.
- ▶ Calibration targets: 9 caplet smiles, 36 swaption smiles, 7 CMS spread option (10Y-2Y) implied volatilities.
- ▶ Computing time: 1:54 min.

	RMSE	
	ATM	Smile
caplets (Black-vol)	0.24%	0.48%
swaptions (Black-vol)	0.25%	0.31%
	$K = 0.25\%$	Smile
CMSSOs (bp vol)	0.7	2.4

## Calibration example

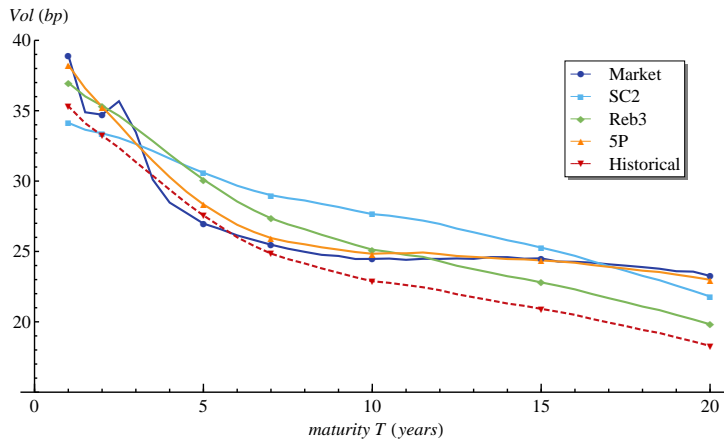
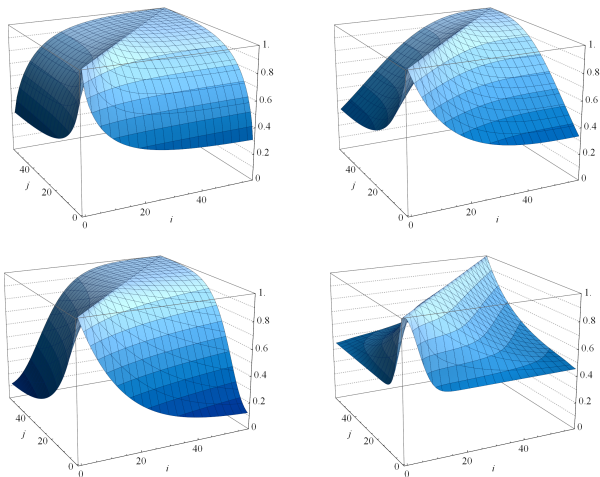


Figure: 10Y-2Y CMS spread option implied volatilities for  $K = 0.25\%$ .

## Calibration example



**Figure:** Calibrated correlation matrices. From top left to bottom right: 5P fitted to historical matrix, SC2, Reb3, 5P.

## References



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Thank you for your attention !