

Consistent Variance Curve Models

Hans Buehler

Deutsche Bank GME Quantitative Products Analytics, London
Technical University, Berlin

<http://www.math.tu-berlin.de/~buehler/> for papers etc
hans.buehler@db.com

Bachelier Congress 2006 Tokyo, August 20th



Deutsche Bank





Volatility as a traded asset

Introduction

- We are interested generally on pricing options on *realised variance* over business days $T_1=t_0 < \dots < t_n=T_2$,

$$RV(T_1, T_2) := \sum_{i=1}^n \left(\log \frac{S_{t_i}}{S_{t_{i-1}}} \right)^2$$

- We simplify and assume that realized variance is in effect the quadratic variation of the log of the stock price $S=(S_t)_t$ to the holder.
 - Quadratic variation is an unbiased estimator of the actual „realized variance“, c.f. Barndorff-Nielsen et al (2004), i.e.

$$\left\langle \log S_{\bullet} / S_{T_1} \right\rangle \approx RV(T_1, T_2)$$



Volatility as a traded asset

Introduction

- The basic product on realized variance is a *Variance Swap*.
 - The (zero strike) variance swap for the period $[T_1, T_2]$ pays the realized variance in this period to the holder.
 - We assume that variance swaps are liquidly traded. Their price process is denoted as

$$V(T_1, T_2) \equiv \left(V_t(T_1, T_2) \right)_{0 \leq t \leq T_2}$$

$$V(T) := V(0, T)$$

Observation time t

- At any time t also define the „floating maturity“ variance swaps for the period $[t, t+x]$ („Musielá-Parametrization“ for variance swaps)

$$U_t(x) := V_t(t, t+x)$$



Volatility as a traded asset

Exotic options on variance

- Since variance swaps are liquid, we do not need to model them.
 - They become input into our model.
- Products we are interested in
 - Vanilla options on variance

$$\left(RV(0, T) - kV_0(T) \right)^+$$

- Forward-started options on variance

$$\left(RV(T, T + \Delta T) - kV_T(T, T + \Delta T) \right)^+$$

- Options on variance swaps (c.f. also options on VIX)

$$\left(V_T(T, T + \Delta T) - kV_0(T, T + \Delta T) \right)^+$$

Price of the
variance swap
for $[T, T + \Delta]$ at
time T

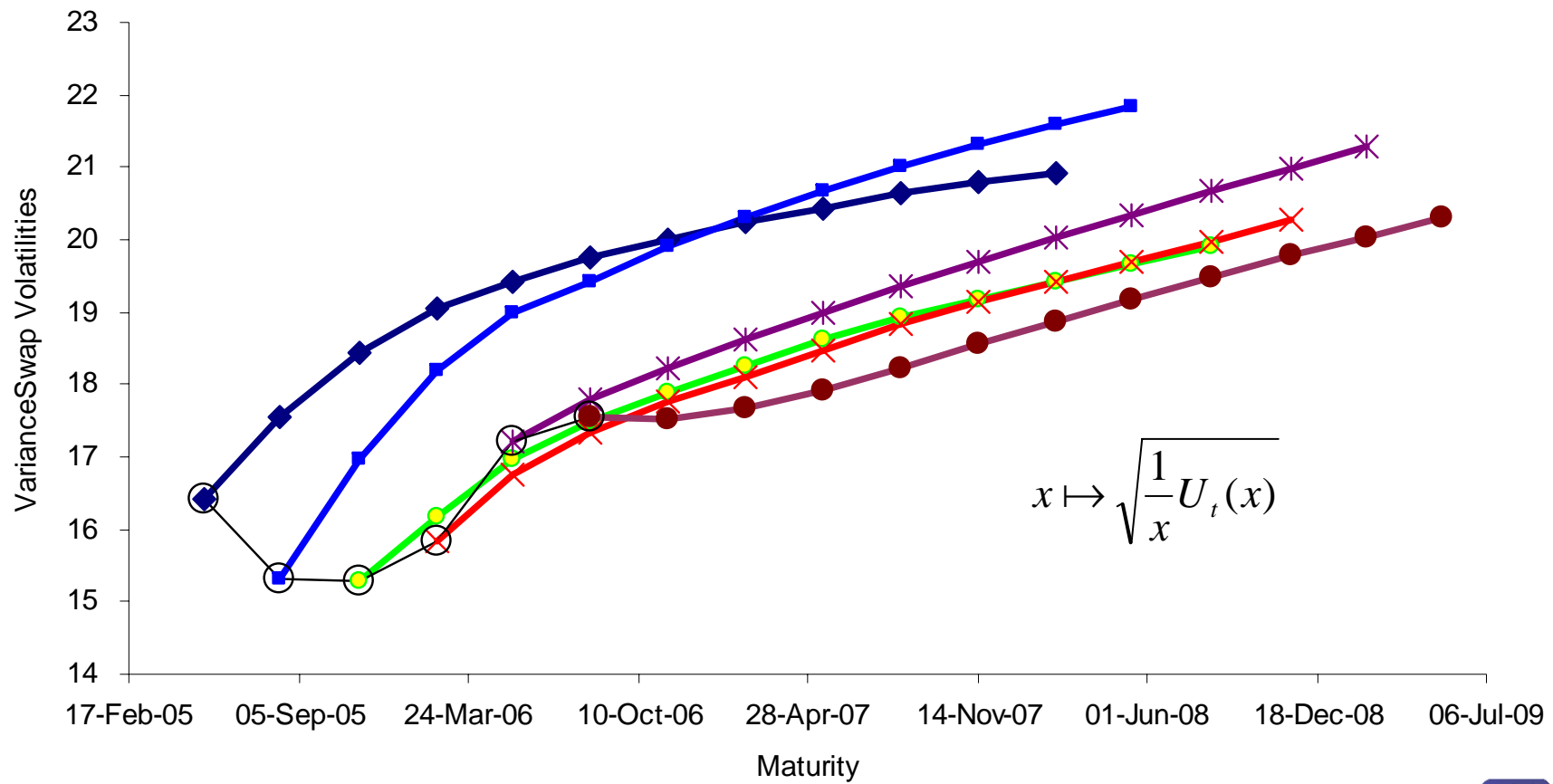
Price of the
variance
swap for
 $[T, T + \Delta]$ today



Variance Curve Models

... a picture of the market

.GDAXI Variance Swap Volatilities (3m)





Options on Variance

Task list

- The idea is to develop a pricing model that describes the entire variance swap term structure.
 - Interest-rate type problem.
- Minimum requirements
 - Free of arbitrage.
 - Allows hedging of exotic options with variance swaps (ie, complete market).
 - Allows to model a corresponding stock price process.
 - Can be implemented.
- Note:
We do not take into account European options on the underlying asset at this stage.



Variance Swap Curve Models



Variance Curve Models

Modelling

- Assume $(\Omega, \mathcal{P}, \mathcal{F})$ is a filtered probability space with d -dimensional extremal Brownian motion W .
- The idea is, to model the curve U as stochastic variable. The first problem is, that we need $x \rightarrow U_t(x)$ to be monotonic increasing for all times t . Hence, we concentrate rather on *forward variance*

$$u_t(x) := \partial_x U_t(x)$$

- Similar to forward rates in interest rate theory.
- Note that forward variance cannot be negative, but might well be zero.



Variance Curve Models

Modelling

- Idea 1: General HJM approach

Let $u = (u(x))_{x \geq 0}$ be a family of processes $u(x) = (u_t(x))_{t \geq 0}$ which are given as

$$du_t(x) := a_t(x)dt + \sum_{j=1, \dots, d} b_t^j(x) dW_t^j$$

(a and b are some suitably integrable and previsible processes).

- Problem 1:

When is u a „*variance curve model*“ in the sense that we can define a local martingale S , such that the market with S and the variance swaps is free of arbitrage?

- Answer (cf later): essentially if $v(T)$ with $v_t(T) := u_t(T-t)$ is a local martingale, i.e.

$$a_t(x) = \partial_x u_t(x)$$



Consistent Variance Curve Models

Next step

- However, HJM-type processes are by far too general:
 - It is much more natural to model u via a finite-dimensional Markov-process (in particular if we want to actually implement such a model).
 - Let us assume we write u as a function G of an m -dimensional Diffusion $Z=(Z_t)_{t \geq 0}$,

$$u_t(x) := G(Z_t; x).$$

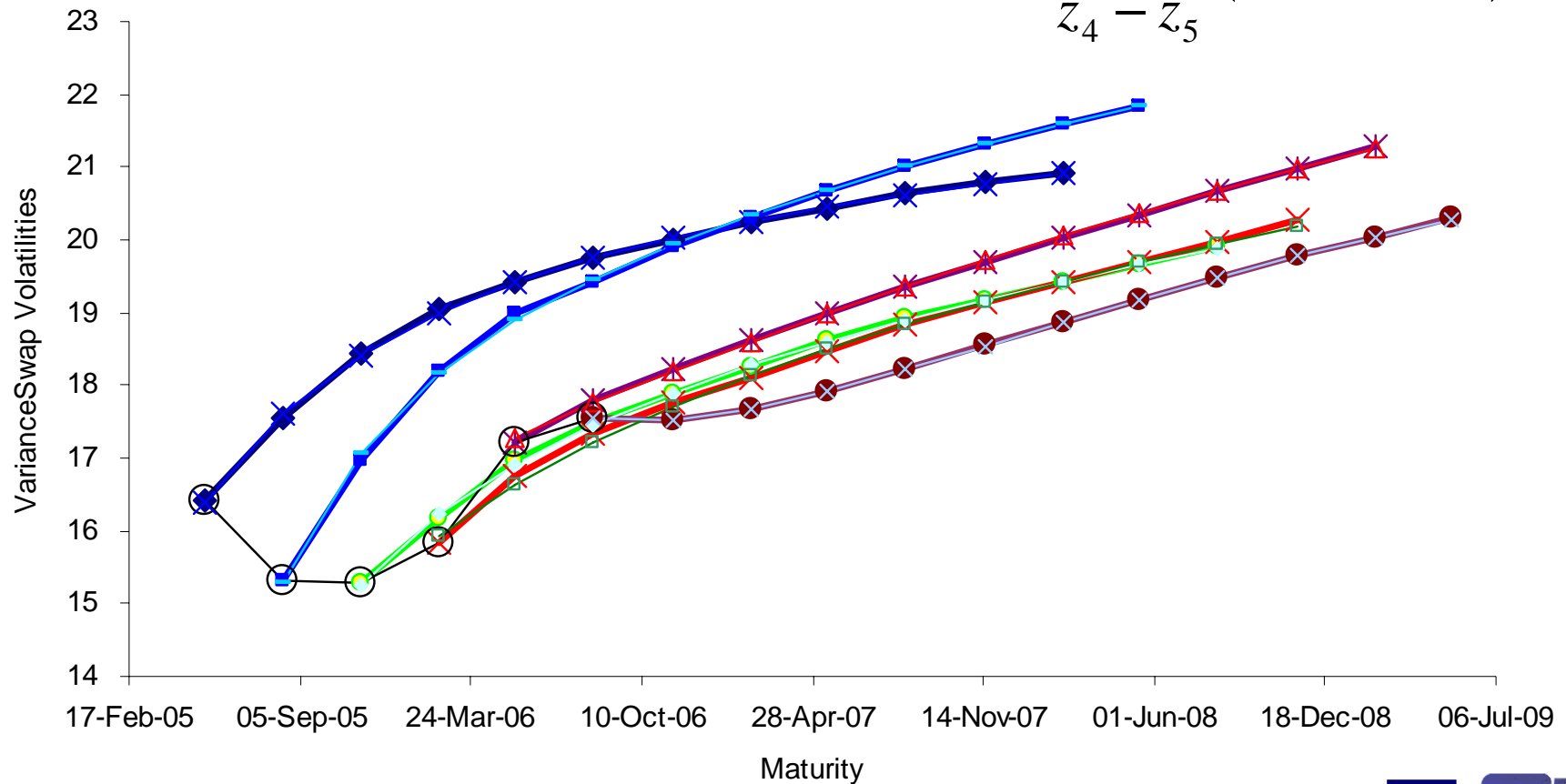
- Problem 2:
Given a “nice” function G , which diffusion Z is “consistent” with G in the sense that $G(Z; \cdot)$ is a variance curve model ?
 - Consistency problem a’la Bjoerk et al for interest rates.



Consistent Variance Curve Models

Fitting a function to the market

$$G(z; x) = z_3 + (z_1 - z_3)e^{-z_4x} + (z_2 - z_3) \frac{z_4}{z_4 - z_5} (e^{-z_5x} - e^{-z_4x})$$





Consistent Variance Curve Models

Definition

■ Definition

We call a non-negative $C^{2,2}$ -Function $G:D \times \mathbb{R}_{\geq 0} \rightarrow \mathbb{R}_{\geq 0}$ a *variance curve functional* if

$$\int_0^T G(z; x) dx < \infty$$

for all T and $z \in D$ where D is an open set in $\mathbb{R}_{\geq 0}^m$.

- We denote by Ξ the set of coefficients $C=(\mu, \sigma)$ for which the SDE

$$dZ_t = \mu(Z_t)dt + \sum_{j=1, \dots, d} \sigma^j(Z_t) dW_t^j$$

has a unique, non-explosive m -dimensional strong solution for each starting point $Z_0 \in D$. We call such a process *parameter process*.

- Note that time-dependent coefficients are part of this definition.



Consistent Variance Curve Models

Definition

- Definition

We call $C=(\mu,\sigma)\in\Xi$ *consistent* with G if

$$u_t(x) := G(Z_t; x)$$

defines for all $Z_0 \in D$ a variance curve model. Then, (C, G) is called a *consistent variance curve model*.

- Proposition

This is the case, iff Z remains in D and if the “heat equation”

$$\partial_x G(z; x) = \mu(z) \partial_z G(z; x) + \frac{1}{2} \sigma^T \sigma(z) \partial_{zz}^2 G(z; x)$$

is satisfied (note that this is a condition on $C=(\mu,\sigma)$ and not on G).



Consistent Variance Curve Models

Associated stock price process

- The associated price process and the Markov-property

Given a consistent variance curve model $C=(\mu, \sigma) \in \Xi$ and a “correlation function” $\rho: \mathbb{R}_{\geq 0} \times D \rightarrow [-1, 1]^d$ with $|\rho|_2 = 1$, define

$$dS_t = \sum_{j=1, \dots, d} S_t \rho^j(S_t; Z_t) \left\{ \sqrt{G(Z_t; 0)} dW_t^j \right\}$$

- The solution S for this SDE always exists, is a local martingale and the system (S, Z) is Markov.
 - The Markov-property allows discussion of completeness.
 - The correlation function controls the skew of the implied vol.
 - Note that the freedom of choosing ρ might be limited if S is to be a true martingale.
- The “short variance” of the stock price is

$$G(Z_t; 0)$$



Consistent Variance Curve Models

Absence of arbitrage

■ Theorem

For each variance curve model u and any correlation function ρ , the market (S, V) of the associated stock price and the variance swaps

$$V = V(T)_{T \geq 0}$$

$$V_t(T) := G(Z_t; T - t) + \int_0^t G(Z_s; 0) ds.$$

\uparrow
 $U_t(T-t)$

\nwarrow
Running variance

is free of arbitrage.

- The simplicity of this condition is a strong advantage to related approaches such as „Stochastic Implied Volatility“.



Hedging

General Markovian markets

■ Setting

Assume we have a market with K tradable instruments $X=(X^1, \dots, X^K)$ and an additional finite variation process $A=(A^1, \dots, A^L)$, such that $Y=(X,A)$ is the unique, strong non-explosive solution of

$$\begin{cases} dX_t &= \sum_{j=1, \dots, d} \Omega^j(X_t, A_t) dW^j \\ dA_t &= C(X_t, A_t) dt \end{cases}$$

- The process A can encode time or, for us, running variance.

■ Problem 3

When is the market of all “hedging-relevant” payoffs on Y complete, i.e. the market of all F^Y -measurable non-negative L^1 random variables ?



Hedging

General Markovian markets

- We say that Y “weakly preserves smoothness”, if

$$PF(t; x, a) := E[F(X_t, A_t) | X_0 = x, A_0 = a]$$

maps smooth F with compact support into C^1 -functions in x .

- Theorem (“Delta-hedging works”):
If the above holds, then the market of all “hedging-relevant” payoffs is complete.
- Proposition:
In the case that Ω und C are differentiable with locally Lipschitz derivatives, this holds.
In particular, the volatility matrix may not have full rank.
 - Open question: can we relax this assumption further?



Hedging

Consistent variance curve models

- This can be used for consistent variance curve models

$$dZ_t = \mu(Z_t)dt + \sum_{j=1, \dots, d} \sigma^j(Z_t) dW_t^j$$

$$dA_t := G(Z_t; 0)dt$$

- We need an additional invertibility criterion on G so that we can recover the state Z locally from the market.



Hedging

Consistent variance curve models

- Then, an option

$$H_t := E \left[H \left(\int_0^T \zeta_s ds \right) \mid \mathcal{F}_t \right]$$

can be replicated with variance swaps, i.e. “VarSwapDeltas” v^1, \dots, v^m exist such that

$$dH_t = \sum_{k=1}^{m'} v_t^k dV_t(T_k)$$

- If H_t is given as a differentiable function $H_t = h(t; V_t, A_t)$, then we have

$$v_t^k = \partial_{V^k} h(t; V_t, A_t)$$



Applications



Consistent Variance Curve Models

Double Heston

- The example we had before

$$G(z; x) = z_3 + (z_1 - z_2)e^{-\kappa x} + (z_2 - z_3) \frac{\kappa}{\kappa - c} \left(e^{-cx} - e^{-\kappa x} \right)$$

- The parameters κ and c must be constant to avoid arbitrage.
- A consistent factor model for this G must have the form

$$dZ_t^1 = \kappa(Z_t^2 - Z_t^1)dt + \sigma_1(Z_t)dW_t$$

$$dZ_t^2 = c(Z_t^3 - Z_t^2)dt + \sigma_2(Z_t)dW_t$$

$$dZ_t^3 = \sigma_3(Z_t)dW_t$$

Note that non-negativity must be ensured.



Consistent Variance Curve Models

Double Heston

- This works

$$\begin{aligned}dZ_t^1 &= \kappa(Z_t^2 - Z_t^1)dt + \nu Z_t^{1\alpha} d\hat{W}_t^1 \\dZ_t^2 &= c(Z_t^3 - Z_t^2)dt + \mu Z_t^{2\beta} d\hat{W}_t^2 \\dZ_t^3 &= \eta Z_t^3 d\hat{W}_t^3\end{aligned}$$

for a correlated vector of Brownian motions and $\alpha, \beta > 1/2$.*

- Complete model where S and the variance swaps are true martingales.

- Calibration strategy

- First fit the functional G , i.e. find Z_0 .
- The use the European prices to infer the correlation/volatility structure using Monte-Carlo calibration.
- Pretty expensive.

* In fact, instead of x^α we use $(x+\varepsilon)^\alpha - \varepsilon^\alpha$.



Consistent Variance Curve Models

Double Heston

■ Alternative

$$\begin{aligned}dZ_t^1 &= \kappa(Z_t^2 - Z_t^1)dt + v\sqrt{Z_t^1}d\hat{W}_t^1 \\dZ_t^2 &= c(Z_t^3 - Z_t^2)dt + \mu\sqrt{Z_t^2}dW_t^2 \\Z_t^3 &= \text{const}\end{aligned}$$

with W^2 independent of W^1 and stock price Brownian motion.

- Affine model.
- Fourier-pricing of Europeans via Riccati equations.



Consistent Variance Curve Models

Fitting the market

- A good functional may fit well, but probably never perfect.
 - We need “post-fitting”.
- Assume that $(C=(\mu,\sigma), G)$ is a consistent pair and assume that we observe in the market a short variance curve $m_0(x)$.

Then,

$$u_t(x) := \frac{m_0(t+x)}{G(Z_0; t+x)} G(Z_t; x) \equiv \frac{m_0(t+x)}{\mathbb{E}[\tilde{u}_t(x)]} \tilde{u}_t(x)$$

“fits the market” in the sense that

$$\mathbb{E}[v_T(T)] = \mathbb{E}[u_T(0)] = m_0(T)$$



Consistent Variance Curve Models

Fitting the market

- For the affine example before, the result is still affine and can be treated with Ricatti.
- Another example (Dupire)

$$G(z; x) = \exp \left\{ z e^{-kx} + \frac{s^2}{4} \left(\frac{1 - e^{-2kx}}{2k} \right) \right\}$$

- Log-normal short variance fitted to the variance swap market.
- Two-factor version cf. Bergomi.



Consistent Variance Curve Models

Conclusions

- The framework
 - Provides means to ensure absence of arbitrage and market completeness.
 - Concept of “consistency” is a desirable property from a conceptual point of view (in Buehler 2006, we also discuss applications of variance curves in Hilbert spaces a’la Filipovic/Teichmann).
 - From a practical point of view, “post-fitting” to the market ensures that we fit the variance swaps perfectly.

- To keep in mind
 - Real markets price in jump risk in the underlying (there is an upward implied vol slope in options on realised variance).
 - When pricing options, use “proper” realized variance, not quadratic variation!

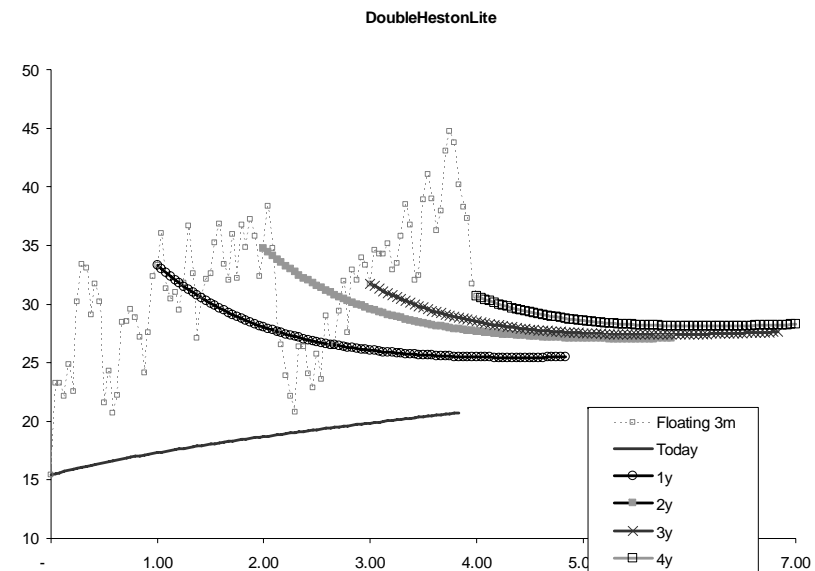
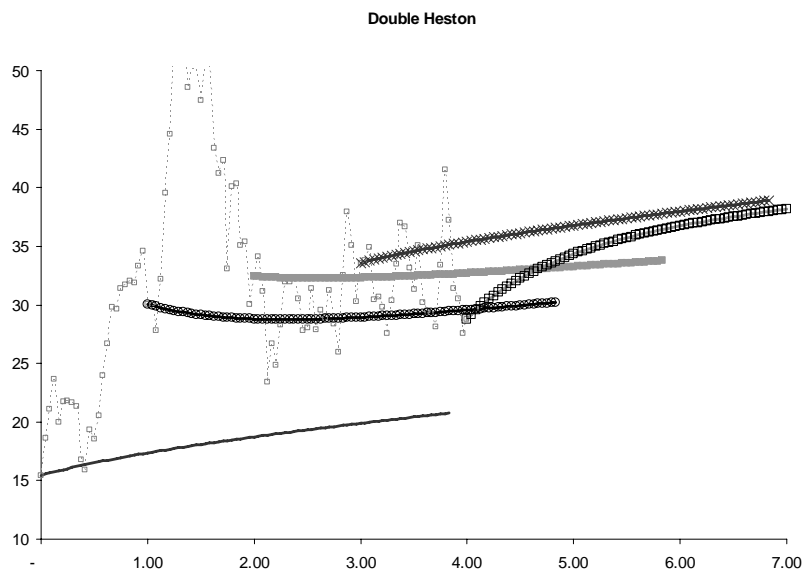


Thank you very much for your attention.
<http://www.math.tu-berlin.de/~buehler/>



Consistent Variance Curve Models

Double Heston – effect of stochastic “mean reversion level”



The long term level of the variance swap curve can move in the three-factor model.
The right hand side shows the one-factor version.

(May 2 data)