

Options On Variance: Pricing And Hedging

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Options On Variance: Pricing And Hedging

Introduction

- This talk is on pricing options on *realised variance*, which is defined over business days $T_1=t_0 < \dots < t_n=T_2$ as

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

- The scaled quantity $1/n RV(T_1, T_2)$ is an unbiased estimator for the variance in $[T_1, T_2]$ if we neglect any drift.
- Second-order Taylor on the log contract gives

$$\frac{1}{2} RV(T_1, T_2) \approx \sum_{i=1, \dots, n} \frac{1}{S_{t_{i-1}}} (S_{t_i} - S_{t_{i-1}}) - \log \frac{S_{T_2}}{S_{T_1}}$$

Delta-Hedging



Static position in Log-contracts

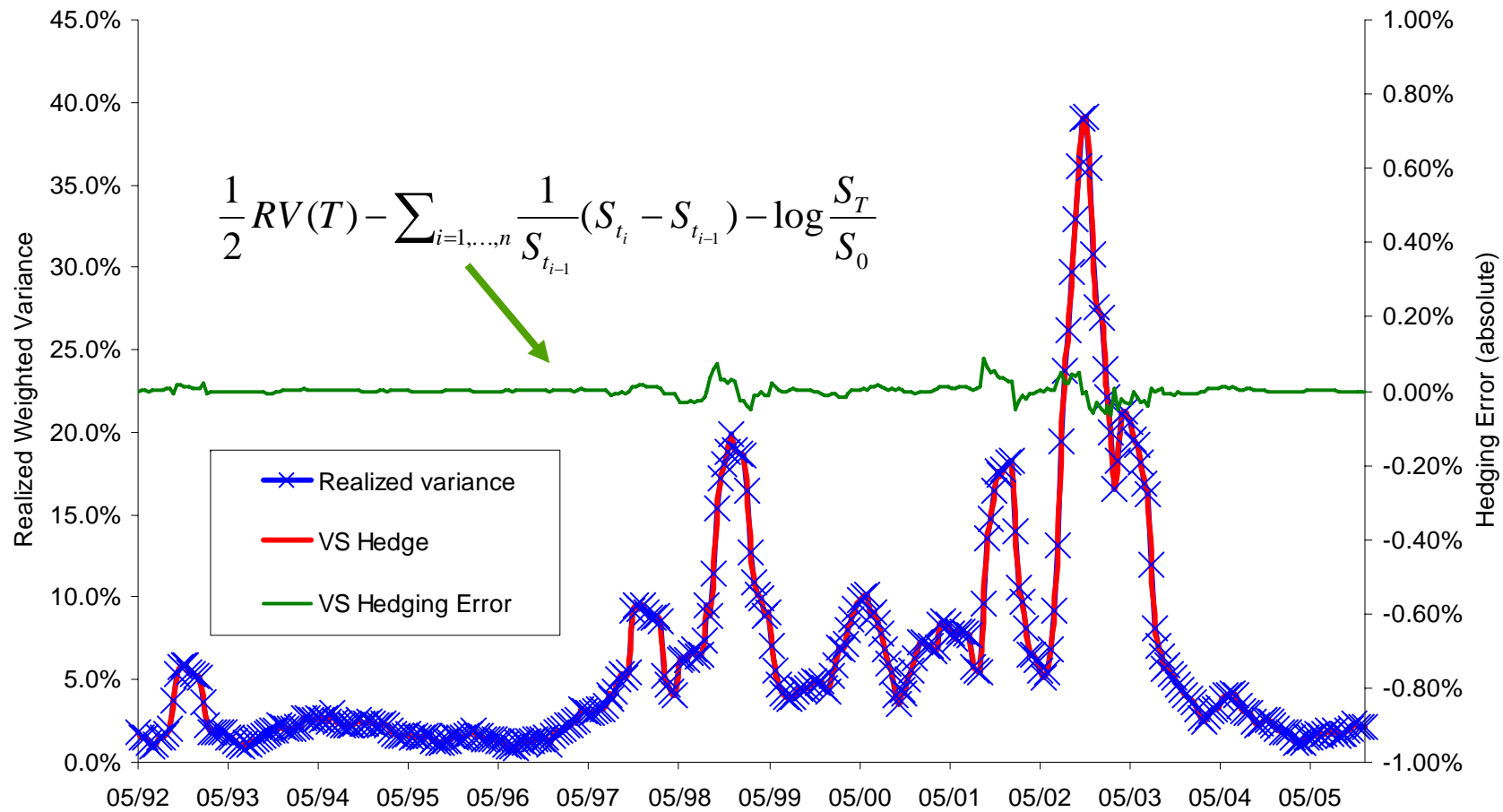




Options On Variance: Pricing And Hedging

Introduction

.STOXX50E Realized Variance Hedge (90 days)





Options On Variance: Pricing And Hedging

Introduction

- The purpose of this talk is to line out ways to evaluate non-linear payoffs on realized variance

- Vanilla options on variance (*)

$$\left(RV(T_1, T_2) - K^2\right)^+, \left(\sqrt{RV(T_1, T_2)} - K\right)^+$$

- Forward started options and options on volatility indices (VIX, VDAX_{new})

$$\left(RV(T_1, T_2) - kV_{T_1}(T_1, T_2)\right)^+$$

- „Hybrids“

$$\left(RV(T_1, T_2) - V_{T_1}(T_1, T_2)\right)^+ 1_{S_{T_1} < B}$$

$V_{T_1}(T_1, T_2)$: Market value of $RV(T_1, T_2)$ at time T_1 (\approx variance swap)

(*) All payoffs in this presentation are in “natural” scale. Business day adjustments might be applied in practise.



Options On Variance: Pricing And Hedging

Questions

- Realized variance:
Dividends, defaults and its relation to quadratic variation.
- Pricing:
Common approaches
- Hedging



Realized Variance



Realized Variance

Impact of dividends

- Assume that rates are zero and that the stock pays at dividend dates $0 < \tau_1 < \dots < \tau_N$ cash dividends Δ_i . Across a dividend date we have

$$S_{\tau_i} = S_{\tau_i^-} - \Delta_i$$

- Bermudez et al (2006): since the stock price can never be below the PV of any certain future cash dividends, we can write S in terms of a martingale X as

$$S_t = \underbrace{\left(S_0 - \sum_i \Delta_i \right)}_{S_0^*} X_t + \sum_{k:t_k > t} \Delta_k$$

Adjusted “volatile” spot level. →

→ “Pure martingale” part

← Stock floor due to future dividends

- Extension to deterministic rates, proportional dividends and credit risk straight forward, Bermudez et al (2006).



Realized Variance

Impact of dividends

- For realized variance, we obtain

Realized variance of the “continuous part”
Note that this is *not* the same as the “proportional dividend” RV.

$$RV(T_1, T_2) = RV_{cont}(T_1, T_2) + \sum_{k: T_1 < \tau_k \leq T_2} \log^2 \frac{S_{\tau_i}}{S_{\tau_i^-}}$$
$$= RV_{cont}(T_1, T_2) + \sum_{k: T_1 < \tau_k \leq T_2} \log^2 \frac{S_{\tau_i}}{S_{\tau_i} + \Delta_i}$$

- There is additional variance arising from the dividends.
- For single stocks, dividends are usually taken out.
- What happens to variance swaps ...?



Realized Variance

Impact of dividends on variance swaps

- Assume that the stock price is continuous between dividends, e.g.

$$\frac{dX_t}{X_t} = \sqrt{v_t} dB_t$$

■ Then,

$$d \log S_t \approx \frac{1}{S_{t-}} dS_t - \frac{1}{2} dRV_{cont} + \log \frac{S_t}{S_t + \Delta_t}$$

Log-contract (points to $d \log S_t$)
Continuous part of RV (points to dRV_{cont})
Delta-Hedge (points to $\frac{1}{S_{t-}} dS_t$)
Additional contribution from Dividends (points to $\log \frac{S_t}{S_t + \Delta_t}$)

- Variance swap pricing needs to take into account dividends, too.
- Note that delta-hedging strategy still works very well.



Realized Variance

A note on default risk

- Variance swaps on single stocks incorporate a „cap“ on realized variance. Hence their payoff is

$$\min\{RV(T_1, T_2), 250\% V_0(T_1, T_2)\} - V_0(T_1, T_2)$$

- The value of this cap is zero in a diffusion framework.
- To incorporate default risk, a simple way is to say

$$(RV(T_1, T_2) - V_0(T_1, T_2))1_{\text{no default}}$$

+

$$(250\% - 1)V_0(T_1, T_2)1_{\text{default}}$$

- Can be implemented very easily.
- But mind effect on pricing via log-contract.



Realized Variance

Relation to quadratic variation

- We know that realised variance is an unbiased estimator of the *quadratic variation* of the returns of S ,

$$QV(T_1, T_2) := \langle \log S_{T_2} \rangle - \langle \log S_{T_1} \rangle$$

- However, care must be taken if a non-linear payoff on realized variance is priced.
 - See also Barndorff-Nielsen et al (2005).



Realized Variance

Relation to quadratic variation

- Jordinson (2006): assume a Black & Scholes model with constant vol σ .
- Then,

$$RV(0, T) = \sum_{i=1}^n \frac{\sigma^2 T}{n} \left(x_i - \frac{1}{2} \frac{\sigma^2 T}{n} \right)^2 = \sigma^2 T \frac{1}{n} \sum_{i=1}^n \left(x_i - \frac{1}{2} \frac{\sigma^2 T}{n} \right)^2$$

where $(x_i)_i$ are independent standardnormal. Hence,

$$\frac{RV(0, T)}{QV(0, T)} \sim \chi^2(n, \lambda) \quad \lambda = \sigma^4 T^2 / n$$

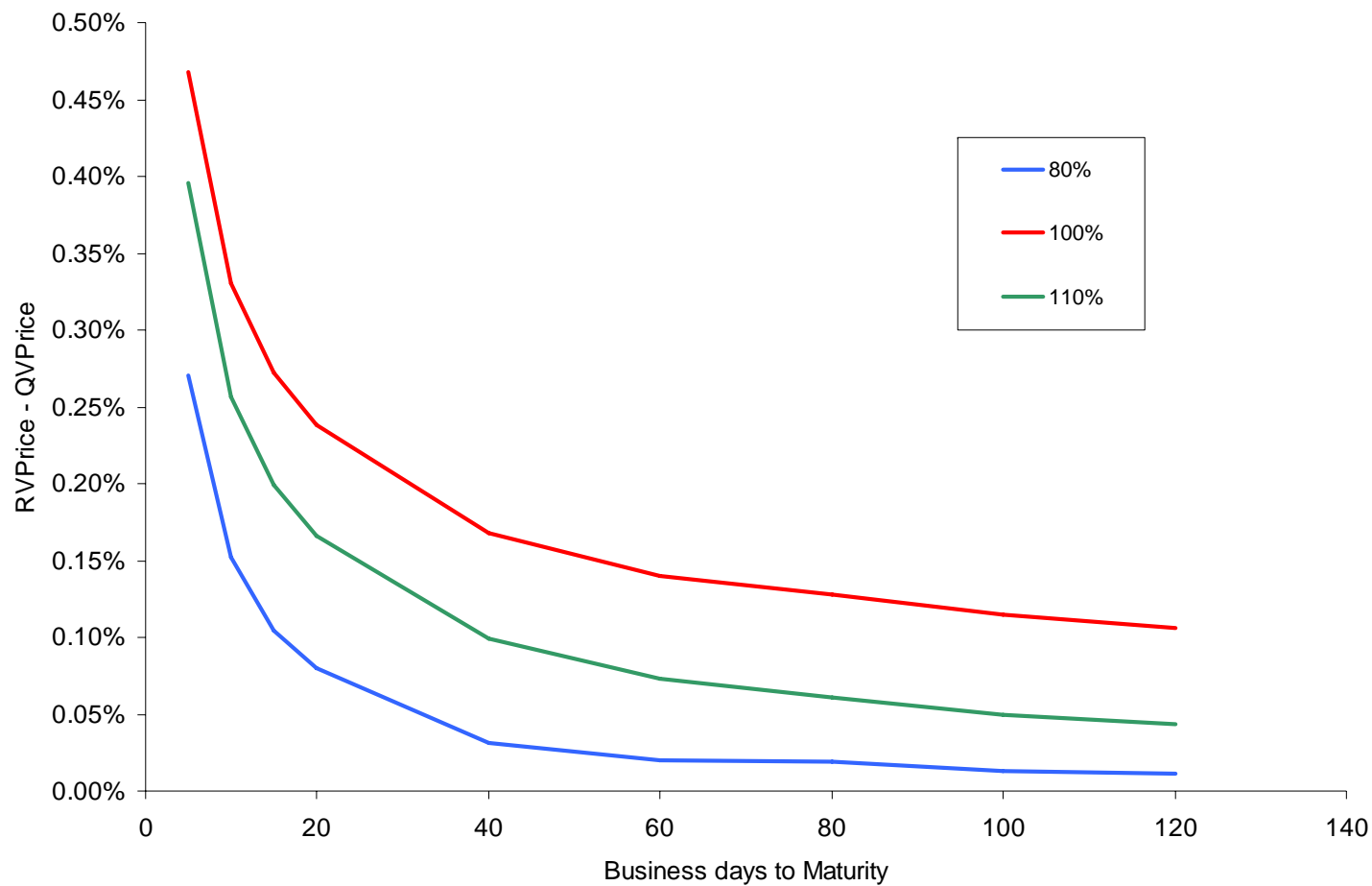
- Hence, the price of an option on realized variance is non-trivial even in BS.
- Above approximation allows direct pricing in BS.



Realized Variance

Relation to quadratic variation

Example difference of prices of options on RV vs QV in Black-Scholes

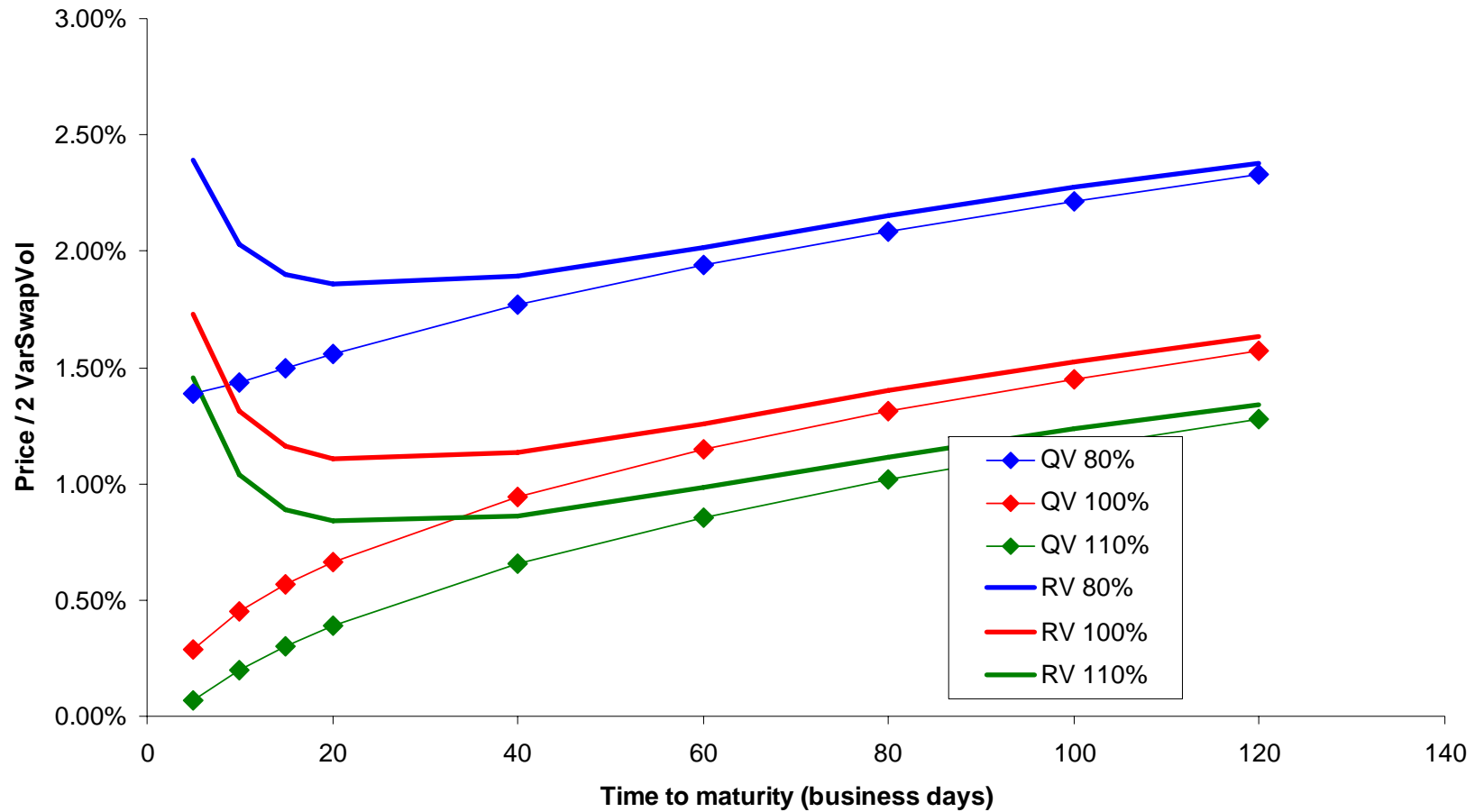




Realized Variance

Relation to quadratic variation

Options on Quadratic Variation vs. Realized Variance in a Heston model





Modelling Realized Variance



Quadratic Variation

Modeling

- Despite the previous finding, it is common practise to model QV instead of RV.
 - As long as we use RV to *price* options on variance, no contradiction.

- Modelling QV - various approaches:
 - Direct modelling of the distribution of QV.
 - Consistent variance swap HJM-type term-structure models.
 - Hybrid term-structure modelling with specific stock price distribution
 - “Fitted” models



Quadratic Variation

Direct Models

- Use the marginal distribution F_T of $QV(0,T)$ directly.

$$RV(0,T) \approx QV(0,T) \sim \text{distribution } F_T$$

- Current approaches to infer the distribution of QV rely on a symmetric smile, in which case we have

$$\text{Call}(T, K) = \int_0^{\infty} \text{BSCall}(K, V) dF_T(V)$$

- Solve directly via numerical inversion of integral equation on discrete grid.
- Carr/Lee (2004) use power payoffs to recover F_T .



Quadratic Variation

Direct Models

■ Good

- No input other than market information required.
- Pricing vanilla options simple and very quick.
- Greeks w.r.t. variance swaps straight forward.

■ But

- Publicly known methods rely on symmetric smile (FX etc)
- $RV \neq QV$
- Maturity dependent; no link between different maturities per se.

- Note: information encoded in European option prices equivalent to Dupire's local volatility → base model for lower bounds for options on variance (that's a conjecture!).



Quadratic Variation

Consistent Variance Curve Models

- HJM-type “consistent term structure models” Buehler (2006)
 - Based on consistent HJM interest rate theory by Bjoerk/Filipovic/Teichman and others.
 - Attempt to model the full term structure movements of variance swaps and to use the latter as natural hedging instruments.

- Notation:

- A variance swap price (ex scaling) is denoted by

$$V_t(T_1, T_2) := E_t[RV(T_1, T_2)]$$

- A more convenient object is the “floating maturity” variance swap

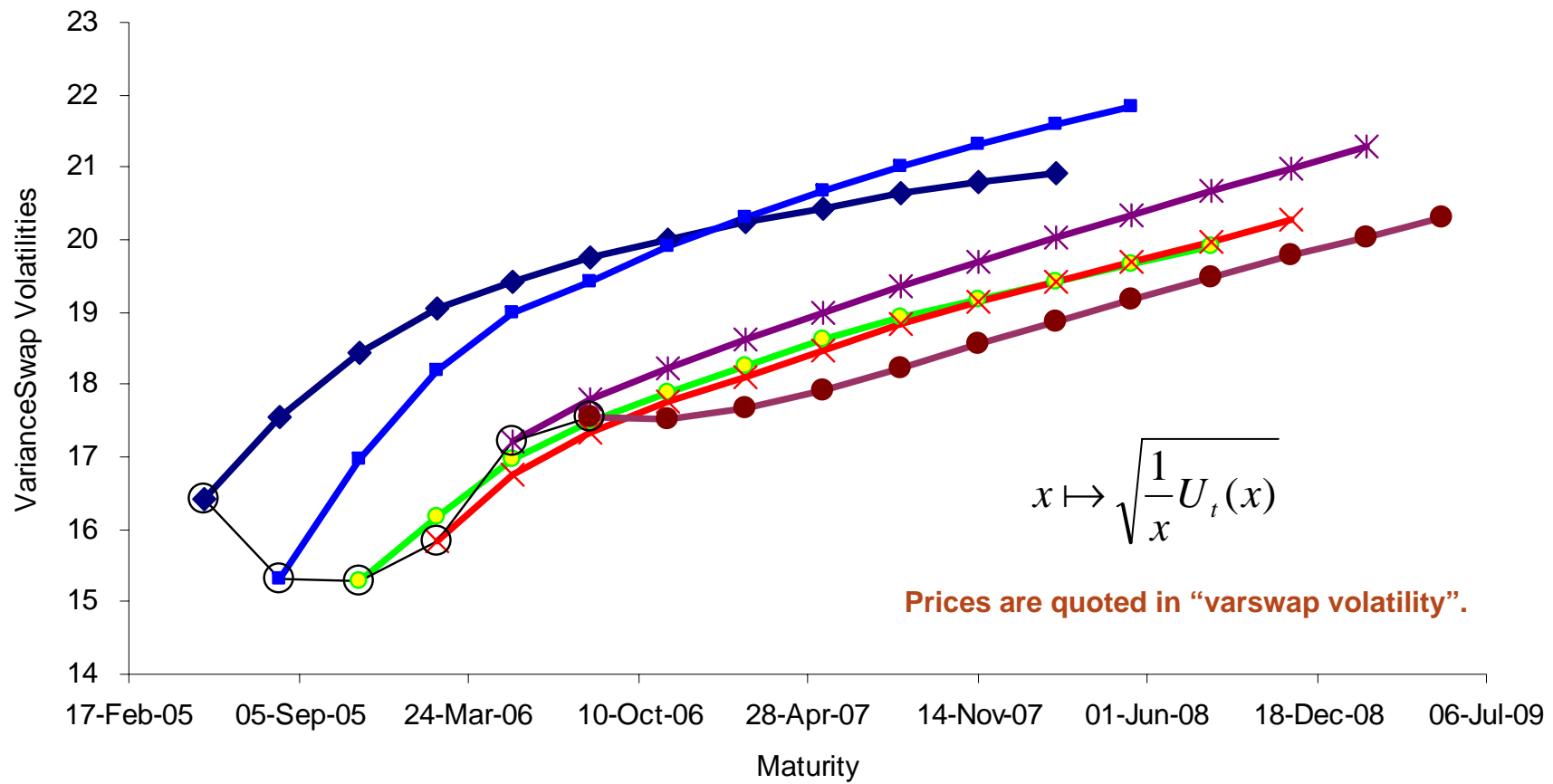
$$U_t(x) := V_t(t, t + x)$$



Quadratic Variation

Consistent Variance Curve Models

.GDAXI Variance Swap Volatilities (3m)





Quadratic Variation

Consistent Variance Curve Models

- “Consistency” means that the variance curves U_t belong to some finite-dimensional function class (“finite dimensional representation”).
 - This ensures controlled term-structure movements.
- Hence, write $U(x)$ as a function of a low-dimensional parameter process Z .
 - Since U as a function of x must be increasing, it is easier to write its derivative as non-negative function G of Z , i.e.

$$U_t(x) \equiv \int_0^x G(y; \bar{Z}_t) dy$$

G is the “variance curve function” and describes a “typical” shape of the implied variance swap curve.

with

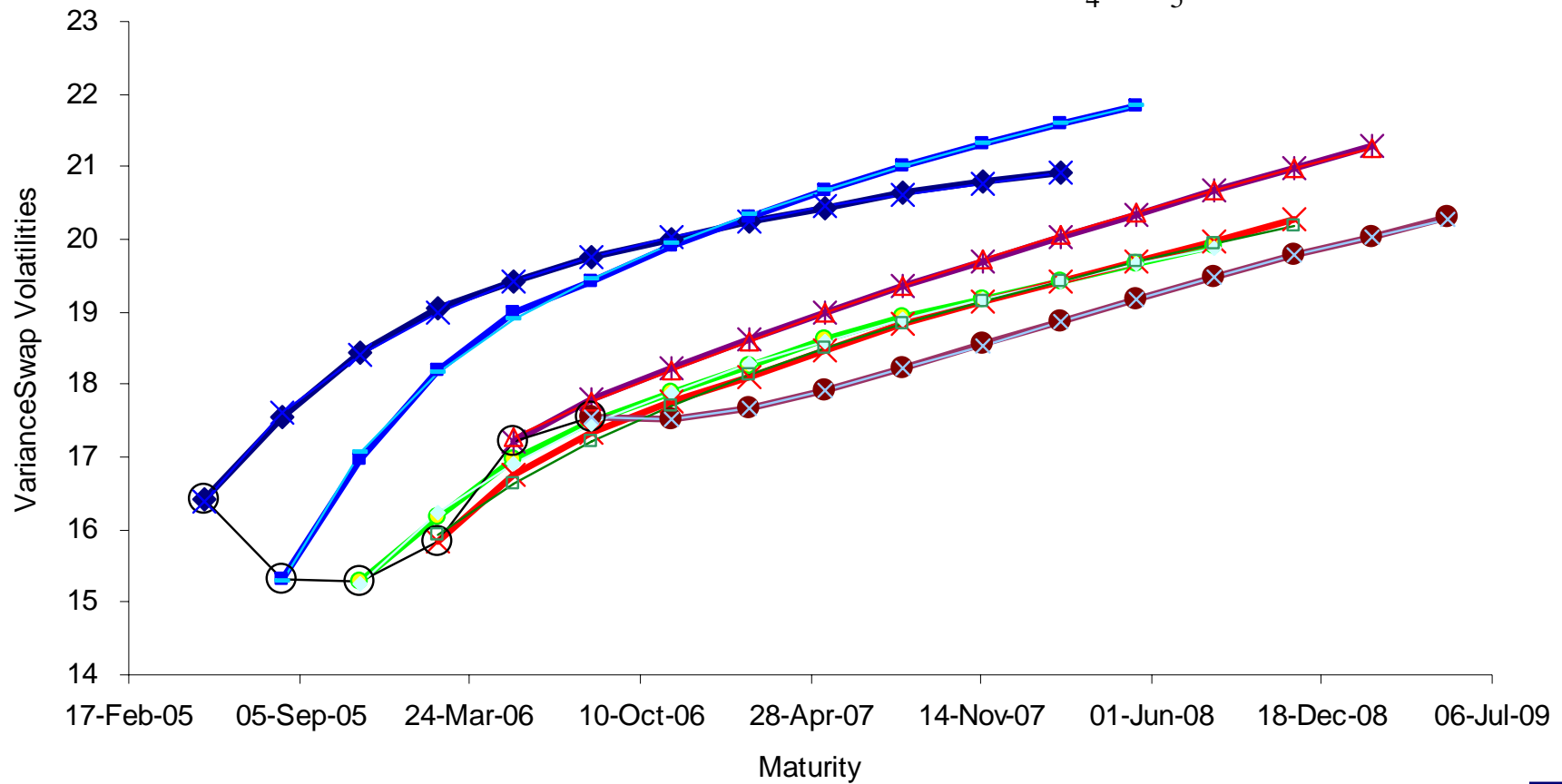
$$d\bar{Z}_t = \mu(\bar{Z}_t)dt + \sum_{j=1}^d \sigma_j(\bar{Z}_t)d\bar{W}_t^j$$



Quadratic Variation

Consistent Variance Curve Models

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





Quadratic Variation

Consistent Variance Curve Models

- The (local) martingale property of the variance swap prices imposes restrictions on the possible coefficients μ and σ of the SDE of Z .

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

- If satisfied, we can always define

$$\frac{dS_t}{S_t} = \sum_{j=1}^d \rho^j(S_t, \bar{Z}_t) (\sqrt{v_t} dW_t^j) \quad v_t := G(0; \bar{Z}_t)$$

Short variance implied by this approach.

where ρ with $\|\rho\|_2=1$ is a local correlation vector which controls the skew of the implied volatility.

- Arbitrage-free model
- Complete under differentiability assumptions on the coefficients of the SDEs.



Quadratic Variation

Consistent Variance Curve Models

- An example (see Bermudez et al 2006) is

$$G(x; z_1, \dots, z_3) := z_3 + (z_1 - z_3)e^{-kx} + (z_2 - z_3) \frac{k}{k-c} (e^{-cx} - e^{-kx})$$

with driving Heston-type process

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

- If W^1 and W^2 are independent, Fourier-transform can be computed using Ricatti equations.
- More elaborate volatility structure requires numerical calibration via MC.



Quadratic Variation

Consistent Variance Curve Models

■ Good

- Fully consistent arbitrage-free joint stock/variance “stochastic volatility” model
- Greeks w.r.t. stock and variance swaps straight forward.
- Consistent term structure dynamics.
- Self-similar future implied volatility surfaces by construction.

■ But

- “Skew” realized via correlation coefficient ρ ; very indirect.
- Cumbersome non-linear calibration for most models except for affine sub-class.



Quadratic Variation

Hybrid models

- Hybrid models – model term structure of variance swaps but use mathematically consistent “locally controlled” stock dynamics for intervals between “reset dates” $t_0 < \dots < t_N$:
 - Assume the “short variance” v is given.
 - On $(t_{k-1}, t_k]$ set

$$S_t := S_{t_{k-1}} X_t^k \quad \frac{dX_t^k}{X_t^k} = \sigma_t^k(X_t^k) dB_t \quad (X_{t_{k-1}}^k = 1)$$

Stochastic or local volatility

such that

$$E_{t_{k-1}} [RV(t_{k-1}, t_k)] = E_{t_{k-1}} \left[\int_{t_{k-1}}^{t_k} \sigma_t^k(X_t^k)^2 dt \right]$$

\equiv

$$V_{t_{k-1}}(t_{k-1}, t_k) := E_{t_{k-1}} \left[\int_{t_{k-1}}^{t_k} v_s ds \right]$$

Note that realized variance is computed using X^k .



Quadratic Variation

Hybrid models

■ Examples:

- Bergomi 2005: two-factor exponential-OU model for ν and CEV distribution for the stock price increments between reset dates.
 - Model intended to hedge Cliquets using (forward) variance swaps
 - Approximation formulas can be used to “fix the smile” for the forward intervals.
 - No closed form for spot Europeans, but good “natural fit”.

- Bermudez et al 2006: one-factor Heston model with Heston forward distribution
 - Pricing tool for assessing cross-correlation risk when pricing Cliquets.
 - Allows to control inter-dependency between the increments in Heston.
 - Analytic Fourier-transform available to calibrate spot European options.
 - Forward distribution simply Heston.



Quadratic Variation

Hybrid models

■ Good

- Mathematically consistent
- Greeks w.r.t. stock and variance swaps straight forward.
- Consistent term structure dynamics.
- Self-similar future implied volatility surfaces at each reset date by construction.

■ But

- Logically inconsistent short variance behaviour across reset dates.
- Reset-date dependent.
- Spot implied vol surface more or less random if models are calibrated towards the forward distribution.



Quadratic Variation

Fitted models

- Dupire 2004: use log-normal variance and fit it to the observed market prices of variance swaps.
 - Assume that at time 0 we observe a variance curve

$$V_0(0, T) = \int_0^T w(s) ds$$

- Define the drift μ in short-variance process

$$\frac{dv_t}{v_t} = \mu_t dt + \sigma_t dW_t$$

such that

$$V_0(0, T) = \mathbb{E} \left[\int_0^T v_s ds \right]$$



Quadratic Variation

Fitted models

- Such a model is always perfectly fitted to the variance swap market. More general, we can extend this “Hull&White-approach” to any short variance model $(x_t)_t$ (Buehler 2006):

- Given x , rescale

$$v_t := w(t) \frac{x_t}{\mathbb{E}[x_t]} \quad V_0(0, T) = \int_0^T w(s) ds$$

such that, obviously,

$$V_0(0, T) = \mathbb{E} \left[\int_0^T v_s ds \right]$$

$w(t)$ is the “forward variance” at time t .

- To construct a stock price process, chose a BM B and set

$$\frac{dS_t}{S_t} = \sqrt{v_t} dB_t$$



Quadratic Variation

Fitted models

■ Examples

- Dupire 2004: Log-normal model.
- Bergomi 2005: Two-factor exponential-OU.
- Buehler 2006: Heston one- and two-factor version
 - Europeans can be priced via Fourier-inversion by solving Ricatti-equations.
 - Analytic Fourier-transform if time-change method is used.
 - Consistent models remain consistent if correction is not “too large”.



Quadratic Variation

Fitted models

■ Good

- Perfect fit to the hedging instrument variance swap for options at all maturities.
- Greeks w.r.t. stock and variance swaps straight forward.
- Consistent term structure dynamics if combined with a consistent model.
- For Heston-based models relatively quick European prices on the stock.

■ But

- Ad hoc definition could lead to inconsistent term-structure movements. But by experience, models are remarkably resistant.
- Skew incorporated only indirectly via correlation parameter.



Jumps

Introducing skew

- The implied volatility skew for options on realized variance is upward sloping.
 - This reflects the additional risk of jumps in the stock price which always contribute to an increase of realized variance.

$$RV(T_1, T_2) \approx QV_{cont}(T_1, T_2) + \sum_{t: T_1 < t \leq T_2} \log^2 \frac{S_t}{S_{t-}}$$

- Hence, more risk premium is charged for OTM calls on variance.
- Adding jumps can improve the fit to the broker market of options on realized variance.
 - “Fitting” can easily be applied to this case.
 - Hedging arguments theoretically less clear.

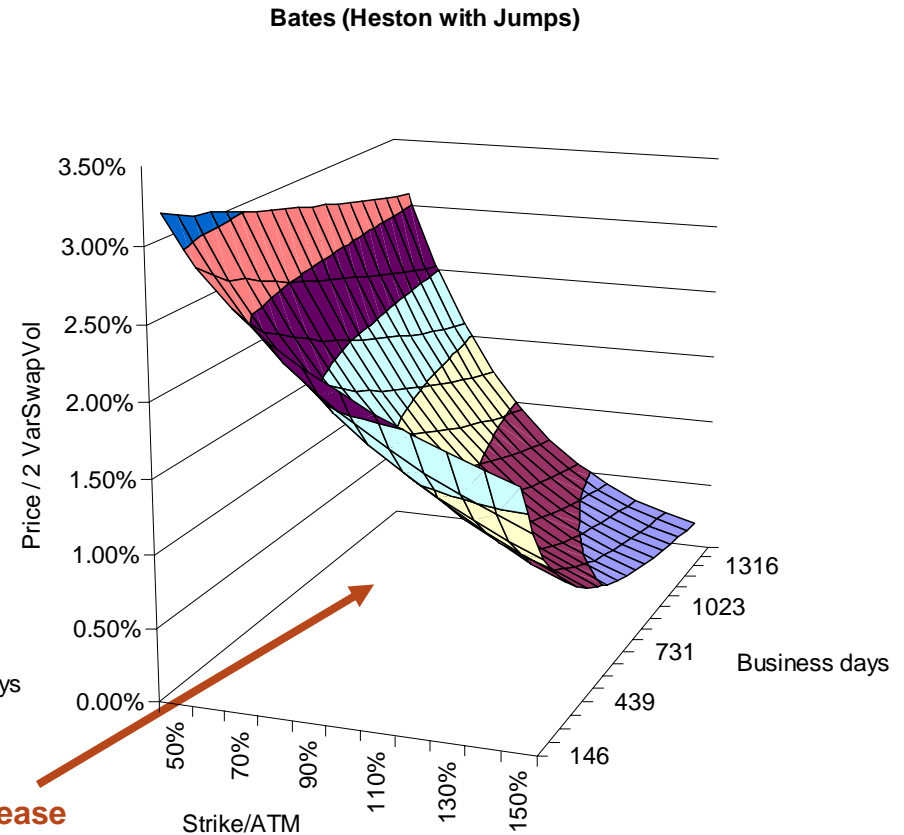
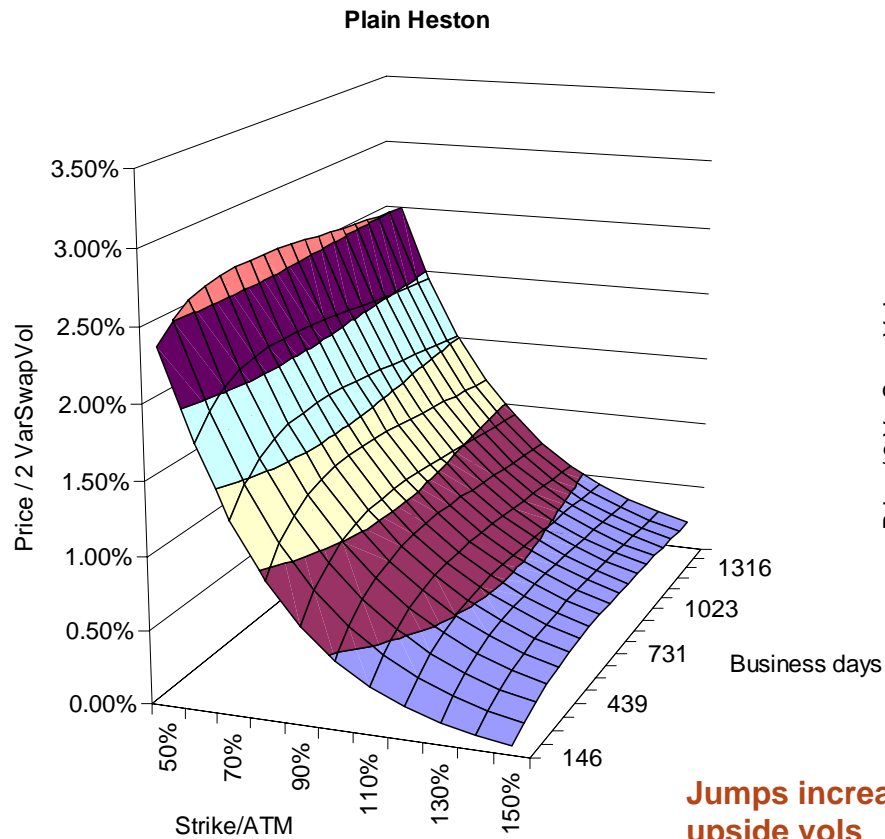
Jumps

Impact of jumps



Bates model vs Heston

VolOfVol in Heston calibrated to Bates equity 1yATM.
Both models have 20% flat variance swap vol.



Jumps increase upside vols (here seen in price)



Hedging



Hedging

... with variance swaps.

- The most natural instrument to hedge an options on variance is a variance swap. Set

$$H_t(T) := E_t[h(RV(0,T))]$$

- Most models we described provide us with a delta with respect to both stock price and the variance swap with the same maturity as the option.
- In theory, for any one-factor SV model, we have

$$dH_t(T) = \Delta_t dS_t + \Theta_t dV_t(0,T)$$

Stock-delta



VarSwap-Delta.





Hedging

... with variance swaps.

- We conduct a test:
calibrate three models ATM and $\pm 5\%$ OTM options on spot
 - Fitted Heston
 - Fitted Bates (Heston with Jumps)
 - Fitted ExpOU with jumps

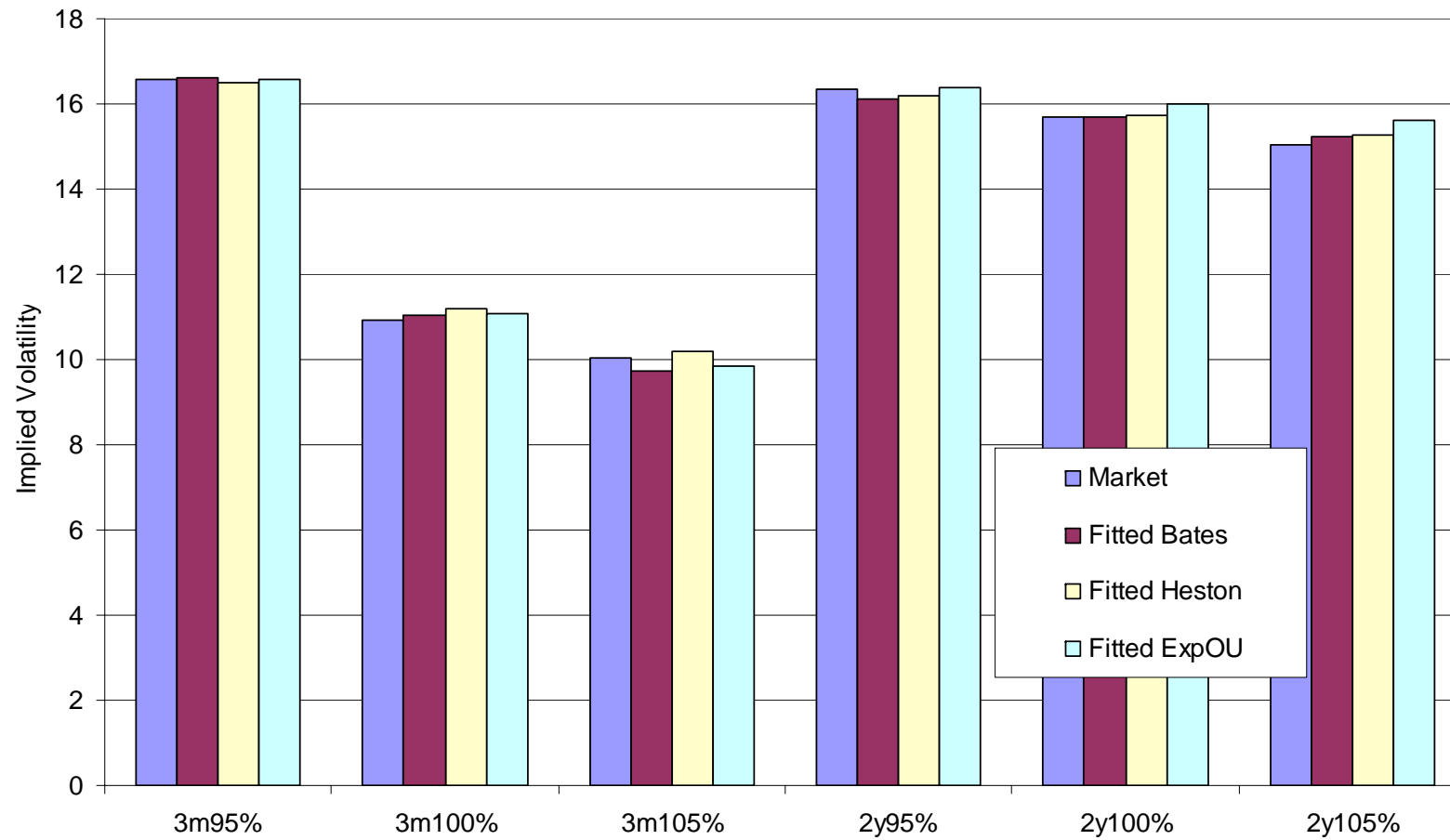
- We want to see
 - Impact of the above on price and variance swap delta reported by the model.



Hedging

... with variance swaps.

.STOXX50E equity market fit 21/11/2006

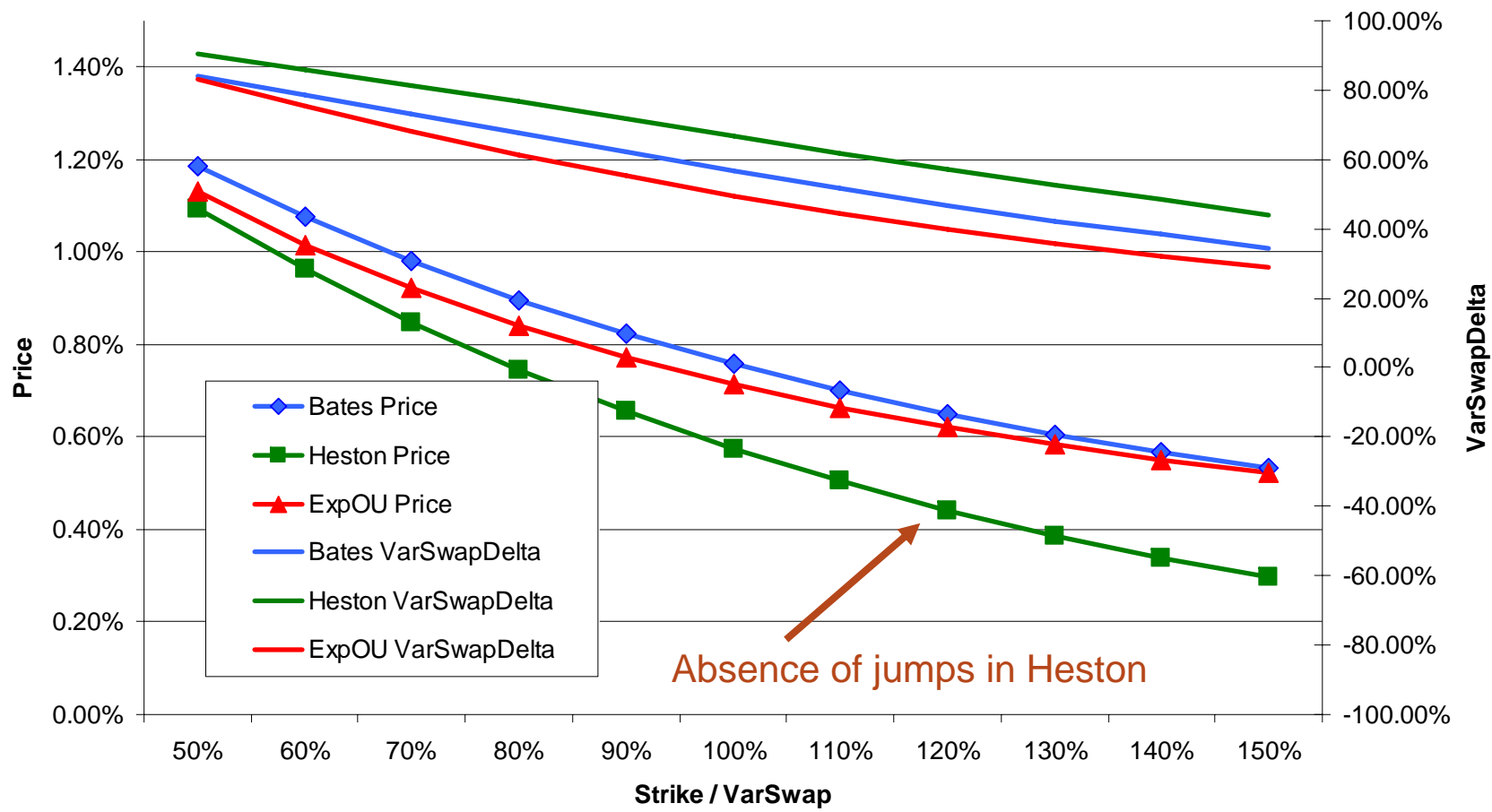




Hedging

... with variance swaps.

3m Options on Variance Price and Variance Swap Delta

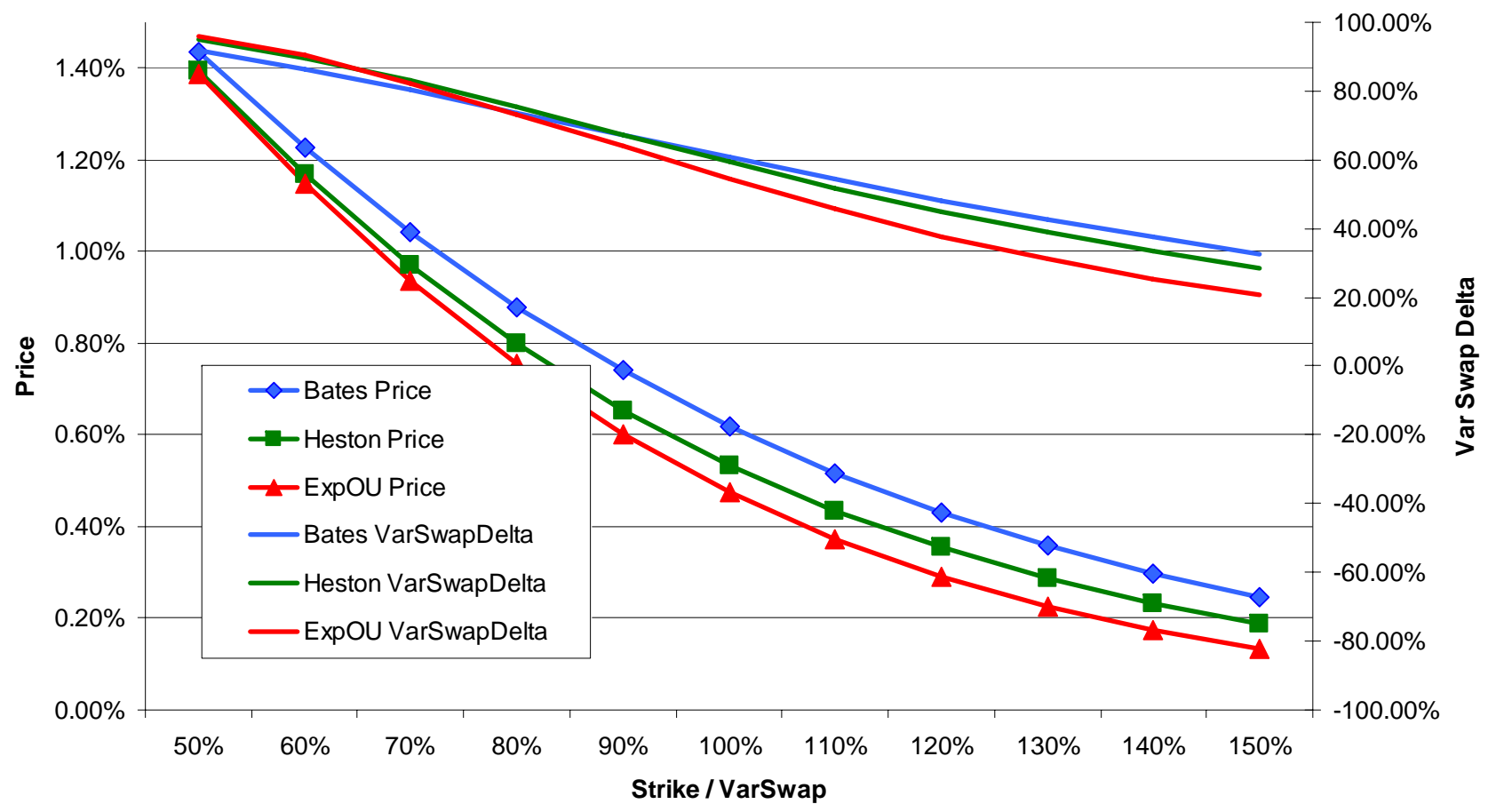




Hedging

... with variance swaps.

2y Options on Variance Price and Variance Swap Delta





Pricing and Hedging in Practise

... more points to look out for

- The idea of hedging with variance swaps is nice, but in practise we face model risk and substantial hedging costs.
 - Approximate variance swap log-profile by far less vanillas and use these to hedge your position.
 - Adapt vol-of-vol accordingly.
 - Diversify



Thank you very much for your attention.
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Quadratic Variation

Fitted models

- This is a close relative to time-changing the process

– Let

$$A(t) := Y^{-1}(V_0(0, T)) \quad Y(t) := E\left[\int_0^t x_s ds\right]$$

and set

$$S_t := X_{A(t)} \quad \frac{dX_t}{X_t} = \sqrt{x_t} dB_t$$

such that once again

$$V_0(0, T) = E[QV(0, T)]$$

Y(t) is the variance swap price given by the model x.

- If we have a closed form for the price of spot European options in the model (X, x) , it is very easy to compute the price for (S, ν) simply by adjusting the maturity of the option accordingly.