

# Static / dynamic properties of stochastic volatility models:

## a structural connection

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## Outline

- Stochastic volatility models produce a smile *AND* generate a dynamics for implied volatilities
- Is there a connection of a structural nature ?
- Enter the *Skew Stickiness Ratio*
  - Long-maturity smiles
    - an expression linking the decay of the ATM and the SSR
  - Historical dynamics of market smiles ?
  - Short-maturity smiles
    - The *realized* skew
    - Materializing the difference between implied and realized skew as a P&L
- Conclusion

## The ATMF skew - 1

- General expression of a stochastic volatility model – no local vol component:

$$dS_t^\omega = (r - q)S_t^\omega dt + \sqrt{\xi_t^t} S_t^\omega dZ_t$$

$$d\xi_t^T = \omega \xi_t^T \sum_k \lambda_{kt}^T(\xi_t) dW_t^k$$

- At order 1 in  $\omega$ : 
$$\xi_t^T = \xi_0^T \left( 1 + \omega \int_0^t \sum_k (\lambda_{k\tau}^T)_0 dW_\tau^k \right)$$

- In particular, for the instantaneous variance:

$$\xi_t^t = \xi_0^t + \delta\xi_t^t$$

$$\delta\xi_t^t = \omega \xi_0^t \int_0^t \sum_k (\lambda_{k\tau}^t)_0 dW_\tau^k$$

- ATMF skew as a function of skewness  $s_T$  of  $x_T = \ln\left(\frac{S_T}{F_T}\right)$ : 
$$\mathbf{S}_T = \frac{d\hat{\sigma}_{KT}}{d \ln K} \Big|_F \approx \frac{s_T}{6\sqrt{T}}$$

with  $s_T = \frac{M_3^T}{(M_2^T)^{3/2}}$ ,  $M_n^T = (x_T - \langle x_T \rangle)^n$

- At order 1 in  $\omega$ : 
$$x_T - \langle x_T \rangle = \int_0^T \sqrt{\xi_0^t} dZ_t + \frac{1}{2} \int_0^T \frac{\delta\xi_t^t}{\sqrt{\xi_0^t}} dZ_t - \frac{1}{2} \int_0^T \delta\xi_t^t dt$$

## The ATMF skew - 2

$$M_2^T = \int_0^T \sqrt{\xi_0^t} dt$$

$$M_3^T = \frac{3}{2} E \left[ \left( \int_0^T \sqrt{\xi_0^t} dZ_t \right)^2 \left( - \int_0^T \delta \xi_t dt + \int_0^T \frac{\delta \xi_t}{\sqrt{\xi_0^t}} dZ_t \right) \right]$$

- Computing the expectation:

$$M_T^3 = 3\omega \int_0^T dt \xi_0^t \int_0^t \sqrt{\xi_0^\tau} \sum_k \rho_{iS} (\lambda_{it}^k)_0 d\tau = 3 \int_0^T dt \left( \int_0^t E \left[ \frac{dS_\tau^0}{S_\tau^0} \delta \xi_t \right] \right)$$

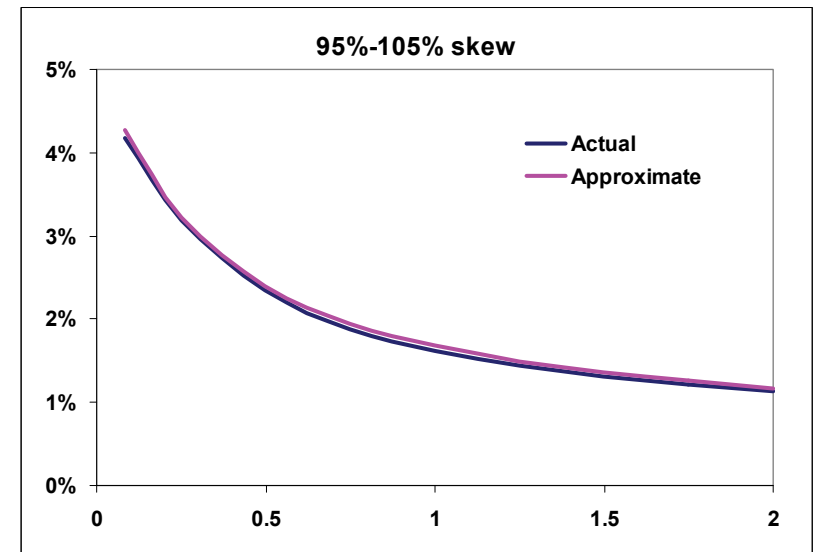
– intuition:  $M_T^3 = \langle (\Sigma r_i)^3 \rangle = \Sigma \langle r_i r_j r_k \rangle = 3 \Sigma_{j>i} \langle r_i r_j^2 \rangle$

- Introduce spot/volatility covariance function:

$$f(\tau, t) = \frac{1}{d\tau} E \left[ \frac{dS_\tau^0}{S_\tau^0} \delta \xi_t \right]$$

- At order 1 in volatility-of-volatility, ATMF skew is given by:

$$S_T = \frac{1}{2\sqrt{T}} \frac{\int_0^T dt \int_0^t f(\tau, t) d\tau}{\left( \int_0^T \xi_0^t dt \right)^{\frac{3}{2}}}$$



## The SSR - 1

- How much does the ATM volatility move when the spot moves – in units of the skew?

- Market-makers use following ratio:  $r_T = \frac{1}{\left. \frac{d\hat{\sigma}_{KT}}{d \ln K} \right|_F} \frac{d\hat{\sigma}_{FT}}{d \ln S}$

- $r_T = 1$  : " sticky-strike " regime

- $r_T = 0$  : " sticky-delta " regime

- Introduce Skew Stickiness Ratio:  $R_T = \frac{1}{\left. \frac{d\hat{\sigma}_{KT}}{d \ln K} \right|_F} \frac{E[d\hat{\sigma}_{FT} d \ln S]}{E[(d \ln S)^2]}$

- Jump / Lévy models :  $R_T = 0$

- Local vol, weak skew :  $R_T = 2$

- Stochastic volatility, short maturity & weak skew:  $R_T = 2$

- At order 1 in volatility-of-volatility

– use VS volatility instead of ATMF volatility:

$$E[d\ln S_\tau d\hat{\sigma}_T] = \frac{1}{2\hat{\sigma}_T T} \int_0^T E[d\ln S d\xi^t] dt = \frac{1}{2\hat{\sigma}_T T} \int_0^T E\left[\frac{dS_0^0}{S_0^0} \delta\xi_t\right] dt = \frac{1}{2\hat{\sigma}_T T} \int_0^T f(0, t) dt$$

– Final expression for the SSR, at order 1 in volatility-of-volatility:

$$R_T = \frac{\int_0^T \xi_0^t dt}{\xi_0^0 T} \frac{T \int_0^T f(0, t) dt}{\int_0^T dt \int_0^t f(\tau, t) d\tau}$$

$$S_T = \frac{1}{2\sqrt{T}} \frac{\int_0^T dt \int_0^t f(\tau, t) d\tau}{\left(\int_0^T \xi_0^t dt\right)^{\frac{3}{2}}}$$

- Take short-maturity limit:

$$R_0 = \lim_{T \rightarrow 0} \frac{T \int_0^T f(0, t) dt}{\int_0^T dt \int_0^t f(\tau, t) d\tau} = 2 \quad S_0 = \lim_{T \rightarrow 0} \frac{1}{2\sqrt{T}} \frac{\int_0^T dt \int_0^t f(\tau, t) d\tau}{\left(\int_0^T \xi_0^t dt\right)^{\frac{3}{2}}} = \frac{f(0, 0)}{4(\xi_0^0)^{3/2}}$$

- Skew tends to a finite value which measures the covariance function at origin
- SSR tends to universal value: 2

## Bounds for the SSR

- Let us make some additional assumptions:
  - VS curve is flat:  $\xi_0^T \equiv \xi_0$
  - Stochastic volatility model is time-homogeneous:  $f(\tau, t) \equiv f(t - \tau)$
  - ATMF skew and SSR take simpler forms:

$$S_T = \frac{\int_0^T (T-t)f(t) dt}{2\xi_0^{3/2}T^2} \quad R_T = \frac{\int_0^T f(t) dt}{\int_0^T (1-\frac{t}{T})f(t) dt}$$

- Assume that  $|f(t)|$  decreases monotonically towards zero. Rewrite  $R_T$  as  $R_T = \frac{g(T)}{\frac{1}{T} \int_0^T g(t) dt}$ ,  $g(t) = \int_0^t f(\tau) d\tau$ 
  - $\frac{g(t)}{g(T)} \leq 1$  implies  $R_T \geq 1$
  - $\frac{g(t)}{g(T)} \geq \frac{t}{T}$  implies  $R_T \leq 2$

⇒ Model-independent range for  $R_T$  :  $R_T \in [1, 2]$

## Scaling of the SSR

- Assume that for large  $t$   $f(t) \propto \frac{1}{t^\gamma}$
- Then, for large  $T$  two types of behaviour, depending on value of  $\gamma$

– Type I : If  $\gamma > 1$

$$S_T \propto \frac{1}{T} \quad \text{and} \quad \lim_{T \rightarrow \infty} R_T = 1$$

– Type II : If  $\gamma < 1$

$$S_T \propto \frac{1}{T^\gamma} \quad \text{and} \quad \lim_{T \rightarrow \infty} R_T = 2 - \gamma$$

– exponential decay: Type I.

- Type I: scaling of  $S_T$  analogous to models with independent increments, however model becomes sticky-strike
- Both types of scaling can be summarized by:

$$S_T \propto \frac{1}{T^{2-R_\infty}}$$

where  $R_\infty = \lim_{T \rightarrow \infty} R_T$



## Type II scaling – in a model ?

- Consider a model of the following type (Bergomi, 2005):  $d\xi_t^T = \omega \xi_t^T \sum w_i e^{-k_i(T-t)} dW_t^i$

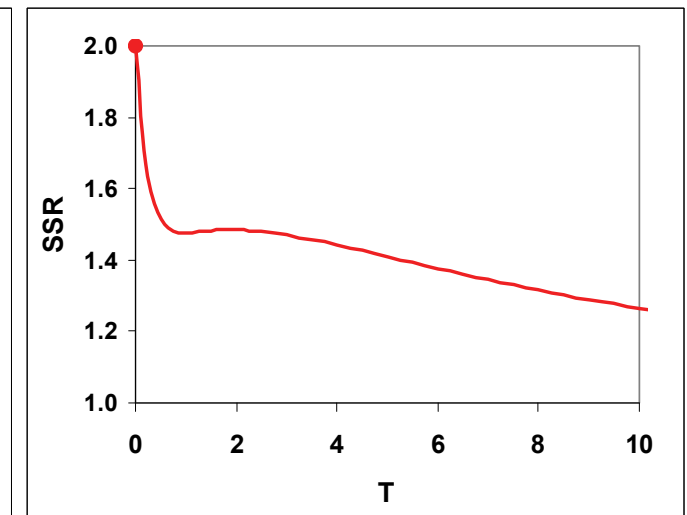
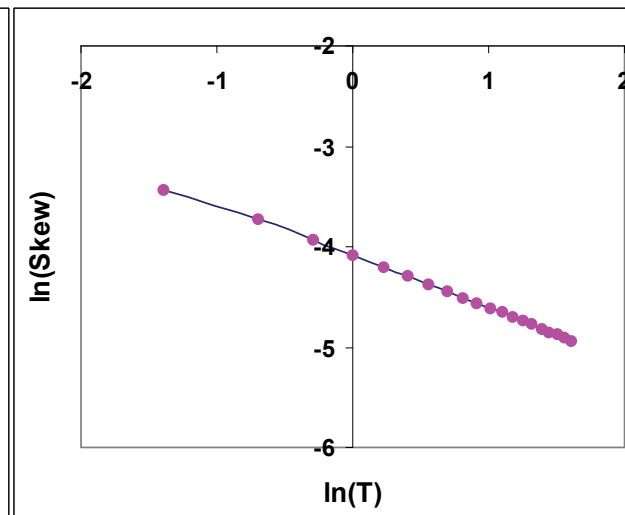
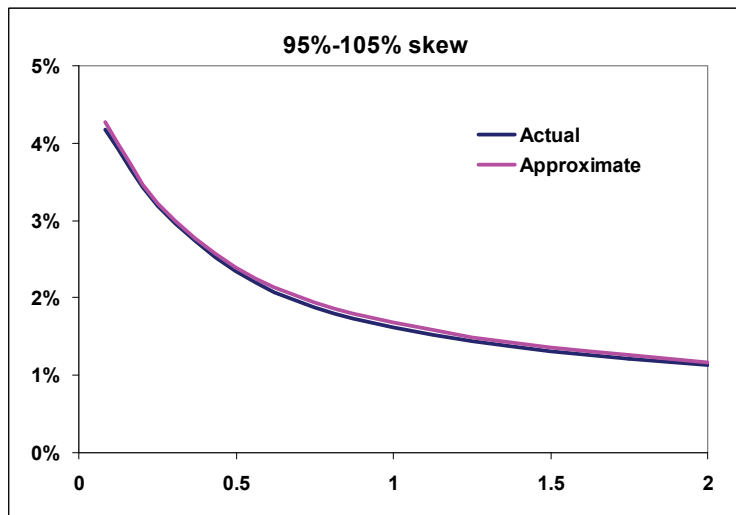
$$f(\tau) = \omega \xi_0^{3/2} \sum w_i \rho_{Si} e^{-k_i \tau}$$

– expressions for skew and SSR:

$$S_T = \frac{\omega}{2} \sum w_i \rho_{Si} \frac{k_i T - (1 - e^{-k_i T})}{(k_i T)^2} \quad R_T = \frac{\sum w_i \rho_{Si} \frac{1 - e^{-k_i T}}{k_i T}}{\sum w_i \rho_{Si} \frac{k_i T - (1 - e^{-k_i T})}{(k_i T)^2}}$$

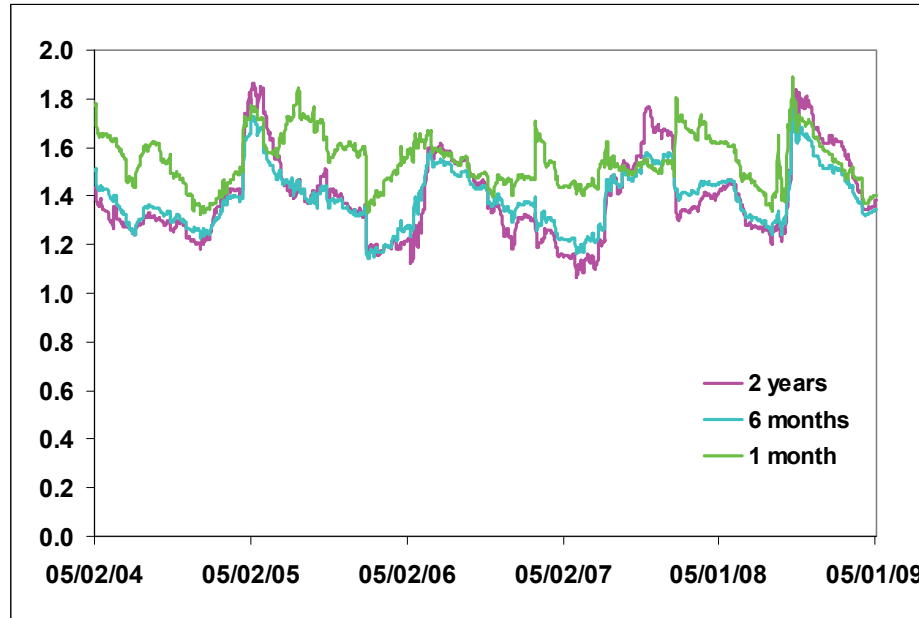
– for large  $\tau$   $f(\tau) \propto e^{-(\min_i k_i) \tau} \Rightarrow$  for (really) large  $T$   $S_T \propto \frac{1}{T}$ ,  $R_T \rightarrow 1$

- By suitably choosing parameter values, get non-trivial scaling over wide range of maturities:
- Example : for intermediate range of maturities  $S_T \propto 1/\sqrt{T}$  and  $R_T \approx 1.5$



## Type II scaling – in the market ?

- Equity index smiles usually exhibit an ATMF skew that typically decays like  $\frac{1}{\sqrt{T}}$ 
  - looks like Type II, but what about the SSR ?
  - For the Eurostoxx, 3-month running estimates of the SSR for maturities 1 month, 6 months, 2 years:



$$R_T = \frac{\sum (\hat{\sigma}_{i+1} - \hat{\sigma}_i) \ln\left(\frac{S_{i+1}}{S_i}\right)}{\sum \frac{d\hat{\sigma}_{KT}^i}{d \ln S} \Big|_F \ln\left(\frac{S_{i+1}}{S_i}\right)^2}$$

⇒ SSR usually in interval [1, 2]

⇒ For longer maturities, average value of the SSR  $\sim 1.5$  – compatible with a decay of the skew  $\sim \frac{1}{\sqrt{T}}$

⇒ Equity volatility markets seem to be of Type II

- One puzzle left: short-maturity SSR always lower than 2 – can this be arbitrated ?
  - is it possible materialize  $2 - R_0$  as a P&L ? (or are we just plain wrong?)

## Arbing the realized SSR – 1

- SSR involves covariance of spot return & ATM vol return, and skew

⇒ Need model

- in which at least spot and ATM vol are dynamic
- that can consistently manage all strikes

- In Black Scholes model, P&L reads  $P \& L = \frac{1}{2} S^2 \frac{d^2 P_K}{dS^2} \left( \left( \frac{\delta S}{S} \right)^2 - \sigma_K^2 \delta t \right)$

– If we sell the 95, buy the 105

- zero Gamma
- " free " Theta
- how " free " is the Theta ? No way of assessing what the " fair " level of skew is

- We would like our P&L to read :

$$P \& L = \frac{1}{2} S^2 \frac{d^2 Q}{dS^2} \left( \left( \frac{\delta S}{S} \right)^2 - \sigma_S^2 \delta t \right) + \frac{1}{2} \sigma_0^2 \frac{d^2 Q}{d\sigma_0^2} \left( \left( \frac{\delta \sigma_0}{\sigma_0} \right)^2 - v^2 \delta t \right) + S \sigma_0 \frac{d^2 Q}{dS d\sigma_0} \left( \frac{\delta S}{S} \frac{\delta \sigma_0}{\sigma_0} - \rho \sigma_S v \delta t \right)$$

where  $\sigma_0$  is the ATM vol and the break-even levels  $\sigma_S, v, \rho$  do not depend on strike (unlike in the BS model).

- Parameterize smile for short, near-the-money options as:

$$\hat{\sigma}(x) = \sigma_0 \left( 1 + \alpha(\sigma_0) x + \frac{\beta(\sigma_0)}{2} x^2 \right), \quad x = \ln(K/S)$$

## Arbing the realized SSR – 2

- For a delta-hedged,  $\sigma_0$ -hedged option, P&L reads:

$$P \& L = \frac{dQ}{dt} \delta t + \frac{1}{2} S^2 \frac{d^2 Q}{dS^2} \left( \frac{\delta S}{S} \right)^2 + \frac{1}{2} \sigma_0^2 \frac{d^2 Q}{d\sigma_0^2} \left( \frac{\delta \sigma_0}{\sigma_0} \right)^2 + S \sigma_0 \frac{d^2 Q}{dS d\sigma_0} \frac{\delta S}{S} \frac{\delta \sigma_0}{\sigma_0}$$

- Our parameterization does not explicitly involve time

– Our Theta is the same as the BS Theta, only distributed differently :

$$\begin{aligned} -\frac{dQ}{dt} \delta t &= \frac{1}{2} S^2 \frac{d^2 P_K}{dS^2} \hat{\sigma}_K^2 \\ &= \frac{1}{2} S^2 \frac{d^2 Q}{dS^2} \sigma_S^2 + \frac{1}{2} \sigma_0^2 \frac{d^2 Q}{d\sigma_0^2} v^2 + S \sigma_0 \frac{d^2 Q}{dS d\sigma_0} \rho \sigma_S v \end{aligned}$$

–  $\frac{dP_{BS}}{dt}, \frac{d^2 P_{BS}}{dS^2}, \frac{d^2 P_{BS}}{d\sigma^2}, \frac{d^2 P_{BS}}{dS d\sigma}$  all start with same prefactor  $\frac{SN'(d)}{\sigma\sqrt{T}}$  with  $d = \frac{-x + \frac{\sigma^2 T}{2}}{\sigma\sqrt{T}}$

– Pull prefactor out and expand at order 2 in  $x$  and order 0 in  $T$  :

$$\begin{aligned} \frac{dQ}{dt} &= -\mathfrak{N} \cdot \sigma_0^2 \left( 1 + \alpha x + \frac{\beta x^2}{2} \right) \\ \frac{1}{2} S^2 \frac{d^2 Q}{dS^2} &= \mathfrak{N} \cdot \left( 1 - 3\alpha x + \left( 6\alpha^2 - \frac{5}{2}\beta \right) x^2 \right) \\ \frac{1}{2} \sigma_0^2 \frac{d^2 Q}{d\sigma_0^2} &= \mathfrak{N} \cdot x^2 \\ S \sigma_0 \frac{d^2 Q}{dS d\sigma_0} &= 2\mathfrak{N} \cdot \left( x - (2\alpha - \alpha') x^2 \right) \end{aligned}$$

$$\text{where } \mathfrak{N} = \frac{SN'(d)}{2\sigma_0\sqrt{T}}, \quad \alpha' = \frac{d\alpha}{d \ln \sigma_0}$$

## Arbing the realized SSR – 3

- Expression of P&L :

$$\begin{aligned}
 P \& L &= \frac{dQ}{dt} \delta t + \frac{1}{2} S^2 \frac{d^2 Q}{dS^2} \left( \frac{\delta S}{S} \right)^2 + \frac{1}{2} \sigma_0^2 \frac{d^2 Q}{d\sigma_0^2} \left( \frac{\delta \sigma_0}{\sigma_0} \right)^2 + S \sigma_0 \frac{d^2 Q}{dS d\sigma_0} \frac{\delta S}{S} \frac{\delta \sigma_0}{\sigma_0} \\
 &= \mathfrak{N} \cdot \left[ -\sigma_0^2 \delta t \left( 1 + \alpha x + \frac{\beta x^2}{2} \right) + \left( 1 - 3\alpha x + \left( 6\alpha^2 - \frac{5}{2} \beta \right) x^2 \right) \left( \frac{\delta S}{S} \right)^2 + x^2 \left( \frac{\delta \sigma_0}{\sigma_0} \right)^2 + 2 \left( x - (2\alpha - \alpha') x^2 \right) \frac{\delta S}{S} \frac{\delta \sigma_0}{\sigma_0} \right]
 \end{aligned}$$

- now, find the break-even level that make P&L vanish  $\forall x$ :  $\left\langle \frac{\delta S^2}{S^2} \right\rangle \equiv \sigma_S^2 \delta t$ ,  $\left\langle \frac{\delta \sigma_0^2}{\sigma_0^2} \right\rangle \equiv v^2 \delta t$ ,  $\left\langle \frac{\delta S}{S} \frac{\delta \sigma_0}{\sigma_0} \right\rangle = \rho \sigma_S v$

$$\begin{aligned}
 P \& L &= \mathfrak{N} \cdot \left[ -\sigma_0^2 \left( 1 + \alpha x + \frac{\beta x^2}{2} \right) + \left( 1 - 3\alpha x + \left( 6\alpha^2 - \frac{5}{2} \beta \right) x^2 \right) \sigma_S^2 + x^2 v^2 + 2 \left( x - (2\alpha - \alpha') x^2 \right) \rho \sigma_S v \right] \delta t \\
 &= \mathfrak{N} \cdot \left[ (-\sigma_0^2 + \sigma_S^2) + (-\sigma_0^2 \alpha - 3\alpha \sigma_S^2 + 2\rho \sigma_S v) x + \left( -\sigma_0^2 \frac{\beta}{2} + \left( 6\alpha^2 - \frac{5}{2} \beta \right) \sigma_S^2 + v^2 - 2(2\alpha - \alpha') \rho \sigma_S v \right) x^2 \right] \delta t
 \end{aligned}$$

- Get following conditions:

$$\sigma_S = \sigma_0$$

$\Rightarrow$  break-even level of spot vol is ATM vol

$$\rho v = 2\alpha \sigma_0$$

$\Rightarrow$  break-even level of spot / ATM vol covariance is given by ATM skew

$$v^2 = \sigma_0^2 (3\beta + 2\alpha^2 - 4\alpha\alpha')$$

$\Rightarrow$  relationship between vol-of-vol and smile curvature is model-dependent

## Arbing the realized SSR – 4

- Examples of admissible models :
- Lognormal dynamics for  $\sigma_0$

–  $\rho, v$  constant implies that  $\alpha(\sigma_0) = \frac{a}{\sigma_0}$ ,  $\beta(\sigma_0) = \frac{b}{\sigma_0^2}$

$$\hat{\sigma}(x) = \sigma_0 \left( 1 + \frac{a}{\sigma_0} x + \frac{b}{2\sigma_0^2} x^2 \right)$$

$$\rho v = 2a$$

$$v = \sqrt{3b + 6a^2}$$

$\Rightarrow$  skew does not depend on vol  
 $\Rightarrow$  curvature is inversely proportional to vol

- Normal dynamics for  $\sigma_0$

–  $v \propto \frac{1}{\sigma_0}$  implies that  $\alpha(\sigma_0) = \frac{a}{\sigma_0^2}$ ,  $\beta(\sigma_0) = \frac{b}{\sigma_0^4}$

$$\hat{\sigma}(x) = \sigma_0 \left( 1 + \frac{a}{\sigma_0^2} x + \frac{b}{2\sigma_0^4} x^2 \right)$$

$$\rho\mu = 2a$$

$$\mu = \sqrt{3b + 10a^2}$$

$\Rightarrow$  skew inversely proportional to vol  
 $\Rightarrow$  curvature is inversely proportional to vol<sup>3</sup>

$$\sigma_S = \sigma_0$$

$$\rho v = 2\alpha\sigma_0$$

$$v^2 = \sigma_0^2 (3\beta + 2\alpha^2 - 4\alpha\alpha')$$

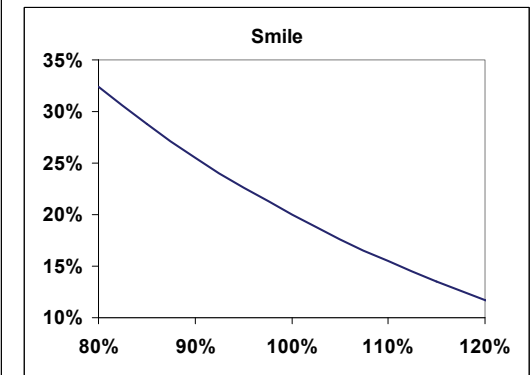
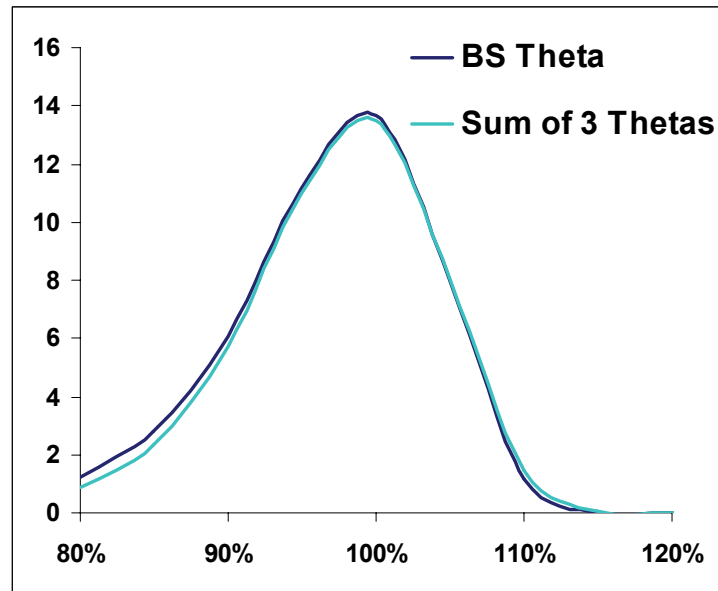
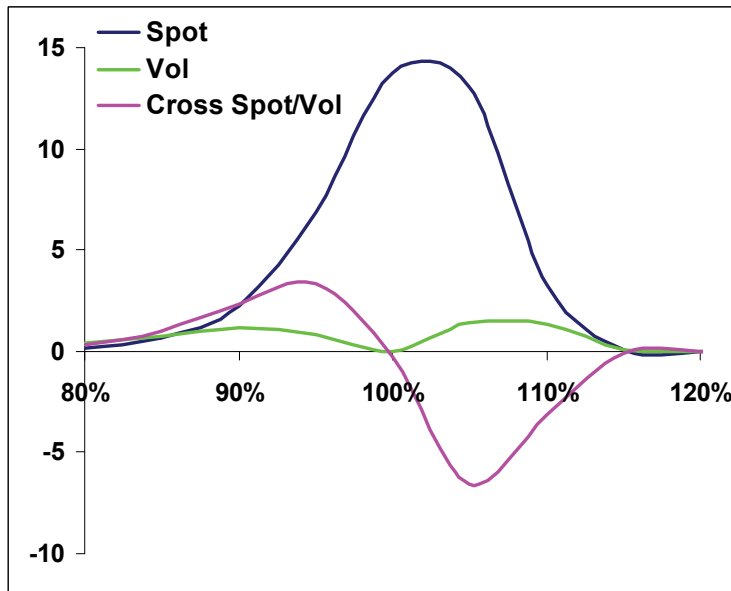
## Arbing the realized SSR – 5

- Use lognormal model for  $\sigma_0$  :
  - P&L accounting:

$$P \& L = \frac{1}{2} S^2 \frac{d^2 Q}{dS^2} \left( \left( \frac{\delta S}{S} \right)^2 - \sigma_0^2 \delta t \right) + \frac{1}{2} \sigma_0^2 \frac{d^2 Q}{d\sigma_0^2} \left( \left( \frac{\delta \sigma_0}{\sigma_0} \right)^2 - (3b + 6a^2) \delta t \right) + S \sigma_0 \frac{d^2 Q}{dS d\sigma_0} \left( \frac{\delta S}{S} \frac{\delta \sigma_0}{\sigma_0} - 2a \sigma_0 \delta t \right)$$

$$\underbrace{S \sigma_0 \frac{d^2 Q}{dS d\sigma_0} \left( \frac{\delta S}{S} \frac{\delta \sigma_0}{\sigma_0} - 2 \sigma_0 \frac{d\hat{\sigma}}{d \ln K} \Big|_S \delta t \right)}_{\text{Arb P\&L}}$$

- Does it work ? How do the 3 Thetas add to the BS Theta ?
  - Take 1 month option,  $\sigma_0=20\%$ ,  $a=-10\%$ ,  $b = 0.4\%$
  - Look at 3 Thetas + sum



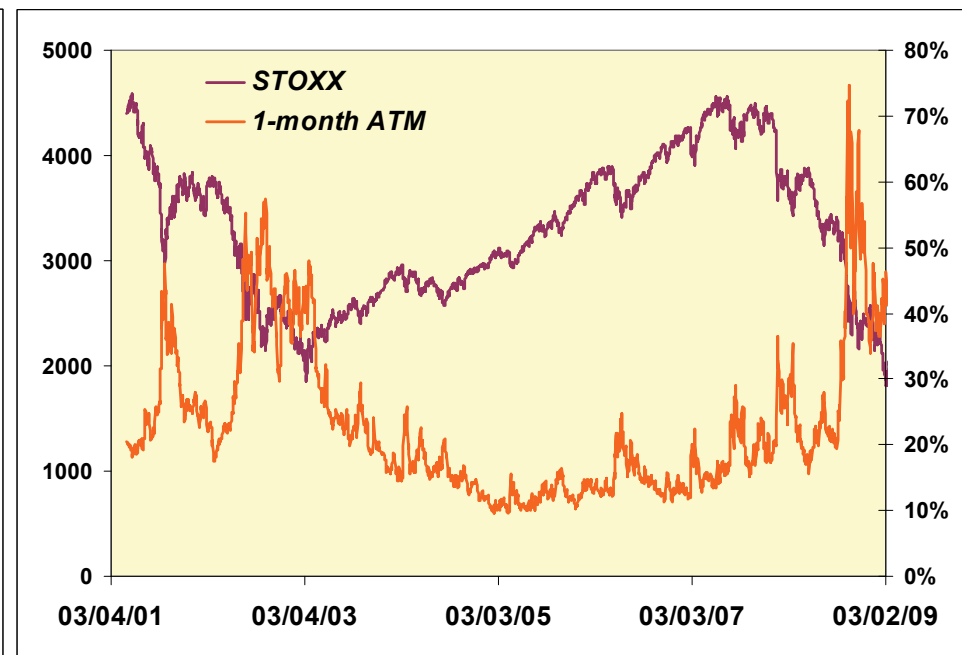
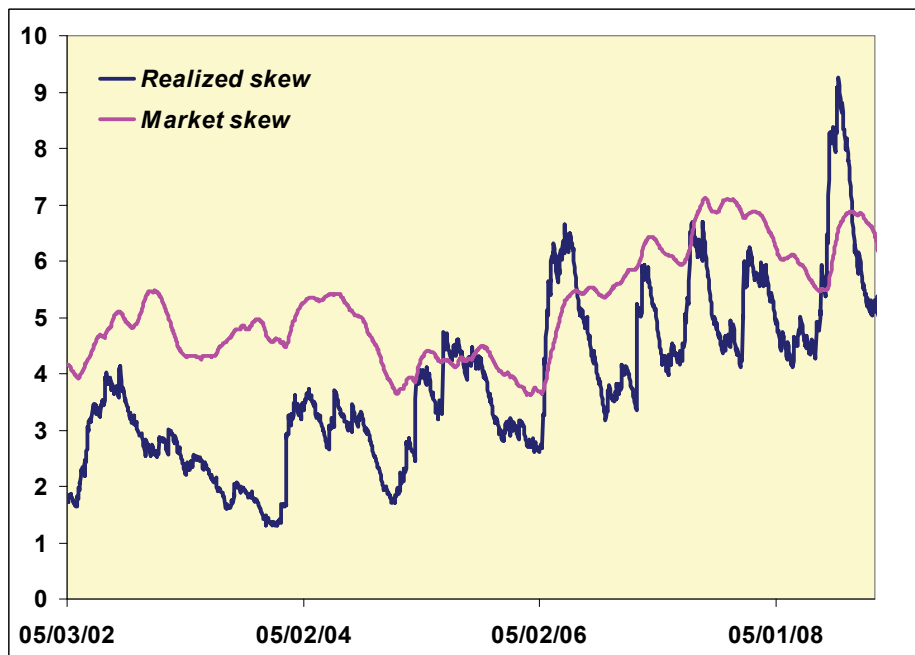
## Arbing the realized SSR – 6

- Try out arbitrage on Stoxx50

– introduce " realized " skew:

$$\left. \frac{d\hat{\sigma}}{d \ln K} \right|_S^{Realized} = \frac{1}{2\sigma_0 \delta t} \left\langle \frac{\delta S}{S} \frac{\delta \sigma_0}{\sigma_0} \right\rangle$$

– implied versus " realized " 1-month 95 / 105 skew



– Back-test strategy:

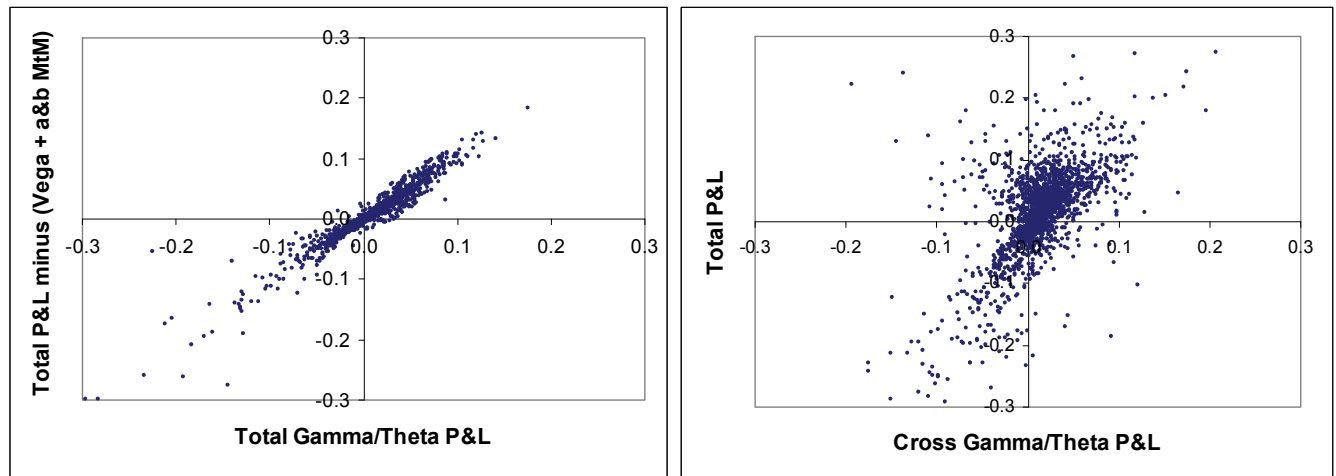
- every day short one 95 option, buy 105 option so that Spot Gamma = zero (~ 0.6)
- carry position until next day, then unwind & restart



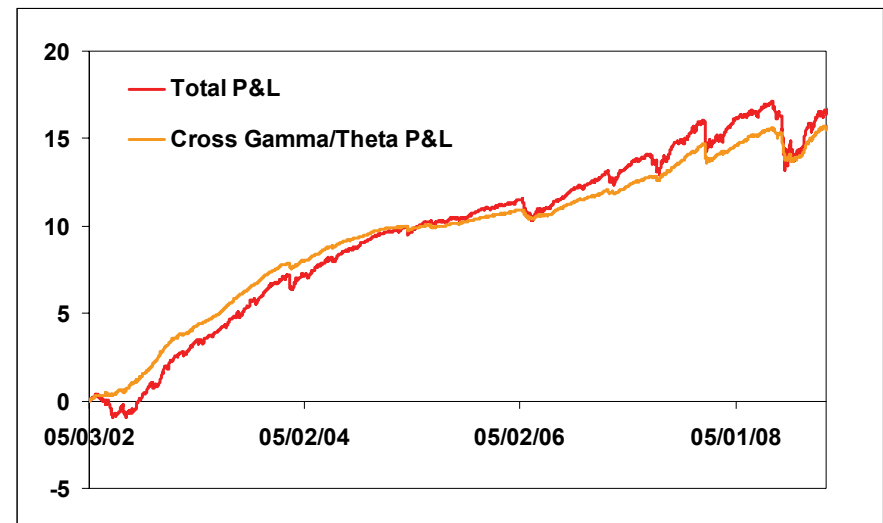
## Arbing the realized SSR – 7

- P&L analysis : in addition to cross-Gamma P&L:
  - Vega P&L
  - vol Gamma / Theta P&L
  - P&L generated by remarking to market  $a$  and  $b$   $\Rightarrow$  tells us whether our model is misspecified

- Scatter plots of daily P&Ls:



- Cumulative P&L:



## Conclusion

- In stochastic volatility models, at order 1 in the vol-of-vol, the Skew Stickiness Ratio and the rate of decay of the ATMF skew are related through the spot/vol covariance function

- For time-homogeneous model + flat VS term-structure:

– SSR is bounded :  $R_T \in [1, 2]$

– 2 types of models

– Type I  $S_T \propto \frac{1}{T}$  and  $\lim_{T \rightarrow \infty} R_T = 1$

– Type II  $S_T \propto \frac{1}{T^\gamma}$  and  $\lim_{T \rightarrow \infty} R_T = 2 - \gamma$

$$S_T \propto \frac{1}{T^{2-R_\infty}}$$

- Index volatility markets' behaviour consistent with Type II

- For short maturity smiles, SSR = 2

– Markets display a realized SSR < 2

– Introduce the notion of " realized skew "

realized covariance of spot & implied vol

–  $2 - R_0$  mismatch can be materialized as a P&L

$$\left. \frac{d\hat{\sigma}}{d \ln K} \right|_S^{\text{Realized}} = \frac{1}{2\sigma_0 \delta t} \left\langle \frac{\delta S}{S} \frac{\delta \sigma_0}{\sigma_0} \right\rangle$$