

# SSVI a la Bergomi

Stefano De Marco<sup>1</sup>, Claude Martini<sup>2</sup>

<sup>1</sup> Ecole Polytechnique

<sup>2</sup> Zeliade Systems

Jim Gatheral 60th birthday conference

# Foreword To Jim

A nice pipeline:

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*Jim's* → *Zeliade (ZQF, Model Validation)* → *Banks, HFs, CCPs*

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- ▶ ..and more to come!



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*So, Jim, on behalf of Zeliade I say: thank you!*

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Reminder on SSVI

Chriss-Morokoff-Gatheral-Fukasawa formula

SSVI a la Bergomi

1st ingredient:  $(e)\text{SSVI}$



# Gatheral SVI

Formula for the implied total variance *at a given maturity  $T$* :

$$v(k) = a + b(\rho(k - m) + \sqrt{(k - m)^2 + \sigma^2})$$

where:

$v$  = implied  $\text{vol}^2 T$ .

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Fits super well (the best 5 parameters model around?).

# Gatheral-Jacquier *Surface* SVI

Formula for the implied total variance *for the whole surface*:

$$w(k, \theta_t) = \frac{\theta_t}{2} (1 + \rho \varphi(\theta_t) k + \sqrt{(\varphi(\theta_t) k + \rho)^2 + \tilde{\rho}^2})$$

where:

$\theta$  : ATM TV,  $\rho$  : (constant) spot vol correlation,  $\varphi$  : (function) curvature.

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$\theta$  taken directly as a parameter: feature quite unique to SSVI. Unlike Bergomi Variance Swap curve parameterization.

(Historically, stems out of SVI. SSVI *slices* are a subfamily of SVI).

# No arbitrage in SSVI

Proposition (GJ, SSVI paper, Theorems 4.1 and 4.2 )

*There is no calendar spread and no butterfly arbitrage if*

$$\partial_t \theta_t \geq 0 \quad (2.1)$$

$$0 \leq \partial_\theta(\theta \varphi(\theta)) \leq \frac{1}{\rho^2}(1 + \bar{\rho})\varphi(\theta), \quad \forall \theta > 0 \quad (2.2)$$

$$\theta \varphi(\theta) \leq \min \left( \frac{4}{1 + |\rho|}, 2\sqrt{\frac{\theta}{1 + |\rho|}} \right), \quad \forall \theta > 0 \quad (2.3)$$

where  $\bar{\rho} = \sqrt{1 - \rho^2}$ .

Condition 2.3 implies that  $\lim_{\theta \rightarrow 0} \theta \varphi(\theta) = 0$ .

# SSVI in practice

Usage: implied vol smoother, risk models

Widely used on Equity (indexes, stocks), works very well

Also on some FI and FX markets

Easy to implement (calibration easier than SVI)

# e(xtended) SSVI

(joint work with Sebas Hendriks)

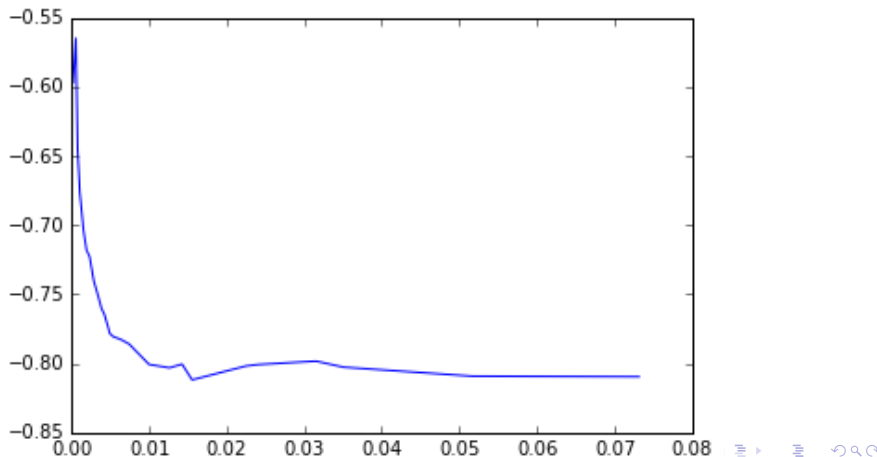
Idea: allows for time ( $\theta$ ) dependent correlation  $\rho$  in SSVI.



# e(xtended) SSVI

(joint work with Sebas Hendriks)

Idea: allows for time ( $\theta$ ) dependent correlation  $\rho$  in SSVI. Motivation: correlation in the calibration of a *joint slice SSVI* model:



# e(xtended) SSVI

eSSVI *slices* are SSVI slices: same no-butterfly arbitrage conditions.

Question: investigate **calendar-spread** arbitrage.

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Question: investigate **calendar-spread** arbitrage.

Starting point: look at **2 SSVI slices with different correlations  $\rho_1, \rho_2$** .

$$\begin{aligned}w_1 &= \frac{\theta_1}{2}(1 + \rho_1\varphi_1k + \sqrt{\varphi_1^2k^2 + 2\rho_1\varphi_1k + 1}) \\w_2 &= \frac{\theta_2}{2}(1 + \rho_2\varphi_2k + \sqrt{\varphi_2^2k^2 + 2\rho_2\varphi_2k + 1})\end{aligned}\tag{2.4}$$

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[ Haute Couture on *parametric quadratic polynomials* here]

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[ Haute Couture on *parametric quadratic polynomials* here]

Proposition (Sufficient conditions for no crossing)

*The 2 smiles don't cross if*

$$\begin{aligned}\theta_2 &\geq \theta_1 \text{ and } \varphi_2 \leq \varphi_1 \\ \frac{\theta_2\varphi_2}{\theta_1\varphi_1} &\geq \max\left(\frac{1+\rho_1}{1+\rho_2}, \frac{1-\rho_1}{1-\rho_2}\right)\end{aligned}$$

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## Proposition

Let

$$\gamma := \frac{1}{\varphi} \frac{d(\theta\varphi)}{d\theta}, \delta := \theta \frac{d(\rho)}{d\theta}$$

Then there is no calendar spread arbitrage in eSSVI iff  $\partial_t \theta_t \geq 0$  and

$$-\gamma \leq \delta + \rho\gamma \leq \gamma$$

and either:

1.  $\gamma \leq 1$
2.  $-\sqrt{2\gamma - 1} \leq \delta + \rho\gamma \leq \sqrt{2\gamma - 1}$



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Can be proven rigorously directly, investigating  $\partial_\theta w$ .

# Representation formula for $\rho(\theta)$

If we restrict to the case where  $0 \leq \gamma \leq 1$ , we can get all possible  $\rho$  satisfying  $-\gamma \leq \delta + \rho\gamma \leq \gamma$  by solving the ODE  $\delta + \rho\gamma = \gamma u$  where  $u$  is any function with values in  $[-1, 1]$ .

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## Proposition

*Assume  $0 \leq \gamma \leq 1$ . Then there is no calendar spread arbitrage in eSSVI iff*

$$\rho(\theta) = \frac{1}{\theta\varphi(\theta)} \int_0^\theta u(\tau) d(\tau\varphi(\tau)) \quad (2.5)$$

*for some  $u \rightarrow [-1, 1]$*

## 2nd ingredient: Chriss-Morokoff-Gatheral-Fukasawa formula

# VIX reminder

For a continuous model:

$$\lim E\left[\sum \log \frac{S_{(k+1)h}}{S_{kh}}\right]^2 = E\left[-2 \log\left(\frac{S_T}{S_0}\right)\right]$$

and one has always the replication formula for the log contract:

$$E\left[-2 \log\left(\frac{S_T}{F_T}\right)\right] = 2 \int_0^{F_T} \frac{P(K, T)}{K^2} dK + 2 \int_{F_T}^{\infty} \frac{C(K, T)}{K^2} dK$$

where we assume that there is no interest rate. Here  $C(K, T)$  (resp.  $P(K, T)$ ) is the price of a Call (resp. Put) with strike  $K$  and time to maturity  $T$ .  $F_T$  is the Forward at maturity  $T$

VIX: synthetic index with a discrete version of this formula (and fixed 30 days time to maturity)

Notation:  $VIX^2(T) = E\left[-2 \log\left(\frac{S_T}{F_T}\right)\right] / T$

# Chriss-Morokoff-Gatheral-Fukasawa formula

In Jim's *Practitioner* book, the following formula is obtained:

$$E[-2 \log(\frac{S_T}{F_T})] = \int \sigma^2(g_2(z)) \frac{e^{-\frac{z^2}{2}}}{\sqrt{2\pi}} dz \quad (3.6)$$

(we drop the  $T$  dependence in the RHS)

where  $g_2$  is the inverse function of the transformation  $k \rightarrow d_2(k, \sigma(k))$  where  $d_2(k, \sigma) = -\frac{k}{\sigma} - \frac{\sigma}{2}$ .

Fukasawa (2010) proved that under no butterfly arbitrage conditions  $d_2(k, \sigma(\cdot))$  is indeed invertible and proved rigorously 3.6.

# General shape of $\sigma(g_2)$



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## Lemma (Fukasawa)

*The inequality  $2g_2(z) \leq z^2$  holds for all  $z \in \mathbb{R}$ . There exists a unique  $z^* > 0$  such that  $2g_2(z^*) = (z^*)^2$ . Moreover, we have  $\sigma(g_2(z)) = z + \sqrt{z^2 - 2g_2(z)}$  below  $z^*$  and  $\sigma(g_2(z)) = z - \sqrt{z^2 - 2g_2(z)}$  above  $z^*$ . In particular,  $\sigma(g_2(z^*)) = z^*$ .*

# $\sigma(g_2)$ in SSVI

SSVI:

$$\sigma^2(g_2(z)) = \frac{\theta}{2}(1 + \rho\varphi g_2 + \sqrt{(\varphi g_2 + \rho)^2 + \bar{\rho}^2})$$

$$\text{so } \theta(1 + \rho\varphi g_2 + \sqrt{(\varphi g_2 + \rho)^2 + \bar{\rho}^2}) = 4(z^2 - g_2 \pm z\sqrt{z^2 - 2g_2})$$

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Setting  $v_2 = \sigma(g_2(z))$  we get the quadratic equation:

$$\theta^2(1 - \rho^2)\varphi^2 \frac{(2z - v_2)^2}{4} = 4[v_2^2 - \theta(1 + \rho\varphi \frac{v_2(2z - v_2)}{2})] \quad (3.7)$$

## Close formula for $\sigma(g_2)$ and the VIX in (e)SSVI

Let  $u := \theta\varphi(\theta)$  and set:

$$a = 1 + \frac{\rho u}{2} - \frac{\bar{\rho}^2 u^2}{16}$$
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$\sigma(g_2(z))^2 = \frac{(b^2 + u^2)z^2 + 4a\theta - 2bz\sqrt{u^2 z^2 + 4a\theta}}{4a^2}$ , integrate in  $z$  wrt Gauss kernel.



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In the sequel:  $V(\theta) := T \text{ VIX}^2(T)$ .

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Assume  $\rho \leq 0$  and no calendar-spread arbitrage. Then:

1.  $\theta \rightarrow V(\theta)$  is non-decreasing.
2.  $V(\theta) \geq \theta$

## Conclusion: (e)SSVI a la Bergomi

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Why a la Bergomi?

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Why a la Bergomi?

$T \text{ VIX}^2(T) = \int_0^T \xi_0(t) dt$  where  $\xi_0$  is the initial Forward Variance curve.

Key input of Bergomi approach.

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Same *principle* for  $\varphi(\theta) = \eta\theta^{-\lambda}$  with  $\lambda \neq 1/2$ .

# $\theta(V)$ formulas

Proposition (ATM implied total variance, uncorrelated sqrt SSVI)

Assume  $\varphi(\theta) = \frac{\eta}{\sqrt{\theta}}$  and  $\rho = 0$ . Then:

$$\theta = \frac{8(1 - \sqrt{1 + \frac{\eta^2}{2} + \frac{\eta^4}{8}(V + \frac{1}{2})}) + \eta^2(V + 2)}{\eta^2(1 + \frac{\eta^2(V-4)}{16})}$$

Proposition (ATM implied total variance, uncorrelated sqrt SSVI, small parameter expansion)

Assume  $\varphi(\theta) = \frac{\eta}{\sqrt{\theta}}$  and  $\rho = 0$ . Then at first order in  $\eta^2$ :

$$\theta = V(1 - \frac{\eta^2(V + 4)}{16})$$

# $\theta(V)$ formulas, ctd

Proposition (Short term ATM implied total variance, sqrt SSVI)

Assume  $\varphi(\theta) = \frac{\eta}{\sqrt{\theta}}$ . Then for small  $\theta$ :

$$\theta = \frac{V}{\left(\frac{(1+\rho^2)}{4}\eta^2 + 1\right)} \left[ 1 + \rho\eta\sqrt{V} \frac{\left(\frac{(3+\rho^2)}{8}\eta^2 + \frac{1}{2}\right)}{\left(\frac{(1+\rho^2)}{4}\eta^2 + 1\right)^{\frac{3}{2}}} - \eta^2 V \frac{\frac{(3\rho^2+1)}{16} + \eta^2 \frac{3(\rho^4+6\rho^2+1)}{64}}{\left(\frac{(1+\rho^2)}{4}\eta^2 + 1\right)^2} + o(V) \right]$$

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with one of these formulas. Same parameters as Bergomi type models.

*Vol and correlation are disentangled.*

# Hurst exponent from short term skew in Rough Bergomi models

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(SSVI)

$$\sqrt{T} \partial_k \sigma_{BS}(k=0) \approx \rho/2\sqrt{\theta} \varphi(\theta) \propto \rho T^{1/2-\lambda}$$



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*Temptative rough (e)SSVI*

Thank you for your attention !

*Thanks and joyeux anniversaire Jim !!!*