

Asymmetric Super-Heston-rough volatility model with Zumbach effect as a scaling limit of quadratic Hawkes processes

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Motivation

“The hunt for a “perfect” statistical model of financial markets is still going on.”

– Pierre Blanc, Jonathan Donier, Jean-Phillippe Boucard (2015).

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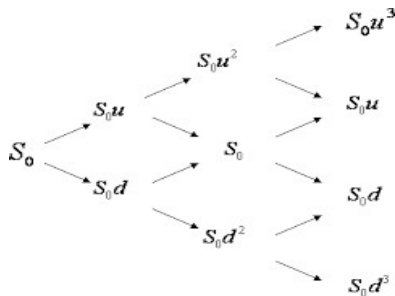
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Aim: Microstructural foundations for macroscopic models.

Multi-period Binomial Model

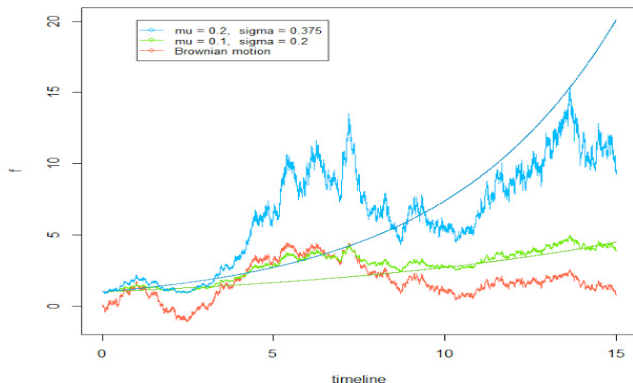
Multi-period Binomial Model



Scaling Limit

$$\frac{dS_t}{S_t} = \mu dt + \sigma dW_t; \quad S_t = S_0 \exp \left(\sigma W_t + \left(\mu - \frac{1}{2} \sigma^2 \right) t \right).$$

Geometric Brownian Motion trajectories



1 Highly endogenous markets

First Stylized Facts

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- 2 Absence of statistical arbitrage

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- 3 Buying/Selling Asymmetry

Hawkes Process Framework for the Microstructure

Bacry, Delattre, Hoffmann, Muzy (2013)

$$P_t = N_t^1 - N_t^2$$

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N_t^i - Hawkes processes with intensity

$$\begin{pmatrix} \lambda_t^1 \\ \lambda_t^2 \end{pmatrix} = \begin{pmatrix} \mu^1 \\ \mu^2 \end{pmatrix} + \int_0^t \Phi(t-s) \begin{pmatrix} dN_s^1 \\ dN_s^2 \end{pmatrix}.$$

$$\Phi = \begin{pmatrix} \phi_1 & c\phi_2 \\ \phi_2 & \phi_1 + (c-1)\phi_2 \end{pmatrix}$$

$$K = \int_0^\infty \Phi(t) dt; \quad \text{stability: } \rho(K) < 1.$$

Getting the Stylized Facts

Euch, Fukasawa, Rosenbaum (2018)

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- 1 Highly endogenous markets: $\rho(K) \rightarrow 1$. Near unstable regime; stochastic volatility.
- 2 Absence of statistical arbitrage: $\mu^1 = \mu^2$; row sums of Φ identical.
- 3 Buying/Selling Asymmetry: $c \neq 1$.

The Heston Model as the Scaling Limit

$$dP_t = \frac{1}{1 - (\|\phi_1\|_1 - \|\phi_2\|_1)} \sqrt{\frac{2}{1+c}} \sqrt{V_t} dW_t$$

$$dV_t = \kappa(v_0 - V_t) dt + \eta \sqrt{V_t} dB_t$$

$$d\langle W, B \rangle_t = \frac{1-c}{\sqrt{2(1+c^2)}} dt.$$

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- Leverage Effect: $c > 1$; negative correlations between price returns and vol increments at macro scale.
- Stochastic volatility only in the near critical regime.

One More Stylized Fact

- Volatility is rough (Gatheral *et. al.* 2018); log-vol \approx fractional BM with Hurst parameter of order 0.1
- Meta orders, split by trading algorithms; Leads to long range correlations
- Tail of the largest eigenvalue:

$$\int_t^\infty (\phi_1 + c\phi_2)(s)ds \sim Ct^{-\alpha}, \alpha \in \left(\frac{1}{2}, 1\right) \dots$$

Scaling Limit is the **Rough Heston** Model

$$V_t = V_0 + \int_0^t (t-s)^{\alpha-1} \kappa(c - V_s) ds + \eta \sqrt{V_s} dB_s$$

TRA: The Zumbach Effect (2010)

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 - past returns negatively affect future vols but not other way around

Encoding the Zumbach Effect

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$$\lambda_t = \mu + \int_0^t \phi(t-s) dN_s + Z_t^2; \quad Z_t = \int_0^t k(t-s) dP_s$$

$$P_t = \sum_{i=1}^{N_t} \xi_i; \quad \xi_i \stackrel{iid}{\sim} \pm 1 \text{ w.p. } \frac{1}{2}.$$

- A univariate Hawkes process is being used here; positive and negative price trends have same impact on volatility. Martingale structure inside the quadratic term.

Super-Heston-Rough-Volatility with Zumbach Effect: Dandapani, Jusselin, Rosenbaum (2021)

Scaling Limit in the near unstable regime:

$$dP_t = \sqrt{V_t} dW_t$$

$$V_t = V_0 + C_1 \int_0^t (t-s)^{\alpha-1} (\theta(s) - V_s + Z_s^2) ds + C_2 \int_0^t (t-s)^{\alpha-1} \sqrt{V_s} dB_s$$

$$Z_t = \int_0^t k(t-s) \sqrt{V_s} dW_s$$

- W, B independent Brownian motions; $\alpha \in (\frac{1}{2}, 1)$.

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- Stochastic Volatility even in the stable regime.

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- W, B independent Brownian motions; $\alpha \in (\frac{1}{2}, 1)$.
- Stochastic Volatility even in the stable regime.
- **Super:** Enhanced volatility tails.

Encoding Asymmetric Impact of Price Trends

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$$\Phi = \begin{pmatrix} \phi_1 & \phi_2 \\ \phi_2 & \phi_1 \end{pmatrix}$$

$$\kappa = \begin{pmatrix} k_1 & -k_2 \\ -k_2 & k_1 \end{pmatrix}; \quad k_1 > k_2 > 0$$

$$M_t^i = N_t^i - \int_0^t \lambda_s^i ds, \quad i = 1, 2.$$

Stability

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$$\alpha = \sqrt{\alpha_1 \alpha_2}; \quad \phi = \phi_1 - \phi_2 \quad \bar{\phi} = \phi_1 + \phi_2$$

$$\text{Stability: } \|\bar{\phi}\|_1 + \frac{(\alpha_1 + \alpha_2)}{2} \|k\|_2^2 < 1$$

Scaling

$$P_{tT}^T = N_{tT}^{(1,T)} - N_{tT}^{(2,T)}, \quad t \in [0, 1], \quad T \geq 0.$$

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$$\begin{aligned} \lambda_{tT}^T &:= \lambda_{tT}^{(1,T)} - \lambda_{tT}^{(2,T)} = \int_0^t \phi^T(T(t-s)) \sqrt{T} d\left(\frac{M_{tT}^T}{\sqrt{T}}\right) \\ &+ \int_0^t \phi^T(T(t-s)) T \lambda_{Ts}^T ds + \alpha \left[\int_0^t k^T(T(t-s)) \sqrt{T} d\left(\frac{M_{tT}^T}{\sqrt{T}}\right) \right]^2 \end{aligned}$$

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$$\phi^T(\cdot) = \phi\left(\frac{\cdot}{T}\right) \frac{\beta}{T}, \quad \bar{\phi}^T(\cdot) = \phi\left(\frac{\cdot}{T}\right) \frac{\bar{\beta}}{T}; \quad k^T(\cdot) = k\left(\frac{\cdot}{T}\right) \frac{\sqrt{\gamma}}{T}$$

Scaled Intensity

$$\begin{aligned}\lambda_{tT}^T &= \beta \int_0^t \phi(t-s) \lambda_{Ts}^T ds + \int_0^t \phi(t-s) \frac{1}{\sqrt{T}} d\left(\frac{M_{tT}^T}{\sqrt{T}}\right) \\ &+ \alpha \gamma \left[\int_0^t k(t-s) d\left(\frac{M_{tT}^T}{\sqrt{T}}\right) \right]^2.\end{aligned}$$

Assumptions

(i) $0 < \bar{\beta} + (\alpha_1 + \alpha_2)\gamma < 1$, $\|k^2\|_1 = \|\phi\|_1 = 1$.

(ii) The function $k \in L^{2+\epsilon}$ for some $\epsilon > 0$ and for any $0 \leq t \leq \hat{t} \leq 1$,

$$\int_0^t |k(\hat{t} - s) - k(t - s)|^2 < C|\hat{t} - t|^r$$

for some $r > 0$ and $C > 0$ and

$$\frac{1}{\eta} \int_0^1 |k(t)|^2 t^{-2\eta} dt + \int_0^1 \int_0^1 \frac{|k(t) - k(s)|^2}{|t - s|^{1+2\eta}} ds dt, \text{ is finite.}$$

for some $\eta \in (0, 1)$.

Scaling Limit in the Stable Regime

Theorem

The family of processes

$$\left\{ \left(\frac{P_{tT}^T}{T}, \frac{P_{tT}^T - T\Lambda_t^T}{\sqrt{T}} \right) : t \in [0, 1] \right\}_{T>0}$$

is C-tight. Any subsequential limit (X, M) satisfies

$$X_t = \int_0^t V_s ds \text{ and } M_t = \int_0^t \sqrt{\bar{V}_s} dB_s,$$

$$V_t = \beta \int_0^t \phi(t-s) V_s ds + \alpha \gamma \left(\int_0^t k(t-s) dM_s \right)^2,$$

$$\bar{V}_t = 2\mu + \bar{\beta} \int_0^t \phi(t-s) \bar{V}_s ds + \frac{(\alpha_1 + \alpha_2)}{2} \gamma \left(\int_0^t k(t-s) \sqrt{\bar{V}_s} dB_s \right)^2.$$

Nearly unstable criteria

For, non-linear Hawkes processes to be stable the mean intensity should converges to a finite constant.

$$\mathbb{E}[\lambda_t^T] \leq \frac{2\mu_T}{1 - \left(\|\bar{\phi}^T\|_1 + \frac{(\alpha_1 + \alpha_2)}{2} \|k^T\|_2^2 \right)}.$$

$$\text{Let } a_T = \|\bar{\phi}^T\|_1 + \frac{(\alpha_1 + \alpha_2)}{2} \|k^T\|_2^2.$$

- **Stable case:** $a_T < 1$ for all $T \in [0, \infty]$.
- **Nearly unstable criteria:** $a_T < 1$ for all $T \in [0, \infty)$ and $a_T \rightarrow 1$ as $T \rightarrow \infty$.

Scaling

$$\frac{1 - a_T}{2\mu_T} P_{tT}^T = \frac{1 - a_T}{2\mu_T} (N_{tT}^{(1,T)} - N_{tT}^{(2,T)}), \quad t \in [0, 1], \quad T \geq 0.$$

$$\begin{aligned} \lambda_{tT}^{*T} &:= \frac{1 - a_T}{2\mu_T} \lambda_{tT}^T = \frac{1 - a_T}{2\mu_T} \int_0^t \phi^T(T(t-s)) T \lambda_{Ts}^T ds \\ &+ \frac{1 - a_T}{2\mu_T} \left(\int_0^t \phi^T(T(t-s)) dM_{Ts}^T + \alpha \left[\int_0^t k^T(T(t-s)) dM_{Ts}^T \right]^2 \right). \end{aligned}$$

$$M_{tT} = M_{tT}^{(1,T)} - M_{tT}^{(2,T)}; \phi^T(.) = \beta_T \phi(.),$$

$$\bar{\phi}^T(.) = \bar{\beta}_T \phi(.), k^T(.) = k\left(\frac{\cdot}{T}\right) \sqrt{\frac{1 - a_T}{T}}.$$

Here, $|\beta_T| < \bar{\beta}_T < 1$, $|\beta_T| = \frac{1}{c_1(1-a_T)+1}$, $\bar{\beta}_T = \frac{1}{c_2(1-a_T)+1}$ and $c_2 > c_1 > 0$ are constants.

Scaled Intensity

$$\begin{aligned}\lambda_{tT}^{*T} &:= \frac{1-a_T}{2\mu_T} \lambda_{tT}^T = \int_0^t \frac{(1-a_T)T\psi_T(T(t-s))}{\sqrt{2\mu_T(1-a_T)T}} dM_s^{*T} \\ &+ \frac{1-a_T}{2\mu_T} \alpha(Z_{tT}^T)^2 + \int_0^t (1-a_T)T\psi_T(T(t-s)) \left(\frac{\alpha}{2\mu_T} (Z_{sT}^T)^2 \right) ds.\end{aligned}$$

$$M_{tT}^{*T} = \sqrt{\frac{1-a_T}{2\mu_T T}} (M_{tT}^{(1,T)} - M_{tT}^{(2,T)}), Z_{tT}^T = \sqrt{2\mu_T} \int_0^t k(t-s) dM_s^{*T},$$

$$\psi_T := \sum_{i \geq 1} (\phi^T)^{\circledast i}, \text{ where } \circledast \text{ is convolution operator.}$$

Assumptions

- $a_T := \left\| \overline{\phi}^T \right\|_1 + \frac{(\alpha_1 + \alpha_2)}{2} \left\| k^T \right\|_2^2 \rightarrow 1$ as $T \rightarrow \infty$.

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- $\phi^T(\cdot) = \beta_T \phi(\cdot), \quad \bar{\phi}^T(\cdot) = \bar{\beta}_T \phi(\cdot), \quad \|\phi\|_1 = 1.$
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- $\tilde{\alpha} x^{\tilde{\alpha}} \int_x^{+\infty} \phi(s) ds \rightarrow K$ as $x \rightarrow +\infty; \quad \tilde{\alpha} \in (\frac{1}{2}, 1).$

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- $\tilde{\alpha} x^{\tilde{\alpha}} \int_x^{+\infty} \phi(s) ds \rightarrow K$ as $x \rightarrow +\infty; \quad \tilde{\alpha} \in (\frac{1}{2}, 1).$
- Let $\delta = K \frac{\Gamma(1-\tilde{\alpha})}{\tilde{\alpha}}$. As $T \rightarrow \infty$

$$(1 - a_T) T^{\tilde{\alpha}} \rightarrow \lambda \delta, \quad \mu_T T^{(1-\tilde{\alpha})} \rightarrow \frac{\mu^*}{\delta}, \quad \bar{\mu}_T T^{(1-\tilde{\alpha})} \rightarrow \frac{\bar{\mu}^*}{\delta}.$$

$$|\beta_T| < \bar{\beta}_T < 1, |\beta_T| = \frac{1}{c_1(1-a_T)+1}, \bar{\beta}_T = \frac{1}{c_2(1-a_T)+1} \text{ and } c_2 > c_1 > 0 \text{ are constants.}$$

Scaling Limit in the nearly-unstable regime

Theorem

$$\left(\frac{1 - a_T}{2\mu_T} \left(\frac{P_{tT}^T}{T} \right), \sqrt{\frac{1 - a_T}{2\mu_T}} \left(\frac{P_{tT}^T - T\Lambda_t^T}{\sqrt{T}} \right) \right)_{t \in [0,1]} \text{ is } C\text{-tight.}$$

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Any subsequential limit (X, M^*) satisfies

$$X_t = \int_0^t V_s ds, \quad M_t^* = \int_0^t \sqrt{V_s} dB_s^1, \quad \overline{M}_t^* = \int_0^t \sqrt{V_s} dB_s^2,$$

$$Z_t^* = \int_0^t k(t-s) dM_s^*, \quad \langle B^1, B^2 \rangle_t = \frac{V_t}{V_t}.$$

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$$Z_t^* = \int_0^t k(t-s) dM_s^*, \quad \langle B^1, B^2 \rangle_t = \frac{V_t}{V_t}.$$

$$V_t = \int_0^t \frac{1}{c_1} f^{\tilde{\alpha}, \lambda}(t-s) \left[\frac{1}{\sqrt{\lambda \mu^*}} dM_s^* + \left(\frac{\mu^*}{\mu^*} + \alpha(Z_s^*)^2 \right) ds \right]$$

$$\overline{V}_t = \int_0^t \frac{1}{c_2} f^{\tilde{\alpha}, \lambda}(t-s) \left[\frac{1}{\sqrt{\lambda \mu^*}} d\overline{M}_s^* + \left(1 + \frac{(\alpha_1 + \alpha_2)}{2} (Z_s^*)^2 \right) ds \right]$$