

« Quadratic » Hawkes processes: A microfoundation for rough vol models?

Fat-tails and Time Reversal Asymmetry

Pierre Blanc, Jonathan Donier, JPB

(building on previous work with

Rémy Chicheportiche & Steve Hardiman)

« Stylized facts »

I. Well known:

- Fat-tails in return distribution $p(r) \underset{|r| \rightarrow \infty}{\sim} \frac{C'}{|r|^{1+\nu}}$

with a (universal?) exponent ν around 4 for many different assets, periods, geographical zones,...

- Fluctuating volatility with « long-memory » -- log-vol is close to a fBM with H small (« rough vol »*), possibly H=0 (« multifractal »**)
- Leverage effect (negative return/vol correlations)

* Gatheral-Jaisson-Rosenbaum; ** Bacry-Muzy

« Stylized facts »

II. The Zumbach effect:

- Intuition: past trends, up or down, increase future vol more than alternating returns (for a fixed HF activity/volatility)
 - Reverse not true (HF vol does not predict more trends)
- Can one create « micromodels » that capture all these effects?
- What is the large scale limit of these models? « Rough » volatility?

Hawkes processes

- A *self-reflexive feedback* framework, mid-way between purely stochastic and agent-based models
- Activity is a Poisson Process with history dependent rate:

$$\lambda_t = \lambda_\infty + \int_{-\infty}^t \phi(t-s) dN_s$$

- Feedback intensity $n \equiv \int_0^\infty \phi(\tau) d\tau < 1$
- Calibration on financial data suggests *near criticality* ($n \rightarrow 1$) and *long-memory* power-law kernel ϕ :
the « Hawkes without ancestors » limit (Brémaud-Massoulié)

Continuous time limit of near-critical Hawkes

- Jaisson-Rosenbaum show that when $n \rightarrow 1$ Hawkes processes converge (in the right scaling regime) to either:
 - i) Heston for short-range kernels
 - ii) Fractional Heston for long-range kernels, with a small Hurst exponent H
- But: still no fat-tails and no TRA...
- J-R suggest results apply to log-vol, but why?
- Calibrated Hawkes processes generate very little TRA, even on short time scales (see below)

Generalized Hawkes processes

- Intuition: not just past activity, but *price moves themselves* feedback onto current level of activity
- The most general quadratic feedback encoding is:

$$\lambda_t = \lambda_\infty + \frac{1}{\psi} \int_{-\infty}^t L(t-s) dP_s + \frac{1}{\psi^2} \int_{-\infty}^t \int_{-\infty}^t K(t-s, t-u) dP_s dP_u$$

- With: $dN_t := \lambda_t dt$; $dP := (+/-) \psi dN$ with random signs
- $L(\cdot)$: leverage effect neglected here (small for intraday time scales)
- $K(\cdot, \cdot)$ is a symmetric, positive definite operator
- Note: $K(t, t) = \phi(t)$ is exactly the Hawkes feedback ($dP^2 = dN$)

Generalized Hawkes processes

$$\lambda_t = \lambda_\infty + \frac{1}{\psi} \int_{-\infty}^t L(t-s) dP_s + \frac{1}{\psi^2} \int_{-\infty}^t \int_{-\infty}^t K(t-s, t-u) dP_s dP_u$$

- 2- and 3-points correlation functions

$$\mathcal{C}(\tau) \equiv \mathbb{E} \left[\frac{dN_t}{dt} \frac{dN_{t-\tau}}{dt} \right] - \bar{\lambda}^2 = \mathbb{E} \left[\lambda_t \frac{dN_{t-\tau}}{dt} \right] - \bar{\lambda}^2,$$

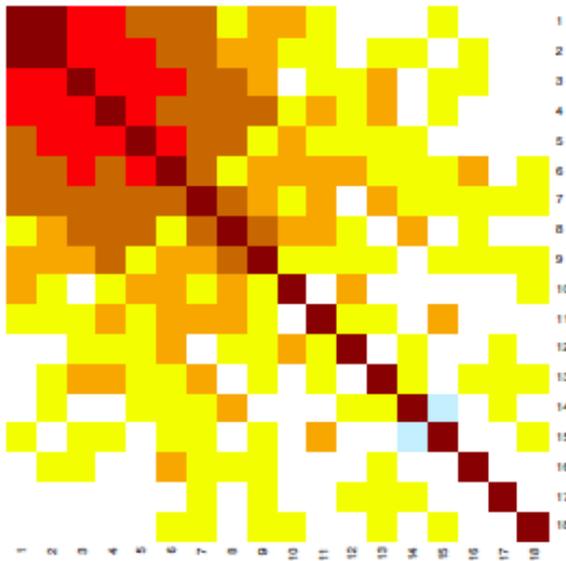
$$\mathcal{D}(\tau_1, \tau_2) \equiv \frac{1}{\psi^2} \mathbb{E} \left[\frac{dN_t}{dt} \frac{dP_{t-\tau_1}}{dt} \frac{dP_{t-\tau_2}}{dt} \right] = \frac{1}{\psi^2} \mathbb{E} \left[\lambda_t \frac{dP_{t-\tau_1}}{dt} \frac{dP_{t-\tau_2}}{dt} \right]$$

$$\mathcal{C}(\tau) = \kappa \bar{\lambda} K(\tau, \tau) + \int_{-\infty}^{\tau} du K(\tau - u, \tau - u) \mathcal{C}(u) + 2 \int_{0^+}^{\infty} du \int_{u^+}^{\infty} dr K(\tau + u, \tau + r) \mathcal{D}(u, r),$$

- And a similar closed equation for $\mathcal{D}(\cdot, \cdot)$, $\mathcal{C}(\cdot)$
- This allows one to do a GMM calibration

Calibration on 5 minutes US stock returns

- Using GMM as a starting point for MLE, we get for $K(s,t)$:

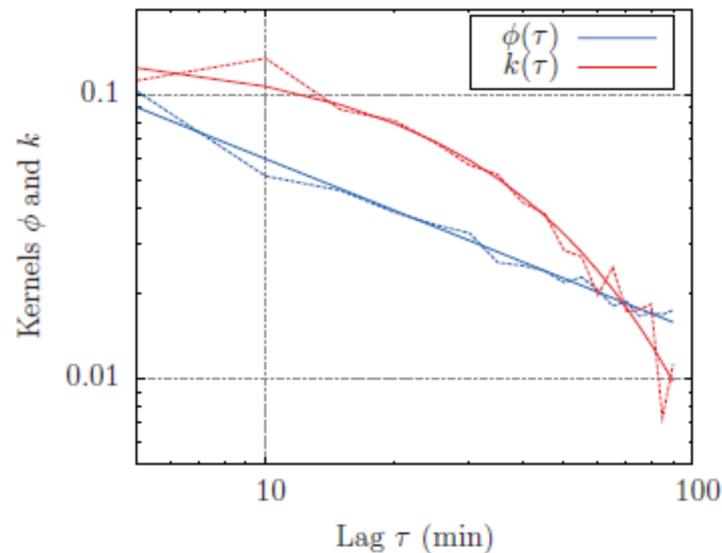


- K is well approximated by Diag + Rank 1:

$$K(\tau, \tau') \approx \phi(\tau)\delta_{\tau-\tau'} + k(\tau)k(\tau')$$

Calibration on 5 minutes US stock returns

$$K(\tau, \tau') \approx \phi(\tau)\delta_{\tau-\tau'} + k(\tau)k(\tau')$$



$$\phi(\tau) = g\tau^{-\alpha} \quad , \quad k(\tau) = k_0 \exp(-\omega\tau),$$
$$g = 0.09, \quad \alpha = 0.60, \quad k_0 = 0.14, \quad \omega = 0.15.$$

Generalized Hawkes processes: Hawkes + « ZHawkes »

$$K(\tau, \tau') \approx \phi(\tau)\delta_{\tau-\tau'} + k(\tau)k(\tau')$$

$$\lambda_t = \lambda_\infty + H_t + Z_t^2,$$

$$H_t := \int_{-\infty}^t \phi(t-s) dN_s, \quad Z_t = \frac{1}{\psi} \int_{-\infty}^t k(t-s) dP_s.$$

Z_t : moving average of price returns, i.e. recent « trends »

→ The Zumbach effect: trends increase future volatilities

The Markovian Hawkes + ZHawkes processes

$$\lambda_t = \lambda_\infty + H_t + Z_t^2,$$

$$H_t := \int_{-\infty}^t \phi(t-s) dN_s, \quad Z_t = \frac{1}{\psi} \int_{-\infty}^t k(t-s) dP_s.$$

With: $k(t) = \sqrt{2n_Z\omega} \exp(-\omega t)$ and $\phi(t) = n_H\beta \exp(-\beta t)$

In the continuum time limit: ($h = H$; $y = Z^2$):

$$dh = [- (1 - n_H) h + n_H (\lambda + y)] \beta dt$$

$$dy = [- (1 - n_Z) y + n_Z (\lambda + h)] \omega dt + [2 \omega n_Z y (\lambda + y + h)]^{1/2} dW$$

→ 2-dimensional generalisation of Pearson diffusions ($n_H = 0$)

→ The y process is asymptotically multiplicative, as assumed in many « log-vol » models (including Rough vols.)

The Markovian Hawkes + ZHawkes processes

$$dh = [- (1 - n_H) h + n_H (\lambda + y)]\beta dt$$

$$dy = [- (1 - n_Z) y + n_Z (\lambda + h)]\omega dt + [2 \omega n_Z y (\lambda + y + h)]^{1/2} dW$$

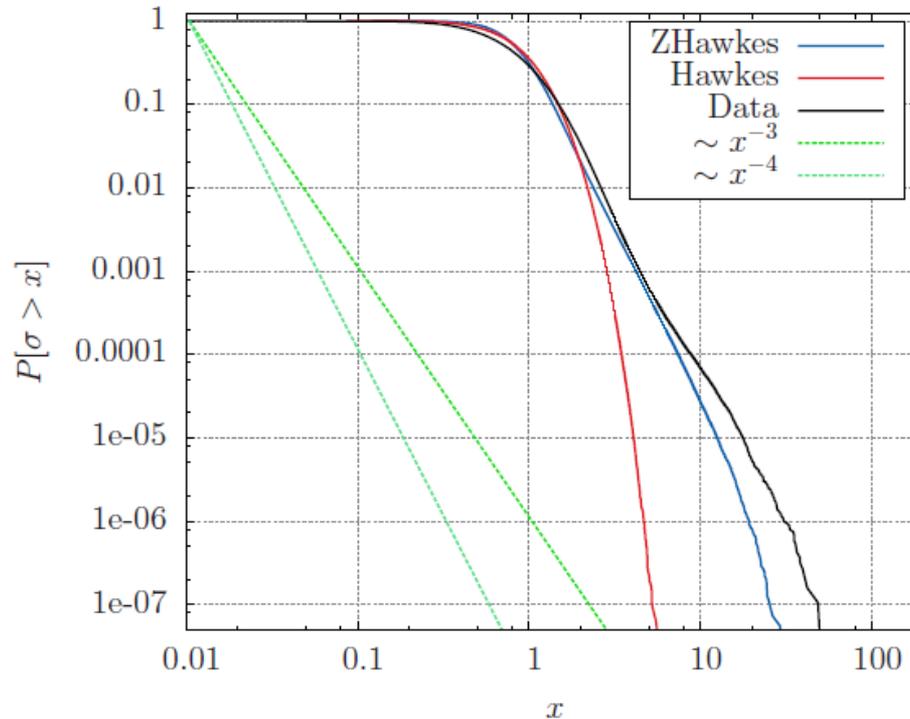
→ The upshot is that the vol/return distribution has a power-law tail with a computable exponent, for example:

$$\beta \gg \omega \rightarrow \nu = 1 + (1 - n_H)/n_Z$$

→ Even when n_Z is smallish, n_H conspires to drive the tail exponent ν in the empirical range ! – see next slide

→ Note: $n_Z > 1$ defines a stationary Hawkes process with infinite mean intensity!
(C. Aubrun, M. Benzaquen, JPB)

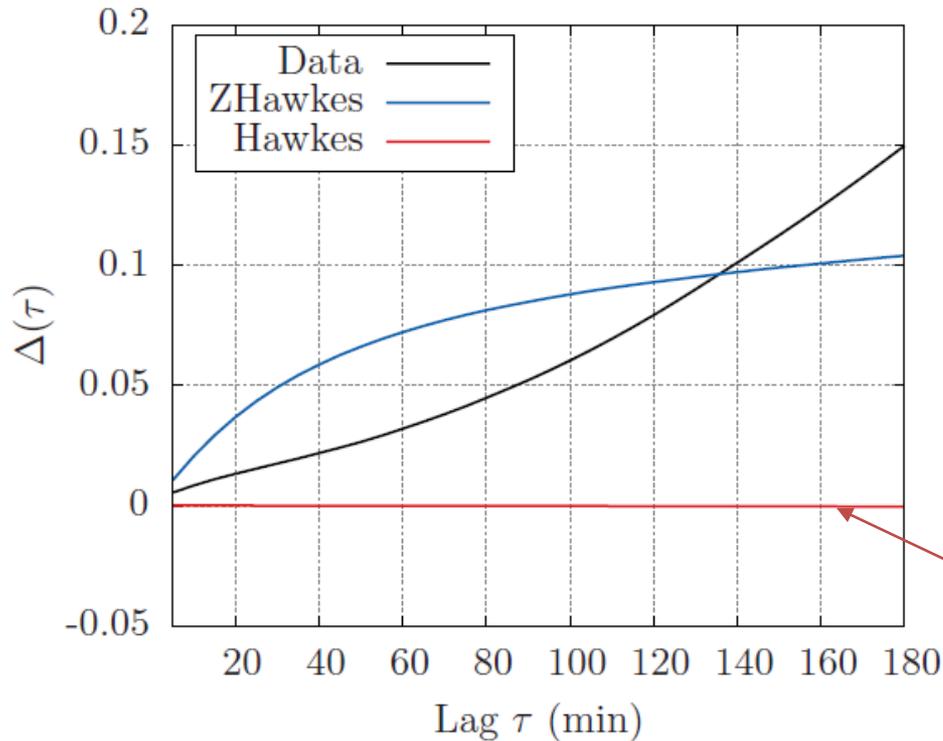
The calibrated Hawkes + ZHawkes process: numerical simulations



Fat-tails are indeed accounted for with $n_z = 0.06!$

Note: $\Delta P_\tau = \pm\psi$ so tails *do not* come from « residuals »

The calibrated Hawkes + ZHawkes process: numerical simulations



$$\Delta(\tau) = \frac{\sum_{\tau'=1}^{\tau} [C(\tau') - C(-\tau')]}{2 \sum_{\tau'=1}^q \max(|C(\tau')|, |C(-\tau')|)}$$

where C is the cross-correlation between σ_{HF} and $|r|$

Close to zero!

The level of TRA is also satisfactorily reproduced

(wrong concavity probably due to intraday non-stationarities not accounted for here)

Generalisation: order book activity & liquidity crises

* Fosset, JPB, Benzaquen, 2021

$$\lambda_t = \alpha_0 + \int_0^t \phi(t-s) dN_s + \int_0^t L(t-s) dP_s + \int_0^t \int_0^t K(t-s, t-u) dP_s dP_u$$

Diagram illustrating the components of the equation:

- 6-vectors (pointing to λ_t and α_0)
- 6 x 6 matrix (pointing to $\phi(t-s)$)
- 6-vectors (pointing to $L(t-s)$ and $K(t-s, t-u)$)

- Consider the two best limits and 6 event-types: MO, LO, CA, described by a 6-dimensional rate vector λ_t (\rightarrow 3 by symmetry)
- These rates depend on past events $d\mathbf{N}$ and past price changes dP
- The second term is a Hawkes feedback (bid/ask symmetric)
- The third term is a « leverage » feedback (bid/ask antisymmetric)
- The last term couples past volatility $K(u,u)$ and past trends $K(u,v)$ to present rates (bid/ask symmetric) – cf. the Zumbach effect

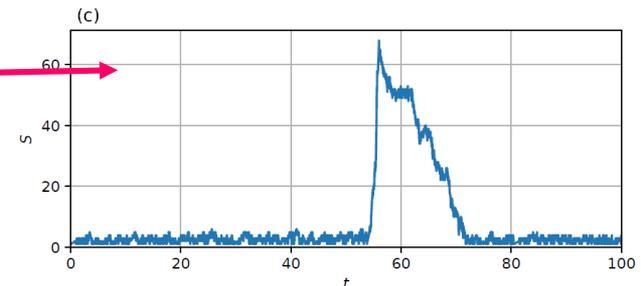
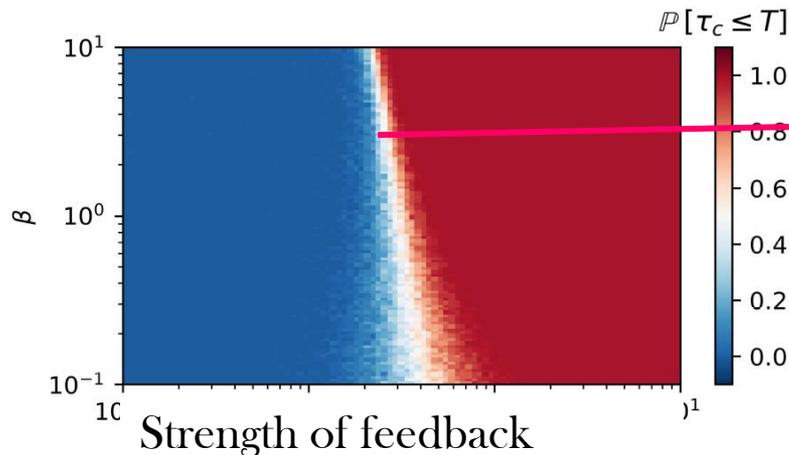
Generalisation: order book activity & liquidity crises

* Fosset, JPB, Benzaquen, 2021

$$\lambda_t = \alpha_0 + \int_0^t \phi(t-s) dN_s + \int_0^t L(t-s) dP_s + \int_0^t \int_0^t K(t-s, t-u) dP_s dP_u$$

6-vectors 6 x 6 matrix 6-vectors

- Calibration on tick by tick data shows a clear influence of past trends and past volatility on event rates, which decrease the volume in the order book → a possible destabilising feedback loop!



Spread dynamics

Conclusion

- Generalized Hawkes Processes: a natural extension of Hawkes processes accounting for « trend » (Zumbach) effects on volatility – a step to close the gap between ABMs and stochastic models
 - Leads naturally to a multiplicative process for volatility
 - Accounts for tails (induced by micro-trends) and TRA
 - Adding the « Zumbach » term in Rough Vol. models leads to a very satisfactory model → see M. Rosenbaum
-

- A lot of work remaining (empirical and mathematical)
- Multivariate generalisation (C. Aubrun & M. Benaïm)
- Real « Micro » foundation ? Higher order terms ?