

A simplified unified propagator model for signed order flow, concave price impact and rough volatility

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21st Feb 2026

Abstract

In this note, we simplify the setup in [MORS26] by exogenously modelling signed order flow as the difference of two integrated hyper-rough Heston process with Hölder exponent in $(\frac{1}{2}, 1)$, which endogenously determines the price process, the propagator mechanism and the long-run permanent price impact. This obviates the need to use nearly unstable heavy-tailed Hawkes processes, and lengthy macroscopic scaling limit arguments with C -tightness.

1.1 Persistent order flow

Similar to Theorem 3.1 in [MORS26]¹, we model positive order flow F_t^+ for an asset as a weak solution the Stochastic Volterra Equation

$$F_t^+ = g(t) + \int_0^t f(t-s)B_{F_s^+}ds \quad (1)$$

for a standard Brownian motion B with $B_0 = 0$, where $f(t) = e^{-\theta t}t^{H-\frac{1}{2}}$ with $H \in (-\frac{1}{2}, \frac{1}{2})$, $\theta > 0$ so $f \in L^1$ and $g(t) = V_0 t$ for some $V_0 > 0$ (note θ replaces the ergodicity parameter λ in Theorem 3.1 in [MORS26]) (weak existence and uniqueness for (1) is established in section 2 of [AJ21]). F_t^+ is a.s. $(2\alpha - \varepsilon) \wedge 1$ -Hölder continuous where $\alpha = H + \frac{1}{2}$ (see e.g. Theorem 3.1 in [JM20] for details)² and F^+ is an increasing process which is non-differentiable if $H \leq 0$.

Lemma 1.1 $\mathbb{E}(F_t^+) < \infty$.

Proof. See Appendix A. ■

Corollary 1.2 $B_{F_t^+}$ is an \mathcal{G}_t^+ -martingale, where $\mathcal{G}_t^+ = \mathcal{F}_{F_t^+}^B$.

Proof. Let $S = F_s^+$ and $T = F_t^+$ for $0 \leq s \leq t$, then T and S are \mathcal{F}^B -stopping times with $0 \leq S \leq T$. To see this, consider two Brownian paths ω, ω' with $\omega(r) = \omega(r')$, then $F_t^+(\omega) \leq u \Rightarrow F_t^+(\omega') \leq u$. By the contrapositive, $F_t^+(\omega') > u \Rightarrow F_t^+(\omega) \geq u$, so $F_t^+(\omega) > u \Rightarrow F_t^+(\omega') \geq u$, hence $1_{F_t^+(\omega) \leq u} = 1_{F_t^+(\omega') \leq u}$, so $\{F_t^+(\omega) \leq u\} \in \mathcal{F}_u^B$.

Then $\mathbb{E}(B_{u \wedge T}^2) = \mathbb{E}(u \wedge T) \leq \mathbb{E}(T) = \mathbb{E}(F_t^+) < \infty$ (from the previous lemma), so $(B_{u \wedge T})$ is bounded in L^2 (and hence $M_u = B_{u \wedge T}$ is an \mathcal{F}_u^B -martingale with $\mathbb{E}(M_u^2) = \mathbb{E}(u \wedge T) \leq \mathbb{E}(T) < \infty$ and hence is UI) so from the OST for U.I. martingales applied to M (e.g. Theorem 5.20 in [Eth18]), $\mathbb{E}(B_T | \mathcal{F}_S^B) = B_S$, and the martingale property follows. ■

1.2 The inversion formula

If h is a function with $h * f \equiv 1$, then we have the inversion formula:

$$\begin{aligned} \int_0^t h(t-s)(F_s^+ - g(s))ds &= \int_0^t h(t-s) \int_0^s f(s-u)B_{F_u^+}duds = \int_0^t \int_u^t h(t-s)f(s-u)dsB_{F_u^+}du \\ &= \int_0^t \int_0^{t-u} h(t-s-u)f(s)dsB_{F_u^+}du \\ &= \int_0^t (h * f)(t-u)B_{F_u^+}du = \int_0^t B_{F_u^+}du \end{aligned}$$

(where Fubini is justified since $F_t^+ < \infty$ a.s. for finite t and B is a.s. continuous), and hence

$$B_{F_t^+} = \frac{d}{dt} \int_0^t h(t-s)(F_s^+ - g(s))ds = \frac{d}{dt} \int_0^t h(t-s)F_s^+ ds - \frac{d}{dt} \int_0^t h(t-s)g(s)ds \quad (2)$$

¹see also section 5.4 in [FGS21]

²we can simulate F^+ using the Monte Carlo scheme in [AA25] using Normal Inverse Gaussian variates

Lebesgue a.e., where the final equality holds since $\int_0^t h(t-s)(F_s^+ - g(s))ds$ and $\int_0^t h(t-s)g(s)ds$ are both absolutely continuous, and hence so is their sum. Moreover, setting $H(t) = \int_0^t h(s)ds$, integrating by parts we see that

$$\int_0^t H(t-s)dF_s^+ = F_s^+ H(t-s)_{s=0}^{s=t} + \int_0^t h(t-s)F_s^+ ds = \int_0^t h(t-s)F_s^+ ds.$$

Taking Laplace transforms, the condition $h * f \equiv 1$ becomes $\hat{f}(\lambda)\hat{h}(\lambda) = \frac{1}{\lambda}$, from which we find that

$$h(t) = \frac{\theta^\alpha}{\Gamma(\alpha)}(1 - \frac{\Gamma(-\alpha, t\theta)}{\Gamma(-\alpha)})$$

where $\Gamma(a, z) = \int_z^\infty s^{a-1}e^{-s}ds$ is the incomplete Gamma function, and $h(t) = O(t^{-H-\frac{1}{2}})$ as $t \rightarrow 0$.

1.3 Signed order flow

Again following [MORS26], we now assume the *signed* order flow for an asset is

$$V_t = F_t^+ - F_t^-$$

where $F_t^- = g(t) + \int_0^t f(t-s)W_{F_s^-}ds$ and W is another Brownian motion independent of B (so F^- is an i.i.d. copy of F^+). Then we see that

$$\frac{d}{dt} \int_0^t h(t-s)V_s ds = \frac{d}{dt} \int_0^t H(t-s)dV_s = \int_0^t h(t-s)dV_s = B_{F_t^+} - W_{F_t^-} \quad (3)$$

since the g terms cancel (and the second equality follows from the Stieltjes-Leibniz rule using the known Hölder continuity of F^\pm and that $|h(u)| \leq Cu^{-\alpha}$ so $\int_0^t |h(t-s)||dV_s| < \infty$), and for $0 \leq t \leq u$ we have

$$\mathbb{E}(F_u^+ | \mathcal{F}_{F_t^+}^B) = g(u) + \mathbb{E}\left(\int_0^u f(u-s)B_{F_s^+}ds | \mathcal{G}_t^+\right) = g(u) + \int_0^t f(u-s)B_{F_s^+}ds + B_{F_t^+} \int_t^u f(u-s)ds$$

(and similarly for $\mathbb{E}(F_u^- | \mathcal{F}_{F_t^-}^W)$), where $\mathcal{G}_t^+ = \mathcal{F}_{F_t^+}^B$, since $(B_{F_t^+})_{t \geq 0}$ is a \mathcal{G}_t^+ -martingale by Corollary 1.2.

1.4 Price dynamics and the propagator model

Making the usual assumption that the asset price $P_t = \kappa \lim_{u \rightarrow \infty} \mathbb{E}(F_u | \mathcal{G}_t)$ where $F_t = F_t^+ - F_t^-$ and $\mathcal{G}_t = \sigma(\mathcal{G}_t^+, \mathcal{G}_t^-)$ with $\mathcal{G}_t^- = \mathcal{F}_{F_t^-}^W$ (for some $\kappa > 0$)³, we find that $f(u-s) \rightarrow 0$ as $u \rightarrow \infty$ since $\theta > 0$, but $\lim_{u \rightarrow \infty} \int_t^u f(u-s)ds = c_{H,\theta}$ where $c_{H,\theta} = \theta^{-\frac{1}{2}-H}\Gamma(\frac{1}{2}+H)$, so

$$P_t = \kappa \lim_{u \rightarrow \infty} \mathbb{E}(F_u^+ - F_u^- | \mathcal{G}_t) = \kappa c_{H,\theta}(B_{F_t^+} - W_{F_t^-})$$

i.e. P is the difference of two i.i.d. hyper-rough Heston models (each with correlation $\rho = 1$ since each process only has one driving Brownian motion), and we recover the Propagator equation

$$P_t = \int_0^t G(t-s)dV_s$$

where $G(t) = \kappa c_{H,\theta}h(t)$, and the *unsigned* order flow is $U_t = F_t^+ + F_t^-$.

1.5 Power-law price impact

In particular, the *market impact function* of an exogenous metaorder executed at constant trading speed 1 up to time t_0 is given by

$$MI(t) = \frac{d}{dt} \int_0^t h(t-s)sds = \frac{\theta^{-\alpha_-}(t\theta\Gamma(-\alpha) - t\theta\Gamma(-\alpha, t\theta) - \Gamma(\alpha_-, 0) + \Gamma(\alpha_-, t\theta))}{\Gamma(-\alpha)\Gamma(\alpha)}$$

for $0 \leq t < t_0$ (where $\alpha_- = \frac{1}{2}-H$) for $0 \leq t \leq t_0$ which is $O(t^{\frac{1}{2}-H})$ as $t \rightarrow 0$ (for $H \neq \frac{1}{2}$) and globally concave in t for $H \in (-\frac{1}{2}, \frac{1}{2})$, consistent with empirical evidence (note the usual square-root impact law corresponds to $H = 0$ here), and we can show that the asymptotic (i.e. permanent) impact as $t \rightarrow \infty$ (given no trading after t_0) is

$$MI(\infty) = \frac{t_0\theta^\alpha}{\Gamma(\frac{1}{2}+H)}$$

³note we are assuming $P_0 = 0$ without loss of generality to ease notation, but we can easily add a positive P_0 term

(this is the asymptotic grey line in the plot below).

For the special case $H = \frac{1}{2}$ that we have previously excluded, $h(t) = \theta$, which corresponds to linear permanent price impact; conversely $\lim_{\theta \rightarrow 0} h(t) = \text{const.} \times t^{-\frac{1}{2}-H}$. Recall from above that F_t^+ has Hölder regularity $(2\alpha-\varepsilon)\wedge 1$, and empirical evidence in [MORS26] and below suggests that $2\alpha \in (\frac{1}{2}, 1)$ which corresponds to $H \in (-\frac{1}{4}, 0)$.

1.6 Controlling liquidity

At-the-moment, we have no control over the level of liquidity in the model above, since the kernel $G(t)$ which controls the amount of transient price impact is endogenously determined via the f function in (1). To resolve this inflexibility, we can make the natural assumption that $P_t = \theta M_t + (1-\theta) \int_0^t G(t-s)dV_s$ for $\theta \in [0, 1]$ where M is an additional martingale, so θ directly controls market liquidity.

1.7 Other extensions

If we augment f to a function of the form $f(t) = e^{-\theta t} t^{H-\frac{1}{2}} + e^{-\theta_2 t} t^{H_2-\frac{1}{2}}$, then we find that

$$h(t) = \frac{1}{\Gamma(\alpha)} \sum_{n=0}^{\infty} (-c)^n \sum_{k=0}^{\infty} \frac{(\beta n)_k}{k!} \delta^k \theta^{-(\mu_n+k)} \frac{\gamma(\mu_n+k, \theta t)}{\Gamma(\mu_n+k)}$$

where $\beta = \frac{1}{2} + H_2$, $c = \Gamma(\beta)/\Gamma(\alpha)$, $\delta = \theta - \theta_2$, $\mu_n = \beta n - \alpha(n+1)$, $\gamma(a, z) = \int_0^z s^{a-1} e^{-s} ds$ is the lower incomplete Gamma function and $(a)_k$ denotes the Pochhammer symbol. We can then use two (possibly different) f functions to drive F^\pm , and if $H_2 \in (0, \frac{1}{2})$, P_t will have a conventional (i.e. non-hyper) rough component, which may be more realistic. We can also assume W and B correlated which gives greater flexibility in modelling skew for P_t .

1.8 Empirical results

Below we have tabulated estimates of 2α (i.e. the Hölder exponent of $(F_t)_{t \geq 0}$) using signed order flow for major US tech stocks using the model-independent $\hat{H}_{L,K}^\pi(X)$ estimator in Eq 7 in Cont&Das[CD24] which is consistent and allows for irregularly spaced data (with their $L/K = [\sqrt{N}]$ where N is the total sample size). For this we have used the free data at <https://data.lobsterdata.com/info/DataSamples.php>: ⁴

	GOOG	AAPL	MSFT	AMZN	INTC
N	49284	118326	410881	57360	404496
$2\hat{\alpha}$	0.692	0.720	0.847	0.735	0.890
\hat{H}	-0.154	-0.140	-0.0767	-0.133	-0.0553

⁴only using columns 4 and 5 from the lobsterdata.csv files which correspond to actual realized trades.

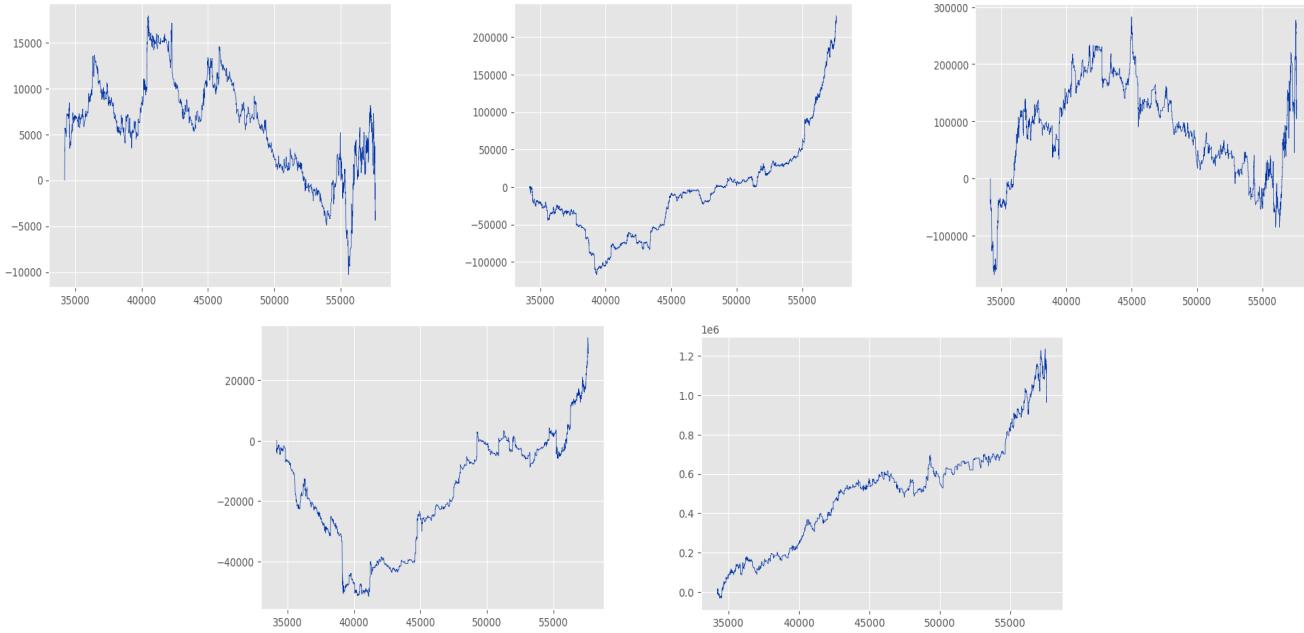


Figure 1: Signed order flow for GOOG, AAPL, MSFT, AMZN and INTC

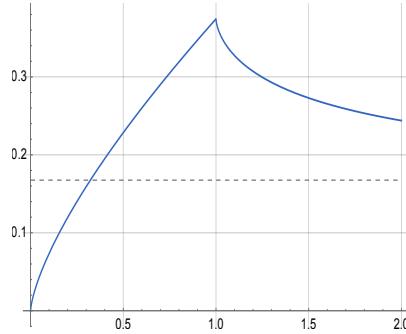


Figure 2: Concave price impact of an exogenous metaorder executed at constant trading speed 1 over $[0, 1]$ with $H = -0.2$, $\theta = 0.1$ and $\kappa = 1$. The blue line asymptotes to the constant level $MI(\infty)$ (grey line) as $t \rightarrow \infty$ which represents the asymptotic permanent price impact.

References

- [AJ21] Abi Jaber, E., “Weak existence and uniqueness for affine stochastic Volterra equations with L^1 -kernels”, *Bernoulli*, 27(3), 2021, 1583–1615
- [AA25] Abi Jaber, E. and E.Attal, “Simulating integrated Volterra square-root processes and Volterra Heston models via Inverse Gaussian”, preprint, 2025
- [CD24] Cont, R. and P.Das, “Rough Volatility: Fact or Artefact?”, *Sankhya: The Indian Journal of Statistics*, Series B, 86(1), 191-223, 2024.
- [Eth18] Etheridge, A., “Continuous martingales and stochastic calculus”, Lecture notes, 2018, <https://www.stats.ox.ac.uk/~etheridg/ctsmartingales.pdf>
- [FGS21] Forde, M., S.Gerhold and B.Smith, “Small-time, large-time and $H \rightarrow 0$ asymptotics for the rough Heston model”, *Mathematical Finance*, 31(1), 203-241, 2021.
- [JM20] Jusselin, P. and M.Rosenbaum, “No-arbitrage implies power-law market impact and rough volatility”, *Mathematical Finance*, Volume 30, Issue 4 pp. 1309-1336
- [KF70] Kolmogorov, A.N. and S.V.Fomin, *Introductory Real Analysis*, revised English ed., translated and edited by R.A.Silverman, Prentice-Hall, Englewood Cliffs, NJ, 1970.

[MORS26] Muhle-Karbe, J., Y.Chahdi, M.Rosenbaum and G.Szymanski, “A unified theory of order flow, market impact, and volatility”, preprint.

[Will91] Williams, D., “Probability with Martingales”, Cambridge Mathematical Textbooks, 1991

A Appendix A

Proof. Set $G_t := \sup_{0 \leq r \leq t} |g(r)|$, $K_t := \int_0^t |f(r)| dr > 0$. For $n \geq 1$ define the level hitting time $\tau_n := \inf\{s \geq 0 : F_s^+ \geq n\}$.

Since F^+ is continuous and nondecreasing, on $\{\tau_n \leq t\}$ we have $F_{\tau_n}^+ = n$ and $F_r^+ \leq n$ for all $0 \leq r \leq \tau_n$. Fix $n \geq 1$ and work on the event $\{\tau_n \leq t\}$. For $0 \leq r \leq \tau_n$ we have $|B_{F_r^+}| \leq \sup_{0 \leq u \leq n} |B_u|$. Setting $s = \tau_n$ and taking absolute values we see that

$$\begin{aligned} n &= g(\tau_n) + \int_0^{\tau_n} f(\tau_n - r) B_{F_r^+} dr \leq |g(\tau_n)| + \int_0^{\tau_n} |f(\tau_n - r)| |B_{F_r^+}| dr \\ &\leq G_t + \left(\int_0^{\tau_n} |f(\tau_n - r)| dr \right) \sup_{0 \leq u \leq n} |B_u| \\ &= G_t + \left(\int_0^{\tau_n} |f(s)| ds \right) \sup_{0 \leq u \leq n} |B_u| \leq G_t + K_t \sup_{0 \leq u \leq n} |B_u|. \end{aligned}$$

Hence for $n > G_t$, on $\{\tau_n \leq t\}$ we have $\{\sup_{0 \leq u \leq n} |B_u| \geq \frac{n-G_t}{K_t}\}$. Since $F_t^+ \geq n$ iff $\tau_n \leq t$, we obtain

$$\mathbb{P}(F_t^+ \geq n) \leq \mathbb{P}\left(\sup_{0 \leq u \leq n} |B_u| \geq \frac{n-G_t}{K_t}\right), \quad n > G_t.$$

By the standard reflection principle bound $\mathbb{P}(\sup_{0 \leq u \leq n} |B_u| \geq a) \leq 4 \exp(-\frac{a^2}{2n})$, for $a > 0$, so for $n > G_t$,

$$\mathbb{P}(F_t^+ \geq n) \leq 4 \exp\left(-\frac{(n-G_t)^2}{2nK_t^2}\right).$$

In particular, if $n \geq 4G_t$ then $(n-G_t)^2/n \geq \frac{1}{2}n$, hence

$$\mathbb{P}(F_t^+ \geq n) \leq 4e^{-\frac{n}{4K_t^2}}, \quad n \geq 4G_t. \quad (\text{A-1})$$

Since $F_t^+ \geq 0$, we also know that $\mathbb{E}(F_t^+) = \int_0^\infty \mathbb{P}(F_t^+ \geq x) dx$. Now set $m := \lceil 4G_t \rceil$. Then

$$\mathbb{E}(F_t^+) \leq m + \int_m^\infty \mathbb{P}(F_t^+ \geq x) dx = m + \sum_{n=m}^\infty \int_n^{n+1} \mathbb{P}(F_t^+ \geq x) dx \leq m + \sum_{n=m}^\infty \mathbb{P}(F_t^+ \geq n). \quad (\text{A-2})$$

Using (A-1) for all $n \geq m$,

$$\sum_{n=m}^\infty \mathbb{P}(F_t^+ \geq n) \leq \sum_{n=m}^\infty 4 \exp\left(-\frac{n}{4K_t^2}\right) < \infty. \quad (\text{A-3})$$

Combining (A-2) and (A-3) yields $\mathbb{E}(F_t^+) < \infty$. ■