

Local limit theorem in non-Gaussian quasi-likelihood inference

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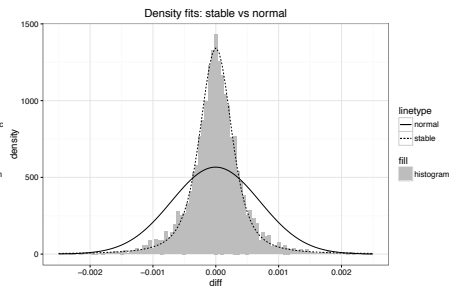
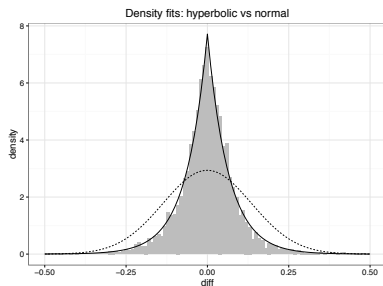
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● NYSE: IBM stock data (©CREST JST)



Primary objective: L^1 -local limit theorem

- **Standard β -stable Lévy process** $J = (J_t)$: $\mathcal{L}(J_1) = S_\beta$, i.e.

$$\mathbb{E}(e^{iuJ_t}) = e^{-t|u|^\beta} = \exp\left(t \int (\cos(uz) - 1) \frac{c_\beta}{|z|^{\beta+1}} dz\right)$$

- ▶ Scaling property: $h^{-1/\beta} J_h \sim S_\beta$, smooth density ϕ_β

- **Locally (small-time) standard β -stable Lévy process** $J = (J_t)$:

$$\mathcal{L}(h^{-1/\beta} J_h) \Rightarrow S_\beta, \quad h \rightarrow 0$$

- ▶ $\mathcal{L}(h^{-1/\beta} J_h)$ admits a b'dd. conti. density f_h (Bertoin and Doney, 1997)
- ▶ Roughly, e.g. the Lévy density: $z \mapsto \frac{c_\beta}{|z|^{\beta+1}} \{1 + o(1)\}$ for $|z| \rightarrow 0$

Easy conditions for “ $\lim_{h \rightarrow 0} \frac{1}{\sqrt{h}} \int |f_h(y) - \phi_\beta(y)| dy = 0$ ” ?

Why?

Inference for SDE from high-frequency data

$$dX_t = a(X_t, \alpha)dt + c(X_{t-}, \gamma)dJ_t$$

- Estimate true $\theta_0 = (\alpha_0, \gamma_0)$ from $(X_{t_j})_{j=0}^n$, where $t_j = jh_n$, $h_n := T/n$
- J is a locally β -stable pure-jump ($1 \leq \beta < 2$) Lévy process
 - ▶ Small-time non-Gaussian, infinite-activity and/or infinite-variation character
- Technical merits
 - ▶ Consistent estimation of stochastic location-scale structure over fixed time domain
 - ▶ Sidestep most stability constraints: stationarity, ergodicity, finite variance, ...

Stable quasi-likelihood: heuristic

- Euler (small-time) approximation:

$$\begin{aligned}\Delta_j X &:= X_{t_j} - X_{t_{j-1}} \stackrel{\mathbb{P}_\theta}{=} \int_{t_{j-1}}^{t_j} a(X_s, \alpha) ds + \int_{t_{j-1}}^{t_j} c(X_{s-}, \gamma) dJ_s \\ &\stackrel{\mathbb{P}_\theta}{\approx} a(X_{t_{j-1}}, \alpha) h_n + c(X_{t_{j-1}}, \gamma) (J_{t_j} - J_{t_{j-1}}) \\ &=: a_{j-1}(\alpha) h_n + c_{j-1}(\gamma) \Delta_j J\end{aligned}$$

- $\epsilon_{nj}(\theta) := \frac{\Delta_j X - a_{j-1}(\alpha) h_n}{h_n^{1/\beta} c_{j-1}(\gamma)} \sim \text{i.i.d. } S_\beta, \text{ approximately}$

Stable Quasi-Maximum Likelihood Estimator $\hat{\theta}_n = (\hat{\alpha}_n, \hat{\gamma}_n)$; **SQMLE**

$$\hat{\theta}_n = (\hat{\alpha}_n, \hat{\gamma}_n) \in \operatorname{argmax}_{\theta \in \Theta} \sum_{j=1}^n \log \left\{ \frac{1}{h_n^{1/\beta} c_{j-1}(\gamma)} \phi_\beta(\epsilon_{nj}(\theta)) \right\}$$

Asymptotic mixed normality of SQMLE

$$dX_t = a(X_t, \alpha)dt + c(X_{t-}, \gamma)dJ_t, \quad (X_{jT/n})_{j=0}^n$$

Theorem. Under some assumptions ($\beta \geq 1$; next slide),

$$\begin{pmatrix} n^{1/\beta-1/2}(\hat{\alpha}_n - \alpha_0) \\ \sqrt{n}(\hat{\gamma}_n - \gamma_0) \end{pmatrix} \xrightarrow{\mathcal{L}} MN(0, \text{diag}[\Sigma_{T,\alpha}(\theta_0)^{-1}, \Sigma_{T,\gamma}(\theta_0)^{-1}])$$

$$\Sigma_{T,\alpha}(\theta_0) := T^{2(1-1/\beta)} \frac{1}{T} \int_0^T \frac{\{\partial_\alpha a(X_t, \alpha_0)\}^{\otimes 2}}{c(X_t, \gamma_0)^2} dt \cdot \int \frac{\{\partial \phi_\beta(y)\}^2}{\phi_\beta(y)} dy,$$

$$\Sigma_{T,\gamma}(\theta_0) := \frac{1}{T} \int_0^T \frac{\{\partial_\gamma c(X_t, \gamma_0)\}^{\otimes 2}}{c(X_t, \gamma_0)^2} dt \cdot \int \frac{\{\phi_\beta(y) + y \partial \phi_\beta(y)\}^2}{\phi_\beta(y)} dy$$

- ▶ Masuda, H. (2017), Non-Gaussian quasi-likelihood estimation of SDE driven by locally stable Lévy process. arXiv:1608.06758 (v3)
- ▶ Asymptotically efficient in several cases (maybe in general).
 - ★ Locally stable Lévy process: Ivanenko, Kulik and M (2015)
 - ★ SDE: Clément and Gloter (2015); Clément, Gloter and Nguyen (2017)

Assumptions

$$dX_t = a(X_t, \alpha)dt + c(X_{t-}, \gamma)dJ_t, \quad (X_{jT/n})_{j=0}^n$$

① Regularity of the coefficients

- ▶ (a, c) smooth enough, with $a(\cdot, \alpha_0)$ and $c(\cdot, \gamma_0)$ globally Lipschitz.
- ▶ $\exists K \geq 0, \forall x, \sup_{\gamma} |c(x, \gamma)|^{-1} \leq K(1 + |x|)^K$

② Identifiability

- ▶ $\mathbb{P}_{\theta_0} \left\{ \left(\begin{array}{c} a(X_t, \alpha) \\ c(X_t, \gamma) \end{array} \right)_{t \leq T} = \left(\begin{array}{c} a(X_t, \alpha_0) \\ c(X_t, \gamma_0) \end{array} \right)_{t \leq T} \right\} = 1 \Rightarrow \theta = \theta_0$

③ **Driving-noise structure** for the pdf f_h of $\mathcal{L}(h^{-1/\beta} J_h)$

- ▶ $1 \leq \beta < 2$
- ▶ $\exists \epsilon > 0, \int |y|^{\beta-\epsilon} |f_h(y) - \phi_\beta(y)| dy \rightarrow 0$
- ▶ $\sqrt{n} \int |f_h(y) - \phi_\beta(y)| dy \rightarrow 0$ for the pdf f_h of $\mathcal{L}(h^{-1/\beta} J_h)$

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- How to verify in an easy manner?:

- ▶ $\exists \epsilon > 0, \int |y|^{\beta-\epsilon} |f_h(y) - \phi_\beta(y)| dy \rightarrow 0$

- ▶ $\sqrt{n} \int |f_h(y) - \phi_\beta(y)| dy \rightarrow 0$

- Here we forget that $1 \leq \beta < 2$, required to handle Euler-approx. error.

Sufficient conditions I

(L) J 's Lévy measure $\nu(dz) = g(z)dz$ for a symmetric g s.t.

- ▶ $g(z) = \frac{c_\beta}{|z|^{\beta+1}} \{1 + \rho(z)\}$ ($z \neq 0$);
- ▶ $\exists \delta, \epsilon_\rho > 0, \exists c_\rho \geq 0, \forall |z| \in (0, \epsilon_\rho], \quad |\rho(z)| \leq c_\rho |z|^\delta$

$$(L) + \sup_z |\rho(z)| < \infty \Rightarrow \exists \epsilon > 0, \quad \int |y|^{\beta-\epsilon} |f_h(y) - \phi_\beta(y)| dy \rightarrow 0$$

(G1) $\rho \in \mathcal{C}^1(\mathbb{R} \setminus \{0\})$ and the pair (c_ρ, β, δ) satisfies either

- ▶ $c_\rho = 0$, or
- ▶ $c_\rho > 0$ and $\delta > \beta$ with $|\rho(z)| + |z\partial\rho(z)| \leq c_\rho |z|^\delta$ ($z \neq 0$)

$$(L) + (G1) + \text{supp}(g) \subset [-K, K] \ (\exists K > 0) \Rightarrow \sqrt{n} \int |f_h(y) - \phi_\beta(y)| dy \rightarrow 0$$

- Additional conditions on ρ lose no generality through the localization.

Sufficient conditions II

(L) J 's Lévy measure $\nu(dz) = g(z)dz$ for a symmetric g s.t.

- ▶ $g(z) = \frac{c_\beta}{|z|^{\beta+1}} \{1 + \rho(z)\}$ ($z \neq 0$);
- ▶ $\exists \delta, \epsilon_\rho > 0, \exists c_\rho \geq 0, \forall |z| \in (0, \epsilon_\rho], \quad |\rho(z)| \leq c_\rho |z|^\delta$

(G2) $\psi_h \in C^1(\mathbb{R} \setminus \{0\})$, and for $\psi_h(u) := h \log \mathbb{E}(e^{iuh^{-1/\beta} J_h})$ and $\varphi_0(u) := e^{-|u|^\beta}$:

- ▶ $\exists c_\psi \geq 0, \quad |\partial_u \psi_h(u)| \lesssim \frac{1}{u} \vee u^{c_\psi} \quad (u > 0);$
- ▶ $\exists \epsilon_\psi(h) \rightarrow 0, \exists r \in [0, 1] \text{ s.t.}$
 - ★ $\int_{(0, \infty)} u^r \varphi_0(u) \left| \partial_u \psi_h(u) + \beta u^{\beta-1} \right| du \leq \epsilon_\psi(h)$
 - ★ $\sqrt{n}(\epsilon_\psi(h) \vee h^{a_\nu})^{\frac{\beta}{\beta+r}} \rightarrow 0 \text{ for } a_\nu := \begin{cases} 1 & (c_\rho = 0) \\ (\delta/\beta) \wedge 1 & (c_\rho > 0). \end{cases}$

$$(L) + (G2) + \sup_z |\rho(z)| < \infty \Rightarrow \sqrt{n} \int |f_h(y) - \phi_\beta(y)| dy \rightarrow 0$$

Two examples satisfying “(L) & (G2)”, but not “(L) & (G1)”

- **Exponentially tempered stable** with index $\beta \in [1, 2)$:

$$\psi_h(u) = \begin{cases} \frac{1}{\pi} \left\{ \lambda h \log \left(1 + \frac{u^2}{\lambda^2 h^2} \right) - 2u \arctan \left(\frac{u}{\lambda h} \right) \right\} & (\beta = 1) \\ 2c_\beta \Gamma(-\beta) \left[(\lambda^2 h^{2/\beta} + u^2)^{\beta/2} \cos \left\{ \beta \arctan \left(\frac{u}{\lambda h^{1/\beta}} \right) \right\} - \lambda^\beta h \right] & (\beta \in (1, 2)) \end{cases}$$

$$\triangleright \int_{(0, \infty)} \varphi_0(u) |\partial_u \psi_h(u) + \beta u^{\beta-1}| du \lesssim h^{1 \wedge (1/\beta)} = h^{1/\beta}$$

- **Generalized hyperbolic** ($\beta = 1$) except for the variance gamma:

$$\partial_u \psi_h(u) + 1 = 1 - \frac{u}{\sqrt{(\eta h)^2 + u^2}} \frac{K_{\lambda+1}}{K_\lambda} \left(\frac{1}{h} \sqrt{(\eta h)^2 + u^2} \right)$$

$$\triangleright \int_{(0, \infty)} \varphi_0(u) |\partial_u \psi_h(u) + 1| du \lesssim \begin{cases} h \log(1/h) & (\lambda \neq -1/2) \\ h & (\lambda = -1/2 : \text{NIG case}) \end{cases}$$

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Summary

$$dX_t = a(X_t, \alpha)dt + c(X_{t-}, \gamma)dJ_t, \quad (X_{jT/n})_{j=0}^n \\ \mathcal{L}(h^{-1/\beta} J_h) \Rightarrow S_\beta$$

- Sufficient conditions for the **local limit theorems** ($1 \leq \beta < 2$):
 - ▶ $\exists \epsilon > 0, \int |y|^{\beta-\epsilon} |f_h(y) - \phi_\beta(y)| dy \rightarrow 0$
 - ▶ $\sqrt{n} \int |f_h(y) - \phi_\beta(y)| dy \rightarrow 0$ for the pdf f_h of $\mathcal{L}(h^{-1/\beta} J_h)$
- In terms of either
 - ▶ Lévy density $g(z)$ of J , or
 - ▶ Lévy-Khintchine exponent $u \mapsto h \log \mathbb{E}(e^{iuh^{-1/\beta} J_h})$
- For verifying the key assumptions in quasi-likelihood inference for (α, γ) .
 - ▶ Local limit results + Coefficients' regularity \Rightarrow Asymptotic mixed normality
 - ▶ No moment conditions and no ergodicity

Some references

- Masuda, H. (2017), Non-Gaussian quasi-likelihood estimation of SDE driven by locally stable Lévy process. [arXiv:1608.06758 \(v3\)](#)
- Bertoin, J. and Doney, R. A. (1997), Spitzer's condition for random walks and Lévy processes. *Ann. Inst. H. Poincaré Probab. Statist.* 33, 167–178.
- Clément, E. and Gloter, A. (2015), Local asymptotic mixed normality property for discretely observed stochastic differential equations driven by stable Lévy processes. *Stochastic Process. Appl.* 125, 2316–2352.
- Clément, E. and Gloter, A., and Nguyen, H. (2017), LAMN property for the drift and volatility parameters of a SDE driven by a stable Lévy process. Preprint: [hal-01472749](#)
- Ivanenko, D. O., Kulik, A. M. and Masuda, H. (2015), Uniform LAN property of locally stable Lévy process observed at high frequency. *ALEA* 12, 835–862.

