Central Limit Theorem for stationary Fleming-Viot particle systems in finite spaces

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Quasistationary distribution in finite spaces

- ► Let E be a finite space.
- ▶ Let $(x_t)_{t \ge 0}$ be the **continuous time** Markov chain with infinitesimal generator

$$Lf(x) = \sum_{y \in \mathsf{E}} p(x,y)[f(y) - f(x)], \qquad p(x,y) \ge 0, \quad \sum_{y \in \mathsf{E}} p(x,y) = 1.$$

▶ Let $D \subset E$ be nonempty, and let $\tau_D = \inf\{t \ge 0 : x_t \notin D\}$.

Quasistationary distribution (QSD)

A probability measure π on D is called **quasistationary** in D if

$$\forall t \geq 0, \qquad \mathbb{P}_{\pi}(\mathbf{x}_t \in \cdot | t < \tau_{\mathsf{D}}) = \pi(\cdot).$$

Perron-Frobenius Theorem

If the substochastic matrix $P_{\mathsf{D}} = \{p(x, y), x, y \in \mathsf{D}\}$ is **irreducible**, then:

- there is a unique QSD π ;
- the spectral radius $1 \lambda \in (0, 1]$ of P_D is a single eigenvalue;
- $\blacktriangleright P_{\mathsf{D}}^*\pi = (1-\lambda)\pi.$

 ${}^{1}P_{\mathsf{D}}$ is seen as an operator on the functions $\mathsf{D} \to \mathbb{R}$, and P_{D}^{*} as an operator on the measures on D .

Yaglom limit

In the sequel we always assume that P_{D} is irreducible.

Yaglom limit

Darroch, Seneta – J. Appl. Probab. '67: for any initial distribution μ on D,

$$\lim_{t \to +\infty} \mathbb{P}_{\mu}(\mathbf{x}_t \in \cdot | t < \tau_{\mathsf{D}}) = \pi(\cdot).$$

- ► The QSD is particularly relevant in the study of **metastability**, where convergence to the **Yaglom limit** occurs on a **shorter time scale** than **exit from** D.
- From Kramers Physica '40 to many works by people at CIRM this week!

Nontrivial computational issue: how to sample from the QSD?

- ▶ **Rejection** Monte-Carlo fails in almost surely finite time.
- ► $t \mapsto \mathbb{P}_{\mu}(\mathbf{x}_t \in \cdot | t < \tau_{\mathsf{D}})$ obeys a **nonlinear** evolution.
- Occupation measure-based algorithm proposed by Aldous, Flannery, Palacios Probab. Engrg. Inform. Sci. '88, see M. Benaïm's talk tomorrow.
- Particle system-based algorithm: Fleming–Viot particle system (Burdzy, Hołyst, March – Comm. Math. Phys. '00).

- ▶ Take *n* copies (**particles**) of the process $(x_t)_{t>0}$ started iid according to μ on D.
- ▶ When one attempts to exit from D, pick its next position uniformly among the positions of the n-1 remaining particles.



We get a well-defined Dⁿ-valued **exchangeable** continuous time Markov chain $(\mathbf{x}_t^1, \ldots, \mathbf{x}_t^n)_{t\geq 0}$, with empirical measure

$$\eta_t^n(x) = \frac{1}{n} \sum_{i=1}^n \mathbb{1}_{\{\mathbf{x}_t^i = x\}}, \qquad x \in \mathsf{D},$$

random process in the set $\mathcal{P}(\mathsf{D})$ of probability measures on D.

► The chain is irreducible: $\eta_t^n \to \eta_t^n$ in distribution (exponential/uniform rates in Cloez, Thai – Stoch. Proc. Appl. '16).

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$$\label{eq:eq:expansion} \begin{split} \eta^n_t(x) = \frac{1}{n} \sum_{i=1}^n \mathbbm{1}_{\{\mathbf{x}^i_t = x\}}, \qquad x \in \mathsf{D}, \end{split}$$

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LLN and CLT

What about the $n \to +\infty$ limit?

Laws of Large Numbers by Asselah, Ferrari, Groisman - J. Appl. Probab. '11.

- For all $t \ge 0$, $\eta_t^n \to \mathbb{P}_{\mu}(\mathbf{x}_t \in \cdot | t < \tau_{\mathsf{D}})$.
- For the stationary distribution: $\eta_{\infty}^n \to \pi$.

Central Limit Theorem by Cérou, Delyon, Guyader, Rousset – arXiv '16, '17, see M. Rousset's talk on Friday.

- ► For all $t \ge 0$, $\sqrt{n}(\eta_t^n \mathbb{P}_{\mu}(\mathbf{x}_t \in \cdot | t < \tau_D))$ converges in distribution to $\mathcal{N}(0, K_t^{\mu})$.
- ▶ If the system starts from the QSD, the covariance operator writes

$$\langle K_t^{\pi} f, f \rangle = \operatorname{Var}_{\pi}(f) + 2\lambda \int_{s=0}^t e^{2\lambda s} \operatorname{Var}_{\pi}(Q_s f) \mathrm{d}s, \qquad \langle \pi, f \rangle = 0,$$

where $Q_s f(x) = \mathbb{E}_x [f(\mathbf{x}_s) \mathbb{1}_{\{s < \tau_{\mathsf{D}}\}}].$

Extension to infinite time horizon not so straightforward.

Our result: stationary Central Limit Theorem.

- $\sqrt{n}(\eta_{\infty}^n \pi)$ converges in distribution to $\mathcal{N}(0, K)$.
- The covariance operator writes

$$\langle Kf, f \rangle = \operatorname{Var}_{\pi}(f) + 2\lambda \int_{s=0}^{\infty} e^{2\lambda s} \operatorname{Var}_{\pi}(Q_s f) ds, \qquad \langle \pi, f \rangle = 0.$$

FV particle system η_t^n





Conditional law
$$\mathbb{P}_{\mu}(\mathbf{x}_t \in \cdot | t < \tau_{\mathsf{D}})$$











- We aim to prove that $\sqrt{n}(\eta_{\infty}^n \pi) \to \mathcal{N}(0, K)$,
- in the space $\mathcal{M}_0(\mathsf{D})$ of signed measures ξ on D such that $\sum_{x \in \mathsf{D}} \xi(x) = 0$.
- We write the infinitesimal generator \mathbf{M}^n of the process $\sqrt{n}(\eta_t^n \pi)$ in $\mathcal{M}_0(\mathsf{D})$.
- We compute the limit $\overline{\mathbf{M}}$ of \mathbf{M}^n . \rightarrow Any limit of $\sqrt{n}(\eta_{\infty}^n - \pi)$ is thus a stationary distribution for $\overline{\mathbf{M}}$.
- We identify M as the infinitesimal generator of a linear diffusion process (*ξ*_t)_{t≥0} on M₀(D), whose unique stationary distribution is a Gaussian measure N(0, K).
- We solve a Lyapunov equation to compute K explicitly.
- **9** We prove that the sequence $\sqrt{n}(\eta_{\infty}^n \pi)$ is tight.

Step 1 of the proof: infinitesimal generators

► The law of η_{∞}^n is the stationary distribution of the $\mathcal{P}(\mathsf{D})$ -valued process $(\eta_t^n)_{t\geq 0}$ with infinitesimal generator

$$\mathbf{L}^{n}\phi(\eta) = \sum_{x,y\in\mathsf{D}} n\eta(x) \left(p(x,y) + q(x)\frac{n\eta(y)}{n-1} \right) \left[\phi\left(\eta + \frac{1}{n}\theta^{x,y}\right) - \phi(\eta) \right],$$

where
$$q(x) = \sum_{y \in \mathsf{E} \setminus \mathsf{D}} p(x, y)$$
 and $\theta^{x, y} = \mathbb{1}_y - \mathbb{1}_x$.

► The law of $\xi_{\infty}^n := \sqrt{n}(\eta_{\infty}^n - \pi)$ is the stationary distribution of the $\mathcal{M}_0(\mathsf{D})$ -valued process $(\xi_t^n)_{t\geq 0} := (\sqrt{n}(\eta_t^n - \pi))_{t\geq 0}$ with infinitesimal generator

$$\mathbf{M}^{n}\psi(\xi) = \sum_{x,y\in\mathsf{D}} n\left(\pi(x) + \frac{\xi(x)}{\sqrt{n}}\right) \left(p(x,y) + q(x)\frac{n}{n-1}\left(\pi(y) + \frac{\xi(y)}{\sqrt{n}}\right)\right)$$
$$\times \left[\psi\left(\xi + \frac{1}{\sqrt{n}}\theta^{x,y}\right) - \psi(\xi)\right].$$

Step 2 of the proof: taking the limit

Take $\psi \in C^{\infty}_{c}(\mathcal{M}_{0}(\mathsf{D}))$. When $n \to +\infty$,

 $\mathbf{M}^{n}\psi(\xi) \to \overline{\mathbf{M}}\psi(\xi) = \langle ((P_{\mathsf{D}}^{\pi})^{*} - (1-\lambda)I)\xi, \nabla\psi\rangle + A_{\mathsf{D}}^{\pi} :: \nabla^{2}\psi(x), \quad \text{uniformly in } \xi,$

with the following notation:

 $\triangleright P_{\rm D}^{\pi}$ is the stochastic matrix with coefficients

 $p_{\mathsf{D}}^{\pi}(x,y) = p(x,y) + q(x)\pi(y), \qquad x, y \in \mathsf{D},$

which defines the π -return process.

- The π -return process describes the **pathwise limit** of the (stationary) Fleming–Viot particle system, it is irreducible and ergodic with respect to π .
- $A_{\rm D}^{\pi}$ is the symmetric operator defined by

$$\langle A_{\mathsf{D}}^{\pi}f,f\rangle = \frac{1}{2} \sum_{x,y \in \mathsf{D}} \pi(x) p_{\mathsf{D}}^{\pi}(x,y) [f(y) - f(x)]^2.$$
 (energy / Dirichlet form)

Consequence: assume tightness, so that $\xi_{\infty}^n \to \overline{\xi}_{\infty}$ (up to a subsequence). Then

$$0 = \mathbb{E}\left[\mathbf{M}^{n}\psi(\boldsymbol{\xi}_{\infty}^{n})\right] = \mathbb{E}[\underbrace{(\mathbf{M}^{n} - \overline{\mathbf{M}})\psi(\boldsymbol{\xi}_{\infty}^{n})}_{\rightarrow 0 \quad \text{uniformly}}] + \mathbb{E}\left[\overline{\mathbf{M}}\psi(\boldsymbol{\xi}_{\infty}^{n})\right] \rightarrow \mathbb{E}[\overline{\mathbf{M}}\psi(\overline{\boldsymbol{\xi}}_{\infty})],$$

therefore the law of $\overline{\xi}_\infty$ is a stationary distribution for $\overline{\mathbf{M}}.$

Step 3 of the proof: identification of the limit

The operator

$$\overline{\mathbf{M}}\psi(\xi) = \langle ((P_{\mathsf{D}}^{\pi})^* - (1-\lambda)I)\xi, \nabla\psi \rangle + A_{\mathsf{D}}^{\pi} :: \nabla^2\psi(x)$$

is the infinitesimal generator of the linear diffusion process $(\overline{\xi}_t)_{t>0}$ on $\mathcal{M}_0(\mathsf{D})$

$$\mathrm{d}\overline{\xi}_t = B_0 \overline{\xi}_t \mathrm{d}t + \Sigma \mathrm{d}\mathrm{w}_t,$$

with $(w_t)_{t\geq 0}$ BM in \mathbb{R}^k , $\Sigma: \mathbb{R}^k \to \mathcal{M}_0(\mathsf{D})$ and $B_0: \mathcal{M}_0(\mathsf{D}) \to \mathcal{M}_0(\mathsf{D})$ defined by

$$\frac{1}{2}\Sigma\Sigma^* := A_{\mathsf{D}}^{\pi} \quad \text{and} \quad B_0 = (P_{\mathsf{D}}^{\pi})^* - (1-\lambda)I.$$

- ▶ By irreducibility of the π -return process, A_{D}^{π} is **positive definite** on $\mathcal{M}_0(\mathsf{D})$.
- There exists $\gamma > 0$ such that any eigenvalue $\tau \in \mathbb{C}$ of B_0 satisfies $\operatorname{Re} \tau \leq -\gamma$.

Consequence

The **unique** stationary distribution of $\overline{\mathbf{M}}$ is the **centered Gaussian measure** on $\mathcal{M}_0(\mathsf{D})$ with covariance operator *K* defined as the unique solution to the **Lyapunov equation**

$$B_0 K + K B_0^* + 2A_{\mathsf{D}}^\pi = 0.$$

- Uniqueness follows from uniform ellipticity.
- Linearity of evolution preserves Gaussian measures.
- Lyapunov equation is Ito's formula for the covariance of (ξ_t)_{t≥0}.
- Existence of a solution for Lyapunov equation follows from spectral stability of the drift.

Step 4 of the proof: computation of the covariance

► The solution K to Lyapunov equation $B_0K + KB_0^* + 2A_D^{\pi} = 0$ is known to write

$$K = 2 \int_{s=0}^{+\infty} e^{sB_0} A_{\mathsf{D}}^{\pi} e^{sB_0^*} \mathrm{d}s.$$

• Take $f : \mathsf{D} \to \mathbb{R}$ such that $\langle \pi, f \rangle = 0$, and compute

$$\langle Kf, f \rangle = 2 \int_{s=0}^{+\infty} \langle A_{\mathsf{D}}^{\pi} \mathrm{e}^{sB_0^*} f, \mathrm{e}^{sB_0^*} f \rangle \mathrm{d}s.$$

- Since B₀ = (P^π_D)* − (1 − λ)I, e^{sB^{*}₀}f = e^{λs}e^{sL^π_D}f = e^{λs}P^π_{s,D}f, where:
 L^π_D = P^π_D − I is the infinitesimal generator of the π-return process,
 P^π_{s,D} = e^{sL^π_D} is the semigroup of the π-return process.
- ► As a consequence,

$$\begin{split} \langle Kf, f \rangle &= 2 \int_{s=0}^{+\infty} \mathrm{e}^{2\lambda s} \underbrace{\langle A_{\mathsf{D}}^{\pi} P_{s,\mathsf{D}}^{\pi} f, P_{s,\mathsf{D}}^{\pi} f \rangle}_{-\frac{1}{2} \frac{\mathrm{d}}{\mathrm{d}s} \mathrm{Var}_{\pi}(P_{s,\mathsf{D}}^{\pi} f)} \, \mathrm{d}s \\ &= \mathrm{Var}_{\pi}(f) + 2\lambda \int_{s=0}^{+\infty} \mathrm{e}^{2\lambda s} \mathrm{Var}_{\pi}(P_{s,\mathsf{D}}^{\pi} f) \mathrm{d}s \end{split}$$

 $\models \langle \pi, f \rangle = 0 \text{ then ensures that } P_{s,\mathsf{D}}^{\pi}f(x) = Q_sf(x) = \mathbb{E}_x[f(\mathbf{x}_s)\mathbb{1}_{\{s < \tau_{\mathsf{D}}\}}].$

Step 5 of the proof: tightness of ξ_{∞}^n

- ► Recall that we denote by Lⁿ the infinitesimal generator of the empirical distribution (η^t_t)_{t≥0} of the FV particle system, so that E[Lⁿφ(ηⁿ_∞)] = 0.
- ► Take $\phi(\eta) = \frac{1}{2} \langle \eta \pi, R(\eta \pi) \rangle$, where R is a symmetric operator.
- Little algebra yields the inequality

 $\mathbb{E}\left[\langle -B'[\mathfrak{\eta}_{\infty}^{n}]\xi_{\infty}^{n},R\xi_{\infty}^{n}\rangle\right] \leq C(R), \qquad \xi_{\infty}^{n}=\sqrt{n}(\mathfrak{\eta}_{\infty}^{n}-\pi),$

where $B'[\eta] := B_0 + \langle \eta - \pi, q \rangle$, $q(x) = \sum_{y \in \mathsf{E} \setminus \mathsf{D}} p(x, y)$.

- Quadratic part consistent with the limit $\mathbf{M}^n \psi \to \overline{\mathbf{M}} \psi$,
- cubic nonlinearity originates from the fact that ϕ is not bounded.

If there were no cubic nonlinearity, then:

- ► take $R = N^{-1}$ where N solves the Lyapunov equation $B_0N + NB_0^* + 2I = 0$,
- ► this yields $\langle -B_0\xi, R\xi \rangle = -\frac{1}{2} \langle (N^{-1}B_0 + B_0^*N^{-1})\xi, \xi \rangle = \|N^{-1}\xi\|^2$,
- ▶ from which you deduce the variance control $c_{N-1}\mathbb{E}[\|\xi_{\infty}^n\|^2] \leq C(N^{-1}).$
- ▶ LLN on η_{∞}^{n} by Asselah, Ferrari, Groisman: with large probability,
 - $\langle \eta_{\infty}^n \pi, q \rangle$ is small,
 - the Lyapunov equation with $B'[\eta_{\infty}^n] := B_0 + \text{perturbation remains solvable.}$
- We get **tightness** but **no variance control**.

In our finite space setting, can we obtain uniform variance control of the form

$$\mathbb{E}\left[\|\eta_{\infty}^{n} - \pi\|^{2}\right] \leq \frac{C}{n} \qquad ?$$

Such an estimate is known (at least) for:

- diffusions with soft killing (Rousset SIAM J. Math. '06),
- discrete space Markov chains with strong mixing condition (Cloez, Thai Stoch. Proc. Appl. '16).

A possible approach: uniform and quantitative control of correlations.

Can we extend our CLT to more general Markov processes?

(ideally, the same level of generality as Cérou, Delyon, Guyader, Rousset - arXiv '16, '17)

• An easy conjecture: any limit of $\xi_{\infty}^n \in M_0(D)$ is a stationary distribution of the 'measure-valued' linear diffusion

$$\mathrm{d}\overline{\xi}_t = ((L_{\mathsf{D}}^{\pi})^* + \lambda)\overline{\xi}_t \mathrm{d}t + \mathrm{d}\mathrm{m}_t,$$

where:

- $L_{\rm D}^{\pi}$ is the infinitesimal generator of the π -return process,
- (m_t)_{t≥0} is a 'measure-valued' martingale with quadratic variation given by the Dirichlet form A[¯]_D of the π-return process.
- ▶ Do spectral properties of the π -return process still hold? What about tightness?
- How to make sense of all this?