

Quantitative mean-field limit for interacting branching diffusions

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Joint work with Felipe Muñoz

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January 2024

Outline

1. Motivation: Lotka-Volterra cross diffusion models
2. Mean-field interacting branching diffusions
3. Main result: convergence rate
4. Ideas of the proof: coupling and optimal transport

1. Motivation: cross diffusion models, local/non-local, Lotka-Volterra

2. Mean-field interacting branching diffusions

3. Main result

4. Idea of the proof

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- ▶ PDE model of **dispersive spatial interaction** between 2 species + competition

$$\begin{cases} \partial_t u - \Delta(d_1 u + a_{11}uu + a_{12}uv) = (r_1 - s_{11}u - s_{12}v)u, \\ \partial_t v - \Delta(d_2 v + a_{21}uv + a_{22}vv) = (r_2 - s_{22}v - s_{21}u)v. \end{cases} \quad (\text{SKT})$$

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- ▶ Global (weak) solutions: 📖 [Galiano, Garzón, Jüngel 2003], 📖 [Chen, Jüngel 2004-2006], 📖 [Chen, Daus, Jüngel 2018], 📖 [Chen, Jüngel, Wang 2022]. Entropic structure.

► Non-local SKT system introduced in [F', Méléard 2016] :

$$\begin{cases} \partial_t u - \Delta(d_1 u + a_{11}(u * G^{1,1})u + a_{12}(v * G^{1,2})u) = (r_1 - s_{11}u * C^{1,1} - s_{12}v * C^{1,2})u, \\ \partial_t v - \Delta(d_2 v + a_{21}(u * G^{2,1})v + a_{22}(v * G^{2,2})v) = (r_2 - s_{22}v * C^{2,2} - s_{21}u * C^{2,1})v. \end{cases}$$

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with $G^{i,j}$, $H^{i,j}$, $C^{i,j}$ regular kernels.

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Question 

Can we quantify the approximation of non-local SKT system by individual based models?

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A simpler population model

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Binary branching diffusions in \mathbb{R}^d (one species) with:

- › **mean-field interaction** in diffusion and drift
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- ›  Individuals carry independent reproduction clock of parameter $r > 0$:
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- ▶  (conditionally) independent killing clock of parameter cN_t^K/K for $c > 0$:
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- ▶ Between birth and death events the individual $X^{n,K}$ evolves as the diffusion process
$$dX_t^{n,K} = b(X_t^{n,K}, H * \mu_t^K(X_t^{n,K})) dt + \sigma(X_t^{n,K}, G * \mu_t^K(X_t^{n,K})) dB_t^n.$$

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Theorem (F' , Méléard '16, particular case)

Under Lipschitz regularity on σ, b, G , and H , moment assumptions on $(\mu_0^K)_K$ and weak convergence $\mu_0^K \rightarrow \mu_0 \in \mathcal{M}_+(\mathbb{R}^d)$, processes $(\mu_t^K)_{t \geq 0}$ converge weakly as $K \rightarrow \infty$ to $(\mu_t)_{t \geq 0}$, the unique weak solution of

$$\partial_t \mu_t = L_{\mu_t}^* \mu_t + (r - c \langle \mu_t, 1 \rangle) \mu_t,$$

with i.c. μ_0 , where

$$L_\mu f(x) = \frac{1}{2} \text{Tr} (a(x, G * \mu(x)) \text{Hess}(f)(x)) + b(x, H * \mu(x)) \cdot \nabla f(x).$$

Remarks:

- ▶ $(N_t^K / K)_{t \geq 0}$ converges to $n_t := \langle \mu_t, 1 \rangle$ solution of **logistic ODE** $\dot{n}_t = n_t(r - cn_t)$
- ▶ $\bar{\mu}_t := \mu_t / \langle \mu_t, 1 \rangle$ satisfies **nonlinear diffusion equation of McKean-Vlasov type**:

$$\partial_t \bar{\mu}_t = L_{\mu_t}^* \bar{\mu}_t$$

with i.c. $\bar{\mu}_0 := \mu_0 / \langle \mu_0, 1 \rangle$.

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Assumptions : same as before, plus:

- ▶ Conditionally on N_0^K , atoms of μ_0^K are i.i.d. $\sim \bar{\mu}_0 := \mu_0 / \langle \mu_0, 1 \rangle$, $\mu_0 \in \mathcal{M}_+(\mathbb{R}^d)$
- ▶ $\int_{\mathbb{R}^d} |x|^q \mu_0(dx) < \infty$ for some $q > 2$ and $\sup_K \mathbb{E}(\langle \mu_0^K, 1 \rangle^4) < \infty$.
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Theorem (F., Muñoz, 2022, EJP)

For all $K, T > 0$ one has

$$\sup_{t \in [0, T]} \mathbb{E}(\|\mu_t^K - \mu_t\|_{\text{BL}^*}) \leq C_T (I_4(K) + R_{q,d}(K))$$

where $C_T > 0$, $R_{q,d}(K)$ is an explicit polynomial function $\rightarrow 0$ as $K \rightarrow \infty$ and

$I_4(K) = \mathbb{E}(|\langle \mu_0^K, 1 \rangle - \langle \mu_0, 1 \rangle|^4)^{\frac{1}{4}}$ is a smaller term.

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$R_{q,d}(K)$ is explicit...

Where does the rate $R_{d,q}(K)$ come from?

Recall:

- ▶ **Wasserstein distance:** $p \in [1, \infty)$, p -Wasserstein distance $W_p(\mu, \nu)$ between $\mu, \nu \in \mathcal{P}(\mathbb{R}^d)$ defined by

$$W_p(\mu, \nu) = \left(\inf_{\pi \in \Pi(\mu, \nu)} \int_{\mathbb{R}^d \times \mathbb{R}^d} |x - y|^p \pi(\mathrm{d}x, \mathrm{d}y) \right)^{\frac{1}{p}}.$$

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For any $X \sim \mu$, $Y \sim \nu$ on common probability space, $W_p(\mu, \nu) \leq \mathbb{E}(|X - Y|^p)^{\frac{1}{p}}$
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 $(Z_i)_{i \geq 1}$ i.i.d. $\sim m \in \mathcal{P}(\mathbb{R}^d)$ with q -moment $M_q < \infty$, $q > 2$.


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 $(Z_i)_{i \geq 1}$ i.i.d. $\sim m \in \mathcal{P}(\mathbb{R}^d)$ with q -moment $M_q < \infty$, $q > 2$.
Then, for some $C_{d,q} > 0$ and all $N \in \mathbb{N} \setminus \{0\}$ FIXED,

$$\mathbb{E} \left(W_2^2 \left(\frac{1}{N} \sum_{i=1}^N \delta_{Z_i}, m \right) \right) \leq C_{d,q} M_q^{\frac{2}{q}} R_{d,q}^2(N).$$

Where does the rate $R_{d,q}(K)$ come from?

For all $K > 0$ one has

$$R_{d,q}(K) = \begin{cases} K^{-\frac{1}{4}} + K^{-\frac{(q-2)}{2q}}, & \text{if } d < 4 \text{ and } q \neq 4, \\ K^{-\frac{1}{4}} (\log(1+K))^{\frac{1}{2}} + K^{-\frac{(q-2)}{2q}}, & \text{if } d = 4 \text{ and } q \neq 4, \\ K^{-\frac{1}{d}} + K^{-\frac{(q-2)}{2q}}, & \text{if } d > 4 \text{ and } q \neq \frac{d}{(d-2)}, \end{cases}$$

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$$\begin{aligned}\|\mu - \nu\|_{\text{BL}^*} &\leq \langle \mu, 1 \rangle W_1(\bar{\mu}, \bar{\nu}) + |\langle \mu, 1 \rangle - \langle \nu, 1 \rangle| \\ &\leq \langle \mu, 1 \rangle W_2(\bar{\mu}, \bar{\nu}) + |\langle \mu, 1 \rangle - \langle \nu, 1 \rangle|\end{aligned}$$

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- Bound easily obtained at $t = 0$ from Fournier-Guillin's result and conditional independence of atoms of μ_0^K given mass $\frac{N_0^K}{K}$:

$$\begin{aligned}\mathbb{E} \left(\|\mu_0^K - \mu_0\|_{\text{BL}^*} \right) &\leq C \mathbb{E} \left(\frac{N_0^K}{K} W_2^2(\bar{\mu}_0^K, \bar{\mu}_0) \right)^{\frac{1}{2}} \mathbb{E} \left(\frac{N_0^K}{K} \right)^{\frac{1}{2}} + \mathbb{E} (|\langle \mu_0^K, 1 \rangle - \langle \mu_0, 1 \rangle|), \\ &\leq C R_{q,d}(K) + \mathbb{E} (|\langle \mu_0^K, 1 \rangle - \langle \mu_0, 1 \rangle|)\end{aligned}$$

Basic ingredients

- If $\bar{\mu}, \bar{\nu} \in \mathcal{P}(\mathbb{R}^d)$ are normalized versions of $\mu, \nu \in \mathcal{M}(\mathbb{R}^d)$, then

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No longer true for $t > 0$!!!

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 - ii) For each $t \geq 0$, conditionally on $\langle \nu_t^K, 1 \rangle$, atoms of ν_t^K are i.i.d. of law $\bar{\mu}_t$ solving $\partial_t \bar{\mu}_t = L_{\mu_t}^* \bar{\mu}_t$.
- Then, result boils down to control \mathbb{E} of :

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- ▶ **To grant ii)** atoms of $\nu_t^K = \frac{1}{K} \sum_{n=1}^{N_t^K} \delta_{Y_t^{n,K}}$ must be **independent diffusions**

$$dY_t^{n,K} = b(Y_t^{n,K}, H * \mu_t(Y_t^{n,K})) dt + \sigma(Y_t^{n,K}, G * \mu_t(Y_t^{n,K})) dB_t^n.$$

so that $\text{Law}(Y_t^{n,K}) = \bar{\mu}_t$, and we must chose $Y_\tau^{n,K} \sim \bar{\mu}_\tau$ at its birth time τ .

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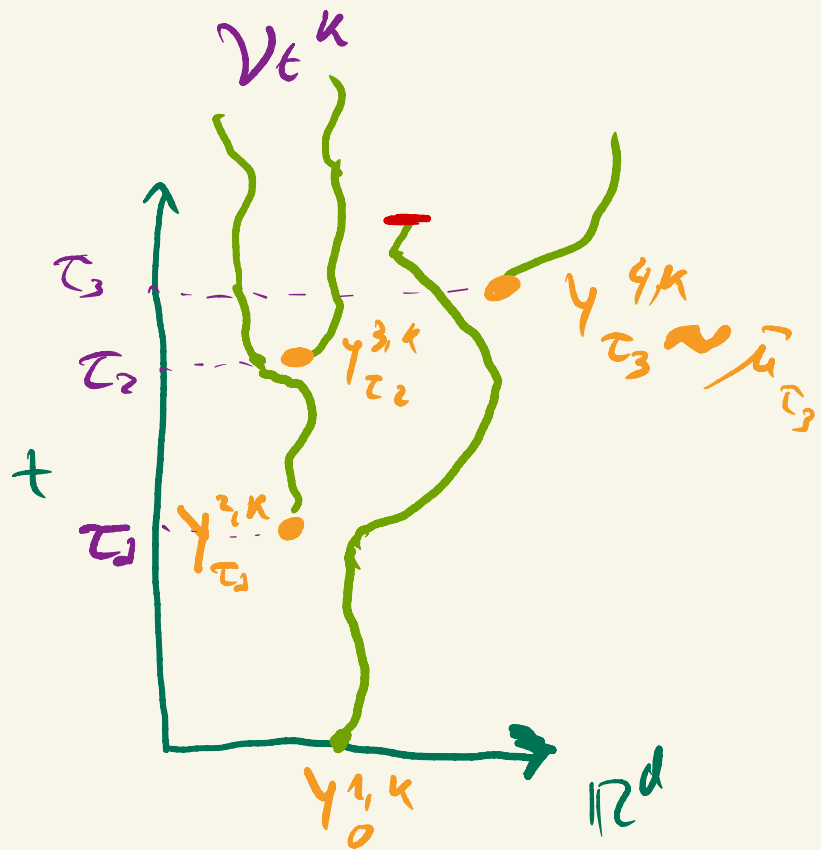
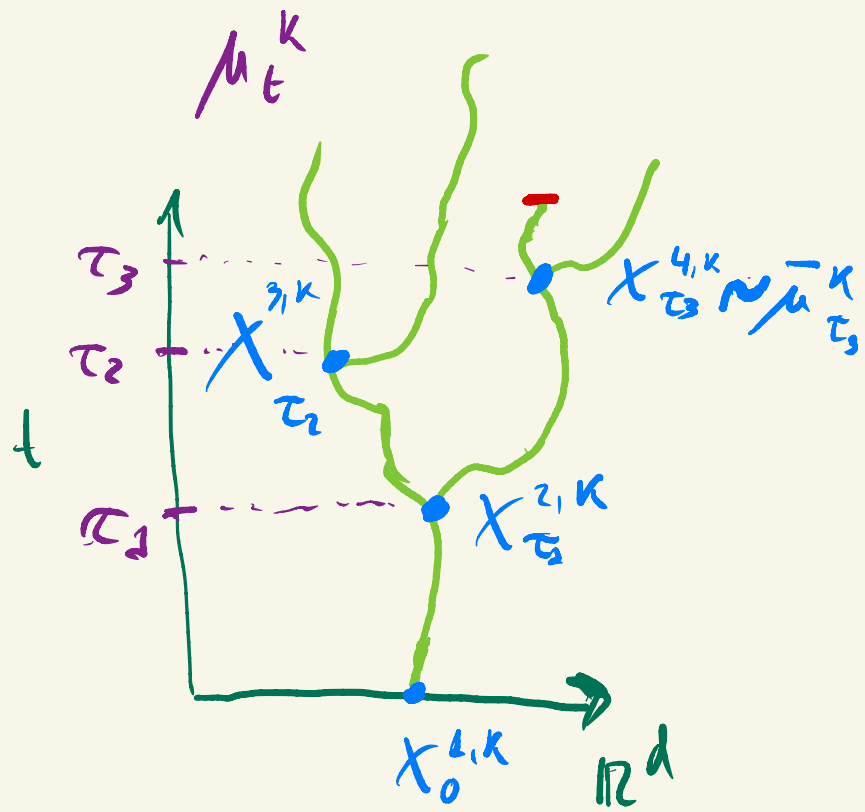
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- ▶ Drawing
- ▶ Triangle ineq. with $\bar{\nu}_{\tau-}^K$ + Gronwall \Rightarrow remainder terms of order $R_{d,q}(K)$ too.



Thank you!