

Large population limit of interacting particle systems for population dynamics

Summer School at ICTS Bengaluru
September 2026

Oliver Tse

Joint work with Jasper Hoeksema, Anastasiia Hraivoronska, Evie Nielen



Population dynamics

Doubly nonlocal Fisher-KPP

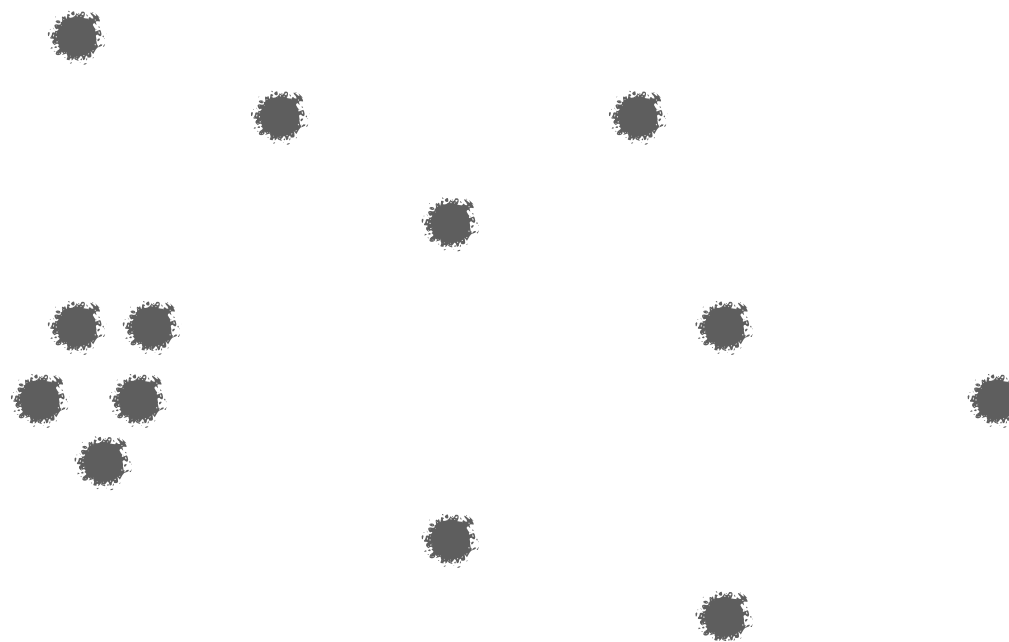
$X \hat{=}$ compact Polish space

$$\partial_t u_t(x) = r_b[u_t](x) - r_d[u_t](x) u_t(x)$$

(FKPP)

$$+ \int_X (u_t(y) - u_t(x)) r_h[u_t](x, y) \pi(dy)$$

$r_b \hat{=}$ birth, $r_d \hat{=}$ death/competition, $r_h \hat{=}$ hopping



Population dynamics

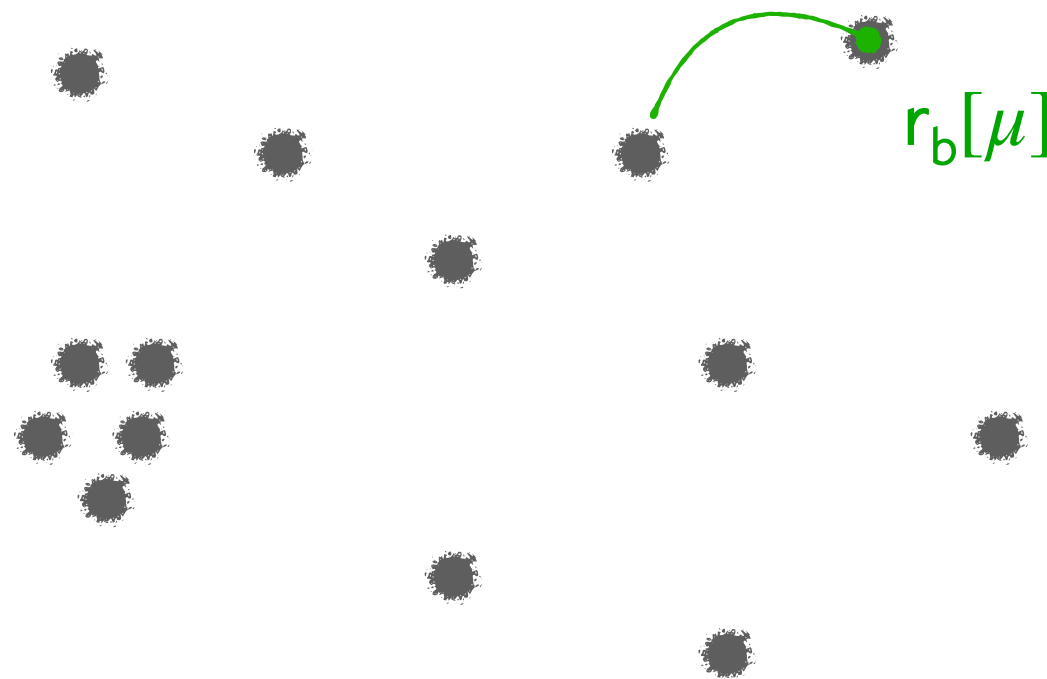
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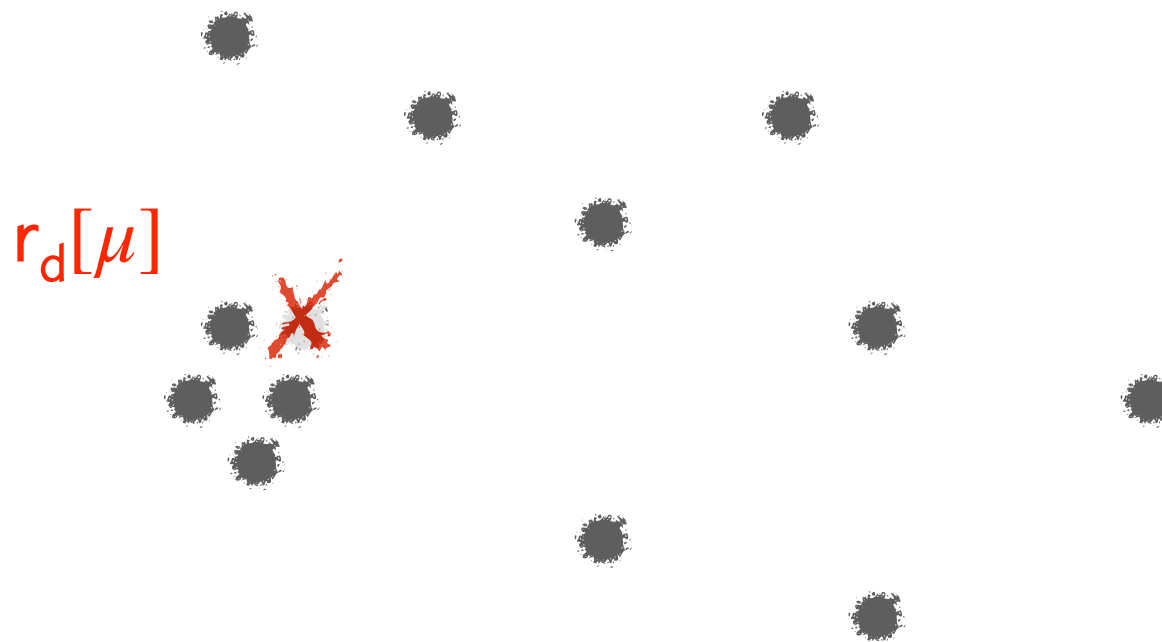
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$$\partial_t u_t(x) = \mathbf{r}_b[u_t](x) - \mathbf{r}_d[u_t](x) u_t(x) + \int_X (u_t(y) - u_t(x)) \mathbf{r}_h[u_t](x, y) \pi(dy)$$

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Population dynamics

Doubly nonlocal Fisher-KPP

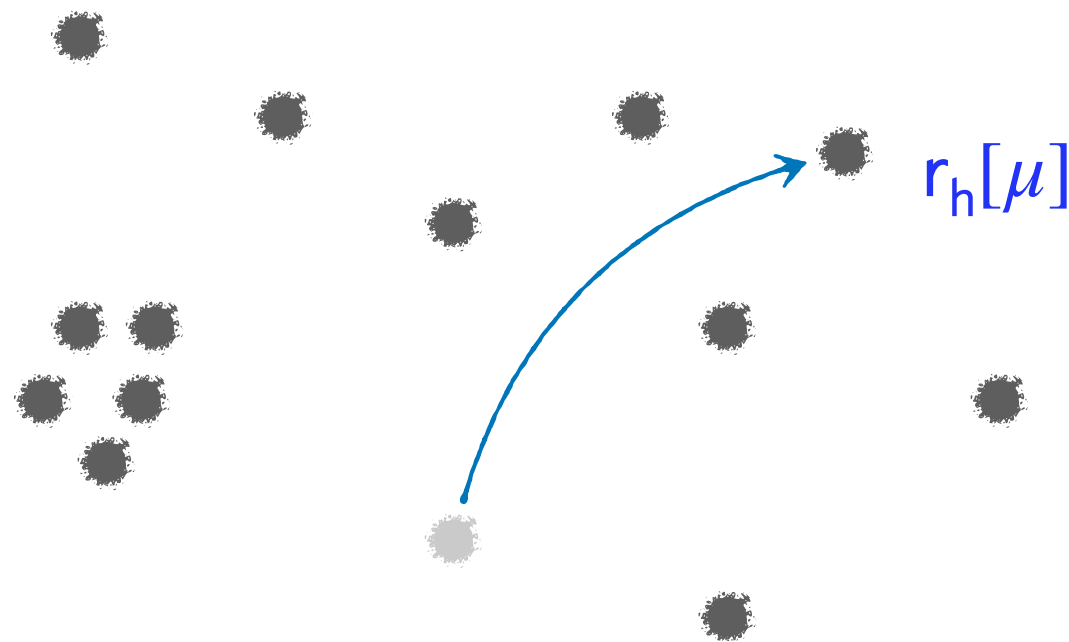
$X \triangleq$ compact Polish space

(FKPP)

$$\partial_t u_t(x) = r_b[u_t](x) - r_d[u_t](x) u_t(x) + \int_X (u_t(y) - u_t(x)) r_h[u_t](x, y) \pi(dy)$$

Reference measure 

$r_b \triangleq$ birth, $r_d \triangleq$ death/competition, $r_h \triangleq$ hopping



Population dynamics

Doubly nonlocal Fisher-KPP

$X \hat{=}$ compact Polish space

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$$\partial_t u_t(x) = r_b[u_t](x) - r_d[u_t](x) u_t(x) + \int_X (u_t(y) - u_t(x)) r_h[u_t](x, y) \pi(dy)$$

Reference measure 

$r_b \hat{=}$ birth, $r_d \hat{=}$ death/competition, $r_h \hat{=}$ hopping

Multicomponent

(SIS)

$$\begin{aligned} \partial_t u_t &= -\phi \star v_t u_t + \beta v_t \\ \partial_t v_t &= \phi \star v_t u_t - \beta v_t \end{aligned} \quad \longrightarrow \quad \partial_t v_t = \phi \star v_t (\alpha - v_t) - \beta v_t$$

Population dynamics

Doubly nonlocal Fisher-KPP

$X \hat{=}$ compact Polish space

(FKPP)

$$\partial_t \mu_t = \mathcal{V}[\mu_t]$$
$$\mathcal{V}[\mu_t] \hat{=} \kappa_{\text{bd}}[\mu_t] + \kappa_{\text{h}}[\mu_t] : \mathcal{M}^+(X) \rightarrow \mathcal{M}^+(X)$$

where

$$\kappa_{\text{bd}}[\mu](dx) = r_{\text{b}}[\mu](dx) - r_{\text{d}}[\mu](x) \mu(dx)$$

$$\kappa_{\text{h}}[\mu](dx) = \int_X r_{\text{h}}[\mu](y, dx) \mu(dy) - \int_X r_{\text{h}}[\mu](x, dy) \mu(dy)$$

$r_{\text{b}} \hat{=}$ birth, $r_{\text{d}} \hat{=}$ death/competition, $r_{\text{h}} \hat{=}$ hopping

Question: Is (FKPP) derived from microscopic dynamics?

Is (FKPP) a (generalized) gradient flow?

Agenda

Configuration space

Jump processes on configuration spaces

Towards the Liouville equation

Agenda

Configuration space

Jump processes on configuration spaces

Towards the Liouville equation

Configuration Spaces

For each $N \in \mathbb{N}_0$, define

$$\mathbb{X}^N := \left\{ (x^1, \dots, x^N) \in X^N : x^i \neq x^j, j \neq i \right\} / \sim$$

where $(x^1, \dots, x^N) \sim (x^{\sigma_1}, \dots, x^{\sigma_N})$ for every permutation σ

Reference measure

$$\pi \in \mathcal{M}^+(X) \longrightarrow \frac{\pi^{\otimes N}}{N!} \in \mathcal{M}^+(\mathbb{X}^N)$$

Configuration space

$$\mathbb{X} := \bigsqcup_{N \in \mathbb{N}_0} \mathbb{X}^N$$

$$\Lambda := e^{-\pi(X)} \sum_{N \in \mathbb{N}_0} \frac{\pi^{\otimes N}}{N!} \in \mathcal{P}(\mathbb{X})$$

Normalised Lebesgue-Poisson

Configuration Spaces

Configuration space

$$\mathbb{X} := \bigsqcup_{N \in \mathbb{N}_0} \mathbb{X}^N$$

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Normalised Lebesgue-Poisson

Observations

- 1) $\mathbf{x} \in \mathbb{X}^N \longleftrightarrow \gamma = \{x^1, \dots, x^N\} \xleftrightarrow{\mathbb{L}} \nu = \sum_{x \in \gamma} \delta_x \in \mathcal{M}^+(X)$
- 2) $\Pi := \mathbb{L}_\# \Lambda \in \mathcal{P}(\mathcal{M}^+(X))$
- 3) $\mathbb{E}[N(\mathbf{x})] = \int_{\mathbb{X}} N(\mathbf{x}) \Lambda(d\mathbf{x}) = \pi(X) = O(1)$

Configuration Spaces

Configuration space

$$\mathbb{X} := \bigsqcup_{N \in \mathbb{N}_0} \mathbb{X}^N$$

$$\Lambda^n := e^{-n\pi(X)} \sum_{N \in \mathbb{N}_0} \frac{(n\pi)^{\otimes N}}{N!} \in \mathcal{P}(\mathbb{X})$$

Scaled Normalised Lebesgue-Poisson

Observations

- 1) $\mathbf{x} \in \mathbb{X}^N \longleftrightarrow \gamma = \{x^1, \dots, x^N\} \xleftrightarrow{\mathbb{L}^n} \nu = \frac{1}{n} \sum_{x \in \gamma} \delta_x \in \mathcal{M}^+(X)$
- 2) $\Pi^n := \mathbb{L}_\#^n \Lambda^n \in \mathcal{P}(\mathcal{M}^+(X))$
- 3) $\mathbb{E}^n[N(\mathbf{x})] = \int_{\mathbb{X}} N(\mathbf{x}) \Lambda^n(d\mathbf{x}) = n\pi(X) = O(n)$

Configuration Spaces

Configuration space

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Scaled Normalised Lebesgue-Poisson

Observations

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- 2) $\Pi^n := \mathbb{L}_\#^n \Lambda^n \in \mathcal{P}(\mathcal{M}^+(X))$
- 3) $\mathbb{E}^n[|\nu|] = \int_{\mathcal{M}^+(X)} |\nu| \Pi^n(d\nu) = \pi(X)$

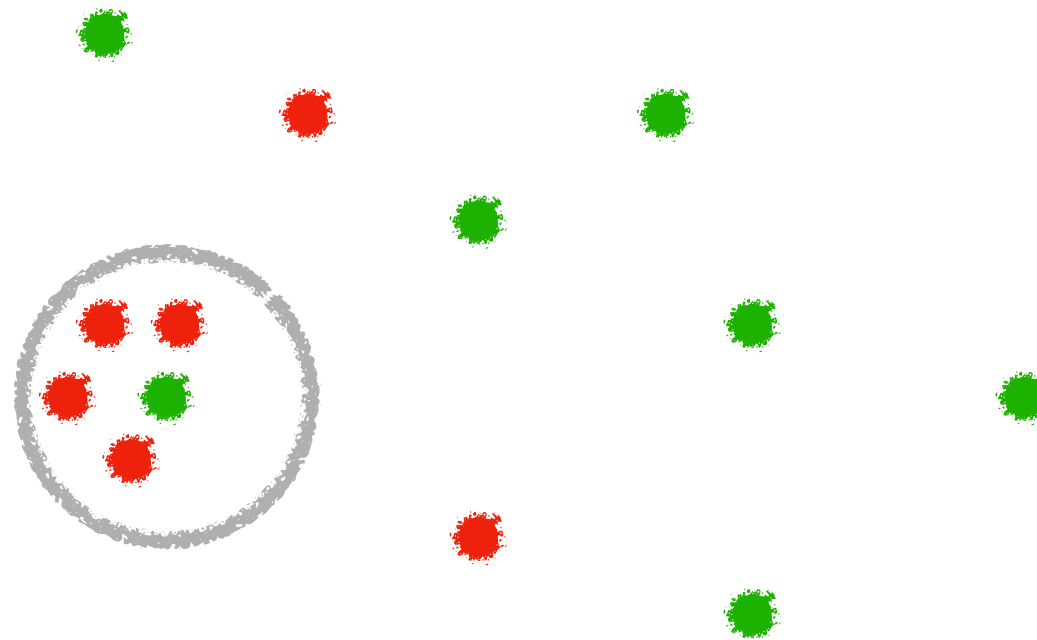
Configuration Spaces

Multicomponent Configurations

$$\mathbb{Y} := \bigsqcup_{\mathbf{N} \in \mathbb{N}_0^2} \mathbb{Y}^{\mathbf{N}}$$

$$\mathbb{Y}^{\mathbf{N}} := \mathbb{X}^{N_1} \times \mathbb{X}^{N_2}$$

$$\mathbb{Y}_{\neq} := \left\{ \gamma \in \mathbb{Y} : \gamma = (\gamma_1, \gamma_2), \gamma_1 \cap \gamma_2 = \emptyset \right\}$$



Configuration Spaces

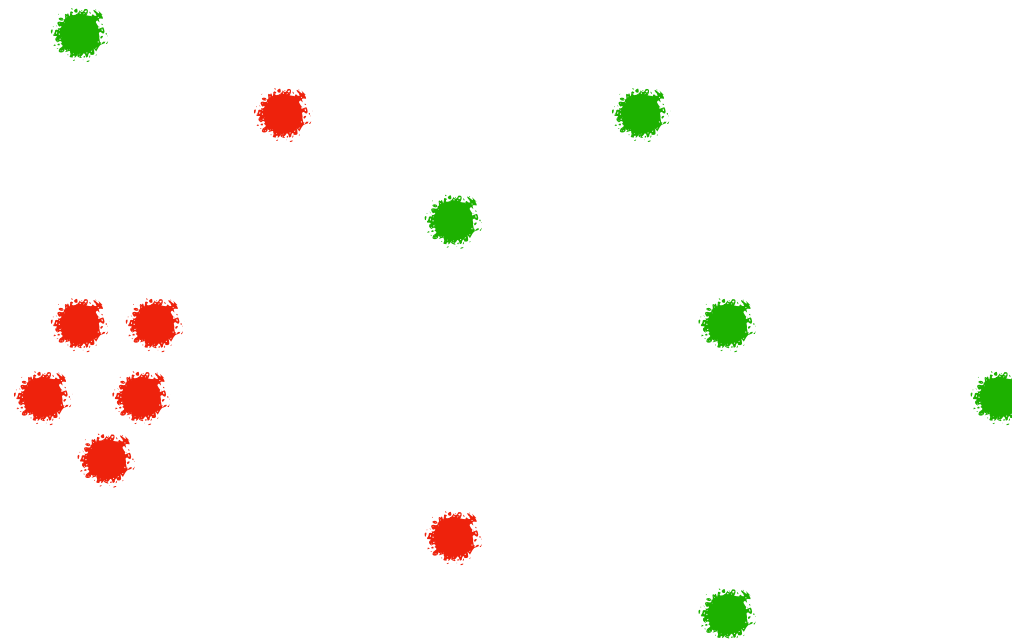
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$$\Lambda_{\neq}^n := \Lambda^n|_{\mathbb{Y}_{\neq}}$$



Agenda

Configuration space

Jump processes on configuration spaces

Towards the Liouville equation

Jump process

Birth

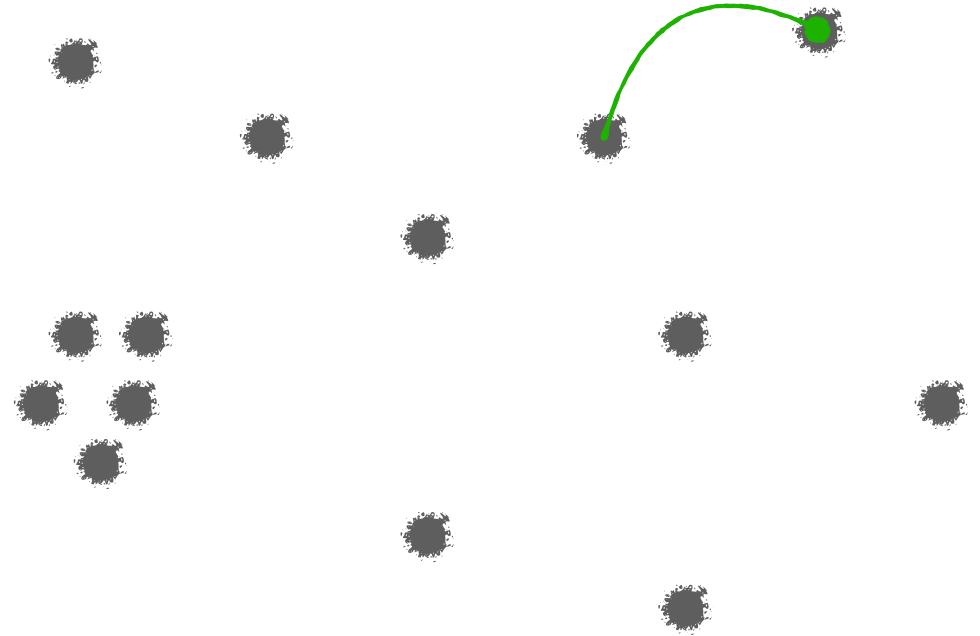
$$\nu \mapsto \nu + \frac{1}{n} \delta_x$$

with rate

$$n \, r_b[\nu](dx)$$

Generator

$$L_b^n F(\nu) := \int_X n \left[F\left(\nu + \frac{1}{n} \delta_x\right) - F(\nu) \right] r_b[\nu](dx)$$



Jump process

Birth

$$\nu \mapsto \nu + \frac{1}{n} \delta_x$$

with rate

$$n \, r_b[\nu](dx)$$

Death

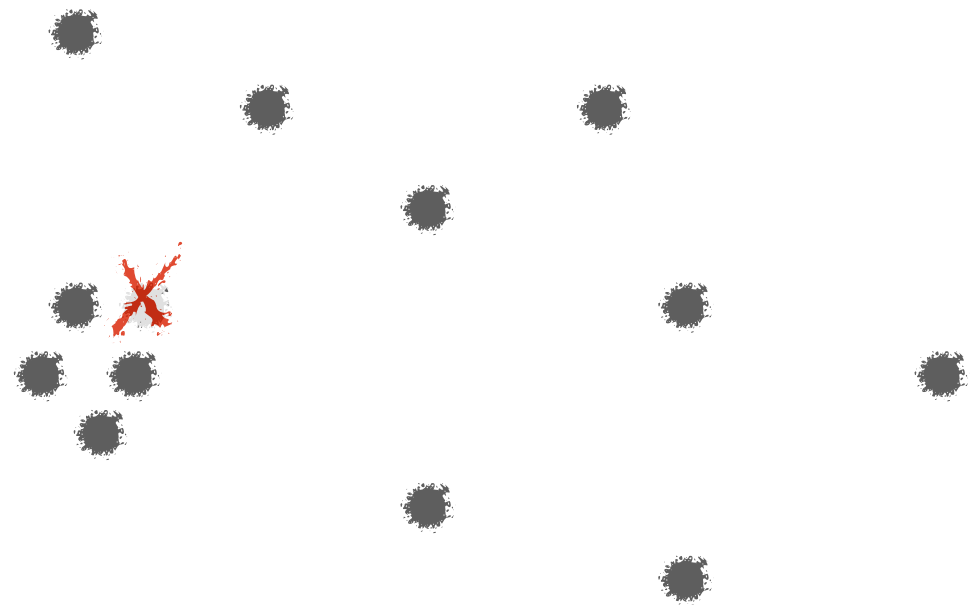
$$\nu \mapsto \nu - \frac{1}{n} \delta_x$$

with rate

$$n \, r_d[\nu](x) \, \nu(dx)$$

Generator

$$L_d^n F(\nu) := \int_X n \left[F\left(\nu - \frac{1}{n} \delta_x\right) - F(\nu) \right] r_d[\nu](x) \, \nu(dx)$$



Jump process

Birth

$$\nu \mapsto \nu + \frac{1}{n} \delta_x$$

with rate

$$n \mathbf{r}_b[\nu](dx)$$

Death

$$\nu \mapsto \nu - \frac{1}{n} \delta_x$$

with rate

$$n \mathbf{r}_d[\nu](x) \nu(dx)$$

Hopping

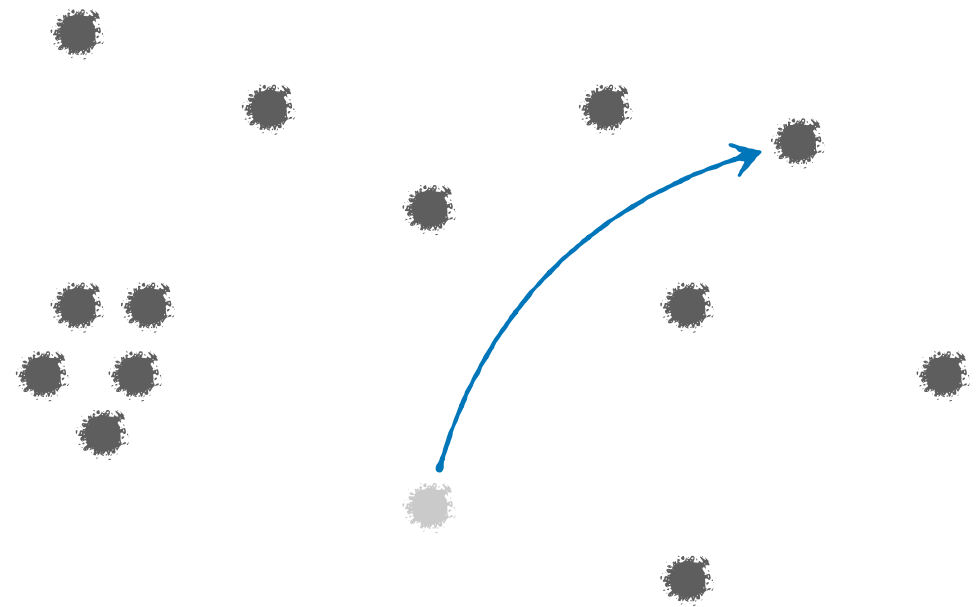
$$\nu \mapsto \nu - \frac{1}{n} \delta_x + \frac{1}{n} \delta_y$$

with rate

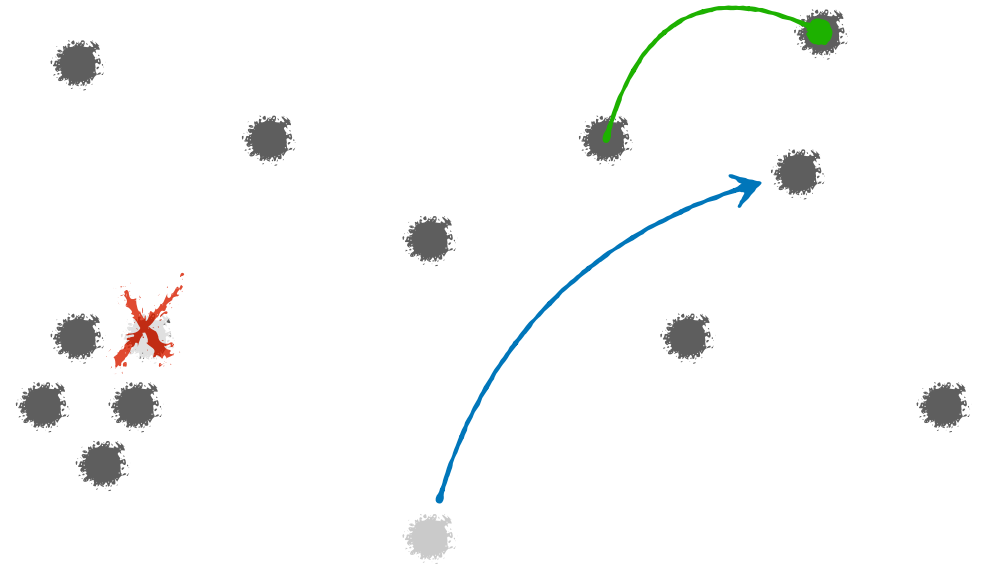
$$n \mathbf{r}_h[\nu](x, dy) \nu(dx)$$

Generator

$$L_h^n F(\nu) := \iint_{X \times X} n \left[F\left(\nu - \frac{1}{n} \delta_x + \frac{1}{n} \delta_y\right) - F(\nu) \right] \mathbf{r}_h[\nu](x, dy) \nu(dx)$$



Jump process



Generator $L^n := L_b^n + L_d^n + L_h^n$

$$L^n F(\nu) = \int_{\mathcal{M}^+(X)} [F(\eta) - F(\nu)] \kappa^n(\nu, d\eta)$$

generates the process $t \mapsto \nu_t \in D([0, T]; \mathcal{M}^+(X))$

where $P_t := \text{Law } \nu_t$ satisfies

(FKE_n) $\partial_t P_t = (L^n)^* P_t = -\overline{\text{div}}_{\mathcal{M}} J_t$

$$J_t = \kappa^n(\nu, d\eta) P_t(d\nu) = P_t \odot \kappa^n$$

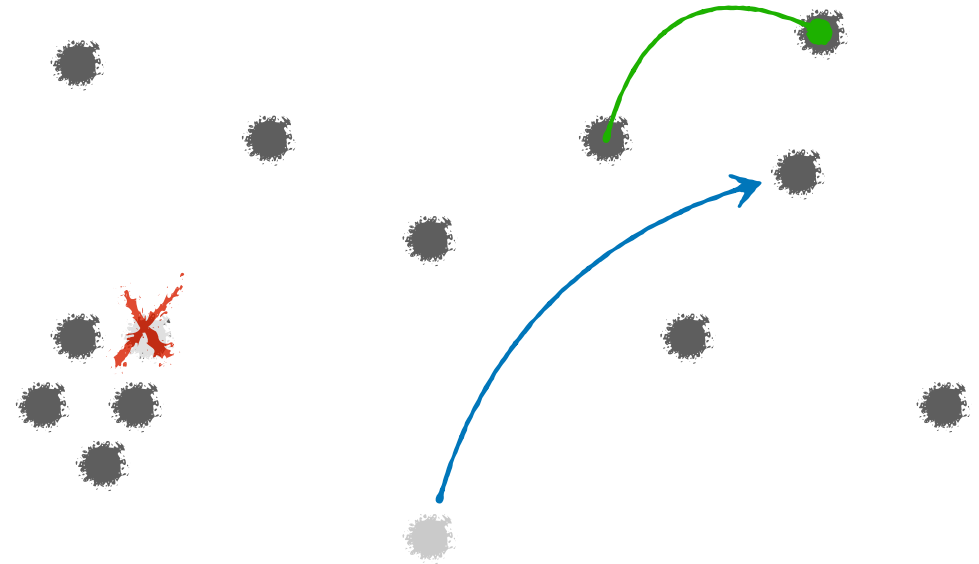
Gradient structure $\iff$ Detailed balance

Mielke-Peletier-Renger 2014

$\kappa^n(\nu, d\eta) \Pi^n(d\nu)$ **is symmetric**

Forward Kolmogorov

Jump process



Generator $L^n := L_b^n + L_d^n + L_h^n$

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$$J_t = \partial_2 \mathcal{R}_n^*(P_t, -\overline{\nabla}_{\mathcal{M}} \mathcal{E}_n(P_t))$$

Forward Kolmogorov

Gradient structure $\iff$ Detailed balance

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$\kappa^n(\nu, d\eta) \Pi^n(d\nu)$ **is symmetric**

$$\mathcal{E}_n(P) = n^{-1} \text{Ent}(P | \Pi^n)$$

Jump process

Gradient structure $\iff$ Detailed balance

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$$(\text{FKE}_n) \quad \partial_t P_t = (L^n)^* P_t = -\overline{\text{div}}_{\mathcal{M}} J_t$$

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$$\mathcal{E}_n(P) = n^{-1} \text{Ent}(P | \Pi^n)$$

Variational Principle

$$(P, J) \in \text{argmin} \left\{ \mathcal{L}_n(Q, G) : \partial_t Q_t + \overline{\text{div}}_{\mathcal{M}} G_t = 0, Q_0 = P_0 \right\}$$

where

$$0 \leq \mathcal{L}_n(P, J) := \int_0^T \mathcal{R}_n(P_t, J_t) + \mathcal{D}_n(P_t) dt + \mathcal{E}_n(P_T) - \mathcal{E}_n(P_0) \leq 0$$

Chain rule
Inequality

Generalized
Minimizing-movement

Jump process

Gradient structure $\Rightarrow$ detailed balance

Mielke-Peletier 2014

$\kappa^n(\nu, d\eta) \Pi^n$ symmetric

$$(\text{FKE}_n) \quad \partial_t P_t = (L^n)^* P_t = - \overline{\text{div}}_{\mathcal{M}} J_t$$

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where

$$\mathcal{L}_n(P, J) := \int_0^T \text{Ent}(J_t | P_t \odot \kappa^n) dt \leq 0$$

Large deviations

Approximations +
Lower semicontinuity

Agenda

Configuration space

Jump processes on configuration spaces

Large population limit

Towards Louville

$$\partial_t P_t^n + \overline{\mathbf{div}}_{\mathcal{M}} J_t^n = 0$$

$$\mathcal{E}_n(P) = n^{-1} \text{Ent}(P | \Pi^n)$$

Recall

$$L_b^n F(\nu) = \int_X n \left[F\left(\nu + \frac{1}{n} \delta_x\right) - F(\nu) \right] \overline{\nabla}_b^n F(\nu, x) r_b[\nu](dx)$$

Cylindrical functions $\mathbf{Cyl}_c(\mathcal{M}^+(X))$

$$F(\nu) = g(\langle f_1, \nu \rangle, \dots, \langle f_m, \nu \rangle)$$

$$g \in C_c^\infty(\mathbb{R}^m)$$

$$f_l \in C(X), \quad l = 1, \dots, m$$

Large population limit

$$\overline{\nabla}_b^n F(\nu, x) \longrightarrow \sum_{i=1}^m \partial_i g(\langle f_1, \nu \rangle, \dots, \langle f_m, \nu \rangle) f_i(x) =: \nabla_{\mathcal{M}} F(\nu, x)$$

$$L_b^n F(\nu) \longrightarrow \int_X \nabla_{\mathcal{M}} F(\nu, x) r_b[\nu](dx)$$

Towards Louville

$$\partial_t P_t^n + \overline{\mathbf{div}}_{\mathcal{M}} J_t^n = 0$$

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Recall

$$L_{\mathbf{d}}^n F(\nu) = \int_X n \left[F\left(\nu - \frac{1}{n} \delta_x\right) - F(\nu) \right] \overline{\nabla}_{\mathbf{d}}^n F(\nu, x) r_{\mathbf{d}}[\nu](x) \nu(dx)$$

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Large population limit

$$\overline{\nabla}_{\mathbf{d}}^n F(\nu, x) \longrightarrow -\nabla_{\mathcal{M}} F(\nu, x)$$

$$L_{\mathbf{d}}^n F(\nu) \longrightarrow -\int_X \nabla_{\mathcal{M}} F(\nu, x) r_{\mathbf{d}}[\nu](x) \nu(dx)$$

Towards Louville

$$\partial_t P_t^n + \overline{\mathbf{div}}_{\mathcal{M}} J_t^n = 0$$

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Recall

$$L_h^n F(\nu) = \iint_{X \times X} n \left[F\left(\nu - \frac{1}{n} \delta_x + \delta_y\right) - F(\nu) \right] \overline{\nabla}_h^n F(\nu, x, y) r_h[\nu](x, dy) \nu(dx)$$

Cylindrical functions $\mathbf{Cyl}_c(\mathcal{M}^+(X))$

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$$f_l \in C(X), \quad l = 1, \dots, m$$

Large population limit

$$\overline{\nabla}_h^n F(\nu, x, y) \longrightarrow \nabla_{\mathcal{M}} F(\nu, y) - \nabla_{\mathcal{M}} F(\nu, x)$$

$$L_h^n F(\nu) \longrightarrow \iint_{X \times X} [\nabla_{\mathcal{M}} F(\nu, y) - \nabla_{\mathcal{M}} F(\nu, x)] r_h[\nu](x, dy) \nu(dx)$$

Large population limit

$$\partial_t P_t^n + \overline{\mathbf{div}}_{\mathcal{M}} J_t^n = 0$$

$$\mathcal{E}_n(P) = n^{-1} \text{Ent}(P | \Pi^n)$$

Large population limit $\mathcal{V}[\nu] \hat{=} \kappa_{\text{bd}}[\nu] + \kappa_{\text{h}}[\nu]$

$$L^n F(\nu) \longrightarrow \int_X \nabla_{\mathcal{M}} F(\nu, x) \mathcal{V}[\nu](dx) \quad \forall F \in \mathbf{Cyl}_c(\mathcal{M}^+(X))$$

Result: The sequence of solutions $(P^n)_{n \in \mathbb{N}}$ to **(FKE)_n** with initial data $P_0 \in \mathcal{P}(\mathcal{M}^+(X))$ such that

$$\sup_{n \in \mathbb{N}} \mathcal{E}_n(P_0) < +\infty$$

has a (narrow) accumulation point P satisfying

$$\textbf{(Li)} \quad \partial_t P_t + \mathbf{div}_{\mathcal{M}}(\mathcal{V} P_t) = 0 \quad \textbf{Liouville}$$

where

$$\sup_{t \in [0, T]} \mathcal{E}_{\infty}(P_t) < +\infty, \quad \mathcal{E}_{\infty}(P) = \int_{\mathcal{M}^+(X)} \mathbf{Ent}(\nu | \pi) P(d\nu)$$

Large population limit

$$\partial_t P_t^n + \overline{\mathbf{div}}_{\mathcal{M}} J_t^n = 0$$

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If $\mathcal{S}_t: \mathcal{M}^+(X) \rightarrow \mathcal{M}^+(X)$ is the flow map for **(FKPP)** $\partial_t \mu_t = \mathcal{V}[\mu_t]$, then

$$P_t = (\mathcal{S}_t)_{\#} P_0 \quad \textbf{for all } t \in [0, 1] \quad \textbf{Superposition Principle}$$

In particular, if $P_0 = \delta_{\nu}$ with $\text{Ent}(\nu_0 | \pi) < +\infty$, then $P_t = \delta_{\mathcal{S}_t(\nu)}$

Large population limit

$$\partial_t P_t^n + \overline{\mathbf{div}}_{\mathcal{M}} J_t^n = 0$$

$$\mathcal{E}_n(P) = n^{-1} \text{Ent}(P | \Pi^n)$$

Result: The sequence of solutions $(P^n)_{n \in \mathbb{N}}$ to **(FKE_n)** with initial data $P_0 \in \mathcal{P}(\mathcal{M}^+(X))$ such that

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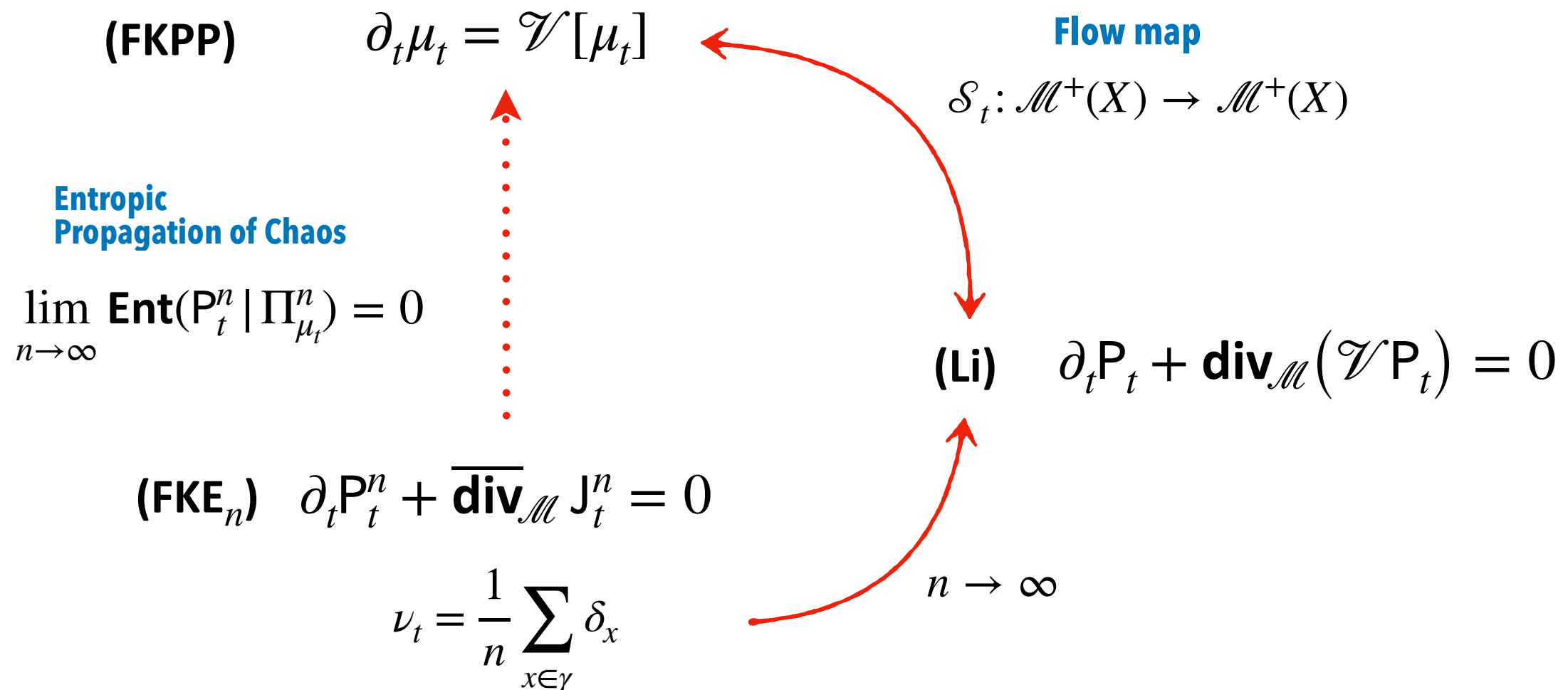
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$$\sup_{t \in [0, T]} \mathcal{E}_{\infty}(P_t) < +\infty, \quad \mathcal{E}_{\infty}(P) = \int_{\mathcal{M}^+(X)} \mathbf{Ent}(\nu | \pi) P(d\nu)$$

If **Detailed Balance** is satisfied then the GGF structure on **(FKE_n)** induces a GGF structure on **(FKPP)** via *Evolutionary Γ -convergence*

Sandier-Serfaty 2004

Wrap Up



Work in progress

Marked configurations
 Singular (detailed balance)
 Infinite configurations
 'Central limit'

Thank you!