

Discrete gradient flow structures for mean-field systems

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- \mathcal{X} a finite set
- N particles on \mathcal{X} distributed according to a **Gibbs measure** $\pi \in \mathcal{P}(\mathcal{X}^N)$

$$\mathbf{x} \in \mathcal{X}^N : \quad \pi(\mathbf{x}) := \frac{1}{Z^N} \exp\left(-U^N(\mathbf{x})\right)$$

- Hamiltonian $U^N : \mathcal{X}^N \rightarrow \mathbf{R}$ of **mean-field type**: $\exists U : \mathcal{P}(\mathcal{X}) \rightarrow \mathbf{R}$

$$U^N(\mathbf{x}) = NU\left(L^N(\mathbf{x})\right) \quad \text{with} \quad L^N(\mathbf{x}) := \frac{1}{N} \sum_{i=1}^N \delta_{x_i}$$

- Example

$$U^N(\mathbf{x}) = \sum_{i=1}^N V(x_i) + \frac{1}{N} \sum_{i,j=1}^N W(x_i, x_j)$$

In terms of U

$$U(\mu) = \sum_{x \in \mathcal{X}} \mu_x K_x(\mu) \quad \text{with} \quad K_x(\mu) = V(x) + \sum_{y \in \mathcal{X}} W(x, y) \mu_y$$

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Introduce a **reversible** dynamic wrt. Gibbs distribution π

- Single particle jumps

$$\mathbf{x}^{i;y} := \mathbf{x} - (x_i - y)\mathbf{e}^i = (x_1, \dots, x_{i-1}, y, x_{i+1}, \dots, x_N).$$

- On the level of **empirical** distributions

$$\text{if } L^N(\mathbf{x}) = \nu \in \mathcal{P}_N(\mathcal{X}) \quad \text{then} \quad L^N(\mathbf{x}^{i;y}) = \nu^{N;x_i,y} := \nu - \frac{1}{N}(\delta_{x_i} - \delta_y)$$

- Make dynamic reversible wrt. π

$$Q^N(\mathbf{x}, \mathbf{x}^{i;y}) = \sqrt{\frac{\pi_{\mathbf{x}^{i;y}}}{\pi_{\mathbf{x}}}} A_{x_i,y}^N(L^N(\mathbf{x})s) = Q^N(L^N(\mathbf{x}); x_i, y)$$

and $\{A_{x,y}^N(\mu)\}_{\mu \in \mathcal{P}(\mathcal{X})}$ a family of irreducible symmetric matrices.

- Generator

$$\mathcal{L}^N f := \sum_{i=1}^N \sum_{y \in \mathcal{X}} (f(\mathbf{x}^{i;y}) - f(\mathbf{x})) Q_{\mathbf{x}, \mathbf{x}^{i;y}}^N.$$

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- Free energy for $\mu^N \in \mathcal{P}(\mathcal{X}^N)$

$$\mathcal{F}^N(\mu) := \mathcal{H}^N(\mu | \pi) = \sum_{x \in \mathcal{X}^N} \mu_x \log \frac{\mu_x}{\pi_x}.$$

- Action of $\mu \in \mathcal{P}(\mathcal{X}^N)$ and $\psi \in \mathbf{R}^{\mathcal{X}^N}$

$$\mathcal{A}^N(\mu, \psi) = \frac{1}{2} \sum_{x, y} (\psi_y - \psi_x)^2 w_{x, y}^N(\mu) = \langle \psi, \mathcal{K}^N(\mu)\psi \rangle$$

with weights $w_{x, y}^N(\mu)$ defined with $\Lambda(a, b) = (a - b)/(\log a - \log b)$ as follows

$$w_{x, y}^N(\mu) := \Lambda\left(\mu_x Q^N(x, y), \mu_y Q^N(y, x)\right) = \Lambda\left(\frac{\mu_x}{\pi_x}, \frac{\mu_y}{\pi_y}\right) Q^N(x, y) \pi_x.$$

- Metric \mathcal{W}^N on $\mathcal{P}(\mathcal{X}^N)$

$$\mathcal{W}^N(\mu, \nu)^2 := \inf_{(c, \psi)} \int_0^1 \mathcal{A}^N(c(t), \psi(t)) dt$$

with the infimum among pairs such that $c(0) = \mu$, $c(1) = \nu$ and

$$\dot{c}_x(t) + \sum_y (\psi_y(t) - \psi_x(t)) w_{x, y}^N(c(t)) = 0 \quad \Leftrightarrow \quad \dot{c}(t) = \mathcal{K}^N(c(t))\psi.$$

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- N -particle Fisher information

$$\mathcal{I}^N(\mu) := \frac{1}{2} \sum_{(x, y) \in E_\mu} w_{x, y}^N(\mu) \left(\log(\mu_x Q^N(x, y)) - \log(\mu_y Q^N(y, x)) \right)^2$$

The evolution of the density $c \in \mathcal{P}(\mathcal{X}^N)$ satisfies

$$\dot{c}_x(t) = \sum_y (c_y(t) Q_{y,x} - c_x(t) Q_{x,y}) = (c(t) Q)_x = - \left(\mathcal{K}^N(c(t)) D \mathcal{F}^N(c(t)) \right)_x.$$

The results of [Maas / Mielke, 2011] show that c is the gradient flow of \mathcal{F}^N wrt. \mathcal{W}^N .

Proposition (Curves of maximal slope)

For $c \in \text{AC}([0, T], (\mathcal{P}(\mathcal{X}^N), \mathcal{W}^N))$ the function \mathcal{J}^N given by

$$\mathcal{J}^N(c) := \mathcal{F}^N(c(T)) - \mathcal{F}^N(c(0)) + \frac{1}{2} \int_0^T \mathcal{I}^N(c(t)) dt + \frac{1}{2} \int_0^T \mathcal{A}^N(c(t), \psi(t)) dt,$$

is non-negative, where ψ_t is such that the continuity equation holds. Moreover, a curve c is a solution to $\dot{c}(t) = c(t) Q^N$ if and only if $\mathcal{J}^N(c) = 0$.

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- Gibbs measures $\{\pi(\mu) \in \mathcal{P}(\mathcal{X})\}_{\mu \in \mathcal{P}(\mathcal{X})}$

$$\pi_x(\mu) = \frac{1}{Z(\mu)} \exp(-H_x(\mu)), \quad \text{with } H_x(\mu) = \frac{\partial}{\partial \mu_x} U(\mu), \quad \text{and } U(\mu) = \sum_{x \in \mathcal{X}} \mu_x K_x(\mu).$$

- $Q(\mu)$ reversible rates wrt. $\pi(\mu)$

$$Q_{xy}(\mu) = \sqrt{\frac{\pi_y(\mu)}{\pi_x(\mu)}} A_{xy}(\mu) \quad \text{with } A(\mu) \in \mathbf{R}^{\mathcal{X} \times \mathcal{X}} \text{ irreducible and symmetric.}$$

- nonlinear ODE for $c \in C^1([0, T], \mathcal{P}(\mathcal{X}))$

$$\dot{c}_x(t) = \sum_{y \neq x} (c_y(t) Q_{yx}(c(t)) - c_x(t) Q_{xy}(c(t))) = (c(t) Q(c(t)))_x$$

- Stationary states π^* are fixed points of

$$\mu \mapsto \pi(\mu) : \quad \pi(\pi^*) = \pi^*.$$

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$$\mathcal{F}(\mu) = \sum_{x \in \mathcal{X}} \mu_x \log \mu_x + U(\mu).$$

Note: $\mathcal{F}(\mu) \neq \mathcal{H}(\mu \mid \pi(\mu))$. However $\partial_{\mu_x} \mathcal{F}(\mu) = \log \frac{\mu_x}{\pi_x(\mu)} + 1 - \log Z(\mu)$.

Onsager operator $\mathcal{K} : \mathbf{R}^{\mathcal{X}} \rightarrow \mathbf{R}^{\mathcal{X}}$ defined for $\psi \in \mathbf{R}^{\mathcal{X}}$ by

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Formal gradient flow

$$\dot{c}(t) = -\mathcal{K}(c(t))DF(c(t)).$$

Dissipation:

$$\frac{d}{dt} \mathcal{F}(c(t)) = -\mathcal{I}(c(t)) = -\frac{1}{2} \sum_{x,y} w_{xy}(c) (\log(c_x Q_{xy}(c)) - \log(c_y Q_{yx}(c)))^2.$$

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Proposition (Metric)

The space $(\mathcal{P}(\mathcal{X}), \mathcal{W})$ with the metric defined by

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and (c, ψ) solves

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Proposition (Curves of maximal slope)

For any $(c(t))_{t \in [0, T]} \in \text{AC}([0, T], (\mathcal{P}(\mathcal{X}), \mathcal{W}))$ holds

$$\mathcal{J}(c) := \mathcal{F}(c(T)) - \mathcal{F}(c(0)) + \frac{1}{2} \int_0^T \mathcal{I}(c(t)) dt + \frac{1}{2} \int_0^T \mathcal{A}(c(t), \psi(t)) dt \geq 0$$

Moreover, $\mathcal{J}(c) = 0$ if and only if $\dot{c} = cQ(c)$. In this case $c(t) \in \mathcal{P}^*(\mathcal{X})$ for all $t > 0$.

- Since $L_{\sharp}^N \mu^N \in \mathcal{P}(\mathcal{P}_N(\mathcal{X}))$, a **lifting** of the die ODE from $\mathcal{P}(\mathcal{X})$ to $\mathcal{P}(\mathcal{P}(\mathcal{X}))$ is necessary to make it compatible
- For randomized initial data law $c(0) = \mathbb{C}(0) \in \mathcal{P}(\mathcal{P}(\mathcal{X}))$ holds

$$\partial_t \mathbb{C}(t, c) + \operatorname{div}_{\mathcal{P}(\mathcal{X})} (\mathbb{C}(t, c) c Q(c)) = 0. \quad (\text{Lio})$$

- free energy \mathbb{F} , action \mathbb{A} , Fisher information \mathbb{I} are defined as averages of their unlifted counterparts:

$$\mathbb{F}(\mathbb{C}) := \int \mathcal{F}(\nu) \mathbb{C}(d\nu).$$

- Consistency of definition of metric

$$\mathbb{W}(\mathbb{M}, \mathbb{N}) := \inf_{(\mathbb{C}, \Psi)} \int_0^1 \mathbb{A}(\mathbb{C}(t), \Psi(t)) dt \stackrel{!}{=} W_{\mathbb{W}}^2(\mathbb{M}, \mathbb{N}) := \inf_{\Pi} \int \mathcal{W}^2(\mu, \nu) \Pi(d\mu, d\nu).$$

- De Giorgi functional $\mathbb{J} : \text{AC}([0, T], (\mathcal{P}(\mathcal{P}(\mathcal{X})), \mathbb{W})) \rightarrow [0, \infty]$

$$\mathbb{J}(\mathbb{C}) = \mathbb{F}(\mathbb{C}(T)) - \mathbb{F}(\mathbb{C}(0)) + \frac{1}{2} \int_0^T \mathbb{I}(\mathbb{C}(t)) dt + \frac{1}{2} \int_0^T \mathbb{A}(\mathbb{C}(t), \Psi(t)) dt \geq 0$$

and $\mathbb{J}(\mathbb{C}) = 0$ if and only if \mathbb{C} solves (Lio).

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Master equation \mathbf{X}_t^N Markov $(\mathcal{L}^N, \mathcal{X}^N)$	$\mathbf{c} \in \text{AC}([0, t], (\mathcal{P}(\mathcal{X}^N), \mathcal{W}^N))$
$\dot{\mathbf{c}}(t) = -\mathcal{K}^N(\mathbf{c}(t))D\mathcal{H}^N(\mathbf{c}(t) \boldsymbol{\pi})$	$\stackrel{\text{de Giorgi}}{\iff} \mathcal{J}^N(\mathbf{c}) = 0$
$\Downarrow L_{\sharp}^N$	$\Downarrow L_{\sharp}^N$
\mathbb{C}^N Markov $(\bar{\mathcal{L}}^N, \mathcal{P}_N(\mathcal{X}))$	$\mathbb{C}^N \in \text{AC}([0, T], (\mathcal{P}(\mathcal{P}_N(\mathcal{X})), \mathcal{W}^N))$
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Strategy

Proof Γ -lim inf estimate for \mathcal{J}^N wrt. \mathbb{J} , whenever $L_{\sharp}^N \mathbf{c} \xrightarrow{d} \mathbb{C}$ on $[0, T]$

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Theorem (Sandier-Serfaty)

Assume that whenever a sequence $c^N \in AC([0, T], (\mathcal{P}(\mathcal{X}^N), \mathcal{W}^N))$ for $t \in [0, T]$ it holds $L_{\#}^N c^N(t) \xrightarrow{d} \mathbb{C}(t) \in \mathcal{P}(\mathcal{P}(\mathcal{X}))$ and

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In addition, assume it holds

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where (c^N, ψ^N) and $(\mathbb{C}(t), \Psi(t))$ are solutions of certain continuity equations.

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Then, whenever $\mathcal{J}^N(c^N) = 0$ and $c^N(0) \xrightarrow{T} \mathbb{C}(0)$ such that $\lim_{N \rightarrow \infty} \mathcal{F}^N(c^N(0)) = \mathbb{F}(\mathbb{C}(0)) - \mathcal{F}_0$, it holds $\mathbb{J}(\mathbb{C}) = 0$ and

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Proposition (\liminf -estimate for free energy)

If $L_{\sharp}^N \mu^N \xrightarrow{d} \mathbb{M}$, then

$$\lim_{N \rightarrow \infty} \frac{1}{N} \mathcal{H}(\mu^N \mid \pi) \geq \int_{\mathcal{P}(\mathcal{X})} (\mathcal{F}(\nu) - \mathcal{F}_0) \mathbb{M}(d\nu) = \mathbb{F}(\mathbb{M}) - \mathcal{F}_0, \quad (\text{A0})$$

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Proof: Decompose relative entropy

$$\frac{1}{N} \mathcal{H}(\mu^N | \pi^N) = \frac{1}{N} \mathcal{H}(\mu^N) + \mathbb{E}_{L_{\#}^N \mu^N} [U] + \frac{1}{N} \log \mathbf{Z}^N$$

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$$\frac{1}{N} \mathcal{H}(\mu^N) \geq -\frac{1}{N} \log |\mathcal{P}_N(\mathcal{X})| - \frac{1}{N} \mathbb{E}_{L_{\#}^N \mu^N} [\log |\mathcal{T}_N|]$$

$$\text{Stirling} \geq -\frac{d \log N}{N} + \mathbb{E}_{L_{\#}^N \mu^N} [\mathcal{H}_{\mathcal{P}(\mathcal{X})}(\bullet)] - \frac{\log(N+1)}{N}$$

By Sanov's Theorem:

$$\lim_{N \rightarrow \infty} \frac{1}{N} \log \mathbf{Z}^N = - \inf_{\nu \in \mathcal{P}(\mathcal{X})} \left\{ \sum_{x \in \mathcal{X}} \nu(x) \log \nu(x) + U(\nu) \right\} =: -\mathcal{F}_0.$$

Proposition (Convergence of metric derivative and slopes)

Let $\mathbf{c}^N \in \text{AC}([0, T], (\mathcal{P}(\mathcal{X}^N), \mathcal{W}^N))$ with $(\mathbf{c}^N, \boldsymbol{\psi}^N)$ solving the continuity equation.
If

$$L_{\#}^N \mathbf{c}^N \xrightarrow{d} \mathbb{C} \quad \text{for some measurable } \mathbb{C} : [0, T] \rightarrow \mathcal{P}(\mathcal{P}(\mathcal{X})),$$

such that

$$\limsup_{N \rightarrow \infty} \int_0^T \frac{1}{N} \mathcal{A}^N(\mathbf{c}^N(t), \boldsymbol{\psi}^N(t)) dt < \infty.$$

Then $\mathbb{C} \in \text{AC}([0, T], \mathcal{P}(\mathcal{P}(\mathcal{X})))$, and it exists $\Psi : [0, T] \times \mathcal{P}(\mathcal{X}) \rightarrow \mathbf{R}^X$, for which (\mathbb{C}, Ψ) solves the continuity equation and it holds

$$\liminf_{N \rightarrow \infty} \int_0^T \frac{1}{N} \mathcal{A}^N(\mathbf{c}^N(t), \boldsymbol{\psi}^N(t)) dt \geq \int_0^T \mathbb{A}(\mathbb{C}(t), \Psi(t)) dt \quad (\text{A1})$$

and

$$\liminf_{N \rightarrow \infty} \int_0^T \frac{1}{N} \mathcal{I}^N(\mathbf{c}^N(t)) dt \geq \int_0^T \mathbb{I}(\mathbb{C}(t)) dt. \quad (\text{A2})$$

Previous results + tightness for particle system imply:

Theorem (Convergence of the particle system to the mean field equation)

Let \mathbf{c}^N be the law of the N -particle system. Moreover assume its initial distribution to be well prepared

$$\frac{1}{N} \mathcal{F}^N(\mathbf{c}^N(0)) \rightarrow \mathbb{F}(\mathbb{C}(0)) - \mathcal{F}_0 \quad \text{with} \quad L_{\sharp}^N \mathbf{c}^N(0) \xrightarrow{d} \mathbb{C}(0) \quad \text{as } N \rightarrow \infty.$$

Then it holds

$$L_{\sharp}^N \mathbf{c}^N(t) \xrightarrow{d} \mathbb{C}(t) \quad \text{for all } t \in (0, \infty),$$

with \mathbb{C} a weak solution to (Lio) and moreover

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Similar results in this spirit:

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Definition (κ -convexity wrt. \mathcal{W})

$\{Q(\mu) \in \mathcal{R}^{\mathcal{X} \times \mathcal{X}}\}_{\mu \in \mathcal{P}(\mathcal{X})}$ is κ -convex with $\kappa \in \mathbf{R}$, if for any constant speed geodesic $c \in \text{AC}([0, 1], (\mathcal{P}(\mathcal{X}), \mathcal{W}))$ holds

$$\mathcal{F}(c(t)) \leq (1-t)\mathcal{F}(c(0)) + t\mathcal{F}(c(1)) - \kappa \frac{t(1-t)}{2} \mathcal{W}^2(c(0), c(1)).$$

Corollary (Two-point space)

Assume $\mathcal{X} = \{0, 1\}$, $p(\mu) := Q(\mu; 0, 1)$ and $q(\mu) := Q(\mu; 1, 0)$ as well as $p'(\mu) = \partial_{\mu_0} p(\mu)$ and $q'(\mu) = \partial_{\mu_1} q(\mu)$ then the κ is give by

$$\kappa = \inf_{\mu \in \mathcal{P}(\mathcal{X})} \left(\frac{p(\mu) + q(\mu)}{2} + 3(\mu(0)p'(\mu) + \mu(1)q'(\mu)) \right. \\ \left. + \Lambda(\mu_0 p(\mu), \mu_1 q(\mu)) \left(\frac{1}{2\mu(0)p(\mu)} + \frac{1}{2\mu(1)q(\mu)} - \frac{p'(\mu)}{p(\mu)} - \frac{q'(\mu)}{q(\mu)} \right) \right).$$

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Mean-field Ising model on $\mathcal{X} = \{0, 1\}$. Define potentials by

$V(0) = V(1) = W(0, 0) = W(1, 1) = 0$ and $W(0, 1) = W(1, 0) = \beta > 0$. Hence $K_0(\mu) = \beta\mu_1$, $K_1(\mu) = \beta\mu_0$ and so

$$\mathcal{F}(\mu) = \sum_{\sigma \in \{0,1\}} (\log \mu_\sigma + K_\sigma(\mu)) \mu_\sigma = \mu_0 \log \mu_0 + \mu_1 \log \mu_1 + 2\beta\mu_0\mu_1.$$

As a function $\mathcal{F} : \mathcal{P}(\mathcal{X}) \rightarrow \mathbf{R}$ is convex for $\beta \leq 1$.

Does the same holds for κ -convexity wrt. \mathcal{W} ?

For the dynamic use for instance Metropolis rates:

$$p_{\text{MC}}(\mu) = \exp(-2\beta(\mu(0) - \mu(1))_+) \quad q_{\text{MC}}(\mu) = \exp(-2\beta(\mu(1) - \mu(0))_+)$$

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κ -convexity.

- Proof lower bound in $\kappa_{MC}(\beta) = 2 - 2\beta$
- Connect κ^N -convexity of N -particle system with κ -convexity of limit system:
Easy:

$$\lim_{N \rightarrow \infty} \kappa^N \leq \kappa$$

Hard: Quantified comparison

$$\kappa = \kappa^N + o_N(1).$$

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 \Rightarrow Fokker-Planck equation
- Quantify the rate of convergence in N
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