

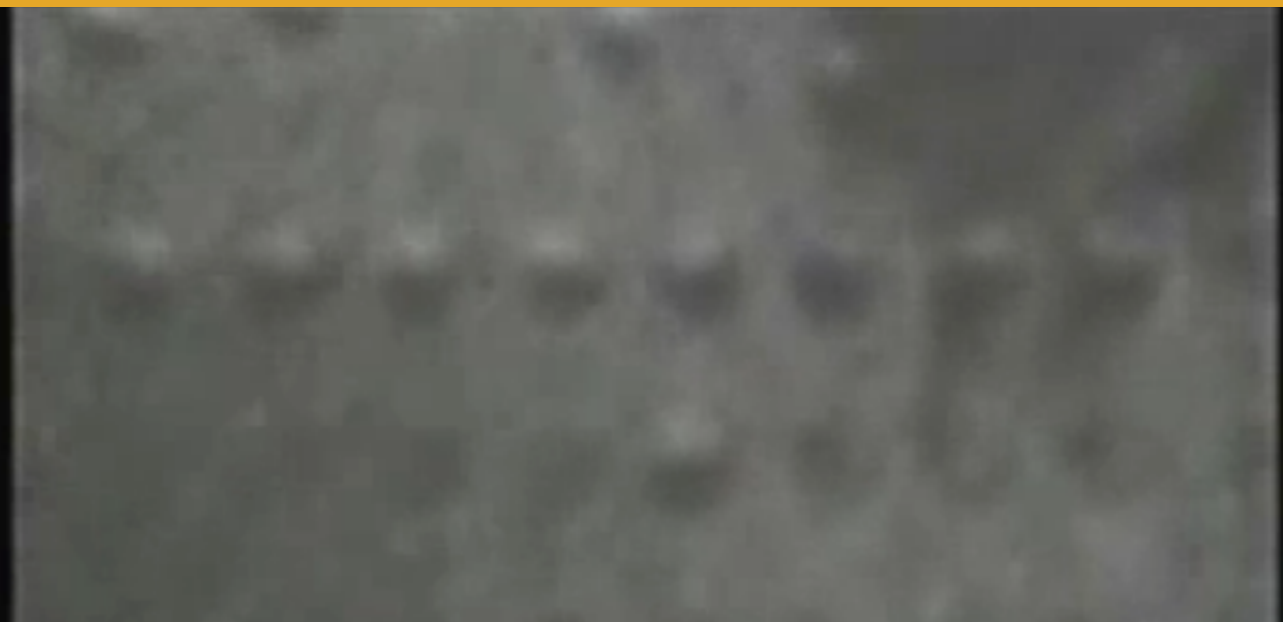
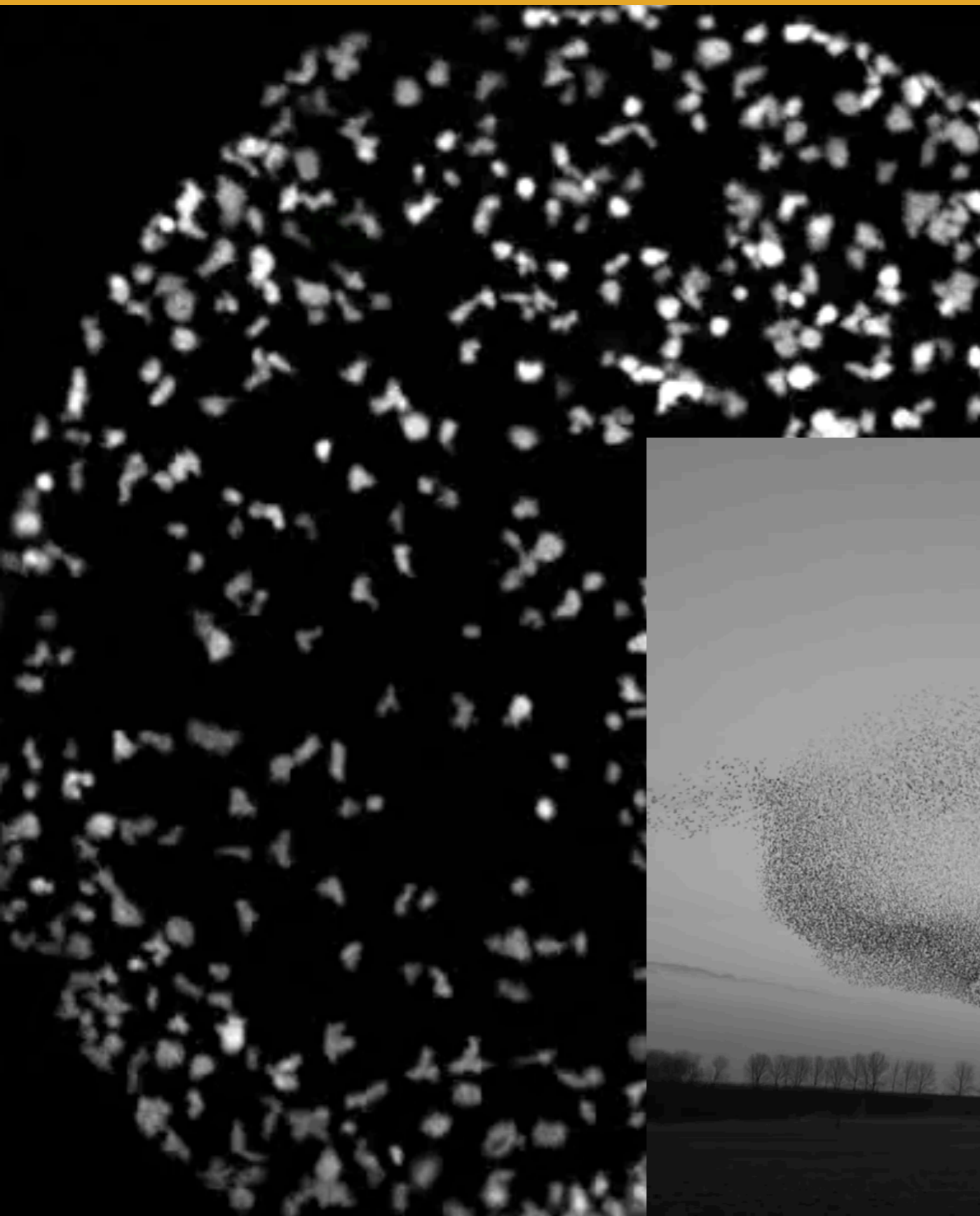
Minimizers and gradient flows in the slow diffusion limit

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nonlocal interactions



interaction energy / aggregation eqn

- $\rho(x,t): \mathbb{R}^d \times \mathbb{R} \rightarrow [0, +\infty)$ nonnegative density
- mass is conserved $\Rightarrow \int \rho(x) dx = M$

interaction energy:

$$\mathcal{K}(\rho) = \frac{1}{2} \int K * \rho d\rho$$

aggregation equation:

$$\frac{d}{dt} \rho = \nabla \cdot ((\nabla K * \rho) \rho)$$

interaction kernels, $K(x) : \mathbb{R}^d \rightarrow \mathbb{R}$

- vortex motion/chemotaxis: $K(x) = \pm |x|^{2-d} / (2-d)$ $|x|^0/0 = \log(|x|)$
- granular media: $K(x) = |x|^3$
- swarming: $K(x) = |x|^a/a - |x|^b/b, \quad -d \leq b \leq a$

minimizers

interaction energy:

$$\mathcal{K}(\rho) = \frac{1}{2} \int K * \rho d\rho$$

aggregation equation:

$$\frac{d}{dt}\rho = \nabla \cdot ((\nabla K * \rho)\rho)$$

previous work:

- competing effects of attraction/repulsion lead to rich structure

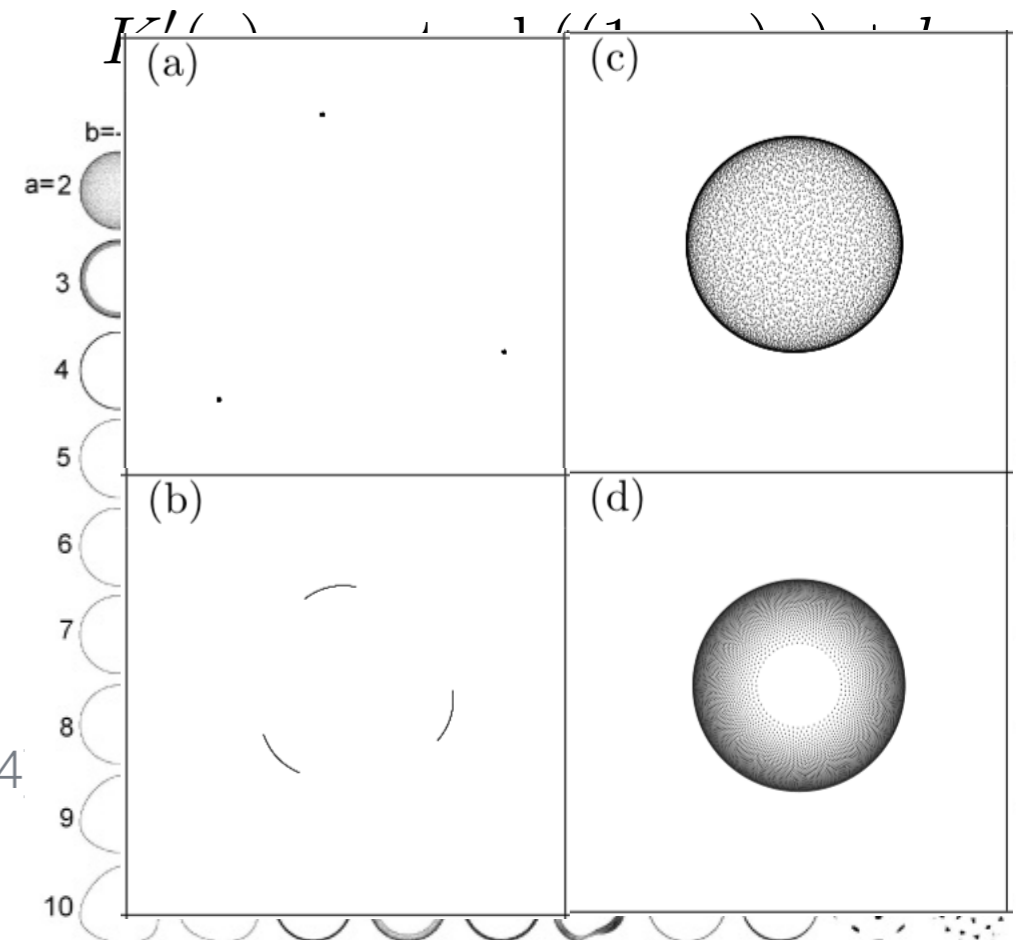
[Kolokolnikov, Sun, Uminsky, Bertozzi, 2011]

- more singular, repulsive $K \rightarrow$ minimizers have higher dimensional support

[Balague, Carrillo, Laurent, Raoul 2012]

- existence of minimizers

[Slepçev, Simione, Topaloglu '14] [Cañizo, Carrillo, Patacchini '14]



set valued minimizers

more recently: set valued minimizers

$$\min \left\{ \int_{\Omega} \int_{\Omega} K(x-y) dx dy : \Omega \subseteq \mathbb{R}^d, |\Omega| = M \right\}$$

Related shape optimization problems (d=3):



- Nonlocal isoperimetric problem [Knüpfer, Moratov '13]. [Lu, Otto '14], [Frank, Lieb '15], ...

$$\min \left\{ \text{Perimeter}(\Omega) + \int_{\Omega} \int_{\Omega} \frac{1}{|x-y|} dx dy : \Omega \subseteq \mathbb{R}^3, |\Omega| = M \right\}$$

set valued minimizers

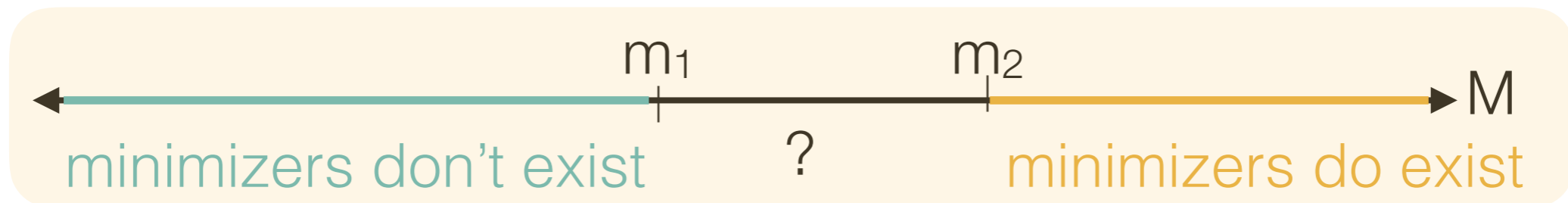
more recently: set valued minimizers $K(x) = |x|^a/a - |x|^b/b$, $-d < b \leq a$

$$\min \left\{ \int K * \rho d\rho : \rho = 1_\Omega \text{ for } \Omega \subseteq \mathbb{R}^d, |\Omega| = M \right\}$$

critical mass:

- $a = 2$, $-d < b \leq 2-d$ [Burchard, Choksi, Topaloglu 2016] $a > 0$, $b = 2-d$ [Frank, Lieb 2017]

there exist $m_1 \leq m_2$ s.t.



- for $a = 2$, $b = 2-d$, we have $m_1 = m_2 = \omega_d$

in general, it is unknown whether $m_1 = m_2$ and how m_1, m_2 depend on a, b .

set valued minimizers

[B

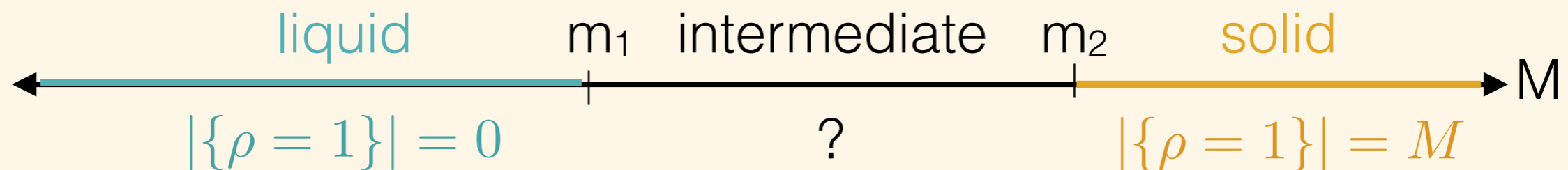
more recently: set valued minimizers $K(x) = |x|^a/a - |x|^b/b$, $-d < b \leq a$

$$\min \left\{ \int K * \rho d\rho : \rho = 1_\Omega \text{ for } \Omega \subseteq \mathbb{R}^d, |\Omega| = M \right\}$$

relaxed problem:

$$\min \left\{ \int K * \rho d\rho : 0 \leq \rho \leq 1, \int \rho = M \right\}$$

- minimizers exist
- $2 \leq a \leq 4$, $2-d \leq b < 0$: minimizers unique [Burchard, Choksi, Topaloglu '16] [Lopes '17]
- $a > 0$, $b = 2-d$: there exist $m_1 \leq m_2$ s.t [Frank, Lieb 2017]



constrained aggregation

constrained aggregation:

$$\left\{ \begin{array}{l} \frac{d}{dt} \rho = \nabla \cdot (\nabla(K * \rho)\rho) \text{ if } \rho < 1 \\ \rho \leq 1 \text{ always} \end{array} \right.$$

aggregation diffusion equation:

$$\frac{d}{dt} \rho = \nabla \cdot ((\nabla K * \rho)\rho) + \Delta \rho^m$$

motivation:

- previous work on congested drift equation (pedestrian crown motion)
[Maury, Roudneff-Chupin, Santambrogio 2010], [Alexander, Kim, Yao 2014]
- competition between nonlocal attraction and repulsion from constraint.
- heuristically, height constraint is singular limit of degenerate diffusion:

Idea: $\Delta \rho^m = \nabla \cdot (\underbrace{m \rho^{m-1}}_D \nabla \rho)$, so as $m \rightarrow +\infty$, $D \rightarrow \begin{cases} +\infty & \text{if } \rho > 1 \\ 0 & \text{if } \rho < 1 \end{cases}$

constrained aggregation

constrained aggregation:

$$\left\{ \begin{array}{l} \frac{d}{dt} \rho = \nabla \cdot (\nabla(K * \rho)\rho) \text{ if } \rho < 1 \\ \rho \leq 1 \text{ always} \end{array} \right.$$

aggregation diffusion equation:

$$\frac{d}{dt} \rho = \nabla \cdot ((\nabla K * \rho)\rho) + \Delta \rho^m$$

$m \rightarrow +\infty?$

results: [C. 2017, C. Kim, Yao 2017]

- the constrained aggregation eqn is well-posed as a W_2 gradient flow

$$\mathcal{E}_\infty(\rho) = \begin{cases} \int K * \rho d\rho & \text{if } \|\rho\|_\infty \leq 1 \\ +\infty & \text{otherwise} \end{cases}$$

- $K(x) = -|x|^{2-d}/(2-d) = \Delta^{-1}$
 - solutions with “set-valued” initial data remain “set-valued”
 - characterize via Hele-Shaw type free boundary problem
 - d=2: quantify convergence to equilibrium

slow diffusion limit

goals

- prove slow diffusion limit,

$$\frac{d}{dt}\rho = \nabla \cdot ((\nabla K * \rho)\rho) + \Delta \rho^m$$

$m \rightarrow +\infty$

$$\left\{ \begin{array}{l} \frac{d}{dt}\rho = \nabla \cdot (\nabla(K * \rho)\rho) \text{ if } \rho < 1 \\ \rho \leq 1 \text{ always} \end{array} \right.$$

$$K(x) = |x|^b/b, \quad K(x) = |x|^a/a - |x|^b/b, \quad 2 - d \leq b \leq a$$

- use numerical methods for aggregation diffusion equations to shed light on properties of minimizers of constrained interaction energy

gradient flow

Def: $\rho(t): \mathbb{R} \rightarrow P_2(\mathbb{R}^d)$ is the **Wasserstein gradient flow** of $E: P_2(\mathbb{R}^d) \rightarrow \mathbb{R}$ if

$$E(\rho(t)) - E(\rho(0)) \leq -\frac{1}{2} \int_0^t |\partial E|(\rho(s)) ds - \frac{1}{2} \int_0^t |\rho'| (s) ds$$

where

$$|\partial E|(\mu) := \limsup_{\nu \rightarrow \mu} \frac{(E(\mu) - E(\nu))^+}{W_2(\mu, \nu)} \quad \text{and} \quad |\rho'| (t) = \lim_{s \rightarrow t} \frac{W_2(\rho(s), \rho(t))}{|s - t|}$$

Analogy with **Euclidean gradient flow**:

$$\begin{aligned} \frac{d}{dt} x(t) = -\nabla E(x(t)) &\iff \begin{cases} \left| \frac{d}{dt} x(t) \right| = |\nabla E(x(t))| \\ \frac{d}{dt} E(x(t)) = -|\nabla E(x(t))| \left| \frac{d}{dt} x(t) \right| \end{cases} \\ &\iff \frac{d}{dt} E(x(t)) \leq -\frac{1}{2} |\nabla E(x(t))| - \frac{1}{2} \left| \frac{d}{dt} x(t) \right| \end{aligned}$$

gradient flow

goal:

show solutions of aggregation diffusion equations converge to congested aggregation equation

or equivalently:

show gradient flows of \mathcal{E}_m converge to the gradient flow of \mathcal{E}_∞

aggregation diffusion equation:

$$\frac{d}{dt}\rho = \nabla \cdot ((\nabla K * \rho)\rho) + \Delta \rho^m$$

interaction energy + Rényi entropy

$$\mathcal{E}_m(\rho) = \int K * \rho d\rho + \frac{1}{m-1} \int \rho^m$$

constrained aggregation:

$$\begin{cases} \frac{d}{dt}\rho = \nabla \cdot (\nabla(K * \rho)\rho) & \text{if } \rho < 1 \\ \rho \leq 1 & \text{always} \end{cases}$$

constrained interaction energy

$$\mathcal{E}_\infty(\rho) = \begin{cases} \int K * \rho d\rho & \text{if } \|\rho\|_\infty \leq 1 \\ +\infty & \text{otherwise} \end{cases}$$

Γ -convergence of gradient flows

Theorem: (Serfaty 2010): Let $\rho_m(x, t)$ be grad flows of \mathcal{E}_m such that

$$\rho_m(x, t) \rightarrow \rho_\infty(x, t) \text{ and } \mathcal{E}_m(\rho_m(x, 0)) \rightarrow \mathcal{E}_\infty(\rho_\infty(x, 0))$$



$$\rho_m(x, 0) \rightarrow \rho_\infty(x, 0) \text{ and } \mathcal{E}_m(\rho_m(x, 0)) \rightarrow \mathcal{E}_\infty(\rho_\infty(x, 0))$$

Recall: $\rho(t): \mathbb{R} \rightarrow \mathcal{P}_2(\mathbb{R}^d)$ is the Wasserstein gradient flow of $E: \mathcal{P}_2(\mathbb{R}^d) \rightarrow \mathbb{R}$ if

$$E(\rho(t)) - E(\rho(0)) \leq -\frac{1}{2} \int_0^t |\partial E|(\rho(s)) ds - \frac{1}{2} \int_0^t |\rho'| (s) ds$$

Γ -convergence of gradient flows

Goal: 1. $\liminf_{m \rightarrow +\infty} E_m(\rho_m(t)) \geq E_\infty(\rho_\infty(t))$

3. $\liminf_{m \rightarrow +\infty} \int_0^t |\partial E_m|^2(\rho_m(s)) ds \geq \int_0^t |\partial E_\infty|^2(\rho_\infty(s)) ds$

$$\mathcal{E}_m(\rho) = \int K * \rho d\rho + \frac{1}{m-1} \int \rho^m$$

$$\mathcal{E}_\infty(\rho) = \begin{cases} \int K * \rho d\rho & \text{if } \|\rho\|_\infty \leq 1 \\ +\infty & \text{otherwise} \end{cases}$$

- (1) follows by interpolation of L^p norms
- (3) is more difficult, due to the lack of convexity (or even ω -convexity) uniformly in m

instead, we must use specific structure of metric slope

$$|\partial \mathcal{E}_m|(\mu_m) = \left\| \nabla K * \mu_m + \frac{\nabla \mu_m^m}{\mu_m} \right\|_{L^2(\mu_m)}$$

slow diffusion limit

Theorem: (C., Topaloglu, in preparation)

Suppose $\rho_m(x, t)$ are gradient flows of \mathcal{E}_m satisfying

$$\rho_m(x, 0) \rightarrow \rho_\infty(x, 0) \text{ and } \mathcal{E}_m(\rho_m(x, 0)) \rightarrow \mathcal{E}_\infty(\rho_\infty(x, 0))$$

Then $\rho_m(x, t) \rightarrow \rho_\infty(x, t)$, the gradient flow of \mathcal{E}_∞ .

We also show...

Theorem:

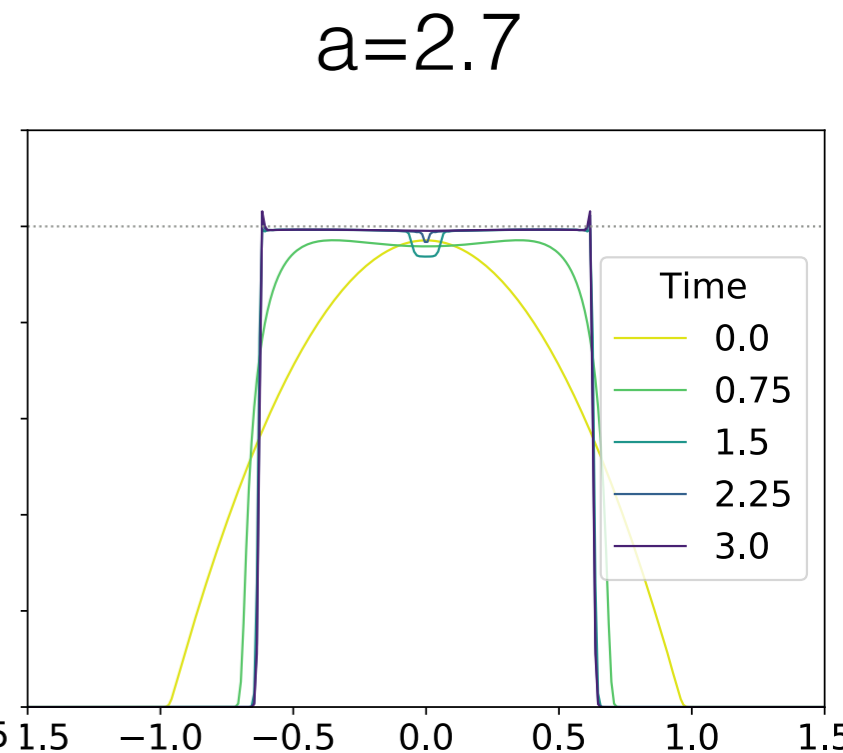
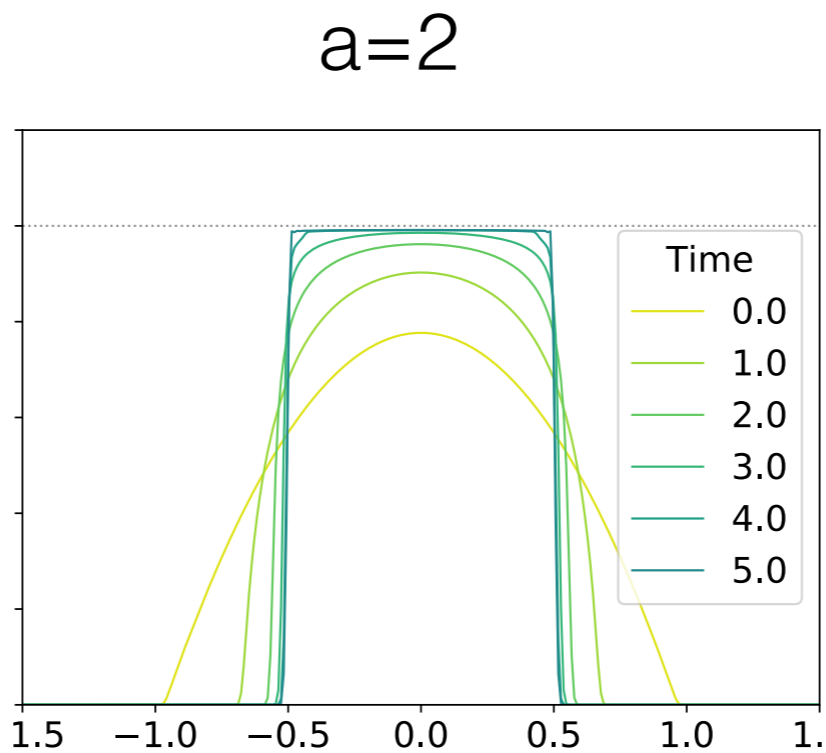
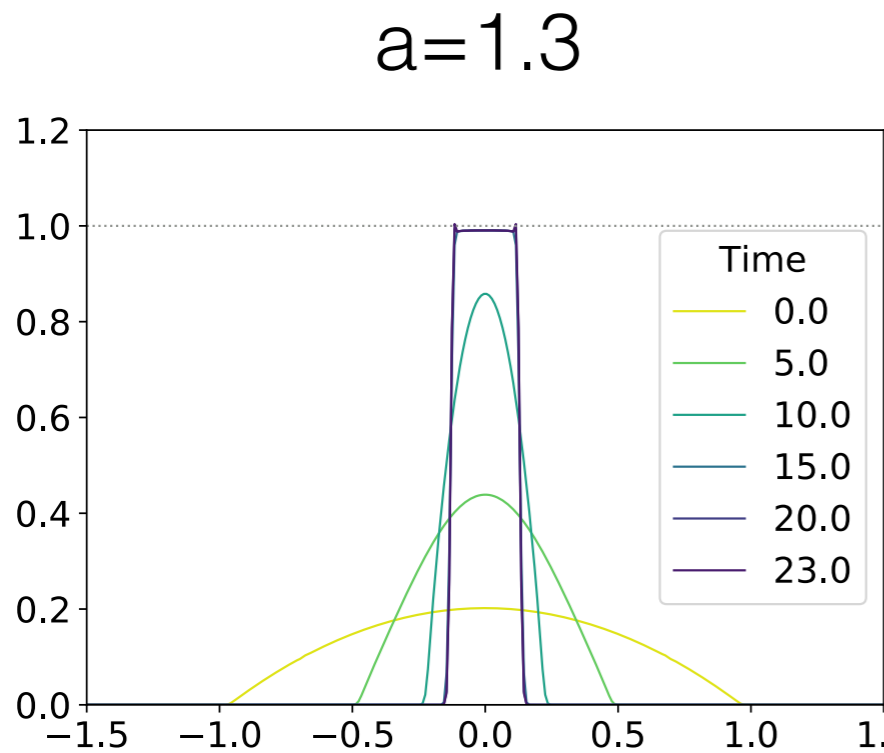
Suppose ρ_m are minimizers of \mathcal{E}_m . Then, up to a subsequence and translations, $\rho_m \rightarrow \rho_\infty$ where ρ_∞ is a minimizer of \mathcal{E}_∞ .

Thus, to gain numerical intuition for properties of minimizers of \mathcal{E}_∞ , we can simulate $\rho_m(x, t)$ for large m .

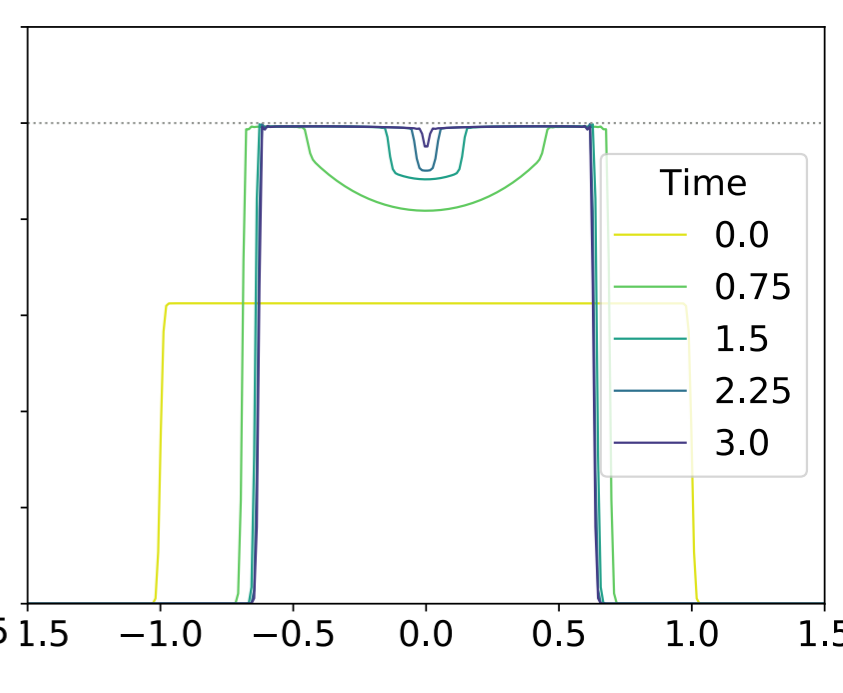
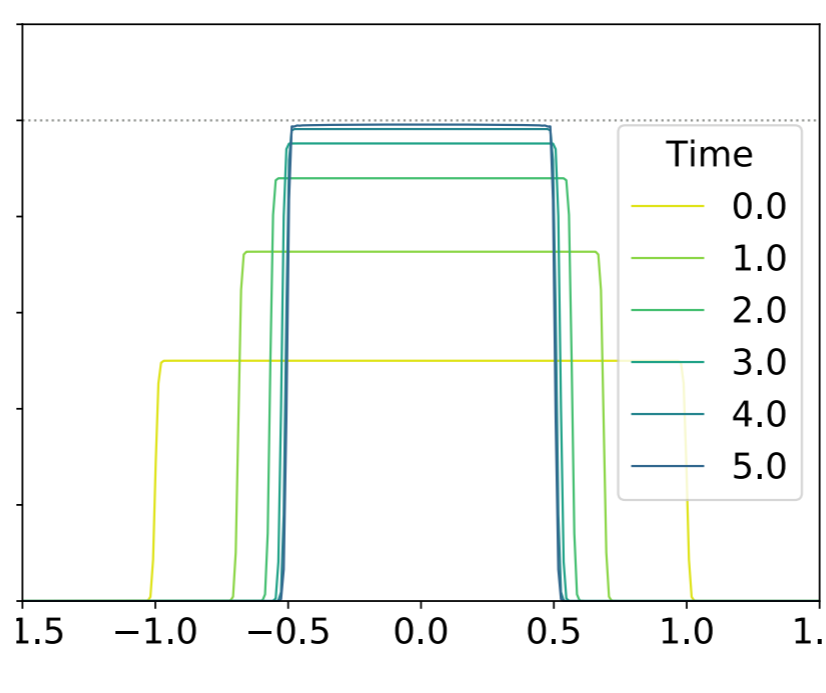
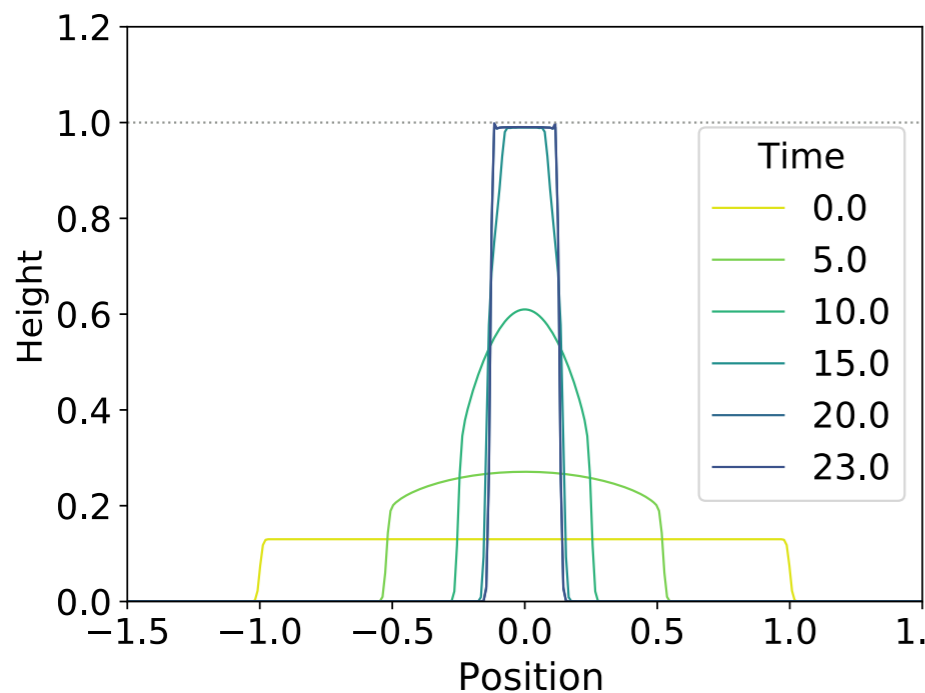
[blob method for diffusion](#) [Carrillo, Craig, Patacchini 2017]

numerics: convergence to equilibrium

barenblatt initial data



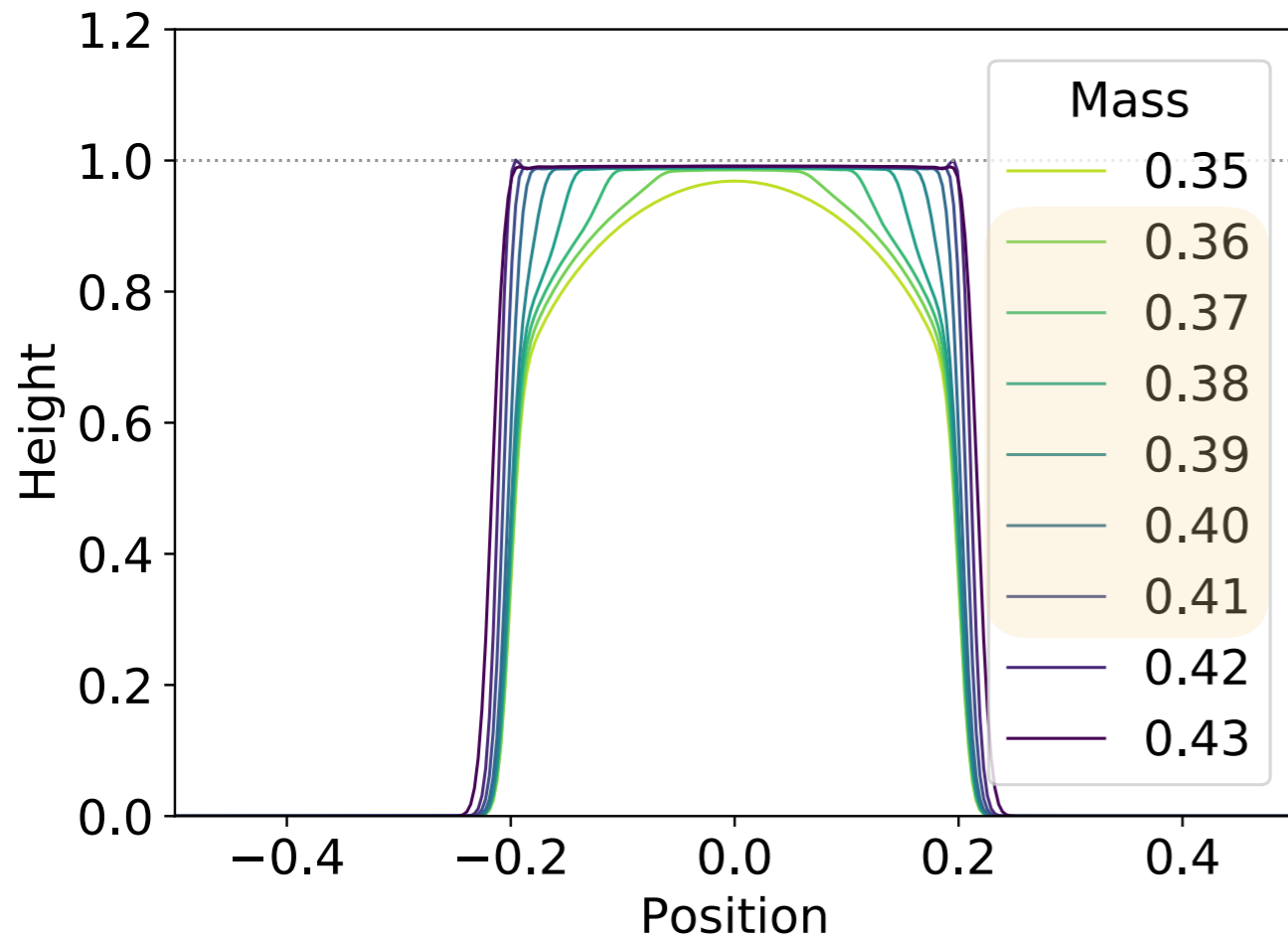
patch initial data



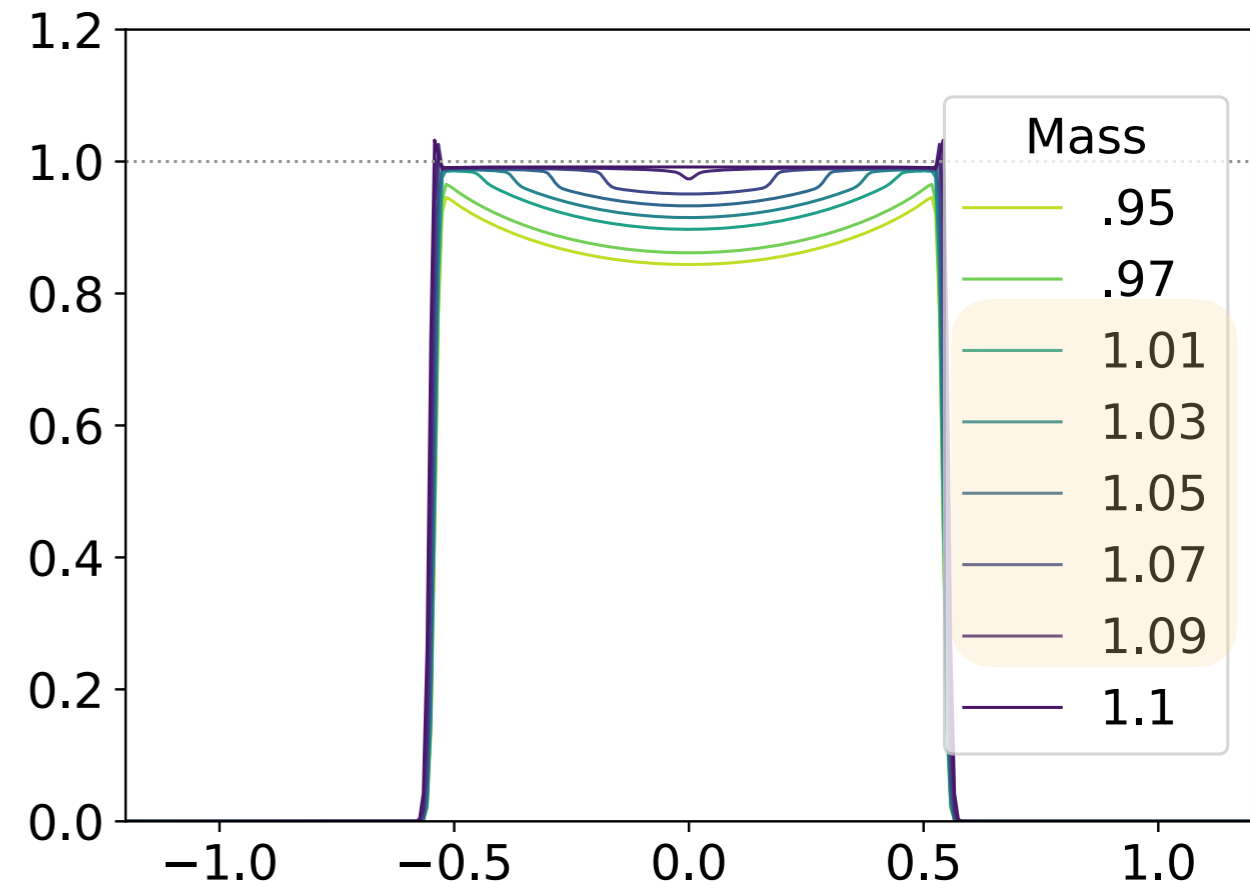
$m = 800, N_x = 500, M = \text{critical mass}$

numerics: equilibria for varying mass

$a=1.4$



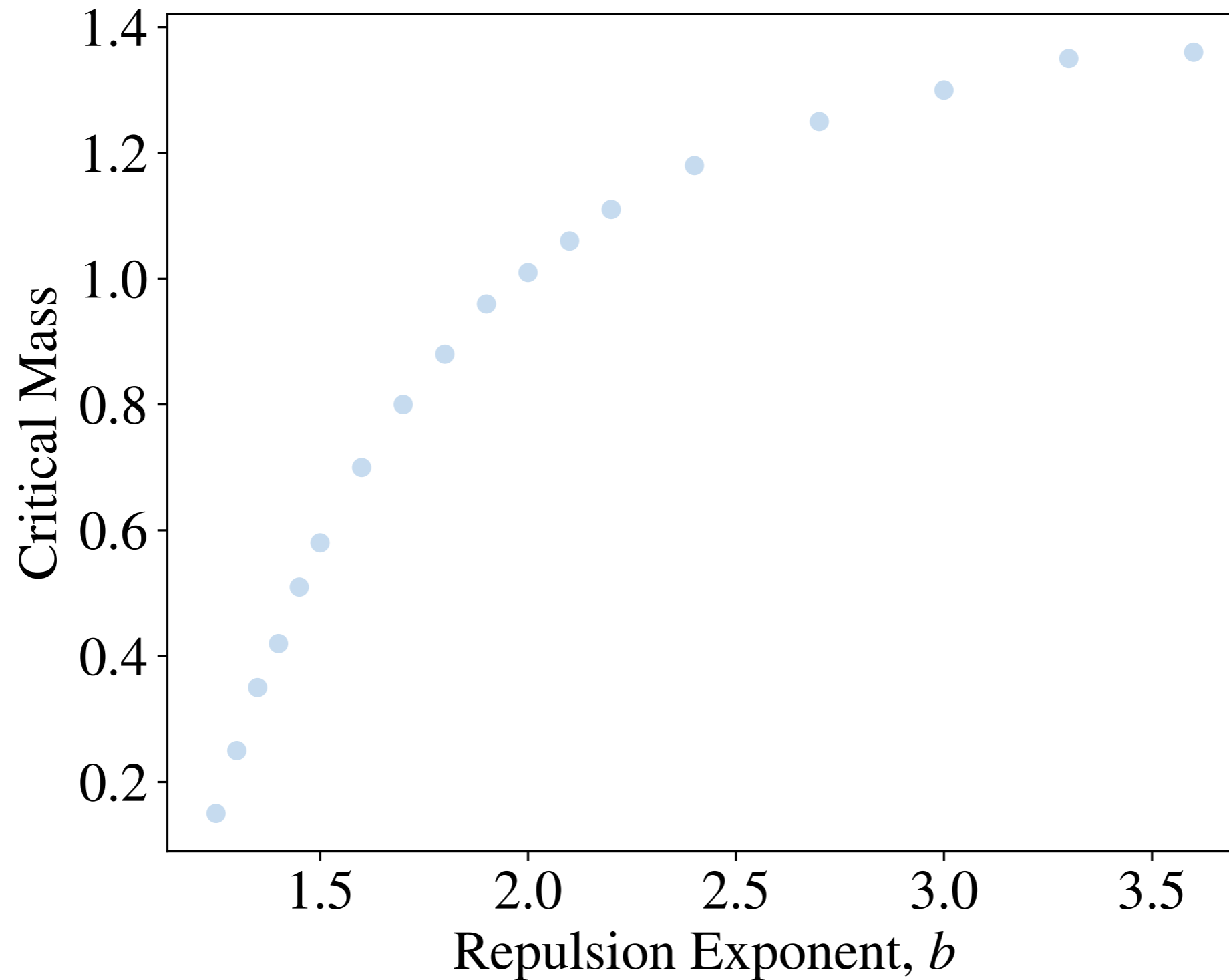
$a=2.2$



numerical evidence for intermediate phase

$m = 800, N_x = 500$

numerics: critical mass for solid state

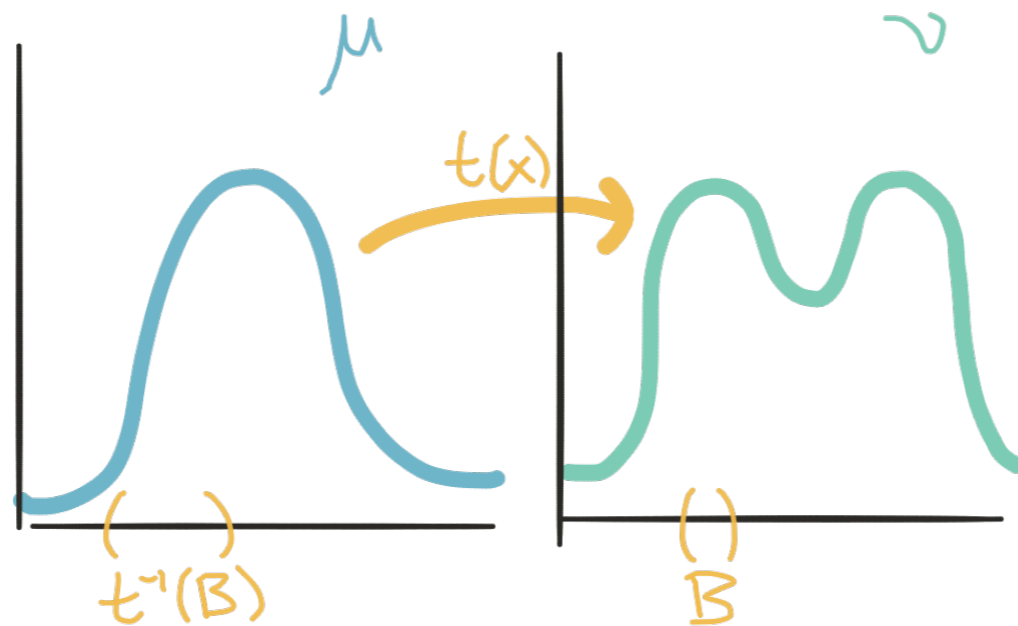


Thank you!

Backup

Wasserstein metric

- Given two probability measures μ and ν on \mathbb{R}^d , $\mathbf{t} : \mathbb{R}^d \rightarrow \mathbb{R}^d$ transports μ onto ν if $\nu(B) = \mu(\mathbf{t}^{-1}(B))$. Write this as $\mathbf{t}\#\mu = \nu$.



- The *Wasserstein distance* between μ and $\nu \in P_{2,ac}(\mathbb{R}^d)$ is

$$W_2(\mu, \nu) := \inf \left\{ \left(\int |t(x) - x|^2 d\mu(x) \right)^{1/2} : t\#\mu = \nu \right\}$$

effort to rearrange μ to look like ν , using t

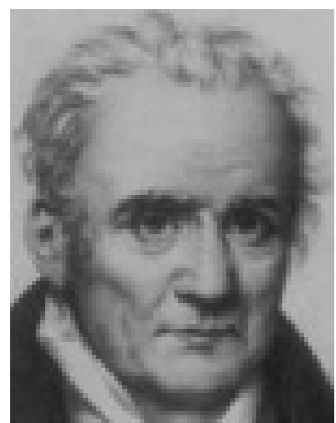
t sends μ to ν

geodesics

Not just a metric space... a **geodesic metric space**: there is a constant speed geodesic $\sigma : [0, 1] \rightarrow \mathcal{P}_2(\mathbb{R}^d)$ connecting any μ and ν .

$$\sigma(0) = \mu, \quad \sigma(1) = \nu, \quad W_2(\sigma(t), \sigma(s)) = |t - s|W_2(\mu, \nu)$$

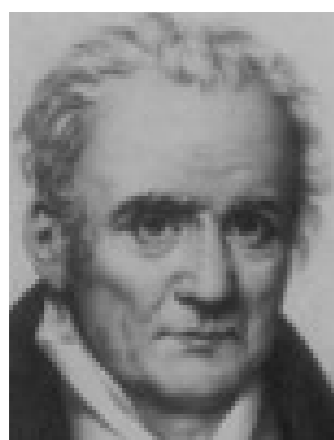
Monge



μ



Kantorovich



μ



linear interpolation $(1 - t)\mu + t\nu$

convexity

Since the Wasserstein metric has **geodesics**, it has a notion of **convexity**.

Recall: in **Euclidean space**, $E: \mathbb{R}^d \rightarrow \mathbb{R}$ is...

λ -convex $E((1-t)x + ty) \leq (1-t)E(x) + tE(y) - t(1-t)\frac{\lambda}{2}|x-y|^2$

Likewise, in the **Wasserstein metric**, $E: P_2(\mathbb{R}^d) \rightarrow \mathbb{R}$ is...

λ -convex $E(\sigma(t)) \leq (1-t)E(\mu) + tE(\nu) - t(1-t)\frac{\lambda}{2}W_2^2(\mu, \nu)$

ω -convex $E(\sigma(t)) \leq (1-t)E(\mu) + tE(\nu)$
 $- \frac{\lambda}{2} [(1-t)\omega(t^2W_2^2(\mu, \nu)) + t\omega((1-t)^2W_2^2(\mu, \nu))]$

$$\int_0^1 \frac{dx}{\omega(x)} = +\infty, \quad \text{e.g. } \omega(x) = x|\log(x)|$$

gradient flow

How does this relate to PDE?

- In general, given a complete metric space (X, d) , a curve $x(t): \mathbb{R} \rightarrow X$ is the **gradient flow** of an energy $E: X \rightarrow \mathbb{R}$ if

$$\frac{d}{dt}x(t) = -\nabla_X E(x(t))$$

- “ $x(t)$ evolves in the direction of steepest descent of E ”

Examples:

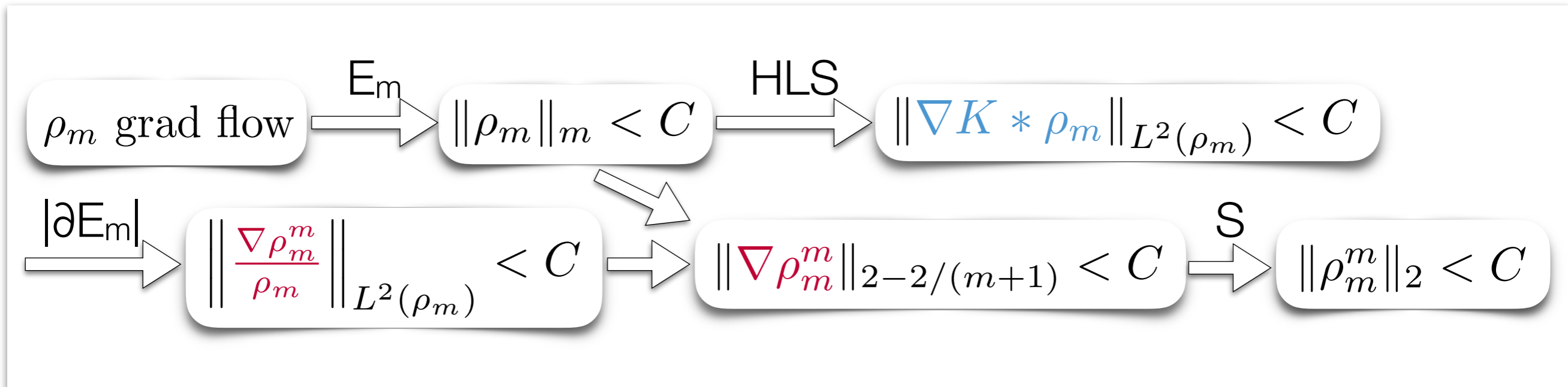
metric	energy functional	gradient flow
$(L^2(\mathbb{R}^d), \ \cdot\ _{L^2})$	$E(f) = \frac{1}{2} \int \nabla f ^2$	$\frac{d}{dt}f = \Delta f$
$(\mathcal{P}_2(\mathbb{R}^d), W_2)$	$E(\rho) = \int \rho \log \rho$	$\frac{d}{dt}\rho = \Delta \rho$
	$E(\rho) = \frac{1}{m-1} \int \rho^m$	$\frac{d}{dt}\rho = \Delta \rho^m$

slow diffusion limit

Sketch of proof (iii): $\liminf \int |\partial E_m|^2(\rho_m(t)) dt \geq \int |\partial E_\infty|^2(\rho_\infty(t))$

$$E_m(\rho) = \frac{1}{2} \int K * \rho d\rho + \frac{1}{m-1} \int \rho^m$$

$$|\partial E_m|(\rho) = \left\| \nabla K * \rho + \frac{\nabla \rho^m}{\rho} \right\|_{L^2(\rho)}$$



With this compactness, we get

$$\nabla K * \rho_m \rightarrow \nabla K * \rho, \quad \frac{\nabla \rho_m^m}{\rho_m} \rightarrow \frac{\nabla \sigma}{\rho}, \quad \liminf |\partial E_m|(\rho_m) \geq \left\| \nabla K * \rho + \frac{\nabla \sigma}{\rho} \right\|_{L^2(\rho)}$$

We conclude by showing $\text{RHS} \geq |\partial E_\infty|(\rho_\infty)$.

gradient flow

Good news: the congested aggregation equation is the Wasserstein gradient flow of the constrained interaction energy:

$$\begin{cases} \frac{d}{dt} \rho = \nabla \cdot (\nabla(K * \rho)\rho) & \text{if } \rho < 1 \\ \rho \leq 1 & \text{always} \end{cases}$$

$$E_\infty(\rho) = \begin{cases} \frac{1}{2} \int K * \rho d\rho & \text{if } \|\rho\|_\infty \leq 1 \\ +\infty & \text{otherwise} \end{cases}$$

Fact: If $K: \mathbb{R}^d \rightarrow \mathbb{R}$ is λ -convex, then E_∞ is λ -convex.

Bad news: $K(x) = \begin{cases} \frac{1}{2\pi} \log |x| & \text{if } d = 2 \\ C_d |x|^{2-d} & \text{otherwise} \end{cases}$ is not λ -convex.

E_∞ falls outside the scope of the existing theory.

ω -convexity

Solution: Even though we don't have

$$E_\infty(\sigma(t)) \leq (1-t)E_\infty(\mu) + tE_\infty(\nu) - \frac{\lambda}{2}t(1-t)W_2^2(\mu, \nu)$$

← λ -convexity

E_∞ does satisfy a similar inequality for a different modulus of convexity

$$E_\infty(\sigma(t)) \leq (1-t)E_\infty(\mu) + tE_\infty(\nu) - \frac{\lambda}{2} [(1-t)\omega(t^2W_2^2(\mu, \nu)) + t\omega((1-t)^2W_2^2(\mu, \nu))]$$

where $\omega(x) = x |\log(x)|$.

← ω -convexity

[Carrillo, McCann, Villani, 2006] [Ambrosio, Serfaty, 2008]

[Carrillo, Lisini, Mainini, 2014]

Inequalities coincide for $\omega(x) = x$; ω -convexity generalizes λ -convexity.

ω -convexity: well-posedness

For merely ω -convex energies, the gradient flow is well-posed.

Theorem (C. 2016): If E is ω -convex for $\omega(x) = x |\log(x)|$, lower semicontinuous, and bounded below, solutions of its W_2 gradient flow

- exist (quantitative JKO)
- are unique
- contract ($\lambda > 0$)/expand ($\lambda \leq 0$) double exponentially: for $W_2(\rho_1(0), \rho_2(0)) \leq 1$,
$$W_2(\rho_1(t), \rho_2(t)) \leq W_2(\rho_1(0), \rho_2(0)) e^{2\lambda t}$$

More generally, for $\omega(x)$ satisfying Osgood's condition, i.e.

$$\int_0^1 \frac{dx}{\omega(x)} = +\infty$$

we obtain the stability estimate

$$F_{2t}(W_2^2(\rho_1(t), \rho_2(t))) \leq W_2^2(\rho_1(0), \rho_2(0))$$
$$\frac{d}{dt} F_t(x) = \lambda \omega(F_t(x)), \quad F_0(x) = x$$

dynamics via free boundary problem

How does congested aggregation equation relate to free boundary problem?

$$\left\{ \begin{array}{l} \frac{d}{dt} \rho = \nabla \cdot (\nabla (K * \rho) \rho) \text{ if } \rho < 1 \\ \rho \leq 1 \text{ always} \end{array} \right.$$

Consider initial data: $\rho(x, 0) = \begin{cases} 1 & \text{if } x \in \Omega, \\ 0 & \text{otherwise.} \end{cases}$

Since $\nabla K * \rho$ causes self-attraction, we expect $\rho(x, t) = 1_{\Omega(t)}(x)$.

Theorem (C., Kim, Yao '18):

Suppose $\rho(x, t)$ solves congested aggregation eqn with $\rho(x, 0) = 1_{\Omega(0)}(x)$.

Then $\rho(x, t) = 1_{\Omega(t)}(x)$, for $\Omega(t) = \{\mathbf{p}(x, t) > 0\}$, where \mathbf{p} a viscosity solution of

$$\left\{ \begin{array}{ll} -\Delta \mathbf{p} = 1 & \text{on } \{\mathbf{p} > 0\} \\ V = -\partial_\nu K * 1_{\{\mathbf{p} > 0\}} - \partial_\nu \mathbf{p} & \text{on } \partial\{\mathbf{p} > 0\}. \end{array} \right.$$

long time behavior

Using free boundary characterization, we can describe long time behavior:

- In **any dimension**, the Riesz Rearrangement Inequality guarantees that the unique minimizer of E_∞ is $1_B(x)$.
- Need to show mass of $\rho(x,t)$ doesn't escape to $+\infty$. To accomplish this, we use an inequality due to Talenti, which holds in **d=2**.

Theorem (C., Kim, Yao 2016):

- Suppose $\rho(x,t)$ solves **congested aggregation eqn** with $\rho(x,0) = 1_{\Omega(0)}(x)$.
- Then, in **two dimensions**,

$$\rho(x, t) \xrightarrow{L^p} 1_B(x) \text{ for all } 1 \leq p < +\infty$$

and

$$|E_\infty(\rho(\cdot, t)) - E_\infty(1_B)| \leq C_{\Omega(0)} t^{-1/6}$$

questions

$$K = \Delta^{-1}$$

Congested aggregation eqn:

$$\left\{ \begin{array}{l} \frac{d}{dt} \rho = \nabla \cdot (\nabla (K * \rho) \rho) \text{ if } \rho < 1 \\ \rho \leq 1 \text{ always} \end{array} \right.$$

- Well-posed? (nonconvex) Wasserstein gradient flow
- Dynamics/long time behavior? gradient flow + viscosity solution theory
- Slow diffusion limit? Gamma convergence

previous work

Congested drift equation:

$$\begin{cases} \frac{d}{dt}\rho = \nabla \cdot ((\nabla V)\rho) & \text{if } \rho < 1 \\ \rho \leq 1 & \text{always} \end{cases}$$

[Maury, Roudneff-Chupin, Santambrogio 2010]

- introduced as a model of crowd motion in an evacuation scenario, where $V(x) = \text{distance to exit}$.
- showed well-posedness as a W_2 gradient flow for $V(x)$ convex.

[Alexander, Kim, Yao 2014]

- for $\Delta V > 0$, characterized patch dynamics via free boundary problem

$$\begin{cases} -\Delta \mathbf{p} = \Delta \Phi & \text{on } \{\mathbf{p} > 0\} \\ V = -\partial_\nu V - \partial_\nu \mathbf{p} & \text{on } \partial\{\mathbf{p} > 0\}. \end{cases}$$

gradient flow

We want to define the gradient flow as $\frac{d}{dt}\rho(t) = -\nabla_{W_2} E(\rho(t))$,
but without a Riemannian structure, we don't have a notion of **gradient**.

- Given $E: P_2(\mathbb{R}^d) \rightarrow \mathbb{R}$, its **local slope** is:

$$|\partial E|(\mu) := \limsup_{\nu \rightarrow \mu} \frac{(E(\mu) - E(\nu))^+}{W_2(\mu, \nu)}$$

- Given $\rho: [0, T] \rightarrow P_2(\mathbb{R}^d)$, its **metric derivative** is:

$$|\rho'| (t) = \lim_{s \rightarrow t} \frac{W_2(\rho(s), \rho(t))}{|s - t|}$$

DEF: $\rho(t): \mathbb{R} \rightarrow P_2(\mathbb{R}^d)$ is the **Wasserstein gradient flow** of $E: P_2(\mathbb{R}^d) \rightarrow \mathbb{R}$ if

$$\frac{d}{dt} E(\rho(t)) \leq -\frac{1}{2} |\partial E(\rho(t))| - \frac{1}{2} |\rho'| (t)$$

gradient flow

$\rho(t): \mathbb{R} \rightarrow P_2(\mathbb{R}^d)$ is the **gradient flow** of energy $E: P_2(\mathbb{R}^d) \rightarrow \mathbb{R}$ if

$$\frac{d}{dt}x(t) = -\nabla_X E(x(t))$$

More precisely, $\rho(t)$ is the **gradient flow** of E if...

- $\exists v(t) \in L^2_{\text{loc}}((0, +\infty), L^2(\rho(t)))$ s.t.
- $-v(t) \in \partial E(\rho(t))$ for a.e. $t > 0$

$$\frac{d}{dt}\rho(x, t) + \nabla$$

The term brackets is analogous to $\xi(v - \rho)$

Tangent space?

- ξ belongs to the **subdifferential** of E at ρ if as $\mu \rightarrow \nu$,

$$E(\nu) - E(\rho) \geq \int \langle \xi, \mathbf{t}_\rho^\nu - \text{id} \rangle d\mu + o(W_2(\rho, \nu))$$

- If E and ρ are **nice**, $\partial E(\rho) = \left\{ \nabla \frac{\partial E}{\partial \rho} \right\}$

- Then solutions of the gradient flow can be characterized via a PDE.

aside: ω -convexity & Euler equations

In fact, when $\omega(x) = x |\log(x)|$, ω -convexity is related to **well-posedness** of **bounded** solutions of the the **Euler equations**.

- λ -convexity in W_2 is analogous to D^2E being **bounded** from below in Euclidean space, or that ∇E is one-sided Lipschitz.
- Likewise, ω -convexity in W_2 is analogous to D^2E being **BMO** in Euclidean space, or that ∇E is log-Lipschitz.
- Log-Lipschitz regularity of the velocity field was precisely what allowed **[Yudovich 1963]** to prove uniqueness of bounded solutions of the two dimensional Euler equations.

ω -convexity

Examples:

$$\frac{d}{dt}\rho = \nabla \cdot ((\nabla K * \rho)\rho) + \Delta \rho^m$$

• Chemotaxis: $K(x) = \begin{cases} \frac{1}{2\pi} \log |x| & \text{if } d = 2 \\ C_d |x|^{2-d} & \text{otherwise} \end{cases}$

ω -convex on L^∞

• Swarming: $K(x) = |x|^a/a - |x|^b/b, \quad 2 - d \leq b < a$

ω -convex on $L^p,$
 $p \geq d/(b+d-2)$

• Granular media: $K(x) = |x|^3$

ω -convex on measures with
fixed center of mass; $\omega(x) = x^{3/2}$

Sufficient condition:

Above the tangent line inequality

$$E(\mu_1) - E(\mu_0) - \frac{d}{d\alpha} E(\mu_\alpha)|_{\alpha=0} \geq \frac{\lambda}{2} \omega(W_2^2(\mu_0, \mu_1))$$

motivation for free boundary problem

How does congested aggregation equation relate to free boundary problem?

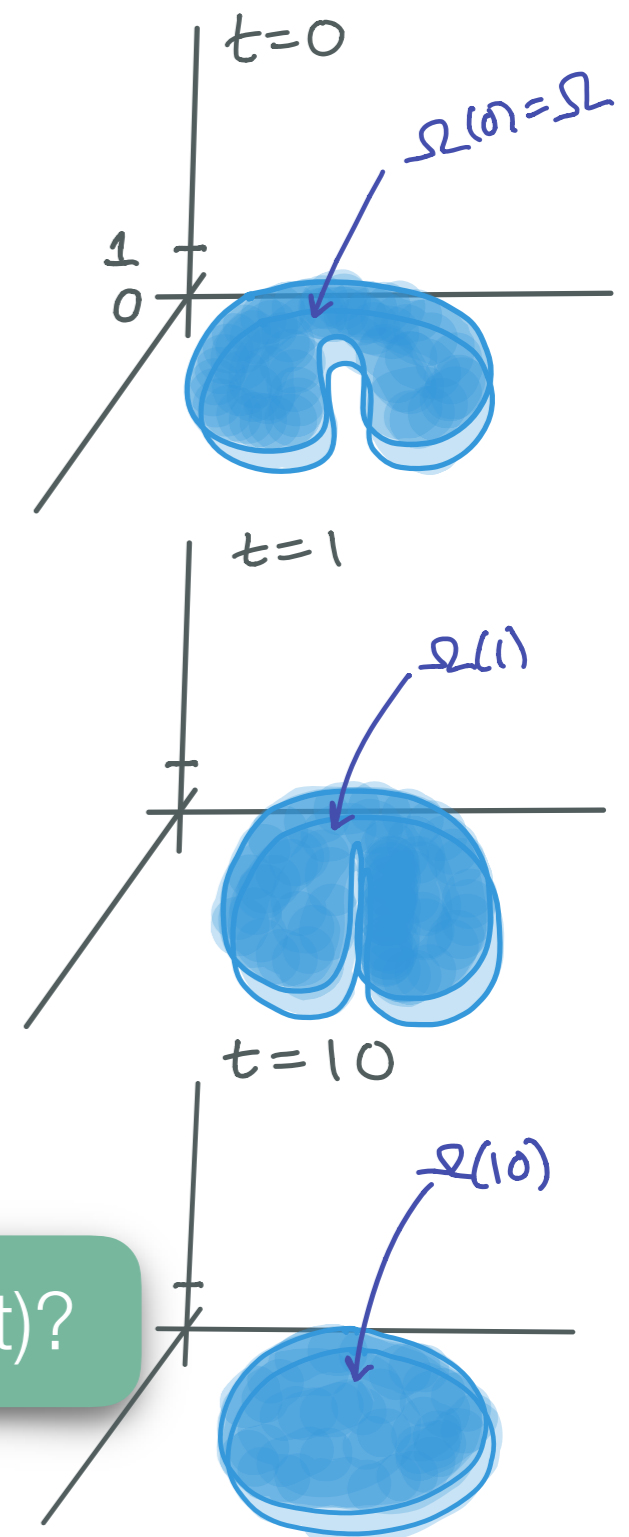
“

$$\begin{cases} \frac{d}{dt}\rho = \nabla \cdot (\nabla(K * \rho)\rho) & \text{if } \rho < 1 \\ \rho \leq 1 & \text{always} \end{cases}$$

”

- Consider **patch solutions**. For a domain Ω , suppose that $\rho(x,t)$ is a solution with initial data
$$\rho(x, 0) = \begin{cases} 1 & \text{if } x \in \Omega, \\ 0 & \text{otherwise.} \end{cases}$$
- Since $K = \Delta^{-1}$, $\nabla K * \rho$ causes **self-attraction**. Thus, we expect $\rho(x,t)$ to remain a characteristic function.
- Let $\Omega(t) = \{\rho = 1\}$ be **congested region**, so $\rho(x,t) = 1_{\Omega(t)}(x)$.

What free boundary problem describes evolution of $\Omega(t)$?



formal derivation

- Here is a **formal** derivation of the related free boundary problem.

- Suppose $\rho(x,t)$ solves “
$$\begin{cases} \frac{d}{dt}\rho = \nabla \cdot (\nabla(K * \rho)\rho) & \text{if } \rho < 1 \\ \rho \leq 1 & \text{always} \end{cases}$$
”

- Since mass is conserved, we expect $\rho(x,t)$ satisfies a continuity equation

$$\frac{d}{dt}\rho = \nabla \cdot \underbrace{((\nabla K * \rho + \nabla \mathbf{p}))\rho}_v$$

where $\nabla \mathbf{p}(x,t)$ is the pressure arising from the **height constraint**.

Height constraint is **active** on the congested region $\{\mathbf{p} > 0\} = \Omega(t)$.

Height constraint is **inactive** outside the congested region $\{\mathbf{p} = 0\} = \Omega(t)^c$.

formal derivation

Given $\frac{d}{dt}\rho = \nabla \cdot \underbrace{((\nabla K * \rho + \nabla \mathbf{p})) \rho}_v$ what happens on congested region?

- Because of hard height constraint, on the congested region $\Omega(t)=\{\rho=1\}$, the velocity field is incompressible, $\nabla \cdot v=0$.
- Since $K = \Delta^{-1}$, $\nabla \cdot v = \Delta K * \rho + \Delta \mathbf{p} = \rho + \Delta \mathbf{p}$, so incompressibility means

$$-\Delta \mathbf{p} = \rho \text{ on } \Omega(t) = \{\rho = 1\}$$

- Using that the height constraint is active on the congested region, $\Omega(t)=\{\mathbf{p}>0\}$, we obtain the following equation for the pressure:

$$-\Delta \mathbf{p} = 1 \text{ on } \{\mathbf{p} > 0\}$$

formal derivation

Given $\frac{d}{dt}\rho = \nabla \cdot \underbrace{((\nabla K * \rho + \nabla \mathbf{p}))\rho}_v$ what about bdy of congested region?

outward normal velocity of $\partial\Omega(t)$

- By conservation of mass,

$$0 = \frac{d}{dt} \int_{\Omega(t)} \rho = \int_{\Omega(t)} \frac{d}{dt} \rho + \int_{\partial\Omega(t)} V \rho$$

- Using that $\rho(x,t)$ solves the above continuity equation, this equals

$$= \int_{\Omega(t)} \nabla \cdot ((\nabla K * \rho + \nabla \mathbf{p}))\rho + \int_{\partial\Omega(t)} V \rho = \int_{\partial\Omega(t)} (\partial_\nu K * \rho + \partial_\nu \mathbf{p} + V)\rho$$

- Since $\rho(x,t) = 1_{\Omega(t)}(x)$, for $\Omega(t) = \{\mathbf{p} > 0\}$, we again obtain an equation for \mathbf{p} ,

$$\partial_\nu K * 1_{\{\mathbf{p} > 0\}} + \partial_\nu \mathbf{p} + V = 0 \text{ on } \partial\{\mathbf{p} > 0\}$$

free boundary problem

Combining the observations that...

- on the congested region,

$$-\Delta \mathbf{p} = 1 \text{ on } \{\mathbf{p} > 0\}$$

- and on the boundary of the congested region,

$$\partial_\nu K * 1_{\{\mathbf{p} > 0\}} + \partial_\nu \mathbf{p} + V = 0 \text{ on } \partial\{\mathbf{p} > 0\}$$

outward normal
velocity of $\partial\Omega(t)$

Remind myself the hoops we had to jump through to even define viscosity solutions

Theorem (C., Kim, Yao 2016):

- Suppose $\rho(x,t)$ solves congested aggregation eqn with $\rho(x,0) = 1_{\Omega(0)}(x)$.
- Then $\rho(x,t) = 1_{\Omega(t)}(x)$, for $\Omega(t) = \{\mathbf{p}(x,t) > 0\}$, where \mathbf{p} a viscosity solution of

$$\begin{cases} -\Delta \mathbf{p} = 1 & \text{on } \{\mathbf{p} > 0\} \\ V = -\partial_\nu K * 1_{\{\mathbf{p} > 0\}} - \partial_\nu \mathbf{p} & \text{on } \partial\{\mathbf{p} > 0\}. \end{cases}$$

future work

Does Keller-Segel converge to congested aggregation?

$$\frac{d}{dt}\rho = \nabla \cdot ((\nabla K * \rho)\rho) + \Delta \rho^m$$

$m \rightarrow +\infty$

$$\begin{cases} \frac{d}{dt}\rho = \nabla \cdot (\nabla(K * \rho)\rho) & \text{if } \rho < 1 \\ \rho \leq 1 & \text{always} \end{cases}$$

- [Alexander, Kim, Yao 2014] showed the analogous result for a convex drift potential $V(x)$.
- **Obstacle:** Lack of uniform convexity as $m \rightarrow +\infty$.

Non-patch solutions?

- Recent work on $m \rightarrow +\infty$ limit in PME-type models for tumor growth with source [Kim and Pozar 2015], [Mellet, Perthame, Quiros 2015] and drift [Kim, Pozar, Woodhouse 2017].
- Can this be extended to include nonlocal interaction?
- **Obstacle:** Nucleation of new congested regions, infinite speed of propagation, neck pinching...

future work

Other characterizations of dynamics?

- Can we show $\frac{d}{dt}\rho = \nabla \cdot \underbrace{((\nabla K * \rho + \nabla \mathbf{p})) \rho}_v$ in a weak sense?
- For the congested drift equation [Maury, Roudneff-Chupin, Santambrogio 2010] showed that the analogous continuity equation holds, where v is obtained by projecting ∇V onto a space of admissible velocities.
- **Obstacle:** With a nonlocal interaction term K , projection would depend nonlocally on ρ .

Height constrained aggregation with non-Newtonian kernels?

- Well-posedness theory extends to a range of interaction kernels
- **Obstacle:** Free boundary problem strongly uses Newtonian structure

Further examples of ω -convex energies? Problems with height constraint?