

# From slow diffusion to a hard height constraint: characterizing congested aggregation

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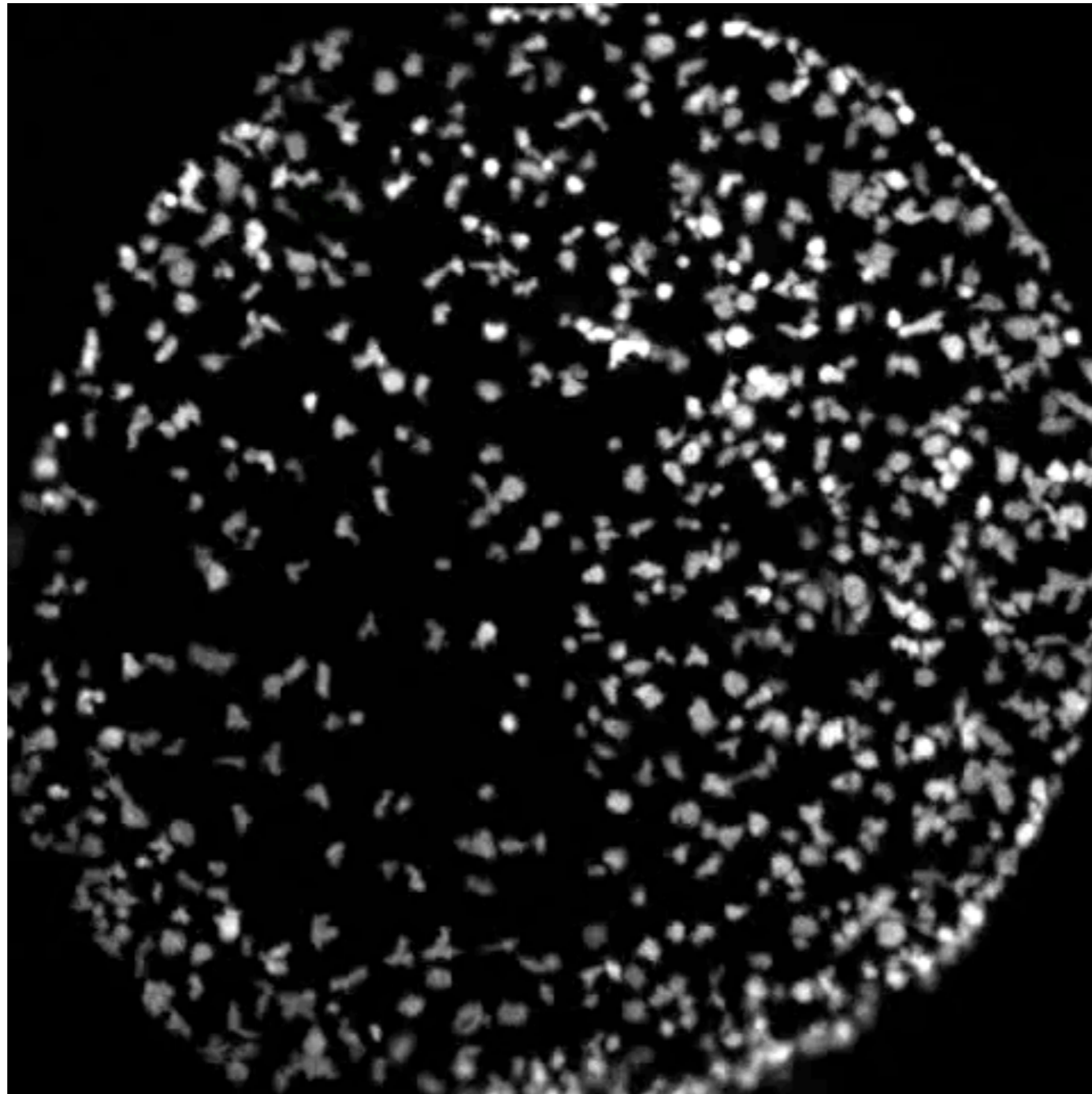
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# collective dynamics

biological chemotaxis (a colony of slime mold)



# plan

- collective dynamics
- optimal transport and Wasserstein gradient flow
- $\omega$ -convexity and height constrained aggregation
- future work

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# motivation

- $\rho(x,t): \mathbb{R}^d \times \mathbb{R} \rightarrow [0, +\infty)$  nonnegative density
- mass is conserved  $\Rightarrow \int \rho(x) dx = 1$

aggregation equation with degenerate diffusion:

$$\frac{d}{dt}\rho = \underbrace{\nabla \cdot ((\nabla K * \rho)\rho)}_{\text{self attraction}} + \underbrace{\Delta \rho^m}_{\text{degenerate diffusion}} \quad \text{for } K(x) : \mathbb{R}^d \rightarrow \mathbb{R} \text{ and } m \geq 1$$

self attraction    degenerate diffusion

interaction kernels:

- granular media:  $K(x) = |x|^3$
- swarming:  $K(x) = |x|^a/a - |x|^b/b, -d < b < a$
- chemotaxis:  $K(x) = \begin{cases} \frac{1}{2\pi} \log |x| & \text{if } d = 2, \\ C_d |x|^{2-d} & \text{otherwise.} \end{cases}$

degenerate diffusion:

$$\Delta \rho^m = \nabla \cdot \underbrace{(m \rho^{m-1} \nabla \rho)}_D$$

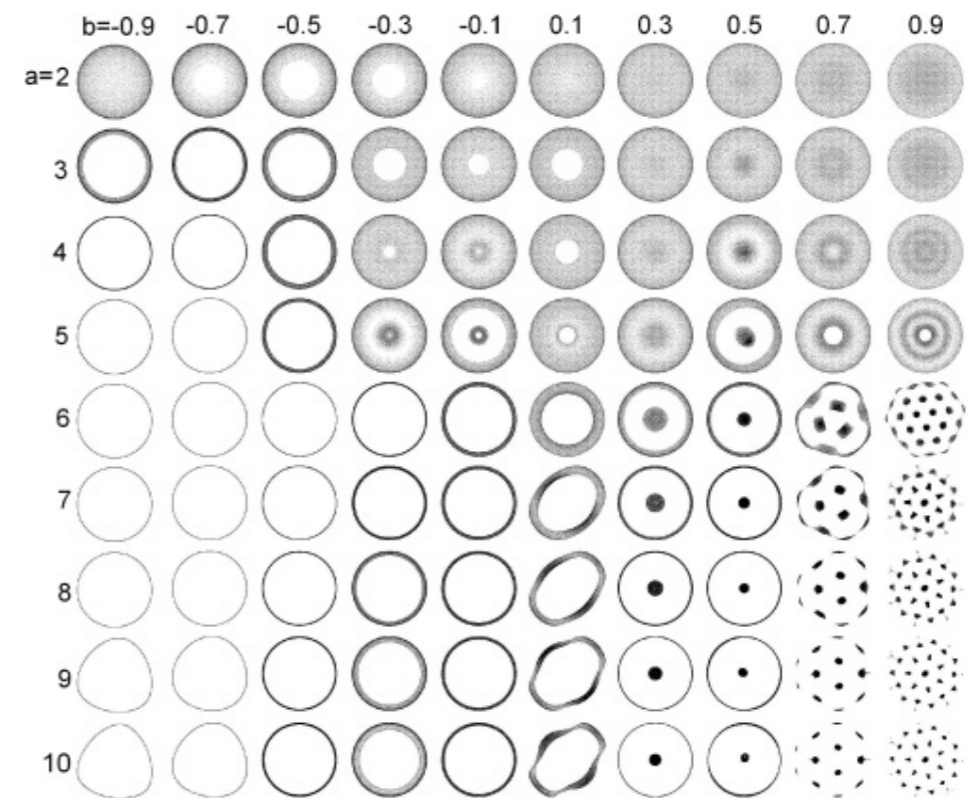
# collective dynamics: mathematics

Aggregation equation with degenerate diffusion:

$$\frac{d}{dt}\rho + \nabla \cdot ((-\nabla K * \rho)\rho) = \Delta \rho^m \quad \text{for } K(x) : \mathbb{R}^d \rightarrow \mathbb{R} \text{ and } m \geq 1$$

## Mathematical interest:

- Nonlinear
- Nonlocal
- Competing effects of attraction/repulsion
- Rich structure of equilibria



[Kolokolnikov, Sun, Uminsky, Bertozzi, 2011]

# collective dynamics: main questions

Aggregation equation with degenerate diffusion:

$$\frac{d}{dt}\rho + \nabla \cdot ((-\nabla K * \rho)\rho) = \Delta \rho^m \quad \text{for } K(x) : \mathbb{R}^d \rightarrow \mathbb{R} \text{ and } m \geq 1$$

Main questions:

1. Do solutions exist?
2. Are they unique? stable?
3. How do they behave in the long time limit?
4. How can we simulate them numerically?

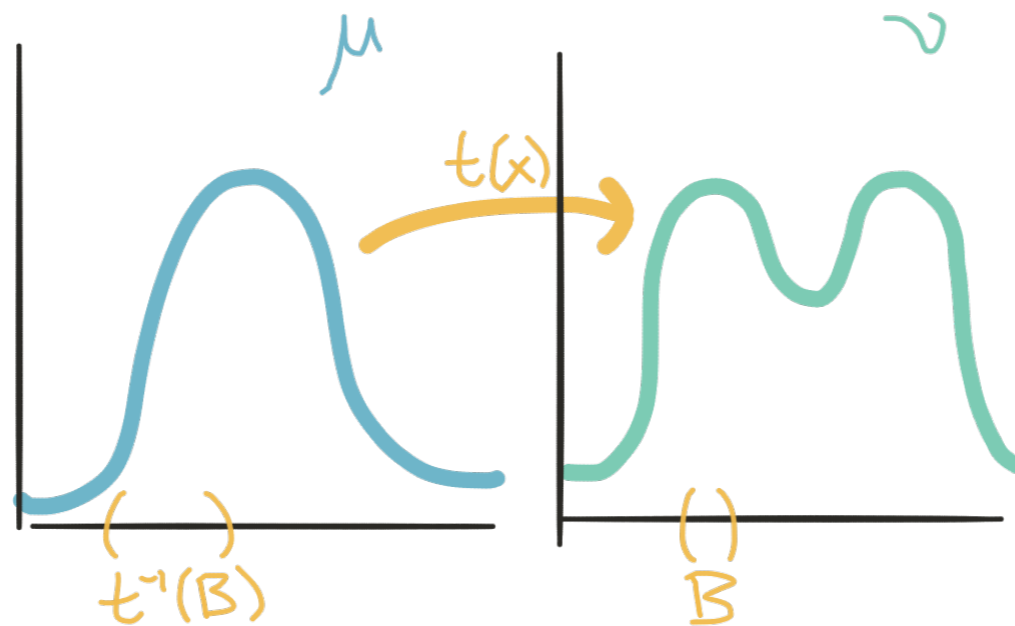
Key tool: optimal transport

# plan

- collective dynamics
- optimal transport and Wasserstein gradient flow
- $\omega$ -convexity and height constrained aggregation
- future work

# Wasserstein metric

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



- The *Wasserstein distance* between  $\mu$  and  $\nu \in P_2(\mathbb{R}^d)$  is

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

For simplicity of notation,  
 $\mu, \nu \ll \mathcal{L}^d$

effort to rearrange  $\mu$  to  
 look like  $\nu$ , using  $t(x)$

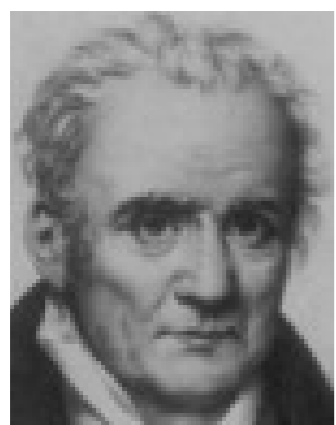
$t$  sends  $\mu$  to  $\nu$

# geodesics

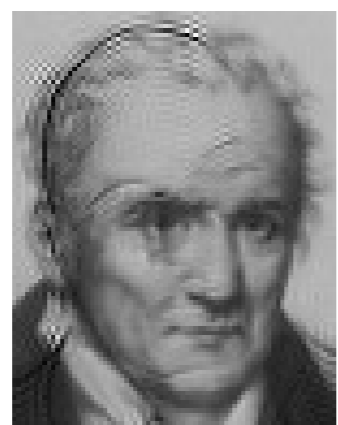
Not just a metric space... a **geodesic metric space**: there is a constant speed geodesic  $\sigma : [0, 1] \rightarrow \mathcal{P}_2(\mathbb{R}^d)$  connecting any  $\mu$  and  $\nu$ .

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

Monge



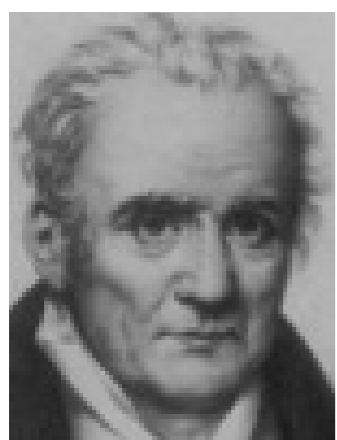
$\mu$



Wasserstein geodesic  $\sigma(t)$

$\nu$

Kantorovich



$\mu$



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

$\nu$

# 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...

convex      Euclidean geodesic      endpoints

$$D^2E \geq 0 \iff E((1-t)x + ty) \leq (1-t)E(x) + tE(y)$$

$\lambda$ -convex

$$D^2E \geq \lambda I_{d \times d} \iff E((1-t)x + ty) \leq (1-t)E(x) + tE(y) - t(1-t)\frac{\lambda}{2}|x-y|^2$$

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

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

Wasserstein geodesic      endpoints

# gradient flow

## How does this relate to PDE? Wasserstein gradient flow.

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



# gradient flow

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

$$\text{“} \frac{d}{dt} \rho(t) = -\nabla_{W_2} E(\rho(t)) \text{”}$$

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

- there exists  $v(t) \in L^2_{\text{loc}}((0, +\infty), L^2(\rho(t)))$  so that

$$\frac{d}{dt} \rho(x, t) + \nabla \cdot (v(x, t) \rho(x, t)) = 0$$

- for a.e.  $t > 0$ ,  $-v(t) \in \partial E(\rho(t))$

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

- If  $E$  and  $\rho$  are nice,  $\partial E(\rho) = \left\{ \nabla \frac{\partial E}{\partial \rho} \right\}$ , and solutions of the gradient flow can be characterized as solutions to a PDE.

# collective dynamics: main questions

Aggregation equation with degenerate diffusion:

$$\frac{d}{dt}\rho + \nabla \cdot ((-\nabla K * \rho)\rho) = \Delta \rho^m \quad \text{for } K(x) : \mathbb{R}^d \rightarrow \mathbb{R} \text{ and } m \geq 1$$

$$E(\mu) = \iint K(x - y) d\mu(x) d\mu(y) + \frac{1}{m - 1} \int \mu(x)^m dx$$

Main questions:

1. Do solutions exist?
2. Are they unique? stable?
3. How do they behave in the long time limit?
4. How can we simulate them numerically?

If  $K(x)$  is  $\lambda$ -convex,  $\lambda \leq 0$ , so is  $E(\mu)$  [CDFLS, 2011].  
But what about when  $K(x)$  isn't  $\lambda$ -convex?

Aggregation equation with degenerate diffusion:

$$\frac{d}{dt}\rho + \nabla \cdot ((-\nabla K * \rho)\rho) = \Delta \rho^m \quad \text{for } K(x) : \mathbb{R}^d \rightarrow \mathbb{R} \text{ and } m \geq 1$$

$$E(\mu) = \iint K(x - y) d\mu(x) d\mu(y) + \frac{1}{m-1} \int \mu(x)^m dx$$

**Theorem** (Ambrosio, Gigli, Savaré 2005): If the energy is  $\lambda$ -convex,

1. Do solutions exist? **Yes (JKO)**
2. Are they unique? **Yes** stable? **contract** ( $\lambda > 0$ )/**expand** ( $\lambda \leq 0$ ) exponentially
3. How do they behave in the long time limit? **For**  $\lambda > 0$ , there is a unique **steady state**, which solutions approach exponentially quickly.
4. How can we simulate them numerically?

# collective dynamics: applications

Aggregation equation with degenerate diffusion:

$$\frac{d}{dt}\rho + \nabla \cdot ((-\nabla K * \rho)\rho) = \Delta \rho^m \quad \text{for } K(x) : \mathbb{R}^d \rightarrow \mathbb{R} \text{ and } m \geq 1$$

Applied interest:

- Slime mold (chemotaxis):  $K(x) = \begin{cases} \frac{1}{2\pi} \log |x| & \text{if } d = 2, \\ C_d |x|^{2-d} & \text{otherwise.} \end{cases}$
- Swarming:  $K(x) = |x|^a/a - |x|^b/b, \quad -d < b < a$  not  $\lambda$ -convex
- Granular media:  $K(x) = |x|^3$  “merely” 0-convex

# plan

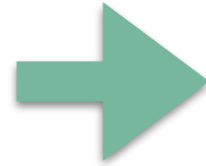
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- $\omega$ -convexity and height constrained aggregation
- future work

# height constrained aggregation

**a new model** (C., Kim, Yao 2016):

inspired by the aggregation equation with degenerate diffusion, we consider a height constrained aggregation equation, for  $K = \Delta^{-1}$

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



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

- Both models have self-attraction from  $\nabla K * \rho$ .
- The role of repulsion is played by hard height constraint instead of degenerate diffusion.
- Heuristically, hard height constraint is singular limit of degenerate diffusion:

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

# height constrained aggregation

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

- **Hard height constraint** appeared in previous work by [Maury, Roudneff-Chupin, Santambrogio 2010]—instead of  $K * \rho(x)$  had  $V(x)$ .
- Has a (formal) Wasserstein gradient flow structure:

equation

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

energy

$$E(\mu) = \iint K(x - y) d\mu(x) d\mu(y) + \frac{1}{m - 1} \int \mu(x)^m dx$$

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

$$E_\infty(\mu) = \begin{cases} \iint K(x - y) d\mu(x) d\mu(y) & \text{if } \|\mu\|_{L^\infty} \leq 1 \\ +\infty & \text{otherwise} \end{cases}$$

Since  $K(x)$  is not  $\lambda$ -convex,  $E_\infty$  falls outside the scope of the existing theory.

# $\omega$ -convexity

Even though we don't have

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

$\lambda$ -convexity

$E_\infty$  does satisfy a similar inequality for a modulus of convexity  $\omega(x) = x |\log(x)|$ .

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

$\omega$ -convexity

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

[Carrillo, Lisini, Mainini, 2014]

- Inequalities coincide for  $\omega(x) = x$ ;  $\omega$ -convexity generalizes  $\lambda$ -convexity.
- Sufficient condition: above the tangent line inequality

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



# collective dynamics: main questions

Aggregation equation with degenerate diffusion:

$$\frac{d}{dt}\rho + \nabla \cdot ((-\nabla K * \rho)\rho) = \Delta \rho^m \quad \text{for } K(x) : \mathbb{R}^d \rightarrow \mathbb{R} \text{ and } m \geq 1$$

$$E(\mu) = \iint K(x - y) d\mu(x) d\mu(y) + \frac{1}{m - 1} \int \mu(x)^m dx$$

Main questions:

1. Do solutions exist?
2. Are they unique? stable?
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# collective d

Aggregation equation

$$\frac{d}{dt}\rho + \nabla \cdot ((-\nabla K$$

$$E(\mu) = \iint$$

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

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

we obtain the stability estimate

$$F_{2t}(W_2^2(\rho_1(t), \rho_2(t))) \leq W_2^2(\rho_1(0), \rho_2(0))$$

$$\frac{d}{dt}F_t(x) = \lambda \omega(F_t(x))$$

from which we recover [AGS, 2005] & [CMV, 2006].

**Theorem** (C. 2016): If the energy is  $\omega$ -convex,  $\omega(x) = x |\log(x)|$ ,

1. Do solutions exist? **Yes (JKO)**
2. Are they unique? **Yes** stable? **expand at most double-exponentially**

$$W_2^2(\rho_1(t), \rho_2(t)) \leq W_2^2(\rho_1(0), \rho_2(0)) e^{2\lambda t}$$

# $\omega$ -convexity: applications

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

- Slime mold singular limit:  $K(x) = \begin{cases} \frac{1}{2\pi} \log |x| & \text{if } d = 2 \\ C_d |x|^{2-d} & \text{otherwise} \end{cases}$   $\leftarrow$   $\omega$ -convex

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

- Slime mold (chemotaxis):  $K(x) = \begin{cases} \frac{1}{2\pi} \log |x| & \text{if } d = 2 \\ C_d |x|^{2-d} & \text{otherwise} \end{cases}$   $\leftarrow$   $\omega$ -convex on bounded measures

- Swarming:  $K(x) = |x|^a/a - |x|^b/b, \quad -d < b < a$   $\leftarrow$   $\omega$ -convex on  $L^p$  measures for  $2-d \leq b < a$

- Granular media:  $K(x) = |x|^3$   $\leftarrow$   $\omega$ -convex on measures with fixed center of mass and  $\omega(x) = x^{3/2}$

# collective dynamics: main questions

Aggregation equation with degenerate diffusion:

$$\frac{d}{dt}\rho + \nabla \cdot ((-\nabla K * \rho)\rho) = \Delta \rho^m \quad \text{for } K(x) : \mathbb{R}^d \rightarrow \mathbb{R} \text{ and } m \geq 1$$

Main questions:

1. Do solutions exist?
2. Are they unique? stable?
3. How do they behave in the long time limit? depends on choice of  $K(x)$
4. How can we simulate them numerically?

# long time behavior: $K = \Delta^{-1}$

For  $K = \Delta^{-1}$  and  $1 \leq m \leq +\infty$ , long time behavior of Keller-Segel equation has been the subject of recent interest.

- **Supercritical power** ( $m \leq 2-2/d$ ):

Profiles of steady states known for certain of  $m$ ; solutions can “blow up” to a Dirac mass in finite time or remain bounded.

[Sugiyama 2006, 2007], [Luckhaus and Sugiyama 2006, 2007],  
[Blanchet, Carlen, Carrillo 2012], [Chen, Liu, Wang 2012]

- **Subcritical power** ( $m > 2-2/d$ ):

All steady states are radially symmetric and decreasing; still, convergence to equilibrium is only known in  $d=1, 2$  and for radial solutions in higher dimensions.

[Carrillo, Hitter, Volzone, Yao 2016], [Kim, Yao 2012]

# long time behavior: $K = \Delta^{-1}$ , $m = +\infty$

In the case of the **height constrained aggregation equation**, we obtained quantitative rates of convergence to equilibrium for patch solutions:

**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

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

- In **any dimension**, the Riesz Rearrangement Inequality guarantees that the unique minimizer of  $E_\infty$  is  $1_B(x)$ .
- The tricky part is showing mass of  $\rho(x,t)$  doesn't escape to  $+\infty$ . To do this, we characterize the dynamics of patch solutions in terms of a free boundary problem and control  $M_2(\rho(t))$  by Talenti inequality ( $d=2$ ).

# collective dynamics: main questions

Aggregation equation with degenerate diffusion:

$$\frac{d}{dt}\rho + \nabla \cdot ((-\nabla K * \rho)\rho) = \Delta \rho^m \quad \text{for } K(x) : \mathbb{R}^d \rightarrow \mathbb{R} \text{ and } m \geq 1$$

Main questions:

1. Do solutions exist?
2. Are they unique? stable?
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4. How can we simulate them numerically?

# numerics

- For nice velocity fields and  $\rho = \frac{1}{N} \sum_{i=1}^N \delta_{x_i(t)}$ ,

$$\frac{d}{dt} \rho(x, t) + \nabla \cdot (v(x, t) \rho(x, t)) = 0$$



$$\frac{d}{dt} x_i(t) = v(x_i(t), t), \quad \forall i = 1, \dots, N$$

- For any  $\rho(x)$ , there exist  $x_1, \dots, x_N$  so that  $W_2 \left( \rho, \frac{1}{N} \sum_{i=1}^N \delta_{x_i} \right) \xrightarrow{N \rightarrow +\infty} 0$

General Numerical Strategy: to approximate a solution  $\rho(x, t)$  of a PDE...

- 1) Approximate  $\rho(x, 0)$  by  $\rho_N(x, 0) = \frac{1}{N} \sum_{i=1}^N \delta_{x_i}$ .
- 2) Compute the solution with initial data  $\rho_N$  by numerically solving the corresponding system of ODEs.
- 3) Use stability of PDE to conclude that the numerical solution  $\rho_N(x, t)$  must be close to  $\rho(x, t)$  on bounded time intervals.

What about when  $v(x, t)$  is not “nice”?



# numerics

None of the  $v(x,t)$  mentioned so far are nice! We need to make them nice.

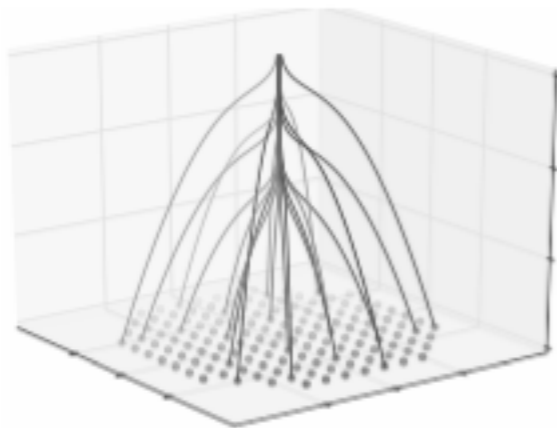
Aggregation equation without diffusion:

- Regularize  $K$  by convolution with a mollifier (“blob”)

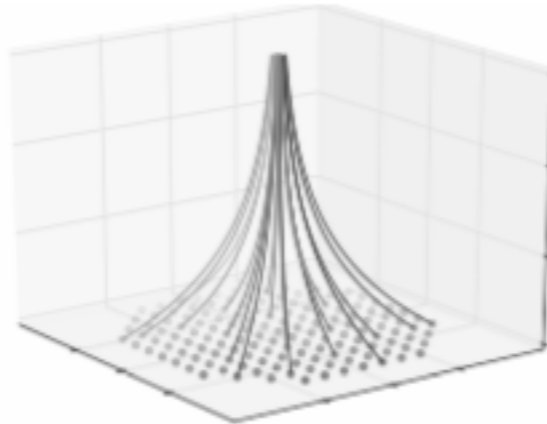
- **Theorem** [C., Bertozzi 2014]: If you remove the mollification as you add particles, the particle “blob” method converges.

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

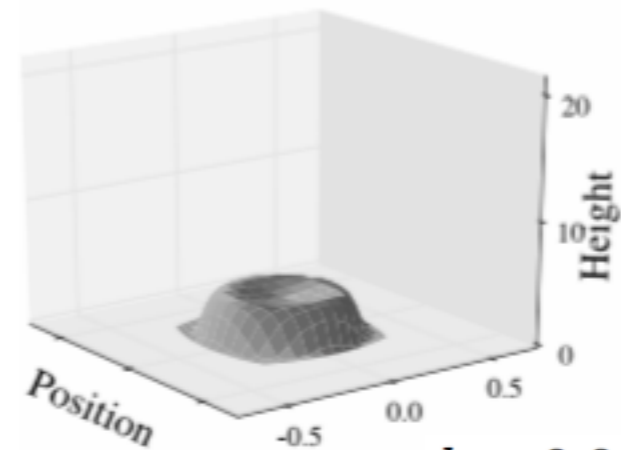
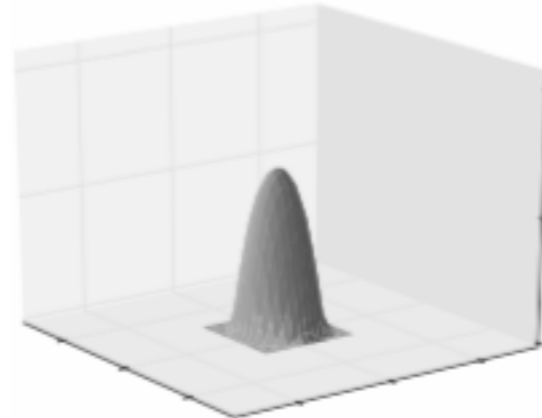
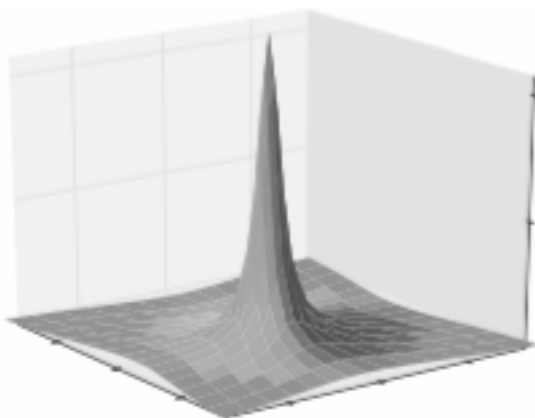
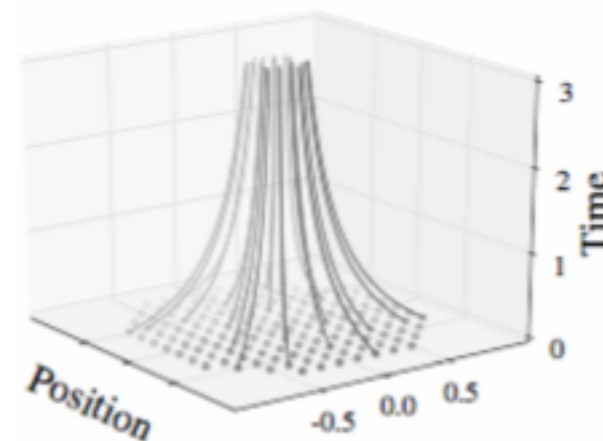
$$K(x) = \log |x|/2\pi$$



$$K(x) = |x|^2/2$$



$$K(x) = |x|^3/3$$



# numerics

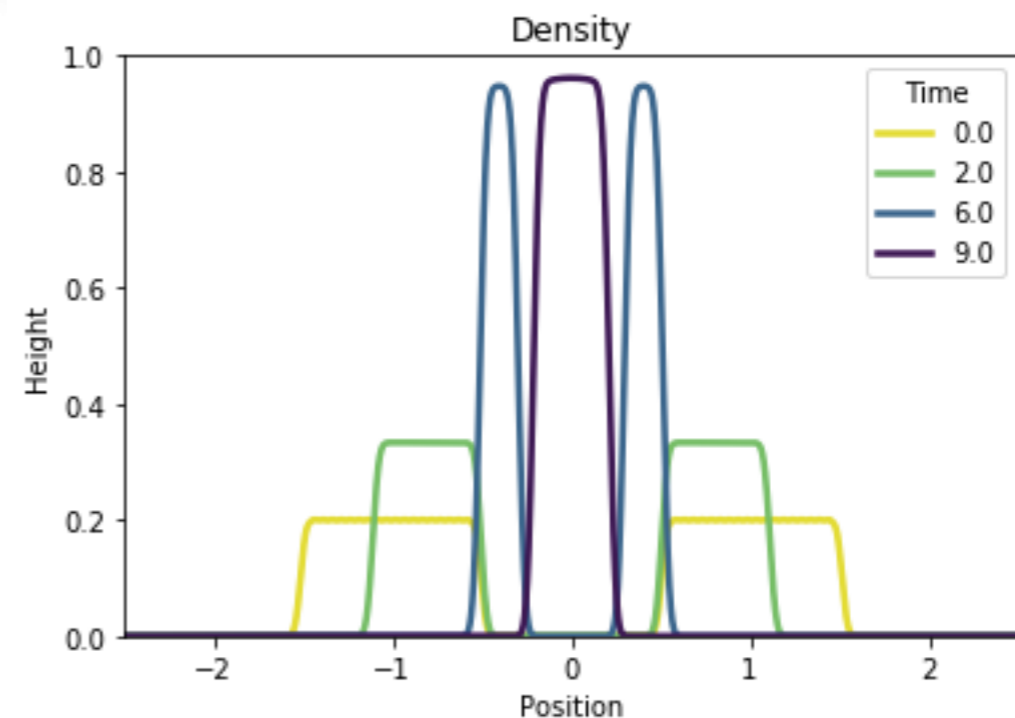
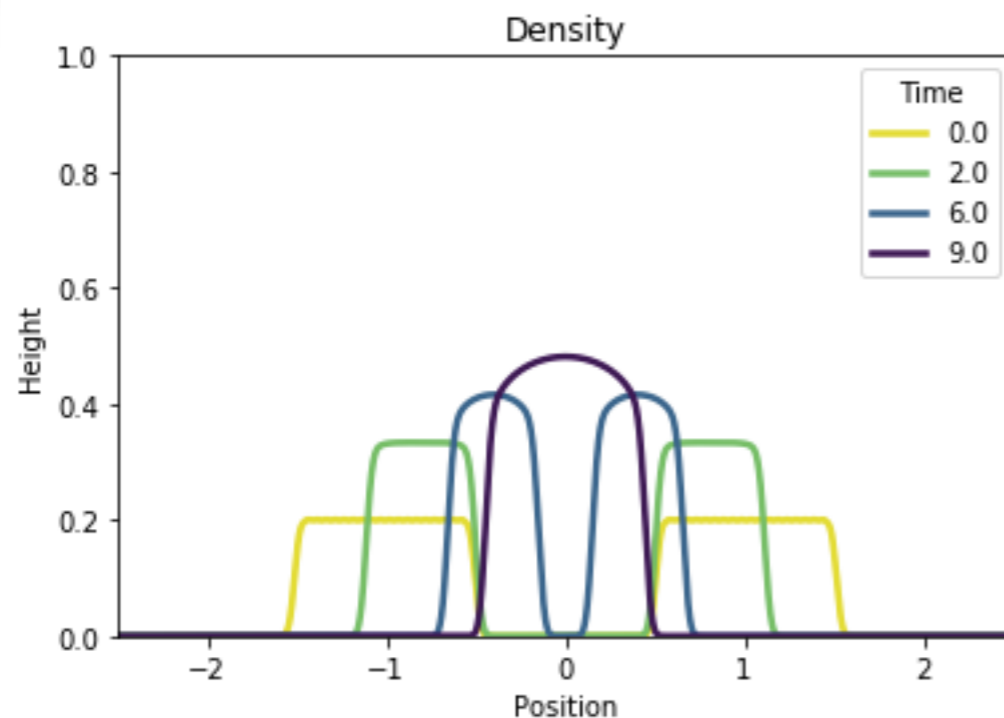
Aggregation equation w/ deg. diffusion:

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

- Regularize both  $K$  and  $v$  by convolution

$$\Delta\rho^m = \nabla \cdot (m\rho^{m-1}\nabla\rho) = \nabla \cdot (\underbrace{(m\rho^{m-2}\nabla\rho)}_v)\rho$$

- Theorem** [Carrillo, C., Patacchini (in progress)]: If you remove the mollification as you add particles, the particle “blob” method  $\Gamma$ -converges.



Newtonian attraction ( $K = \Delta^{-1}$ ) and  $m=2$  and  $m=100$  diffusion

# future work:

Does Keller-Segel converge to congested aggregation?

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

$m \rightarrow +\infty$

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

- For  $V(x)$  convex, [Alexander, Kim, Yao 2014] showed

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

$m \rightarrow +\infty$

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

- Connecting Keller-Segel and the congested aggregation eqn would lead to greater insight in long-time behavior of supercritical ( $m > 2 - 2/d$ ) Keller-Segel.

Further examples of  $\omega$ -convex energies?

More applications with a height constraint?

Thank you!

# motivation for free boundary problem

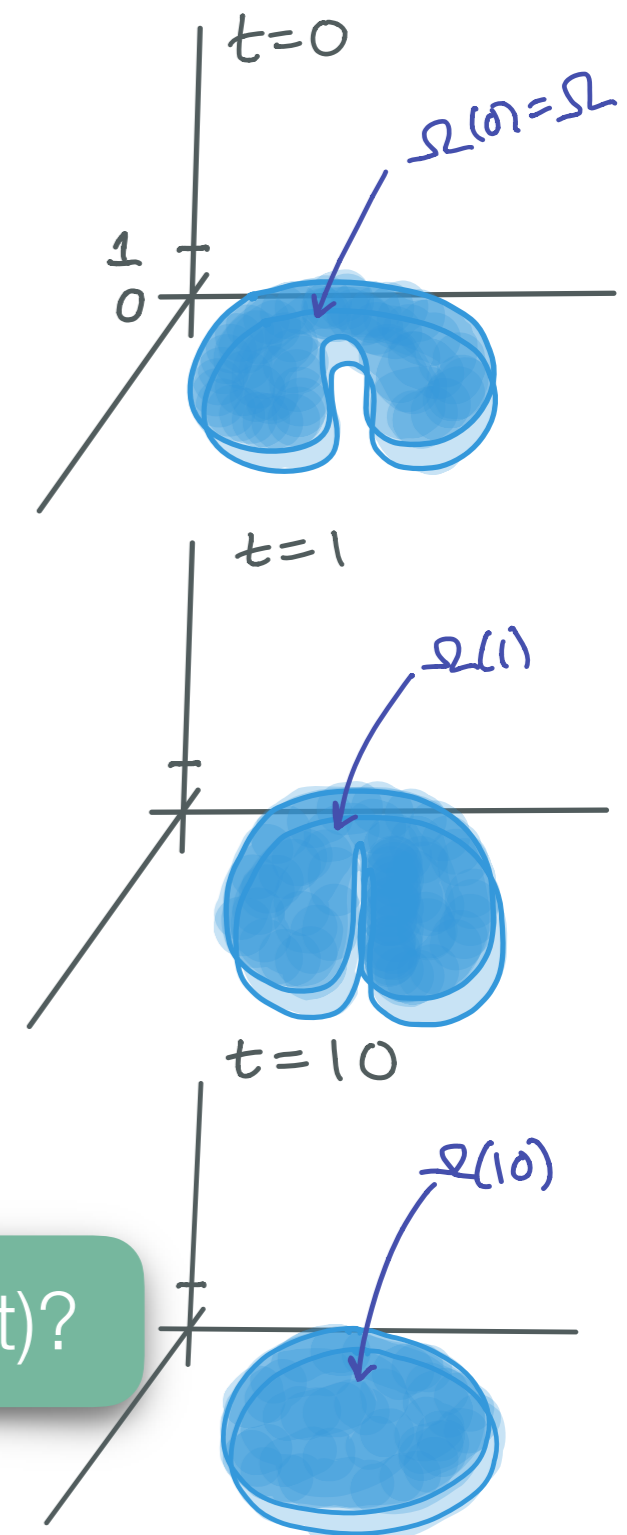
How does congested aggregation equation relate to free boundary problem?

“

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

”

- Consider **patch solutions**. For a domain  $\Omega$ , 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

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

- Using that  $\rho(x,t) = 1_{\Omega(t)}(x)$ , 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,

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

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

**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

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

# collective dynamics

- $\rho(x, t) : \mathbb{R}^d \times \mathbb{R} \rightarrow [0, +\infty)$  nonnegative density
- Mass is conserved (assume  $\int \rho(x) dx = 1$ ), and  $\rho(x, t)$  evolves according to a continuity equation:

$$\frac{d}{dt} \rho(x, t) + \nabla \cdot (v(x, t) \rho(x, t)) = 0$$

- Particle approximation:
  - Suppose  $\rho = \frac{1}{N} \sum_{i=1}^N \delta_{x_i(t)}$
  - For “nice” velocity fields,  $\rho(x, t)$  solves the continuity equation iff

$$\frac{d}{dt} x_i(t) = v(x_i(t), t), \quad \forall i = 1, \dots, N$$

# collective dynamics: slime mold

In the case of the slime mold, we have 1) self-attraction and 2) diffusion.

## 1) Self-Attraction

- At the particle level, we may formulate self-attraction as

$$\frac{d}{dt}x_i(t) = -\frac{1}{N} \sum_{j=1}^N \nabla K(x_i(t) - x_j(t))$$

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

- Since  $\rho = \frac{1}{N} \sum_{i=1}^N \delta_{x_i(t)}$ , we write the resulting velocity field as

$$-\frac{1}{N} \sum_{j=1}^N \nabla K(x - x_j(t)) = -\int \nabla K(x - y) d\rho(y) = -\nabla K * \rho(x)$$

# collective dynamics: slime mold

In the case of the slime mold, we have 1) self-attraction and 2) diffusion.

## 2) Diffusion

- Combining self-attraction with diffusion gives the Keller-Segel equation

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

- More generally, we can consider degenerate diffusion for  $m \geq 1$

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

$$\Delta\rho^m = \nabla \cdot \underbrace{(m\rho^{m-1})}_D \nabla\rho$$