

# Deterministic particle approximations for transport models with nonlinear mobility

**Marco Di Francesco**

University of L'Aquila

Work in collaboration with Simone Fagioli and Emanuela Radici (University of L'Aquila)

Entropies, the Geometry of Nonlinear Flows, and their Applications -  
Banff, April 9-13 2018

# Table of contents

- 1 Motivation
- 2 The follow-the-leader model
- 3 Particle approximation of the nonlocal model

# Table of contents

- 1 Motivation
- 2 The follow-the-leader model
- 3 Particle approximation of the nonlocal model

# Transport/Diffusion models with nonlinear mobility

**Continuity equations with local/nonlocal transport and/or diffusion:**

$$\partial_t \rho + \operatorname{div}(\rho V[\rho, \nabla \rho]) = 0, \quad \rho = \rho(t, x), \quad x \in \mathbb{R}^d, \quad t \geq 0.$$

# Transport/Diffusion models with nonlinear mobility

**Continuity equations with local/nonlocal transport and/or diffusion:**

$$\partial_t \rho + \operatorname{div}(\rho V[\rho, \nabla \rho]) = 0, \quad \rho = \rho(t, x), \quad x \in \mathbb{R}^d, \quad t \geq 0.$$

If  $V[\rho, \nabla \rho]$  is a *gradient*, we have a **gradient flow**:

$$\partial_t \rho + \operatorname{div}(\rho \nabla f[\rho]) = 0.$$

# Transport/Diffusion models with nonlinear mobility

**Continuity equations with local/nonlocal transport and/or diffusion:**

$$\partial_t \rho + \operatorname{div}(\rho V[\rho, \nabla \rho]) = 0, \quad \rho = \rho(t, x), \quad x \in \mathbb{R}^d, \quad t \geq 0.$$

If  $V[\rho, \nabla \rho]$  is a *gradient*, we have a **gradient flow**:

$$\partial_t \rho + \operatorname{div}(\rho \nabla f[\rho]) = 0.$$

In relevant examples, the *speed*  $|V[\rho, \nabla \rho]|$  is *cut-off at high densities* no matter what the direction is: **prevention of overcrowding**:

$$\partial_t \rho + \operatorname{div}(\rho v(\rho) \nabla g[\rho]) = 0.$$

# Transport/Diffusion models with nonlinear mobility

**Continuity equations with local/nonlocal transport and/or diffusion:**

$$\partial_t \rho + \operatorname{div}(\rho V[\rho, \nabla \rho]) = 0, \quad \rho = \rho(t, x), \quad x \in \mathbb{R}^d, \quad t \geq 0.$$

If  $V[\rho, \nabla \rho]$  is a *gradient*, we have a **gradient flow**:

$$\partial_t \rho + \operatorname{div}(\rho \nabla f[\rho]) = 0.$$

In relevant examples, the *speed*  $|V[\rho, \nabla \rho]|$  is *cut-off at high densities* no matter what the direction is: **prevention of overcrowding**:

$$\partial_t \rho + \operatorname{div}(\rho v(\rho) \nabla g[\rho]) = 0.$$

Typical law for  $v(\rho)$

Let  $\rho_{\max} > 0$  be a maximal density. Then,

$$v : [0, \rho_{\max}] \rightarrow [0, v_{\max}], \quad v \text{ decreasing, and } v(\rho_{\max}) = 0.$$

# Examples in biology

## Aggregation/Swarming phenomena

Avoiding blow-up of the density when aggregation phenomena dominate, with the goal of detecting *pattern-formation* for large times rather than concentration to Dirac masses:

$$\partial_t \rho + \operatorname{div}(\rho v(\rho) \nabla(a(\rho) + W * \rho)) = 0,$$

with  $a(\rho)$  a nonlinear diffusion function and  $W = W(x)$  an aggregative kernel.



# Examples in biology

## Aggregation/Swarming phenomena

Avoiding blow-up of the density when aggregation phenomena dominate, with the goal of detecting *pattern-formation* for large times rather than concentration to Dirac masses:

$$\partial_t \rho + \operatorname{div}(\rho v(\rho) \nabla(a(\rho) + W * \rho)) = 0,$$

with  $a(\rho)$  a nonlinear diffusion function and  $W = W(x)$  an aggregative kernel.

## Modified chemotaxis modelling

To prevent concentration in the Keller-Segel system:

$$\begin{aligned} \rho_t &= D_\rho \Delta \rho - \chi \operatorname{div}(\rho(\rho_{\max} - \rho) \nabla c) \\ \varepsilon c_t &= D_c \Delta c + \alpha \rho - \beta c. \end{aligned}$$

## Examples in real-world applications

Traffic flow: extended LRW equation

$d = 1$ , vehicles moving in the same direction, external potential  $V = V(x)$  describing possible heterogeneities on the road:

$$\partial_t \rho + \partial_x (\rho v(\rho) V'(x)) = 0.$$

## Examples in real-world applications

### Traffic flow: extended LRW equation

$d = 1$ , vehicles moving in the same direction, external potential  $V = V(x)$  describing possible heterogeneities on the road:

$$\partial_t \rho + \partial_x(\rho v(\rho) V'(x)) = 0.$$

### Pedestrian motion

$d = 1, 2$ , pedestrians moving with speed  $v(\rho)$  and direction  $\nabla \varphi / |\nabla \varphi|$ , where  $\varphi$  is determined nonlocally from the overall density. Examples:

- (Hughes)  $|\nabla \varphi| = c(\rho)$ , with cost function  $c(\rho)$  increasing with the density.

- (Colombo et al.)  $\rho_t + \operatorname{div} \left( \rho v(\rho) \left( v(x) - \varepsilon \frac{\nabla \eta * \rho}{\sqrt{1 + |\nabla \eta * \rho|^2}} \right) \right) = 0.$

## Examples in real-world applications

### Traffic flow: extended LRW equation

$d = 1$ , vehicles moving in the same direction, external potential  $V = V(x)$  describing possible heterogeneities on the road:

$$\partial_t \rho + \partial_x (\rho v(\rho) V'(x)) = 0.$$

### Pedestrian motion

$d = 1, 2$ , pedestrians moving with speed  $v(\rho)$  and direction  $\nabla \varphi / |\nabla \varphi|$ , where  $\varphi$  is determined nonlocally from the overall density. Examples:

- (Hughes)  $|\nabla \varphi| = c(\rho)$ , with cost function  $c(\rho)$  increasing with the density.
- (Colombo et al.)  $\rho_t + \operatorname{div} \left( \rho v(\rho) (v(x) - \varepsilon \frac{\nabla \eta * \rho}{\sqrt{1 + |\nabla \eta * \rho|^2}}) \right) = 0.$

Other applications include:

- Phase segregation with long range interactions
- Simplified models for Fermi-Dirac condensates

Comparison with other approaches preventing high densities (Katy Craig's talk).

## Microscopic vs. macroscopic description

- All aforementioned applications are best formulated in a (discrete) microscopic setting, as they describe *individual-based phenomena*.

## Microscopic vs. macroscopic description

- All aforementioned applications are best formulated in a (discrete) microscopic setting, as they describe *individual-based phenomena*.
- The (continuum) macroscopic description has two main advantages:

## Microscopic vs. macroscopic description

- All aforementioned applications are best formulated in a (discrete) microscopic setting, as they describe *individual-based phenomena*.
- The (continuum) macroscopic description has two main advantages:
  - Broad range of numerical methods available.

# Microscopic vs. macroscopic description

- All aforementioned applications are best formulated in a (discrete) microscopic setting, as they describe *individual-based phenomena*.
- The (continuum) macroscopic description has two main advantages:
  - Broad range of numerical methods available.
  - Detecting large time behaviour (patterns, diffusive decay, concentrations, etc.).



# Microscopic vs. macroscopic description

- All aforementioned applications are best formulated in a (discrete) microscopic setting, as they describe *individual-based phenomena*.
- The (continuum) macroscopic description has two main advantages:
  - Broad range of numerical methods available.
  - Detecting large time behaviour (patterns, diffusive decay, concentrations, etc.).
- The continuum description needs to be rigorously validated, via *many particle limits*.

# Microscopic vs. macroscopic description

- All aforementioned applications are best formulated in a (discrete) microscopic setting, as they describe *individual-based phenomena*.
- The (continuum) macroscopic description has two main advantages:
  - Broad range of numerical methods available.
  - Detecting large time behaviour (patterns, diffusive decay, concentrations, etc.).
- The continuum description needs to be rigorously validated, via *many particle limits*.
- First step in the understanding of those limits: no diffusion. Deterministic particle limits.

# Microscopic vs. macroscopic description

- All aforementioned applications are best formulated in a (discrete) microscopic setting, as they describe *individual-based phenomena*.
- The (continuum) macroscopic description has two main advantages:
  - Broad range of numerical methods available.
  - Detecting large time behaviour (patterns, diffusive decay, concentrations, etc.).
- The continuum description needs to be rigorously validated, via *many particle limits*.
- First step in the understanding of those limits: no diffusion. Deterministic particle limits.

## Main difficulty

The velocity field is typically not continuous w.r.t. tight convergence, because of the local dependency on  $\rho$ .

## Related literature

Case  $\nu$  constant (linear mobility):

- Nonlocal interaction equations with regular kernels: Dobrushin (1979)
- Nonlocal interaction equations (with singular kernels): Carrillo et al.
- Deterministic diffusion (linear case): Russo (1990)
- Nonlinear diffusion: Gosse-Toscani (2006)
- Multidimensional deterministic diffusion via gamma convergence: Carrillo, Craig, Patacchini...
- Diffusion and nonlocal interactions: Matthes et al.

Case  $\nu$  non-constant (nonlinear mobility):

- General setting with velocity field continuous w.r.t. measure topology: Piccoli-Rossi 2013

# Table of contents

- 1 Motivation
- 2 The follow-the-leader model
- 3 Particle approximation of the nonlocal model

# A discrete lagrangian description

General continuum model:

$$\rho_t + \operatorname{div}(\underbrace{\rho v(\rho) \nabla(V + W * \rho)}_{\text{velocity field}}) = 0.$$

# A discrete lagrangian description

General continuum model:

$$\rho_t + \operatorname{div}(\underbrace{\rho v(\rho)}_{\text{velocity field}} \nabla(V + W * \rho)) = 0.$$

Discrete counterpart with  $N$  particles  $x_1, \dots, x_N$ :

$$\dot{x}_i(t) = v(R_i(t)) \left( \nabla V(x_i(t)) + (\nabla W * \tilde{\rho}^N(x_i(t))) \right),$$

with

- $R_i$  a suitable reconstruction of the density around the particle  $x_i$ ,
- $\tilde{\rho}^N$  a suitable replacement for the global discrete density.

# A discrete lagrangian description

General continuum model:

$$\rho_t + \operatorname{div}(\underbrace{\rho v(\rho)}_{\text{velocity field}} \nabla(V + W * \rho)) = 0.$$

Discrete counterpart with  $N$  particles  $x_1, \dots, x_N$ :

$$\dot{x}_i(t) = v(R_i(t)) \left( \nabla V(x_i(t)) + (\nabla W * \tilde{\rho}^N(x_i(t))) \right),$$

with

- $R_i$  a suitable reconstruction of the density around the particle  $x_i$ ,
- $\tilde{\rho}^N$  a suitable replacement for the global discrete density.

Fact 1: with  $d = 1$  the reconstruction of the density is much easier, and one-dimensional cases are relevant to some of our target applications (e.g. traffic flow).



# A discrete lagrangian description

General continuum model:

$$\rho_t + \operatorname{div}(\underbrace{\rho v(\rho)}_{\text{velocity field}} \nabla(V + W * \rho)) = 0.$$

Discrete counterpart with  $N$  particles  $x_1, \dots, x_N$ :

$$\dot{x}_i(t) = v(R_i(t)) \left( \nabla V(x_i(t)) + (\nabla W * \tilde{\rho}^N(x_i(t))) \right),$$

with

- $R_i$  a suitable reconstruction of the density around the particle  $x_i$ ,
- $\tilde{\rho}^N$  a suitable replacement for the global discrete density.

Fact 1: with  $d = 1$  the reconstruction of the density is much easier, and one-dimensional cases are relevant to some of our target applications (e.g. traffic flow).

Fact 2: the nonlocal part can be discretized as

$$\tilde{\rho}^N(t) = \frac{M}{N} \sum_i \delta_{x_i(t)}, \quad \nabla W * \tilde{\rho}^N(x_i(t)) = \frac{M}{N} \sum_i \nabla W(x_i(t) - x_j(t)),$$

(where  $M$  is the total mass).

# Scalar conservation laws: the result<sup>1</sup>

The unique entropy solution to the  $d = 1$  conservation law

$$\rho_t + (\rho v(\rho))_x = 0,$$

with given  $\rho(t = 0) \in L^\infty$ , compactly supported, and with mass  $M$ , is approximated in  $L^1$  (strongly) as  $N \rightarrow +\infty$  by the discrete density

$$\rho^N(t, x) = \sum_{i=0}^{N-1} R_i(t) \chi_{[x_i(t), x_{i+1}(t))}, \quad R_i(t) = \frac{\ell_N}{x_{i+1}(t) - x_i(t)}, \quad \ell_N = M/N,$$

where  $x_i$ ,  $i = 1, \dots, N$  solve

$$\begin{aligned} \dot{x}_N(t) &= v(0) \\ \dot{x}_i(t) &= v(R_i(t)), \quad i = 0, \dots, N-1, \end{aligned}$$

with initial condition  $\bar{x}_0, \dots, \bar{x}_N$  such that

$$\int_{\bar{x}_i}^{\bar{x}_{i+1}} \rho(t=0, x) dx = \ell_N.$$

<sup>1</sup>DF-Rosini, ARMA 2015

# Scalar conservation laws: remarks

- Result holds in presence of *vacuum*.

## Scalar conservation laws: remarks

- Result holds in presence of *vacuum*.
- Essential for the result is a control of the  $BV$  norm of  $\rho^N$ .

## Scalar conservation laws: remarks

- Result holds in presence of *vacuum*.
- Essential for the result is a control of the  $BV$  norm of  $\rho^N$ .
- If we assume  $\rho \mapsto \rho v'(\rho)$  non-increasing, then the result does not require the initial norm to be finite ( $L^\infty$ - $BV$  smoothing effect).

## Scalar conservation laws: remarks

- Result holds in presence of *vacuum*.
- Essential for the result is a control of the  $BV$  norm of  $\rho^N$ .
- If we assume  $\rho \mapsto \rho v'(\rho)$  non-increasing, then the result does not require the initial norm to be finite ( $L^\infty$ - $BV$  smoothing effect).
- Otherwise, we need  $\rho(t=0) \in BV(\mathbb{R})$ . The  $BV$  norm can be proven to be *non-increasing*.

## Scalar conservation laws: remarks

- Result holds in presence of *vacuum*.
- Essential for the result is a control of the  $BV$  norm of  $\rho^N$ .
- If we assume  $\rho \mapsto \rho v'(\rho)$  non-increasing, then the result does not require the initial norm to be finite ( $L^\infty$ - $BV$  smoothing effect).
- Otherwise, we need  $\rho(t=0) \in BV(\mathbb{R})$ . The  $BV$  norm can be proven to be *non-increasing*.
- The scheme leads to the unique *entropy solution*. Surprising, since the scheme does not display the shock structure of a conservation law (kinetic velocity  $\neq$  characteristic velocity).

## Scalar conservation laws: remarks

- Result holds in presence of *vacuum*.
- Essential for the result is a control of the  $BV$  norm of  $\rho^N$ .
- If we assume  $\rho \mapsto \rho v'(\rho)$  non-increasing, then the result does not require the initial norm to be finite ( $L^\infty$ - $BV$  smoothing effect).
- Otherwise, we need  $\rho(t=0) \in BV(\mathbb{R})$ . The  $BV$  norm can be proven to be *non-increasing*.
- The scheme leads to the unique *entropy solution*. Surprising, since the scheme does not display the shock structure of a conservation law (kinetic velocity  $\neq$  characteristic velocity).
- Approach can be extended to
  - Dirichlet boundary conditions,
  - Second order traffic models,
  - (Small  $BV$ -norm solutions to the) Hughes model for pedestrians.



# Table of contents

- 1 Motivation
- 2 The follow-the-leader model
- 3 Particle approximation of the nonlocal model

# The continuum nonlocal model

We study<sup>2</sup>

$$\begin{aligned} \partial_t \rho - \partial_x(\rho v(\rho) K' * \rho) &= 0 \\ \rho(t=0) &= \bar{\rho} \in L_c^\infty(\mathbb{R}; [0, \rho_{\max}]) \cap BV(\mathbb{R}) \end{aligned} \tag{1a}$$

Assumptions on  $K$

$K \in C^2(\mathbb{R})$ ,  $K(-x) = K(x)$ ,  $K' > 0$  on  $(0, +\infty)$ ,  $K'' \in \text{Lip}_{loc}(\mathbb{R})$ .

Assumptions on  $v$

$v \in C^1([0, +\infty))$ ,  $v$  decreasing on  $[0, \rho_{\max}]$ ,  $v \equiv 0$  on  $[\rho_{\max}, +\infty)$ .

Same atomization algorithm as before  $\Rightarrow$  Initial positions of  $N + 1$  particles

$\bar{x}_0 < \dots < \bar{x}_N$ .

<sup>2</sup>DF, Fagioli, Radici - submitted

# The discrete model

$$\dot{x}_i(t) = -\frac{1}{N} \underbrace{v(R_i(t))}_{\text{forward density}} \sum_{j>i} K'(x_i(t) - x_j(t)) - \frac{1}{N} \underbrace{v(R_{i-1}(t))}_{\text{backward density}} \sum_{j<i} K'(x_i(t) - x_j(t))$$

$$\text{Discrete density: } R_i(t) = \frac{1}{N(x_{i+1}(t) - x_i(t))}$$

## Properties:

- Discrete maximum principle  $x_{i+1}(t) - x_i(t) \geq \frac{M}{\rho_{\max} N}$ , i.e.  $R_i(t) \leq \rho_{\max}$
- Global existence
- $x_0(t) \geq \bar{x}_0$ ,  $x_N(t) \leq \bar{x}_N$  (confined support)

# Estimates

$$\rho^N(t, x) = \sum_{i=0}^{N-1} R_i(t) \chi_{[x_i(t), x_{i+1}(t))}$$

$$TV[\rho^N(\cdot, t)] = R_0(t) + \sum_{i=0}^{N-2} |R_{i+1}(t) - R_i(t)| + R_{N-1}(t)$$

## Uniform BV estimate

There exists a constant  $C > 0$  depending only on  $K$ ,  $v$ , and  $\text{meas}(\text{supp}[\bar{\rho}])$ , such that

$$TV[\rho^N(\cdot, t)] \leq TV[\bar{\rho}]e^{Ct}, \quad \text{for all } t \geq 0.$$

- The proof crucially uses the monotonicity of  $v$  and the splitting of the use of 'upwind' densities in the velocity field.
- The overcrowding prevention effect is also crucial: without it, particles would collapse into a single point mass.

# Convergence of the scheme

- The previous  $BV$ -estimate allows to control space-oscillations.
- As for time-oscillations, we prove Lipschitz equi-continuity in the 1-Wasserstein distance

$$W_1(\rho^N(t, \cdot), \rho^N(s, \cdot)) \leq C|t - s|,$$

with  $C > 0$  independent of  $N$ .

- Rossi-Savaré 2003 (Aubin/Lions-type compactness theorem) implies strong compactness of  $\rho^N$  in  $L^1([0, +\infty) \times \mathbb{R})$ .

# Entropy solutions

Similarly to scalar conservation laws, we define

## Definition

$\rho : [0, +\infty) \times \mathbb{R} \rightarrow [0, +\infty)$  is an entropy solution to (1a) with initial condition  $\bar{\rho}$  if  $\rho \in L^\infty([0, +\infty); L^1 \cap L^\infty(\mathbb{R}))$  and, for all constants  $c \geq 0$  and for all  $\varphi \in C_c([0, +\infty) \times \mathbb{R})$  with  $\varphi \geq 0$  one has

$$\int_{\mathbb{R}} |\bar{\rho}(x) - c| \varphi(0, x) dx + \int_0^{+\infty} \int_{\mathbb{R}} (|\rho - c| \varphi_t - \text{sign}(\rho - c) [(f(\rho) - f(c))K' * \rho \varphi_x - f(c)K'' * \rho \varphi]) dx dt \geq 0,$$

where  $f(z) = zv(z)$ .

Notice that entropy solutions are weak solutions.

## Theorem

- There exists no more than one entropy solution to (1a) with initial condition  $\bar{\rho}$
- $\rho^N \rightarrow \rho$  as  $N \rightarrow +\infty$  and  $\rho$  is an entropy solution (proof: very technical).

# Non uniqueness of weak solutions

Consider  $v(\rho) = (1 - \rho)_+$  and the initial condition

$$\bar{\rho}(x) = \chi_{[-1, -1/2]} + \chi_{[1/2, 1]}.$$

Let  $\rho_s(t, x) = \bar{\rho}(x)$  for all  $t \geq 0$ .

- $\rho_s$  is a (stationary) weak solution to (1a)
- $\rho_s$  is *not* an entropy solution. Proof: use test functions that concentrate around  $-1/2$  and  $1/2$  to violate the entropy condition. Extra assumption needed:  $K'' > 0$  on the support of  $\bar{\rho}$ .
- We know the scheme converges to an entropy solution, therefore there are *at least two weak solutions* with this initial condition.
- Why is  $\rho_s$  not satisfying the entropy condition: the discontinuities at  $\pm 1/2$  are *not admissible*.
- The scheme catches this behavior because particles at  $\pm 1/2$  are forced to *move*.

# Simulations



## Extension to the diffusive case<sup>3</sup>

$$\partial_t \rho = \partial_{xx} \varphi(\rho) + \partial_x (\rho v(\rho) K' * \rho)$$

- On bounded intervals with no-slip boundary conditions
- Initial datum away from the vacuum state
- Convergence to weak solutions via  $BV$  estimate
- Diffusion term may be degenerate,  $\varphi$  is required to be *non-decreasing*
- Assumption on  $K$  are the same as before.

---

<sup>3</sup>Fagioli, Radici - to appear on M3AS

# End of the talk

Thanks for your attention!