Rigorous derivation of nonlinear scalar conservation laws from follow-the-leader type models via many particle limit

## Massimiliano D. Rosini

ICM, Uniwersytet Warszawski

Work in collaboration with Marco Di Francesco, Università degli Studi dell'Aquila

Micro and Macro Systems in Life Sciences Będlewo, 8-12 June 2015

## Table of contents

### 1

The motivating problem: traffic flow

- Macroscopic models
- Follow the leader approximation

### 2

Convergence result

- A preliminary lemma
- Weak convergence
- Uniform estimates
- Time continuity
- Convergence to entropy solutions

#### 3 Remarks and future projects

## Table of contents.



#### The motivating problem: traffic flow

- Macroscopic models
- Follow the leader approximation
- 2
- Convergence result
- A preliminary lemma
- Weak convergence
- Uniform estimates
- Time continuity
- Convergence to entropy solutions
- 3 Remarks and future projects

## Modelling traffic flow.

#### Practical goals.

- Rational planning and management of vehicle fluxes.
- Reduce environmental pollution and cities congestion.

#### Mathematical modelling.

- Microscopic approach. Each car is a 'moving particle' satisfying an ODE.
- Macroscopic approach. Averaged quantities satisfying PDEs.

#### Advantages of the macroscopic approach.

- Very powerful description of queues tails in terms of shocks.
- Suitable with very large number of vehicles.
- Easy to validate and implement (low number of parameters).
- Suitable to *real time* prediction, estimation, optimization and control.

## Modelling traffic flow.

#### Practical goals.

- Rational planning and management of vehicle fluxes.
- Reduce environmental pollution and cities congestion.

#### Mathematical modelling.

- Microscopic approach. Each car is a 'moving particle' satisfying an ODE.
- Macroscopic approach. Averaged quantities satisfying PDEs.

### Continuum hypothesis is not satisfied!

The number of cars is far lower than that of molecules, for example, in gas dynamics (1 mole of gas contains  $6 \times 10^{23}$  molecules), the continuum assumption is not justified and the macroscopic formulation is not a priori justified!

#### Link between macroscopic and microscopic approach.

It provides a validation of the macroscopic approach and of the use of data collection from GPS devises when the number of detected 'reference cars' is very large.

Rosini (ICM, UW)

## Table of contents.



#### The motivating problem: traffic flow Macroscopic models

Follow the leader approximation

2

- Convergence result
- A preliminary lemma
- Weak convergence
- Uniform estimates
- Time continuity
- Convergence to entropy solutions

3 Remarks and future projects

The macroscopic variables are:



The macroscopic variables satisfy:

by definition

 $f = \rho \mathbf{V}$ 

by conservation of # cars  $\rho_t + f_x = 0$ 

# We have 3 variables and 2 equations!

### Two macroscopic approaches.

 First order models close the system by giving an explicit expression of 1 of the 3 unknowns in terms of the remaining 2 (equation of state).
 Example: Lighthill-Whitham-Richards (1955, 1956).

$$\rho_t + [\rho \mathbf{v}(\rho)]_x = \mathbf{0}.$$

- Equilibrium models.
- Velocity as function of the density.
- Second order models close the system by adding a further PDE. Example: Aw-Rascle-Zhang (1999, 2002).

 $\rho_t + (\rho \mathbf{v})_x = \mathbf{0}, \qquad [\mathbf{v} + \mathbf{p}(\rho)]_t + \mathbf{v}[\mathbf{v} + \mathbf{p}(\rho)]_x = \mathbf{0}.$ 

- Continuum analogue of Newton's law.
- Velocity evolves via a separate PDE.

Justification of the macroscopic models.

- A posteriori. Descriptive power (no physical laws).
- A priori. Validation via microscopic models (no continuum hypothesis).

$$\rho_t + [\rho \mathbf{v}(\rho)]_x = \mathbf{0}, \qquad \rho(t=\mathbf{0}) = \bar{\rho}$$

#### Relevant parameters.

- Maximum density of vehicles  $\rho_{max} > 0$ . We normalize  $\rho_{max} = 1$ .
- Maximum possible speed v<sub>max</sub> > 0.
- Total length of the vehicles (constant in time)

$$L=\int_{\mathbb{R}}\rho(t,x)\mathrm{d}x>0.$$

### Main assumptions on the velocity map v.

- $v \in \mathbf{C}^1([0, 1]; [0, v_{\max}]).$
- v strictly decreasing on [0, 1].

• 
$$v(0) = v_{\max}, v(1) = 0.$$

### Initial condition.

•  $\rho(t = 0) = \overline{\rho} \in L^{\infty}(\mathbb{R}), \ \overline{\rho} \ge 0, \ \overline{\rho} \text{ with compact support.}$ 

#### Greenshields 1935:

$$v(\rho) = v_{\max}(1-\rho)$$



Greenberg 1959, 'renormalized' version

$$v(\rho) = v_{\max} \left[ \log \left( \frac{1+\alpha}{\alpha} \right) \right]^{-1} \log \left( \frac{1+\alpha}{\rho+\alpha} \right) \qquad \alpha > 0$$

$$v = \int_{1}^{1} \int_{\rho} \int_{0}^{1} \int_{1}^{1} \int_{\rho} \int_{0}^{1} \int_$$

#### Pipes-Munjal 1967

$$v(
ho) = v_{\max} (1 - 
ho^{lpha})$$
  $lpha > 0$ 

Example:  $\alpha = 0.2$ 



#### Pipes-Munjal 1967

$$v(
ho) = v_{\max} (1 - 
ho^{lpha})$$
  $lpha > 0$ 

Example:  $\alpha = 2$ 



## A quick review on the mathematical theory.

Discontinuous solutions, weak solutions: for all φ ∈ C<sup>1</sup><sub>c</sub>([0, +∞) × ℝ),

$$\int_{\mathbb{R}}\int_{\mathbb{R}_+}\left[\rho(t,x)\phi_t(t,x)+f(\rho(t,x))\phi_x(t,x)\right]\mathrm{d}t\,\mathrm{d}x+\int_{\mathbb{R}}\bar{\rho}(x)\phi(0,x)\,\mathrm{d}x=0$$

Non uniqueness of weak solutions. Uniqueness of *entropy solutions* (Oleinik 1963, Kružkov 1970): for all test functions φ ≥ 0 and for all k ∈ R,

$$\int_{\mathbb{R}} \int_{\mathbb{R}_{+}} \left[ \left| \rho(t, x) - k \right| \phi_{t}(t, x) + \operatorname{sgn} \left( \rho(t, x) - k \right) \left[ f\left( \rho(t, x) \right) - f(k) \right] \phi_{x}(t, x) \right] dt dx + \int_{\mathbb{R}} \phi(0, x) \left| \bar{\rho}(x) - k \right| dx \ge 0$$
(1)

- Initial trace. Uniqueness if (1) is satisfied with *φ*(*t* = 0) = 0 and initial trace is reached in the weak-\* L<sup>∞</sup>-topology, provided *f* is not affine a.e. (Chen-Rascle 2000).
- Oleinik condition. Entropy solutions are characterised by

$$f'(\rho)_x \leq \frac{1}{t},$$
 in  $\mathcal{D}'((0,+\infty) \times \mathbb{R}).$ 

## Constructing entropy solutions.

#### Regularization strategy.

• Vanishing viscosity, see Dafermos 2000 and the references therein.

#### Numerical methods.

- Finite differences. Glimm 1965.
- Wave front tracking method. Dafermos 1972.

#### Mesoscopic approximation.

• Kinetic approximation. Lions-Perthame-Tadmor 1994.

#### Microscopic probabilistic approach.

Exclusion processes (list incomplete!). Rost 1982, Ferrari-Fouque 1987.

#### Microscopic deterministic approach.

- Sticky particles. Brenier-Grenier 1998.
- Lagrangian, follow-the-leader type systems.

Formal derivation: Whitham 1974, Colombo-Rossi 2013.

Rigorous derivation: Di Francesco-Rosini 2015.

## Table of contents.



The motivating problem: traffic flow

- Macroscopic models
- Follow the leader approximation
- Convergence result
  - A preliminary lemma
  - Weak convergence
  - Uniform estimates
  - Time continuity
  - Convergence to entropy solutions
- 3 Remarks and future projects

### The follow-the-leader particle approximation.

Fix the initial condition  $\bar{\rho}$  and let *L* be its total mass.

- Fix  $n \in \mathbb{N}$  and let  $\ell = \ell_n = L2^{-n}$  be the length of each platoon of cars.
- Consider  $N + 1 = N_n + 1 = 2^n + 1$  (ordered) *reference cars*  $x_0, x_1, ..., x_N$  corresponding to the end points of the *N* platoons.
- The cars  $x_0, \ldots, x_N$  evolve according to

$$\begin{split} \dot{x}_N(t) &= v_{\max}, \\ \dot{x}_i(t) &= v \left( \frac{\ell}{x_{i+1}(t) - x_i(t)} \right), \qquad i = N-1, \dots, 0. \end{split}$$

• The initial conditions of the cars are taken by atomization of  $\bar{\rho}$ , i.e.

$$\begin{aligned} x_0(t=0) &= \bar{x}_0 = \min(\operatorname{spt}(\bar{\rho})), \\ x_i(t=0) &= \bar{x}_i = \sup\left\{x \in \mathbb{R} \colon \int_{\bar{x}_{i-1}}^x \bar{\rho}(y) \, \mathrm{d}y < \ell\right\}, \qquad i = 1, \dots, N. \end{aligned}$$

### Atomization of the initial condition.

The initial condition  $\bar{\rho}$  is split into *N* parts with equal integral  $\ell$ .



## Large particle limit.

### Empirical measure

$$\tilde{\rho}^n(t) = \sum_{i=1}^N \ell_n \, \delta_{x_i(t)}$$

### Discrete density

$$\hat{\rho}^{n}(t,x) = \sum_{i=1}^{N} y_{i}^{n}(t) \chi_{[x_{i}(t),x_{i+1}(t))}(x), \qquad \qquad y_{i}^{n}(t) = \frac{\ell_{n}}{x_{i+1}(t) - x_{i}(t)}$$

#### Goal:

Prove that  $\tilde{\rho}^n(t)$  and  $\hat{\rho}^n(t, \cdot)$  converge to the unique entropy solution  $\rho$  of the LWR equation with  $\bar{\rho}$  as initial condition.

### Empirical measure and discrete density.



### Heuristic derivation.

Let ρ be the entropy solution of

$$\rho_t + [\rho \mathbf{v}(\rho)]_{\mathbf{x}} = \mathbf{0}. \tag{2}$$

Let *F* be the cumulative distribution of *ρ*:

$$F(x,t) = \int_{-\infty}^{x} \rho(t,y) \mathrm{d}y.$$

• Let X be the pseudo inverse of F:

$$X(t,z) = \inf \left\{ x \in \mathbb{R} \colon F(x) > z \right\}, \qquad z \in [0,L].$$

• Formally, X(t, z) satisfies F(t, X(t, z)) = z and

$$F_{x} = \rho \\ F_{t} = -\rho v(\rho)$$
  $\Rightarrow$   $1 = F(t, X(t, z))_{z} = F_{x} X_{z} = \rho X_{z} \\ 0 = F(t, X(t, z))_{t} = F_{t} + F_{x} X_{t} = \rho(X_{t} - v(\rho))$   $\Rightarrow X_{t} = v\left(\frac{1}{X_{z}}\right).$  (3)

● Forward *z*-finite difference of (3) with step ℓ gives

$$X_t(t,z) = v\left(\frac{\ell}{X(t,z+\ell) - X(t,z)}\right), \qquad z = 0,\ldots,(N-1)\ell.$$
(4)

The follow-the-leader system (4) is the discrete Lagrangian version of (2).

Rosini (ICM, UW)

Conservation laws via particle systems

## Table of contents.



The motivating problem: traffic flow

- Macroscopic models
- Follow the leader approximation

### 2

#### Convergence result

- A preliminary lemma
- Weak convergence
- Uniform estimates
- Time continuity
- Convergence to entropy solutions

Remarks and future projects

## Convergence Theorem.

Let  $\rho$  be the unique entropy solution of

$$\rho_t + [\rho \mathbf{v}(\rho)]_{\mathbf{x}} = \mathbf{0}, \qquad \qquad \rho(t = \mathbf{0}) = \bar{\rho},$$

where  $\bar{\rho}$  is in  $\mathcal{M}_{L} \cap L^{\infty}(\mathbb{R})$ ,  $\nu$  is in  $C^{1}(\mathbb{R}_{+})$ , strictly decreasing with  $\nu(0) = \nu_{max} > 0$ .

### Theorem (MDF and MDR, ARMA 2015.)

lf

•  $\bar{
ho} \in \mathsf{BV}(\mathbb{R})$ ,

or

$$\mathbb{R}_+ \ni \rho \mapsto \rho \mathbf{v}'(\rho) \in \mathbb{R}_+ \text{ is non-increasing.}$$

Then,

- the sequence  $\hat{\rho}^n \to \rho$  almost everywhere and in  $L^1_{loc}([0, +\infty) \times \mathbb{R})$ .
- the sequence ρ̃<sup>n</sup> → ρ in the topology L<sup>1</sup><sub>loc</sub>([0, +∞); d<sub>L,1</sub>), where d<sub>L,1</sub> is the scaled 1-Wasserstein distance.

 $\mathcal{M}_{L} = \{\rho \text{ Radon measure on } \mathbb{R} \text{ with compact support: } \rho \geq 0, \ \rho(\mathbb{R}) = L \}$  $d_{L,1}(\rho_{1},\rho_{2}) = L d_{1}(\rho_{1}/L,\rho_{2}/L) = \|F_{\rho_{1}} - F_{\rho_{2}}\|_{L^{1}(\mathbb{R})}$ 

### Ingredients: cumulative distributions.



### Ingredients: pseudo inverses.



### Ingredients: discrete Lagrangian density.

$$\check{\rho}^n = \hat{\rho}^n \circ \hat{X}^n$$
 is **PC**



Rosini (ICM, UW)

## Strategy of the proof.

- Prove that (X̃<sup>n</sup>)<sub>n</sub> has a strong limit X in L<sup>1</sup><sub>loc</sub>([0, +∞[×[0, L]), equivalent to (ρ̃<sup>n</sup>)<sub>n</sub> converging to a measure ρ in L<sup>1</sup><sub>loc</sub>([0, +∞[; d<sub>L1</sub>).
  - Prove that (X̂<sup>n</sup>)<sub>n</sub> converges in L<sup>1</sup><sub>loc</sub>([0, +∞[×[0, L])) to the same limit X, i.e. (ρ̂<sup>n</sup>)<sub>n</sub> converges to ρ in L<sup>1</sup><sub>loc</sub>([0, +∞[; d<sub>L,1</sub>).
- Prove that X has difference quotients bounded below by 1, i.e. ρ is actually in L<sup>∞</sup> and is a.e. bounded by 1.
  - This easily implies weak-∗ convergence of (*p˜*<sup>n</sup>)<sub>n</sub> to a limit *p˜* in L<sup>∞</sup>.
  - It remains to prove that *ρ* ∘ *F* = *ρ*, and that such limit is the unique entropy solution to LWR. This requires stronger estimates on *ρ*<sup>n</sup>.
- (iii) Case  $\bar{\rho} \in \mathbf{BV}$ : direct uniform **BV** estimate of  $\hat{\rho}^n$ .
- (iv) Prove that  $\rho$  is a weak solution: it follows from letting  $n \to +\infty$  in the formulation of the FTL system

$$\tilde{X}_t^n = v(\check{\rho}^n).$$

(v) Prove that ρ is an entropy solution: in the discrete setting, use strong L<sup>1</sup> compactness to pass it to the limit.

## Table of contents.



The motivating problem: traffic flow

- Macroscopic models
- Follow the leader approximation



#### Convergence result

- A preliminary lemma
- Weak convergence
- Uniform estimates
- Time continuity
- Convergence to entropy solutions

Remarks and future projects

## Discrete maximum principle.

The global existence for the FTL system is guaranteed by the following

### Lemma (Discrete maximum principle)

For all  $i = 0, \ldots, N-1$ , we have  $\ell \le x_{i+1}(t) - x_i(t)$  for all times  $t \ge 0$ .

### Proof.

- Replace v with its extension  $V = v\chi_{[0,1]}$ , V is still Lipschitz.
- Assume by contradiction  $x_{i+1}(t_1) x_i(t_1) = \ell$ , and  $x_{i+1}(t) x_i(t) < \ell$  on  $t \in (t_1, t_2]$ .
- Integrate FTL on [t<sub>1</sub>, t]:

$$x_i(t) = x_i(t_1) + \int_{t_1}^t V\left(\frac{\ell}{x_{i+1}(\tau) - x_i(\tau)}\right) \mathrm{d}\tau = x_i(t_1).$$

- $\ell > x_{i+1}(t) x_i(t) \ge x_{i+1}(t_1) x_i(t_1) = \ell$ , contradiction!
- By uniqueness, the same holds for the system with v.

## Table of contents.

#### 1

The motivating problem: traffic flow

- Macroscopic models
- Follow the leader approximation

### 2

#### Convergence result

A preliminary lemma

#### Weak convergence

- Uniform estimates
- Time continuity
- Convergence to entropy solutions

#### Remarks and future projects

## Strong compactness of $\hat{X}^n$ and $\tilde{X}^n$ .

#### 1-dimensional Wasserstein distance

 $\begin{array}{l} F_i \text{ cumulative distribution of } \rho_i \\ X_i \text{ pseudo-inverse of } F_i \end{array} \right\} \Rightarrow d_{L,1}(\rho_1, \rho_2) = \|X_1 - X_2\|_{L^1([0,L])}$ 

#### Proposition

There exists a unique  $X \in L^{\infty}(\mathbb{R}_+ \times [0, L])$ , monotone non-decreasing and right continuous with respect to *z*, such that

$$(\hat{X}^n)_n$$
 and  $(\tilde{X}^n)_n$  converge to  $X$  in  $L^1_{loc}(\mathbb{R}_+ \times [0, L])$ ,

and for any t, s > 0

$$TV[X(t)] \le |\bar{x}_N - \bar{x}_0 + v_{\max} t|, \qquad (5a)$$

$$\|X(t)\|_{\mathbf{L}^{\infty}([0,L];\mathbb{R})} \le \max\{|\bar{x}_{0}|,|\bar{x}_{N}+v_{\max}|t|\},$$
(5b)

$$\int_{0}^{L} |X(t,z) - X(s,z)| \, \mathrm{d}z \le v_{\max} \, L \, |t-s|. \tag{5c}$$

Moreover,  $(\tilde{X}^n)_n$  converges to X a.e. on  $\mathbb{R}_+ \times [0, L]$ .

## Strong compactness of $\hat{X}^n$ and $\tilde{X}^n$ .

#### Proof.

Fix  $T > t > s \ge 0$ . Estimates (5a) and (5b) are immediately proven for  $\tilde{X}^n$ .

$$\begin{split} &\int_0^L \left| \tilde{X}^n(t,z) - \tilde{X}^n(s,z) \right| \mathrm{d}z = \sum_{i=0}^{N_n-1} \ell_n \left[ x_i^n(t) - x_i^n(s) \right] \\ &= \sum_{i=0}^{N_n-1} \ell_n \left[ \int_s^t v\left( y_i^n(\tau) \right) \mathrm{d}\tau \right] \leq v_{\max} \, L\left(t-s\right). \end{split}$$

By Helly's theorem,  $\tilde{X}^n$  converges strongly as in the statement up to a subsequence. As  $\tilde{X}^n$  is monotone in *n*, the whole sequence converges to a unique limit *X*.

$$\begin{split} &\int_{0}^{L} \left| \hat{X}^{n}(t,z) - \tilde{X}^{n}(t,z) \right| \mathrm{d}z = \sum_{i=0}^{N_{n}-1} y_{i}^{n}(t)^{-1} \int_{i\,\ell_{n}}^{(i+1)\,\ell_{n}} \left[ z - i\,\ell_{n} \right] \,\mathrm{d}z \\ &= \frac{\ell_{n}}{2} \sum_{i=0}^{N_{n}-1} \left[ x_{i+1}^{n}(t) - x_{i}^{n}(t) \right] = \frac{\ell_{n}}{2} \left[ x_{N_{n}}^{n}(t) - x_{0}^{n}(t) \right] \leq \frac{\ell_{n}}{2} \left[ \bar{x}_{\max} - \bar{x}_{\min} + v_{\max} T \right], \end{split}$$

Hence,  $\hat{X}^n$  and  $\tilde{X}^n$  have the same limit.

### $L^{\infty}$ bound for the limit measure.

• Let  $0 \le z_1 < z_2 \le L$ .

• For *n* sufficiently large, let  $i \ell_n \le z_1 < (i+1) \ell_n$  and  $\ell_n j \le z_2 < \ell_n (j+1)$ 

• The discrete maximum principle implies

$$\frac{\tilde{X}^{n}(t, z_{2}) - \tilde{X}^{n}(t, z_{1})}{z_{2} - z_{1}} \geq \frac{x_{j}^{n}(t) - x_{i}^{n}(t)}{(j+1)\ell_{n} - i\ell_{n}} \geq \frac{(j-i)\ell_{n}}{(j+1)\ell_{n} - i\ell_{n}}$$
$$= 1 - \frac{1}{j-i+1} \geq 1 - \frac{1}{(z_{2}\ell_{n}^{-1} - 1) - z_{1}\ell_{n}^{-1} + 1} = 1 - \frac{\ell_{n}}{z_{2} - z_{1}}.$$

• By sending 
$$n \to +\infty$$
, we get

$$\partial_z X(t,\cdot) \geq 1$$
 in  $\mathcal{D}'$ .

• Let  $F(t, \cdot)$  be the pseudo inverse of  $X(t, \cdot)$ . Then  $\rho = F_x$  satisfies

 $\rho(t, x) \leq 1.$ 

•  $\tilde{\rho}^n$  and  $\hat{\rho}^n$  converge to  $\rho$  in  $L^1_{loc}([0, +\infty); d_{L,1})$ .

*ρ*<sup>n</sup> = *ρ̂<sup>n</sup>* ∘ *X̂<sup>n</sup>* converges to some limit *ρ* up to a subsequence in the weak-\* L<sup>∞</sup> topology.

In the sequel, we shall make an extensive use of the discrete equations

$$\dot{y}_{i}^{n}(t) = -\frac{y_{i}^{n}(t)^{2}}{\ell_{n}} \bigg[ v(y_{i+1}^{n}(t)) - v(y_{i}^{n}(t)) \bigg], \qquad i = 0, \dots, N-2,$$
  
$$\dot{y}_{N-1}^{n}(t) = -\frac{y_{N-1}^{n}(t)^{2}}{\ell_{n}} \bigg[ v_{\max} - v(y_{N-1}^{n}(t)) \bigg].$$

## Table of contents.



The motivating problem: traffic flow

- Macroscopic models
- Follow the leader approximation



#### Convergence result

- A preliminary lemma
- Weak convergence
- Uniform estimates
- Time continuity
- Convergence to entropy solutions

Remarks and future projects

### **BV** contraction for **BV** initial data.

### Proposition (BV contractivity for BV initial data)

Assume  $\bar{\rho} \in \mathbf{BV}$ . Then for any  $n \in \mathbb{N}$ 

$$\operatorname{TV}\left[\hat{\rho}^{n}(t)\right] = \operatorname{TV}\left[\check{\rho}^{n}(t)\right] \leq \operatorname{TV}\left[\bar{\rho}\right] \qquad \qquad \text{for all } t \geq 0.$$

**Proof.** (We omit the index *n* for simplicity)

$$\begin{aligned} \frac{\mathrm{d}}{\mathrm{d}t} \mathrm{TV}\left[\hat{\rho}(t)\right] &= \frac{\mathrm{d}}{\mathrm{d}t} \left[ y_0 + y_{N-1} + \sum_{i=0}^{N-2} |y_i - y_{i+1}| \right] \\ &= \dot{y}_0 + \dot{y}_{N-1} + \sum_{i=0}^{N-2} \mathrm{sgn}\left[y_i - y_{i+1}\right] [\dot{y}_i - \dot{y}_{i+1}] \\ &= \left[ 1 + \mathrm{sgn}\left[y_0 - y_1\right] \right] \dot{y}_0 + \left[ 1 - \mathrm{sgn}\left[y_{N-2} - y_{N-1}\right] \right] \dot{y}_{N-1} \\ &+ \sum_{i=1}^{N-2} \left[ \mathrm{sgn}\left[y_i - y_{i+1}\right] - \mathrm{sgn}\left[y_{i-1} - y_i\right] \right] \dot{y}_i. \end{aligned}$$

Proof (continued): We analise all the terms above, and we get

$$\begin{split} & \left[1 + \operatorname{sgn}\left[y_0 - y_1\right]\right] \dot{y}_0 = -\left[1 + \operatorname{sgn}\left[y_0 - y_1\right]\right] \frac{y_0^2}{\ell} \left[v(y_1) - v(y_0)\right] \le 0, \\ & \left[1 - \operatorname{sgn}\left[y_{N-2} - y_{N-1}\right]\right] \dot{y}_{N-1} = -\left[1 - \operatorname{sgn}\left[y_{N-2} - y_{N-1}\right]\right] \frac{y_{N-1}^2}{\ell} \left[v_{\max} - v(y_{N-1})\right] \le 0, \\ & \left[\operatorname{sgn}\left[y_i - y_{i+1}\right] - \operatorname{sgn}\left[y_{i-1} - y_i\right]\right] \dot{y}_i = \\ & = -\left[\operatorname{sgn}\left[y_i - y_{i+1}\right] - \operatorname{sgn}\left[y_{i-1} - y_i\right]\right] \frac{y_i^2}{\ell} \left[v(y_{i+1}) - v(y_i)\right] \le 0. \end{split}$$

Therefore, TV  $[\hat{\rho}(t)] \leq$  TV  $[\bar{\rho}]$  for all  $t \geq$  0.

### Lemma (Discrete Oleinik-type condition)

Assume v satisfies the additional assumption  $\rho v'(\rho)$  non-increasing. Then, for any  $i = 0, ..., N_n - 2$  we have

$$t y_i^n(t) \left[ v \left( y_{i+1}^n(t) \right) - v \left( y_i^n(t) \right) \right] \le \ell_n \qquad \text{for all } t \ge 0. \tag{6}$$

#### Remark

Condition (6) in terms of  $x_i(t)$ 

$$\frac{v\left(y_{i+1}^{n}(t)\right) - v\left(y_{i}^{n}(t)\right)}{x_{i+1}(t) - x_{i}(t)} \leq \frac{1}{t} \qquad \qquad \text{for all } t \geq 0.$$

(7) is a discrete analogous of

$$v(\rho)_x \leq \frac{1}{t}.$$

However, the sharp form of the Oleinik condition for the scalar conservation law is (cf. Hoff 1983)

$$f'(\rho)_{\mathsf{X}} = (\mathsf{V}(\rho) + \rho \mathsf{V}'(\rho))_{\mathsf{X}} \leq \frac{1}{t}.$$

(7)

**Notation:** (we omit the dependence on *n* and *t*)

$$z_i \doteq t y_i [v(y_{i+1}) - v(y_i)], \qquad i = 0, \dots, N-2,$$
  
$$z_{N-1} \doteq t y_{N-1} [v_{\max} - v(y_{N-1})].$$

The Lemma is proven once we provide the estimate  $z_i \leq \ell$  for i = 1, ..., N - 1. • STEP 0:  $z_{N-1} \leq \ell$ .

$$\begin{aligned} \dot{z}_{N-1} &= y_{N-1} \left[ v_{\max} - v(y_{N-1}) \right] + t \, \dot{y}_{N-1} \left[ v_{\max} - v(y_{N-1}) \right] - t \, y_{N-1} \, v'(y_{N-1}) \, \dot{y}_{N-1} \\ &= y_{N-1} \left[ v_{\max} - v(y_{N-1}) \right] - \frac{t \, y_{N-1}^2}{\ell} \left[ v_{\max} - v(y_{N-1}) \right]^2 \\ &+ \frac{t \, v'(y_{N-1}) \, y_{N-1}^3}{\ell} \left[ v_{\max} - v(y_{N-1}) \right] \\ &\leq y_{N-1} \left[ v_{\max} - v(y_{N-1}) \right] \left[ 1 - \frac{z_{N-1}}{\ell} \right]. \end{aligned}$$

Since  $z_{N-1}(0) = 0$ , from the above estimate we get  $z_{N-1}(t) \le \ell$  for all  $t \ge 0$ .

### Proof of the discrete Oleinik condition (continued).

$$\bullet \; \text{Step 1: } z_{i+1} \leq \boldsymbol{\ell} \Rightarrow z_i \leq \boldsymbol{\ell}.$$

$$\begin{aligned} \dot{z}_{i} &= y_{i} \left[ v \left( y_{i+1} \right) - v \left( y_{i} \right) \right] + t \dot{y}_{i} \left[ v \left( y_{i+1} \right) - v \left( y_{i} \right) \right] + t y_{i} \left[ v'(y_{i+1}) \dot{y}_{i+1} - v'(y_{i}) \dot{y}_{i} \right] \\ &= y_{i} \left[ v \left( y_{i+1} \right) - v \left( y_{i} \right) \right] - \frac{t y_{i}^{2}}{\ell} \left[ v(y_{i+1}) - v(y_{i}) \right]^{2} \\ &+ t y_{i} \left[ -\frac{v'(y_{i+1}) y_{i+1}^{2}}{\ell} \left[ v(y_{i+2}) - v(y_{i+1}) \right] + \frac{v'(y_{i}) y_{i}^{2}}{\ell} \left[ v(y_{i+1}) - v(y_{i}) \right] \right] \\ &= y_{i} \left[ v \left( y_{i+1} \right) - v \left( y_{i} \right) \right] - \frac{y_{i}}{\ell} \left[ v(y_{i+1}) - v(y_{i}) \right] z_{i} - \frac{v'(y_{i+1}) y_{i} y_{i+1}}{\ell} z_{i+1} + \frac{v'(y_{i}) y_{i}^{2}}{\ell} z_{i}. \end{aligned}$$

Since  $\operatorname{sgn}_+[z_i] = \operatorname{sgn}_+[v(y_{i+1}) - v(y_i)] = \operatorname{sgn}_+[y_i - y_{i+1}]$  for all t > 0, from the assumption on  $z_{i+1}$  we easily obtain

$$\frac{\mathrm{d}}{\mathrm{d}t}[z_{i}]_{+} = y_{i} \left[ v \left( y_{i+1} \right) - v \left( y_{i} \right) \right]_{+} - \frac{y_{i}}{\ell} \left[ v \left( y_{i+1} \right) - v \left( y_{i} \right) \right]_{+} \left[ z_{i} \right]_{+} 
- \frac{v'(y_{i+1}) y_{i} y_{i+1}}{\ell} \operatorname{sgn}_{+}[z_{i}] z_{i+1} + \frac{v'(y_{i}) y_{i}^{2}}{\ell} \left[ z_{i} \right]_{+} 
\leq y_{i} \left[ v \left( y_{i+1} \right) - v \left( y_{i} \right) \right]_{+} \left[ 1 - \frac{[z_{i}]_{+}}{\ell} \right] - v'(y_{i+1}) y_{i} y_{i+1} \operatorname{sgn}_{+}[z_{i}] + \frac{v'(y_{i}) y_{i}^{2}}{\ell} \left[ z_{i} \right]_{+}.$$

### Proof of the discrete Oleinik condition (continued).

The additional condition on v gives  $-v'(y_{i+1})y_{i+1} \leq -v'(y_i)y_i$  for  $y_i \geq y_{i+1}$ , and then

$$\begin{aligned} \frac{\mathrm{d}}{\mathrm{d}t}[z_i]_+ &\leq y_i \left[ v\left(y_{i+1}\right) - v\left(y_i\right) \right]_+ \left[ 1 - \frac{[z_i]_+}{\ell} \right] - v'(y_i) y_i^2 \operatorname{sgn}_+[z_i] + \frac{v'(y_i) y_i^2}{\ell} [z_i]_+ \\ &= y_i \left[ \left[ v\left(y_{i+1}\right) - v\left(y_i\right) \right]_+ - v'(y_i) y_i \right] \left[ 1 - \frac{[z_i]_+}{\ell} \right]. \end{aligned}$$

Now, as  $v' \le 0$ , and since  $z_i(0) = 0$ , we get that  $z_i(t)_+ \le \ell$  for all  $t \ge 0$ . • STEP 2:  $z_{N-2} \le \ell$ . From analogous computations as in previous step, by using the monotonicity of  $y \mapsto y v'(y)$  and Step 0, we get

$$\frac{\mathrm{d}}{\mathrm{d}t}[z_{N-2}]_{+} \leq y_{N-2} \left[ \left[ v \left( y_{N-1} \right) - v \left( y_{N-2} \right) \right]_{+} - v'(y_{N-2}) y_{N-2} \right] \left[ 1 - \frac{[z_{N-2}]_{+}}{\ell} \right]$$

Again,  $v' \le 0$  and  $z_{N-2}(0) = 0$  imply that  $z_{N-2}(t)_+ \le \ell$  for all  $t \ge 0$ . • **CONCLUSION**. The estimate (6) is proven recursively: Step 2 provides the first step with i = N-2, whereas Step 1 proves that the estimate holds for all  $i \in \{0, ..., N-3\}$ .

**Remark:** The discrete Oleinik condition provides a uniform **BV** estimate away from t = 0. Here we use that solutions have compact support.

Rosini (ICM, UW)

Conservation laws via particle systems

## Table of contents.



The motivating problem: traffic flow

- Macroscopic models
- Follow the leader approximation

### 2

#### Convergence result

- A preliminary lemma
- Weak convergence
- Uniform estimates
- Time continuity
- Convergence to entropy solutions

#### Remarks and future projects

## A technical problem.

Typically (e.g. the wave-front-tracking algorithm for conservation laws)  $L^1$ -continuity of the approximating sequence gives the desired compactness via Helly's Theorem. Here we are able to prove such an estimate for  $\check{\rho}^n$ , but not for  $\hat{\rho}^n$ .

### Proposition (Uniform L<sup>1</sup>-continuity in time of $\check{\rho}^n$ )

For any  $\delta > 0$  we have

$$\int_0^L \left| \check{\rho}^n(t,z) - \check{\rho}^n(s,z) \right| \mathrm{d} z \le C |t-s| \qquad \qquad \text{for all } t,s \ge \delta,$$

with some C depending on  $\delta$ .

#### Proof.

(Sketched) A direct computation of the l.h.s. and the discrete maximum principle give

$$\int_{0}^{L} \left| \check{\rho}^{n}(t,z) - \check{\rho}^{n}(s,z) \right| \mathrm{d}z = \sum_{i=0}^{N_{n}-1} \ell_{n} \left| y_{i}^{n}(t) - y_{i}^{n}(s) \right| \leq \int_{s}^{t} \left[ \mathrm{TV} \left[ v \left( \check{\rho}^{n}(\tau) \right) \right] + v_{\max} \right] \mathrm{d}\tau.$$

### How to solve the technical problem?

### Proposition (Uniform Wasserstein time continuity of $\hat{\rho}^n$ )

#### For any $n \in \mathbb{N}$ we have

$$\mathcal{O}_{L,1}\left(\hat{
ho}^n(t),\hat{
ho}^n(s)
ight)\leq 2\,L\,v_{ ext{max}}\left|t-s
ight|$$

for all 
$$s, t \ge 0$$
.

#### Proof.

$$\begin{aligned} d_{L,1}\left(\hat{\rho}^{n}(t),\hat{\rho}^{n}(s)\right) &= \left\|\hat{X}^{n}(t) - \hat{X}^{n}(s)\right\|_{L^{1}([0,L];\mathbb{R})} = \sum_{i=0}^{N_{n}-1} \int_{i\ell_{n}}^{(i+1)\ell_{n}} \left[\hat{X}^{n}(t,z) - \hat{X}^{n}(s,z)\right] dz \\ &= \sum_{i=0}^{N_{n}-1} \ell_{n} \left[x_{i}^{n}(t) - x_{i}^{n}(s)\right] + \sum_{i=0}^{N_{n}-1} \left[y_{i}^{n}(t)^{-1} - y_{i}^{n}(s)^{-1}\right] \int_{i\ell_{n}}^{(i+1)\ell_{n}} (z - i\ell_{n}) dz \\ &= \sum_{i=0}^{N_{n}-1} \ell_{n} \int_{s}^{t} v\left(y_{i}^{n}(\tau)\right) d\tau + \sum_{i=0}^{N_{n}-1} \frac{\ell_{n}^{2}}{2} \int_{s}^{t} \frac{d}{d\tau} \left[y_{i}^{n}(\tau)^{-1}\right] d\tau \\ &\leq L v_{\max}\left(t - s\right) + \frac{\ell_{n}}{2} \int_{s}^{t} \left[v_{\max} - v\left(y_{0}^{n}(\tau)\right)\right] d\tau \leq 2L v_{\max}\left(t - s\right). \end{aligned}$$

## A generalisation of Aubin-Lions Lemma.

The desired compactness then follows from the following

#### Theorem (Generalized Aubin-Lions lemma, Rossi-Savaré 2003)

Let T, L > 0 and  $I \subset \mathbb{R}$  be a bounded open convex interval. Assume  $w : \mathbb{R} \to \mathbb{R}$  is a Lipschitz continuous and strictly monotone function. Let  $(\rho^n)_{n \in \mathbb{N}}$  be a nonnegative sequence in  $\mathbf{L}^{\infty}$  ( $(0, T) \times \mathbb{R}; \mathbb{R}$ ), with compact support and fixed mass L > 0, such that:

•  $\rho^n$ :  $(0, T) \to L^1(\mathbb{R}; \mathbb{R})$  is measurable for all  $n \in \mathbb{N}$ ;

• spt 
$$(\rho^n(t)) \subseteq I$$
 for all  $t \in ]0, T[$  and  $n \in \mathbb{N}$ ;

• 
$$\sup_{n \in \mathbb{N}} \int_0^T \left[ \left\| w\left(\rho^n(t)\right) \right\|_{\mathbf{L}^1(l;\mathbb{R})} + \mathrm{TV}\left[ w\left(\rho^n(t)\right) \right] \right] \mathrm{d}t < +\infty;$$

 There exists a constant C depending only on T such that d<sub>L,1</sub> (ρ<sup>n</sup>(s), ρ<sup>n</sup>(t)) ≤ C |t − s| for all s, t ∈ ]0, T[ and n ∈ N.

Then,  $(\rho^n)_{n \in \mathbb{N}}$  is strongly relatively compact in  $L^1((0, T) \times \mathbb{R}; \mathbb{R})$ .

## Table of contents.



The motivating problem: traffic flow

- Macroscopic models
- Follow the leader approximation

### 2

#### Convergence result

- A preliminary lemma
- Weak convergence
- Uniform estimates
- Time continuity
- Convergence to entropy solutions



Another important property that we need to check is

$$\check{\rho}(t, F(t, x)) = \rho(t, x),$$
 on  $\operatorname{spt}(\rho)$ ,

where

- $\check{\rho}$  is the strong limit of  $\check{\rho}^n$ ,
- $\rho$  is the strong limit of  $\hat{\rho}^n$ ,
- F is the cumulative distribution of  $\rho$ .

This is ensured by the strong compactness of both  $\check{\rho}^n$  and  $\hat{\rho}^n$ .

### Proposition

The limit function  $\rho$  of  $\hat{\rho}^n$  is a weak solution of the LWR equation with i.c.  $\bar{\rho}$ .

#### Proof.

Let  $\phi \in \mathbf{C}^{\infty}_{\mathbf{c}}$  ([0, + $\infty$ [  $\times \mathbb{R}$ ;  $\mathbb{R}$ ). We have

$$\begin{split} &\int_{\mathbb{R}_{+}} \int_{0}^{L} \left[ v\left( \check{\rho}^{n}(t,z) \right) \phi_{x}\left( t, \tilde{X}^{n}(t,z) \right) \right] \mathrm{d}z \mathrm{d}t = \sum_{i=0}^{N_{n}-1} \int_{\mathbb{R}_{+}} \int_{i\,\ell_{n}}^{(i+1)\,\ell_{n}} \left[ v\left( y_{i}^{n}(t) \right) \phi_{x}\left( t, x_{i}^{n}(t) \right) \right] \mathrm{d}z \,\mathrm{d}t \\ &= \sum_{i=0}^{N_{n}-1} \int_{\mathbb{R}_{+}} \int_{i\,\ell_{n}}^{(i+1)\,\ell_{n}} \left[ \frac{\mathrm{d}}{\mathrm{d}t} \phi\left( t, x_{i}^{n}(t) \right) - \phi_{t}\left( t, x_{i}^{n}(t) \right) \right] \mathrm{d}z \,\mathrm{d}t \\ &= -\int_{0}^{L} \phi\left( 0, \tilde{X}^{n}(0,z) \right) \mathrm{d}z - \int_{\mathbb{R}_{+}} \int_{0}^{L} \phi_{t}\left( t, \tilde{X}^{n}(t,z) \right) \mathrm{d}z \,\mathrm{d}t. \end{split}$$

By the strong convergence of  $\tilde{X}^n$  and  $\check{\rho}^n$ , by chancing variable x = X(t, z), and by using  $\check{\rho}(t, F(t, x)) = \rho(t, x)$  a.e. on  $\operatorname{spt}(\rho)$ , we get the definition of weak solution for LWR.

#### Conclusion of the proof.

- We need to prove that the limit  $\rho$  is an entropy solution in the Kružkov sense.
- This is trivial in the uniformly concave case f'' ≤ -ε < 0, as in that case f'(ρ) ≃ v(ρ), and one can obtain the sharp Oleinik condition f'(ρ)<sub>x</sub> ≤ 1/t in the limit.
- In the general case, we need to use the definition of entropy solution by Kružkov. This follows from the inequality

$$\int_{\mathbb{R}_+} \int_{\mathbb{R}} \left[ \left| \hat{\rho}^n(t,x) - k \right| \phi_t(t,x) + \operatorname{sgn}(\hat{\rho}^n(t,x) - k) \left[ f(\hat{\rho}^n(t,x)) - f(k) \right] \phi_x(t,x) \right] \mathrm{d}x \, \mathrm{d}t \ge o(1),$$

as  $n \to +\infty$ , which is very technical and is omitted here.

## Table of contents.

#### 1

The motivating problem: traffic flow

- Macroscopic models
- Follow the leader approximation
- Convergence res
  - A preliminary lemma
  - Weak convergence
  - Uniform estimates
  - Time continuity
  - Convergence to entropy solutions

#### Remarks and future projects

## Concluding remarks.

- We remark that our set of assumptions on *v* allows for degenerate concave fluxes at zero.
- In the case of linear velocity v, e.g.  $v(\rho) = v_{max}(1 \rho)$ , the convergence to a weak solution can be obtained without the need of the **BV** estimates, as the velocity term in the pseudo-inverse PDE is linear. This is somehow intrinsic in using a Lagrangian description.
- In order to get continuity in time for the sequence  $\hat{\rho}^n$ , the most natural try would be getting L<sup>1</sup>-continuity. Encouraged by the L<sup>1</sup> time equi-continuity of  $\check{\rho}^n$ , we have attempted at proving such a property in many ways without success. This is the reason why use the generalized Aubin-Lions lemma, which allows to take advantage of the Wasserstein equi-continuity of  $\hat{\rho}^n$ , and still get the same L<sup>1</sup>-compactness in the end. The only drawback of this strategy is that we can't get any L<sup>1</sup> time continuity for the limit.
- Our approach has the advantage of providing a piecewise constant approximation with a non increasing number of jumps. The price to pay for such a simplification is that we lose the classical shock structure at a microscopic *level*. Indeed, the explicit solution to the FTL system even for simple Riemann-type initial conditions is not immediate.

- Extending the results to Dirichlet boundary conditiont (phantom moving particles at the boundary).
- Use this strategy to attack the existence theory of similar models, e.g. with discontinuous flux.

## Numerical simulation.

THANK YOU FOR YOUR ATTENTION