

IPAM Tutorials 2020 on Autonomous Vehicles

Macroscopic models for Autonomous Vehicles

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Outline of the talk

1 Introduction to macroscopic models

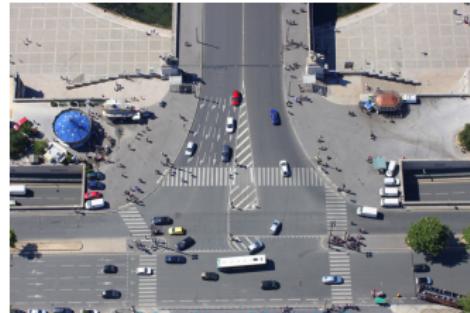
2 Macroscopic models for Autonomous Vehicles

- PDE-ODE models
- Non-local models

Traffic flow modeling: approaches

Microscopic

- individual agents
- ODE system
- many parameters
- low and high densities
- comp. cost \sim number of agents



@fotolia - MurieleB

Macroscopic

- continuous flow
- PDEs
- few parameters
- high densities
- analytical theory
- comp. cost \sim domain size



@fotolia - VRD

Macroscopic variables

Along a road (network) we need to measure aggregate quantities:

- the (mean) **traffic density** ρ : number of vehicles per unit space
- the (mean) **velocity** v : distance covered by vehicles per unit time
- the **traffic flow** $q = \rho v$: number of vehicles per unit time

Data sources:



magnetic loop detectors, video recordings, floating car data, etc



Loop detector data

Example:

355#M3E;	2;	255;	Ma;	01/09/15;	00:00;	1240;	C;	109;	10;
355#M3E;	2;	255;	Ma;	01/09/15;	00:00;	1407;	D;	96;	54;
305#M3e;	2;	255;	Ma;	01/09/15;	00:00;	1449;	D;	100;	10;
709#M7i;	264;	687;	Ma;	01/09/15;	00:00;	1874;	D;	104;	37;
709#M7i;	264;	687;	Ma;	01/09/15;	00:00;	2248;	D;	83;	36;
709#M7i;	264;	687;	Ma;	01/09/15;	00:00;	2368;	D;	78;	39;
577#M6C;	16;	474;	Ma;	01/09/15;	00:00;	4237;	D;	120;	40;
:	:	:	:	:	:	:	:	:	:

where:

- loop code
- road milestone
- distance from the milestone (m)
- day of the week
- date
- time (hours:minutes)
- seconds and hundredth
- lane
- vehicle speed (km/h)
- vehicle length (dm)

Floating Car Data

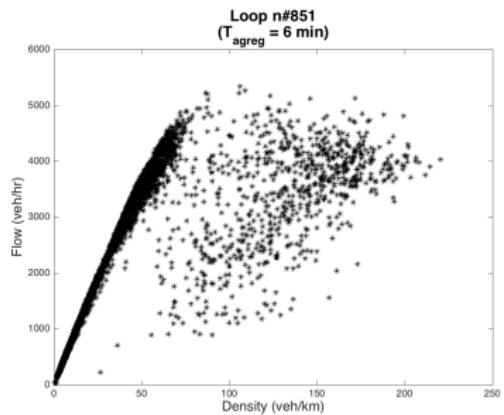
Example:

AVE34e...	2013-03-19T06:09:32	110	43.5452	6.94092	38	A8-Mandelieu-StIsidore	1
COY8c6...	2013-03-19T07:06:58	128	43.5454	6.94102	36	A8-Mandelieu-StIsidore	1
COYf7...	2013-03-19T07:09:20	121	43.5449	6.94064	36	A8-Mandelieu-StIsidore	1
COY3cd...	2013-03-19T07:10:50	124	43.5446	6.94014	36	A8-Mandelieu-StIsidore	1
AVE61d...	2013-03-19T07:16:30	123	43.5448	6.94056	39	A8-Mandelieu-StIsidore	1
AVE61d	2013-03-19T07:16:33	125	43.5455	6.94135	36	A8-Mandelieu-StIsidore	1
COYbdb	2013-03-19T07:23:30	124	43.5448	6.94045	36	A8-Mandelieu-StIsidore	1
COY2ba	2013-03-19T07:23:44	105	43.5458	6.94154	36	A8-Mandelieu-StIsidore	1
:	:	:	:	:	:	:	⋮

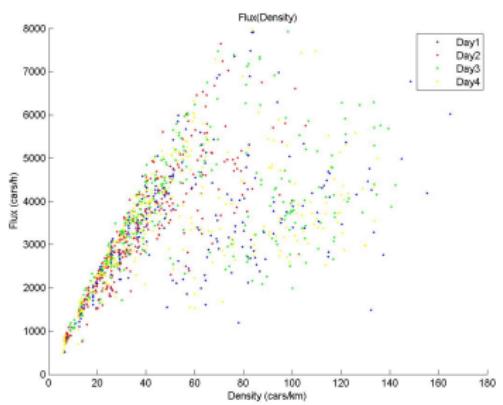
where:

- device code
- date and time
- speed
- GPS position and orientation
- other location info

Typical fundamental diagram



loops



CFD

Macroscopic traffic flow models

$$\left[\text{number of vehicles in } [a, b] \text{ at time } t \right] = \int_a^b \rho(t, x) \, dx$$

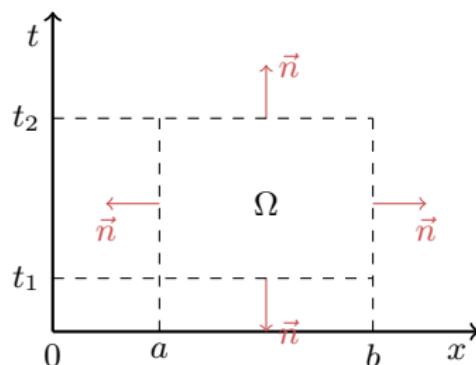
must be conserved!

$$\begin{aligned} \int_a^b \rho(t_2, x) dx &= \int_a^b \rho(t_1, x) dx \\ &+ \int_{t_1}^{t_2} q(t, a) dt - \int_{t_1}^{t_2} q(t, b) dt \end{aligned}$$



divergence theorem

$$\begin{aligned} \int_{\Omega} \operatorname{div}_{(t,x)}(\rho, q) &= \int_{\partial\Omega} (\rho, q) \cdot \vec{n} \\ &\Downarrow \\ \int_{t_1}^{t_2} \int_a^b (\partial_t \rho + \partial_x q) \, dx \, dt &= 0 \end{aligned}$$



conservation law

Basic requirements

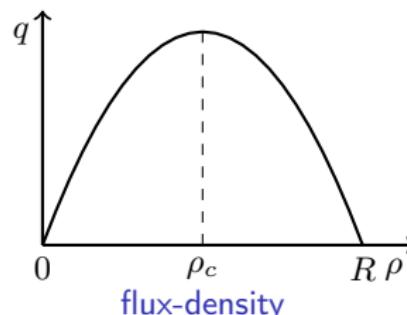
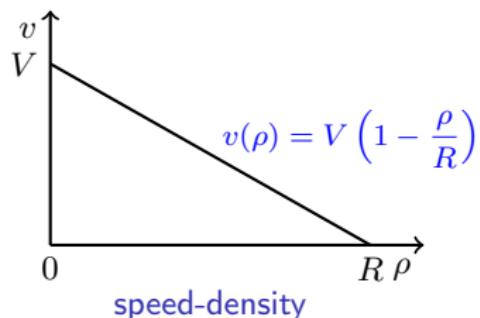
$$\partial_t \rho(t, x) + \partial_x q(t, x) = 0$$

- No information propagates faster than vehicles (anisotropy)
- Fundamental relation: $q(t, x) = \rho(t, x)v(t, x)$.
- Density and mean velocity must be non-negative and bounded:
 $0 \leq \rho(t, x), v(t, x) < +\infty, \forall x, t > 0$.
- Different from fluid dynamics:
 - preferred direction
 - no conservation of momentum / energy
 - no viscosity
 - continuum assumption?

The Lighthill-Whitham-Richards (LWR) model (mid 50s)

- Mass conservation equation: $\partial_t \rho + \partial_x (\rho v) = 0$
- Phenomenological speed-density relation: $v(t, x) = v(\rho(t, x))$

fundamental diagram



V maximal speed, R maximal or *jam* density, ρ_c critical density:

- flux is increasing for $\rho \leq \rho_c$: free-flow phase
- flux is decreasing for $\rho \geq \rho_c$: congestion phase

The Lighthill-Whitham-Richards (LWR) model (mid 50s)

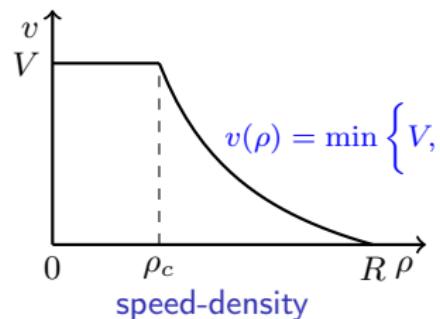
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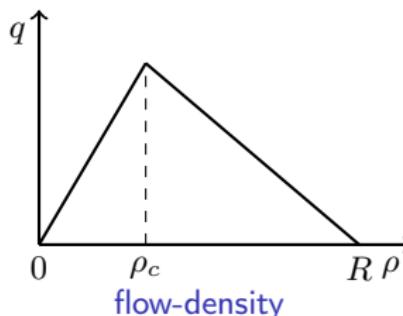
- Phenomenological speed-density relation:

$$v(t, x) = v(\rho(t, x))$$

fundamental diagram



$$v(\rho) = \min \left\{ V, w \left(\frac{R}{\rho} - 1 \right) \right\}$$

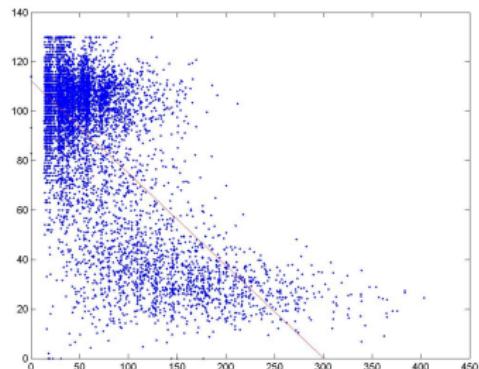


V maximal speed, R maximal or *jam* density, ρ_c critical density:

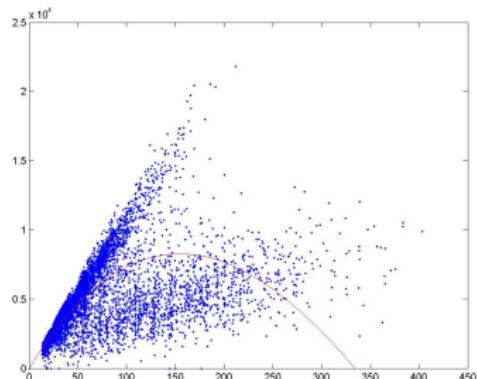
- flux is increasing for $\rho \leq \rho_c$: **free-flow phase**
- flux is decreasing for $\rho \geq \rho_c$: **congestion phase**

$$v = v(\rho)?$$

Experimental data:



speed-density



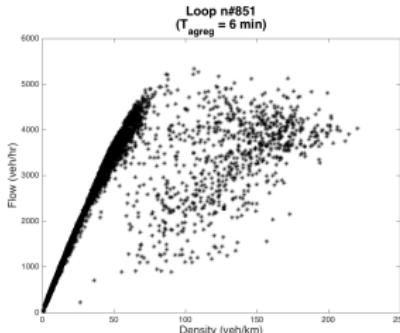
flow-density

A8 (Antibes-Nice), March 2013

Improved models

- Additional equation for v : second order models

[Payne (1971); Helbing (1996); Aw-Rascle (2000);
Zhang (2002); Colombo (2002); Lebacque-et-al (2005); ...]



- Phase transition

[Colombo (2002); Goatin (2006); Blandin-et-al (2011); ...]

- Stochastic and averaged models

[Boel-Mihaylova (2006); Wang-Ni-Chen-Li (2010); Chen-Wang (2011); Sumalee&al (2011); Jabari-Liu (2011); ...]

- Multi-population

[Wong-Wong (2002); Benzoni-Colombo (2003); VanLint-Hoogendorn-Schreurer (2008); Nair-Mahmassani-Hooks (2011); Fan-Work (2015); Gashaw-Goatin-Harri (2016); ...]

- Non-local models

[Blandin-Goatin (2016); Friedrich-Kolb-Göttlich (2018)]

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Moving bottleneck¹

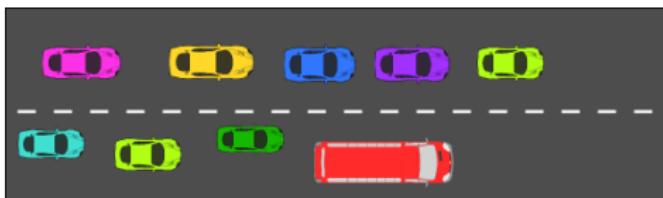
PDE-ODE model for interaction between a vehicle and surrounding traffic

$$\begin{cases} \partial_t \rho + \partial_x(\rho v(\rho)) = 0, & x \in \mathbb{R}, t > 0, \\ \rho(0, x) = \rho_0(x), & x \in \mathbb{R}, \\ f(\rho(t, y(t))) - \dot{y}(t)\rho(t, y(t)) \leq F_\alpha(\dot{y}(t)), & t > 0, \\ \dot{y}(t) = \min\{u(t), v(\rho(t, y(t)+))\}, & t > 0, \\ y(0) = y_0. \end{cases}$$

where $y = y(t) \in \mathbb{R}$ vehicle trajectory

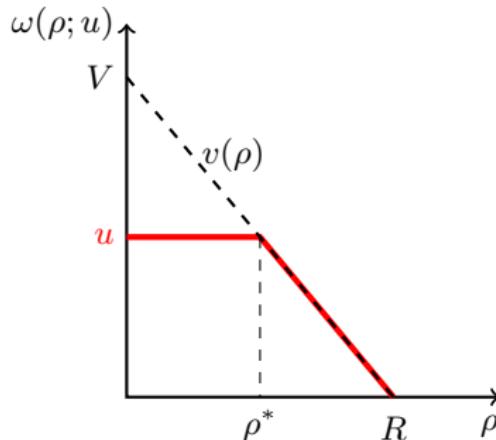
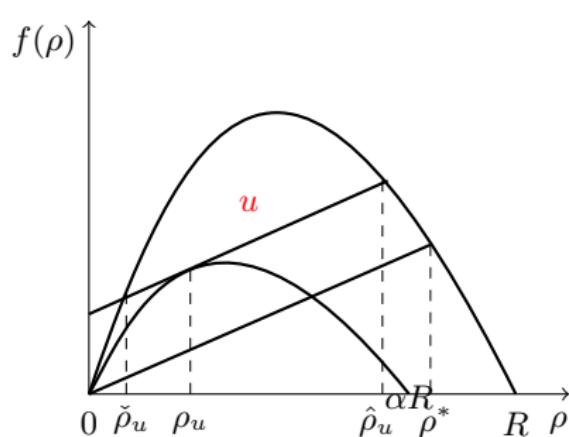
$\alpha \in [0, 1]$ road capacity reduction rate

$u(t) \in [0, V]$ AV desired velocity (control)



¹[Lebacque-Lesort-Giorgi, TRR, 1998; DelleMonache-Goatin, JDE, 2014;
Garavello-Goatin-Piccoli-Liard, JDE, 2020]

The model²



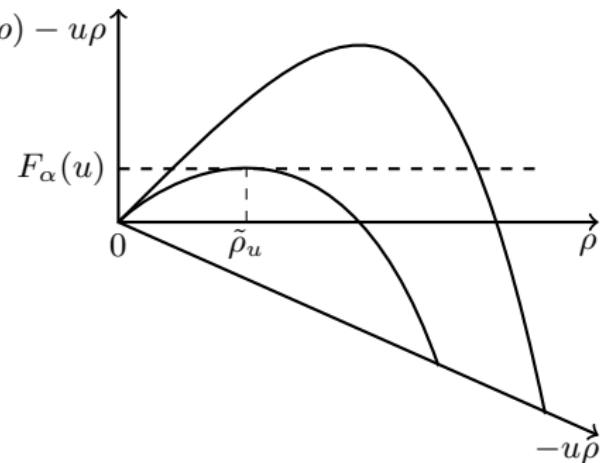
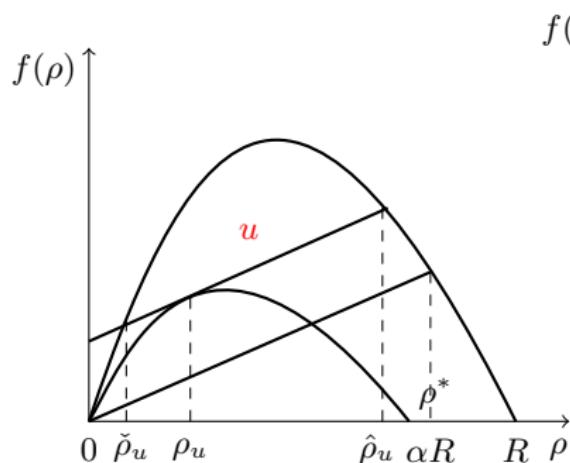
- $v(\rho) = V(1 - \frac{\rho}{R})$ mean traffic velocity
- $f(\rho) = \rho v(\rho)$ flux function, strictly concave
- $\omega(\rho; u) = \begin{cases} u & \text{if } \rho \leq \rho^* \doteq R(1 - u/V) \\ v(\rho) & \text{if } \rho > \rho^* \end{cases}$ AV speed

²[DelleMonache-Goatin, JDE, 2014]

The Riemann Problem

In the vehicle reference frame³ $X = x - ut$ (u constant)

$$\begin{cases} \partial_t \rho + \partial_X (f(\rho) - u\rho) = 0, \\ \rho(0, x) = \begin{cases} \rho_L & \text{if } X < 0, \\ \rho_R & \text{if } X > 0, \end{cases} \\ f(\rho(t, 0)) - u\rho(t, 0) \leq f_\alpha(\rho_u) - u\rho_u = \frac{\alpha R}{4V}(V - u)^2 \doteq F_\alpha(u) \end{cases}$$



³[Colombo-Goatin, JDE, 2007]

The Riemann Problem

Definition (DelleMonache-Goatin, JDE 2014)

- If $f(\mathcal{R}(\rho_L, \rho_R)(\textcolor{red}{u})) \leq F_\alpha + \textcolor{red}{u}\mathcal{R}(\rho_L, \rho_R)(\textcolor{red}{u})$, then

$$\mathcal{R}^\alpha(\rho_L, \rho_R) = \mathcal{R}(\rho_L, \rho_R) \quad \text{and} \quad y(t) = \omega(\rho; \textcolor{red}{u})t.$$

- If $f(\mathcal{R}(\rho_L, \rho_R)(\textcolor{red}{u})) > F_\alpha + \textcolor{red}{u}\mathcal{R}(\rho_L, \rho_R)(\textcolor{red}{u})$, then

$$\mathcal{R}^\alpha(\rho_L, \rho_R)(x) = \begin{cases} \mathcal{R}(\rho_L, \hat{\rho}_{\textcolor{blue}{u}}) & \text{if } x < \textcolor{red}{u}t, \\ \mathcal{R}(\check{\rho}_{\textcolor{blue}{u}}, \rho_R) & \text{if } x \geq \textcolor{red}{u}t, \end{cases} \quad \text{and} \quad y(t) = \textcolor{red}{u}t.$$

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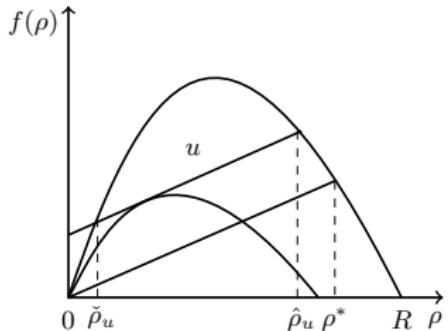
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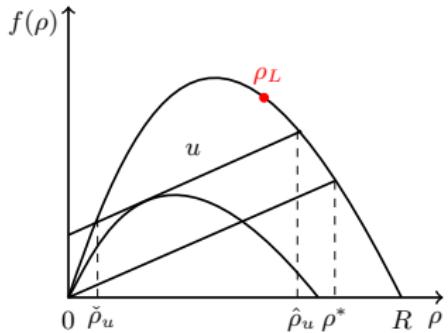
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$\implies (\hat{\rho}_u, \check{\rho}_u)$ non-classical shock traveling with speed $\textcolor{red}{u}$

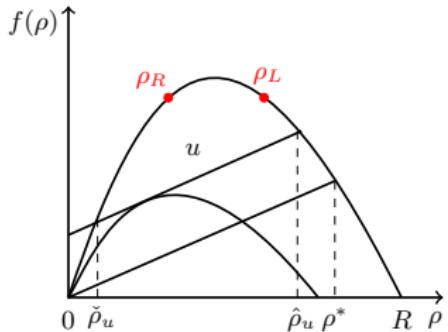
The Riemann Problem: example



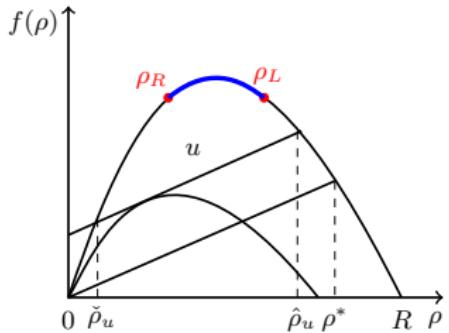
The Riemann Problem: example



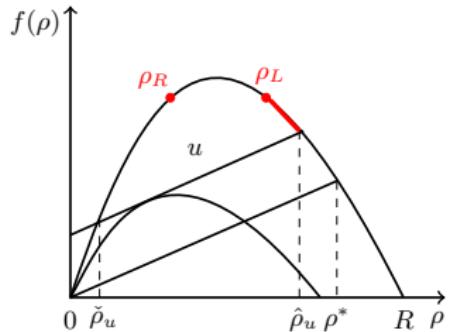
The Riemann Problem: example



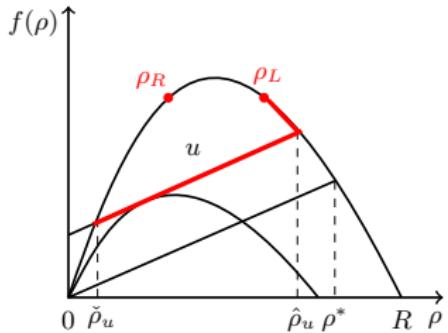
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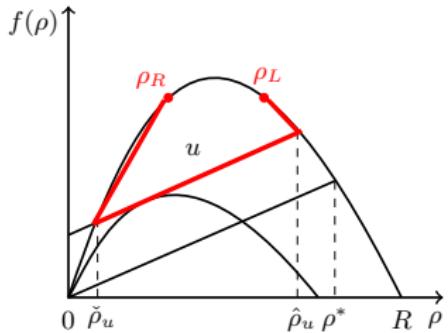
The Riemann Problem: example



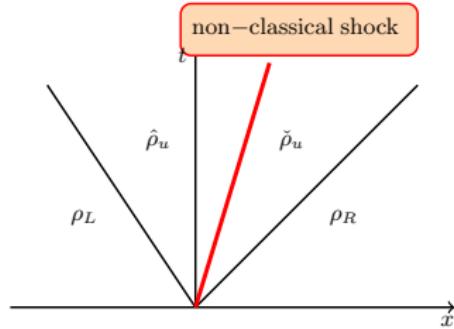
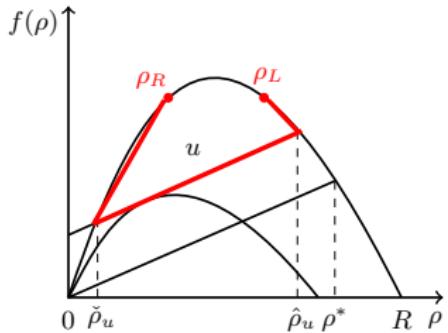
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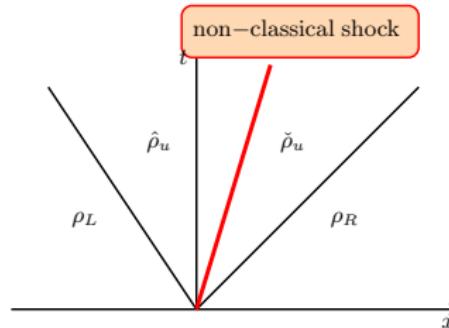
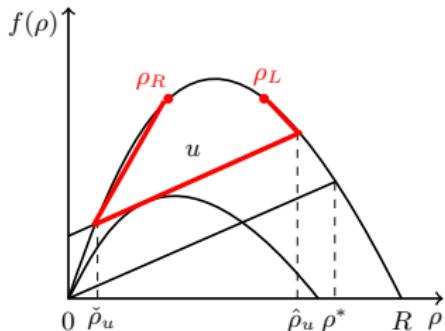
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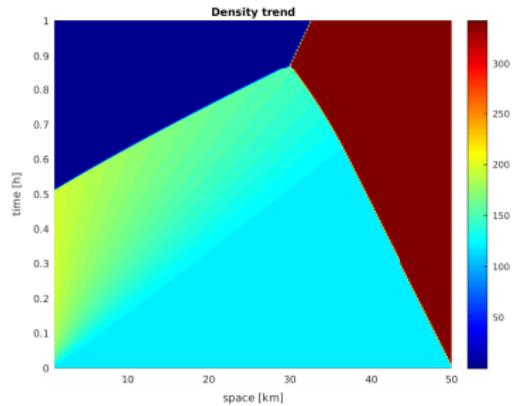
The non-classical shock $(\hat{\rho}_u, \check{\rho}_u)$ satisfies the Rankine-Hugoniot condition but violates the Lax entropy condition

$$f'(\hat{\rho}_u) < u < f'(\check{\rho}_u)$$

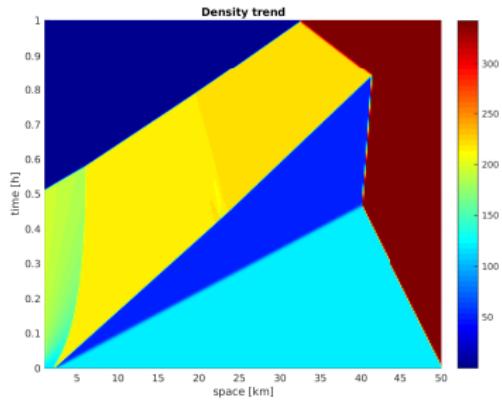
Example

$V = 140 \text{ km/h}$, $u = 80 \text{ km/h}$, $R = 400 \text{ vehicles/km}$, $\alpha = 0.6$:

Optimal control (MPC) [Piacentini-Goatin-Ferrara, IFAC CTS, 2018]



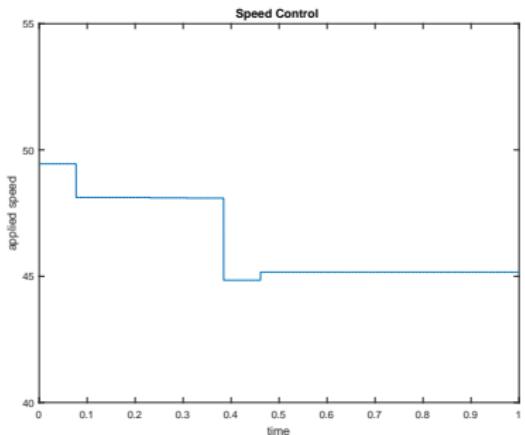
without control



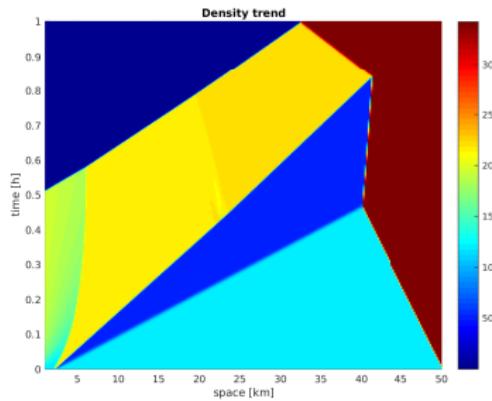
with control

	ATT [h]	jam [km]	TFC [liters]	TFC reduction %
Uncontrolled	0.9107	10.18	$2.7413 \cdot 10^4$	0
Controlled	0.8579	7.66	$2.6852 \cdot 10^4$	2.05

Optimal control (MPC) [Piacentini-Goatin-Ferrara, IFAC CTS, 2018]



speed control law



with control

	ATT [h]	jam [km]	TFC [liters]	TFC reduction %
Uncontrolled	0.9107	10.18	$2.7413 \cdot 10^4$	0
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Model Predictive Control (MPC)

- Time $t = 0$: a prediction of the future evolution of the system is done over the prediction horizon $\Delta\tau = 15\text{min}$
- The **cost** is evaluated on the prediction as a function of the control variable and minimized wrt the controlled vehicle speed $\textcolor{red}{u}$
- The optimal value is applied for a time interval $\Delta t = 5\text{min} < \Delta\tau$
- At $t + \Delta t$ another prediction is done by considering the new situation and so on.

Some references

- R. Borsche, R.M. Colombo, M. Garavello. [Mixed systems: ODEs – Balance laws](#), JDE (2012).
- C. Chalons, M.L. Delle Monache, P. Goatin. [A conservative scheme for non-classical solutions to a strongly coupled PDE-ODE problem](#), Interfaces and Free Boundaries (2017).
- M. Garavello, P. Goatin, T. Liard, B. Piccoli. [A multiscale model for traffic regulation via autonomous vehicles](#), JDE, to appear.
- C. Lattanzio, A. Maurizi, B. Piccoli. [Moving bottlenecks in car traffic flow: A PDE-ODE coupled model](#), SIAM J. Math. Analysis (2011).
- G. Piacentini, P. Goatin, A. Ferrara. [A macroscopic model for platooning in highway traffic](#), SIAM J. Appl. Math. (2020).
- S. Villa, P. Goatin, C. Chalons. [Moving bottlenecks for the Aw-Rascle-Zhang traffic flow model](#), DCDS B (2017).

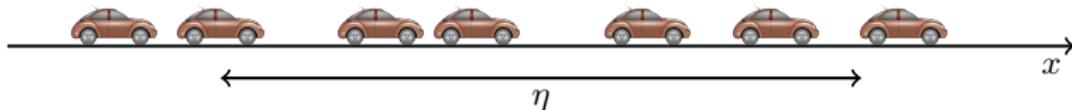
A model with non-local velocity⁴

LWR model with downstream non-local velocity

$$\partial_t \rho(t, x) + \partial_x (\rho(t, x) V(t, x)) = 0$$

where

$$V(t, x) = v \left(\int_x^{x+\eta} \rho(t, y) w_\eta(y - x) dy \right), \quad \eta > 0$$



with $w_\eta \in \mathbf{C}^1([0, \eta]; \mathbb{R}^+)$ $w'_\eta \leq 0$ and $\int_0^\eta w_\eta(x) dx = 1$

$v : [0, \rho_{\max}] \rightarrow \mathbb{R}^+$ s.t. $-A \leq v' \leq 0$, $v(0) = v_{\max}$, $v(\rho_{\max}) = v_{\min}$

⁴[Blandin-Goatin, NumMath 2016; Goatin-Scialanga, NNM 2016]

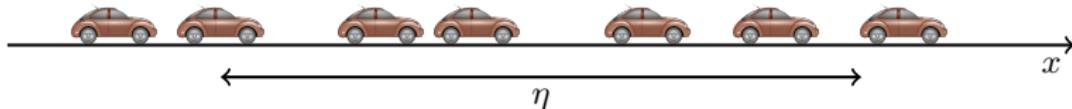
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[Friedrich-Kolb-Göttlich, NHM 2018]: weighted average velocity

$$V(t, x) = \int_x^{x+\eta} v(\rho(t, y)) w_\eta(y - x) dy, \quad \eta > 0$$

(better suited on networks)

⁴[Blandin-Goatin, NumMath 2016; Goatin-Scialanga, NHM 2016]

Well-posedness

Theorem

[Blandin-Goatin, NumMath 2016; Goatin-Scialanga, NHM 2016; Chiarello-Goatin, M2AN 2018]

Let $\rho_0 \in \text{BV}(\mathbb{R}; [0, \rho_{\max}])$. Then the Cauchy problem

$$\begin{cases} \partial_t \rho + \partial_x (\rho V(t, x)) = 0 & x \in \mathbb{R}, t > 0 \\ \rho(0, x) = \rho_0(x) & x \in \mathbb{R} \end{cases}$$

admits a unique weak (entropy) solution ($\rho \in \mathbf{L}^1 \cap \mathbf{L}^\infty \cap \text{BV}$), such that

$$\min_{\mathbb{R}} \{\rho_0\} \leq \rho(t, x) \leq \max_{\mathbb{R}} \{\rho_0\} \quad \text{for a.e. } x \in \mathbb{R}, t > 0$$

Well-posedness

Theorem

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Remark

Space-time dependent flux function

$$f = f(t, x, \rho) = \rho V(t, x)$$

Dependence on the location of the kernel support

We set $v(\rho) = 1 - \rho$ and

downstream: $V_d(t, x) = 1 - \int_x^{x+\eta} \rho(t, y) w_\eta(y - x) dy$

center: $V_c(t, x) = 1 - \int_{x-\eta/2}^{x+\eta/2} \rho(t, y) w_\eta(y - x) dy$

upstream : $V_u(t, x) = 1 - \int_{x-\eta}^x \rho(t, y) w_\eta(y - x) dy$

Dependence on the location of the kernel support

Rarefaction

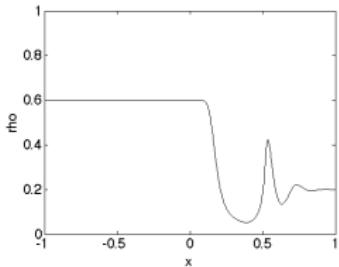
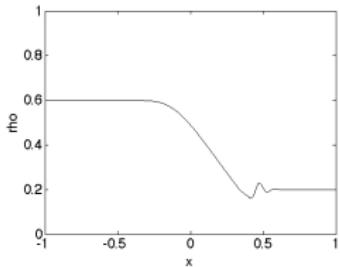
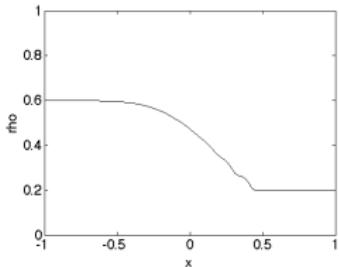
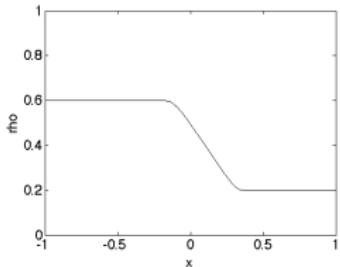


Figure: $w_\eta(x) = 1/\eta$ with downstream, central and upstream supports respectively and initial data $\rho_L = 0.6$, $\rho_R = 0.2$

Dependence on the kernel support

Shock

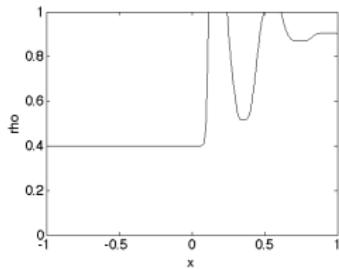
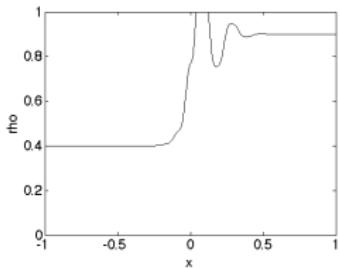
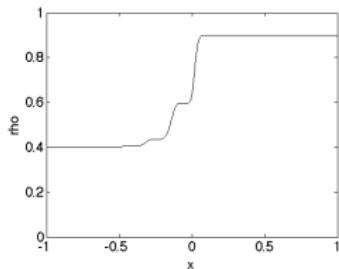
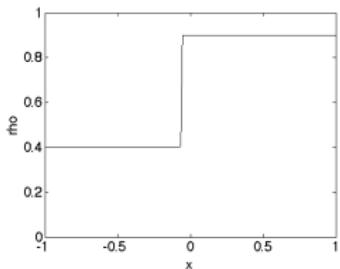


Figure: $w_\eta(x) = 1/\eta$ with downstream, central and upstream supports respectively and initial data $\rho_L = 0.4$, $\rho_R = 0.9$

Dependence on the kernel support

Oscillating initial datum

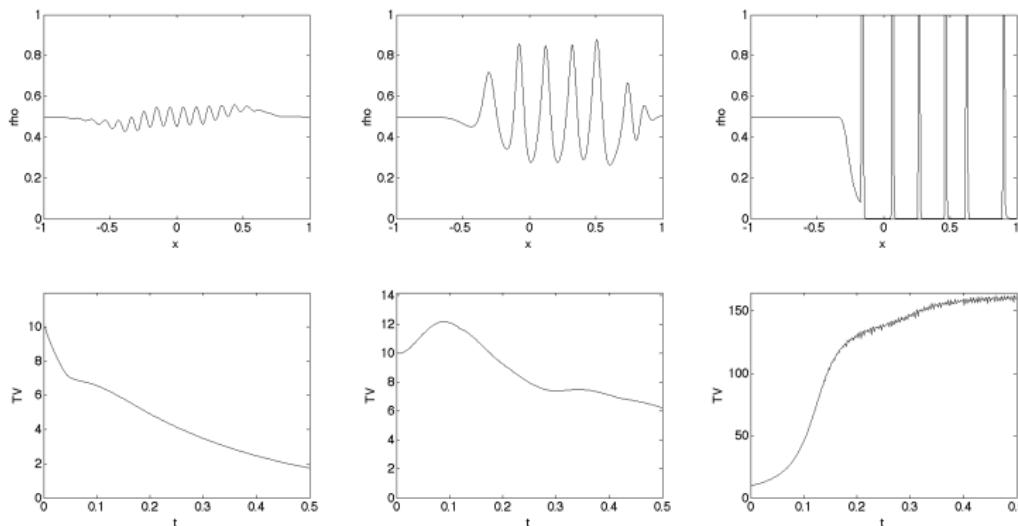


Figure: $w_\eta(x) = 1/\eta$ with downstream, central and upstream supports respectively

Kernel monotonicity

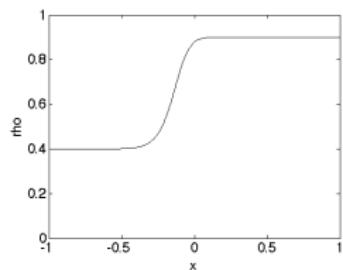
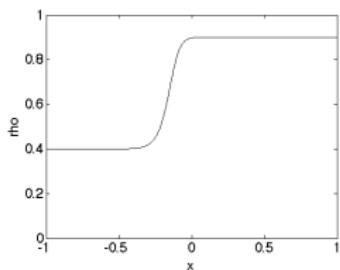
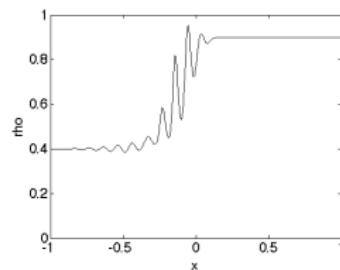
(a) $w_\eta(x) = 1/\eta$ (b) $w_\eta(x) = 2(\eta - x)/\eta^2$ (c) $w_\eta(x) = 2x/\eta^2$

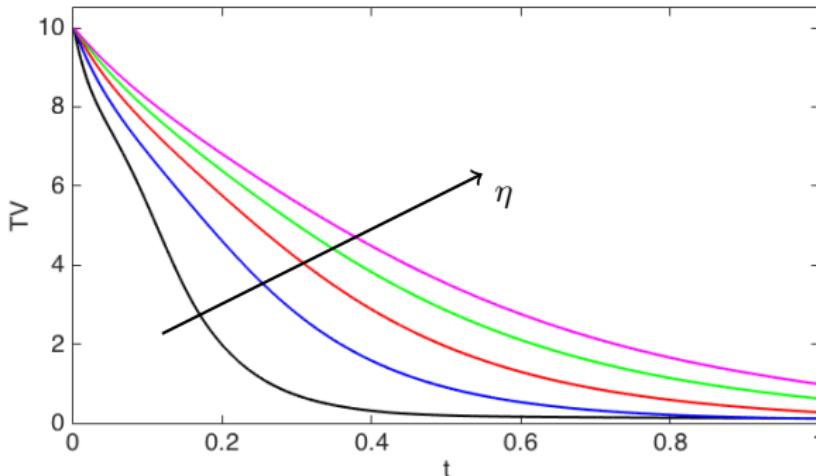
Figure: $\rho(t = 0.5, \cdot)$ corresponding to $\rho_L = 0.4$, $\rho_R = 0.9$

What this non-local model teaches us

- You should always look forward
- You should care more what is closer

What this non-local model teaches us

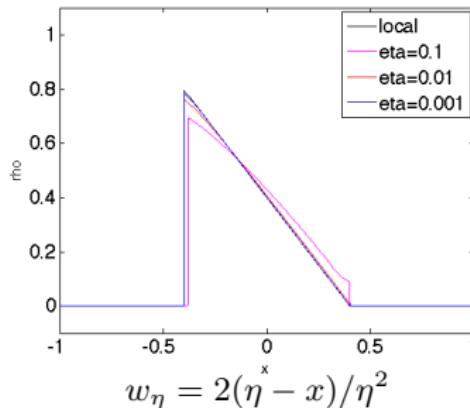
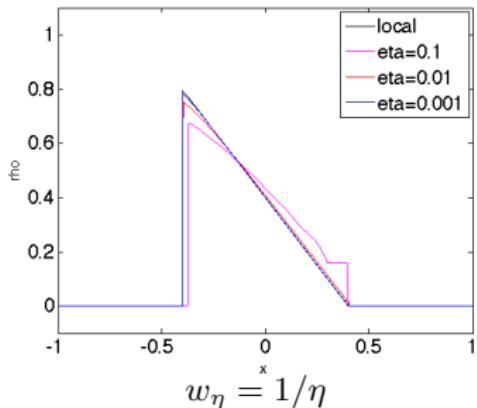
- You should always look forward
- You should care more what is closer
- The farther you see, the worse it is



Limit $\eta \searrow 0^5$

$$\partial_t \rho + \partial_x (\rho v(\rho * w_\eta)) = 0 \quad \rightarrow \quad \partial_t \rho + \partial_x (\rho v(\rho)) = 0 \quad ??$$

We consider $v(\rho) = 1 - \rho$ and $\rho_0(x) = \begin{cases} 0.8 & \text{if } -0.5 < x < -0.1 \\ 0 & \text{otherwise} \end{cases}$

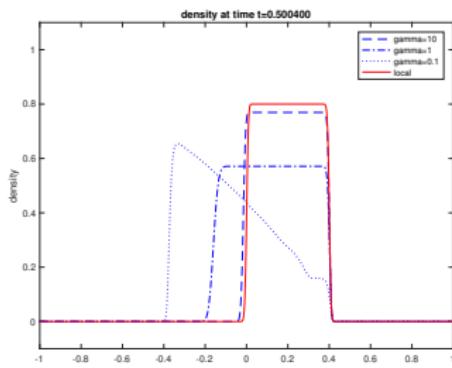


⁵[Colombo-Crippa-Spinolo, 2018; Colombo-Crippa-Marconi-Spinolo, 2019]

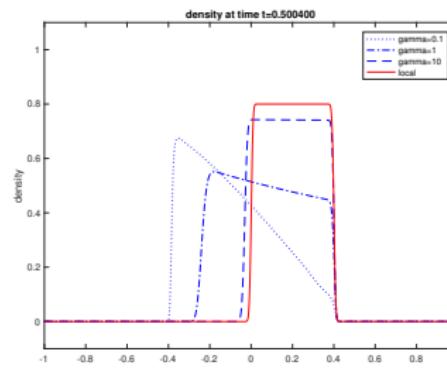
Limit $\eta \rightarrow +\infty$ ⁶

$$\partial_t \rho + \partial_x (\rho v(\rho * w_\eta)) = 0 \quad \rightarrow \quad \partial_t \rho + \partial_x (\rho v(0)) = 0$$

We consider $v(\rho) = 1 - \rho$ and $\rho_0(x) = \begin{cases} 0.8 & \text{if } -0.5 < x < -0.1 \\ 0 & \text{otherwise} \end{cases}$



$$w_\eta = 1/\eta$$



$$w_\eta = 2(\eta - x)/\eta^2$$

⁶ [Chiarello-Goatin, M2AN 2018]

A multi-class model with non-local velocity⁷

Multi-class traffic model with downstream non-local velocity

$$\partial_t \rho_i(t, x) + \partial_x (\rho_i(t, x) v_i((r * \omega_i)(t, x))) = 0, \quad i = 1, \dots, M,$$

where

$$r(t, x) := \sum_{i=1}^M \rho_i(t, x), \quad v_i(\xi) := v_i^{\max} \psi(\xi),$$

$$(r * \omega_i)(t, x) := \int_x^{x+\eta_i} r(t, y) \omega_i(y - x) dy,$$

$$\omega_i \in \mathbf{C}^1([0, \eta_i]; \mathbb{R}^+), \quad \omega'_i \leq 0, \quad \int_0^{\eta_i} \omega_i(y) dy = J_i.$$

$$(\mathbf{H}) \quad W_0 := \max_{i=1, \dots, M} \omega_i(0). \quad 0 < v_1^{\max} \leq v_2^{\max} \leq \dots \leq v_M^{\max}.$$

$$\psi : \mathbb{R}^+ \rightarrow \mathbb{R}^+ \text{ smooth} \quad \psi' \leq 0 \text{ s.t. } \psi(0) = 1 \text{ and } \psi(r) = 0 \text{ for } r \geq 1.$$

- Local multi-class model: [Benzoni-Colombo, EJAM 2003]

⁷[Chiarello-Goatin, NHM 2019]

AV and human-driven mixed traffic

Circular road modeled by the space interval $[-1, 1]$ with periodic boundary conditions at $x = \pm 1$.

$$\begin{cases} \partial_t \rho_1(t, x) + \partial_x (\rho_1(t, x) v_1^{\max} \psi((r * \omega_1)(t, x))) = 0, & \text{autonomous}, \\ \partial_t \rho_2(t, x) + \partial_x (\rho_2(t, x) v_2^{\max} \psi((r * \omega_2)(t, x))) = 0, & \text{human - driven} \\ \rho_1(0, x) = \beta (0.5 + 0.3 \sin(5\pi x)), \\ \rho_2(0, x) = (1 - \beta) (0.5 + 0.3 \sin(5\pi x)), \end{cases}$$

with

$$\omega_1(x) = \frac{1}{\eta_1}, \quad \eta_1 = 1,$$

$$\omega_2(x) = \frac{2}{\eta_2} \left(1 - \frac{x}{\eta_2} \right), \quad \eta_2 = 0.01,$$

$$\psi(\xi) = \max \{1 - \xi, 0\}, \quad \xi \geq 0,$$

$$v_1^{\max} = v_2^{\max} = 1.$$

$\beta \in [0, 1]$ AV penetration rate

AV and human-drive mixed traffic

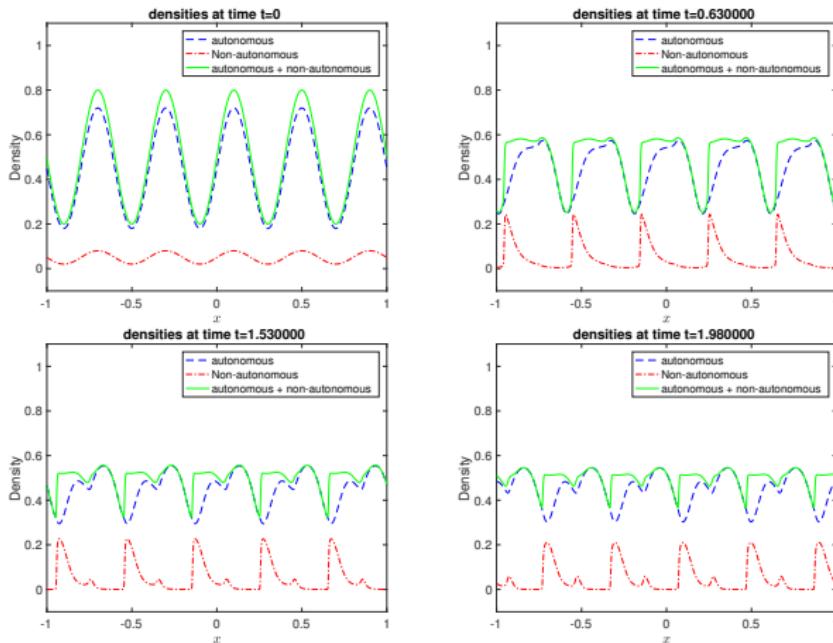


Figure: Density profiles corresponding to $\beta = 0.9$ at different times.

Impact of connected autonomous vehicles⁸

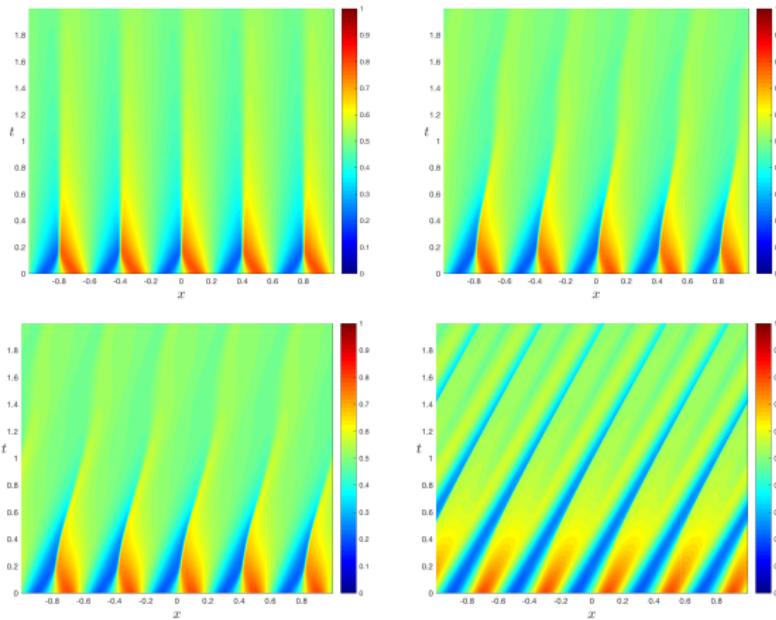


Figure: (t, x) -plots of the total traffic density $r(t, x)$ corresponding to $\beta = 0, 0.2, 0.3, 0.9$.

⁸[Chiarello-Goatin, NHM 2019]

Related results

- Numerical schemes

- First order
 - Lax-Friedrichs
[Colombo-Amorim-Texeira, ESAIM M2AS 2015; Blandin-Goatin, NumMath 2016; Aggarwal-Colombo-Goatin, SINUM 2015; ...]
 - Upwind
[Friedrich-Kolb-Göttlich, NHM 2018; Chiarello-Goatin, NHM 2019]
 - Lagrangian-Remap
[Chiarello-Goatin-Villada, Comput. Appl. Math. 2020]
- High order
 - Discontinuous Galerkin and Finite Volume WENO
[Chalons-Goatin-Villada, SISC 2018; Chiarello-Goatin-Villada, HYP2018 Proceedings (2020)]

- Extension to networks

[Chiarello-Friedrichs-Goatin-Göttlich-Kolb, European J. Appl. Math. (2019)]

- Micro-macro limits

[Goatin-Rossi, CMS 2017; Chiarello-Friedrich-Goatin-Göttlich, SIAM J. Appl. Math.]