

Analysis Motivated by Vehicular Traffic and Crowd Dynamics

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Benasque – August 22nd, 2017

Conservation Laws

Introduction

Vehicular Traffic

Macroscopic Models

Braess Paradox

Crowd Dynamics

Modeling Crowd

Controlling Crowd

Predators – Prey

Introduction – Analytic Theory

$$\partial_t u + \operatorname{div}_x f(t, x, u) = g(t, x, u)$$

$t \in \mathbb{R}_+$ time
 $x \in \mathbb{R}^N$ space
 $u \in \mathbb{R}^n$ unknown

f smooth flux
 g smooth source

Euler	Statement	1755
Riemann	Regular Solutions	1860

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Scalar MultiD $n = 1, N \geq 1$

$$\partial_t u + \operatorname{div}_x f(t, x, u) = g(t, x, u)$$

Existence

(Kružkov: Mat.Sb., 1970)

Uniqueness

(Kružkov: Mat.Sb., 1970)

Dependence on data

(Kružkov: Mat.Sb., 1970)

Dependence on f, g

(Colombo, Mercier, Rosini: CMS, 2009)

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Systems in 1D $n \geq 1, N = 1$

$$\partial_t u + \partial_x f(u) = 0$$

Existence

(Glimm: CPAM, 1965)

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(Bressan & c.: 1999, 2000)

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$$n \geq 1, N \geq 1?$$

Introduction – Key Features

1. Evolution
2. Irreversible
3. Finite Speed
4. Conservation
5. Singularities

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Simplest Case

$$n = 1, N = 1,$$
$$f = f(u), g \equiv 0$$

$$\begin{cases} \partial_t u + \partial_x f(u) = 0 \\ u(0, x) = u_o(x) \end{cases}$$

Introduction – Key Features

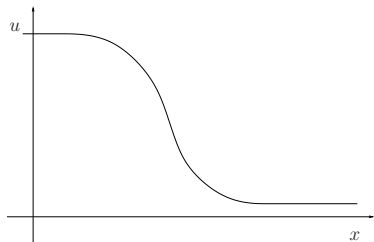
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$$f(u) = \lambda u \quad \partial_t u + \lambda \partial_x u = 0 \quad u(t, x) = u_o(x - \lambda t)$$



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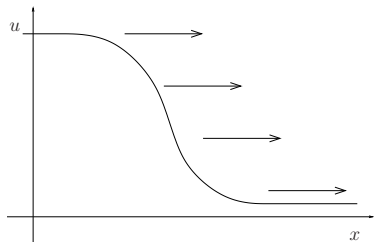
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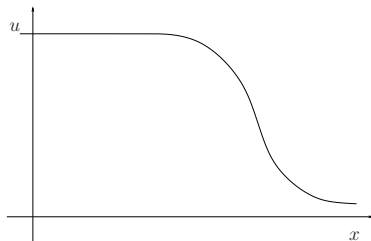
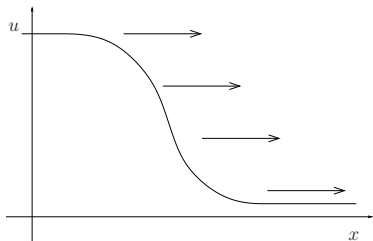
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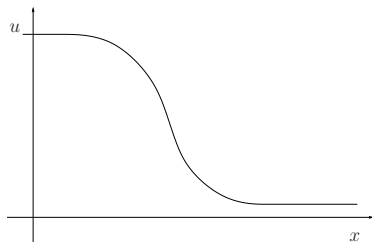
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f non linear

$$\partial_t u + f'(u) \partial_x u = 0$$

$$u(t, x) = ?$$



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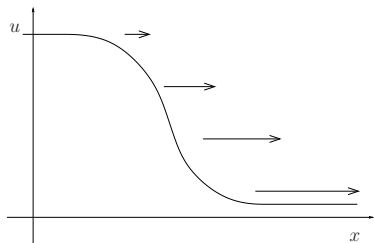
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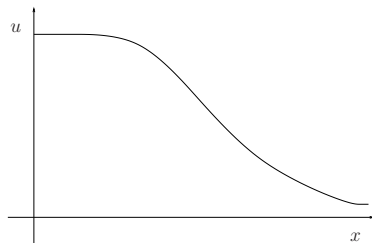
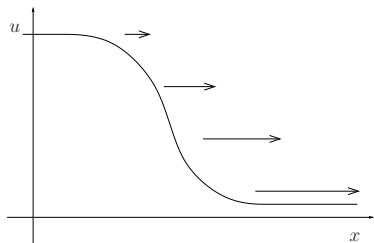
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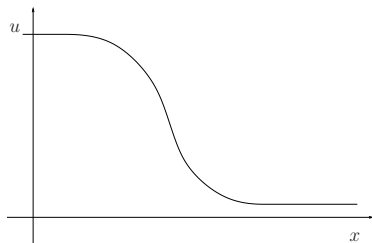
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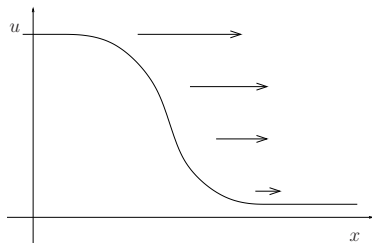
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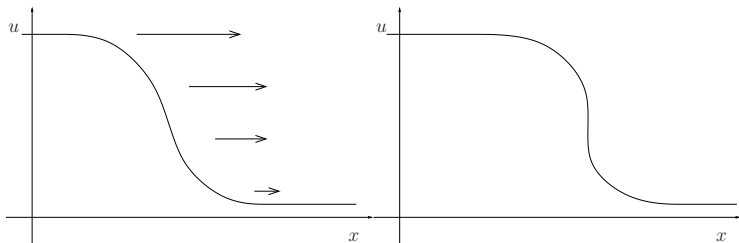
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Introduction

Discontinuities!

Discontinuities!

Viscosity

Entropy

Stability

Discontinuities!

Viscosity

Entropy

Stability

They all agree!

Vehicular Traffic

Vehicular Traffic – Macroscopic Models

t = time
 x = space

$\rho = \left\{ \begin{array}{l} \text{(density)} \\ \text{occupancy} \end{array} \right.$

cars are conserved

Vehicular Traffic – Macroscopic Models

t = time
 x = space

$\rho = \begin{cases} \text{(density)} \\ \text{occupancy} \end{cases}$

$$\partial_t \rho + \partial_x (\rho v) = 0$$

$$v = ?$$

Vehicular Traffic – Macroscopic Models

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$$v = ?$$

LWR

(Lighthill, Whitham: Proc. London. A., 1955)

(Richards: Operations Res., 1956)

v decreasing

$$v(0) = v_{\max}$$

$$v(R) = 0$$

Vehicular Traffic – Macroscopic Models

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► Second Order Models

- *Requiem* (Daganzo: Transp. Research B, 1995)
- *Resurrection* (Aw, Rascle: SIAM Appl. Math., 2000)
- (Zhang: Transp. Research B, 2002)

Vehicular Traffic – Macroscopic Models

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▶ Second Order Models

▶ 2-Phase Models

- ▶ (Colombo: SIAM Appl. Math., 2002)
- ▶ (Colombo, Marcellini, Rascle: SIAM Appl. Math., 2010)
- ▶ (Blandin, Work, Goatin, Piccoli, Bayen: SIAM Appl. Math., 2010)

Vehicular Traffic – Macroscopic Models

t = time
 x = space

$\rho = \begin{cases} \text{(density)} \\ \text{occupancy} \end{cases}$

$$\partial_t \rho + \partial_x (\rho v) = 0$$

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- ▶ Second Order Models
- ▶ 2-Phase Models
- ▶ Multi-Population Models
 - ▶ (Zhang, Jin: Transp. Research Rec., 2002)
 - ▶ (Benzoni-Gavage, Colombo: EJAM, 2003)

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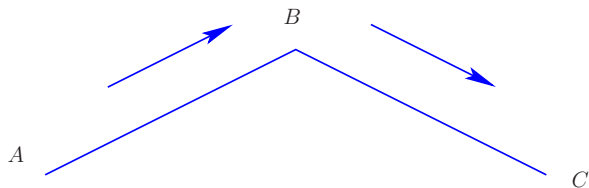
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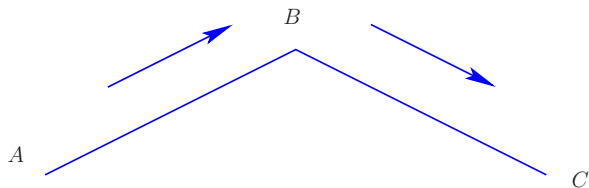
- ▶ Second Order Models
- ▶ 2-Phase Models
- ▶ Multi-Population Models
- ▶ **Networks**
 - ▶ (Garavello, Kahn, Piccoli: Book, 2016)

Vehicular Traffic



From	To	Time
A	B	$\frac{\# \text{ cars}}{100}$
B	C	45

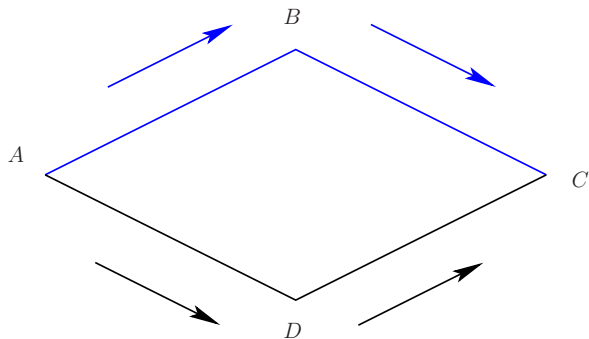
Vehicular Traffic



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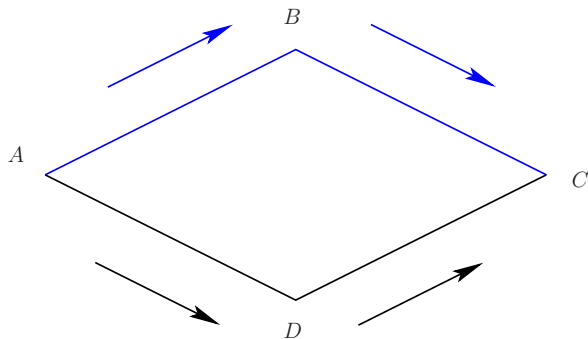
Route ABC : 4000 cars need $\frac{4000}{100} + 45 = 85$

Vehicular Traffic



From	To	Time
A	B	$\frac{\# \text{ cars}}{100}$
B	C	45
A	D	45
D	B	$\frac{\# \text{ cars}}{100}$

Vehicular Traffic

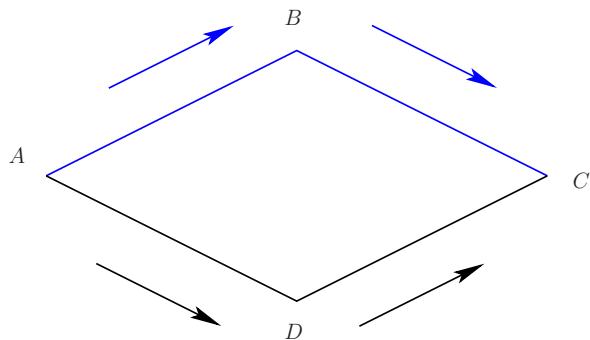


From	To	Time
A	B	$\frac{\# \text{ cars}}{100}$
B	C	45
A	D	45
D	B	$\frac{\# \text{ cars}}{100}$

$$\text{Route } ABC: \frac{\# \text{ cars}}{100} + 45$$

$$\text{Route } ADC: 45 + \frac{\# \text{ cars}}{100}$$

Vehicular Traffic

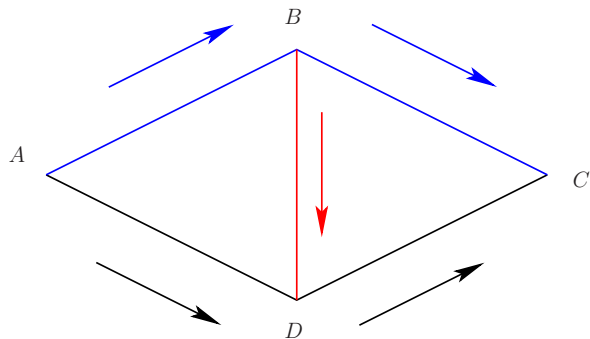


From	To	Time
A	B	$\frac{\# \text{ cars}}{100}$
B	C	45
A	D	45
D	B	$\frac{\# \text{ cars}}{100}$

$$\text{Route } ABC: \frac{\# \text{ cars}}{100} + 45 \Rightarrow \frac{2000}{100} + 45 = 65$$

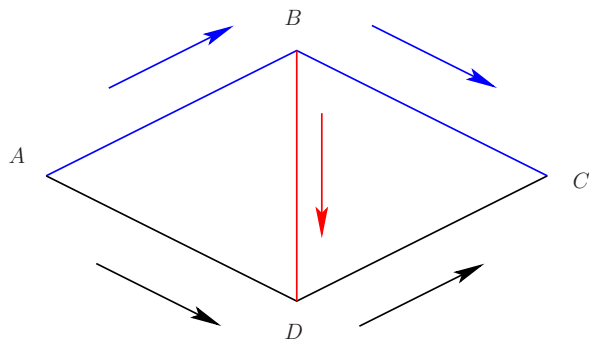
$$\text{Route } ADC: 45 + \frac{\# \text{ cars}}{100} \Rightarrow \frac{2000}{100} + 45 = 65$$

Vehicular Traffic



From	To	Time
A	B	$\frac{\# \text{ cars}}{100}$
B	C	45
A	D	45
D	B	$\frac{\# \text{ cars}}{100}$
B	D	0

Vehicular Traffic



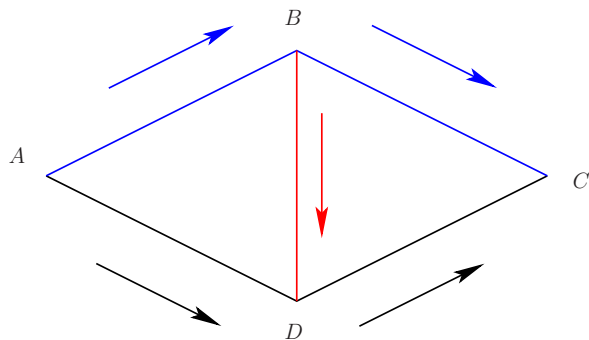
From	To	Time
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$$\text{Route } ABC: \frac{\# \text{ cars}}{100} + 45$$

$$\text{Route } ADC: 45 + \frac{\# \text{ cars}}{100}$$

$$\text{Route } ABDC: \frac{\# \text{ cars}}{100} + 0 + \frac{\# \text{ cars}}{100}$$

Vehicular Traffic



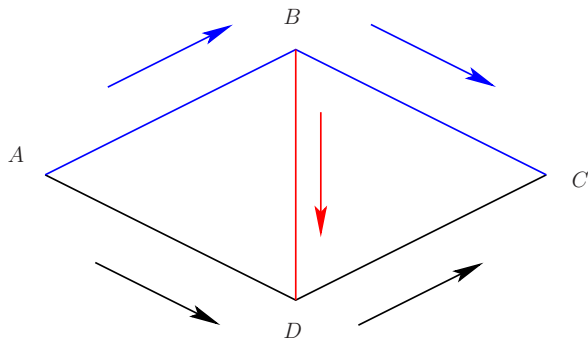
From	To	Time
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B	C	45
A	D	45
D	B	$\frac{\# \text{ cars}}{100}$
B	D	0

$$\text{Route } ABC: \frac{\# \text{ cars}}{100} + 45$$

$$\text{Route } ADC: 45 + \frac{\# \text{ cars}}{100}$$

$$\text{Route } ABDC: \frac{\# \text{ cars}}{100} + 0 + \frac{\# \text{ cars}}{100} \Rightarrow \frac{4000}{100} + \frac{4000}{100} = 80$$

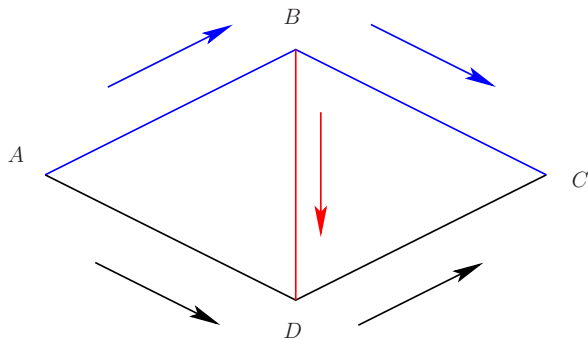
Vehicular Traffic – Braess Paradox



From	To	Time
A	B	$\frac{\# \text{ cars}}{100}$
B	C	45
A	D	45
D	B	$\frac{\# \text{ cars}}{100}$
B	D	0

Only ABC	80
ABC + ADC	65
ABC + ADC + ABDC	80

Vehicular Traffic – Braess Paradox



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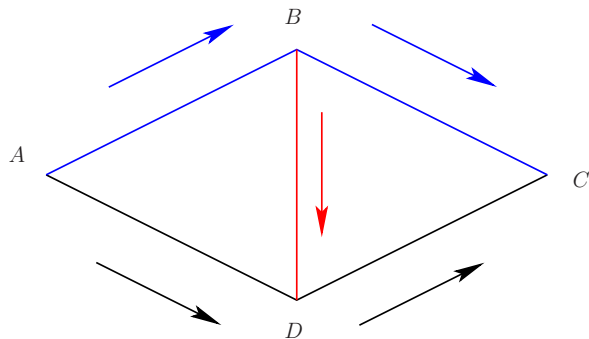
Only ABC 80

ABC + ADC 65

ABC + ADC + ABDC 80

Nash equilibrium vs. optimality

Vehicular Traffic – Braess Paradox



From	To	Time
A	B	$\frac{\# \text{ cars}}{100}$
B	C	45
A	D	45
D	B	$\frac{\# \text{ cars}}{100}$
B	D	0

Real!

Stuttgart Highway segment closed

1968

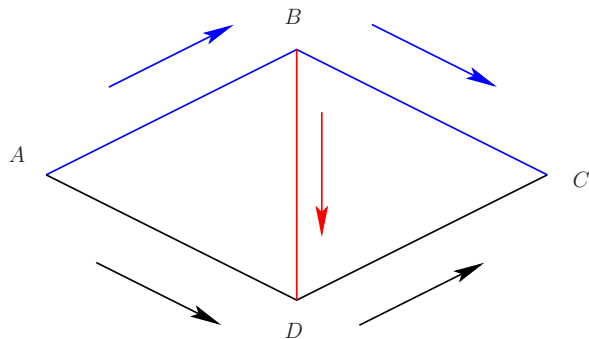
New York 42nd street closed

22.04.1990

Seoul 6 lanes highway substituted by a park

2008

Vehicular Traffic – Braess Paradox



From	To	Time
A	B	$\frac{\# \text{ cars}}{100}$
B	C	45
A	D	45
D	B	$\frac{\# \text{ cars}}{100}$
B	D	0

(Colombo, Holden: JOTA, 2016)

Characterization?
Dynamics?

Crowd Dynamics

Crowd Dynamics

$$\partial_t \rho + \operatorname{div}_x (\rho v(\rho) \vec{v}(x)) = 0 \quad \left\{ \begin{array}{l} v = \text{speed modulus} \\ \vec{v} = \text{velocity direction} \end{array} \right.$$

(Colombo, Facchi, Maternini: HYP2008 Proceedings, 2009)

Crowd Dynamics

$$\partial_t \rho + \operatorname{div}_x \left(\rho \vec{V}(\rho, x) \right) = 0$$

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$$\partial_t \rho + \operatorname{div}_x \left(\rho \vec{V}(\rho, x) \right) = 0$$

$$\partial_t \rho + \operatorname{div}_x \left(\rho v(\rho) \left(\vec{v}(x) + \begin{array}{l} \text{avoid} \\ \text{high} \\ \text{density} \end{array} \right) \right) = 0$$

Crowd Dynamics

$$\partial_t \rho + \operatorname{div}_x \left(\rho \vec{V}(\rho, x) \right) = 0$$

$$\partial_t \rho + \operatorname{div}_x \left(\rho v(\rho) \left(\vec{v}(x) - \frac{\kappa \operatorname{grad}_x (\rho * \eta)}{\sqrt{1 + \|\operatorname{grad}_x (\rho * \eta)\|^2}} \right) \right) = 0$$

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Theorem: If:

- v is smooth, decreasing, $v(0) = V$, $v(\rho) = 0$;
- \vec{v} is smooth;
- η is smooth with compact support;

Crowd Dynamics

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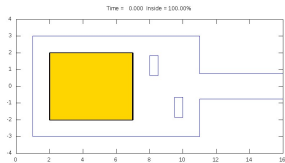
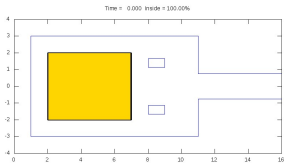
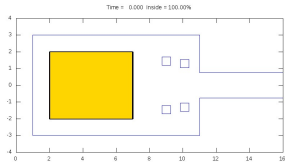
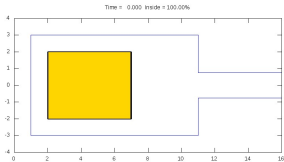
Then: Existence & Uniqueness in \mathbf{L}^1
Lipschitz Continuity from Data and Equation
Viability (discomfort)

(Colombo, Garavello, Lécureux–Mercier: M3AS, 2012)

(Colombo, Lécureux–Mercier: Acta Math. Sc., 2012)

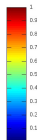
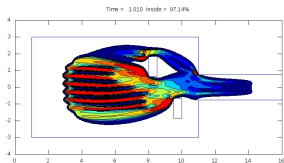
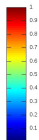
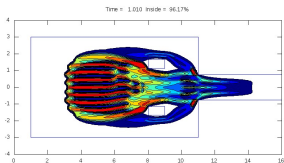
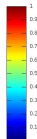
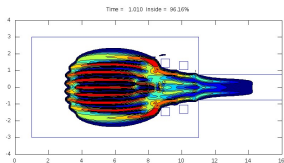
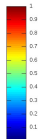
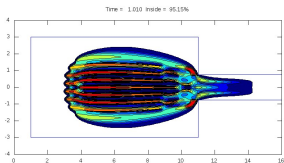
(Göttlich, Hoher, Schindler, Schleper: Appl.Mat.Mod., 2014)

Crowd Dynamics – Time to Exit



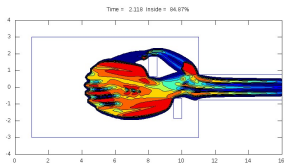
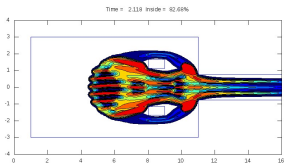
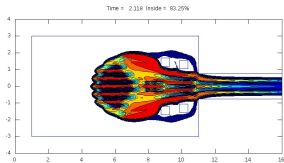
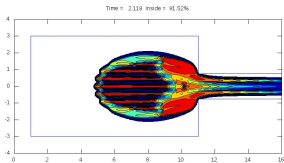
$t = 0.000$

Crowd Dynamics – Time to Exit



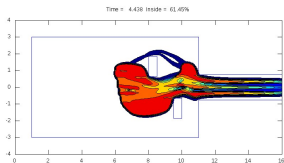
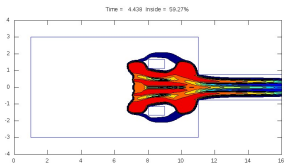
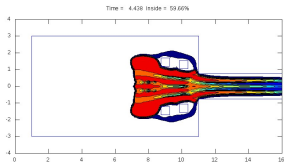
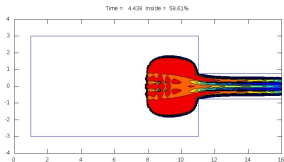
$t = 1.010$

Crowd Dynamics – Time to Exit



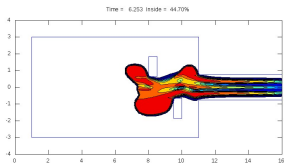
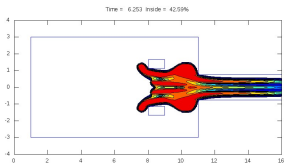
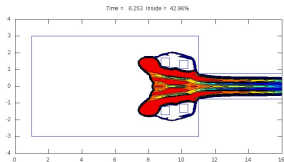
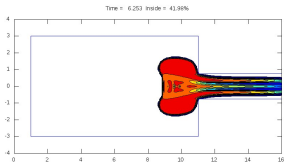
$t = 2.118$

Crowd Dynamics – Time to Exit



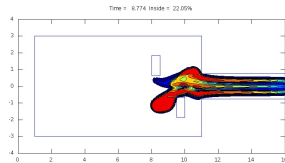
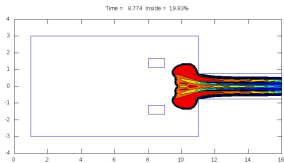
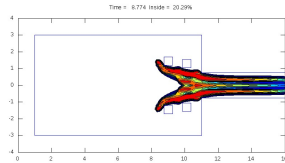
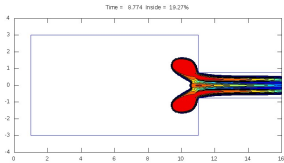
$t = 4.438$

Crowd Dynamics – Time to Exit



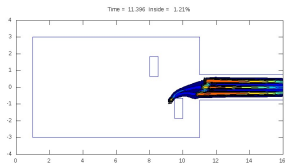
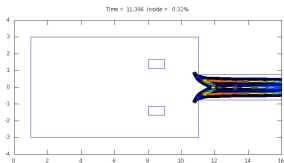
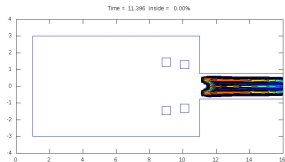
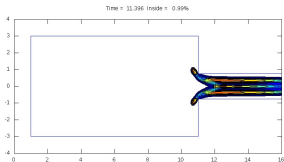
$t = 6.253$

Crowd Dynamics – Time to Exit



$t = 8.774$

Crowd Dynamics – Time to Exit



$t = 11.396$

Crowd Dynamics – Lane Formation

Two populations moving in opposite directions

Crowd Dynamics – Lane Formation

$$\partial_t \rho^1 + \operatorname{div}_x \left[\rho^1 v(\rho^1) \left(\vec{v}^1(x) - \frac{\varepsilon_{11} \nabla(\rho^1 * \eta)}{\sqrt{1 + \|\nabla(\rho^1 * \eta)\|^2}} - \frac{\varepsilon_{12} \nabla(\rho^2 * \eta)}{\sqrt{1 + \|\nabla(\rho^2 * \eta)\|^2}} \right) \right] = 0$$

$$\partial_t \rho^2 + \operatorname{div}_x \left[\rho^2 v(\rho^2) \left(\vec{v}^2(x) - \frac{\varepsilon_{21} \nabla(\rho^1 * \eta)}{\sqrt{1 + \|\nabla(\rho^1 * \eta)\|^2}} - \frac{\varepsilon_{22} \nabla(\rho^2 * \eta)}{\sqrt{1 + \|\nabla(\rho^2 * \eta)\|^2}} \right) \right] = 0$$

$$\vec{v}^1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix} + \delta \quad \eta(x, y) = [1 - (2x)^2]^3 [1 - (2y)^2]^3 \chi_{[-0.5, 0.5]^2}(x, y)$$

$$\vec{v}^2 = \begin{bmatrix} -1 \\ 0 \end{bmatrix} + \delta \quad v(\rho) = 4(1 - \rho) \quad \begin{array}{ll} \varepsilon_{11} = 0.3 & \varepsilon_{12} = 0.7 \\ \varepsilon_{21} = 0.7 & \varepsilon_{22} = 0.3 \end{array}$$

Crowd Dynamics – Shepherd Dog (Consensus)

$$\text{Given: } \begin{cases} \partial_t \rho + \operatorname{div}_x (\rho v(x, \rho, p)) = 0 & \text{HCL} \\ \dot{p} = \varphi \left(t, p, (\rho(t) * \eta)(p) \right) & \text{ODE} \end{cases}$$

Crowd Dynamics – Shepherd Dog (Consensus)

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IF: $v \in \mathbf{C}^2([0, R] \times \mathbb{R}^N \times \mathbb{R}^N; \mathbb{R}^N)$ is such that ...
 $\eta \in \mathbf{C}_c^1(\mathbb{R}^N; \mathbb{R})$
 φ Caratheodory, Locally Lipschitz, Sublinear

Crowd Dynamics – Shepherd Dog (Consensus)

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 φ Caratheodory, Locally Lipschitz, Sublinear

Then: There exists a solution (u, w) , with

- ✓ $\rho = \rho(t, x)$ weak entropy solution to HCL
- ✓ $p = p(t)$ Caratheodory solution to ODE
- ✓ stability estimates

$$\begin{aligned} & \|(\rho_1 - \rho_2)(t)\|_{\mathbf{L}^1} + \|(p_1 - p_2)(t)\| \\ \leq & C(t) \cdot \left(\|\partial_\rho(v_1 - v_2)\|_{\mathbf{L}^\infty} + \|\operatorname{div}_x(v_1 - v_2)\|_{\mathbf{L}^1} \right. \\ & \left. + \|\varphi_1 - \varphi_2\|_{\mathbf{L}^\infty} + \|\eta_1 - \eta_2\|_{\mathbf{L}^1} \right. \\ & \left. + \|\bar{\rho}_1 - \bar{\rho}_2\|_{\mathbf{L}^1} + \|\bar{p}_1 - \bar{p}_2\| \right) \end{aligned}$$

Crowd Dynamics – Shepherd Dog (Consensus)

(Colombo, Mercier: JNLS, 2012)

Crowd Dynamics – Shepherd Dog (Consensus)

(Colombo, Mercier: JNLS, 2012)

Crowd Dynamics – Policemen vs. Hooligans

$$\begin{cases} \partial_t \rho_i + \operatorname{div}_x \left[\rho^i (1 - \rho^i) \left(-w^i(x, \rho) + \mathcal{A}^i(\rho) \right) \right] = 0 & i = 1, 2 \\ \dot{\rho}^k = I_k(\rho) + \mathcal{B}_k(\rho) & k = 1, \dots, 4 \end{cases}$$

$$\mathcal{A}^1(\rho) = \frac{\varepsilon_{11} \eta^*(\rho^1 - \bar{\rho}) \nabla_x(\rho^1 * \eta)}{\sqrt{1 + \|\eta^*(\rho^1 - \bar{\rho}) \nabla_x(\rho^1 * \eta)\|^2}} + \frac{\varepsilon_{12} \eta^*(\rho^2 - \rho^1) \nabla_x(\rho^2 * \eta)}{\sqrt{1 + \|\eta^*(\rho^2 - \rho^1) \nabla_x(\rho^2 * \eta)\|^2}},$$

$$\mathcal{A}^2(\rho) = \frac{\varepsilon_{22} \eta^*(\rho^2 - \bar{\rho}) \nabla_x(\rho^2 * \eta)}{\sqrt{1 + \|\eta^*(\rho^2 - \bar{\rho}) \nabla_x(\rho^2 * \eta)\|^2}} + \frac{\varepsilon_{21} \eta^*(\rho^1 - \rho^2) \nabla_x(\rho^1 * \eta)}{\sqrt{1 + \|\eta^*(\rho^1 - \rho^2) \nabla_x(\rho^1 * \eta)\|^2}},$$

$$\mathcal{B}_k(\rho)(\rho) = \varepsilon_1 \frac{\nabla_x((\bar{\eta} * \rho^1)(\bar{\eta} * \rho^2))(\rho^k)}{\sqrt{1 + \|\nabla_x((\bar{\eta} * \rho^1)(\bar{\eta} * \rho^2))(\rho^k)\|^2}}$$

(Borsche, Colombo, Garavello, Meurer: JNLS, 2015)

Crowd Dynamics – Policemen vs. Hooligans

(Borsche, Colombo, Garavello, Meurer: JNLS, 2015)

Crowd Dynamics – 3D!

Crowd Dynamics – 3D!

Film

Predators – Prey

Hyperbolic

Parabolic

vs.

Predators

Prey

Hyperbolic Predators vs. Parabolic Prey

$$\begin{array}{l} \text{Predators: } u = u(t) \\ \text{Prey: } w = w(t) \end{array} \quad \left\{ \begin{array}{l} \partial_t u \\ \partial_t w \end{array} \right. \quad \begin{array}{l} = (\alpha w - \beta) u \\ = (\gamma - \delta u) w \end{array}$$

α = predators birth rate due to prey

β = predators mortality rate

γ = prey birth rate

δ = prey mortality rate due to predators

Hyperbolic Predators vs. Parabolic Prey

$$\begin{array}{l} \text{Predators: } u = u(t, x) \\ \text{Prey: } w = w(t, x) \end{array} \quad \left\{ \begin{array}{l} \partial_t u \\ \partial_t w - \mu \Delta w \end{array} \right. \quad \begin{array}{l} = (\alpha w - \beta) u \\ = (\gamma - \delta u) w \end{array}$$

α = predators birth rate due to prey

β = predators mortality rate

γ = prey birth rate

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Prey

diffuse

Hyperbolic Predators vs. Parabolic Prey

$$\begin{array}{l} \text{Predators: } u = u(t, x) \\ \text{Prey: } w = w(t, x) \end{array} \quad \left\{ \begin{array}{l} \partial_t u + \operatorname{div}_x (u v(w)) = (\alpha w - \beta) u \\ \partial_t w - \mu \Delta w = (\gamma - \delta u) w \end{array} \right.$$

α = predators birth rate due to prey

β = predators mortality rate

γ = prey birth rate

δ = prey mortality rate due to predators

$$v(w) = \kappa \frac{\operatorname{grad}(w * \eta)}{\sqrt{1 + \|\operatorname{grad}(w * \eta)\|^2}} \quad \begin{array}{l} \text{Predators} \\ \text{Prey} \\ \text{diffuse} \end{array}$$

Hyperbolic Predator vs. Parabolic Prey

There exists $\mathcal{R}: \mathbb{R}_+ \times \mathcal{X}_+ \rightarrow \mathcal{X}_+$ with the properties:

Hyperbolic Predator vs. Parabolic Prey

There exists $\mathcal{R}: \mathbb{R}_+ \times \mathcal{X}_+ \rightarrow \mathcal{X}_+$ with the properties:

1. $\mathcal{X}_+ = (\mathbf{L}^1 \cap \mathbf{L}^\infty \cap \mathbf{BV})(\mathbb{R}^N; \mathbb{R}) \times (\mathbf{L}^1 \cap \mathbf{L}^\infty)(\mathbb{R}^N; \mathbb{R})$
2. \mathcal{R} is a semigroup
3. $t \rightarrow \mathcal{R}_t(u_o, w_o)$ solves the system
4. $t \rightarrow \mathcal{R}_t(u_o, w_o)$ is continuous in time
5. $(u_o, w_o) \rightarrow \mathcal{R}_t(u_o, w_o)$ is locally Lipschitz continuous
6. Growth estimates
7. Propagation speed

(Colombo, Rossi: Comm.Math.Sc., 2015)

(Colombo, Marcellini, Rossi: NHM, 2016)

(Rossi, Schleper: M2AN, 2016)

Hyperbolic Predator vs. Parabolic Prey

