

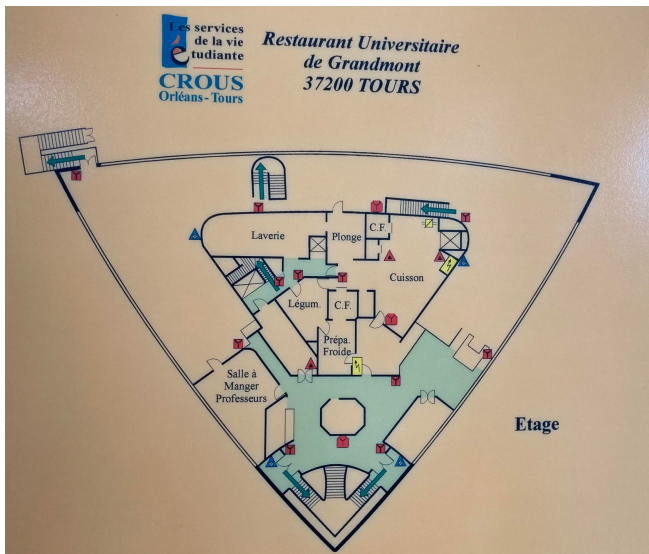
Pedestrian crowd models

T. GIRARD

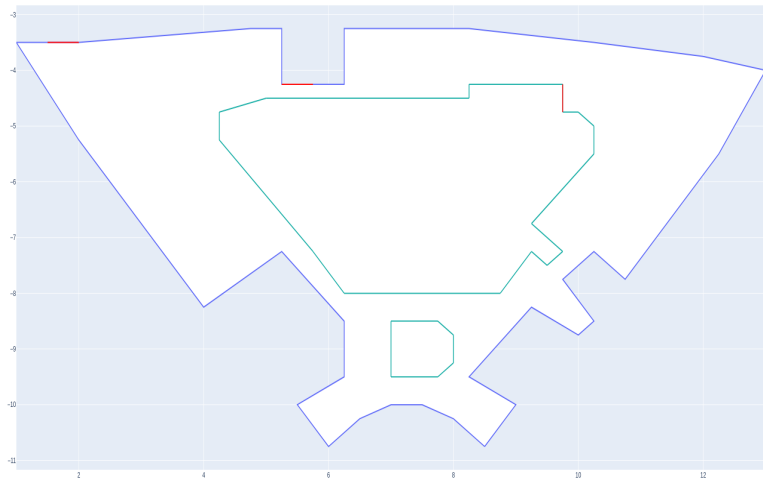
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I want to model the university restaurant of Tours in the context of evacuation...



... or its simplified version !



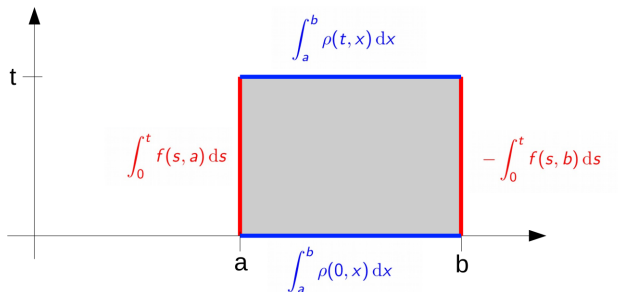
Outline

- 1 Transport of pedestrian : the LWR model
- 2 The direction of pedestrians in one dimension
- 3 The two-dimensional problem

We want to model a moving crowd. The crowd is represented as a pedestrian density ρ between 0 and 1.



The agents flux is represented by the flux function f .



$$\int_a^b \rho(t, x) dx = \int_a^b \rho(0, x) dx + \int_0^t f(s, a) ds - \int_0^t f(s, b) ds$$

$$\int_a^b \int_0^t \partial_t \rho(s, x) ds dx = - \int_0^t \int_a^b \partial_x f(s, x) dx ds$$

The flux is equal to the density multiply by the speed of agents.

$$f(s, x) := \rho(s, x)v(s, x)$$

The velocity v is itself governed by the local density:

$$v(s, x) := v_{\max}(1 - \rho)$$

We set $v_{\max} = 1$ and recover:

$$f(s, x) := f(\rho(s, x)) := \rho(s, x)(1 - \rho(s, x))$$

- M. J. Lighthill and G. B. Whitham, On kinematic waves. ii. a theory of traffic flow on long crowded roads, (1955).

We end up with:

$$\int_a^b \int_0^t \partial_t \rho(s, x) + \partial_x f(\rho(s, x)) \, dx \, ds = 0$$

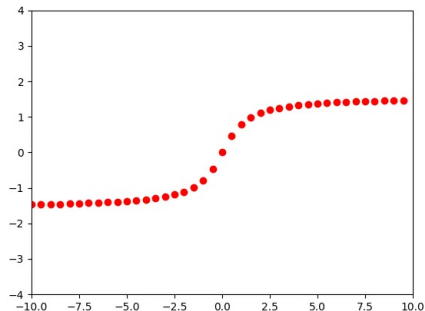
Short version, a scalar conservation law:

$$\rho_t + f(\rho)_x = 0$$

where $f(\rho) = \rho(1 - \rho)$. What's known on this equation ?

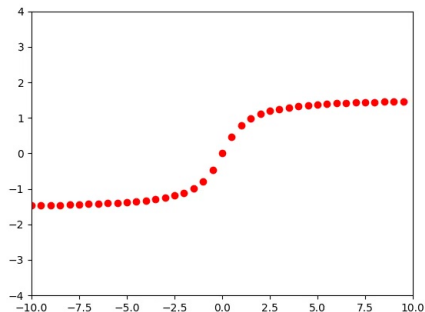
- **Non-existence of continuous solutions**

We use a method of characteristics to propagate the initial datum:



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So we consider weak solutions :

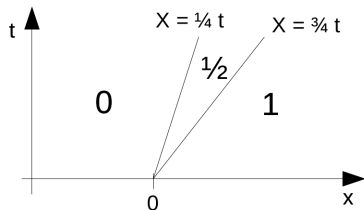
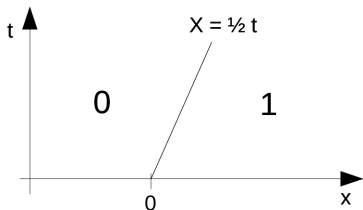
$$\forall \phi \in \mathcal{C}_c^\infty, \iint_{(0,T) \times \mathbb{R}} \rho \phi_t + f(\rho) \phi_x \, dt \, dx = 0$$

- Non-uniqueness of weak solutions

Consider

$$\begin{cases} \rho_t + [\rho^2/2]_x = 0 \\ \rho(0, x) = \mathbb{1}_{(0, +\infty)} \end{cases} \quad (1)$$

Then the two density functions ρ described below are weak solutions:



Krushkov : entropy conditions

We say that $\rho \in L^\infty$ is an entropy solution to

$$\begin{cases} \rho_t + f(\rho)_x = 0 \\ \rho(0, \cdot) = \rho_0(\cdot) \in L^\infty \end{cases}$$

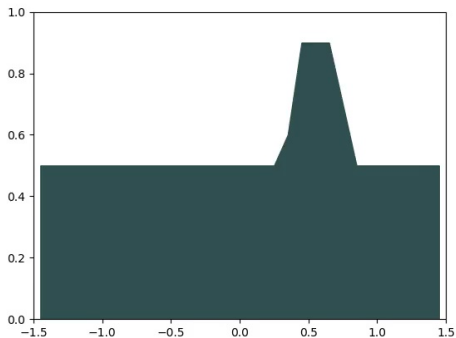
if

$|\rho - k|_t + (\text{sign}(\rho - k)(f(\rho) - f(k)))_x \leq 0$ in the distributional sense.

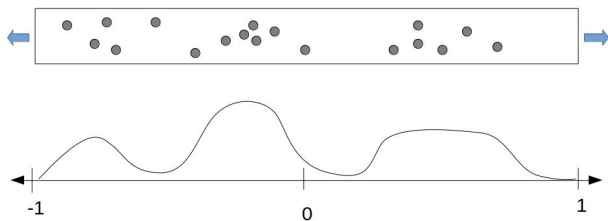
So $\forall k \in \mathbb{R}, \forall \phi \in \mathcal{C}_c^\infty$

$$\begin{aligned} \iint_{(0, T) \times \mathbb{R}} |\rho - k| \phi_t + \text{sign}(\rho - k)(f(\rho) - f(k)) \phi_x \, dt \, dx \\ + \int_{\mathbb{R}} |\rho_0 - k| \phi(0, x) \, dx \geq 0 \end{aligned}$$

Interpretation of Kruskov entropy condition in the context of traffic:
The admissible shocks correspond to the traffic jams.



How to model the psychology of crowds ?



A corridor with two doors located at $x = \pm 1$.

The simplest example : an empty corridor.

We consider a pedestrian located at $x \in [-1, 1]$ at time $t = 0$. This pedestrian computes the time required to reach each of the exits $T^{-1}(x)$ and $T^1(x)$. Naturally, the pedestrian will choose the lowest exit time.

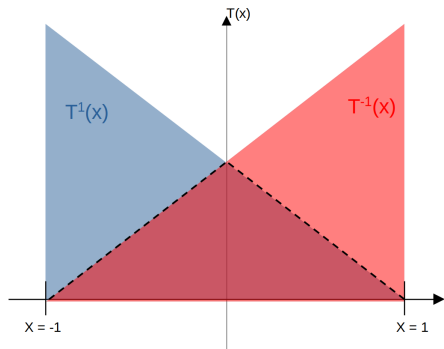
If we repeat this process for any x , we can define $u(x)$ the time to exit the corridor if one start at location x :

$$u(x) = \min\{T^{-1}(x), T^1(x)\}.$$

If the max speed is 1, we have

$$T^{-1}(x) = |x + 1| \text{ and}$$

$$T^1(x) = |x - 1|.$$



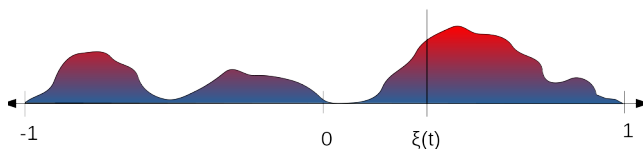
We consider a cost function c depending of the local density. We suppose each agent seeks to minimize not its exit time but its total cost towards the choosen exit. We have again

$$u(x) = \min\{T^{-1}(x), T^1(x)\}.$$

But this time, at speed 1 the total cost towards the exit $x = 1$ is

$$T^1(x) = \int_x^1 c(\rho(0, y)) dy$$

Problem (for later) : the supposed speed of pedestrians is still constant but the speed should vary with the density...

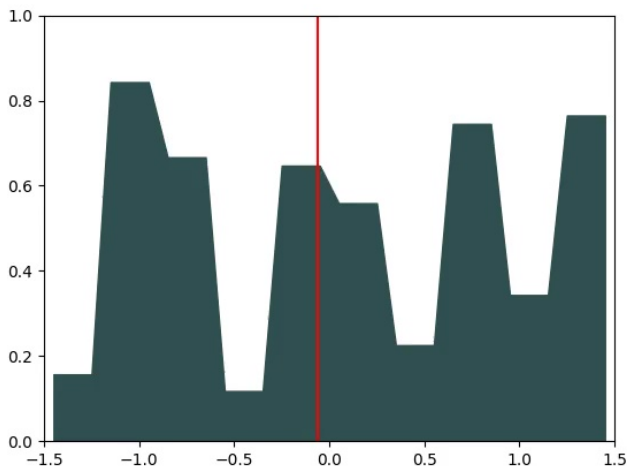


We want to solve:

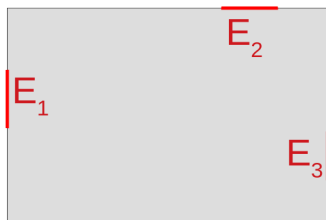
$$\begin{cases} \rho_t + [\text{sign}(x - \xi(t))\rho v(\rho)]_x = 0 \\ \int_{-1}^{\xi(t)} c(\rho(t, x)) dx = \int_{\xi(t)}^1 c(\rho(t, x)) dx. \end{cases}$$

The curve ξ is called the turning curve.

In the one-dimensional case, a fixed point argument proves the existence of (ρ, ξ) a solution. (But c has to be affine...)



We want to study the crowd evacuation of an initial density ρ_0 in the room when, at $t = 0$, the agents want to exit the room minimizing their exit time.



$$E = E_1 \cup E_2 \cup E_3$$

Suppose $V(t, x)$ is a vector field corresponding to the choice of direction of an agent located in x at time t . Then the transport equation follows from LWR:

$$\rho_t + \operatorname{div}_x(V(t, x)\rho v(\rho)) = 0$$

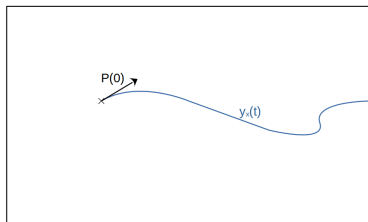
How do we compute V ?

For a fixed and constant density ρ , we use an optimal control problem.

Fix a density ρ in a given domain Ω . Let $\vec{p}(\cdot) \in \mathcal{C}^1([0, +\infty), \mathcal{S}^1)$. We call \vec{p} a "control". Consider the trajectory y_x solution of the Cauchy problem:

$$\begin{cases} \dot{y}_x(t) = v(\rho(y_x(t)))\vec{p}(t) \\ y_x(0) = x. \end{cases}$$

We call all these trajectories the "controlled trajectories" and denote by Y the set of all controlled trajectories.



Now we want to look at the exit time of a pedestrian located at x . If the pedestrian follows the trajectory y_x we compute the total exit time :

$$\int_0^{\infty} \mathbb{1}_{\Omega}(y_x(t)) dt.$$

Following the 1D case, we can also add a cost function c .

$$J(x, y_x(\cdot)) := \int_0^{\infty} c(\rho(y_x(t))) \mathbb{1}_{\Omega}(y_x(t)) dt.$$

If we consider that pedestrian always chose the best option to leave the domain, we end up with the following optimisation problem for the total cost:

$$u(x) = \inf_{y_x \in Y} J(x, y_x(\cdot)).$$

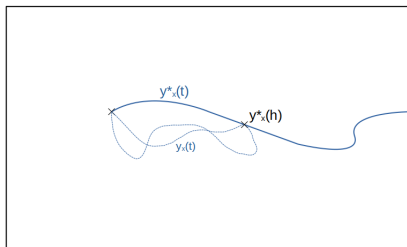
Suppose that the infimum is a minimum reached for an optimal control $y_x^*(\cdot)$.

The pedestrian at x should follow the direction field $V(x) = \dot{y}_x^*(0)$.

Can we compute $\dot{y}_x^*(0)$ for any x ?

The dynamic programming principle :

$$\forall h > 0, u(x) = \inf_{y_x \in \mathcal{Y}} \left\{ \int_0^h c(\rho(y_x(t))) \mathbb{1}_\Omega(y_x(t)) dt + u(y_x(h)) \right\}$$



Notice that, given a $y_x(\cdot)$ and if we suppose that u is differentiable, we have:

$$u(y_x(h)) = u(x) + \int_0^h \nabla u_{y_x(t)} \cdot \dot{y}_x(t) dt.$$

Using both equalities, we get:

$$\inf_{y_x \in Y} \left\{ \int_0^h c(\rho(y_x(t))) + \nabla u_{y_x(t)} \cdot \dot{y}_x(t) dt \right\} = 0.$$

Recall that $c > 0$, if y_x^* exists, then heuristically we should have

$$\dot{y}_x^*(0) = -\lambda \nabla u(x).$$

Furthermore,

$$\int_0^h c(\rho(y_x^*(t))) + \nabla u_{y_x^*(t)} \cdot \dot{y}_x^*(t) dt = 0.$$

This should be true for any h and, we get for all t ,

$$c(\rho(y_x^*(t))) + \nabla u_{y_x^*(t)} \cdot \dot{y}_x^*(t) = 0.$$

In particular, when $t = 0$, if $\dot{y}_x^*(0) = -\lambda \nabla u(x)$, we get:

$$\|\nabla u(x)\| = \frac{c(\rho(x))}{\|\dot{y}_x^*(0)\|} = \frac{c(\rho(x))}{v(\rho(x))}.$$

The Hamilton-Jacobi-Bellman approach:

Solving the optimisation problem $u(x) = \inf_{y_x \in Y} J(x, y_x(\cdot))$
 \Leftrightarrow solving the eikonal equation $\|\nabla u\| = \frac{c(\rho)}{v(\rho)}$.

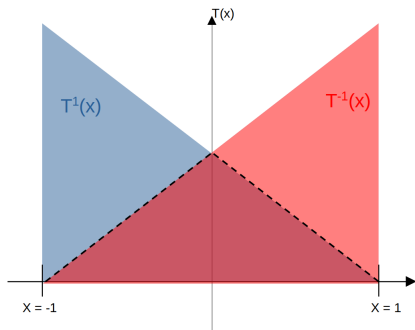
This is more or less the approach of the principle of least action in physics.

For a rigorous proof, Guy and Manu Chasseigne wrote a book where you can find it (and much more) !

Return to one-dimensional case : the solutions of Hamilton-Jacobi equations. If we try to solve the simple case

$$\begin{cases} |\partial_x u| = 1 \\ u(x = \pm 1) = 0. \end{cases}$$

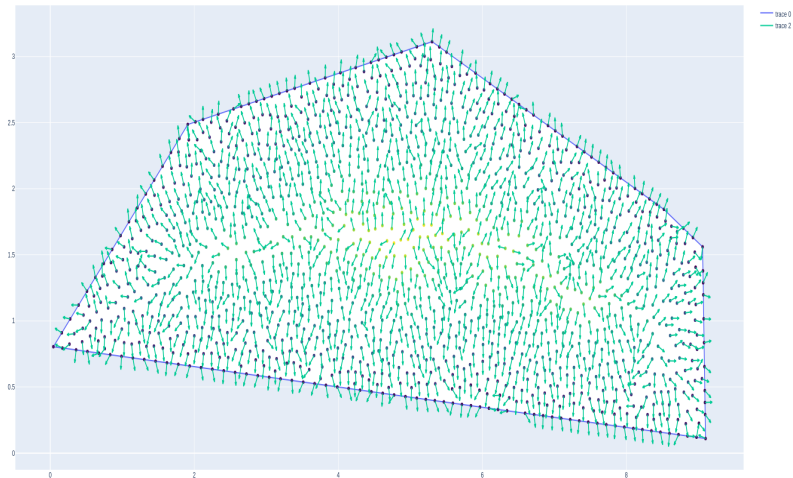
Can we say that $|\partial_x u| = 1$ almost everywhere ?



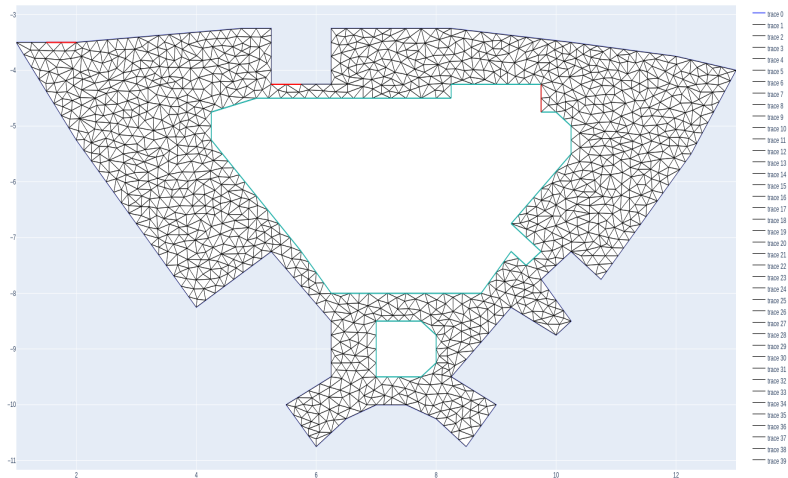
We would lose the uniqueness...

Solution : the notion of viscosity solution.

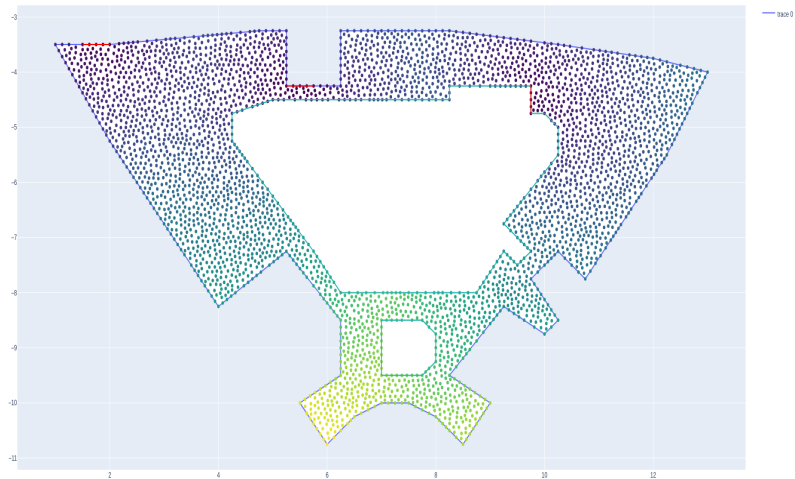
We can approach the eikonal equation's solution via a fast marching numerical scheme.



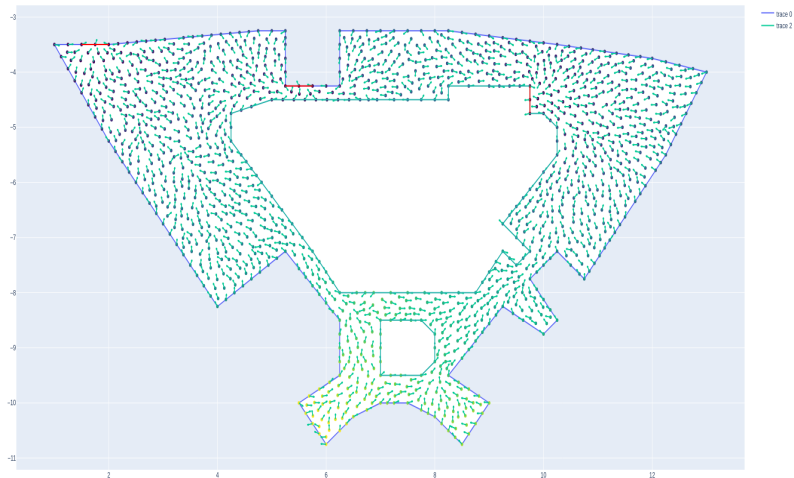
The simulation for the university restaurant of Tours:



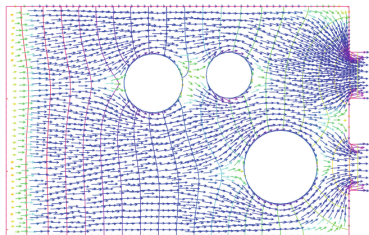
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To summarize, we should find the solutions of:



2D: directions \vec{V}
Gokieli & al.'19

$$\left\{ \begin{array}{l} \rho_t + \operatorname{div}_x \left(\frac{-\nabla \phi}{|\nabla \phi|} \rho v(\rho) \right) = 0 \\ |\nabla_x \phi| = \frac{c(\rho)}{v(\rho)} \\ \phi(x \in E) = 0 \\ (\nabla_x \phi \cdot n_D)^+ = 0 \text{ if } x \in \partial D \setminus E \\ \rho(0, x) = \rho(x) \end{array} \right. \quad (2)$$

where n_D is the normal unit vector to the boundary of the domain D and c is a given cost function depending on the local density.

work in progress...

Thank you.