Colloquium du CERMICS



The Cauchy problem treated as a convex optimization problem

Yann Brenier (CNRS, DMA-ENS, PSL)

5 mars 2018

The Cauchy problem treated as a convex optimization problem

Yann Brenier CNRS, DMA-ENS, PSL en association avec le projet CNRS-INRIA "MOKAPLAN"

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Cauchy by convex optimization

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$$\partial_t U + \nabla \cdot (F(U)) = 0, \ \ U = U(t, x) \in \mathcal{W} \subset \mathbb{R}^m,$$

$$t > 0, \ x \in \mathbb{R}^{d}, \ \nabla = \left(\frac{\partial}{\partial x_{i}}\right)_{i=1}^{d}, \text{ such that } \sum_{\beta=1}^{m} \partial_{\beta} \mathcal{E}(W) \partial_{\alpha} F^{i\beta}(W) = \partial_{\alpha} Q^{i}(W), \ \forall W \in \mathcal{W},$$

for some smooth pair $(\mathcal{E}, Q) : \mathcal{W} \to \mathbb{R}^{1+d}$ (with \mathcal{W} open convex and \mathcal{E} , called "entropy", strictly convex), which implies $\partial_t(\mathcal{E}(U)) + \nabla \cdot (Q(U)) = 0$, for all smooth solutions U.

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A famous example: the equations of an isothermal gas (Euler, 1755)

$$\partial_t \rho + \nabla \cdot q = 0, \ \partial_t q + \nabla \cdot (\frac{q \otimes q}{\rho}) + \nabla \rho = 0, \ \mathcal{E} = \frac{|q|^2}{2\rho} + \rho \log \rho, \ \rho > 0, \ q \in \mathbb{R}^d.$$

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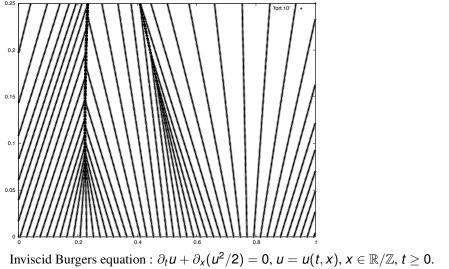
Typically, these systems are locally well-posed, with generic formation of shock waves.

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Cauchy by convex optimization

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Formation of two shock waves. (Vertical axis: $t \in [0, 1/4]$, horizontal axis: $x \in \mathbb{T}$.)

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The problem is not trivial since there may be many weak solutions starting from U_0 which are not entropy-preserving (by "convex integration" à la De Lellis-Székelyhidi).

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The resulting saddle-point problem

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The resulting saddle-point problem

$$\inf_{U} \sup_{A} \int_{[0,T]\times\mathbb{T}^{d}} \mathcal{E}(U) - \partial_{t}A \cdot U - \nabla A \cdot F(U) \\ - \int_{\mathbb{T}^{d}} A(0, \cdot) \cdot U_{0}$$

where $A = A(t, x) \in \mathbb{R}^m$ is smooth with $A(T, \cdot) = 0$. Here U_0 is the initial condition and T the final time.

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where $A = A(t, x) \in \mathbb{R}^m$ is smooth with $A(T, \cdot) = 0$. Here U_0 is the initial condition and T the final time.

N.B. The supremum in *A* exactly encodes that *U* is a weak solution with initial condition U_0 , all test functions *A* acting like Lagrange multipliers.

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leads to a *concave* maximization problem in A, namely

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$$= \sup_{A} \int_{[0,T]\times\mathbb{T}^d} -K(\partial_t A, \nabla A) - \int_{\mathbb{T}^d} A(0, \cdot) \cdot U_0$$

 $\mathcal{K}(\mathcal{E},\mathcal{B}) = \sup_{\mathcal{V}\in\mathcal{W}\subset\mathbb{R}^m} \mathcal{E}\cdot\mathcal{V} + \mathcal{B}\cdot\mathcal{F}(\mathcal{V}) - \mathcal{E}(\mathcal{V}), \ (\mathcal{E},\mathcal{B})\in\mathbb{R}^m\times\mathbb{R}^{m\times d}.$

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$$= \sup_{A} \int_{[0,T]\times\mathbb{T}^d} -\mathcal{K}(\partial_t A, \nabla A) - \int_{\mathbb{T}^d} A(0, \cdot) \cdot U_0$$

 $\mathcal{K}(E,B) = \sup_{V \in \mathcal{W} \subset \mathbb{R}^m} E \cdot V + B \cdot F(V) - \mathcal{E}(V), \ (E,B) \in \mathbb{R}^m \times \mathbb{R}^{m \times d}.$

Notice that *K* is automatically convex.

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Here is the paradox!

How a convex optimization problem could be compatible with a well-posed evolution problem?

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How a convex optimization problem could be compatible with a well-posed evolution problem? For instance, if *K* were just a square, we would get

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which would correspond to an ill-posed equation for A:

 $\partial_{tt}^2 A + \triangle A = 0.$

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Answer: in our construction, K is convex degenerate!

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A consistency result

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Theorem: If U is a smooth solution to the Cauchy problem and T is not too large

A consistency result

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- (*) more precisely if, $\forall t, x, V \in \mathcal{W}, \mathcal{E}''(V) (T t)F''(V) \cdot \nabla(\mathcal{E}'(U(t, x))) > 0$, which requires, in particular, $\mathcal{E}''(V) TF''(V) \cdot \nabla(\mathcal{E}'(U_0(x))) > 0$.

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...But what about shocks and large T???

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Then, the maximization problem in A simply reads

$$\sup_{A}\int_{[0,T]\times\mathbb{T}}-\frac{(\partial_{t}A)^{2}}{2(1-\partial_{x}A)}-\int_{\mathbb{T}}A(0,\cdot)u_{0}.$$

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Introducing $\rho = 1 - \partial_x A \ge 0$, $q = \partial_t A \in \mathbb{R}$, this problem is equivalent to

$$-\inf\{\int_{[0,T]\times\mathbb{T}}\frac{q^2}{2\rho}-qu_0, \ (\rho\geq 0,q) \text{ subject to } \partial_t\rho+\partial_xq=0, \ \rho(T,\cdot)=1\}.$$

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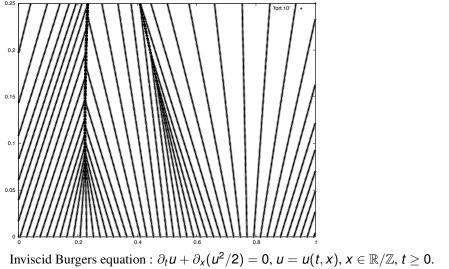
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i.e. the "ballistic" version (à la Ghoussoub) of the optimal transport problem with quadratic cost (à la Benamou-B.), and, as well, an elementary "mean-field game".

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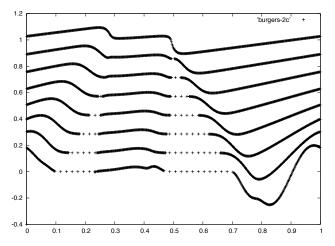


Formation of two shock waves. (Vertical axis: $t \in [0, 1/4]$, horizontal axis: $x \in \mathbb{T}$.)

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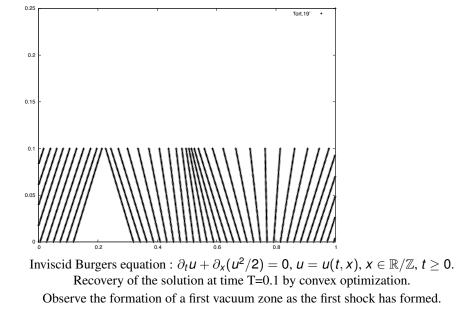
Numerics: 2 lines of code differ from a standard (Benamou-B.) OT solver!

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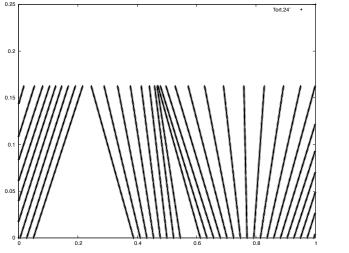
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Inviscid Burgers equation : $\partial_t u + \partial_x (u^2/2) = 0$, u = u(t, x), $x \in \mathbb{R}/\mathbb{Z}$, $t \ge 0$. Recovery of the solution at time T=0.16 by convex optimization.

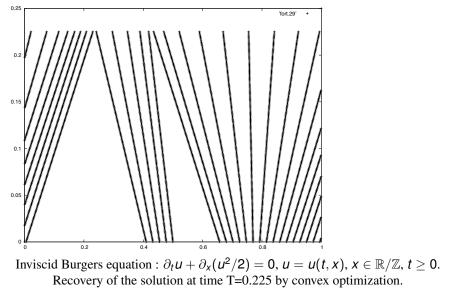
Observe the formation of a second vacuum zone as the second shock has formed.

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Observe the extension of the two vacuum zones.

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Analogy with mountain climbing:

going from Everest to Lhotse without following the crest! (Credit to Th. Gallouët.)

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Merci de votre attention!

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Pour plus de détails, voir Y.B. ArXiv Oct. 2017.

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