Low congested regions and networks and optimal reinforcement for a membrane

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Traffic with congestion

Congestion effects have been extensively studied since '50 by Wardrop (discrete case) and Beckmann (continuous model).

Data:

- a bounded Lipschitz region $\Omega \subset \mathbb{R}^d$;
- two probability measures f^+ and f^- .

The model, in a stationary regime, reduces to:

$$\min \left\{ \int_{\Omega} H(\sigma) \, dx : -\operatorname{div} \sigma = f \text{ in } \Omega, \ \sigma \cdot n = 0 \text{ on } \partial \Omega \right\},$$

where:

- σ is the traffic flux;
- $H: \mathbb{R}^d \to [0, +\infty]$ is the congestion function, i.e. a convex nonnegative function with $\lim_{|s| \to +\infty} H(s)/|s| = +\infty$.



A continuous model

In the model:

$$\min \left\{ \int_{\Omega} H(\sigma) \, dx : -\operatorname{div} \sigma = f \text{ in } \Omega, \ \sigma \cdot n = 0 \text{ on } \partial \Omega \right\},$$

- the boundary condition $\sigma \cdot n = 0$ on $\partial \Omega$ models the zero normal flux on the boundary.
- the PDE $-\operatorname{div} \sigma = f^+ f^-$ captures the equilibrium between the traffic flux and the difference f.

Recently, Carlier, Jimenez, Santambrogio have presented a model equivalent to Beckmann's problem.



Membrane model

We consider an elastic membrane under the action of an exterior load f and fixed at its boundary; this amounts to solve the variational problem

$$\min \left\{ \int_{\Omega} \left(\frac{1}{2} |\nabla u|^2 - fu \right) dx : u \in H_0^1(\Omega) \right\}$$

or equivalently the elliptic PDE

$$-\Delta u = f \text{ in } \Omega, \qquad u \in H_0^1(\Omega).$$

Duality

The two previous models are connected by the following:

Lemma

X,Y Banach, $A:X\to Y$ linear, $F:Y\to \overline{\mathbb{R}}$ convex. Then for every $f\in X'$

$$\inf_{\sigma \in Y'} \Big\{ F^*(\sigma) : A^*\sigma = f \Big\} = -\inf_{u \in X} \Big\{ F(Au) - \langle f, u \rangle \Big\}.$$

Taking $Au = \nabla u$ the transport problem with congestion H can be written in its dual form

A new model

In Ω we consider

- two congestion functions $H_1 \leq H_2$;
- a penalization cost *m*.

For every region C we consider the cost function

$$F(C) = \min \left\{ \int_{\Omega \setminus C} H_2(\sigma) \, dx + \int_C H_1(\sigma) \, dx : \sigma \in \Gamma_f \right\}$$

where

$$\Gamma_f = \{ \sigma \in L^1(\Omega; \mathbb{R}^d) : -\operatorname{div} \sigma = f \text{ in } \Omega, \ \sigma \cdot n = 0 \text{ on } \partial \Omega \}.$$

Goal: find a low congested region $C \subset \Omega$ solving

$$\min\big\{F(C)+m(C)\ :\ C\subset\Omega\big\}.$$



Perimeter penalization

We consider the case m(C) = kPer(C), k > 0

Theorem

Assume that the cost F(C) is finite for at least a subset C of $\overline{\Omega}$ with finite perimeter. Then there exists at least an optimal set C_{opt} .

Regularity:

- since $H_2 \ge H_1$, implies that ∂C has nonnegative mean curvature;
- when d=2 and Ω is convex, the optimal regions $\mathcal C$ are convex.



Volume penalization

We consider the case m(C) = k |C|, k > 0

Passing from sets C to density function $0 \le \theta(x) \le 1$ we obtain the relaxed formulation

$$\min_{\sigma,\theta} \left\{ \int_{\Omega} \theta H_1(\sigma) \, dx + \int_{\Omega} (1-\theta) H_2(\sigma) \, dx + k \int_{\Omega} \theta \, dx \; : \; \sigma \in \Gamma_f \right\}.$$

After the elimination of the variable θ we end up with a non convex integrand.

A new relaxation is necessary, so we have

$$\min \left\{ \int_{\Omega} \left(H_2(\sigma) \wedge \left(H_1(\sigma) + k \right) \right)^{**} dx \ : \ \sigma \in \Gamma_f \right\}.$$



Volume penalization

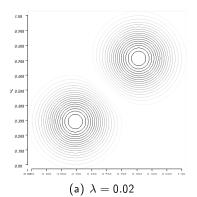
Recall: the functions H_1 and H_2 are superlinear.

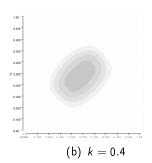
We denote by $\bar{\sigma}$ the optimal solution of the problem we have that:

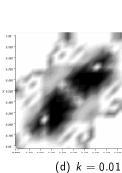
$$\frac{\Omega}{\left(H_2 \wedge \left(H_1 + k\right)\right)^{**}(\overline{\sigma})} = H_2(\overline{\sigma}) \Rightarrow \theta = 0;$$

$$\left(H_2 \wedge \left(H_1 + k\right)\right)^{**}(\overline{\sigma}) = H_1(\overline{\sigma}) + k \Rightarrow \theta = 1;$$

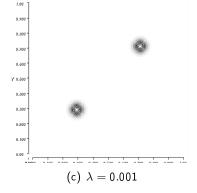
$$\left(H_2 \wedge \left(H_1 + k\right)\right)^{**}(\overline{\sigma}) < \left(H_2 \wedge \left(H_1 + k\right)\right)(\overline{\sigma}) \Rightarrow 0 < \theta < 1.$$











One dimensional sets

We consider the case $C \in \mathcal{A}_L$,

$$\mathcal{A}_L = \left\{ S \subset \Omega, \ S \ \text{closed connected}, \ \mathcal{H}^1(S) \leq L \right\}$$

where \mathcal{H}^1 is the one-dimensional Hausdorff measure. For every $S \in \mathcal{A}_L$ we define the energy functional

$$\mathcal{E}_f(S) = \inf \left\{ \int_{\Omega} \frac{1}{2} |\nabla u|^2 \, dx + \int_{S} \frac{1}{2} |\nabla u|^2 \, d\mathcal{H}^1 - \int_{\Omega} u \, df : u \in C_c^{\infty}(\Omega) \right\}$$

so that the optimization problem we deal with is

$$\max\big\{\mathcal{E}_f(S)\ :\ S\in\mathcal{A}_L\big\}.$$



One dimensional sets

Since limits of one-dimensional sets are in general measures, it is convenient to define the energy functional \mathcal{E}_f even for a measure μ , by setting

$$\mathcal{E}_f(\mu) = \inf \Big\{ \int_{\Omega} \frac{1}{2} |\nabla u|^2 \ dx + \int_{\Omega} \frac{1}{2} |\nabla u|^2 \ d\mu - \int_{\Omega} u \ df \ : \ u \in \mathit{C}^{\infty}_c(\Omega) \Big\}.$$

- Theorem: Assume that $\mathcal{E}_f(\mu)$ is finite for at least a $\mu \in \mathcal{M}_L^+(\Omega)$. Then the maximization problem $\max \left\{ \mathcal{E}_f(\mu) : \mu \in \mathcal{M}_L^+(\Omega) \right\}$ admits at least a solution.
- Theorem: Let μ be a solution of the maximization problem. Then there exists a one-dimensional closed connected set S such that the absolutely continuous part of μ with respect to $\mathcal{H}^1 \lfloor S$ is also a solution. In other words the solution μ is of the form $\mu = a(x)\mathcal{H}^1 \lfloor S$, with $S \in \mathcal{A}_L$ and $a(x) \geq 1$.

One dimensional sets

Open question

It is possible to find an example in which we have the case a(x) > 1?

If we consider $f^+ = \delta_A$, $f^- = \delta_B$ and L >> |A - B|, what is the optimal μ ? On which set is it concentrated?