Optimal transport methods in the least gradient problem

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The least gradient problem

Let $\Omega \subset \mathbb{R}^N$ be an open bounded set with Lipschitz boundary. Take $f \in L^1(\partial \Omega)$. Consider the following minimisation problem:

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It is linked to the study of minimal surfaces, but also (among others) to optimal design, conductivity imaging, and optimal transport.

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Classical approach

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- Existence of solutions for f continuous \mathcal{H}^{N-1} -a.e. on $\partial \Omega$;
- Uniqueness and continuity of solutions for $f \in C(\partial \Omega)$;
- Hölder regularity of solutions for Hölder boundary data (no similar result for Sobolev regularity).

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Equivalence with the Beckmann problem

Suppose that $\Omega \subset \mathbb{R}^2$ is convex. Then, the least gradient problem is equivalent to the *Beckmann problem* (Rybka, Sabra, G. 2017):

$$\min\bigg\{\int_{\overline{\Omega}}|p|: \quad p\in\mathcal{M}(\overline{\Omega};\mathbb{R}^2), \quad \operatorname{div} p=0, \quad p\cdot\nu|_{\partial\Omega}=g\bigg\},\$$

where $g = \frac{\partial f}{\partial \tau}$, in the following sense: from a solution to the LGP we may construct a solution to the Beckmann problem, and vice versa if the solution to the Beckmann problem gives no mass to the boundary.

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The equivalence is formally given by $p = R_{-\frac{\pi}{2}}Du$.

Equivalence with an optimal transport problem

Again on convex domains, the Beckmann problem is equivalent to the optimal transport problem with source and target measures on $\partial \Omega$:

$$\min\left\{\int_{\overline{\Omega}\times\overline{\Omega}}|x-y|\,d\gamma:\,\gamma\in\mathcal{M}^+(\overline{\Omega}\times\overline{\Omega}),\,(\Pi_x)_{\#}\gamma=g^+,(\Pi_y)_{\#}\gamma=g^-\right\}$$

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From every solution p to the Beckmann problem we can construct a solution to the OTP with transport density σ_{γ} (and vice versa) and

$$\sigma_{\gamma} = |\mathbf{p}|.$$

These two results were put together by Dweik and Santambrogio (2019). Their main idea was as follows:

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- We need less than continuity of f for uniqueness of solutions we only require that either $(\partial_{\tau} f)^+$ or $(\partial_{\tau} f)^-$ is atomless;
- Because $\sigma_{\gamma} = |p| = |Du|$, L^{p} estimates for σ_{γ} correspond to $W^{1,p}$ estimates for u. This is (so far) the only way to prove Sobolev regularity of solutions to LGP.

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- For equivalence between the Beckmann problem and the OTP, we do not need convexity, only that all the transport rays for some optimal transport plan lie inside the domain. This leads to admissibility conditions for boundary data.
- A modified version of this framework works also in more surprising situations, such as on annuli which are not contractible.

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We need to take care of two things. The first one is that when $\partial \Omega$ is not connected, we have equivalence between the Beckmann problem and the following variant of the LGP:

$$\min\left\{\int_{\Omega}|Du|: u\in BV(\Omega), \quad \partial_{\tau}(Tu)=g\right\}.$$

The trace is specified only up to vertical translations on connected components of $\partial \Omega!$

LGP on annuli

The second one is that we need to ensure that a solution to the optimal transport problem gives no mass to the boundary. For this, we need some admissibility conditions for boundary data.

Theorem 1 (Dweik, G.)

Suppose that $\Omega \subset \mathbb{R}^2$ is an annulus. Under certain admissibility conditions on $f \in BV(\partial \Omega)$, there exists a solution $u \in BV(\Omega)$ to the least gradient problem. Moreover:

- If one of $(\partial_{\tau} f)^{\pm}$ is atomless, the solution is unique;
- If $f \in W^{1,p}(\partial \Omega)$, then $u \in W^{1,p}(\Omega)$.

Thank you for your attention!

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