

Convex representation for curvature dependent functionals

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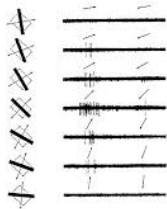
Introduction

- ▶ based on “roto-translation” group;
- ▶ a simple formula for curvature-dependent line energies;
- ▶ a general relaxation for functions;
- ▶ tightness result (C^2 sets);
- ▶ dual formulation and link with previous works [Bredies-Pock-Wirth'15];
- ▶ numerical results

Curvature information: a “natural” idea

Experiments and discovery of Hubel-Wiesel (62, 77)

V1 physiology:
direction
selectivity



Observation: the brain reacts to orientation. Corresponding cells are stacked and connected together to provide sensitivity to curvature. First mathematical theories: Koenderink-van Doorn (87), Hoffman (89), Zucker (2000), Petitot-Tondut (98/2003), Citti-Sarti (2003/2006).

Main idea: use the sub-Riemannian structure of the roto-translation group $((a, R) \in SE(2) \simeq \mathbb{R}^2 \times SO(2) \sim \mathbb{R}^2 \times \mathbb{S}^1$ in dimension 2) to describe the geometry of the visual cortex \rightarrow sub-Riemannian diffusion and mean curvature motion (Citti-Sarti 3/6, Duits-Franken 10, Boscaïn et al 14, Citti et al, 2015) for inpainting.

Variational approaches

To complete contours, Mumford (94) suggested to use the “elastica” functional

$$\int_{\gamma} \kappa^2 d\mathcal{H}^1$$

for contour completion. (Idea suggested by psychological experiments, *cf* for instance Kanizsa 1980.) General theory by Masnou-Morel 98.

Issues: not lower semicontinuous. Studied by Bellettini-Mugnai 2004/2005, Nardi (PhD 2011), Dayrens-Masnou 16, Ambrosio-Masnou 2003. *[Examples]*

Minimisation is computationally challenging. A few approaches trying to exploit the roto-translation metric: Schoenemann with Cremers (2007), Kahl and Cremers (2009), Masnou and Cremers (2011): discrete approach on a graph (or LP) where vertices encode position and orientation (also, El Zehiry-Grady 2010, ...);

Length computation by JM Mirebeau (anisotropic Eikonal equations, 2014)

Variational approaches

Bredies-Pock-Wirth 2013, 2015: “vertex” penalization (“TVX”), then general energies $\int_{\gamma} f(x, \tau, \kappa)$, f convex, $f \geq 1$. Need to “lift” the image in $\mathbb{R}^2 \times \mathbb{S}^1 \times \mathbb{R}$ where last component = curvature, with compatibility condition.

This work: a new (and simpler) representation for the latter approach (with $f(\kappa)$).

Example: a C^2 curve

$\gamma(t)$ planar curve, with $|\dot{\gamma}| = 1$ ($\dot{\gamma} = \tau_\gamma$), and $\ddot{\gamma} = \kappa_\gamma \tau_\gamma^\perp$.

Lifted as $\Gamma(t) = (\gamma(t), \theta(t))$ where $\tau_\gamma = (\cos \theta, \sin \theta)$.

Then: the length of $\Gamma(t)$ in $\Omega \times \mathbb{S}^1$ is

- ▶ Finite: sub-Riemannian structure, local metric is infinite in direction θ^\perp (we will also take into account orientation);
- ▶ Given by $\int_0^L \sqrt{\dot{\gamma}^2 + \dot{\theta}^2} dt = \int_0^L \sqrt{1 + \kappa^2} dt$: encoding curvature penalization information.

Example: a C^2 curve

Let now $f : \mathbb{R} \rightarrow \mathbb{R}$ be convex, assume $f \geq 1$, and consider the energy

$$\int_0^L f(\kappa) = \int_0^L f(\dot{\Gamma}^\theta(t)) dt.$$

Observe that if one considers a reparametrization $\lambda(s)$, $s \in [0, a]$, of the curve Γ , then $\lambda^\times(s)$ is a reparametrization of γ , $\lambda^\times = |\dot{\lambda}^\times| \tau$, $\kappa = d\theta/dt = \dot{\lambda}^\theta ds/dt = \dot{\lambda}^\theta / |\dot{\lambda}^\times|$ hence the energy becomes

$$\int_0^L f(\kappa) dt = \int_0^a f(\dot{\lambda}^\theta / |\dot{\lambda}^\times|) |\dot{\lambda}^\times| ds.$$

Example: a C^2 curve

Denoting σ the measure (charge) in $\mathcal{M}^1(\Omega \times \mathbb{S}^1; \mathbb{R}^3)$ defined by the curve $\Gamma(t)$:

$$\int_{\Omega \times \mathbb{S}^1} \psi \cdot \sigma = \int_0^L \psi(\Gamma(t)) \cdot \dot{\Gamma}(t) dt,$$

one obtains that

$$\int_0^L f(\kappa) = \int_{\Omega \times \mathbb{S}^1} \bar{h}(\sigma^x \cdot \theta, \sigma^\theta)$$

where

$$\bar{h}(s, t) = \begin{cases} sf(t/s) & \text{if } s > 0, \\ f^\infty(t) & \text{if } s = 0, \\ +\infty & \text{else.} \end{cases}$$

where $f^\infty(t) = \lim_{s \rightarrow 0} sf(t/s)$ is the recession function of f .

Example: a C^2 curve

It is standard that if f is convex lsc, then also h is, with

$$\bar{h}(s, t) = \sup \{as + bt : a + f^*(b) \leq 0\}.$$

In addition, as $\sigma^x = \lambda\theta$ where λ is a positive measure in $\Omega \times \mathbb{S}^1$, introducing for $p = (p^x, p^\theta) \in \mathbb{R}^3$

$$h(\theta, p) = \begin{cases} \bar{h}(p^x \cdot \theta, p^\theta) & \text{if } p^x \cdot \theta = |p^x| \Leftrightarrow p^x \parallel \theta, p^x \cdot \theta \geq 0 \\ +\infty & \text{else,} \end{cases}$$

which encodes the sub-Riemannian structure of $\Omega \times \mathbb{S}^1$: we also have

$$\int_0^L f(\kappa) = \int_{\Omega \times \mathbb{S}^1} \bar{h}(\sigma^x \cdot \theta, \sigma^\theta) = \int_{\Omega \times \mathbb{S}^1} h(\theta, \sigma).$$

Example: a C^2 curve

Now, observe that $\operatorname{div} \sigma = \delta_{\Gamma(L)} - \delta_{\Gamma(0)}$, in particular if γ is a closed curve or has its endpoints on $\partial\Omega$, then $\operatorname{div} \sigma = 0$.

Obviously, if one considers the marginal $\bar{\sigma} = \int_{\mathbb{S}^1} \sigma^x \in \mathcal{M}^1(\Omega; \mathbb{R}^2)$ defined by

$$\int_{\Omega \times \mathbb{S}^1} (\psi, 0) \cdot \sigma = \int_{\Omega} \psi \cdot \bar{\sigma}$$

for any $\psi \in C_c(\Omega; \mathbb{R}^2)$, then it also has zero divergence (as it vanishes if $\psi = \nabla\phi$ for some ϕ). In dimension 2, it follows that (assuming Ω is connected) there exists a BV function u such that $Du^\perp = \bar{\sigma}$. In our case, u is the characteristic function of a set E with $\partial E \cap \Omega = \gamma([0, T]) \cap \Omega$.

Generalization to BV functions

One can define for any $u \in BV(\Omega)$

$$F(u) = \inf \left\{ \int_{\Omega \times \mathbb{S}^1} h(\theta, \sigma) : \operatorname{div} \sigma = 0, \int_{\mathbb{S}^1} \sigma^x = Du^\perp \right\}.$$

If we assume that $f(t) \geq \sqrt{1+t^2}$, then one sees that $\bar{h}(s, t) \geq \sqrt{s^2+t^2}$ and $\int_{\Omega \times \mathbb{S}^1} h(\theta, \sigma) \geq \int_{\Omega \times \mathbb{S}^1} |\sigma|$. It easily follows that the “inf” is a min, and that F defines a convex, lower semicontinuous function on BV with $F(u) \geq |Du|(\Omega)$.

From the example above, we readily see that if E is a C^2 set, then

$$F(\chi_E) \leq \int_{\partial E} f(\kappa) d\mathcal{H}^1.$$

Tightness of the representation

We can show the following result:

Theorem if E is a C^2 set, then

$$F(\chi_E) = \int_{\partial E} f(\kappa) d\mathcal{H}^1.$$

Proof: we need to show \geq . In other words, we need to show the obvious fact that if σ is a measure with $\int_{\mathbb{S}^1} \sigma^x = D\chi_E^\perp$, then σ , above ∂E , consists at least in the measure defined by the lifted curve above ∂E (with its orientation as third component).

Maybe there is a simple way to do this (as it is obvious). We used S. Smirnov's theorem which shows that if σ is a measure with $\operatorname{div} \sigma = 0$, then it is a superposition of curves.

Smirnov's Theorem A (1994)

If $\operatorname{div} \sigma = 0$ then it can be decomposed in the following way:

$$\sigma = \int_{\mathfrak{C}_1} \lambda d\mu(\lambda), \quad |\sigma| = \int_{\mathfrak{C}_1} |\lambda| d\mu(\lambda),$$

where λ are of the form

$$\lambda_\gamma = \tau_\gamma \mathcal{H}^1 \llcorner \gamma$$

for rectifiable (possibly closed) curves $\gamma \subset \Omega \times \mathbb{S}^1$ of length at most one.
(\mathfrak{C}_1 is the corresponding set.)

[We do not need here the more precise "Theorem B"]

Smirnov's Theorem A (1994)

Thanks to the fact that the decomposition is convex (ie with $|\sigma| = \int_{\mathfrak{C}_1} |\lambda| d\mu(\lambda)$) we can show that $|\sigma|$ -a.e., for μ -a.e. curve λ one has $\sigma/|\sigma| = \lambda/|\lambda|$ $|\lambda|$ -a.e., and in particular λ^x is oriented along θ , and

$$\int_{\Omega \times \mathbb{S}^1} h(\theta, \sigma) = \int_{\mathfrak{C}_1} \left(\int_{\Omega \times \mathbb{S}^1} h(\theta, \lambda) \right) d\mu(\lambda) = \int_{\mathfrak{C}_1} \left(\int_{\gamma} h(\theta, \tau_{\gamma}) d\mathcal{H}^1 \right) d\mu(\lambda_{\gamma}).$$

The horizontal projection λ^x is a rectifiable curve, and one can deduce that its curvature is a bounded measure.

For this we reparametrize λ with the length of λ^x : that is we define $\tilde{\lambda}(t) = \lambda(s(t))$ in such a way that $\mathcal{H}^1(\tilde{\lambda}^x([0, t])) = t$ [if simple]. Then we show that $\tilde{\lambda}^{\theta}(t)$, which is the orientation of the tangent [because the energy is finite], has bounded variation.

Tightness

Then one can show that if

$$\Gamma^+ = \{x \in \partial E \cap \lambda^x(0, L) : \text{the curves have the same orientation} \}$$

then a.e. on Γ^+ , the absolutely continuous part of the curvature $\kappa = \dot{\tilde{\lambda}}^\theta$ coincides with κ_E . Using that for any set I ,

$$\int_{\lambda^x(I)} f(\kappa^a) \leq \int_{I \times \mathbb{S}^1} h(\theta, \lambda),$$

which more or less follows because this is precisely the way we have built h , we can deduce since $\kappa^a = \kappa_E$ a.e.:

$$\int_{\partial E} f(\kappa_E) d\mathcal{H}^1 = \int_{\mathfrak{c}_1} \int_{\partial E \cap \lambda^x} f(\kappa^a) d\mu(\lambda) \leq \int_{\mathfrak{c}_1} \int_{\partial E \times \mathbb{S}^1} h(\theta, \lambda) d\mu(\lambda)$$

which implies our inequality.

Tightness

- ▶ More cases?
- ▶ We know that F can be below the standard (L^1) relaxation of $\int_{\partial E} f(\kappa)$ (Bellettini-Mugnai 04/05, Dayrens-Masnou 16) (simple examples).

Dual representation

We can compute the dual problem of

$$F(u) = \inf \left\{ \int_{\Omega \times \mathbb{S}^1} h(\theta, \sigma) : \operatorname{div} \sigma = 0, \int_{\mathbb{S}^1} \sigma^x = Du^\perp \right\}.$$

by the standard perturbation technique, which consists in defining

$$G(p) = \inf \left\{ \int_{\Omega \times \mathbb{S}^1} h(\theta, \sigma + p) : \operatorname{div} \sigma = 0, \int_{\mathbb{S}^1} \sigma^x = Du^\perp \right\},$$

showing (exactly as for F) that $p \mapsto G(p)$ is (weakly-*) lsc and therefore that $G^{**} = G$, and in particular

$$F(u) = G(0) = \sup_{\eta \in C_0^0(\Omega \times \mathbb{S}^1; \mathbb{R}^3)} -G^*(\eta)$$

Dual representation

Then, it remains to compute $G^*(\eta)$:

$$\begin{aligned} G^*(\eta) &= \sup_{\substack{\rho, \sigma: \operatorname{div} \sigma = 0 \\ \int_{\mathbb{S}^1} \sigma = Du^\perp}} \int_{\Omega \times \mathbb{S}^1} \eta \cdot \rho - h(\theta, \sigma + \rho) \\ &= \sup_{\substack{\sigma: \operatorname{div} \sigma = 0 \\ \int_{\mathbb{S}^1} \sigma = Du^\perp}} - \int_{\Omega \times \mathbb{S}^1} \eta \cdot \sigma + \sup_{\rho} \eta \cdot (\sigma + \rho) - h(\theta, \sigma + \rho) \end{aligned}$$

We find $\underline{\theta} \cdot \eta^x + f^*(\eta^\theta) \leq 0$, and then $\eta = \psi(x) + \nabla \varphi(x, \theta)$ so that:

$$F(u) = \sup \left\{ \int_{\Omega} \psi \cdot Du^\perp : \psi \in C_c^0(\Omega; \mathbb{R}^2), \right. \\ \left. \exists \varphi \in C_c^1(\Omega \times \mathbb{S}^1), \underline{\theta} \cdot (\nabla_x \varphi + \psi) + f^*(\partial_\theta \varphi) \leq 0 \right\}.$$

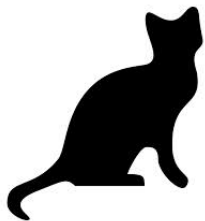
→ SAME as Bredies-Pock-Wirth' 2015... This is how we find out that this representation is a simpler variant of theirs...

Numerical discretization

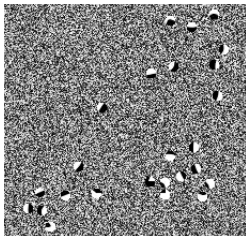
This is work in progress. We have a few approaches which work in theory but yield poorly concentrated measures σ . And better approaches which are not clearly justified.

We use both the primal and dual representation and solve the discretized problem using a saddle-point optimisation.

Examples: shape completion



(a) Original shape



(b) Input



(c) Inpainted shape

Figure : Weickert's cat: Shape completion using the function $f_2 = \sqrt{1 + k|\kappa|^2}$.

Examples: shape denoising



Examples: shape denoising



(a) AC, $\lambda = 8$



(b) AC, $\lambda = 4$



(c) AC, $\lambda = 2$



(d) EL, $\lambda = 8$



(e) EL, $\lambda = 4$

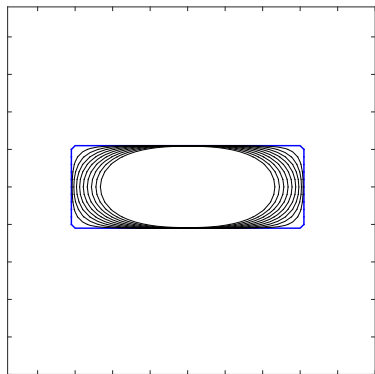


(f) EL, $\lambda = 2$

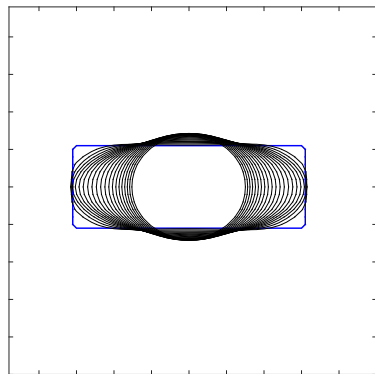
Figure : Shape denoising: First row: Using the function $f_1 = 1 + k|\kappa|$, second row: Using the function $f_3 = 1 + k|\kappa|^2$.

Examples: Willmore flow

(cf for instance Dayrens-Masnou-Novaga 2016)



(a) AC



(b) EL

Figure : Motion by the gradient flow of different curvature depending energies. Energy $1 + |\kappa|$ gives the same as standard mean curvature flow for convex curves. Elastica/Willmore flow converges to a circle (shrinkage is still present due to the length term).

Conclusion, perspectives

- ▶ We have introduced a relatively simple systematic way to represent curvature-dependent energies in $2D$;
- ▶ It simplifies the (energetically equivalent) framework of [Bredies-Pock-Wirth 15];
- ▶ Open questions: characterize the functions for which the relaxation is tight (conjecture: functions with “continuous” curvature?);
- ▶ Discretization needs some improvement (issues: measure with orientation constraint).

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Thank you for your attention