

Approximations of the The Mumford-Shah Functional for Unit Vector Fields

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Warwick, Jan 2009

Introduction

The Functional
Motivation

Algorithms

Splitting & Projection
Splitting & Penalisation
Penalisation without Splitting

Examples

Comparison: Projection vs. Penalisation
Comparison: CB vs. RGB

Vectorial Mumford-Shah Functional

(sphere valued functions)

For $\Omega \subseteq \mathbb{R}^d$, $d \geq 2$, find

$$(\mathbf{v}, \Gamma) := \underset{\substack{K \subset \Omega \text{ closed,} \\ \mathbf{u} \in H^1(\Omega \setminus K, \mathbb{S}^2)}}{\operatorname{argmin}} E(\mathbf{u}, K),$$

$$E(\mathbf{u}, K) := \frac{\alpha}{2} \int_{\Omega \setminus K} |\nabla \mathbf{u}|^2 d\mathbf{x} + \frac{\gamma}{2} \int_{\Omega} |\mathbf{u} - \mathbf{g}|^2 d\mathbf{x} + \beta \mathcal{H}^{d-1}(K),$$

$\mathbf{g} \in L^\infty(\Omega, \mathbb{S}^2)$: original image, K : edge set (closed).

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3 Terms: **smoothing**, **fidelity**, **edge length**.

Weak Formulation

(sphere valued functions)

For $\mathbf{g} \in L^\infty(\Omega; \mathbb{S}^2)$, $\mathbf{u} \in SBV(\Omega, \mathbb{S}^2)$, minimise

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Thm: (Carriero-Leaci, 91): \exists solution \mathbf{u} , and $(\mathbf{u}, \bar{S}_{\mathbf{u}})$ solves original problem.

Vectorial Ambrosio-Tortorelli Functional

(sphere valued functions)

For $\mathbf{u}, \mathbf{g} \in H^1(\Omega; \mathbb{S}^2)$, $s \in H^1(\Omega, [0, 1])$, minimise

$$AT_\varepsilon(\mathbf{u}, s) := \frac{\alpha}{2} \int_\Omega (s^2 + k_\varepsilon) |\nabla \mathbf{u}|^2 dx + \frac{\gamma}{2} \int_\Omega |\mathbf{u} - \mathbf{g}|^2 dx \\ + \beta \int_\Omega \varepsilon |\nabla s|^2 + \frac{1}{4\varepsilon} (1 - s)^2 dx.$$

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Thm (Ambrosio-Tortorelli, 90/92):

$$0 < k_\varepsilon = o(\varepsilon) \implies AT_\varepsilon(\mathbf{u}, s) \xrightarrow{\Gamma} \bar{E}(\mathbf{u}, K).$$

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- ▶ With AT: Strong nonlinear coupling between \mathbf{u} and s .
- ▶ Extending work on harmonic maps to \mathbb{S}^2
(convex functional, non-convex constraint):
Alouges (97), Bartels (05), Bartels-Prohl (07).

Colour Image Segmentation

- ▶ Chromaticity & Brightness (CB) Colour Model:

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RGB image $\mathbf{g} : \Omega \rightarrow [0, 255]^3 \subset \mathbb{N}^3$:
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- ▶ Sources: Mumford-Shah (89)
Chan-Kang-Shen (01): TV, CB
Tang-Sapiro-Caselles (01): Harmonic flow, CB
Osher-Vese (04): p -harmonic flow, CB
Bartels-Prohl (06): PM, CB
Bellettini-Coscia (94), Bourdin (99): MS, grayscale

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- ▶ Simplified Ericksen's energy

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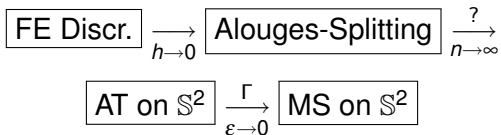
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- ▶ Sources: Lin (89), Lin-Luskin (89)
Virga (94), Alouges (97), Bartels-Prohl (07)

Algorithm 1: Splitting & Projection



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$$\boxed{\text{FE Discr.}} \xrightarrow{h \rightarrow 0} \boxed{\text{Alouges-Splitting}} \xrightarrow[n \rightarrow \infty]{?}$$
$$\boxed{\text{AT on } \mathbb{S}^2} \xrightarrow[\varepsilon \rightarrow 0]{\Gamma} \boxed{\text{MS on } \mathbb{S}^2}$$

Easy to solve numerically.
Exact sphere constraint.

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 - ▶ projecting $\mathbf{u} - \mathbf{w}$ never increases energy
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- ▶ 2nd step: \exists by convexity & coercivity, $0 \leq s \leq 1$ by max-principle (or testing with cutoff functions).

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- ▶ **Problem:** Cannot identify limit

$$\lim_{n \rightarrow \infty} \left(|\nabla \mathbf{u}_{n+1}|^2 \mathbf{s}_n, \varphi \right),$$

so cannot show (\mathbf{u}, \mathbf{s}) stationary point of $AT_\varepsilon(\cdot, \cdot)$.

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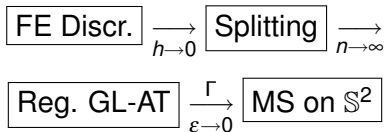
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- ▶ **Problem:** As before.

Algorithm 2: Splitting & Penalisation



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$$\boxed{\text{Reg. GL-AT}} \xrightarrow[\varepsilon \rightarrow 0]{\Gamma} \boxed{\text{MS on } \mathbb{S}^2}$$

Harder to solve numerically.
Approx. sphere constraint.

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- ▶ Identifying limits needs higher regularity of s , so add

$$\frac{\eta_\varepsilon}{\rho} \int_{\Omega} |\Delta s|^p \, d\mathbf{x}.$$

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- ▶ For $p > d$, $\mathbf{u}_n \rightarrow \mathbf{u}$ in H^1 , and $s_n \rightarrow s$ in L^∞ , so identifying limits works.

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- ▶ Sphere constraint only approximated.
- ▶ Two additional parameters.

Algorithm 3: Penalisation without Splitting

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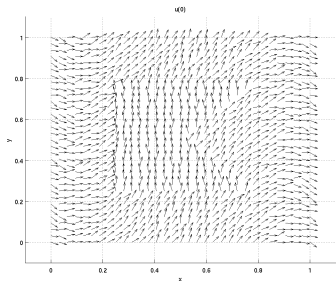
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- ▶ **Advantage**: No additional regularisation term needed.
- ▶ **Disadvantage**: No convergent algorithm known.

Projection



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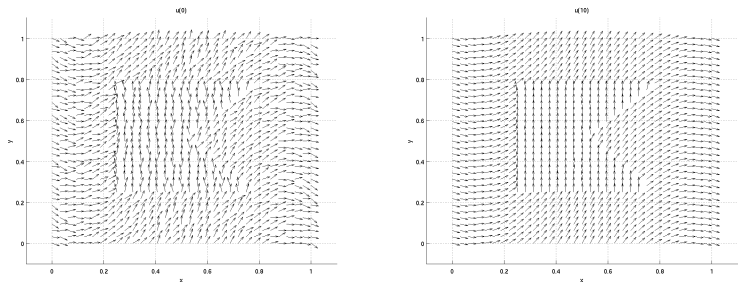
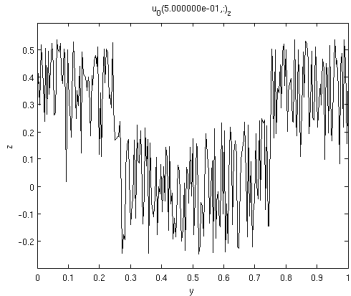


Figure: $\mathbf{U}_0 \equiv \mathbf{g}$ (left) and \mathbf{U}_{10} (right).

Projection



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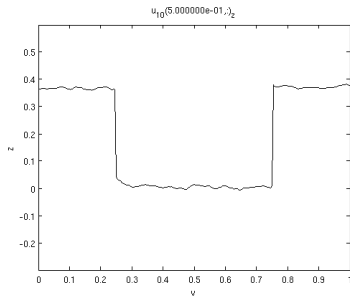
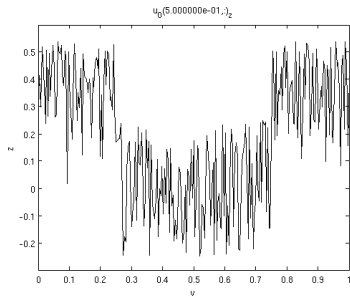
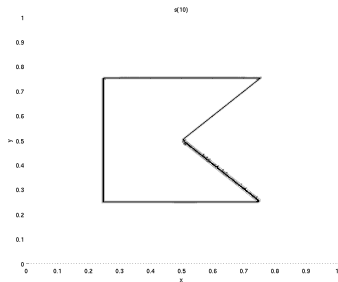


Figure: Section at $x = 0.5$ through $\mathbf{U}_0 \equiv \mathbf{g}$ and \mathbf{U}_{10} .

Projection vs. Penalisation



Projection vs. Penalisation

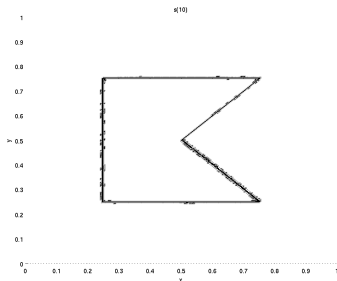
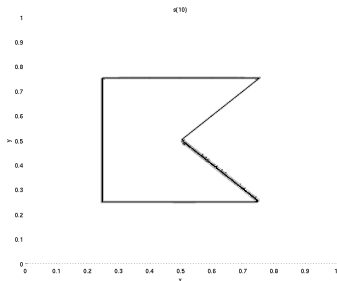
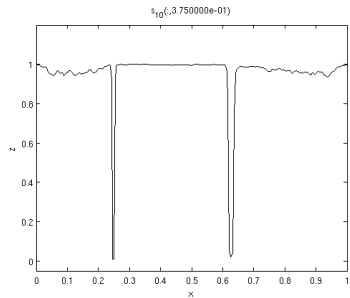


Figure: S_{10} , projection (left) and penalisation (right).

Projection vs. Penalisation



Projection vs. Penalisation

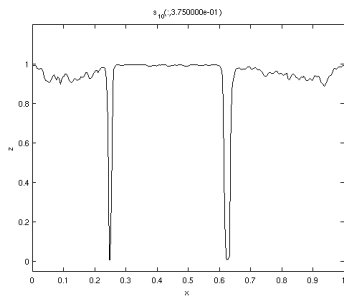
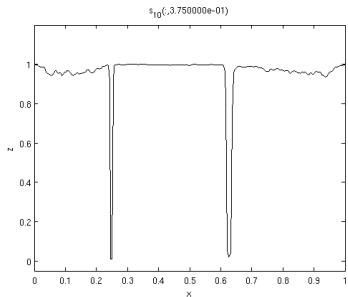


Figure: Section at $y = 0.375$ through S_{10} , projection (left) and penalisation (right).

Penalisation

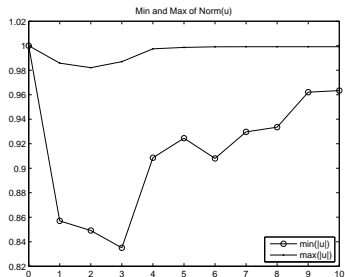


Figure: Min and max of $\|\mathbf{U}\|$.

CB Model

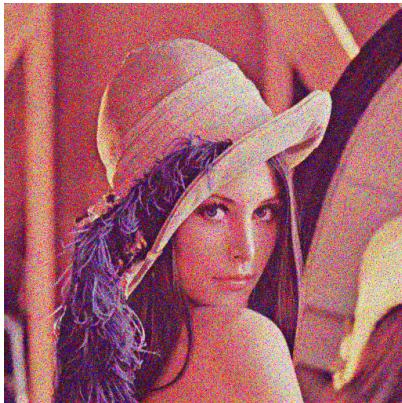
$$\begin{aligned} AT_\varepsilon(\mathbf{u}, v, s) &:= \frac{\alpha}{2} \int_{\Omega} (s^2 + k_\varepsilon) |\nabla \mathbf{u}|^2 d\mathbf{x} + \frac{\gamma}{2} \int_{\Omega} |\mathbf{u} - \mathbf{g}|^2 d\mathbf{x} \\ &+ \frac{\alpha_1}{2} \int_{\Omega} (s^2 + k_\varepsilon) |\nabla v|^2 d\mathbf{x} + \frac{\gamma_1}{2} \int_{\Omega} |v - b|^2 d\mathbf{x} \\ &+ \beta \int_{\Omega} \varepsilon |\nabla s|^2 + \frac{1}{4\varepsilon} (1 - s)^2 d\mathbf{x}, \end{aligned}$$

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- ▶ \mathbf{g}, \mathbf{u} : Original and iterate chromaticity,
- ▶ b, v : Original and iterate brightness,
- ▶ s : Joint edge-indicator.

CB

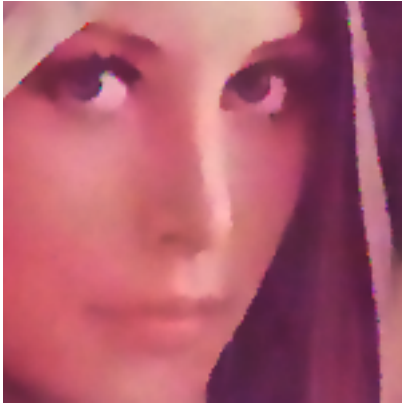


CB



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CB vs. RGB



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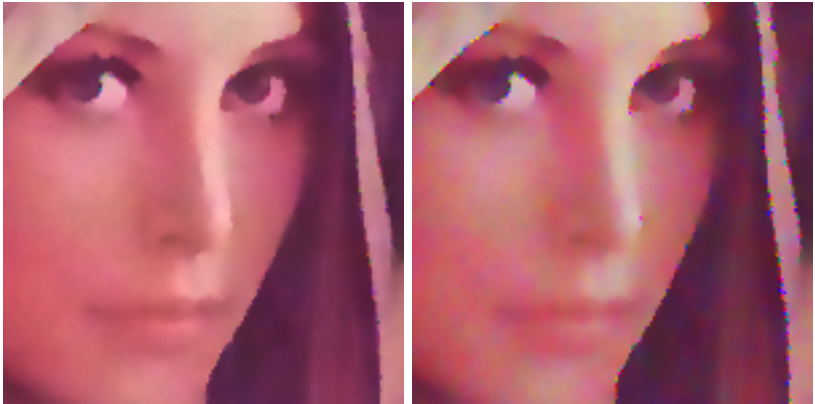


Figure: Crop of \mathbf{U}_{10} , CB (left) and RGB (right).

CB vs. RGB



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Thank You