

A predual formulation for PDE constrained optimization with total variation regularization

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Problem statement

Optimal control problem

$$(P) \quad \begin{cases} \min_{u \in \text{BV}(\Omega)} \frac{1}{2} \int_{\Omega} |y - z|^2 dx + \beta \int_{\Omega} |Du| dx \\ \text{s.t.} \quad Ay = u \end{cases}$$

- A is partial differential operator (with b.c.) on $\Omega \subset \mathbb{R}^d$
- Assumption: A is isometry between state and control space (otherwise add $L^2(\Omega)$ regularization for u)
- Motivation: Control cost proportional to control changes

Bounded Variation

Functions of bounded variation

$$\text{BV}(\Omega) = \left\{ u \in L^2(\Omega); \int_{\Omega} |Du| dx < \infty \right\}$$

$$\int_{\Omega} |Du| dx = \sup \left\{ \int_{\Omega} u \operatorname{div} \vec{v} dx; \vec{v} \in (C_0^\infty(\Omega))^d, |\vec{v}|_{\ell^\infty} \leq 1 \right\}$$

- Nondifferentiable, but convex
- Optimality conditions via Fenchel duality
- Dual of $\text{BV}(\Omega)$: space of measures
- $\text{BV}(\Omega)$ not reflexive \Rightarrow use predual of $\text{BV}(\Omega)$ (Hilbert space)

Formal Fenchel duality

Set

- $\Lambda(u) = Du$
- $\mathcal{F}(u) = \frac{1}{2} \int_{\Omega} |A^{-1}u - z|^2 dx$
- $\mathcal{G}(\vec{q}) = \beta \int_{\Omega} |\vec{q}| dx$

Fenchel duality

Problem (P) equivalent to

$$(P^*) \quad \sup -\mathcal{F}^*(\Lambda^* u^*) - \mathcal{G}^*(-u^*)$$

Solutions \bar{u}, \bar{u}^* related by

$$\Lambda^* \bar{u}^* = \mathcal{F}'(\bar{u})$$

Formal Fenchel dual

Fenchel conjugates:

- $\Lambda^*(u^*) = -\operatorname{div} u^*$
- $\mathcal{F}^*(p^*) = \frac{1}{2} \|A^* p^*\|^2 + \langle A^* p^*; z \rangle$
- $\mathcal{G}^*(\vec{q}^*) = I_\beta := \begin{cases} 0 & \text{if } |\vec{q}^*(x)|_{\ell^\infty} \leq \beta \text{ for a.e. } x \in \Omega, \\ \infty & \text{otherwise.} \end{cases}$

Formal dual problem

$$(P^*) \quad \begin{cases} \inf_{u^*} \frac{1}{2} \|A^* \operatorname{div} u^*\|^2 - \langle A^* \operatorname{div} u^*, z \rangle \\ \text{s.t. } |u^*|_{\ell^\infty} \leq \beta \text{ for a.e. } x \in \Omega. \end{cases}$$

Primal solution from extremality relations:

$$u = Az - AA^* \operatorname{div} u^*$$

Prudal Problem

Constant functions in kernel of div : Add $L^2(\Omega)$ regularization for u^* :

Regularized prudal problem (P_γ^*)

$$\left\{ \begin{array}{l} \min_{u^* \in H_{\text{div}}^A} \frac{1}{2} \|A^* \text{div } u^*\|_{L^2(\Omega)}^2 + \frac{\gamma}{2} \|u^*\|_{(L^2(\Omega))^d}^2 - \langle A^* \text{div } u^*, z \rangle_{L^2(\Omega)} \\ \text{s.t. } |u^*|_{\ell^\infty} \leq \beta \text{ for a.e. } x \in \Omega. \end{array} \right.$$

$$H_{\text{div}}^A := \{ \vec{v} \in (L^2(\Omega))^d ; \text{div } \vec{v} \in \mathcal{D}(A^*), \vec{v} \cdot \nu = 0 \text{ on } \partial\Omega \}$$

$\Rightarrow (P_\gamma^*)$ has unique solution for $\gamma > 0$, dual of (P_0^*) is (P)

Pre dual Regularization

Pre dual problem: Minimization problem with box constraints

⇒ solve with semismooth Newton method

- Insufficient regularity of Lagrange multipliers for box constraints
⇒ Moreau-Yosida-regularization of box constraint
- $\max(0, \cdot), \min(0, \cdot) : L^q \rightarrow L^p$ Newton diff'able only if $q > p$
⇒ add H^1 -regularization for u^*

Optimality System

Regularized prudal problem

$$\begin{aligned} \min_{u^* \in (H_0^1(\Omega))^d \cap H_{\text{div}}^A} & \frac{1}{2} \|A^* \operatorname{div} u^*\|_{L^2(\Omega)}^2 - \langle A^* \operatorname{div} u^*, z \rangle_{L^2(\Omega)} \\ & + \frac{\gamma}{2} \|u^*\|_{L^2(\Omega)}^2 + \frac{\delta}{2} \|\nabla u^*\|_{L^2(\Omega)}^2 \\ & + \frac{1}{2c} |\max(0, c(u^* - \vec{\beta}))|^2 + \frac{1}{2c} |\min(0, c(u^* + \vec{\beta}))|^2 \end{aligned}$$

Optimality conditions

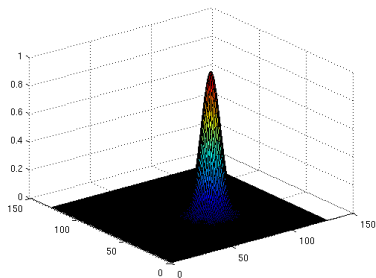
$$\begin{aligned} (-\operatorname{div})^* A A^* \operatorname{div} \bar{u}^* + \gamma \bar{u}^* - \delta \Delta \bar{u}^* \\ + \max(0, c(\bar{u}^* - \vec{\beta})) + \min(0, c(\bar{u}^* + \vec{\beta})) = (-\operatorname{div})^* A z \end{aligned}$$

Numerical Examples

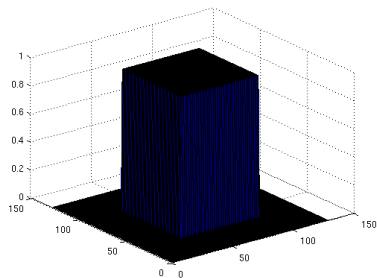
- Model Problem $Ay = y - \Delta y$ in $\Omega = [-1, 1]^2 \subset \mathbb{R}^2$ with Neumann boundary conditions
- Discretization by finite differences $((-\operatorname{div})_h = (\nabla_h)^T)$
- Comparison with $H^1(\Omega)$ -regularized problem

$$\begin{cases} \min_{u \in H^1(\Omega)} \frac{1}{2} \int_{\Omega} |y - z|^2 dx + \frac{\beta}{2} \int_{\Omega} |\nabla u|^2 dx \\ \text{s.t. } Ay = u \end{cases}$$

Target state

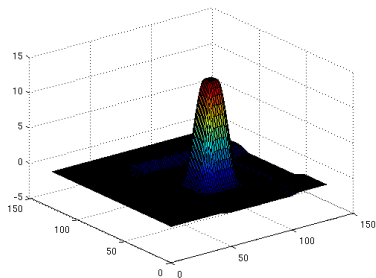


$$z_a(x, y) = e^{-50[(x-0.2)^2+(y+0.1)^2]}$$

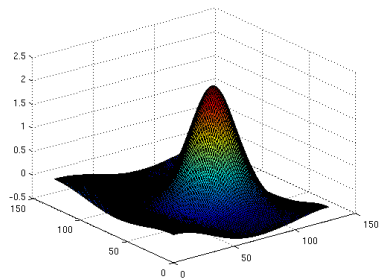


$$z_b(x, y) = \chi_{\{|x| < \frac{1}{2}, |y| < \frac{1}{2}\}}$$

Optimal control for target z_a : BV vs. H^1 penalty



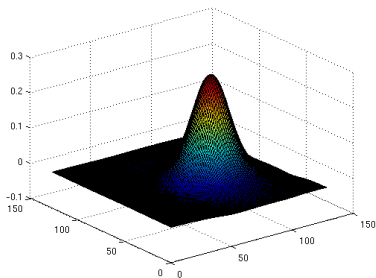
BV(Ω)



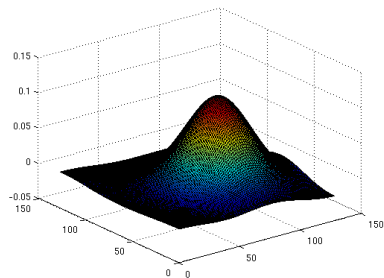
$H^1(\Omega)$

$$(\beta = 10^{-4}, \gamma = \delta = 10^{-1}, c = 10^6)$$

Optimal state for target z_a : BV vs. H^1 penalty



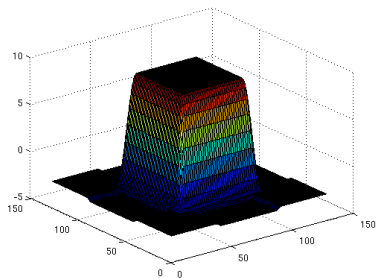
BV(Ω)



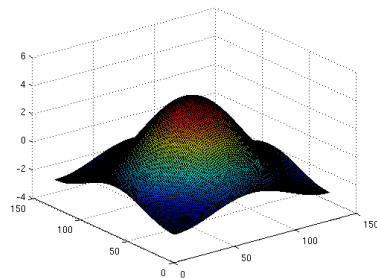
$H^1(\Omega)$

$$(\beta = 10^{-4}, \gamma = \delta = 10^{-1}, c = 10^6)$$

Optimal control for target z_b : BV vs. H^1 penalty



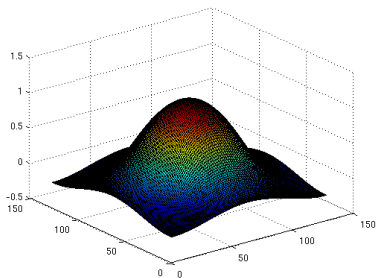
BV(Ω)



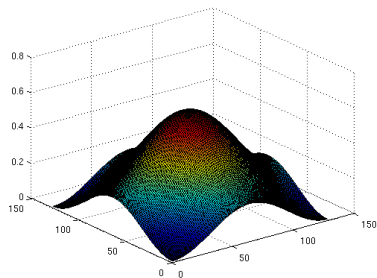
$H^1(\Omega)$

$$(\beta = 10^{-3}, \gamma = \delta = 10^{-1}, c = 10^6)$$

Optimal state for target z_b : BV vs. H^1 penalty



BV(Ω)



$H^1(\Omega)$

$$(\beta = 10^{-3}, \gamma = \delta = 10^{-1}, c = 10^6)$$

Comparison with L^1 -Regularization

Cost of control proportional to magnitude of control: L^1 -Penalty

Optimal control problem

$$\begin{cases} \min_{u \in L^1(\Omega)} \frac{1}{2} \int_{\Omega} |y - z|^2 dx + \beta \int_{\Omega} |u| dx \\ \text{s.t.} \quad Ay = u \end{cases}$$

Comparison with L^1 -Regularization

Set $\Lambda(y, u) = (0, u) \Rightarrow \Lambda^*(y^*, u^*) = (0, u^*)$, use Fenchel duality

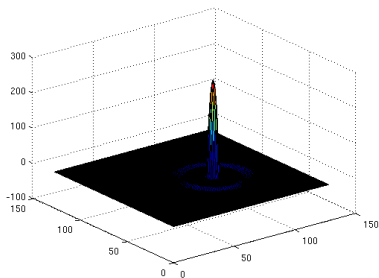
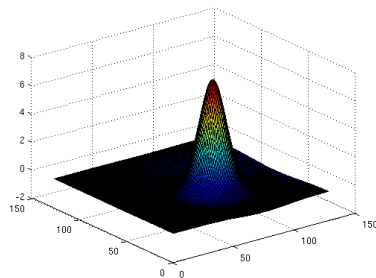
Pre dual problem

$$\begin{cases} \inf_{u^*} \frac{1}{2} \|A^* u^*\|_{L^2(\Omega)}^2 + \langle A^* u^*, z \rangle_{L^2(\Omega)} \\ \text{s.t. } |u^*| \leq \beta \text{ for a.e. } x \in \Omega. \end{cases}$$

Optimality conditions of Moreau-Yosida regularization

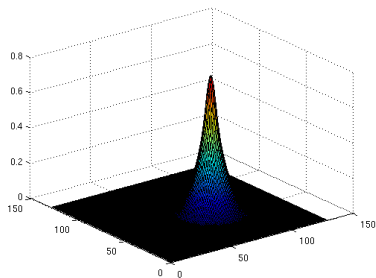
$$AA^* \bar{u}^* + \max(0, c(\bar{u}^* - \vec{\beta})) + \min(0, c(\bar{u}^* + \vec{\beta})) = Az$$

Optimal control for target z_a : L^1 vs. L^2 penalty

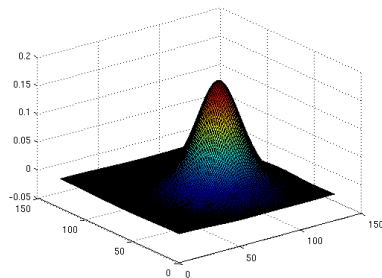
 $L^1(\Omega)$  $L^2(\Omega)$

$$(\beta = 10^{-3}, c = 10^6)$$

Optimal state for target z_a : L^1 vs. L^2 penalty



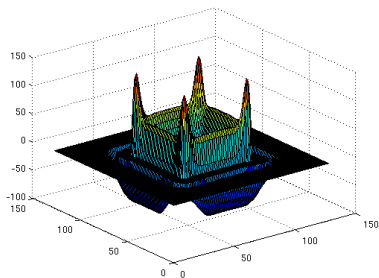
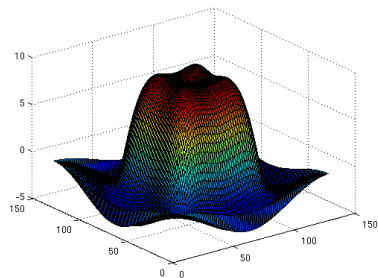
$L^1(\Omega)$



$L^2(\Omega)$

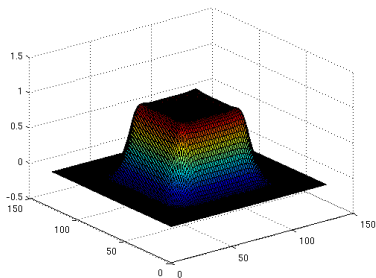
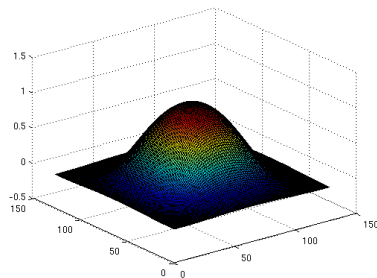
$$(\beta = 10^{-3}, c = 10^6)$$

Optimal control for target z_b : L^1 vs. L^2 penalty

 $L^1(\Omega)$  $L^2(\Omega)$

$$(\beta = 10^{-3}, c = 10^6)$$

Optimal state for target z_b : L^1 vs. L^2 penalty

 $L^1(\Omega)$  $L^2(\Omega)$

$$(\beta = 10^{-3}, c = 10^6)$$