

Minimum effort optimal control of elliptic PDEs

Christian Clason¹ Kazufumi Ito² Karl Kunisch¹

¹Institute for Mathematics and Scientific Computing, Karl-Franzens-Universität Graz

²Department of Mathematics, North Carolina State University

25th IFIP TC7 Conference
Berlin, September 15, 2011

- 1 Problem formulation
- 2 Regularization
- 3 Numerical solution
 - Semi-smooth Newton method
 - Parameter choice
- 4 Numerical examples

Minimum effort control problem

$$\begin{cases} \min_{u \in L^\infty} \frac{1}{2} \|y - z\|_{L^2}^2 + \frac{\alpha}{2} \|u\|_{L^\infty}^2 \\ \text{subject to } Ay = u \end{cases}$$

- $\Omega \subset \mathbb{R}^n$
- A linear elliptic operator with boundary conditions
- $z \in L^2(\Omega)$ given target
- Minimize maximal control magnitude (e.g., system design)
- Well-studied for ODEs [Neustadt '62], very little for PDEs

Minimum effort control problem

Reformulation

$$(P) \quad \left\{ \begin{array}{l} \min_{c \in \mathbb{R}, u \in L^\infty} \frac{1}{2} \|y - z\|_{L^2}^2 + \frac{\alpha}{2} c^2 \\ \text{subject to } Ay = u \\ \|u\|_{L^\infty} \leq c \end{array} \right.$$

- Equivalent formulation
- Different from control constraints
- Related to sup-norm minimization of state
[Grund/Rösch '01, Prüfert/Schiela '09, C/I/K '10]

Minimum effort control problem

Reformulation

$$(\mathcal{P}) \quad \begin{cases} \min_{c \in \mathbb{R}, u \in L^\infty} \frac{1}{2} \|y - z\|_{L^2}^2 + \frac{\alpha}{2} c^2 \\ \text{subject to } Ay = cu \\ \|u\|_{L^\infty} \leq 1 \end{cases}$$

- Equivalent formulation
- Existence of minimizer $(y^*, u^*, c^*) \in H_0^1(\Omega) \times L^\infty(\Omega) \times \mathbb{R}_+$
- Uniqueness iff $c^* \neq 0$ (iff $J(y^*, c^*) < \frac{1}{2} \|z\|_{L^2}^2$)

Optimality system

Standard subdifferential calculus yields existence of $p^* \in H_0^1(\Omega)$,

$$\text{(OS)} \quad \left\{ \begin{array}{l} \langle -p^*, u - u^* \rangle_{L^2} \geq 0 \quad \text{for all } \|u\|_{L^\infty} \leq 1, \\ \alpha c^* - \langle u^*, p^* \rangle_{L^2} = 0, \\ y^* - z + A^* p^* = 0, \\ Ay^* - c^* u^* = 0. \end{array} \right.$$

Optimality system

$$\text{(OS)} \quad \begin{cases} \langle -p^*, u - u^* \rangle_{L^2} \geq 0 & \text{for all } \|u\|_{L^\infty} \leq 1, \\ \alpha c^* - \langle u^*, p^* \rangle_{L^2} = 0, \\ y^* - z + A^* p^* = 0, \\ Ay^* - c^* u^* = 0. \end{cases}$$

pointwise inspection:

$$u^*(x) = \text{sign}(p^*(x)) = \begin{cases} 1 & \text{if } p^*(x) > 0, \\ -1 & \text{if } p^*(x) < 0, \\ t \in [-1, 1] & \text{if } p^*(x) = 0 \end{cases}$$

Reduced optimality system

$$(OS') \quad \begin{cases} AA^* p^* + c^* \text{sign}(p^*) = Az, \\ \alpha c^* - \|p^*\|_{L^1} = 0. \end{cases}$$

$$\text{sign}(p)(x) = \begin{cases} 1 & \text{if } p(x) > 0, \\ -1 & \text{if } p(x) < 0, \\ t \in [-1, 1] & \text{if } p(x) = 0 \end{cases}$$

↪ **not differentiable** (even in generalized sense)

Regularized problem

$$(\mathcal{P}_\beta) \quad \begin{cases} \min_{c \in \mathbb{R}_+, u \in L^\infty} \frac{1}{2} \|y - z\|_{L^2}^2 + \frac{\beta c}{2} \|u\|_{L^2}^2 + \frac{\alpha}{2} c^2 \\ \text{subject to } Ay = cu \quad \text{in } \Omega, \\ \|u\|_{L^\infty} \leq 1. \end{cases}$$

- **Existence** of minimizer $(y_\beta, u_\beta, c_\beta) \in H_0^1(\Omega) \times L^\infty(\Omega) \times \mathbb{R}_+$ from standard arguments ($c_\beta > 0$ assumed)
- **Uniqueness** for α sufficiently large; uniqueness for any α if u_β or c_β given (**bilinear structure**)
- Motivation: Huber-type smoothing of reduced optimality system

Regularized optimality system

$$(\text{OS}_\beta) \quad \left\{ \begin{array}{l} \langle \beta u_\beta - p_\beta, u - u_\beta \rangle_{L^2} \geq 0 \quad \text{for all } \|u\|_{L^\infty} \leq 1, \\ \alpha c_\beta + \frac{\beta}{2} \|u_\beta\|_{L^2}^2 - \langle u_\beta, p_\beta \rangle_{L^2} = 0, \\ y_\beta - z + A^* p_\beta = 0, \\ Ay_\beta - c_\beta u_\beta = 0. \end{array} \right.$$

Regularized optimality system

$$(\text{OS}_\beta) \quad \left\{ \begin{array}{l} \langle \beta u_\beta - p_\beta, u - u_\beta \rangle_{L^2} \geq 0 \quad \text{for all } \|u\|_{L^\infty} \leq 1, \\ \alpha c_\beta + \frac{\beta}{2} \|u_\beta\|_{L^2}^2 - \langle u_\beta, p_\beta \rangle_{L^2} = 0, \\ y_\beta - z + A^* p_\beta = 0, \\ Ay_\beta - c_\beta u_\beta = 0. \end{array} \right.$$

pointwise inspection:

$$u_\beta(x) = \text{sign}_\beta(p_\beta(x)) := \begin{cases} 1 & \text{if } p_\beta(x) > \beta, \\ -1 & \text{if } p_\beta(x) < -\beta, \\ \frac{1}{\beta} p_\beta(x) & \text{if } |p_\beta(x)| \leq \beta \end{cases}$$

Reduced optimality system

$$(OS') \quad \begin{cases} AA^* p_\beta + c_\beta \operatorname{sign}_\beta(p_\beta) = Az, \\ \alpha c_\beta - \|p_\beta\|_{L^1_\beta} = 0. \end{cases}$$

$$\operatorname{sign}_\beta(p)(x) := \begin{cases} 1 & \text{if } p(x) > \beta, \\ -1 & \text{if } p(x) < -\beta, \\ \frac{1}{\beta} p(x) & \text{if } |p(x)| \leq \beta \end{cases}$$

$$\|p\|_{L^1_\beta} := \int_{\Omega} |p(x)|_\beta \, dx, \quad |p(x)|_\beta := \begin{cases} p(x) - \frac{\beta}{2} & \text{if } p(x) > \beta, \\ -p(x) - \frac{\beta}{2} & \text{if } p(x) < -\beta, \\ \frac{1}{2\beta} p(x)^2 & \text{if } |p(x)| \leq \beta. \end{cases}$$

Convergence

Lemma (Monotonicity)

For $\beta < \beta'$ and $J(y, c) = \frac{1}{2} \|y - z\|_{L^2}^2 + \frac{\alpha}{2} c^2$, we have

$$c_{\beta'} \|u_{\beta'}\|_{L^2}^2 \leq c_{\beta} \|u_{\beta}\|_{L^2}^2$$

$$J(y_{\beta}, c_{\beta}) \leq J(y_{\beta'}, c_{\beta'})$$

Theorem (Convergence)

As $\beta \rightarrow 0$, subsequence of $(y_{\beta}, u_{\beta}, c_{\beta}) \rightarrow (y^*, u^*, c^*)$

- weakly- \star in $H_0^1(\Omega) \times L^{\infty}(\Omega) \times \mathbb{R}_+$
- strongly in $H_0^1(\Omega) \times L^q(\Omega) \times \mathbb{R}_+$ for any $q \in [1, \infty)$

Convergence

Theorem (Convergence)

As $\beta \rightarrow 0$, subsequence of $(y_\beta, u_\beta, c_\beta) \rightarrow (y^*, u^*, c^*)$

- weakly- \star in $H_0^1(\Omega) \times L^\infty(\Omega) \times \mathbb{R}_+$
- strongly in $H_0^1(\Omega) \times L^q(\Omega) \times \mathbb{R}_+$ for any $q \in [1, \infty)$

Corollary (Convergence rate)

For $\beta \rightarrow 0$ and $J(y, c) = \frac{1}{2} \|y - z\|_{L^2}^2 + \frac{\alpha}{2} c^2$, we have

$$J(y_\beta, c_\beta) - J(y^*, c^*) = o(\beta)$$

Semi-smooth Newton method

Consider (OS'_β) as $T(p, c) = 0$ for $T : H_0^2 \times \mathbb{R}_+ \rightarrow H^{-2}(\Omega) \times \mathbb{R}$,

$$T(p, c) = \begin{pmatrix} AA^*p + c \operatorname{sign}_\beta(p) - Az \\ \alpha c - \|p\|_{L^1_\beta} \end{pmatrix}$$

- T differentiable with respect to c
- $t \mapsto |t|_\beta$ differentiable, globally Lipschitz derivative $t \mapsto \operatorname{sign}_\beta(t)$
 \Rightarrow Nemytskii operators
 - $p \mapsto \|p\|_{L^1_\beta}$ differentiable on $L^p(\Omega)$, $p \geq 4$
 - $p \mapsto \operatorname{sign}_\beta(p)$ semi-smooth from $L^p(\Omega)$ to $L^q(\Omega)$, $p > q \geq 1$

$\Rightarrow T$ semi-smooth [Ulbrich '02, Schiela '08]

Semi-smooth Newton method

Newton derivatives

$$D_N(\|p\|_{L^1_\beta})h = \langle \text{sign}_\beta(p), h \rangle_{L^2}$$

$$(D_N \text{sign}_\beta(p)h)(x) = \begin{cases} 0, & \text{if } |p(x)| > \beta \\ \frac{1}{\beta}h(x), & \text{if } |p(x)| \leq \beta \end{cases}$$

Semi-smooth Newton step

$$\begin{pmatrix} AA^* \delta p + c^k \frac{1}{\beta} \chi_{\{|p^k(x)| \leq \beta\}} \delta p + \text{sign}_\beta(p^k) \delta c \\ \alpha \delta c - \langle \text{sign}_\beta(p^k), \delta p \rangle \end{pmatrix} = -T(p^k, c^k)$$

Semi-smooth Newton method

Semi-smooth Newton step

$$\begin{pmatrix} AA^* \delta p + c^k \frac{1}{\beta} \chi_{\{|p^k(x)| \leq \beta\}} \delta p + \text{sign}_\beta(p^k) \delta c \\ \alpha \delta c - \langle \text{sign}_\beta(p^k), \delta p \rangle \end{pmatrix} = -T(p^k, c^k)$$

Theorem

If A^ is positive definite, semi-smooth Newton method converges locally superlinearly to solution of (OS'_β) for every $\alpha, \beta > 0$.*

Regularization parameter choice

Local convergence only, radius shrinks with β

↪ Continuation scheme:

- 1 Set β_0 large, q_0 small, $t < 1$
 - 2 If Newton method converges (no change in active sets):
 - Accept solution, $n \leftarrow n + 1$
 - Decrease $\beta_n \leftarrow \beta_{n-1} q_m$
- else
- Reject solution, $m \leftarrow m + 1$
 - Increase $q_m \leftarrow q_{m-1}^t, \beta_n \leftarrow q_m \beta_{n-1}$

Stopping criterion?

Stopping criterion

- Observation: $\beta \mapsto c_\beta$ saturates
- Lemma: $\beta \mapsto c_\beta \|u_\beta\|_{L^2}^2$ monotonically increasing for $\beta \rightarrow 0$

\rightsquigarrow Choose efficiency level $0 \ll \mu < 1$, stop at β_{n^*} with

$$c_{\beta_{n^*}} \|u_{\beta_{n^*}}\|_{L^2}^2 > \mu c^* \|u^*\|_{L^2}^2$$

u^*, c^* unknown \rightsquigarrow model function approach

Stopping criterion

Model function

$$m(\beta) = \frac{K_1}{(K_2 + \beta)^2} \approx c_\beta \|u_\beta\|_{L^2}^2$$

- Motivated by simplified ODE model of optimality conditions
- K_1, K_2 from interpolation of two successive iterates β_n, β_{n-1}

\rightsquigarrow Choose $0 < \mu < 1$, stop at n^* with

$$c_{\beta_{n^*}} \|u_{\beta_{n^*}}\|_{L^2}^2 > \mu m_{n^*}(0)$$

Numerical examples

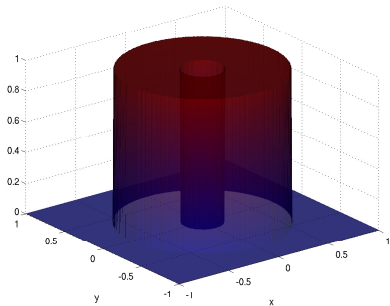
- Model problem: $\Omega = [-1, 1]^2$,

$$\begin{cases} -\nu \Delta y + \langle b, \nabla y \rangle = u, \\ y|_{\partial\Omega} = 0, \end{cases}$$

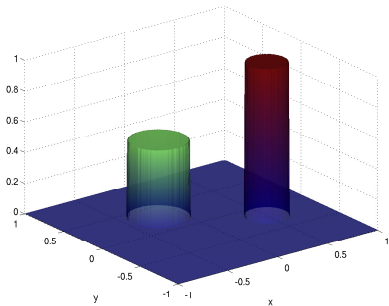
$$\nu = 10^{-1}, b = (-1, 0)^T$$

- Finite differences on uniform grid, $N = 256$ nodes
- $\beta_0 = 1, q_0 = 10^{-1}, t = 0.5, \mu = 0.99$, max. 10 SSN iterations
- Termination at $\beta_{n^*} \approx 2 \cdot 10^{-7}$

Examples: Targets

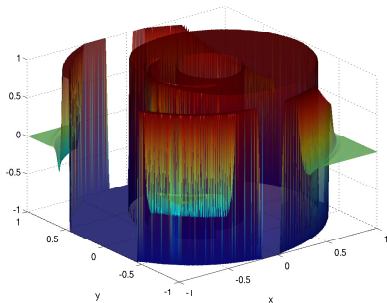


(a) z_1

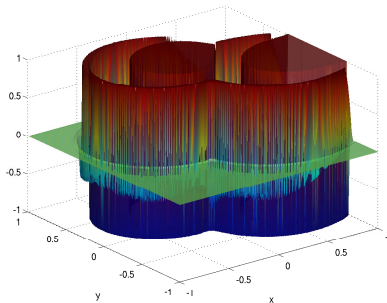


(b) z_2

Examples: Controls ($\alpha = 5 \cdot 10^{-3}$)

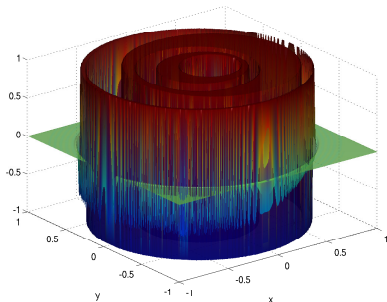


(a) $u_{1,\alpha}$ ($c_{1,\alpha} = 1.9622$)

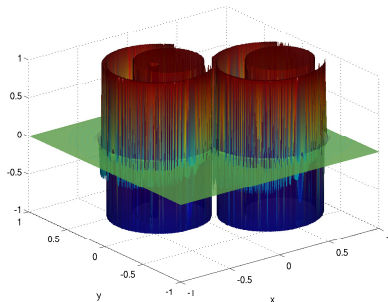


(b) $u_{2,\alpha}$ ($c_{2,\alpha} = 0.8788$)

Examples: Controls ($\alpha = 5 \cdot 10^{-4}$)

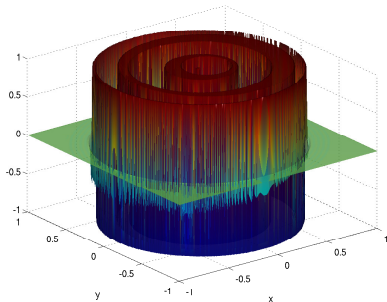


(a) $u_{1,\alpha}$ ($c_{1,\alpha} = 4.4236$)

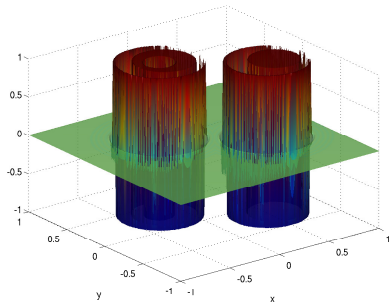


(b) $u_{2,\alpha}$ ($c_{2,\alpha} = 2.6066$)

Examples: Controls ($\alpha = 5 \cdot 10^{-5}$)

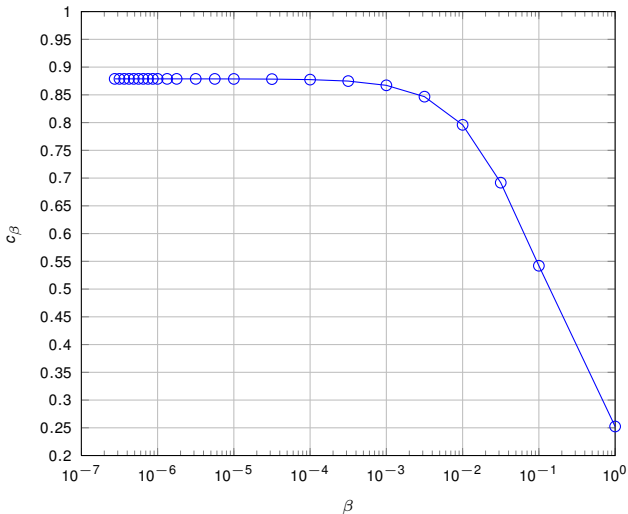


(a) $u_{1,\alpha}$ ($c_{1,\alpha} = 9.8518$)

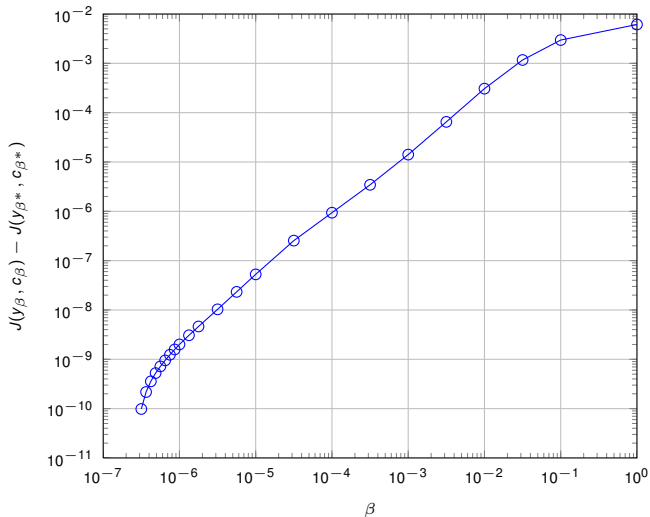


(b) $u_{2,\alpha}$ ($c_{2,\alpha} = 6.8161$)

Examples: Convergence c_β



Examples: Convergence $J(y_\beta, c_\beta)$



Conclusion

Minimum effort problem for PDEs:

- Non-differentiable (in generalized sense)
- Efficient approximation via semi-smooth Newton method with path following
- Regularized controls are “bang-zero-bang”: $\text{sign}_{\beta}(0) = 0$
- (Limit controls are true “bang-bang”: $\text{sign}(0) \neq 0$)

Preprint, MATLAB code:

<http://www.uni-graz.at/~clason/publications.html>