

Convergence Behavior of Barrier-Penalty Methods Applied to Optimal Control with PDEs

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Scope of the Talk

- Problem Formulation, Basic Properties
- Penalties for Control Bounds, Approximate Reduction
- Discretization of Reduced Problems
- Numerical Examples
- Outlook

1. Problem Formulation, Basic Properties

1.1. Distributed Controls (see G/Kunz/Meischner[3], G/Winkler[4])

$\Omega \subset \mathbb{R}^n$, $n = 2, 3$ bounded convex domain,
 $q, b, d \in L_\infty(\Omega)$ given.

Considered optimal control problem

$$\begin{aligned} \hat{J}(y, u) &:= \frac{1}{2} \int_{\Omega} (y - q)^2 + \frac{\alpha}{2} \int_{\Omega} u^2 \rightarrow \min! \\ \text{s.t. } -\Delta y &= u \quad \text{in } \Omega, \\ y + \frac{\partial y}{\partial n} &= 0 \quad \text{on } \Gamma := \partial\Omega, \\ u &\in U_{ad}, \end{aligned} \tag{1}$$

$\alpha > 0$ regularization parameter

$U_{ad} \subset L^2(\Omega)$ set of admissible controls (convex, closed).

$$U_{ad} := \{u \in L^2(\Omega) : u \leq b \quad \text{a.e. in } \Omega\} \quad (2)$$

in case of only control constraints - or

$$U_{ad} := \{u \in L^2(\Omega) : u \leq b, \quad y \leq d \quad \text{a.e. in } \Omega\} \quad (3)$$

in case of state and control constraints.

One-sided bounds only to simplify the presentation.

General assumption: $U_{ad} \neq \emptyset$.

State equations of (1) in the weak sense in $V := H^1(\Omega)$,
 $a(\cdot, \cdot) : V \times V \rightarrow \mathbb{R}$ defined by

$$a(y, v) := \int_{\Omega} \nabla y \cdot \nabla v + \int_{\Gamma} y v \quad \forall y, v \in V. \quad (4)$$

Lemma

For any $u \in U_{ad}$ there is a unique $y \in V$ such that

$$a(y, v) = (u, v) \quad \forall v \in V. \quad (5)$$

$Su := y$ and $V \hookrightarrow L^2(\Omega)$ defines a linear, continuous mapping $S : L^2(\Omega) \rightarrow L^2(\Omega)$.

Thus, problem (1) can be reduced to

$$J(u) := \hat{J}(Su, u) \rightarrow \min! \quad \text{s.t. } u \in U_{ad}. \quad (6)$$

U_{ad} is nonempty, closed and convex and J is continuous and strongly convex.

Remark

Since Ω is convex, the solution $y \in V$ of (5) has the additional regularity $y \in H^2(\Omega)$. Hence, S is a linear, continuous mapping from $L^2(\Omega)$ to $H^2(\Omega)$.

Theorem

Problem (6) possesses a unique optimal solution \bar{u} and $(S\bar{u}, \bar{u}) \in V \times U_{ad}$ is the unique optimal solution of (1). Further, system

$$\begin{aligned}(\bar{y} - q, y) - a(y, \bar{v}) &= 0 \quad \forall y \in V, \\ a(\bar{y}, v) - (\bar{u}, v) &= 0 \quad \forall v \in V, \\ \alpha(\bar{u}, u - \bar{u}) + (u - \bar{u}, \bar{v}) &\geq 0 \quad \forall u \in U_{ad}\end{aligned} \tag{7}$$

is necessary and sufficient for $\bar{u} \in U_{ad}$ to be optimal.

Remark

The inequality in system (7) is equivalent to

$$\bar{u} = P(\bar{u} - \sigma(\bar{v} + \alpha\bar{u})) \quad (8)$$

for any $\sigma > 0$, where P is the $L^2(\Omega)$ -ortho-projection onto U_{ad} . For sufficiently small $\sigma > 0$ the mapping

$$Tu := P(u - \sigma((S')^*(Su - q) + \alpha u)) \quad \forall u \in U \quad (9)$$

is a contraction. (Here $S' = S$)

Especially for $\sigma = 1/\alpha$ formula (8) yields $\bar{u} = P(-\frac{1}{\alpha}\bar{v})$ i.e., \bar{u} is expressed by the remaining variables (see Hinze[6]).

If only control constraints occur then P is easy to evaluate,
not so in case of state constraints!

1.2. Case of Boundary Controls (see G/Winkler[5])

Consider the semi-linear boundary control problem

$$\begin{aligned} \tilde{J}(y, u) &:= \frac{1}{2} \int_{\Omega} (y - q)^2 + \frac{\alpha}{2} \int_{\Gamma} u^2 \rightarrow \min! \\ \text{subject to} \quad -\Delta y + f(\cdot, y) &= 0 \quad \text{in } \Omega, \\ \frac{\partial y}{\partial n} + y &= u \quad \text{on } \Gamma, \quad u \in U_{ad}. \end{aligned} \tag{10}$$

with

$$U_{ad} := \{u \in L^2(\Gamma) : a \leq u \leq b \quad \text{a.e. on } \Gamma\}$$

$q \in L^2(\Omega)$, $a, b \in \mathbb{R}$, $a < b$ given. $f \in C^2(\Omega \times \mathbb{R}) \rightarrow \mathbb{R}$, twice L-continuously differentiable and Minty monotone, i. e.

$$(f(x, s) - f(x, t))(s - t) \geq 0 \quad \forall x \in \bar{\Omega}, \quad s, t \in \mathbb{R} \tag{11}$$

The state equation (10) is understood as weak formulation

$$a(y, v) + (f(\cdot, y), v)_\Omega = (u, v)_\Gamma \quad \forall v \in V. \quad (12)$$

Lemma

For any $u \in U$ problem (12) possesses a unique solution $y \in V$. The operator $S : U \rightarrow V$ defined by $Su := y$ is Lipschitz continuous, i.e. there is some $c > 0$ such that

$$\|Su - S\tilde{u}\|_V \leq c \|u - \tilde{u}\|_U \quad \forall u, \tilde{u} \in U, \quad (13)$$

and weakly sequentially continuous, i.e.

$$u_k \rightharpoonup u \text{ in } U \quad \implies \quad y_k \rightharpoonup y \text{ in } V \quad y_k := Su_k, \quad y := Su. \quad (14)$$

Furthermore holds $y_k \rightarrow y$ in $L^2(\Omega)$.

The control-to-state mapping leads to the reduced problem

$$J(u) := \frac{1}{2} \|Su - q\|_{0,\Omega}^2 + \frac{\alpha}{2} \|u\|_{0,\Gamma}^2 \rightarrow \min! \quad \text{s.t. } u \in U_{ad} \quad (15)$$

Theorem

Problem (15) possesses an optimal solution $\bar{u} \in U_{ad}$. Any optimal solution $\bar{u} \in U_{ad}$ of (15) satisfies

$$\langle J'(\bar{u}), u - \bar{u} \rangle \geq 0 \quad \forall u \in U_{ad}, \quad (16)$$

which is equivalent to

$$(S\bar{u} - q, S'(u - \bar{u}))_{\Omega} + \alpha (\bar{u}, u - \bar{u})_{\Gamma} \geq 0 \quad \forall u \in U_{ad}. \quad (17)$$

2. Penalties for Control Bounds

Let

$$U_{ad} = \{u \in L^2(\Omega) : u \leq b \quad \text{a.e. in } \Omega\}$$

Barrier-penalty modification

$$\tilde{J}(u, s) := J(u) + \int_{\Omega} \phi(u(x) - b(x), s) dx \quad (18)$$

of the objective. Here $s > 0$ denotes the penalty parameter with $s \rightarrow 0+$ and $\phi : \mathbb{R} \rightarrow \overline{\mathbb{R}}$ some barrier-penalty function that satisfies

$$\frac{\partial}{\partial t} \phi(t, s) = \psi \left(\frac{t}{s} \right) \quad \forall t \in \text{dom } \phi(\cdot, s) \quad (19)$$

with an appropriate function $\psi : \mathbb{R} \rightarrow \overline{\mathbb{R}}$.

Results for the finite dimensional case

Problem

$J : \mathbb{R}^n \rightarrow \mathbb{R}$ and $g : \mathbb{R}^n \rightarrow \mathbb{R}^m$, sufficiently smooth and consider

$$J(u) \rightarrow \min! \quad \text{s.t.} \quad u \in \mathbb{R}^n, g(u) \leq 0. \quad (20)$$

Auxiliary problem via general barrier-penalty approach

Problem

$$\tilde{J}(u, s) := J(u) + \sum_{i=1}^m \phi_i(g_i(u), s) \quad \text{s.t.} \quad u \in \mathbb{R}^n \quad (21)$$

Theorem

Let \bar{u} a local solution of (20) and $\bar{u}(s)$ a related local solution of (21). Further, assume that LICQ and strict complementarity locally hold. Then

$$\lim_{s \rightarrow 0+} \bar{u}(s) = \bar{u}$$

and with some $c > 0$ and $s_0 > 0$ we have

$$\|\bar{u}(s) - \bar{u}\| \leq c s \quad \forall s \in (0, s_0].$$

Difficulties: $c = c(n)$ with $c(n) \rightarrow \infty$ for $n \rightarrow \infty$,
local application of implicit function theorem
(using second order sufficiency conditions).

In optimal control: no 'uniform' strict
complementarity.

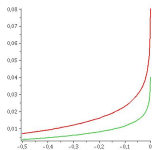
For optimal control consider either the quadratic loss

$$\psi(t) := \max\{0, t\}, \quad \phi(t, s) = \frac{1}{2s} \max^2\{0, t\}$$

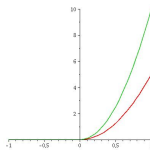
or the smoothed exact penalty

$$\psi(t) := \delta \left(1 + \frac{t}{\sqrt{1+t^2}} \right) \quad \phi(t, s) = \delta \left(t + \sqrt{s^2 + t^2} \right).$$

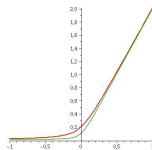
$\delta \geq \|\bar{v}\|_\infty + \varepsilon$ with some $\varepsilon > 0$, optimal multiplier \bar{v} .



log-barrier



quadratic loss



smoothed exact

Log-barrier analyzed by Schiela [8].

Back to optimal control

For the smoothed exact as well as for the quadratic loss functional under mild additional assumptions holds

Theorem (G/Kunz/Meischner[3])

For any $s > 0$ the penalty problem

$$\tilde{J}(u, s) \rightarrow \min! \quad \text{s.t. } u \in L^2(\Omega) \quad (22)$$

possesses a unique solution $\bar{u}(s)$ and there holds $\lim_{s \rightarrow 0+} \bar{u}(s) = \bar{u}$.

Further, if only control constraints occur then

$$\|\bar{u} - \bar{u}(s)\| = O(\sqrt{s}).$$

Theorem (Improved estimate, Winkler[Dipl.thesis 2011])

Assume that only control constraints are considered. Then in case of the quadratic loss penalty some $c > 0$ exists such that

$$\|u(s) - \bar{u}\| \leq c s$$

holds.

Proof (main idea)

The unconstrained auxiliary problem yields

$$J'(u(s)) + \psi\left(\frac{u(s) - b}{s}\right) = 0.$$

Direct calculations lead to

$$Tu(s) = Pu(s) \quad \text{with} \quad Tu := P(u - \sigma J'(u)).$$

Taking $\sigma > 0$ such that T contracts. With $\bar{u} = T\bar{u}$ and with $u^0 := u(s)$ we obtain

$$\begin{aligned}\|u(s) - \bar{u}\| &\leq \frac{1}{1-\kappa} \|Tu(s) - u(s)\| \\ &= \frac{1}{1-\kappa} \|Pu(s) - u(s)\|.\end{aligned}$$

Now,

$$\|u(s) - Pu(s)\| \leq s \|\bar{v}\| \quad \forall s > 0 \quad (\text{shown in C/K/M[3]})$$

completes the proof. □

Remark

The penalty multiplier $\bar{v}(s) := \psi((\bar{u}(s) - b)/s)$ approximates the optimal multiplier \bar{v} by

$$\|\bar{v}(s) - \bar{v}\| \leq c \|\bar{u}(s) - \bar{u}\|.$$

Problem (22) by (18) leads to the necessary and sufficient optimality conditions

$$\begin{aligned}(\bar{y}(s) - q, y) - a(y, \bar{v}(s)) &= 0 \quad \forall y \in V, \\ -a(\bar{y}(s), v) + (\bar{u}(s), v) &= 0 \quad \forall v \in V, \\ \alpha \bar{u}(s) + \bar{v}(s) + \psi((\bar{u}(s) - b)/s) &= 0 \quad \text{a.e. in } \Omega.\end{aligned}\tag{23}$$

Theorem

For any $s > 0$ system (23) possesses a unique solution $(\bar{y}(s), \bar{v}(s), \bar{u}(s))$.

The structure of ψ allows to find $\bar{u}(s)$ in dependence of $\bar{v}(s)$. Due to $\bar{v} \in H^2(\Omega)$ and $H^2(\Omega) \hookrightarrow C(\bar{\Omega})$ this can be done by pointwise.

Let $\bar{u}(s) := g(\bar{v}(s), s)$. Then (23) leads to the reduced optimality system

$$\begin{aligned}(\bar{y}(s) - q, y) - a(y, \bar{v}(s)) &= 0 \quad \forall y \in V, \\ -a(\bar{y}(s), v) + (g(\bar{v}(s), s), v) &= 0 \quad \forall v \in V.\end{aligned}\tag{24}$$

Theorem

For any $s > 0$ the system (24) possesses a unique solution $(\bar{u}(s), \bar{v}(s)) \in V \times V$ and $\bar{u}(s) := g(\bar{v}(s), s)$ forms the optimal solution of the problem (22).

Consider the boundary control case (10).

Assumption: There exist $\delta, \varepsilon > 0$ such that

$$J''(\bar{u})[h, h] \geq \delta \|h\|_{0,\Gamma}^2 \quad \forall \|h\|_{0,\Gamma} \leq \varepsilon.$$

Theorem

Let $\{s_k\} \subset \mathbb{R}_+$ with $s_k \rightarrow 0$ for $k \rightarrow \infty$. Then any related sequence $u_k := u(s_k)$ is bounded in U and therefore $\{u_k\}$ weakly compact. Any weakly convergent $\{u_k\}_{\mathcal{K}} \subset \{u_k\}$ converges also strongly in U to \bar{u} , i.e. $\lim_{k \in \mathcal{K}, k \rightarrow \infty} \|u(s_k) - \bar{u}\|_{0,\Gamma} = 0$. Further, some $\sigma \in (0, 1)$, $s_0 > 0$ exist such that

$$\|u(s_k) - \bar{u}\|_{0,\Gamma} \leq \frac{2\sigma}{1-\sigma} \|\bar{v}\|_{0,\Gamma} s_k \quad \forall k \in \mathcal{K}, s_k \in (0, s_0],$$

where \bar{v} denotes the optimal Lagrange multiplier at \bar{u} related to the constraint $u \leq b$.

3. Discretization of Reduced Problems

Consider conforming finite element discretizations

$$V_h \subset V$$

applied to the control reduced systems (24).

This leads to the finite dimensional nonlinear systems

$$\begin{aligned}(\bar{y}_h - q, y_h) - a(y_h, \bar{v}_h) &= 0 \quad \forall y_h \in V_h, \\ -a(\bar{y}_h, v_h) + (g(\bar{v}_h, s), v_h) &= 0 \quad \forall v_h \in V_h.\end{aligned}\tag{25}$$

Like in the continuous case system (25) defines uniquely the solution $(\bar{y}_h(s), \bar{v}_h(s)) \in V_h \times V_h$.

Remark

Unlike in full discretization no a-priori discretization of the control space U is used. The penalty yields an approximate projection.

Convergence analysis

follows widely standard arguments of conforming FEM discretization.

Galerkin orthogonality

$$\begin{aligned}(\bar{y}_h - \bar{y}, y_h) - a(y_h, \bar{v}_h - \bar{v}) &= 0 \quad \forall y_h \in V_h, \\ -a(\bar{y}_h - \bar{y}, v_h) + (g(\bar{v}_h, s) - g(\bar{v}, s), v_h) &= 0 \quad \forall v_h \in V_h.\end{aligned}\tag{26}$$

Define

Galerkin projections

$\tilde{y}_h, \tilde{v}_h \in V_h$ of $\bar{y}, \bar{v} \in V$ by

$$a(\bar{y} - \tilde{y}_h, v_h) = 0 \quad \forall v_h \in V_h$$

and

$$a(y_h, \bar{v} - \tilde{v}_h) = 0 \quad \forall y_h \in V_h.$$

Lemma (Schiela[8])

From (26) with the Galerkin projections follows

$$(\bar{y}_h - \bar{y}, \tilde{y}_h - \bar{y}_h) - (g(\bar{v}_h, s) - g(\bar{v}, s), \tilde{v}_h - \bar{v}_h) = 0.$$

Theorem

There exist some constant $c > 0$ such that

$$\|\bar{y} - \bar{y}_h\|^2 + \|\bar{v} - \bar{v}_h\|^2 \leq c (\|\bar{y} - \tilde{y}_h\|^2 + \|\bar{v} - \tilde{v}_h\|^2)$$

with the Galerkin projections $\tilde{y}_h, \tilde{v}_h \in V_h$.

Remark

If no state constraints are given then this constant is independent of the embedding parameter $s > 0$.

4. Numerical Example

Consider piecewise linear conforming finite elements

$$V_h \subset V$$

with a criss-cross triangulation applied to the control reduced systems (24).

Example 1

$$J(y, u) := \frac{1}{2} \|y - q\|_0^2 + \frac{\alpha}{2} \|u\|_0^2 \rightarrow \min!$$

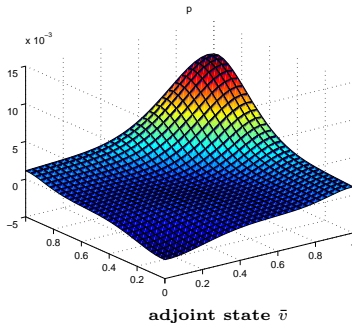
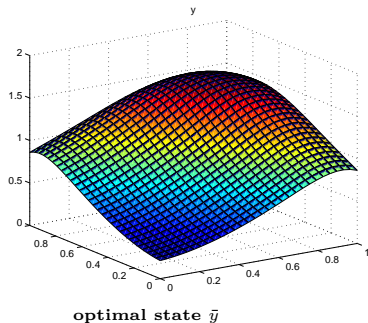
$$\text{s.t.} \quad -\Delta y = u \quad \text{in } \Omega = [0, 1]^2$$

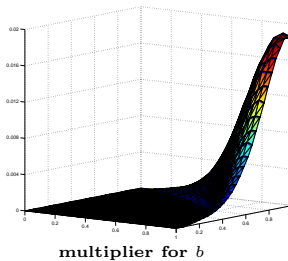
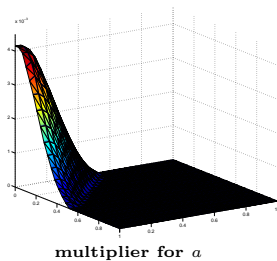
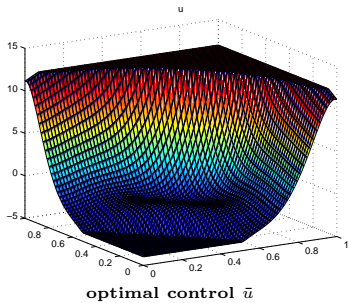
$$\frac{\partial y}{\partial n} + y = 0 \quad \text{on } \Gamma,$$

$$u \in U_{ad} := \{u \in U : -4 \leq u \leq 12 \quad \text{a.e. in } \Omega\}$$

with $q(x_1, x_2) = x_1 + x_2$.

Solution obtained with loss penalty for $s = 10^{-10}$ and overall $N = 900$.

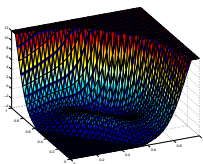




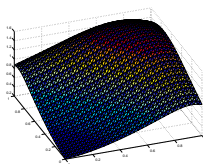
Similar results for the smoothed exact penalty

$$\psi(t) = \delta \left(1 + \frac{t}{\sqrt{1+t^2}} \right).$$

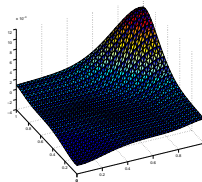
Unlike in the quadratic loss penalty case here feasibility is obtained for sufficiently small $s > 0$.



optimal control \bar{u}



optimal state \bar{y}



adjoint state \bar{v}

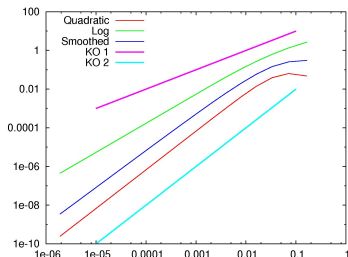
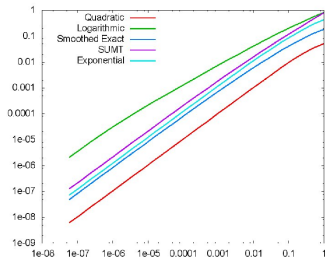
Experimental order of convergence in long-step path following with **one Newton step** per iteration and reduction

$$s_{k+1} = 0.5 s_k$$

s	$EOC_0(y)$	$EOC_1(y)$	$EOC_0(u)$
1	1.0097	0.9419	1.3334
2^{-4}	1.00	0.9955	1.0279
2^{-8}	1.00	1.00	1.0026
2^{-12}	1.00	1.00	1.00
2^{-16}	1.00	1.00	1.00
2^{-20}	1.00	1.00	1.00

Accuracies for a 1D-example with known exact solution

s	quadratic loss		log-barrier		smoothed exact	
	$\ \bar{u}(s) - \bar{u}\ $	EOC	$\ \bar{u}(s) - \bar{u}\ $	EOC	$\ \bar{u}(s) - \bar{u}\ $	EOC
2^{-5}	0.1869	0.9171	4.4882	0.6552	0.8625	0.77546
2^{-10}	0.0061	0.9972	0.3717	0.7499	0.0408	0.9272
2^{-15}	0.0002	0.9999	0.0257	0.7864	0.0014	0.9869
2^{-20}	6.0315 E-06	1.0000	0.0014	0.9115	4.4722 E-05	0.9991
2^{-25}	1.8848 E-07	1.0000	4.5904 E-05	0.9954	1.3985 E-06	0.9999

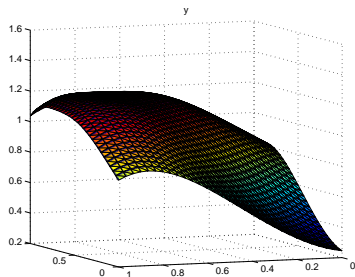


Modification of Example 1 by the additional state constraint

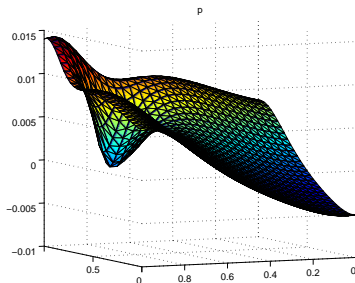
$$y(x) \leq 1.2 \quad \text{in } \Omega$$

and with no control bounds.

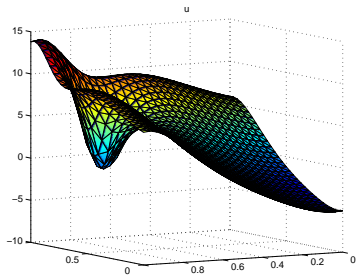
Numerical results for $N = 256$ and $s = 10^{-3}$



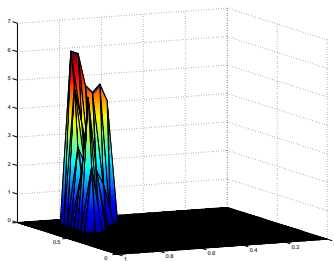
optimal state y



adjoint state v



optimal control u



approximate multiplier

Example 2

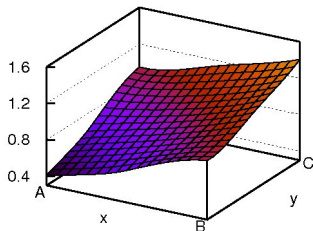
$$J(y, u) := \frac{1}{2} \|y - q\|_0^2 + \frac{\alpha}{2} \|u\|_0^2 \rightarrow \min!$$

$$\text{s.t.} \quad -\Delta y + y^3 = 0 \quad \text{in } \Omega = [0, 1]^2,$$

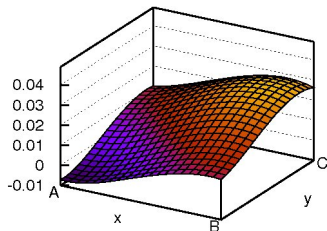
$$\frac{\partial y}{\partial n} + y = u \quad \text{on } \Gamma,$$

$$u \in U_{ad} := \{u \in L^2(\Gamma) : 0 \leq u \leq 2.2 \quad \text{a.e. on } \Gamma\}$$

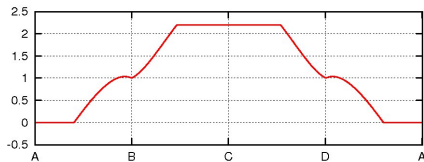
with $q(x_1, x_2) = x_1 + x_2$.



optimal state



adjoint state



optimal control

Solution obtained for $N = 200$ (each direction), $s = 10^{-40}$.

Obtained experimental order of convergence

s	$\ u_h(s) - \bar{u}_h\ $	EOC
2^{-05}	$1.93e - 02$	0.84
2^{-10}	$6.79e - 04$	0.99
2^{-15}	$2.13e - 05$	1.00
2^{-20}	$6.66e - 07$	1.00
2^{-25}	$2.08e - 08$	1.00
2^{-30}	$6.49e - 10$	1.01

$$\mathbf{Error} \ e(s) := \|u_h(s) - \bar{u}_h\|$$

Outlook

- extend the analytical proof for the numerically observed convergence rate $O(s)$ to more general types of barrier-penalty functions;
- deriving specific solution techniques for the elimination of u_h for reduced FEM discretizations;
- finding sharp bounds for the radius of convergence of Newton's method;
- studying long-step path-following Newton methods;
- study of stability and convergence properties in the case of state constraints.



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