

Mathematics of PDE constrained optimization  
Discrete concepts  
3. Constraints, interplay of discretisation and relaxation

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## Topic

**Optimal control of pdes with pointwise constraints:**

$$\min_{u \in U_{\text{ad}}, y \in Y_{\text{ad}}} J(y, u) \text{ s.t. } PDE(y) = B(u)$$

**Analysis: Casas 85,93 (pointwise state constraints), Casas & Fernandez 93 (pointwise constraints on gradient)**

**Numerical analysis (pointwise state constraints):**

**A priori:**

**Original problem: Casas & Mateos; Deckelnick & H.; Meyer;....**

**Relaxation: Group of Rösch; Group of Tröltzsch; Hintermüller & H.; H. & Meyer; H. & Schiela; ...**

**A posteriori: Benedix, Vexler & Wollner; Günther & H.; Hintermüller, Hoppe & Kieweg.**

**Numerical analysis (pointwise constraints on gradient): Deckelnick, Günther, & H.**

## State and control constraints

### Model problem

$$\min_{u \in U_{ad}} J(u) = \frac{1}{2} \int_D |y - z|^2 + \frac{\alpha}{2} \|u\|_U^2$$

subject to  $y = \mathcal{G}(Bu)$  and  $y \leq b$  in  $D$ .

Here,  $U_{ad} \subseteq U$  closed and convex,  $\alpha > 0$ ,  $z$ ,  $b$ , sufficiently smooth, and  $y = \mathcal{G}(Bu)$  iff

$$Ay = Bu \text{ in } D, \text{ plus b.c. (plus i.c.)}$$

- elliptic case:  $D = \Omega$  and  $Ay := -\sum_{i,j=1}^d \partial_{x_j} (a_{ij} y_{x_i}) + \sum_{i=1}^d b_i y_{x_i} + cy$  uniformly elliptic operator,
- parabolic case:  $D = (0, T] \times \Omega$  and  $Ay := y_t - \sum_{i,j=1}^d \partial_{x_j} (a_{ij} y_{x_i}) + \sum_{i=1}^d b_i y_{x_i} + cy$  with strongly elliptic leading part.

Slater condition:  $\exists \tilde{u} \in U_{ad}$  such that  $\mathcal{G}(B\tilde{u}) < b$  in  $\bar{D}$ .

## Optimality conditions (Casas 86,93)

Let  $u \in U_{ad}$  denote the unique optimal control and  $y = \mathcal{G}(Bu)$ . Then there exist  $\mu \in \mathcal{M}(\bar{D})$  and some  $p$  such that there holds

$$\begin{aligned} \int_D p A v &= \int_D (y - z) v + \int_{\bar{D}} v d\mu \quad \forall v \in X, \\ \langle B^* p + \alpha u, v - u \rangle_{U^*, U} &\geq 0 \quad \forall v \in U_{ad}, \\ \mu &\geq 0, \quad y \leq b \text{ in } D \text{ and } \int_{\bar{D}} (b - y) d\mu = 0, \end{aligned}$$

where

- elliptic case:  $p \in W^{1,s}(\Omega)$  for all  $s < d/(d-1)$  and  $X = H^2(\Omega)$  with  $\sum_{i,j=1}^d a_{ij} v_{x_i} v_j = 0$  on  $\partial\Omega$ ,
- parabolic case:  $p \in L^s(W^{1,\sigma})$  for all  $s, \sigma \in [1, 2)$  with  $2/s + d/\sigma > d + 1$  and  $X = \{v \in C^0(\bar{Q}); v(0, \cdot) = 0\} \cap \{v \in L^2(H^2), v_t \in L^2(H^1)\}$ .

## Discretization – a variational concept

**Discrete optimal control problem:**

$$\begin{aligned} \min_{u \in U_{ad}} J_h(u) &:= \frac{1}{2} \int_D |y_h - z|^2 + \frac{\alpha}{2} \|u\|_U^2 \\ \text{subject to } y_h &= \mathcal{G}_h(Bu) \text{ and } y_h \leq l_h b. \end{aligned}$$

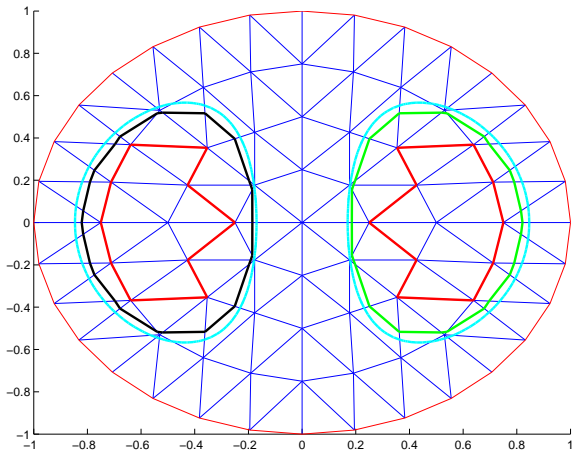
Here,  $y_h(u) = \mathcal{G}_h(Bu)$  denotes the

- p.l. and continuous fe approximation to  $y(u)$  (elliptic case),
- $dg(0)$  in time and p.l. and continuous fe in space approximation to  $y(u)$  (parabolic case), i.e.

$$a(y_h, v_h) = \langle Bu, v_h \rangle \text{ for all } v_h \in X_h.$$

**We do not discretize the control!**

## Variational versus conventional discretization





# Variational discretization for time-dependent problems

**Movie time-dependent problems**

## Discrete optimality conditions

Let  $u_h \in U_{ad}$  denote the unique variational–discrete optimal control,  $y_h = \mathcal{G}(Bu_h)$ . There exist  $\mu \in \mathbb{R}^k$  and  $p_h \in X_h$  such that with

- $\mu_h = \sum_{j=1}^{nv} \mu_j \delta_{x_j}$  (elliptic case,  $x_j$  fe nodes,  $k = nv$ ),
- $\mu_h = \sum_{i=1}^m \sum_{j=1}^{nv} \mu_{ij} \delta_{x_j} \circ \frac{1}{|I_i|} \int_{I_i} \bullet dt$  (parabolic case,  $x_j$  fe nodes,  $I_i$  dg intervals,  $k = nv + m$ ),

we have

$$\begin{aligned}
 a(v_h, p_h) &= \int_D (y_h - z)v_h + \int_{\bar{D}} v_h d\mu_h & \forall v_h \in X_h, \\
 \langle B^* p_h + \alpha u_h, v - u_h \rangle_{U^*, U} &\geq 0 & \forall v \in U_{ad}, \\
 \mu_j &\geq 0, y_h \leq I_h b, \text{ and } \int_{\bar{D}} (I_h b - y_h) d\mu_h = 0.
 \end{aligned}$$

Here,  $\delta_x$  denotes the Dirac measure concentrated at  $x$  and  $I_h$  is the usual Lagrange interpolation operator.



## Results

Let  $u_h \in U_{ad}$  denote the variational–discrete optimal solution with corresponding state  $y_h \in X_h$  and  $\mu_h \in \mathcal{M}(\bar{D})$ . Then for  $h$  small enough

$$\|y_h\|, \|u_h\|_U, \|\mu_h\|_{\mathcal{M}(\bar{D})} \leq C.$$

For the proof a discrete counterpart to the Slater condition is needed, which is deduced from uniform convergence of the discrete states associated to the Slater point  $B\tilde{u}$ .

## Results, cont.

Let  $u$  denote the solution of the continuous problem and  $u_h$  the variational discrete optimal control. Then

$$\begin{aligned} \alpha \|u - u_h\|^2 + \|y - y_h\|^2 &\leq \\ &\leq C(\|\mu\|_{\mathcal{M}(\bar{D})}, \|\mu_h\|_{\mathcal{M}(\bar{D})}) \left\{ \|y - y_h(u)\|_\infty + \|y^h(u_h) - y_h\|_\infty \right\} + \\ &\quad + C(\|u\|, \|u_h\|) \left\{ \|y - y_h(u)\| + \|y^h(u_h) - y_h\| \right\}. \end{aligned}$$

Here,  $y_h(u) = \mathcal{G}_h(Bu)$ ,  $y^h(u_h) = \mathcal{G}(Bu_h)$ .

We need uniform estimates for discrete approximations.

## Error estimates, parabolic case

Deckelnick, H. (JCM 2010)

Controls  $u \in L^2(0, T)^m$ , and  $f_i \in H^1(\Omega)$  given actuations.

$$Bu := \sum_{i=1}^m u_i(t) f_i(x), \quad y_0 \in H^2(\Omega).$$

Then  $y = \mathcal{G}(Bu) \in \{v \in L^\infty(H^2), v_t \in L^2(H^1)\}$  and we have with  $y_h = \mathcal{G}_h(Bu)$  and time stepping  $\delta t \sim h^2$

$$\|y - y_h\|_\infty \leq C \begin{cases} h\sqrt{|\log h|}, & (d = 2) \\ \sqrt{h}, & (d = 3) \end{cases}$$

This is not an *off-the-shelf* result! It yields

$$\alpha \|u - u_h\|^2 + \|y - y_h\|^2 \leq C \begin{cases} h\sqrt{|\log h|}, & (d = 2) \\ \sqrt{h}, & (d = 3). \end{cases}$$

## Error estimates, elliptic case

Deckelnick, H. (SINUM 2007, ENUMATH 2007)

- $Bu \in L^2(\Omega)$ :

$$\|u - u_h\|_U, \|y - y_h\|_{H^1} = \begin{cases} O(h^{\frac{1}{2}}), & \text{if } d = 2, \\ O(h^{\frac{1}{4}}), & \text{if } d = 3, \end{cases}$$

- $Bu \in W^{1,s}(\Omega)$ :

$$\|u - u_h\|_U, \|y - y_h\|_{H^1} \leq Ch^{\frac{3}{2} - \frac{d}{2s}} \sqrt{|\log h|}.$$

- $Bu \in L^\infty(\Omega)$ :

$$\|u - u_h\|_U, \|y - y_h\|_{H^1} \leq Ch |\log h|.$$

- $U = L^2(\Omega)$ ,  $U_{ad} = \{u \leq d\}$ ,  $u_h$  p.c.:

$$\|u - u_h\|_U, \|y - y_h\|_{H^1} \leq C \begin{cases} h |\log h|, & \text{if } d = 2, \\ \sqrt{h}, & \text{if } d = 3. \end{cases}$$

Similar results obtained by C. Meyer for discrete controls.

## Numerical experiment 1

$$\Omega := B_1(\mathbf{0}), \alpha > 0,$$

$$z(x) := 4 + \frac{1}{\pi} - \frac{1}{4\pi}|x|^2 + \frac{1}{2\pi} \log |x|, \quad b(x) := |x|^2 + 4,$$

and  $u_0(x) := 4 + \frac{1}{4\alpha\pi}|x|^2 - \frac{1}{2\alpha\pi} \log |x|$ .

$$J(u) := \frac{1}{2} \int_{\Omega} |y - z|^2 + \frac{\alpha}{2} \int_{\Omega} |u - u_0|^2,$$

where  $y = \mathcal{G}(u)$ .

Unique solution  $u \equiv 4$  with corresponding state  $y \equiv 4$  and multipliers

$$p(x) = \frac{1}{4\pi}|x|^2 - \frac{1}{2\pi} \log |x| \quad \text{and} \quad \mu = \delta_0.$$

## Experimental order of convergence

<i>RL</i>	$\ u - u_h\ $	$\ y - y_h\ $
1	0.788985	0.536461
2	0.759556	1.147861
3	0.919917	1.389378
4	0.966078	1.518381
5	0.986686	1.598421

## Relaxing constraints – Lavrentiev (H., Meyer COAP 2008)

**Lavrentiev Regularization: relax  $y \leq b$  to  $\lambda u + y \leq b$  ( $\lambda > 0$ ). Numerical analysis yields**

- $Bu^\lambda \in L^2(\Omega)$  uniformly:

$$\|u - u_h^\lambda\| \sim \|u - u^\lambda\| + \|u^\lambda - u_h^\lambda\| \sim \sqrt{\lambda} + h^{1-d/4},$$

- $Bu^\lambda \in W^{1,s}(\Omega)$  uniformly for all  $s \in (1, \frac{d}{d-1})$ :

$$\|u - u_h^\lambda\| \sim \|u - u^\lambda\| + \|u^\lambda - u_h^\lambda\| \sim \sqrt{\lambda} + h^{2-d/2-\epsilon},$$

- $Bu^\lambda \in L^\infty(\Omega)$ ,  $Bu_h^\lambda \in L^\infty(\Omega)$  uniformly:

$$\|u - u_h^\lambda\| \sim \|u - u^\lambda\| + \|u^\lambda - u_h^\lambda\| \sim \sqrt{\lambda} + h |\log h|.$$

## Relaxing constraints – penalization (Hintermüller, H.)

Relax  $y \leq b$  with  $\frac{\gamma}{2} \int_{\Omega} |(y - b)^+|^2 dx$  in cost functional.

- $Bu^\gamma \in L^2(\Omega)$  uniformly:

$$\begin{aligned} \|u - u_h^\gamma\| &\sim \|u - u^\gamma\| + \|u^\gamma - u_h^\gamma\| \sim \\ &\sim \left( h^{1-d/p} + \frac{1}{\sqrt{\gamma}} h^{-d/2} \right)^{1/2} + h^{1-d/4}, \end{aligned}$$

- $Bu^\gamma \in W^{1,s}(\Omega)$  for all  $s \in (1, \frac{d}{d-1})$  uniformly:

$$\begin{aligned} \|u - u_h^\gamma\| &\sim \|u - u^\gamma\| + \|u^\gamma - u_h^\gamma\| \sim \\ &\sim \left( h^{1-d/p} + \frac{1}{\sqrt{\gamma}} h^{-d/2} \right)^{1/2} + h^{2-d/2-\epsilon}, \end{aligned}$$

- $Bu^\gamma \in L^\infty(\Omega), Bu_h^\gamma \in L^\infty(\Omega)$  uniformly:

$$\begin{aligned} \|u - u_h^\gamma\| &\sim \|u - u^\gamma\| + \|u^\gamma - u_h^\gamma\| \sim \\ &\sim \left( h^{1-d/p} + \frac{1}{\sqrt{\gamma}} h^{-d/2} \right)^{1/2} + h |\log h|. \end{aligned}$$





## Relaxing constraints – barriers

**Barriers:** relax  $y \leq b$  by adding  $-\mu \int_{\Omega} \log(b - y) dx$  to cost functional ( $\mu > 0$ ).  
**Numerical analysis yields**

- $Bu^{\mu} \in L^2(\Omega)$  uniformly:

$$\|u - u_h^{\mu}\| \sim \|u - u^{\mu}\| + \|u^{\mu} - u_h^{\mu}\| \sim \sqrt{\mu} + h^{1-d/4},$$

- $Bu^{\mu} \in W^{1,s}(\Omega)$  for all  $s \in (1, \frac{d}{d-1})$  uniformly:

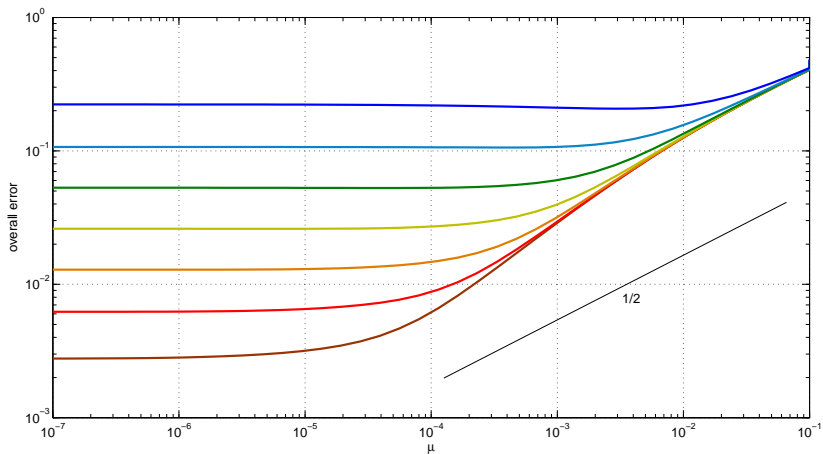
$$\|u - u_h^{\mu}\| \sim \|u - u^{\mu}\| + \|u^{\mu} - u_h^{\mu}\| \sim \sqrt{\mu} + h^{2-d/2-\epsilon},$$

- $Bu^{\mu} \in L^{\infty}(\Omega), Bu_h^{\mu} \in L^{\infty}(\Omega)$  uniformly:

$$\|u - u_h^{\mu}\| \sim \|u - u^{\mu}\| + \|u^{\mu} - u_h^{\mu}\| \sim \sqrt{\mu} + h |\log h|.$$

This is work in progress with Anton Schiela.

## Relaxing constraints – barriers, numerical results



**Consequence: Grid size  $h$  and parameters  $(\lambda, \gamma, \mu)$  should be coupled;**

**Lavrentiev:**  $\sqrt{\lambda} \sim h^{2-d/2}$ ,

**Barriers:**  $\sqrt{\mu} \sim h^{2-d/2}$ ,

**Penalization ( $p = \infty$ ):**  $\frac{1}{\sqrt{\gamma}} \sim h^{1+d/2}$  (optimal ?).

## Constraints on the gradient

Consider

$$\min_{u \in U_{\text{ad}}} J(u) = \frac{1}{2} \int_{\Omega} |y - z|^2 + \frac{\alpha}{r} \int_{\Omega} |u|^r \left( + \frac{\alpha}{2} \int_{\Omega} |u|^2 \right)$$

where  $y = \mathcal{G}(u)$ , i.e. solves the pde, and  $\nabla y \in Y_{\text{ad}}$ .

Here

$$Y_{\text{ad}} = \{z \in C^0(\bar{\Omega})^d \mid |z(x)| \leq \delta, x \in \bar{\Omega}\},$$

and

$$\begin{aligned} r = 2 : & \quad U_{\text{ad}} = \{u \in L^2(\Omega) \mid a \leq u \leq b \text{ a.e. in } \Omega\} (a, b \in L^\infty), \\ r > d : & \quad U_{\text{ad}} = L^r(\Omega). \end{aligned}$$

Then  $U_{\text{ad}} \subset L^r(\Omega)$  for  $r > d \Rightarrow \nabla y \in C^0(\bar{\Omega})^d$ .

Slater condition:

$$\exists \hat{u} \in U_{\text{ad}} \mid |\nabla \hat{y}(x)| < \delta, x \in \bar{\Omega}, \text{ where } \hat{y} \text{ solves the pde with } u = \hat{u}.$$

## Optimality conditions (Casas & Fernandez)

An element  $u \in U_{\text{ad}}$  is a solution if and only if there exist  $\bar{\mu} \in \mathcal{M}(\bar{\Omega})^d$  and  $p \in L^t(\Omega)$  ( $t < \frac{d}{d-1}$ ) such that

$$\begin{aligned} \int_{\Omega} p \mathcal{A}z - \int_{\Omega} (y - z)z &= \int_{\bar{\Omega}} \nabla z \cdot d\bar{\mu} & \forall z \in W^{2,t'}(\Omega) \cap W_0^{1,t'}(\Omega) \\ \int_{\bar{\Omega}} (z - \nabla y) \cdot d\bar{\mu} &\leq 0 & \forall z \in Y_{\text{ad}}, \end{aligned}$$

$$\begin{aligned} \int_{\Omega} (p + \alpha u)(\tilde{u} - u) &\geq 0 & \forall \tilde{u} \in U_{\text{ad}} \text{ for } r = 2, \text{ or} \\ p + \alpha((u+)|u|^{r-2}u) &= 0 & \text{in } \Omega \text{ for } r > d. \end{aligned}$$

**Structure of multiplier:**  $\bar{\mu} = \frac{1}{\delta} \nabla y \mu$ , where  $\mu \in \mathcal{M}(\bar{\Omega}) \geq 0$  is concentrated on  $\{x \in \bar{\Omega} \mid |\nabla y(x)| = \delta\}$ .

## FE discretization, conventional

**Piecewise linear, continuous Ansatz for the state  $y_h = \mathcal{G}_h(u) \in X_h$ .**

**The discrete control problem reads**

$$\begin{aligned} \min_{u \in U_{\text{ad}}} J_h(u) &:= \frac{1}{2} \int_{\Omega} |y_h - z|^2 + \frac{\alpha}{r} \int_{\Omega} |u|^r \left( + \frac{\alpha}{2} \int_{\Omega} |u|^2 \right) \\ \text{subject to } y_h &= \mathcal{G}_h(u) \text{ and } \left( \frac{1}{|T|} \int_T \nabla y_h \right)_{T \in \mathcal{T}_h} \in Y_{\text{ad}}^h, \end{aligned}$$

**where**

$$Y_{\text{ad}}^h := \{c_h : \bar{\Omega} \rightarrow \mathbb{R}^d \mid c_h|_T \text{ is constant and } |c_h|_T| \leq \delta, T \in \mathcal{T}_h\}.$$

## FE discretization, conventional, optimality conditions

The variational discrete problem has a unique solution  $u_h \in U_{\text{ad}}$ . There exist  $\mu_T \in \mathbb{R}^d$ ,  $T \in \mathcal{T}_{h,X}$  and  $p_h \in X_h$  such that with  $y_h = \mathcal{G}_h(u_h)$  we have

$$a(v_h, p_h) = \int_{\Omega} (y_h - z)v_h + \sum_{T \in \mathcal{T}_{h,X}} |T| \nabla v_h|_T \cdot \mu_T \quad \forall v_h \in X_h,$$

$$\sum_{T \in \mathcal{T}_{h,X}} |T| (c_{hT} - \nabla y_h|_T) \cdot \mu_T \leq 0 \quad \forall c_h \in C_h,$$

$$p_h + \alpha((u_h +)|u_h|^{r-2}u_h) = 0 \quad \text{in } \Omega.$$

Structure of the multiplier:  $\vec{\mu}_T = \mu_T \frac{1}{\delta} \nabla y_{hT}$ , where  $\mu_T \in \mathbb{R}$ . Furthermore,  $\mu_T \geq 0$  and  $\mu_T > 0$  only if  $|\nabla y_{hT}| = \delta$ .



## Results

Deckelnick, Günther, H. (Oberwolfach Report 2008): Let  $u_h \in U_{\text{ad}}$  be the variational discrete optimal solution with corresponding state  $y_h \in X_h$  and adjoint variables  $p_h \in X_h$ ,  $\vec{\mu}_T (T \in \mathcal{T}_h)$ .

Then for  $h$  small enough

- $\|y_h\|, \|u_h\|_{L^r}, \|p_h\|_{L^{r-1}}, \sum_{T \in \mathcal{T}_{h,x}} |T| |\mu_T| \leq C,$
- $\|y - y_h\| \leq Ch^{\frac{1}{2}(1-\frac{d}{r})}, \|u - u_h\|_{L^r} \leq Ch^{\frac{1}{r}(1-\frac{d}{r})},$  and  
 $\|u - u_h\|_{L^2} \leq Ch^{\frac{1}{2}(1-\frac{d}{r})}.$

These results are also valid for a piecewise constant Ansatz of the control.

## FE discretization, Raviart Thomas

Mixed fe approximation of the state with lowest order Raviart–Thomas element, i.e.

$$(y_h, v_h) = \mathcal{G}_h(u) \in Y_h \times V_h$$

denotes the solution of

$$\begin{aligned} \int_{\Omega} A^{-1} v_h \cdot w_h + \int_{\Omega} y_h \operatorname{div} w_h &= 0 & \forall w_h \in V_h \\ \int_{\Omega} z_h \operatorname{div} v_h - \int_{\Omega} a_0 y_h z_h + \int_{\Omega} u z_h &= 0 & \forall z_h \in Y_h. \end{aligned}$$

## FE discretization, cont.

The discrete control problem reads

$$\begin{aligned} \min_{u \in U_{\text{ad}}} J_h(u) &:= \frac{1}{2} \int_{\Omega} |y_h - z|^2 + \frac{\alpha}{2} \int_{\Omega} |u|^2 \\ \text{subject to } (y_h, v_h) &= \mathcal{G}_h(u) \text{ and } \left( \frac{1}{|T|} \int_T \mathbf{A}^{-1} v_h \right)_{T \in \mathcal{T}_h} \in Y_{\text{ad}}^h, \end{aligned}$$

where

$$Y_{\text{ad}}^h := \{c_h : \bar{\Omega} \rightarrow \mathbb{R}^d \mid c_h|_T \text{ is constant and } |c_h|_T| \leq \delta, T \in \mathcal{T}_h\}.$$

## FE discretization, optimality conditions

The discrete problem has a unique solution  $u_h \in U_{\text{ad}}$ . Furthermore, there are  $\vec{\mu}_T \in \mathbb{R}^d$  and  $(p_h, \chi_h) \in Y_h \times V_h$  such that with  $(y_h, v_h) = \mathcal{G}_h(u_h)$  we have

$$\int_{\Omega} \mathbf{A}^{-1} \chi_h \cdot \mathbf{w}_h + \int_{\Omega} p_h \operatorname{div} \mathbf{w}_h + \sum_{T \in \mathcal{T}_h} \vec{\mu}_T \cdot \int_T \mathbf{A}^{-1} \mathbf{w}_h = 0 \quad \forall \mathbf{w}_h \in V_h$$

$$\int_{\Omega} z_h \operatorname{div} \chi_h - \int_{\Omega} a_0 p_h z_h + \int_{\Omega} (y_h - z) z_h = 0 \quad \forall z_h \in Y_h.$$

$$\int_{\Omega} (p_h + \alpha u_h) (\tilde{u} - u_h) \geq 0 \quad \forall \tilde{u} \in U_{\text{ad}}$$

$$\sum_{T \in \mathcal{T}_h} \vec{\mu}_T \cdot (\mathbf{c}_h|_T - \int_T \mathbf{A}^{-1} \mathbf{v}_h) \leq 0 \quad \forall \mathbf{c}_h \in Y_{\text{ad}}^h.$$

Structure of the multiplier:  $\vec{\mu}_T = \mu_T \frac{1}{\delta} \int_T \mathbf{A}^{-1} \mathbf{v}_h$ , where  $\mu_T \in \mathbb{R}$ . Furthermore,  $\mu_T \geq 0$  and  $\mu_T > 0$  only if  $|\int_T \mathbf{A}^{-1} \mathbf{v}_h| = \delta$ .

## Results

Deckelnick, Günther, H. (Numer. Math 2008): Let  $u_h \in U_{\text{ad}}$  be the optimal solution of the discrete problem with corresponding state  $(y_h, v_h) \in Y_h \times V_h$  and adjoint variables  $(p_h, \chi_h) \in Y_h \times V_h$ ,  $\bar{\mu}_T, T \in \mathcal{T}_h$ .

Then for  $h$  small enough

- $\|y_h\|, \sum_{T \in \mathcal{T}_h} |\bar{\mu}_T| \leq C$ , and
- $\|u - u_h\| + \|y - y_h\| \leq Ch^{\frac{1}{2}} |\log h|^{\frac{1}{2}}$ .

## Constraints on the gradient, example

We take  $\Omega = B_2(0)$  and consider

$$\min J(u) = \frac{1}{2} \|y - z\|_{L^2(\Omega)}^2 + \frac{1}{2} \|u\|_{L^2(\Omega)}^2$$

with pointwise bounds on the constraints, i.e.  $\{a \leq u \leq b\}$ , where  $a, b \in L^\infty(\Omega)$ , and pointwise bounds on the gradient, i.e.  $|\nabla y(x)| \leq \delta := 1/2$ . State and control satisfy

$$-\Delta y = f + u \text{ in } \Omega, \quad y = 0 \text{ on } \partial\Omega.$$

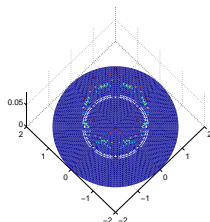
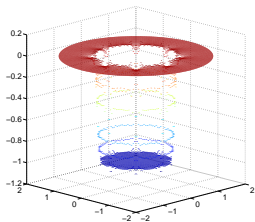
Data:

$$z(x) := \begin{cases} \frac{1}{4} + \frac{1}{2} \ln 2 - \frac{1}{4} |x|^2 & , 0 \leq |x| \leq 1 \\ \frac{1}{2} \ln 2 - \frac{1}{2} \ln |x| & , 1 < |x| \leq 2 \end{cases} \quad f(x) := \begin{cases} 2 & , 0 \leq |x| \leq 1 \\ 0 & , 1 < |x| \leq 2 \end{cases}$$

Solution:

$$y(x) \equiv z(x) \text{ and } u(x) = \begin{cases} -1 & , 0 \leq |x| \leq 1 \\ 0 & , 1 < |x| \leq 2 \end{cases}$$

# Numerical experiment, piecewise constant control Ansatz



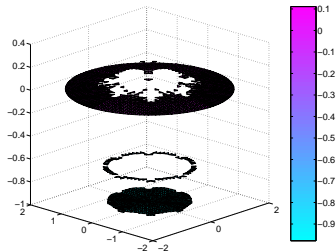
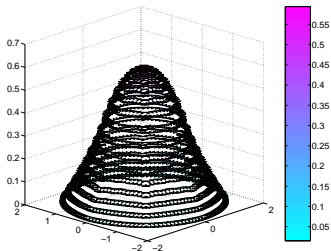
## Experimental order of convergence

<i>RL</i>	$\ u - u_h\ _{L^4}$	$\ u - u_h\ $	$\ y - y_h\ $
1	0.76678	0.72339	1.90217
2	0.33044	0.64248	1.25741
3	0.27542	0.54054	1.23233
4	0.28570	0.53442	1.16576

Results show the predicted behaviour, since  $r = \infty$ .



# Numerical solution, mixed finite elements



## Experimental order of convergence

<i>RL</i>	$\ u - u_h\ $	$\ y - y_h\ $	$\ y^P - y_h^P\ $
1	0.98576	1.06726	1.08949
2	0.51814	1.02547	1.09918
3	0.50034	1.01442	1.08141

Superscript  $P$  denotes post-processed piecewise linear state. It attains the same order of convergence but yields significantly smaller approximation error.



**Thank you very much for your attention**