

# $L^\infty$ -error estimates in non-convex domains with application to optimal control

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“PDE Constrained Optimization - recent challenges and future developments”  
Hamburg, März 2008



## Model problem

We discuss the optimal control problem

$$J(\bar{u}) = \min J(u),$$

$$J(u) := F(Su, u),$$

$$F(y, u) := \frac{1}{2} \|y - y_d\|_{L^2(\Omega)}^2 + \frac{\nu}{2} \|u\|_{L^2(\Omega)}^2,$$

where the associated state  $y = Su$  to the control  $u$  is the weak solution of the state equation

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

and the control variable is constrained by

$$a \leq u(x) \leq b \quad \text{for a.a. } x \in \Omega.$$

$\Omega$  is assumed to be two-dimensional and non-convex.

# Results from literature for $L^\infty$ -estimates

## Results in convex domains

- [Meyer/Rösch, 2004] discrete control, no postprocessing, 2D

$$\|\bar{u} - \bar{u}_h\|_{L^\infty(\Omega)} = O(h)$$

- [Hinze, 2004] semidiscrete approach

$$\begin{aligned} \|\bar{u} - \bar{u}_h\|_{L^\infty(\Omega)} &\lesssim \|(S - S_h)y_d\|_{L^\infty(\Omega)} + \|(S^* - S_h^*)S\bar{u}\|_{L^\infty(\Omega)} \\ &+ \begin{cases} h^2 |\log h|^{1/2} \|\bar{u}\|_{L^2(\Omega)} & (2D) \\ h^{3/2} \|\bar{u}\|_{L^2(\Omega)} & (3D) \end{cases} \end{aligned}$$

# $L^\infty$ -error estimates for the state equation

State equation:

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

Results in literature:

- [Frehse/Rannacher, '76]

$$\|y - y_h\|_{L^\infty(\Omega)} \leq Ch^2 |\log h| \|y\|_{W^{2,\infty}(\Omega)}$$

- [Schatz/Wahlbin, '78/'79]

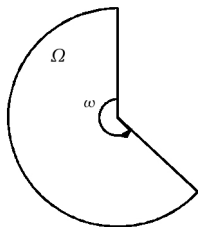
$$\|y - y_h\|_{L^\infty(\Omega)} \leq Ch^{2-\epsilon}$$

for “smooth” right-hand side.

Aim:

- Estimate for non-convex domains.
- Separate the constant from the norm of the function to be approximated.

## Corner singularities (2D)



The solution  $y$  of

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

is **not** contained in the Sobolev space  $W^{2,2}(\Omega)$ , if  $\omega > \pi$ .

Instead, one can write

$$y = y_r + y_s$$

where  $y_r \in W^{2,2}(\Omega)$  and

$$y_s = \xi(r)\gamma r^\lambda \sin(\lambda\varphi) \quad \text{with } \lambda = \frac{\pi}{\omega}.$$

$\xi(r)$  is a smooth cut-off function and  $\gamma$  a coefficient.

# Regularity

The solution can be described in **weighted Sobolev spaces**

$$V_{\beta}^{k,p}(\Omega) = \left\{ y \in \mathcal{D}' : \sum_{|\alpha| \leq k} \left( \int_{\Omega} |r^{\beta-|\alpha|+k} D^{\alpha} y|^p \right)^{1/p} < \infty \right\}$$

$$V_{\gamma}^{k,\infty}(\Omega) = \left\{ y \in \mathcal{D}' : \sup_{\substack{x \in \Omega \\ |\alpha| \leq k}} \left( |r^{\gamma-|\alpha|+k} D^{\alpha} y(x) \right) < \infty \right\}.$$

One has the a priori estimates

$$\|y\|_{V_{\beta}^{2,2}(\Omega)} \lesssim \|f\|_{L^2(\Omega)} \quad \text{for } \beta > 1 - \lambda,$$

$$\|y\|_{V_{\gamma}^{2,\infty}(\Omega)} \lesssim \|f\|_{C^{0,\sigma}(\Omega)} \quad \text{for } \gamma > 2 - \lambda.$$

## Mesh grading

In order to get an optimal rate of convergence, mesh grading is necessary. We set the element size according to

$$h_T = \begin{cases} h^{1/\mu} & \text{if } r_T = 0 \\ hr^{1-\mu} & \text{if } r_T > 0 \end{cases}$$

where  $h$  is the global mesh parameter and  $r$  is the distance to the corner. One has to choose

$$\mu < \begin{cases} \lambda & \text{for optimal convergence rate in } L^2(\Omega). \\ \frac{\lambda}{2} & \text{for optimal convergence rate in } L^\infty(\Omega). \end{cases}$$

# Subsets of $\Omega$

Divide  $\Omega$  in different subsets

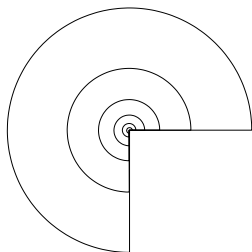
$$\Omega = \bigcup_{j=0}^l \Omega_j$$

where

$$\Omega_j = \{x : d_{j+1} \leq |x - v| \leq d_j\} \quad j \neq l$$

$$\Omega_l = \{x : |x - v| \leq d_l\}$$

and  $d_j = 2^{-j}$ ,  $d_l \sim h^{2/\lambda}$ .





## Local error estimates

### Lemma

For  $J = l, l - 1$  one has

$$\|y - y_h\|_{L^\infty(\Omega_J)} \lesssim |\log h|^{1/2} h^2 \|y\|_{V_\gamma^{2,\infty}(\Omega'_J)} + |\log h|^{1/2} d_J^{-1} \|y - y_h\|_{L^2(\Omega'_J)}.$$

For  $J \neq l, l - 1$  the inequality

$$\|y - y_h\|_{L^\infty(\Omega_J)} \lesssim |\log h| \min_{\chi \in V_h} \|y - \chi\|_{L^\infty(\Omega'_J)} + d_J^{-1} \|y - y_h\|_{L^2(\Omega'_J)}$$

is valid.

### Lemma

The inequality

$$d_J^{-1} \|y - y_h\|_{L^2(\Omega'_J)} \lesssim h^2 |\log h|^{1/2} \|y\|_{V_\gamma^{2,\infty}(\Omega)}$$

is valid.

## $L^\infty$ -estimate for the state equation

### Theorem

Let  $y$  be the solution of the boundary value problem with a right-hand side  $f \in C^{0,\sigma}(\Omega)$ . The finite element error can be estimated by

$$\|y - y_h\|_{L^\infty(\Omega)} \leq Ch^2 |\log h| \|f\|_{C^{0,\sigma}(\Omega)}$$

on finite element meshes with grading parameter  $\mu < \lambda/2$ .

A mesh is graded, if  $\mu < 1$ . This means, that **mesh grading is necessary** for all corners with

$$\omega > \frac{\pi}{2}.$$

## Back to optimal control

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

subject to

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

and

$$a < u < b \quad \text{a. e. in } \Omega.$$

with  $a < b$ ,  $\nu > 0$ ,  $\Omega \subset \mathbb{R}^2$  non-convex with one reentrant corner.

## Optimality system

The optimal control  $\bar{u}$  is the unique solution of the system

$$\begin{aligned}\bar{y} &= S\bar{u}, \\ \bar{p} &= S^*(\bar{y} - y_d), \\ \bar{u} &= \Pi_{[a,b]} \left( -\frac{1}{\nu} \bar{p} \right)\end{aligned}$$

with the projection

$$\Pi_{[a,b]} f(x) := \max(a, \min(b, f(x))).$$

The control is discretized by **piecewise constant** functions. State and adjoint state are approximated with **piecewise linear** functions. Denote by  $S_h$  the solution operator of the discrete system and by  $R_h$  the operator

$$R_h u(x) = u(S_T) \quad \text{if } x \in T, \quad S_T \text{ centroid of } T.$$

# Discretization of the OCP

Find  $\bar{u}_h \in U_h^{ad}$  such that

$$J_h(\bar{u}_h) = \min_{u_h \in U_h^{ad}} \frac{1}{2} \|S_h u_h - y_d\|_{L^2(\Omega)}^2 + \frac{\nu}{2} \|u_h\|_{L^2(\Omega)}^2$$

with

$$\begin{aligned}U_h &= \{u \in L^2(\Omega) : u|_T \in \mathcal{P}_0 \ \forall T \in \mathcal{T}_h\} \\U^{ad} &= \{u \in L^2(\Omega) : a \leq u(x) \leq b \text{ a.e. in } \Omega\} \\U_h^{ad} &= U_h \cap U^{ad}\end{aligned}$$

Discrete optimality system:

$$\begin{aligned}\bar{y}_h &= S_h \bar{u}_h \\ \bar{p}_h &= S_h^*(\bar{y}_h - y_d) \\ \bar{u}_h &= \Pi_{U_h^{ad}} \left( -\frac{1}{\nu} R_h \bar{p}_h \right).\end{aligned}$$

# $L^\infty$ -error estimate for OCP

## Theorem

On a family of meshes with grading parameter  $\mu < \frac{\lambda}{2}$  the estimates

$$\|\bar{y}_h - \bar{y}\|_{L^\infty(\Omega)} \leq ch^2 |\log h| \left( \|\bar{u}\|_{C^{0,\sigma}(\bar{\Omega})} + \|y_d\|_{C^{0,\sigma}(\bar{\Omega})} \right)$$

$$\|\bar{p}_h - \bar{p}\|_{L^\infty(\Omega)} \leq ch^2 |\log h| \left( \|\bar{u}\|_{C^{0,\sigma}(\bar{\Omega})} + \|y_d\|_{C^{0,\sigma}(\bar{\Omega})} \right)$$

are valid.

Proof:

$$\begin{aligned} \|\bar{y} - \bar{y}_h\|_{L^\infty(\Omega)} &= \|S\bar{u} - S_h\bar{u}_h\|_{L^\infty(\Omega)} \\ &\leq \|S\bar{u} - S_h\bar{u}\|_{L^\infty(\Omega)} + \|S_h\bar{u} - S_hR_h\bar{u}\|_{L^\infty(\Omega)} + \|S_hR_h\bar{u} - S_h\bar{u}_h\|_{L^\infty(\Omega)}. \end{aligned}$$

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are valid.

Proof:

$$\begin{aligned} \|\bar{y} - \bar{y}_h\|_{L^\infty(\Omega)} &= \|S\bar{u} - S_h\bar{u}_h\|_{L^\infty(\Omega)} \\ &\leq \|S\bar{u} - S_h\bar{u}\|_{L^\infty(\Omega)} + \|S_h\bar{u} - S_hR_h\bar{u}\|_{L^\infty(\Omega)} + \|S_hR_h\bar{u} - S_h\bar{u}_h\|_{L^\infty(\Omega)}. \end{aligned}$$

FE - error

# $L^\infty$ -error estimate for OCP

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Proof:

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Supercloseness



# Supercloseness

The approximate solution  $\bar{u}_h$  is closer to the interpolant  $R_h\bar{u}$  than to the solution  $\bar{u}$ :

[Apel, Rösch, Winkler, 2005]

On a family of meshes with grading parameter  $\mu < \lambda$  the estimate

$$\|\bar{u}_h - R_h\bar{u}\|_{L^2(\Omega)} \leq ch^2 (\|\bar{u}\|_{L^\infty(\Omega)} + \|y_d\|_{L^\infty(\Omega)})$$

holds true.

Further, one has

$$\|S_h\|_{L^2(\Omega) \rightarrow L^\infty(\Omega)} \leq C$$

# $L^\infty$ -error estimate for OCP

## Theorem

On a family of meshes with grading parameter  $\mu < \frac{\lambda}{2}$  the estimates

$$\|\bar{y}_h - \bar{y}\|_{L^\infty(\Omega)} \leq ch^2 |\log h| \left( \|\bar{u}\|_{C^{0,\sigma}(\bar{\Omega})} + \|y_d\|_{C^{0,\sigma}(\bar{\Omega})} \right)$$

$$\|\bar{p}_h - \bar{p}\|_{L^\infty(\Omega)} \leq ch^2 |\log h| \left( \|\bar{u}\|_{C^{0,\sigma}(\bar{\Omega})} + \|y_d\|_{C^{0,\sigma}(\bar{\Omega})} \right)$$

are valid.

Proof:

$$\begin{aligned} \|\bar{y} - \bar{y}_h\|_{L^\infty(\Omega)} &= \|S\bar{u} - S_h\bar{u}_h\|_{L^\infty(\Omega)} \\ &\leq \|S\bar{u} - S_h\bar{u}\|_{L^\infty(\Omega)} + \|S_h\bar{u} - S_hR_h\bar{u}\|_{L^\infty(\Omega)} + \|S_hR_h\bar{u} - S_h\bar{u}_h\|_{L^\infty(\Omega)}. \end{aligned}$$

## Regularized Dirac function

Let us define a regularized Dirac function for a fixed  $a \in T$  as a function with the properties

- 1  $(\delta^h(a), v_h) = v_h(a) \quad \forall v_h \in V_h,$
- 2  $\text{supp } \delta^h(a) \subset \bar{T},$
- 3  $\delta^h(a) \in \mathcal{P}_1(T),$
- 4  $\|\delta^h(a)\|_{L^2(T^*)} = O(h_T^{-1}).$
- 5  $\|\delta^h(a)\|_{L^\infty(T^*)} \leq C|T|^{-1}$

An example of such a function is [Scott, 73]

$$\delta^h(x) = \begin{cases} |T|^{-1} \hat{\varphi}(F^{-1}(x)) & \text{if } x \in T^* \\ 0 & \text{if } x \notin T \end{cases}$$

with  $F : \hat{T} \rightarrow T$  and  $\hat{\varphi}$  a polynomial of degree one, so that

$$\int_{\hat{T}} \hat{p} \hat{\varphi} \, d\hat{x} = \hat{p}(\hat{a}) \quad \forall \hat{p} \in \mathcal{P}_1(\hat{T}).$$

# Regularized Green function

The regularized Green function  $z^h$  is defined as solution of

$$a(v, z^h) = (\delta^h(a), v) \quad \forall v \in V.$$

We denote by  $z_h^h$  its discrete counterpart,

$$a(v_h, z_h^h) = (\delta^h(a), v_h) \quad \forall v_h \in V_h.$$

Here  $a$  is the bilinear form  $a(u, v) = \int_{\Omega} \nabla u \nabla v$ .

## Lemma

The estimate

$$\|z_h^h\|_{L^\infty(\Omega)} + \|z_h^h\|_{H_0^1(\Omega)} \leq C |\log h|$$

holds on a finite element mesh with  $\mu < \lambda$ .

## Estimate for $\|S_h \bar{u} - S_h R_h \bar{u}\|_{L^\infty(\Omega)}$

### Lemma

The inequality

$$\|S_h \bar{u} - S_h R_h \bar{u}\|_{L^\infty(\Omega)} \leq Ch^2 |\log h| (\|\bar{u}\|_{L^\infty(\Omega)} + \|\bar{y}_d\|_{L^\infty(\Omega)})$$

is satisfied.

**Proof:** For an arbitrary, but fixed  $a \in \Omega$ , we find

$$\begin{aligned} |S_h \bar{u}(a) - S_h R_h \bar{u}(a)| &= |(\delta^h(a), S_h \bar{u} - S_h R_h \bar{u})| \quad (\text{Definiton of } \delta_h(a)) \\ &= |a(S_h \bar{u} - S_h R_h \bar{u}, z_h^h)| \quad (\text{Definiton of } z_h^h) \\ &= |(z_h^h, \bar{u} - R_h \bar{u})| \quad (\text{Test function } z_h^h) \end{aligned}$$

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**Auxiliary result [Apel, Rösch, Winkler, 2005]:**

On a mesh with grading parameter  $\mu < \lambda$  the estimate

$$(v_h, \bar{u} - R_h \bar{u})_{L^2(\Omega)} \leq Ch^2 \left( \|v_h\|_{L^\infty(\Omega)} + \|v_h\|_{H_0^1(\Omega)} \right) (\|\bar{u}\|_{L^\infty(\Omega)} + \|\bar{y}_d\|_{L^\infty(\Omega)})$$

can be proved for all  $v_h \in V_h$ .

## Estimate for $\|S_h \bar{u} - S_h R_h \bar{u}\|_{L^\infty(\Omega)}$

### Lemma

The inequality

$$\|S_h \bar{u} - S_h R_h \bar{u}\|_{L^\infty(\Omega)} \leq Ch^2 |\log h| (\|\bar{u}\|_{L^\infty(\Omega)} + \|\bar{y}_d\|_{L^\infty(\Omega)})$$

is satisfied.

**Proof:** For an arbitrary, but fixed  $a \in \Omega$ , we find

$$\begin{aligned} |S_h \bar{u}(a) - S_h R_h \bar{u}(a)| &= |(\delta^h(a), S_h \bar{u} - S_h R_h \bar{u})| \quad (\text{Definiton of } \delta_h(a)) \\ &= |a(S_h \bar{u} - S_h R_h \bar{u}, z_h^h)| \quad (\text{Definiton of } z_h^h) \\ &= |(z_h^h, \bar{u} - R_h \bar{u})| \quad (\text{Test function } z_h^h) \end{aligned}$$

Auxiliary result + estimate for regularized Green function  $\Rightarrow$

$$\begin{aligned} |S_h \bar{u}(a) - S_h R_h \bar{u}(a)| &\leq Ch^2 \left( \|z_h^h\|_{L^\infty(\Omega)} + \|z_h^h\|_{H_0^1(\Omega)} \right) (\|\bar{u}\|_{L^\infty(\Omega)} + \|\bar{y}_d\|_{L^\infty(\Omega)}) \\ &\leq Ch^2 |\ln h| (\|\bar{u}\|_{L^\infty(\Omega)} + \|\bar{y}_d\|_{L^\infty(\Omega)}) \end{aligned}$$

# $L^\infty$ -error estimate for OCP

## Theorem

On a family of meshes with grading parameter  $\mu < \frac{\lambda}{2}$  the estimates

$$\|\bar{y}_h - \bar{y}\|_{L^\infty(\Omega)} \leq ch^2 |\log h| \left( \|\bar{u}\|_{C^{0,\sigma}(\bar{\Omega})} + \|y_d\|_{C^{0,\sigma}(\bar{\Omega})} \right)$$

$$\|\bar{p}_h - \bar{p}\|_{L^\infty(\Omega)} \leq ch^2 |\log h| \left( \|\bar{u}\|_{C^{0,\sigma}(\bar{\Omega})} + \|y_d\|_{C^{0,\sigma}(\bar{\Omega})} \right)$$

is valid.

Proof:

$$\begin{aligned} \|\bar{y} - \bar{y}_h\|_{L^\infty(\Omega)} &= \|S\bar{u} - S_h\bar{u}_h\|_{L^\infty(\Omega)} \\ &\leq \|S\bar{u} - S_h\bar{u}\|_{L^\infty(\Omega)} + \|S_h\bar{u} - S_hR_h\bar{u}\|_{L^\infty(\Omega)} + \|S_hR_h\bar{u} - S_h\bar{u}_h\|_{L^\infty(\Omega)} \\ &\leq ch^2 |\log h| \left( \|\bar{u}\|_{C^{0,\sigma}(\bar{\Omega})} + \|y_d\|_{C^{0,\sigma}(\bar{\Omega})} \right) \end{aligned}$$



# Postprocessing for the control

Postprocessing step [Meyer, Rösch, 2004]

$$\tilde{u}_h = \Pi_{[a,b]} \left( -\frac{1}{\nu} p_h \right).$$

## Theorem

On a family of meshes with grading parameter  $\mu < \frac{\lambda}{2}$  the inequality

$$\|\bar{u} - \tilde{u}_h\|_{L^\infty(\Omega)} \leq Ch^2 |\ln h| \left( \|\bar{u}\|_{C^{0,\sigma}(\bar{\Omega})} + \|y_d\|_{C^{0,\sigma}(\bar{\Omega})} \right)$$

is satisfied.

## Numerical example (by G. Winkler)

Consider

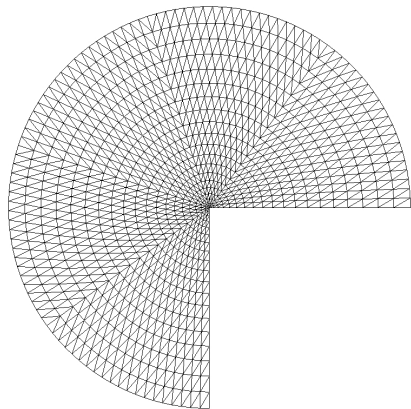
$$\begin{aligned} -\Delta y + y &= u + f && \text{in } \Omega \\ -\Delta p + p &= y - y_d && \text{in } \Omega \\ u &= \Pi_{[-0.3,1]} \left( -\frac{1}{\nu} p \right) \end{aligned}$$

with  $\nu = 10^{-4}$  and homogeneous Dirichlet boundary conditions for  $y$  and  $p$ . The data  $f$  and  $y_d$  are chosen such that

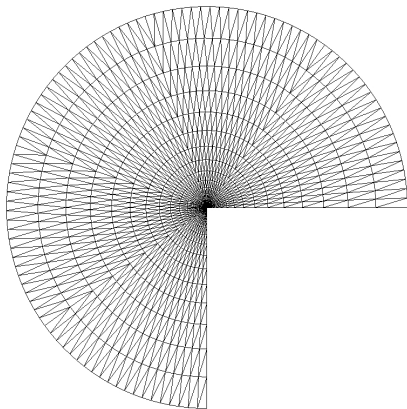
$$\begin{aligned} y(r, \varphi) &= \left( r^{2/3} - r^{5/2} \right) \sin \frac{2}{3} \varphi \\ p(r, \varphi) &= \nu \left( r^{2/3} - r^{5/2} \right) \sin \frac{2}{3} \varphi \end{aligned}$$

are the exact solutions of the optimal control problem.

## Different mesh grading

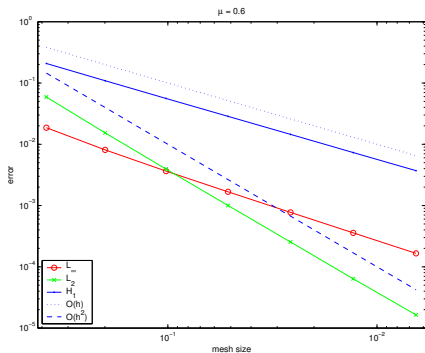


$$\mu = 0.6$$

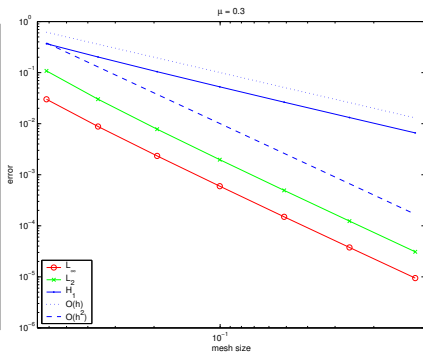


$$\mu = 0.3$$

# Errors in the state for $\mu = 0.6$ and $\mu = 0.3$

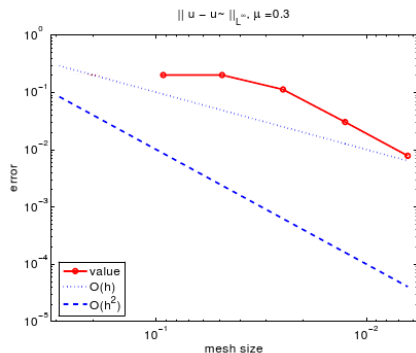


$$\mu < \lambda$$



$$\mu < \frac{\lambda}{2}$$

# Error in the control for $\mu = 0.3$



ndof	$\ u - \tilde{u}_h\ _{L^\infty(\Omega)}$	
425	2.00e-01	.....
1617	1.12e-01	0.84
6305	3.02e-02	1.89
24897	7.77e-03	1.96

# Conclusion and Outlook

## Conclusion

- $L^\infty$ -error estimate for Dirichlet problem in non-convex domains
- $L^\infty$ -error estimate for linear-quadratic OCP in non-convex domains

## Current work

Extension of the results to 3D

- $L^2$ -error estimates in polyhedral domains with reentrant edge  
→ GAMM
- open:  $L^\infty$ -error estimate for three-dimensional, non-convex domains.