

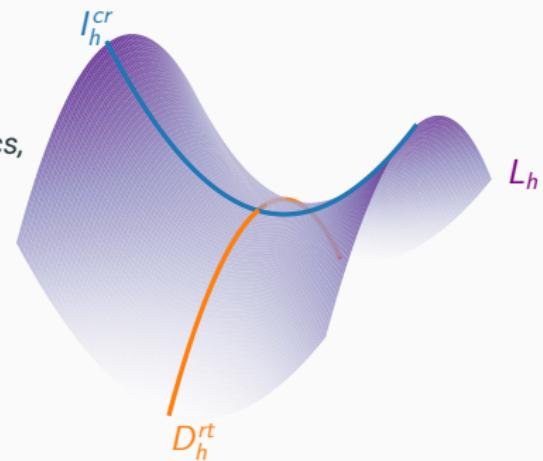
# ***A priori and a posteriori error identities for convex minimization problems based on convex duality relations***

## Lecture 4

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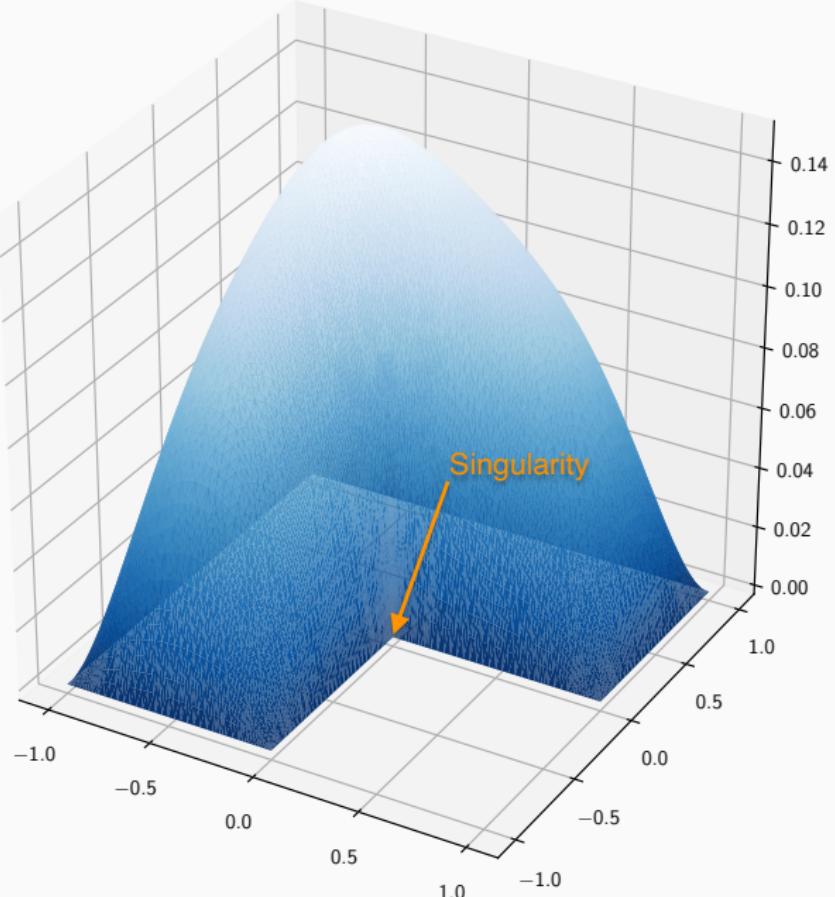
### ◆ Lecture 4: Applications

- Adaptive mesh-refinement
  - Adaptive algorithm;
  - Examples.
- *A priori* error identity on the basis of discrete convex duality;
  - Discrete primal-dual gap identity;
  - Examples.

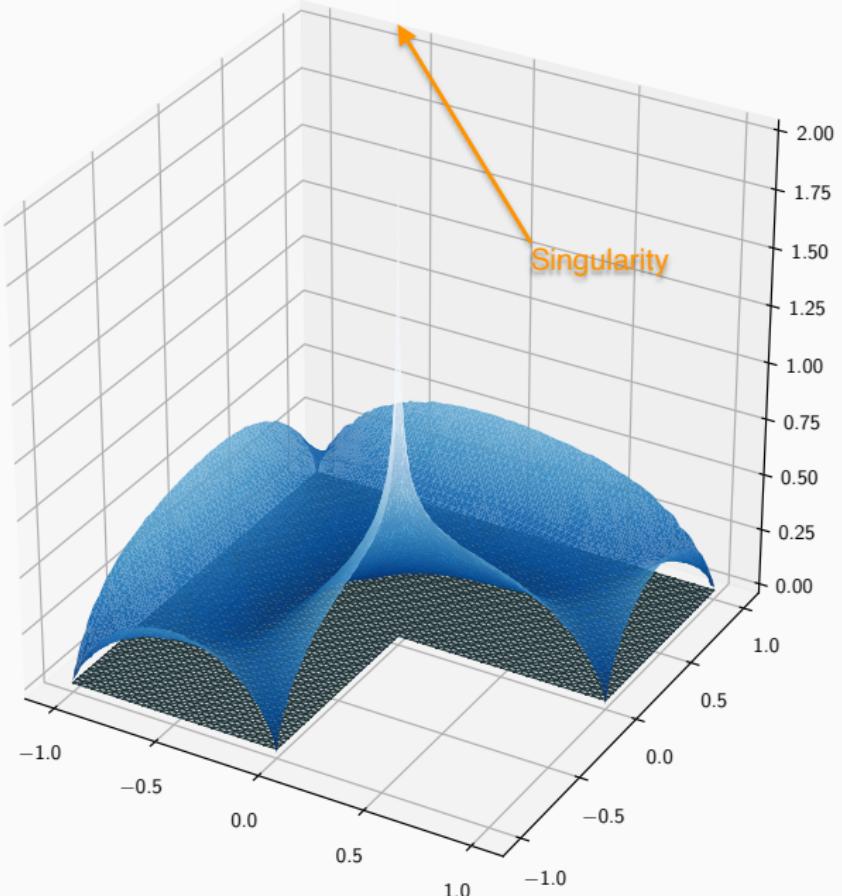
## **Adaptive mesh-refinement**

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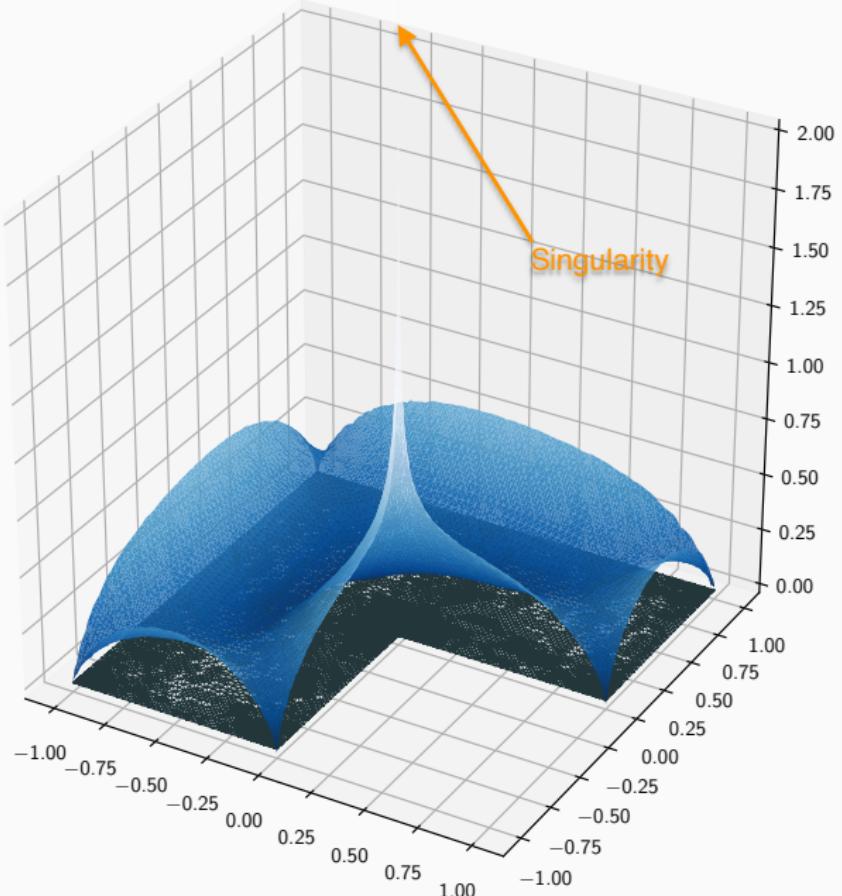
# Adaptive mesh-refinement



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# Adaptive mesh-refinement



# Adaptive algorithm

## Algorithm: (SOLVE–ESTIMATE–MARK–REFINE)

Let  $\varepsilon_{\text{STOP}} > 0$ ,  $\theta \in (0, 1]$ , and  $\mathcal{T}_0$  an initial triangulation. Then, for every  $k \geq 0$ :

(‘SOLVE’) Compute discrete primal solution and discrete dual solution

$$u_k^{cr} \in S_D^{1,cr}(\mathcal{T}_k) \quad \& \quad z_k^{rt} \in \mathcal{RT}_N^0(\mathcal{T}_k);$$

(‘ESTIMATE’) Post-process  $u_k^{cr} \in S_D^{1,cr}(\mathcal{T}_k)$  and  $z_k^{rt} \in \mathcal{RT}_N^0(\mathcal{T}_k)$  to obtain

$$\bar{u}_k^{cr} \in \text{dom}(l) \quad \& \quad \bar{z}_k^{rt} \in \text{dom}(-D) \cap W_N^{p'}(\text{div}; \Omega);$$

Compute (*local*) refinement indicators  $\{\eta_{\text{gap},T}^2(\bar{u}_k^{cr}, \bar{z}_k^{rt})\}_{T \in \mathcal{T}_k}$ , defined by

$$\eta_{\text{gap},T}^2(\bar{u}_k^{cr}, \bar{z}_k^{rt}) = \left\{ \begin{array}{l} \int_T \left\{ \phi^*(\cdot, \bar{z}_k^{rt}) - \bar{z}_k^{rt} \cdot \nabla \bar{u}_k^{cr} + \phi(\cdot, \nabla \bar{u}_k^{cr}) \right\} dx \\ \int_T \left\{ \psi^*(\cdot, \text{div } \bar{z}_k^{rt}) - \text{div } \bar{z}_k^{rt} \bar{u}_k^{cr} + \psi(\cdot, \bar{u}_k^{cr}) \right\} dx \end{array} \right\}$$

for all  $T \in \mathcal{T}_k$ . If  $\rho_{\text{tot}}^2(\bar{u}_k^{cr}, \bar{z}_k^{rt}) = \eta_{\text{gap}}^2(\bar{u}_k^{cr}, \bar{z}_k^{rt}) \leq \varepsilon_{\text{STOP}}$ , then STOP;

(‘MARK’) Use *marking strategy* to find  $\mathcal{M}_k \subseteq \mathcal{T}_k$  on which error is ‘concentrated’;

(‘REFINE’) Perform ‘refinement’ of  $\mathcal{T}_k$  to get  $\mathcal{T}_{k+1}$  s.t. each  $T \in \mathcal{M}_k$  is ‘refined’ in  $\mathcal{T}_{k+1}$ .

Increase  $k \rightarrow k + 1$  and continue with (‘SOLVE’).

◆ Two Cases:

- **Case I:** Primal problem can be approximated cheaper:

- Compute  $u_k^{cr} \in \mathcal{S}_D^{1,cr}(\mathcal{T}_k)$  using **problem-dependent method**;
- Compute  $z_k^{rt} \in \mathcal{RT}_N^0(\mathcal{T}_k)$  via **discrete reconstruction formula (for free!)**, e.g.,

$$z_k^{rt} = D_t \phi_{h_k}(\cdot, \nabla_{h_k} u_k^{cr}) + \frac{D_t \psi_{h_k}(\cdot, \Pi_{h_k} u_k^{cr})}{d} (\text{id}_{\mathbb{R}^d} - \Pi_{h_k} \text{id}_{\mathbb{R}^d});$$

- **Examples:** Poisson problem,  $p$ -Dirichlet problem, Obstacle problem, ...

- **Case II:** Dual problem can be approximated cheaper:

- Compute  $z_k^{rt} \in \mathcal{RT}_N^0(\mathcal{T}_k)$  using **problem-dependent method**;
- Compute  $u_k^{cr} \in \mathcal{S}_D^{1,cr}(\mathcal{T}_k)$  via **discrete reconstruction formula (for free!)**, e.g.,

$$u_k^{cr} = D_t \psi_{h_k}^*(\cdot, \text{div } z_k^{rt}) + D_t \phi_{h_k}^*(\cdot, \Pi_{h_k} z_k^{rt}) \cdot (\text{id}_{\mathbb{R}^d} - \Pi_{h_k} \text{id}_{\mathbb{R}^d});$$

- **Example:** Elasto-plastic torsion problem, i.e.,  $\phi := \frac{1}{2} |\cdot|^2 + I_{|\cdot| \leq 1}^{\Omega} \notin C^0(\mathbb{R}^d)$  and  $\psi(x, \cdot) := (t \mapsto -f(x)t) \in C^1(\mathbb{R})$  for a.e.  $x \in \Omega$ .

# Adaptive algorithm

## Algorithm: (SOLVE–ESTIMATE–MARK–REFINE)

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for all  $T \in \mathcal{T}_k$ . If  $\eta_{\text{gap}}^2(\bar{u}_k^{cr}, \bar{z}_k^{rt}) \leq \varepsilon_{\text{STOP}}$ , then STOP;

(‘MARK’) Use *marking strategy* to find  $\mathcal{M}_k \subseteq \mathcal{T}_k$  on which error is ‘concentrated’;

(‘REFINE’) Perform ‘refinement’ of  $\mathcal{T}_k$  to get  $\mathcal{T}_{k+1}$  s.t. each  $T \in \mathcal{M}_k$  is ‘refined’ in  $\mathcal{T}_{k+1}$ .

Increase  $k \rightarrow k + 1$  and continue with (‘SOLVE’).

◆ **Non-conformity:**

- Since

$$u_k^{cr} \in S_D^{1,cr}(\mathcal{T}_k) \not\subseteq W_D^{1,p}(\Omega),$$

in general, we have that

$$u_k^{cr} \notin \text{dom}(I).$$

- Even if

$$z_k^{rt} \in \mathcal{RT}_N^0(\mathcal{T}_k) \subseteq W_N^{p'}(\text{div}; \Omega),$$

in general, we have that

$$z_k^{rt} \notin \text{dom}(-D).$$

◆ **Post-processing:** Obtaining

$$\bar{u}_k^{cr} \in \text{dom}(I) \quad \& \quad \bar{z}_k^{rt} \in \text{dom}(-D) \cap W_N^{p'}(\text{div}; \Omega),$$

is problem-dependent and often involves:

- Quasi-interpolation operators (e.g., node-averaging operator);
- Truncation and scaling arguments that preserve regularity.

◆ **Node-averaging operator:**  $\Pi_h^{\text{av}} : \mathbb{P}^1(\mathcal{T}_h) \rightarrow \mathcal{S}_D^1(\mathcal{T}_h)$ , for every  $v_h \in \mathbb{P}^1(\mathcal{T}_h)$  defined by

$$\Pi_h^{\text{av}} v_h := \sum_{\nu \in \mathcal{N}_h} \langle v_h \rangle_\nu \varphi_\nu ,$$

$$\langle v_h \rangle_\nu := \begin{cases} \frac{1}{\text{card}(\mathcal{T}_h(\nu))} \sum_{T \in \mathcal{T}_h(\nu)} (v_h|_T)(\nu) & \text{if } \nu \in \Omega \cup \Gamma_N , \\ 0 & \text{if } \nu \in \Gamma_D , \end{cases}$$

where  $\mathcal{T}_h(\nu) := \{T \in \mathcal{T}_h \mid \nu \in T\}$  and  $(\varphi_\nu)_{\nu \in \mathcal{N}_h}$  nodal basis of  $\mathcal{S}^1(\mathcal{T}_h)$ .

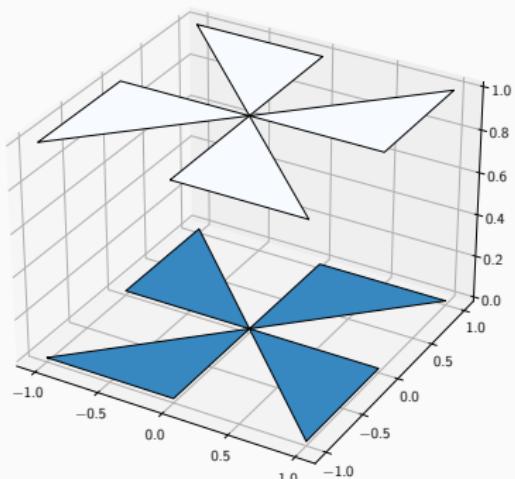
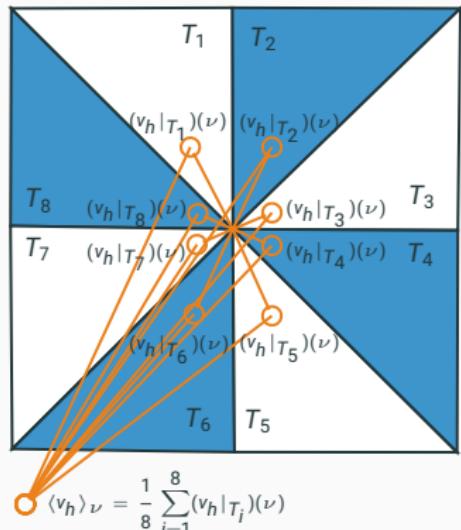


Figure:  $v_h \in \mathbb{P}^0(\mathcal{T}_h)$ .

◆ **Node-averaging operator:**  $\Pi_h^{\text{av}} : \mathbb{P}^1(\mathcal{T}_h) \rightarrow \mathcal{S}_D^1(\mathcal{T}_h)$ , for every  $v_h \in \mathbb{P}^1(\mathcal{T}_h)$  defined by

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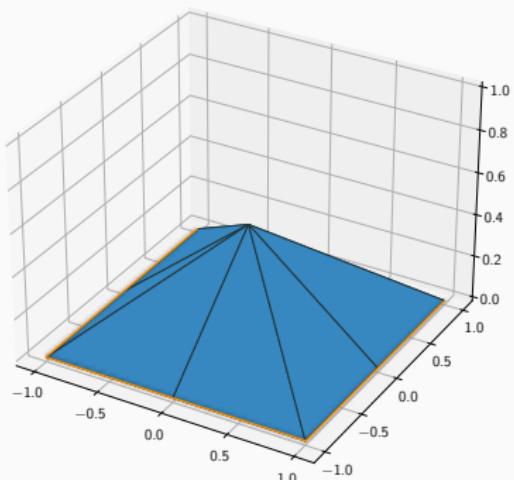
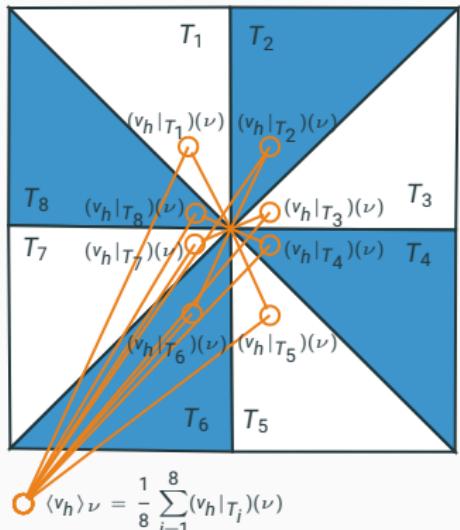


Figure:  $\Pi_h^{\text{av}} v_h \in \mathcal{S}_D^1(\mathcal{T}_h)$  if  $\Gamma_D = \partial\Omega$ .

◆ **Node-averaging operator:**  $\Pi_h^{\text{av}} : \mathbb{P}^1(\mathcal{T}_h) \rightarrow \mathcal{S}_D^1(\mathcal{T}_h)$ , for every  $v_h \in \mathbb{P}^1(\mathcal{T}_h)$  defined by

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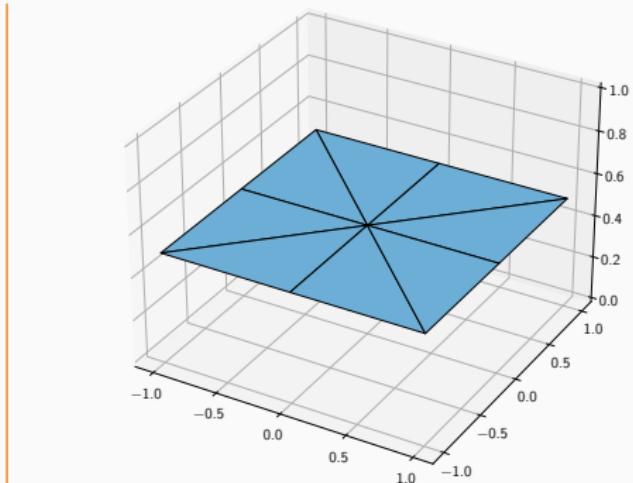
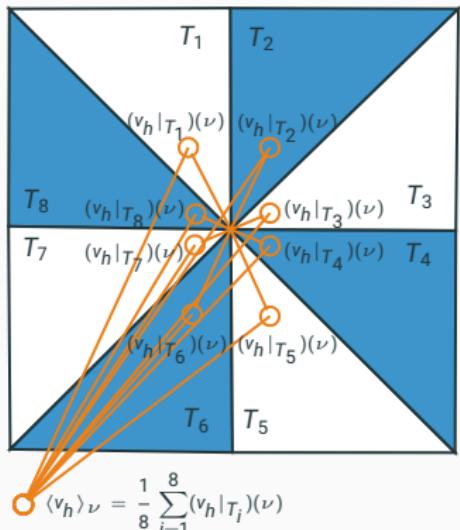


Figure:  $\Pi_h^{\text{av}} v_h \in \mathcal{S}_D^1(\mathcal{T}_h)$  if  $\Gamma_D = \emptyset$ .

◆ **Node-averaging operator:**  $\Pi_h^{\text{av}} : \mathbb{P}^1(\mathcal{T}_h) \rightarrow \mathcal{S}_D^1(\mathcal{T}_h)$ , for every  $v_h \in \mathbb{P}^1(\mathcal{T}_h)$  defined by

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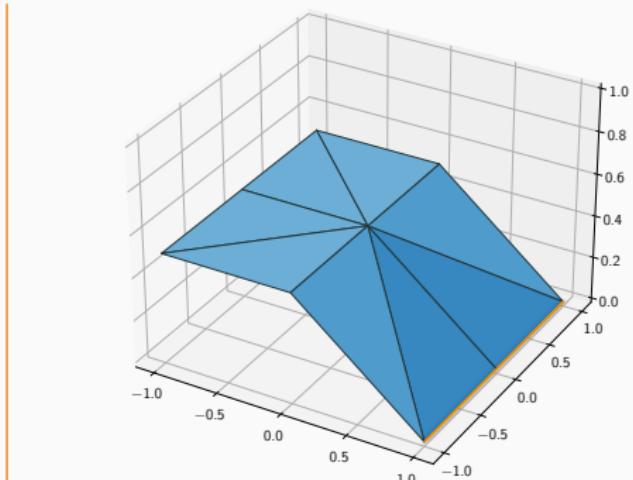
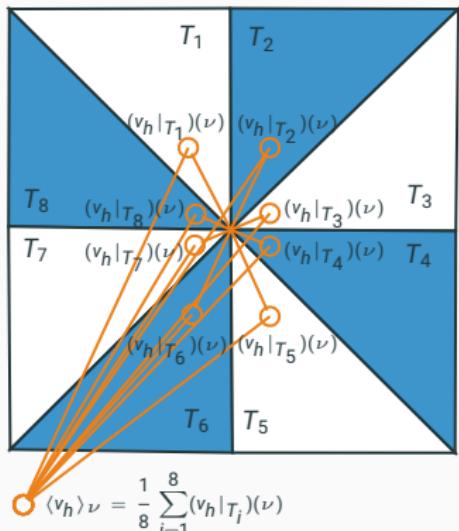


Figure:  $\Pi_h^{\text{av}} v_h \in \mathcal{S}_D^1(\mathcal{T}_h)$  if  $\Gamma_D = \{1\} \times [-1, 1]$ .

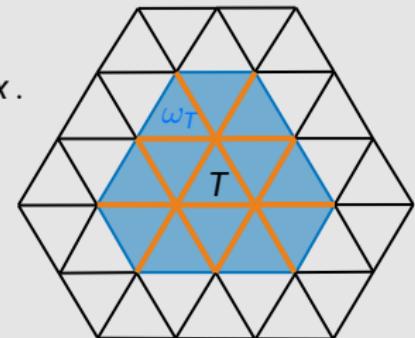
**Lemma: ((local) best-approximation property)**

For every  $v_h \in \mathcal{S}_D^{1,\text{cr}}(\mathcal{T}_h)$ , it holds that

$$\int_T |\nabla_h v_h - \nabla \Pi_h^{\text{av}} v_h|^2 dx \lesssim \inf_{v \in W_D^{1,2}(\Omega)} \int_{\omega_T} |\nabla_h v_h - \nabla v|^2 dx.$$

In particular, for every  $v_h \in \mathcal{S}_D^{1,\text{cr}}(\mathcal{T}_h)$  and  $v \in W_D^{1,2}(\Omega)$ , it holds that

$$\int_T |\nabla \Pi_h^{\text{av}} v_h - \nabla v|^2 dx \lesssim \int_{\omega_T} |\nabla_h v_h - \nabla v|^2 dx.$$



◆ **Proof.** (cf. [1, Bartels, K.]) For every  $v_h \in \mathcal{S}_D^{1,\text{cr}}(\mathcal{T}_h)$  and  $v \in W_D^{1,2}(\Omega)$ , it holds that

$$\begin{aligned} \int_T |\nabla_h v_h - \nabla \Pi_h^{\text{av}} v_h|^2 dx &\lesssim \sum_{s \in \mathcal{S}_h^i \cup \mathcal{S}_h^r : s \cap T \neq \emptyset} \int_s h_T^{-1} |[v_h]_s|^2 ds \\ &= \sum_{s \in \mathcal{S}_h^i \cup \mathcal{S}_h^r : s \cap T \neq \emptyset} \int_s h_T^{-1} \left| [v_h - v]_s - \int_s [v_h - v]_s ds \right|^2 ds \\ &\lesssim \int_{\omega_T} |\nabla_h v_h - \nabla v|^2 dx. \end{aligned}$$

■

# Adaptive algorithm

## Algorithm: (SOLVE–ESTIMATE–MARK–REFINE)

Let  $\varepsilon_{\text{STOP}} > 0$ ,  $\theta \in (0, 1]$ , and  $\mathcal{T}_0$  an initial triangulation. Then, for every  $k \geq 0$ :

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for all  $T \in \mathcal{T}_k$ . If  $\eta_{\text{gap}}^2(\bar{u}_k^{cr}, \bar{z}_k^{rt}) \leq \varepsilon_{\text{STOP}}$ , then STOP;

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Increase  $k \rightarrow k + 1$  and continue with (‘SOLVE’).

- ◆ **Marking strategies:** For *bulk parameter*  $\theta \in (0, 1]$ , find (minimal wrt. cardinality)

$$\mathcal{M}_k \subseteq \mathcal{T}_k,$$

s.t. *bulk criterion* is satisfied.

- **Max marking:** Choose  $\mathcal{M}_k \subseteq \mathcal{T}_k$  s.t.

$$\min_{T \in \mathcal{M}_k} \eta_{\text{gap}, T}^2(\bar{u}_k^{cr}, \bar{z}_k^{rt}) \geq \theta^2 \max_{T \in \mathcal{T}_h} \eta_{\text{gap}, T}^2(\bar{u}_k^{cr}, \bar{z}_k^{rt}).$$

- **Dörfler marking:** (cf. [3, Dörfler, '96]) Choose  $\mathcal{M}_k \subseteq \mathcal{T}_k$  s.t.

$$\sum_{T \in \mathcal{M}_k} \eta_{\text{gap}, T}^2(\bar{u}_k^{cr}, \bar{z}_k^{rt}) \geq \theta \sum_{T \in \mathcal{T}_k} \eta_{\text{gap}, T}^2(\bar{u}_k^{cr}, \bar{z}_k^{rt}).$$

**Construction:** For sorted set of triangles  $\mathcal{T}_k^{\text{sorted}} := \{T_1, \dots, T_{\text{card}(\mathcal{T}_k)}\}$ , where

$$i \leq j \quad \Leftrightarrow \quad \eta_{\text{gap}, T_i}^2(\bar{u}_k^{cr}, \bar{z}_k^{rt}) \geq \eta_{\text{gap}, T_j}^2(\bar{u}_k^{cr}, \bar{z}_k^{rt}),$$

we define  $\mathcal{M}_k = \{T_i \in \mathcal{T}_k^{\text{sorted}} \mid i = 1, \dots, N_k^*\}$ , where

$$N_k^* = \min \left\{ i = 1, \dots, \text{card}(\mathcal{T}_h) \mid \sum_{i=1}^{N_k^*} \eta_{\text{gap}, T_i}^2(\bar{u}_k^{cr}, \bar{z}_k^{rt}) \geq \theta \sum_{T \in \mathcal{T}_k} \eta_{\text{gap}, T}^2(\bar{u}_k^{cr}, \bar{z}_k^{rt}) \right\}.$$

# Adaptive algorithm

## Algorithm: (SOLVE–ESTIMATE–MARK–REFINE)

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for all  $T \in \mathcal{T}_k$ . If  $\eta_{\text{gap}}^2(\bar{u}_k^{cr}, \bar{z}_k^{rt}) \leq \varepsilon_{\text{STOP}}$ , then STOP;

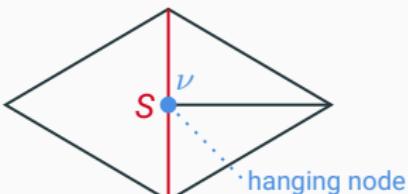
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Increase  $k \rightarrow k + 1$  and continue with (‘SOLVE’).

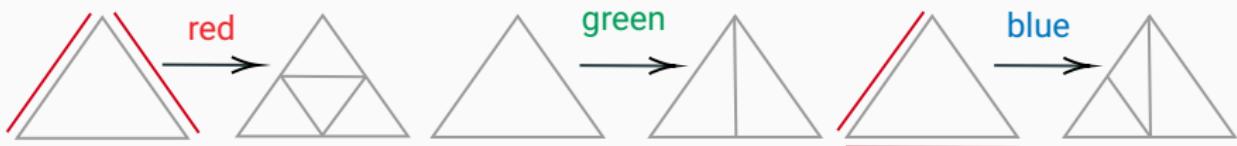
◆ Regular/irregular refinement:

- **Regular refinement:** Refine all marked elements (i.e., all  $T \in \mathcal{M}_k$ );
- **Irregular refinement:** Refine further unmarked elements to avoid *hanging nodes*, i.e.,



◆ Refinement routines:

- **Red-Green-Blue refinement:** (cf. [2, Carstensen, '04])



- **Newest-Vertex-Bisection:** (cf. [4, Mitchell, '88])



## **Examples**

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## Examples: Poisson problem

### Algorithm: (SOLVE–ESTIMATE–MARK–REFINE)

Let  $\varepsilon_{\text{STOP}} > 0$ ,  $\theta \in (0, 1]$ , and  $\mathcal{T}_0$  an initial triangulation. Then, for every  $k \geq 0$ :

- (‘SOLVE’) Compute  $u_k^{cr} \in \mathcal{S}_D^{1,cr}(\mathcal{T}_k)$  using *linear solver* and  $z_k^{rt} \in \mathcal{RT}_N^0(\mathcal{T}_k)$  using *discrete reconstruction formula*.
- (‘ESTIMATE’) Post-process  $u_k^{cr} \in \mathcal{S}_D^{1,cr}(\mathcal{T}_k)$  via *node-averaging* to obtain

$$\bar{u}_k^{cr} = \Pi_{h_k}^{\text{av}} u_k^{cr} \in \text{dom}(I);$$

Compute *(local) refinement indicators*  $\{\eta_{\text{gap},T}^2(\bar{u}_k^{cr}, z_k^{rt})\}_{T \in \mathcal{T}_k}$ , defined by

$$\eta_{\text{gap},T}^2(\bar{u}_k^{cr}, z_k^{rt}) = \frac{1}{2} \int_T |\mathbf{z}_k^{rt} - \nabla \bar{u}_k^{cr}|^2 \, dx$$

for all  $T \in \mathcal{T}_k$ . If  $\eta_{\text{gap}}^2(\bar{u}_k^{cr}, z_k^{rt}) \leq \varepsilon_{\text{STOP}}$ , then **STOP**;

- (‘MARK’) Use *Dörfler marking* to find  $\mathcal{M}_k \subseteq \mathcal{T}_k$  on which error is ‘concentrated’;
  - (‘REFINE’) Perform red-green-blue-refinement of  $\mathcal{T}_k$  to get  $\mathcal{T}_{k+1}$  s.t. each  $T \in \mathcal{M}_k$  is red-refined in  $\mathcal{T}_{k+1}$ .
- Increase  $k \rightarrow k + 1$  and continue with (‘SOLVE’).

## Examples: Poisson problem

### ♦ Implementation details wrt. step ('SOLVE'):

- **Computation of discrete primal solution:** Compute

$$u_k^{cr} \in \mathcal{S}_D^{1,cr}(\mathcal{T}_k),$$

s.t. for every  $v_k \in \mathcal{S}_D^{1,cr}(\mathcal{T}_k)$ , it holds that

$$\int_{\Omega} \nabla_{h_k} u_k^{cr} \cdot \nabla_{h_k} v_k dx = \int_{\Omega} f_{h_k} \Pi_{h_k} v_k dx,$$

using, e.g., *CG method preconditioned with incomplete LU factorization*.

### Computational costs:

$$N_k := \text{ndof}(\mathcal{S}_D^{1,cr}(\mathcal{T}_k)) = \text{card}(\mathcal{S}_h \setminus \mathcal{S}_h^{\Gamma_D}).$$

- **Computation of discrete dual solution:** Compute

$$z_k^{rt} := \nabla_{h_k} u_k^{cr} - \frac{f_{h_k}}{d} (\text{id}_{\mathbb{R}^d} - \Pi_{h_k} \text{id}_{\mathbb{R}^d}) \in \mathcal{RT}_N^0(\mathcal{T}_k),$$

i.e.,  $z_k^{rt} \in \text{dom}(-D) \cap W_N^2(\text{div}; \Omega)$  if  $f = f_{h_k} \in \mathbb{P}^0(\mathcal{T}_k)$ .

### Computational costs: for free!

## Examples: Poisson problem

### ◆ Example with corner singularity:

Let  $\Omega := (-1, 1)^2 \setminus ([0, 1] \times [-1, 0])$ ,  $\Gamma_D = \partial\Omega$ , and  $f = 1 \in L^2(\Omega)$ . Then,  $u \in W_D^{1,2}(\Omega)$  with

$$z = \nabla u \in (W^{\frac{2}{3}, 2}(\Omega))^2.$$

⇒ Expected for uniform mesh-refinement (i.e.,  $\theta = 1$ ):  $\eta_{\text{gap}}^2(\bar{u}_k^{\text{cr}}, z_k^{\text{rt}}) \sim h_k^{\frac{4}{3}} \sim N_k^{-\frac{2}{3}}$ .

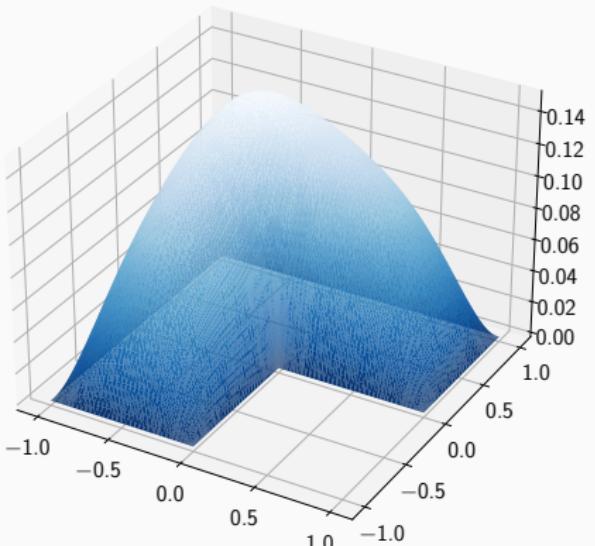


Figure:  $u \in W_D^{1,2}(\Omega)$ .

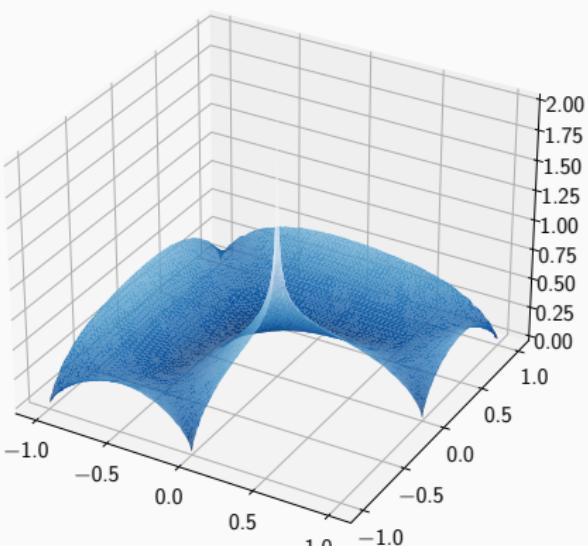
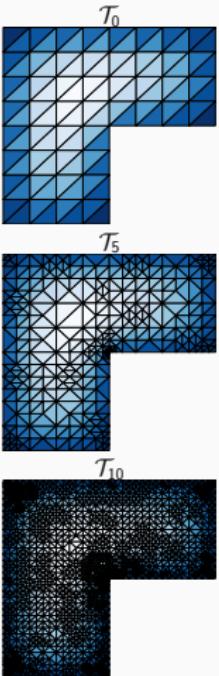
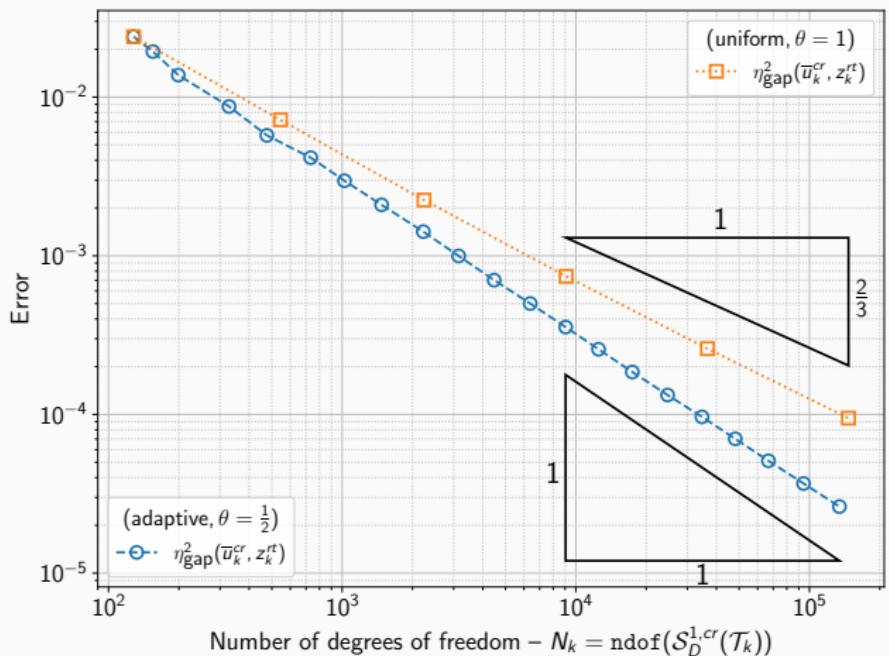


Figure:  $|\nabla u| \in L^2(\Omega)$ .

## Examples: Poisson problem



### Observations:

- Adaptive mesh refinement yields optimal convergence rate  $h_k^2 \sim N_k^{-1}$ .
- Uniform mesh refinement yields reduced convergence rate  $h_k^{\frac{4}{3}} \sim N_k^{-\frac{2}{3}}$ .

## Examples: $p$ -Dirichlet problem

### Algorithm: (SOLVE–ESTIMATE–MARK–REFINE)

Let  $\varepsilon_{\text{STOP}} > 0$ ,  $\theta \in (0, 1]$ , and  $\mathcal{T}_0$  an initial triangulation. Then, for every  $k \geq 0$ :

- ('SOLVE') Compute  $u_k^{cr} \in \mathcal{S}_D^{1,cr}(\mathcal{T}_k)$  using *non-linear solver* and  $z_k^{rt} \in \mathcal{RT}_N^0(\mathcal{T}_k)$  using *discrete reconstruction formula*.
  - ('ESTIMATE') Post-process  $u_k^{cr} \in \mathcal{S}_D^{1,cr}(\mathcal{T}_k)$  via *node-averaging* to obtain
$$\bar{u}_k^{cr} = \Pi_{h_k}^{\text{av}} u_k^{cr} \in \text{dom}(I);$$
Compute *(local) refinement indicators*  $\{\eta_{\text{gap},T}^2(\bar{u}_k^{cr}, z_k^{rt})\}_{T \in \mathcal{T}_k}$ , defined by
$$\eta_{\text{gap},T}^2(\bar{u}_k^{cr}, z_k^{rt}) = \int_T \left\{ \frac{1}{p'} |z_k^{rt}|^{p'} - z_k^{rt} \cdot \nabla \bar{u}_k^{cr} + \frac{1}{p} |\nabla \bar{u}_k^{cr}|^p \right\} dx$$
for all  $T \in \mathcal{T}_k$ . If  $\eta_{\text{gap}}^2(\bar{u}_k^{cr}, z_k^{rt}) \leq \varepsilon_{\text{STOP}}$ , then **STOP**;
  - ('MARK') Use *Dörfler marking* to find  $\mathcal{M}_k \subseteq \mathcal{T}_k$  on which error is '*concentrated*';
  - ('REFINE') Perform *red-green-blue*-refinement of  $\mathcal{T}_k$  to get  $\mathcal{T}_{k+1}$  s.t. each  $T \in \mathcal{M}_k$  is *red*-refined in  $\mathcal{T}_{k+1}$ .
- Increase  $k \rightarrow k + 1$  and continue with ('SOLVE').

◆ Implementation details wrt. step ('SOLVE'):

- Computation of discrete primal solution: Compute

$$u_k^{cr} \in \mathcal{S}_D^{1,cr}(\mathcal{T}_k),$$

s.t. for every  $v_k \in \mathcal{S}_D^{1,cr}(\mathcal{T}_k)$ , it holds that

$$\int_{\Omega} D\phi_k(\nabla_{h_k} u_k^{cr}) \cdot \nabla_{h_k} v_k dx = \int_{\Omega} f_{h_k} \Pi_{h_k} v_k dx,$$

where  $\phi_k \in C^1(\mathbb{R}^d)$  is defined by

$$\phi_k(0) = 0 \quad \text{and} \quad D\phi_k(t) = (h_k^2 + |t|)^{p-2} t \quad \text{for all } t \in \mathbb{R}^d,$$

using, e.g., *Newton scheme with line search*.

**Computational costs:** (in each Newton step)

$$N_k := \text{ndof}(\mathcal{S}_D^{1,cr}(\mathcal{T}_k)) = \text{card}(\mathcal{S}_h \setminus \mathcal{S}_h^{r_D}).$$

- Computation of discrete dual solution: Compute

$$z_k^{rt} := D\phi_k(\nabla_{h_k} u_k^{cr}) - \frac{f_{h_k}}{d} (\text{id}_{\mathbb{R}^d} - \Pi_{h_k} \text{id}_{\mathbb{R}^d}) \in \mathcal{RT}_N^0(\mathcal{T}_k),$$

i.e.,  $z_k^{rt} \in \text{dom}(-D) \cap W_N^{p'}(\text{div}; \Omega)$  if  $f = f_{h_k} \in \mathbb{P}^0(\mathcal{T}_k)$ .

**Computational costs:** for free!

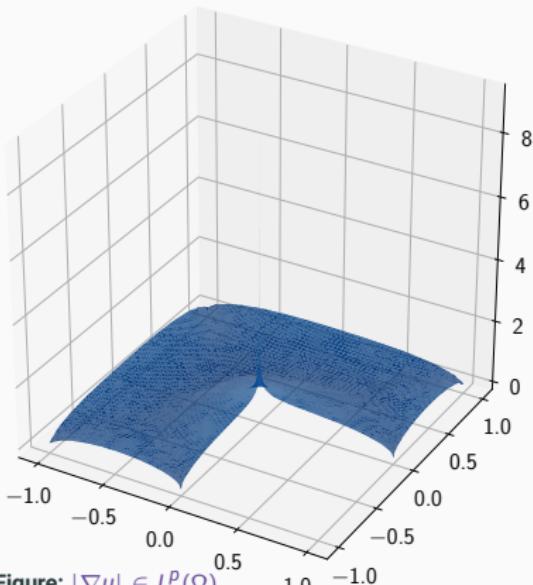
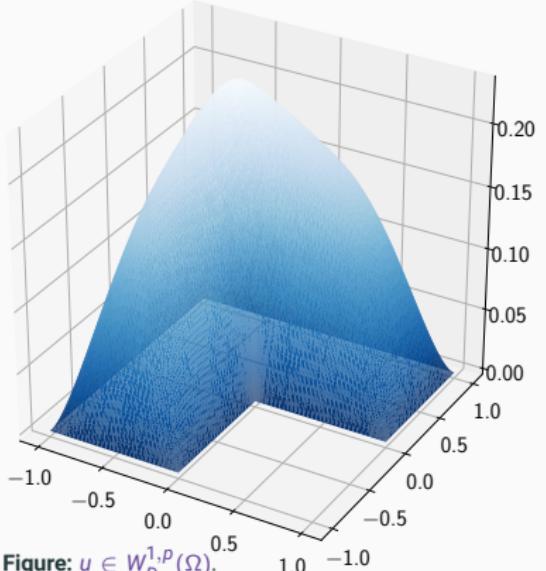
## Examples: $p$ -Dirichlet problem

### ◆ Example with corner singularity:

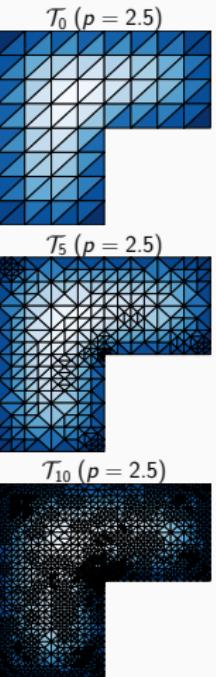
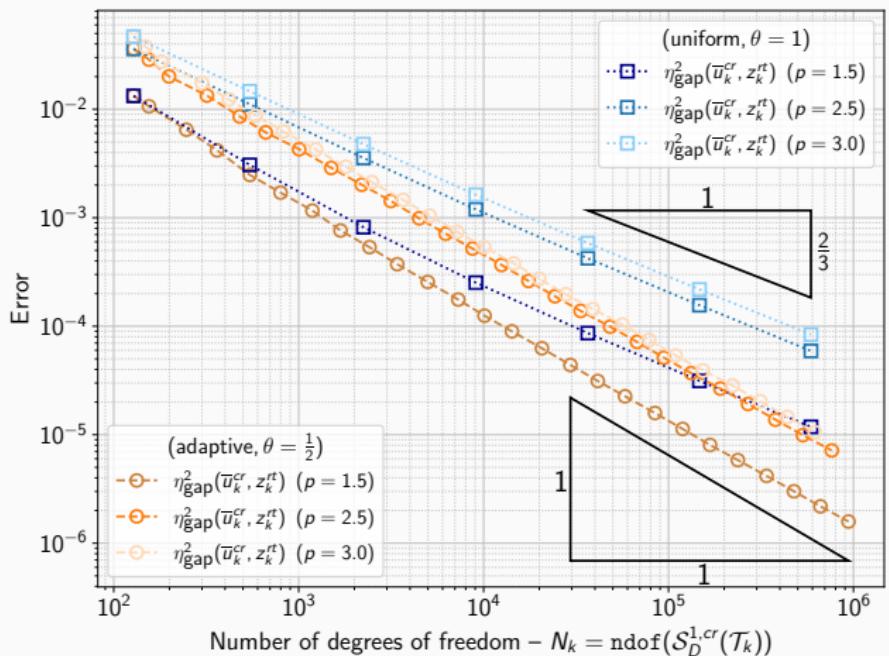
Let  $\Omega = (-1, 1)^2 \setminus ([0, 1] \times [-1, 0])$ ,  $\Gamma_D = \partial\Omega$ , and  $f = 1 \in L^{p'}(\Omega)$ ,  $p \in (1, \infty)$ . Then,  $u \in W_D^{1,p}(\Omega)$  with

$$V(\nabla u) = V^*(z) \in (W^{\frac{2}{3}, 2}(\Omega))^2.$$

⇒ Expected for uniform mesh-refinement (i.e.,  $\theta = 1$ ):  $\eta_{\text{gap}}^2(\bar{u}_k^{\text{cr}}, z_k^{\text{rt}}) \sim h_k^{\frac{4}{3}} \sim N_k^{-\frac{2}{3}}$ .



# Examples: $p$ -Dirichlet problem



## Observations:

- Adaptive mesh refinement yields optimal convergence rate  $h_k^2 \sim N_k^{-1}$ .
- Uniform mesh refinement yields reduced convergence rate  $h_k \sim N_k^{-\frac{2}{3}}$ .

## Examples: Obstacle problem

### Algorithm: (SOLVE–ESTIMATE–MARK–REFINE)

Let  $\varepsilon_{\text{STOP}} > 0$ ,  $\theta \in (0, 1]$ , and  $\mathcal{T}_0$  an initial triangulation. Then, for every  $k \geq 0$ :

- ('SOLVE') Compute  $u_k^{cr} \in S_D^{1,cr}(\mathcal{T}_k)$  using *non-linear solver* and  $z_k^{rt} \in \mathcal{R}\mathcal{T}_N^0(\mathcal{T}_k)$  using *discrete reconstruction formula*.
- ('ESTIMATE') Post-process  $u_k^{cr} \in S_D^{1,cr}(\mathcal{T}_k)$  via *node-averaging* and *cut-off* to obtain

$$\bar{u}_k^{cr} := \max\{0, \Pi_{h_k}^{av} u_k^{cr}\} \in \text{dom}(I);$$

Compute (*local*) *refinement indicators*  $\{\eta_{\text{gap}, T}^2(\bar{u}_k^{cr}, z_k^{rt})\}_{T \in \mathcal{T}_k}$ , defined by

$$\eta_{\text{gap}, T}^2(\bar{u}_k^{cr}, z_k^{rt}) = \left\{ \begin{array}{l} \frac{1}{2} \int_T |z_k^{rt} - \nabla \bar{u}_k^{cr}|^2 \, dx \\ - \int_T (f_{h_k} + \operatorname{div} z_k^{rt}) \bar{u}_k^{cr} \, dx \end{array} \right\}$$

for all  $T \in \mathcal{T}_k$ . If  $\eta_{\text{gap}}^2(\bar{u}_k^{cr}, z_k^{rt}) \leq \varepsilon_{\text{STOP}}$ , then **STOP**;

- ('MARK') Use *Dörfler marking* to find  $\mathcal{M}_k \subseteq \mathcal{T}_k$  on which error is '*concentrated*';
  - ('REFINE') Perform *red-green-blue*-refinement of  $\mathcal{T}_k$  to get  $\mathcal{T}_{k+1}$  s.t. each  $T \in \mathcal{M}_k$  is *red*-refined in  $\mathcal{T}_{k+1}$ .
- Increase  $k \rightarrow k + 1$  and continue with ('SOLVE').

## Examples: Obstacle problem

### ♦ Implementation details wrt. ('SOLVE'):

- Computation of discrete primal solution: Compute

$$u_k^{cr} \in \mathcal{S}_D^{1,cr}(\mathcal{T}_k) \quad \& \quad \lambda_k^{cr} \in \Pi_{h_k}(\mathcal{S}_D^{1,cr}(\mathcal{T}_k))$$

s.t.  $\lambda_k^{cr} \leq 0$  a.e. in  $\Omega$  and for every  $v_k \in \mathcal{S}_D^{1,cr}(\mathcal{T}_k)$ , it holds that

$$\int_{\Omega} \nabla_{h_k} u_k^{cr} \cdot \nabla_{h_k} v_k \, dx = \int_{\Omega} (f_{h_k} - \lambda_k^{cr}) \Pi_{h_k} v_k \, dx,$$

using, e.g., *primal-dual active set strategy*, i.e., *semi-smooth Newton method*.

**Computational costs:** (in each semi-smooth Newton step)

$$N_k = \text{ndof}(\mathcal{S}_D^{1,cr}(\mathcal{T}_k)) + \text{ndof}(\Pi_{h_k}(\mathcal{S}_D^{1,cr}(\mathcal{T}_k))).$$

- Computation of discrete dual solution: Compute

$$z_k^{rt} = \nabla_{h_k} u_k^{cr} - \frac{f_{h_k} - \lambda_k^{cr}}{d} (\text{id}_{\mathbb{R}^d} - \Pi_{h_k} \text{id}_{\mathbb{R}^d}) \in \mathcal{RT}_N^0(\mathcal{T}_k).$$

i.e.,  $z_k^{rt} \in \text{dom}(-D) \cap W_N^2(\text{div}; \Omega)$  if  $f = f_{h_k} \in \mathbb{P}^0(\mathcal{T}_k)$ .

**Computational costs:** for free!

## Examples: Obstacle problem

### ◆ Example with corner singularity:

Let  $\Omega := (-1, 1)^2$ ,  $\Gamma_D = \partial\Omega$ ,  $f = 1 \in L^2(\Omega)$ , and  $\chi = 1 - |\cdot|_\infty \in \mathcal{S}^1(\mathcal{T}_0)$ . Then,  $u \in W_D^{1,2}(\Omega)$  with

$$z = \nabla u \in (W^{\frac{1}{2}, 2}(\Omega))^2.$$

⇒ Expected for uniform mesh-refinement (i.e.,  $\theta = 1$ ):  $\eta_{\text{gap}}^2(\bar{u}_k^{cr}, z_k^{rt}) \sim h_k \sim N_k^{-\frac{1}{2}}$ .

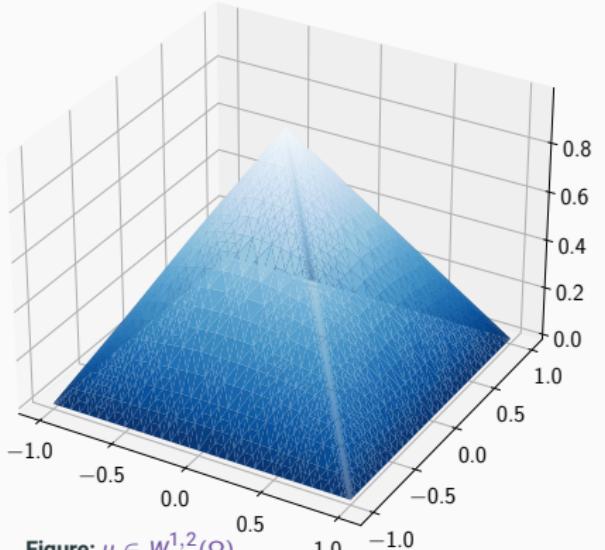


Figure:  $u \in W_D^{1,2}(\Omega)$ .

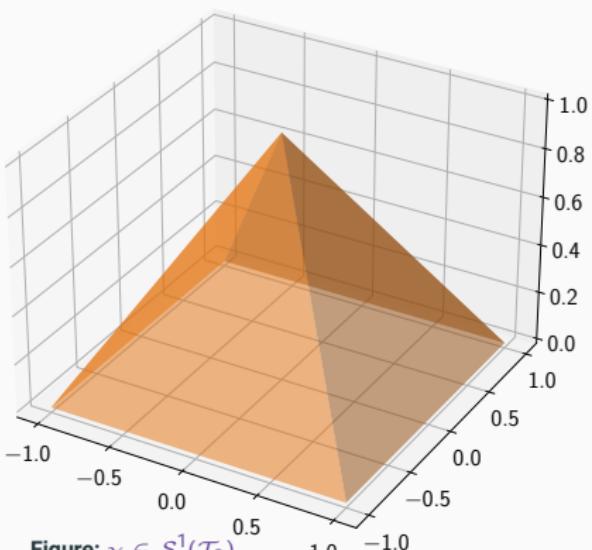
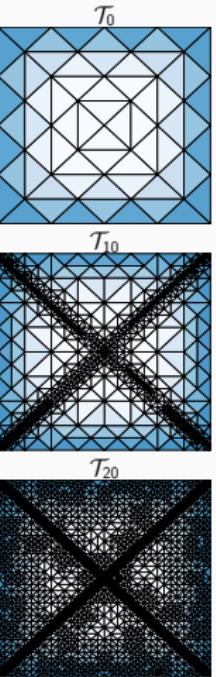
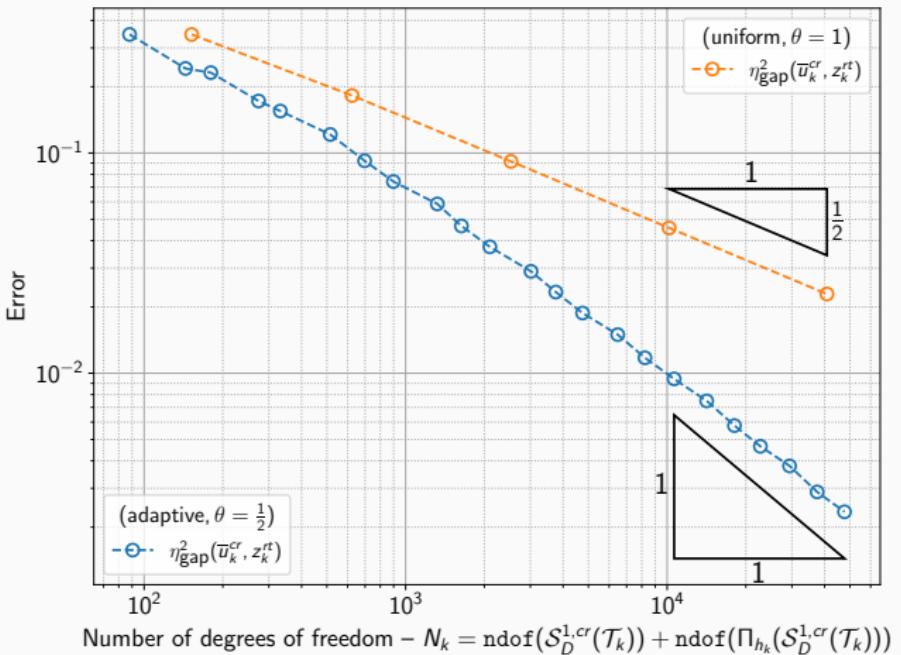


Figure:  $\chi \in \mathcal{S}^1(\mathcal{T}_0)$ .

## Examples: Obstacle problem



### Observations:

- Adaptive mesh refinement yields optimal convergence rate  $h_k^2 \sim N_k^{-1}$ .
- Uniform mesh refinement yields reduced convergence rate  $h_k^1 \sim N_k^{\frac{1}{2}}$ .

## ◆ A posteriori error analysis: (on the basis of convex duality)

- Primal-dual gap identity: for every  $\mathbf{v} \in \text{dom}(I)$  and  $\mathbf{y} \in \text{dom}(-D)$ , it holds that

$$\rho_{\text{tot}}^2(\mathbf{v}, \mathbf{y}) = \eta_{\text{gap}}^2(\mathbf{v}, \mathbf{y}).$$

- Advantages/Challenges:

systematic;    widely applicable;    }  
 simple;        identical;        }      vs.    ? numerically practicable.

## ◆ A posteriori error analysis: (on the basis of discrete convex duality)

- Discrete reconstruction formulas: imply numerical practicability.

$$z_h^{rt} = D_t \phi_h(\cdot, \nabla_h u_h^{cr}) + \frac{D_t \psi_h(\cdot, \Pi_h u_h^{cr})}{d} (\text{id}_{\mathbb{R}^d} - \Pi_h \text{id}_{\mathbb{R}^d}),$$

$$u_h^{cr} = D_t \psi_h^*(\cdot, \text{div} z_h^{rt}) + D_t \phi_h^*(\cdot, \Pi_h z_h^{rt}) \cdot (\text{id}_{\mathbb{R}^d} - \Pi_h \text{id}_{\mathbb{R}^d}).$$

*A priori* error identity  
on the basis of discrete convex duality

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# Discrete primal-dual gap identity

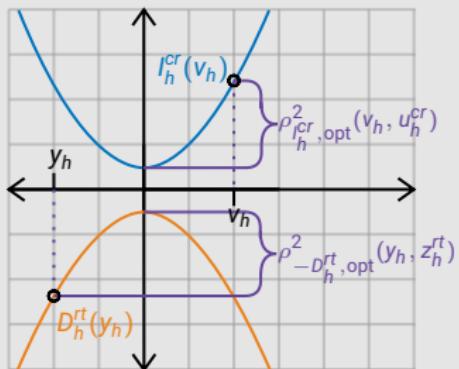
**Definition:** (discrete primal-dual total error & discrete primal-dual gap estimator)

- The *discrete primal-dual total error*

$\rho_{\text{tot},h}^2 : \text{dom}(I_h^{cr}) \times \text{dom}(-D_h^{rt}) \rightarrow [0, +\infty)$   
is defined by

$$\rho_{\text{tot},h}^2(v_h, y_h) := \rho_{h,\text{opt}}^2(v_h, u_h^{cr}) + \rho_{-D_h^{rt},\text{opt}}^2(y_h, z_h^{rt})$$

for all  $(v_h, y_h)^\top \in \text{dom}(I_h^{cr}) \times \text{dom}(-D_h^{rt})$ .

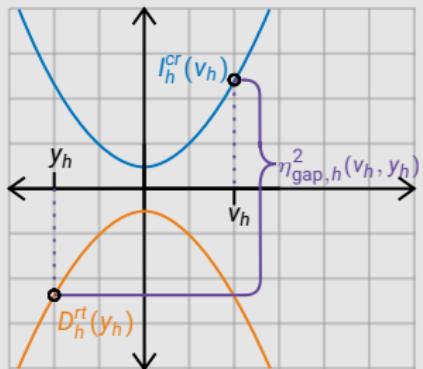


- The *discrete primal-dual gap estimator*

$\eta_{\text{gap},h}^2 : \text{dom}(I_h^{cr}) \times \text{dom}(-D_h^{rt}) \rightarrow [0, +\infty)$   
is defined by

$$\eta_{\text{gap},h}^2(v_h, y_h) := I_h^{cr}(v_h) - D_h^{rt}(y_h)$$

for all  $(v_h, y_h)^\top \in \text{dom}(I_h^{cr}) \times \text{dom}(-D_h^{rt})$ .



# Discrete primal-dual gap identity

**Theorem: (discrete primal-dual gap identity)**

If a *discrete strong duality relation* applies, i.e.,

$$I_h^{cr}(u_h^{cr}) = D_h^{rt}(z_h^{rt}),$$

then for every  $(v_h, y_h)^\top \in \text{dom}(I_h^{cr}) \times \text{dom}(-D_h^{rt})$ , it holds that

$$\rho_{\text{tot},h}^2(v_h, y_h) = \eta_{\text{gap},h}^2(v_h, y_h).$$

◆ **Proof.** For every  $(v_h, y_h)^\top \in \text{dom}(I_h^{cr}) \times \text{dom}(-D_h^{rt})$ , we have that

$$\begin{aligned}\rho_{\text{tot},h}^2(v_h, y_h) &= \rho_{I_h^{cr}, \text{opt}}^2(v_h, u_h^{cr}) + \rho_{-D_h^{rt}, \text{opt}}^2(y_h, z_h^{rt}) \\ &\stackrel{\text{def}}{=} [I_h^{cr}(v_h) - I_h^{cr}(u_h^{cr})] \\ &\quad + [D_h^{rt}(z_h^{rt}) - D_h^{rt}(y_h)] \\ &= I_h^{cr}(v_h) - D_h^{rt}(y_h) \\ &\quad - \underbrace{[D_h^{rt}(z_h^{rt}) - I_h^{cr}(u_h^{cr})]}_{=0} \\ &= \eta_{\text{gap},h}^2(v_h, y_h).\end{aligned}$$

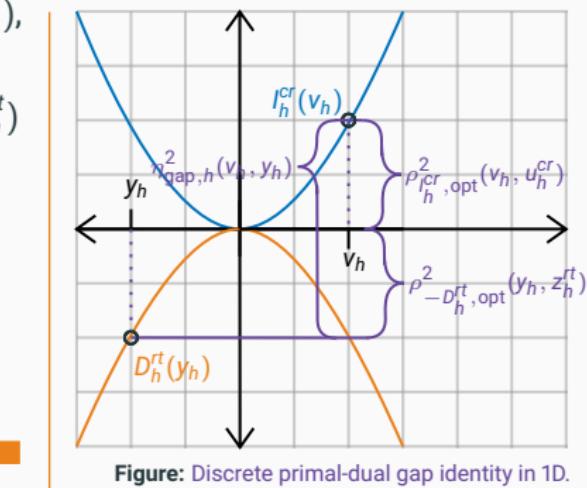


Figure: Discrete primal-dual gap identity in 1D.

### Theorem: (*a priori* error identity)

If  $u \in W_D^{1+s,2}(\Omega)$ ,  $s \in [0, 1]$ , i.e.,  $z \in (W^{s,2}(\Omega))^d$ , and  $f_h = \Pi_h f \in \mathbb{P}^0(\mathcal{T}_h)$ , then

$$\begin{aligned} & \frac{1}{2} \int_{\Omega} |\nabla_h \Pi_h^{cr} u - \nabla_h u_h^{cr}|^2 dx + \frac{1}{2} \int_{\Omega} |\Pi_h \Pi_h^{rt} z - \Pi_h z_h^{rt}|^2 dx \\ &= \frac{1}{2} \int_{\Omega} |\Pi_h z - \Pi_h \Pi_h^{rt} z|^2 dx \lesssim h^{2s} [z]_{s,2}^2. \end{aligned}$$

### Proof.

- For every  $v_h \in \text{dom}(I_h^{cr})$  and  $y_h \in \text{dom}(-D_h^{rt})$ , we have that

$$\begin{aligned} \rho_{I_h^{cr}, \text{opt}}^2(v_h, u_h^{cr}) &= \frac{1}{2} \int_{\Omega} |\nabla_h v_h - \nabla_h u_h^{cr}|^2 dx, \\ \rho_{-D_h^{rt}, \text{opt}}^2(y_h, z_h^{rt}) &= \frac{1}{2} \int_{\Omega} |\Pi_h y_h - \Pi_h z_h^{rt}|^2 dx. \end{aligned}$$

- For every  $v_h \in \text{dom}(I_h^{cr})$  and  $y_h \in \text{dom}(-D_h^{rt})$ , we have that

$$\eta_{\text{gap},h}^2(v_h, y_h) = \frac{1}{2} \int_{\Omega} |\nabla_h v_h - \Pi_h y_h|^2 dx.$$

## Examples: Poisson problem

- It holds that  $\Pi_h^{rt} z \in \text{dom}(-D_h^{rt})$ , since

$$\begin{aligned}\operatorname{div} \Pi_h^{rt} z &= \Pi_h \operatorname{div} z = \Pi_h[-f] = -f_h && \text{a.e. in } \Omega, \\ \Pi_h^{rt} z \cdot n &= \pi_h[z \cdot n] = 0 && \text{a.e. in } \Gamma_N.\end{aligned}$$

- It holds that  $\Pi_h^{cr} u \in \text{dom}(I_h^{cr})$ , since

$$\begin{aligned}\nabla_h \Pi_h^{cr} u &= \Pi_h \nabla u = \Pi_h z && \text{a.e. in } \Omega, \\ \pi_h \Pi_h^{cr} u &= \pi_h u = 0 && \text{a.e. in } \Gamma_D.\end{aligned}$$

For every  $v_h = \Pi_h^{cr} u \in \text{dom}(I_h^{cr})$  and  $y_h = \Pi_h^{rt} z \in \text{dom}(-D_h^{rt})$ , we conclude that

$$\begin{aligned}\frac{1}{2} \int_{\Omega} |\nabla_h \Pi_h^{cr} u - \nabla_h u_h^{cr}|^2 dx + \frac{1}{2} \int_{\Omega} |\Pi_h \Pi_h^{rt} z - \Pi_h z_h^{rt}|^2 dx &= \frac{1}{2} \int_{\Omega} |\nabla_h \Pi_h^{cr} u - \Pi_h \Pi_h^{rt} z|^2 dx \\ &= \frac{1}{2} \int_{\Omega} |\Pi_h z - \Pi_h \Pi_h^{rt} z|^2 dx \\ &\leq \frac{1}{2} \int_{\Omega} |z - \Pi_h^{rt} z|^2 dx \\ &\lesssim h^{2s} [z]_{s,2}^2.\end{aligned}$$



## Examples: Poisson problem

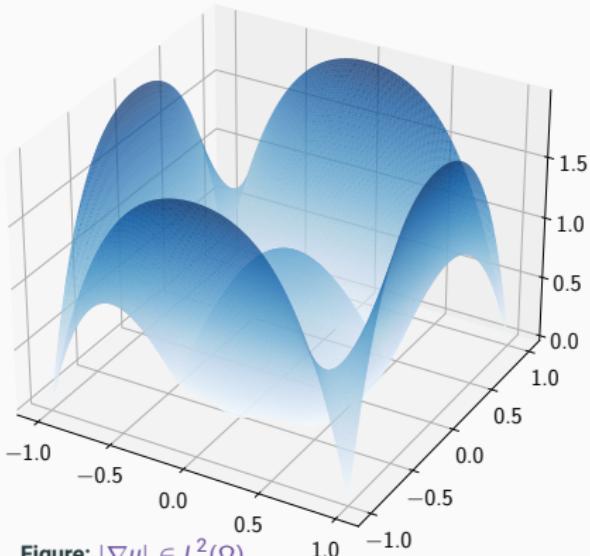
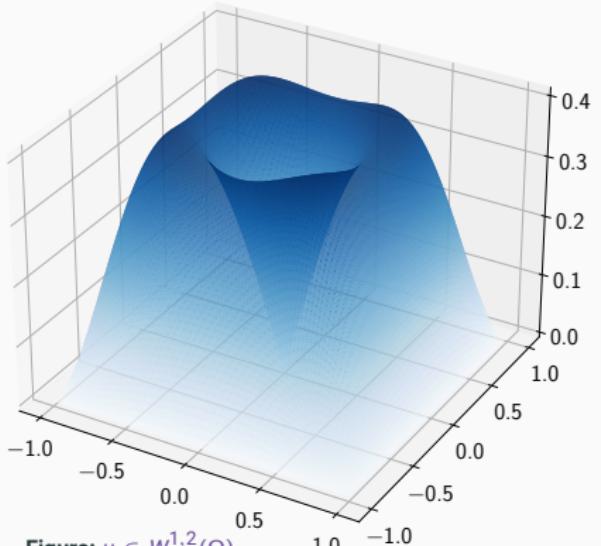
### Example with central singularity:

Let  $\Omega := (-1, 1)^2$ ,  $\Gamma_D = \partial\Omega$ , and  $f \in L^2(\Omega)$  s.t.  $u \in W_D^{1,2}(\Omega)$  is given via

$$u(x_1, x_2) := (1 - x_1^2)(1 - x_2^2)|(x_1, x_2)^\top|^{s+\delta},$$

where  $s \in (0, 1]$  and  $\delta = 1.0 \times 10^{-5}$ , so that  $z = \nabla u \in (W^{s,2}(\Omega))^2$ .

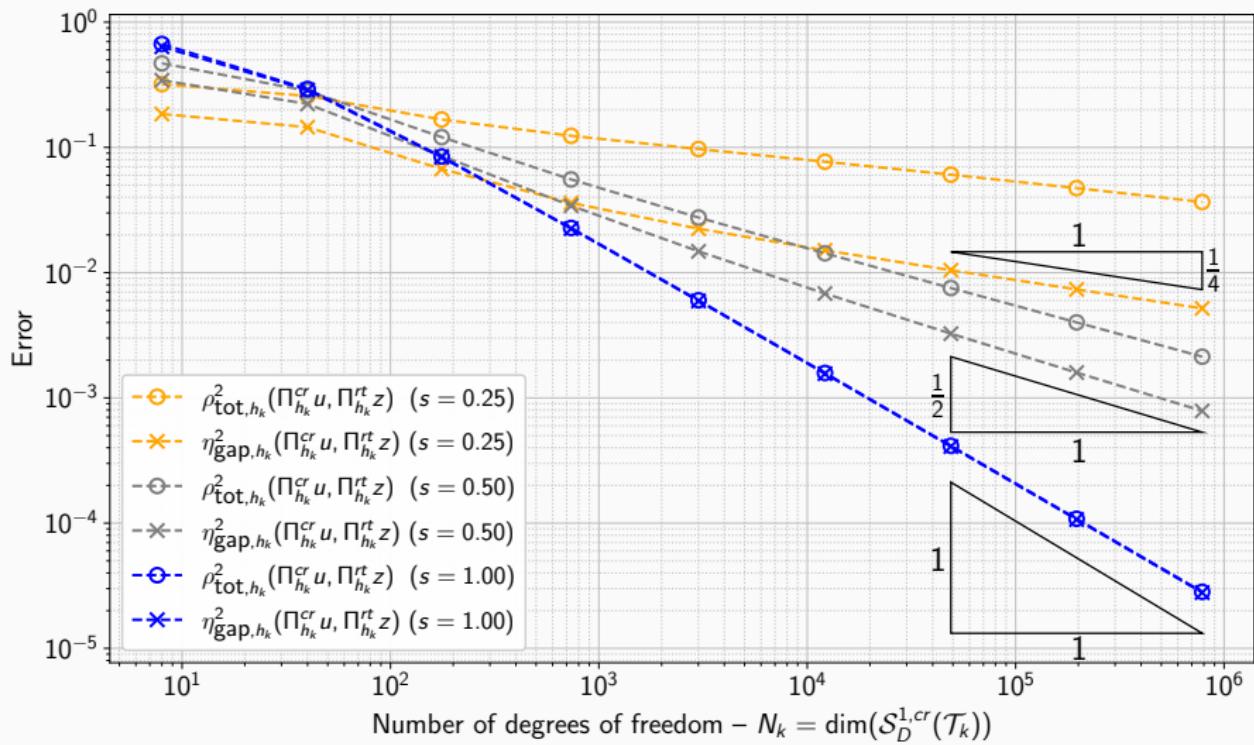
⇒ Expected for uniform mesh-refinement (i.e.,  $\theta = 1$ ):  $\eta_{\text{gap}, h_k}^2(\Pi_{h_k}^{\text{cr}} u, \Pi_{h_k}^{\text{rt}} z) \sim h_k^{2s} \sim N_k^{-s}$ .



# Examples: Poisson problem

## Experiment:

### Results:



## Theorem: (*a priori* error equivalence)

If  $V(\nabla u) \in (W^{s,2}(\Omega))^d$ ,  $s \in (0, 1]$ , i.e.,  $V^*(z) \in (W^{s,2}(\Omega))^d$ , and  $f_h = \Pi_h f \in \mathbb{P}^0(\mathcal{T}_h)$ , then

$$\begin{aligned} & \int_{\Omega} |V(\nabla_h \Pi_h^{cr} u) - V(\nabla_h u_h^{cr})|^2 dx + \int_{\Omega} |V^*(\Pi_h \Pi_h^{rt} z) - V^*(\Pi_h z_h^{rt})|^2 dx \\ & \sim \int_{\Omega} \left\{ \frac{1}{p} |\Pi_h \nabla u|^p - \Pi_h \nabla u \cdot \Pi_h \Pi_h^{rt} z + \frac{1}{p'} |\Pi_h \Pi_h^{rt} z|^{p'} \right\} dx \lesssim h^{2s} [V^*(z)]_{s,2}^2. \end{aligned}$$

### ◆ Proof.

- For every  $v_h \in \text{dom}(I_h^{cr})$  and  $y_h \in \text{dom}(-D_h^{rt})$ , we have that

$$\begin{aligned} \rho_{I_h^{cr}, \text{opt}}^2(v_h, u_h^{cr}) & \sim \int_{\Omega} |V(\nabla_h v_h) - V(\nabla_h u_h^{cr})|^2 dx, \\ \rho_{-D_h^{rt}, \text{opt}}^2(y_h, z_h^{rt}) & \sim \int_{\Omega} |V^*(\Pi_h y_h) - V^*(\Pi_h z_h^{rt})|^2 dx. \end{aligned}$$

- For every  $v_h \in \text{dom}(I_h^{cr})$  and  $y_h \in \text{dom}(-D_h^{rt})$ , we have that

$$\eta_{\text{gap},h}^2(v_h, y_h) = \int_{\Omega} \left\{ \frac{1}{p} |\nabla_h v_h|^p - \nabla_h v_h \cdot \Pi_h y_h + \frac{1}{p'} |\Pi_h y_h|^{p'} \right\} dx.$$

## Examples: $p$ -Dirichlet problem

→ For every  $v_h = \Pi_h^{cr} u \in \text{dom}(I_h^{cr})$  and  $y_h = \Pi_h^{rt} z \in \text{dom}(-D_h^{rt})$ , we conclude that

$$\begin{aligned}
& \int_{\Omega} |V(\nabla_h \Pi_h^{cr} \nabla u) - V(\nabla_h u_h^{cr})|^2 dx + \int_{\Omega} |V^*(\Pi_h \Pi_h^{rt} z) - V^*(\Pi_h z_h^{rt})|^2 dx \\
& \sim \int_{\Omega} \left\{ \frac{1}{p} |\nabla_h \Pi_h^{cr} u|^p - \nabla_h \Pi_h^{cr} u \cdot \Pi_h \Pi_h^{rt} z + \frac{1}{p'} |\Pi_h \Pi_h^{rt} z|^{p'} \right\} dx \\
& \leq \underbrace{\int_{\Omega} \left\{ \frac{1}{p} |\nabla u|^p - \nabla u \cdot \Pi_h \Pi_h^{rt} z + \frac{1}{p'} |\Pi_h \Pi_h^{rt} z|^{p'} \right\} dx}_{= \nabla u \cdot z - \frac{1}{p'} |z|^{p'}} \\
& \quad = |z|^{p'-2} z \\
& = \int_{\Omega} \left\{ \underbrace{-\frac{1}{p'} |z|^{p'}}_{\leq -\frac{1}{p'} |\Pi_h \Pi_h^{rt} z|^{p'}} + \underbrace{\nabla u \cdot (z - \Pi_h \Pi_h^{rt} z)}_{= |\Pi_h \Pi_h^{rt} z|^{p'-2} \Pi_h \Pi_h^{rt} z \cdot (z - \Pi_h \Pi_h^{rt} z)} + \frac{1}{p'} |\Pi_h \Pi_h^{rt} z|^{p'} \right\} dx \\
& \leq \int_{\Omega} (|z|^{p'-2} z - |\Pi_h \Pi_h^{rt} z|^{p'-2} \Pi_h \Pi_h^{rt} z) \cdot (z - \Pi_h \Pi_h^{rt} z) dx \\
& \sim \int_{\Omega} |V^*(z) - V^*(\Pi_h \Pi_h^{rt} z)|^2 dx \\
& \lesssim h^{2s} [V^*(z)]_{s,2}^2.
\end{aligned}$$



## Examples: $p$ -Dirichlet problem

### Example with central singularity:

Let  $\Omega := (-1, 1)^2$ ,  $\Gamma_D = \partial\Omega$ , and  $f \in L^{p'}(\Omega)$ ,  $p \in (1, +\infty)$ , s.t.  $u \in W_D^{1,p}(\Omega)$  is given via

$$u(x_1, x_2) := (1 - x_1^2)(1 - x_2^2)|(x_1, x_2)^\top|^{1+2\frac{s-1}{p}+\delta},$$

where  $s \in (0, 1]$  and  $\delta = 1.0 \times 10^{-5}$ , so that  $V^*(z) = V(\nabla u) \in (W^{s,2}(\Omega))^2$ .

⇒ Expected for uniform mesh-refinement (i.e.,  $\theta = 1$ ):  $\eta_{\text{gap}, h_k}^2(\Pi_{h_k}^{cr} u, \Pi_{h_k}^{rt} z) \sim h_k^{2s} \sim N_k^{-s}$ .

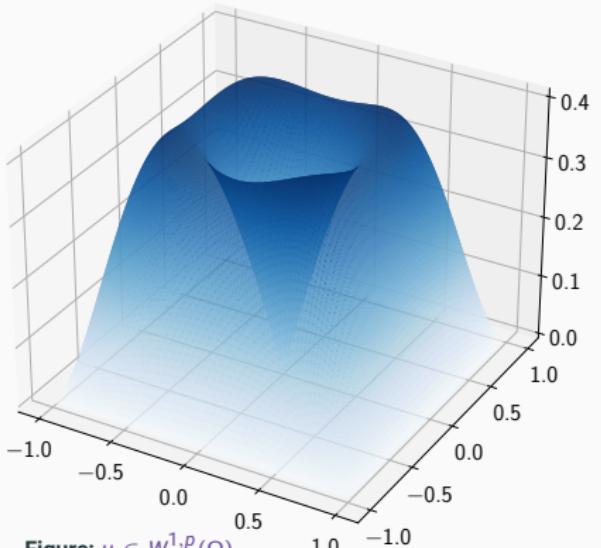


Figure:  $u \in W_D^{1,p}(\Omega)$ .

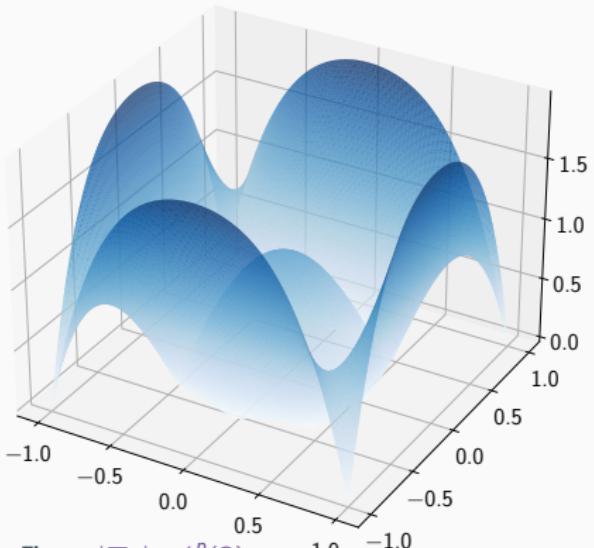
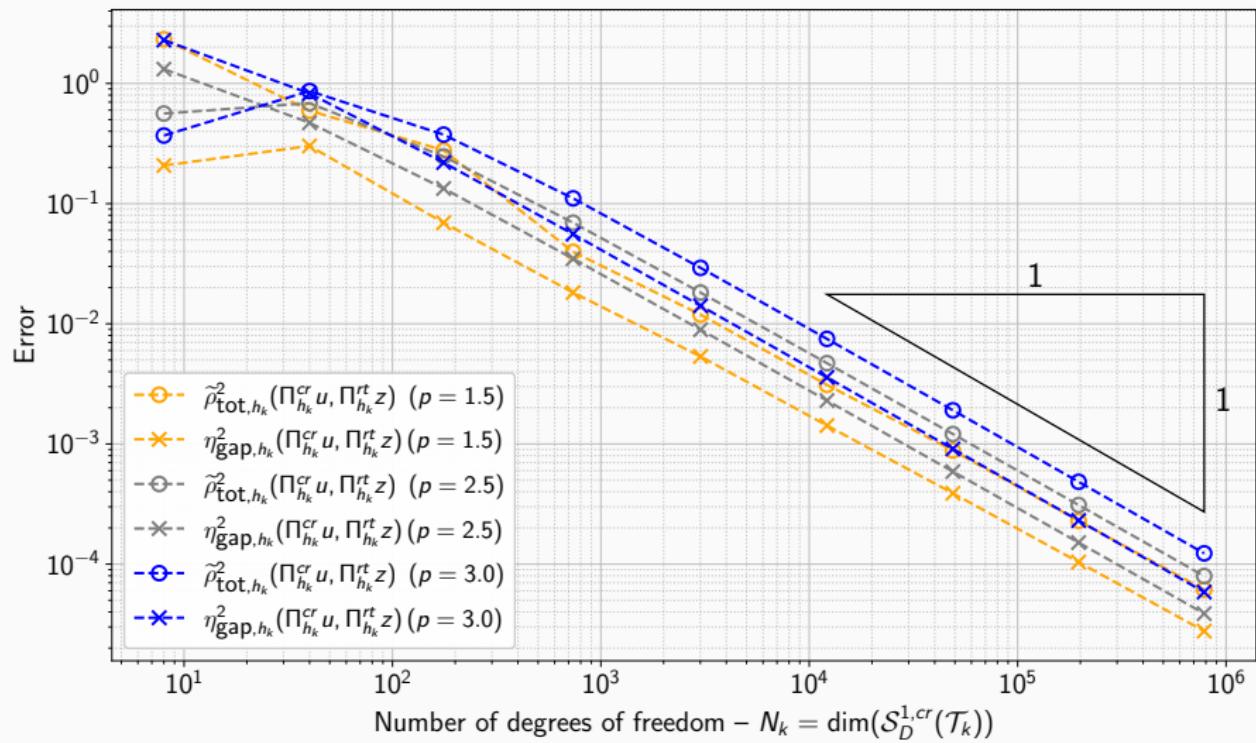


Figure:  $|\nabla u| \in L^p(\Omega)$ .

# Examples: $p$ -Dirichlet problem

## Experiment:

### Results:



## Examples: Obstacle problem

### Theorem: (*a priori* error identity)

If  $z \in (W^{s,2}(\Omega))^d \cap W_N^2(\operatorname{div}; \Omega)$  and  $f_h = \Pi_h f \in \mathbb{P}^0(\mathcal{T}_h)$ , then

$$\begin{aligned} & \frac{1}{2} \int_{\Omega} |\nabla_h \Pi_h^{cr} u - \nabla_h u_h^{cr}|^2 dx - \int_{\Omega} (f_h + \operatorname{div} z_h^{rt}) \Pi_h \Pi_h^{cr} u dx \\ & + \frac{1}{2} \int_{\Omega} |\Pi_h \Pi_h^{rt} z - \Pi_h z_h^{rt}|^2 dx - \int_{\Omega} (f + \operatorname{div} z) \Pi_h u_h^{cr} dx \\ & = \frac{1}{2} \int_{\Omega} |\Pi_h z - \Pi_h \Pi_h^{rt} z|^2 dx + \int_{\Omega} (f + \operatorname{div} z) (u - \Pi_h \Pi_h^{cr} u) dx \leq h^{2s} [z]_{s,2}^2. \end{aligned}$$

### Proof.

- For every  $v_h \in \operatorname{dom}(I_h^{cr})$  and  $y_h \in \operatorname{dom}(-D_h^{rt})$ , we have that

$$\rho_{I_h^{cr}, \text{opt}}^2(v_h, u_h^{cr}) = \frac{1}{2} \int_{\Omega} |\nabla_h v_h - \nabla_h u_h^{cr}|^2 dx - \int_{\Omega} (f_h + \operatorname{div} z_h^{rt}) \Pi_h v_h dx,$$

$$\rho_{-D_h^{rt}, \text{opt}}^2(y_h, z_h^{rt}) = \frac{1}{2} \int_{\Omega} |\Pi_h y_h - \Pi_h z_h^{rt}|^2 dx - \int_{\Omega} (f_h + \operatorname{div} y_h) \Pi_h u_h^{cr} dx.$$

- For every  $v_h \in \operatorname{dom}(I_h^{cr})$  and  $y_h \in \operatorname{dom}(-D_h^{rt})$ , we have that

$$\eta_{\text{gap},h}^2(v_h, y_h) = \frac{1}{2} \int_{\Omega} |\nabla_h v_h - \Pi_h y_h|^2 dx - \int_{\Omega} (f_h + \operatorname{div} y_h) \Pi_h v_h dx.$$

## Examples: Obstacle problem

- It holds that  $\Pi_h^{rt} z \in \text{dom}(-D_h^{rt})$ , since

$$\begin{aligned} f_h + \operatorname{div} \Pi_h^{rt} z &= \Pi_h(f + \operatorname{div} z) \leq 0 && \text{a.e. in } \Omega, \\ \Pi_h^{rt} z \cdot n &= \pi_h[z \cdot n] = 0 && \text{a.e. in } \Gamma_N. \end{aligned}$$

- It holds that  $\Pi_h^{cr} u \in \text{dom}(I_h^{cr})$ , since

$$\begin{aligned} \nabla_h \Pi_h^{cr} u &= \Pi_h \nabla u = \Pi_h z && \text{a.e. in } \Omega, \\ \pi_h \Pi_h^{cr} u &= \pi_h u = 0 && \text{a.e. in } \Gamma_D, \\ \Pi_h \Pi_h^{cr} u &\geq 0 && \text{a.e. in } \Omega. \end{aligned}$$

For  $v_h = \Pi_h^{cr} u \in \text{dom}(I_h^{cr})$  and  $y_h = \Pi_h^{rt} z \in \text{dom}(-D_h^{rt})$ , we conclude that

$$\begin{aligned} &\frac{1}{2} \int_{\Omega} |\nabla_h \Pi_h^{cr} u - \nabla_h u_h^{cr}|^2 dx - \int_{\Omega} (f_h + \operatorname{div} z_h^{rt}) \Pi_h \Pi_h^{cr} u dx \\ &\quad + \frac{1}{2} \int_{\Omega} |\Pi_h \Pi_h^{rt} z - \Pi_h z_h^{rt}|^2 dx - \int_{\Omega} (f + \operatorname{div} z) \Pi_h u_h^{cr} dx \\ &= \frac{1}{2} \int_{\Omega} |\nabla_h \Pi_h^{cr} u - \Pi_h \Pi_h^{rt} z|^2 dx - \int_{\Omega} (f_h + \operatorname{div} \Pi_h^{rt} z) \Pi_h \Pi_h^{cr} u dx \\ &= \frac{1}{2} \int_{\Omega} |\Pi_h z - \Pi_h \Pi_h^{rt} z|^2 dx + \int_{\Omega} (f + \operatorname{div} z) (u - \Pi_h \Pi_h^{cr} u) dx. \end{aligned}$$

## Examples: Obstacle problem

- On the one hand, it holds that

$$\frac{1}{2} \int_{\Omega} |\Pi_h z - \Pi_h \Pi_h^{cr} z|^2 dx \lesssim h^{2s} [z]_{s,2}^2.$$

- On the other hand, it holds that

$$\begin{aligned} \int_{\Omega} (f + \operatorname{div} z) (u - \Pi_h \Pi_h^{cr} u) dx &= \int_{\Omega} (f + \operatorname{div} z) (u - \Pi_h^{cr} u) dx \\ &\quad + \int_{\Omega} (f + \operatorname{div} z) (\Pi_h^{cr} u - \Pi_h \Pi_h^{cr} u) dx, \end{aligned}$$

where

- It holds that

$$\int_{\Omega} (f + \operatorname{div} z) (u - \Pi_h^{cr} u) dx \lesssim \left( \int_{\Omega} |u - \Pi_h^{cr} u|^2 dx \right)^{\frac{1}{2}} \lesssim h^{2s} [z]_{s,2}.$$

- Due to  $\Pi_h^{cr} u - \Pi_h \Pi_h^{cr} u = \nabla_h \Pi_h^{cr} u \cdot (\operatorname{id}_{\mathbb{R}^d} - \Pi_h \operatorname{id}_{\mathbb{R}^d})$  and  $u = 0$  on  $\{\operatorname{div} z < -f\}$ , it holds that

$$\begin{aligned} - \int_{\Omega} (f + \operatorname{div} z) (\Pi_h^{cr} u - \Pi_h \Pi_h^{cr} u) dx &= \int_{\Omega} (f + \operatorname{div} z) (\nabla_h \Pi_h^{cr} u - \nabla u) \cdot (\operatorname{id}_{\mathbb{R}^d} - \Pi_h \operatorname{id}_{\mathbb{R}^d}) dx \\ &\lesssim h \left( \int_{\Omega} |\nabla_h \Pi_h^{cr} u - \nabla u|^2 dx \right)^{\frac{1}{2}} \lesssim h^{s+1} [z]_{s,2}. \end{aligned}$$



## Examples: Obstacle problem

### ◆ Example with central singularity:

Let  $\Omega := (-1, 1)^2$ ,  $\Gamma_D = \partial\Omega$ , and  $f \in L^2(\Omega)$  s.t.  $u \in W^{1,2}(\Omega)$  is given via

$$u(x) = \begin{cases} \frac{|x|^2 - 2}{2} - \log|x| - \frac{1}{2} & \text{if } |x| > 1, \\ 0 & \text{if } |x| \leq 1, \end{cases}$$

so that  $z = \nabla u \in (W^{1,2}(\Omega))^2$ .

⇒ Expected for uniform mesh-refinement (i.e.,  $\theta = 1$ ):  $\eta_{\text{gap}, h_k}^2 (\Pi_{h_k}^{cr} u, \Pi_{h_k}^{rt} z) \sim h_k^2 \sim N_k^{-1}$ .

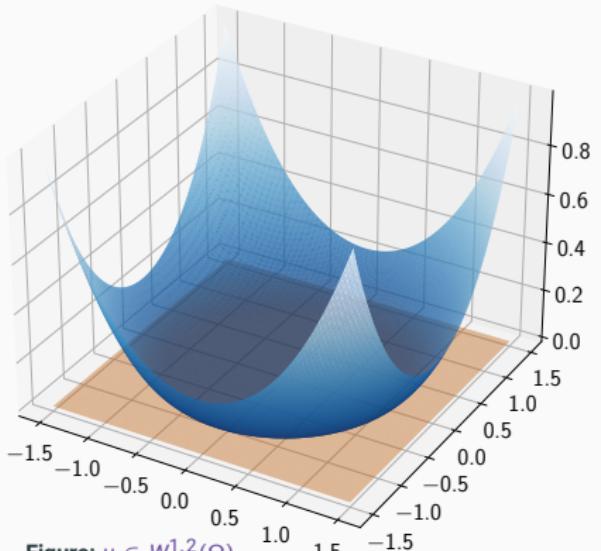


Figure:  $u \in W^{1,2}(\Omega)$ .

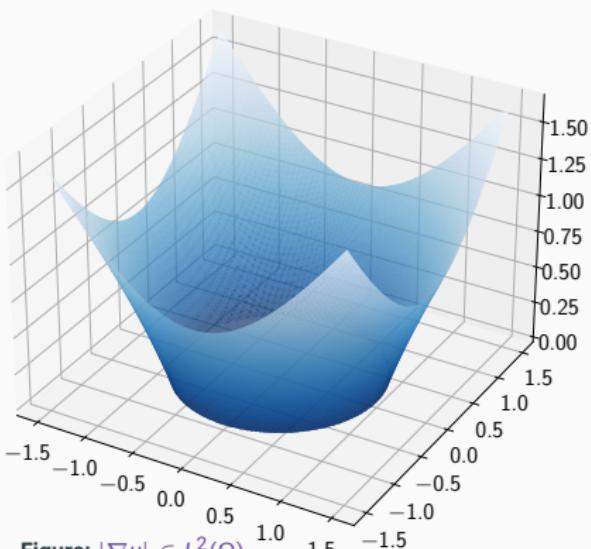
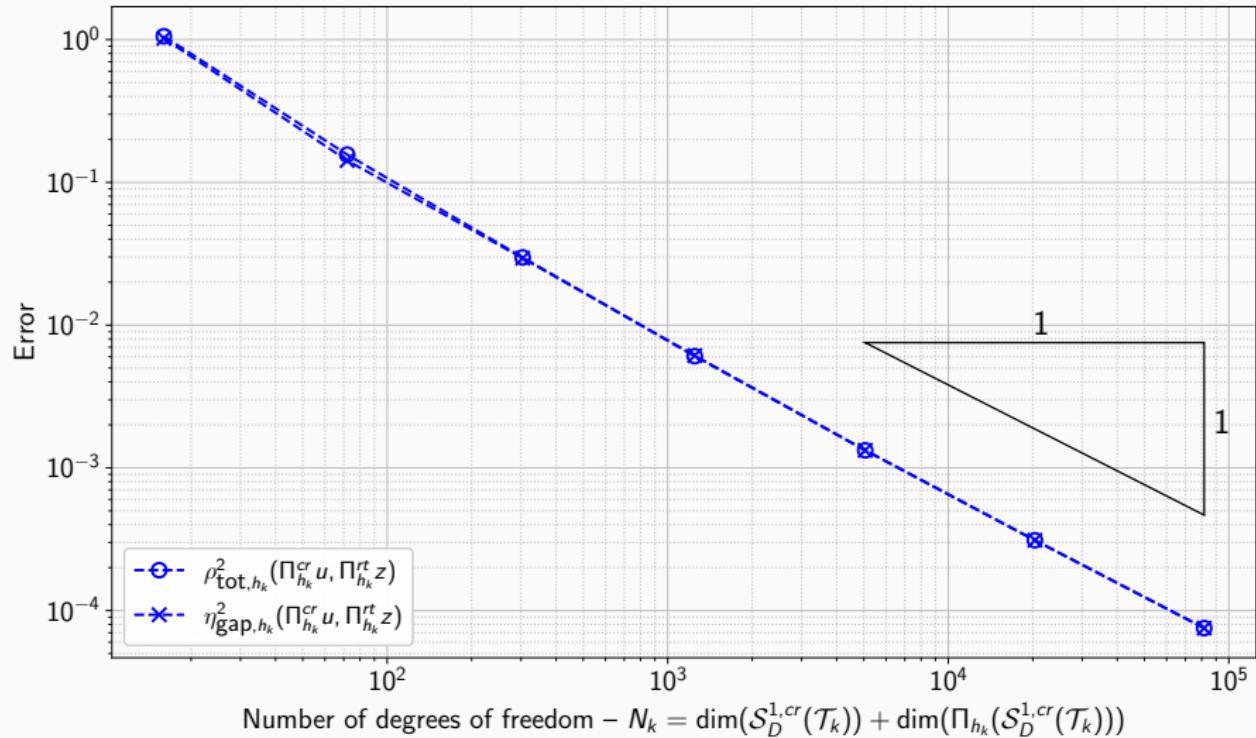


Figure:  $|\nabla u| \in L^2(\Omega)$ .

## Examples: Obstacle problem

### Experiment:

#### Results:



## ◆ A posteriori error analysis: (on the basis of convex duality)

- Primal-dual gap identity: for every  $\mathbf{v} \in \text{dom}(I)$  and  $\mathbf{y} \in \text{dom}(-D)$ , it holds that

$$\rho_{\text{tot}}^2(\mathbf{v}, \mathbf{y}) = \eta_{\text{gap}}^2(\mathbf{v}, \mathbf{y}).$$

- Advantages/Challenges:

systematic;    widely applicable;    }  
 simple;        identical;        }      vs.    ? numerically practicable.

## ◆ A posteriori error analysis: (on the basis of discrete convex duality)

- Discrete reconstruction formulas: imply numerical practicability.

$$z_h^{rt} = D_t \phi_h(\cdot, \nabla_h u_h^{cr}) + \frac{D_t \psi_h(\cdot, \Pi_h u_h^{cr})}{d} (\text{id}_{\mathbb{R}^d} - \Pi_h \text{id}_{\mathbb{R}^d}),$$

$$u_h^{cr} = D_t \psi_h^*(\cdot, \text{div} z_h^{rt}) + D_t \phi_h^*(\cdot, \Pi_h z_h^{rt}) \cdot (\text{id}_{\mathbb{R}^d} - \Pi_h \text{id}_{\mathbb{R}^d}).$$

- Discrete primal-dual gap identity: leads to a priori error identity, i.e.,

$$\rho_{\text{tot},h}^2(\Pi_h^{cr} u, \Pi_h^{rt} z) = \eta_{\text{gap},h}^2(\Pi_h^{cr} u, \Pi_h^{rt} z).$$

Thank You for today!

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