

Condition number bounds for IETI-DP methods that are explicit in h and p

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RICAM-Report 2019-31

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We study the convergence behavior of Dual-Primal Isogeometric Tearing and Interconnecting (IETI-DP) methods for solving large-scale algebraic systems arising from multi-patch Isogeometric Analysis. We focus on the Poisson problem on two dimensional computational domains. We provide a convergence analysis that covers several choices of the primal degrees of freedom: the vertex values, the edge averages, and the combination of both. We derive condition number bounds that show the expected behavior in the grid size h and that are quasi-linear in the spline degree p .

Keywords: Isogeometric Analysis; FETI-DP; p -Robustness.

AMS Subject Classification: 65N55, 65N30, 65F08, 65D07

1. Introduction

Isogeometric Analysis (IgA), see Ref. 16, is a method for solving partial differential equations (PDEs) in a way that integrates better with standard computer aided design (CAD) software than classical finite element (FEM) simulation. Both the computational domain and the solution of the PDE are represented as linear combination of tensor-product B-splines or non-uniform rational B-splines (NURBS). Since only simple domains can be represented by just one such spline function, usually the computational domain is decomposed into subdomains, in IgA usually called patches, where each patch is parameterized with its own spline function.

For IgA, like for any other discretization method for PDEs, fast iterative solvers are of interest. Domain decomposition solvers are a natural choice for multi-patch

IgA. Nowadays, Finite Element Tearing and Interconnecting (FETI) methods have become the most popular non-overlapping domain decomposition methods. After the introduction of the FETI technology by C. Farhat and F.-X. Roux in 1991¹¹, many FETI versions have been developed for different applications, see, e.g., Refs. 33, 25, 19 for a comprehensive presentation of domain decomposition methods and, in particular, of FETI methods. The most advanced and most widely used versions are certainly the Dual-Primal FETI methods (FETI-DP), which have been introduced in Ref. 10. The Balancing Domain Decomposition by Constraints (BDDC) methods introduced in Ref. 8 can be seen as the primal counterpart of the FETI-DP methods. In Ref. 21, it has been proven that the spectra of the preconditioned systems for FETI-DP and BDDC are essentially the same. Moreover, that paper gives an abstract framework for proving condition number bounds which we use in this paper.

The Ref. 18 extends FETI-DP to isogeometric discretizations. The proposed method is sometimes called Dual-Primal Isogeometric Tearing and Interconnecting (IETI-DP) method. Convergence analysis for IETI-DP methods^{13,14} and BDDC methods for IgA⁴ has been provided for numerous contexts. The given condition number bounds are explicit in grid and patch size, sometimes also in other parameters, like diffusion parameters. Convergence theory that also covers the dependence on the spline degree, is not available in the context of IgA.

Condition number estimates for FETI methods applied to spectral element discretizations have been provided in Refs. 17, 23, which show that the condition number only grows poly-logarithmic in the polynomial degree. Comparable bounds are also known for Schwarz type^{12,29} and substructuring^{2,24} methods for hp finite element discretizations. For an overview and more references, see Ref. 20. However, for hp -FETI-DP algorithms, such bounds are not known to the authors. Often the analysis for spectral methods is carried out by showing that the stiffness matrix of interest is spectrally equivalent to a stiffness matrix for a suitably constructed low-order problem. While such an analysis yields appealing upper bounds for spectral methods, a straight-forward extension of the analysis to IgA does not yield reasonable statements concerning the dependence on the spline degree. An alternative is to work directly on the function spaces of interest. One ingredient for any FETI analysis are energy bounds for the discrete harmonic extension, which can be done by explicitly constructing bounded extension operators, cf. Ref. 22 for the case of h -FEM, Refs. 5, 28 for spectral elements, and the paper on hand for spline spaces.

One of the strengths of IgA is k -refinement which allows the construction of discretizations that show the approximation power of a high-order method for the costs (in terms of the degrees of freedom) of a low order method. Certainly, an efficient realization also requires a fast solver that is (almost) robust in the spline degree p . In the last years, p -robust approximation error estimates for single-patch domains^{32,26} and multi-patch domains³¹, as well as p -robust linear solvers, like the

fast diagonalization method²⁷ or multigrid methods for single-patch domains^{9,15,7} and multi-patch domains³¹ have been proposed.

In the present paper, we provide a convergence analysis for a standard IETI-DP solver. The convergence analysis covers three cases for the choice of the primal degrees of freedom: Vertex values (Algorithm A), edge averages (Algorithm B) and a combination of both (Algorithm C). For all cases, we show that the condition number of the preconditioned system is bounded by

$$Cp \left(1 + \log p + \max_k \log \frac{H^{(k)}}{h^{(k)}} \right)^2,$$

where p is the spline degree, $H^{(k)}$ is the diameter of the k th patch, $h^{(k)}$ is the grid size on the k th patch, and the constant C only depends on the geometry function, the quasi-uniformity of the grids on the individual patches and the maximum number of patches sharing any vertex. This means that the constant C is independent of the number of patches, their diameters, the grid sizes, the spline degree and the smoothness of the splines, i.e., all choices of the smoothness between C^0 and C^{p-1} are covered.

The remainder of this paper is organized as follows. In Section 2, we introduce the model problem and discuss its isogeometric discretization. The proposed IETI-DP solver is presented in Section 3. In Section 4, we give the convergence theory. Then, we proceed with numerical experiments that illustrate our theoretical findings in Section 5. In Section 6, we conclude with some final remarks.

2. The model problem

We consider a standard Poisson problem with homogeneous Dirichlet boundary conditions as model problem. Let $\Omega \subset \mathbb{R}^2$ be an open, bounded and simply connected domain with Lipschitz boundary. For a given right-hand side $f \in L_2(\Omega)$, we want to find a function $u \in H_0^1(\Omega)$ such that

$$\underbrace{\int_{\Omega} \nabla u(x) \cdot \nabla v(x) \, dx}_{a(u, v) :=} = \underbrace{\int_{\Omega} f(x) v(x) \, dx}_{\langle f, v \rangle :=} \quad \text{for all } v \in H_0^1(\Omega). \quad (2.1)$$

Here and in what follows, we denote by $L_2(\Omega)$ and $H^s(\Omega)$, $s \in \mathbb{R}$, the usual Lebesgue and Sobolev spaces, respectively. $H_0^1(\Omega) \subset H^1(\Omega)$ is the subspace of functions vanishing on $\partial\Omega$, the boundary of Ω . These spaces are equipped with the standard scalar products $(\cdot, \cdot)_{L_2(\Omega)}$ and $(\cdot, \cdot)_{H^1(\Omega)} := (\nabla \cdot, \nabla \cdot)_{L_2(\Omega)}$, seminorms $|\cdot|_{H^s(\Omega)}$ and norms $\|\cdot\|_{L_2(\Omega)}$ and $\|\cdot\|_{H^s(\Omega)}$.

We assume that the physical domain Ω is composed of K non-overlapping

patches $\Omega^{(k)}$, i.e., we have

$$\bar{\Omega} = \bigcup_{k=1}^K \overline{\Omega^{(k)}} \quad \text{and} \quad \Omega^{(k)} \cap \Omega^{(\ell)} = \emptyset \quad \text{for all } k \neq \ell,$$

where \bar{T} denotes the closure of the set T . Each patch $\Omega^{(k)}$ is parameterized by a geometry mapping

$$G_k : \widehat{\Omega} := (0, 1)^2 \rightarrow \Omega^{(k)} := G_k(\widehat{\Omega}) \subset \mathbb{R}^2, \quad (2.2)$$

which can be continuously extended to the closure of the parameter domain $\widehat{\Omega}$. In IgA, the geometry mapping is typically represented using B-splines or NURBS. For the analysis, we only require the following assumption.

Assumption 1. There are patch sizes $H^{(k)} > 0$ for $k = 1, \dots, K$ and a constant $C_1 > 0$ such that

$$\|\nabla G_k\|_{L_\infty(\widehat{\Omega}^{(k)})} \leq C_1 H^{(k)} \quad \text{and} \quad \|(\nabla G_k)^{-1}\|_{L_\infty(\widehat{\Omega}^{(k)})} \leq C_1 \frac{1}{H^{(k)}}$$

holds for all $k = 1, \dots, K$.

The following assumption guarantees that the patches form an admissible decomposition, i.e., that there are no T-junctions.

Assumption 2. For any two patch indices $k \neq \ell$, the set $\overline{\Omega^{(k)}} \cap \overline{\Omega^{(\ell)}}$ is either a common edge (including the neighboring vertices), a common vertex, or empty.

If two patches $\Omega^{(k)}$ and $\Omega^{(\ell)}$ share a common edge, we denote that edge by $\Gamma^{(k,\ell)} = \Gamma^{(\ell,k)}$, and its pre-images by $\widehat{\Gamma}^{(k,\ell)} := G_k^{-1}(\Gamma^{(k,\ell)})$ and $\widehat{\Gamma}^{(\ell,k)} := G_\ell^{-1}(\Gamma^{(k,\ell)})$. Moreover, we define

$$\mathcal{N}_\Gamma(k) := \{\ell : \Omega^{(k)} \text{ and } \Omega^{(\ell)} \text{ share an edge}\}.$$

Analogously, if two patches $\Omega^{(k)}$ and $\Omega^{(\ell)}$ share only a vertex, we denote that vertex by $\mathbf{x}^{(k,\ell)} = \mathbf{x}^{(\ell,k)}$, and its corresponding pre-images by $\widehat{\mathbf{x}}^{(k,\ell)} := G_k^{-1}(\mathbf{x}^{(k,\ell)})$ and $\widehat{\mathbf{x}}^{(\ell,k)} := G_\ell^{-1}(\mathbf{x}^{(k,\ell)})$. Moreover, we define $\mathcal{P}(\mathbf{x}) := \{k : \mathbf{x} \in \overline{\Omega^{(k)}}\}$ and

$$\mathcal{N}_\mathbf{x}(k) := \{\ell : \Omega^{(k)} \text{ and } \Omega^{(\ell)} \text{ share a vertex or an edge}\}.$$

For the analysis, we need that the number of neighbors is bounded.

Assumption 3. There is a constant $C_2 > 0$ such that $|\mathcal{N}_\mathbf{x}(k)| \leq C_2$ for all patches $k = 1, \dots, K$.

The provided assumptions guarantee that the pre-images in the parameter domain $\widehat{\Gamma}_D^{(k)} := G_k^{-1}(\partial\Omega \cap \partial\Omega^{(k)})$ of the (Dirichlet) boundary $\Gamma_D = \partial\Omega$ consists of whole edges.

Now, we introduce the isogeometric function spaces. Let $p \in \mathbb{N} := 1, 2, 3, \dots$ be a given spline degree. For simplicity, we assume that the spline degree is uniformly throughout the overall domain. We use B-splines as discretization space. To keep the paper self-contained, we introduce the basic spline notation. The splines are defined based on a p -open knot vector:

$$\Xi = (\xi_1, \dots, \xi_{n+p+1}) = \underbrace{(\zeta_1, \dots, \zeta_1)}_{m_1}, \underbrace{(\zeta_2, \dots, \zeta_2)}_{m_2}, \dots, \underbrace{(\zeta_{N_Z}, \dots, \zeta_{N_Z})}_{m_{N_Z}},$$

where the multiplicities satisfy $m_1 = m_{N_Z} = p + 1$, and $m_i \in \{1, \dots, p\}$ for $i = 2, \dots, N_Z - 1$ and the breakpoints satisfy $\zeta_1 < \zeta_2 < \dots < \zeta_{N_Z}$. We call

$$Z = (\zeta_1, \dots, \zeta_{N_Z}) \quad \text{and} \quad M = (m_1, \dots, m_{N_Z})$$

the vector of breakpoints associated to Ξ and the vector of multiplicities associated to Ξ , respectively. We denote the standard B-spline basis as obtained by the Cox-de Boor formula, cf. (2.1) and (2.2) in Ref. 6, by $(B[p, \Xi, i])_{i=1}^n$. The corresponding spline space is given as linear span of these basis functions, i.e.,

$$S[p, \Xi] := \text{span}\{B[p, \Xi, 1], \dots, B[p, \Xi, n]\}.$$

To obtain the isogeometric function space, we choose for each patch two p -open knot vectors $\Xi^{(k,1)}$ and $\Xi^{(k,2)}$ over $(0, 1)$. On the parameter domain $\widehat{\Omega}$, we define the tensor-product spline space by $\widehat{V}^{(k)}$ and its transformation to the physical domain by $V^{(k)}$:

$$\widehat{V}^{(k)} := \{v \in S[p, \Xi^{(k,1)}] \otimes S[p, \Xi^{(k,2)}] : v|_{\widehat{\Gamma}_D^{(k)}} = 0\} \quad \text{and} \quad V^{(k)} := \widehat{V}^{(k)} \circ G_k^{-1}, \quad (2.3)$$

where $v|_T$ denotes the restriction of v to T (trace operator). We introduce a basis for the space $\widehat{V}^{(k)}$ by choosing the basis functions of the standard B-spline basis that vanish on the Dirichlet boundary $\widehat{\Gamma}_D^{(k)}$. We order the total number of $N^{(k)} = N_{\Gamma}^{(k)} + N_{\Gamma}^{(k)}$ basis functions such that the first $N_{\Gamma}^{(k)}$ are supported only in the interior of the patch and the following $N_{\Gamma}^{(k)}$ basis function contribute to the boundary of the patch:

$$\begin{aligned} \widehat{\Phi}^{(k)} &:= (\widehat{\phi}_i^{(k)})_{i=1}^{N^{(k)}}, \\ \{\widehat{\phi}_i^{(k)}\} &= \{\widehat{\phi} : \exists j_1, j_2 : \widehat{\phi}(x, y) = B[p, \Xi^{(k,1)}, j_1](x) B[p, \Xi^{(k,2)}, j_2](y) \wedge \widehat{\phi}|_{\widehat{\Gamma}_D^{(k)}} = 0\}, \\ \widehat{\phi}_i^{(k)}|_{\partial\widehat{\Omega}} = 0 &\Leftrightarrow i \in \{1, \dots, N_{\Gamma}^{(k)}\}, \quad \text{and} \quad \widehat{\phi}_i^{(k)}|_{\partial\widehat{\Omega}} \neq 0 \Leftrightarrow i \in N_{\Gamma}^{(k)} + \{1, \dots, N_{\Gamma}^{(k)}\}. \end{aligned} \quad (2.4)$$

Following the pull-back principle used for defining the function space on the physical domain, we define the basis for $V^{(k)}$ by

$$\Phi^{(k)} := (\phi_i^{(k)})_{i=1}^{N^{(k)}} \quad \text{and} \quad \phi_i^{(k)} := \widehat{\phi}_i^{(k)} \circ G_k^{-1}.$$

We assume that the underlying grids on each of the patches are quasi-uniform.

Assumption 4. There are grid sizes $\widehat{h}^{(k)} > 0$ for $k = 1, \dots, K$ and a constant $C_3 > 0$ such that

$$C_3 \widehat{h}^{(k)} \leq \zeta_{i+1}^{(k,\delta)} - \zeta_i^{(k,\delta)} \leq \widehat{h}^{(k)}$$

holds for all $i = 1, \dots, N_Z^{(k,\delta)} - 1$ and all $\delta = 1, 2$.

The grid size on the physical domain is defined via $h^{(k)} := \widehat{h}^{(k)} H^{(k)}$.

To be able to set up a H^1 -conforming discretization on the whole domain Ω , we assume that the function spaces are fully matching.

Assumption 5. For any two patches $\Omega^{(k)}$ and $\Omega^{(\ell)}$ sharing a common edge $\Gamma^{(k,\ell)}$, the following statement holds true. For any basis function $\phi_i^{(k)}$ that does not vanish on $\Gamma^{(k,\ell)}$, there is exactly one basis function $\phi_j^{(\ell)}$ such that they agree on $\Gamma^{(k,\ell)}$, i.e.,

$$\phi_i^{(k)}|_{\Gamma^{(k,\ell)}} \neq 0 \quad \Rightarrow \quad \exists j : \phi_i^{(k)}|_{\Gamma^{(k,\ell)}} = \phi_j^{(\ell)}|_{\Gamma^{(k,\ell)}}. \quad (2.5)$$

This assumption is satisfied if the the spline degrees p , the knot vectors, and the geometry mappings agree on all interfaces. Based on this assumption, we define the overall function space as

$$V := \{u \in H_0^1(\Omega) : u|_{\Omega^{(k)}} \in V^{(k)} \text{ for } k = 1, \dots, K\}. \quad (2.6)$$

Using these function spaces, we obtain the Galerkin discretization of the variational problem (2.1), which reads as follows. Find $u \in V$ such that

$$a(u, v) = \langle f, v \rangle \quad \text{for all } v \in V, \quad (2.7)$$

where $a(\cdot, \cdot)$ and $\langle f, \cdot \rangle$ are as defined in (2.1). By choosing a basis for V , the variational problem (2.7) can be written in a matrix-vector form. For the construction of a IETI-DP method, we omit this step and directly work with the discrete variational problem (2.7).

3. The IETI-DP solver

In this section, we derive a IETI-DP solver for the discretized variational problem (2.7). First, we observe that the bilinear form $a(\cdot, \cdot)$ and the linear form $\langle f, \cdot \rangle$ are the sum of contributions of each of the patches, i.e.,

$$a(u, v) = \sum_{k=1}^K a^{(k)}(u, v), \quad \text{where } a^{(k)}(u, v) := \int_{\Omega^{(k)}} \nabla u(x) \cdot \nabla v(x) \, dx$$

and

$$\langle f, v \rangle = \sum_{k=1}^K \langle f^{(k)}, v \rangle, \quad \text{where } \langle f^{(k)}, v \rangle := \int_{\Omega^{(k)}} f(x) v(x) \, dx.$$

By discretizing $a^{(k)}(\cdot, \cdot)$ and $\langle f^{(k)}, \cdot \rangle$ using the basis $\Phi^{(k)}$, we obtain the matrix-vector system

$$A^{(k)} \underline{u}^{(k)} = \underline{f}^{(k)}, \quad (3.1)$$

where $A^{(k)} = [a^{(k)}(\phi_j^{(k)}, \phi_i^{(k)})]_{i,j=1}^{N^{(k)}}$ is a stiffness matrix and $\underline{f}^{(k)} = [\langle f^{(k)}, \phi_i^{(k)} \rangle]_{i=1}^{N^{(k)}}$ is a load vector.

Since we are interested in the solution of the original problem (2.7), we need to enforce continuity. Doing this is an operation on the interfaces only. Note that the basis $\Phi^{(k)}$ has been defined such that the first $N_{\Gamma}^{(k)}$ basis functions are supported only in the interior of the patch and the following $N_{\Gamma}^{(k)}$ basis functions do contribute to the boundary of the patch, see (2.4). Following this decomposition, we obtain

$$A^{(k)} = \begin{pmatrix} A_{\Pi}^{(k)} & A_{\Pi\Gamma}^{(k)} \\ A_{\Gamma\Gamma}^{(k)} & A_{\Gamma\Gamma}^{(k)} \end{pmatrix} \quad \text{and} \quad \underline{f}^{(k)} = \begin{pmatrix} \underline{f}_{\Pi}^{(k)} \\ \underline{f}_{\Gamma}^{(k)} \end{pmatrix}.$$

Using this decomposition, we rewrite the linear systems (3.1) as Schur complement:

$$S^{(k)} \underline{w}^{(k)} = \underline{g}^{(k)}, \quad (3.2)$$

where

$$S^{(k)} := A_{\Gamma\Gamma}^{(k)} - A_{\Gamma\Gamma}^{(k)} (A_{\Pi\Pi}^{(k)})^{-1} A_{\Pi\Gamma}^{(k)} \quad \text{and} \quad \underline{g}^{(k)} := \underline{f}_{\Gamma}^{(k)} - A_{\Gamma\Gamma}^{(k)} (A_{\Pi\Pi}^{(k)})^{-1} \underline{f}_{\Pi}^{(k)}. \quad (3.3)$$

The overall solution is then obtain by

$$\underline{u}^{(k)} = \begin{pmatrix} \underline{u}_{\Pi}^{(k)} \\ \underline{u}_{\Gamma}^{(k)} \end{pmatrix} = \begin{pmatrix} (A_{\Pi\Pi}^{(k)})^{-1} A_{\Pi\Gamma}^{(k)} \underline{w}^{(k)} \\ \underline{w}^{(k)} \end{pmatrix}. \quad (3.4)$$

By combining the linear systems (3.2) for all patches, we obtain the linear system

$$S \underline{w} = \underline{g}, \quad (3.5)$$

where

$$S := \begin{pmatrix} S^{(1)} & & \\ & \ddots & \\ & & S^{(K)} \end{pmatrix}, \quad \underline{w} := \begin{pmatrix} \underline{w}^{(1)} \\ \vdots \\ \underline{w}^{(K)} \end{pmatrix} \quad \text{and} \quad \underline{g} := \begin{pmatrix} \underline{g}^{(1)} \\ \vdots \\ \underline{g}^{(K)} \end{pmatrix}.$$

As a next step, we give an interpretation of the linear system in a variational sense. Let

$$W^{(k)} := \{v|_{\partial\Omega^{(k)}} : v \in V^{(k)}\} \quad \text{and} \quad W := \prod_{k=1}^K W^{(k)}$$

be the function spaces on the skeleton. Let $\mathcal{H}_h^{(k)} : W^{(k)} \rightarrow V^{(k)}$ be the discrete harmonic extension, i.e., such that

$$(\mathcal{H}_h^{(k)} w^{(k)})|_{\partial\Omega^{(k)}} = w^{(k)}, \quad \text{and} \quad a(\mathcal{H}_h^{(k)} w^{(k)}, v^{(k)}) = 0 \quad \text{for all} \quad v^{(k)} \in V_0^{(k)},$$

where $V_0^{(k)} := \{v \in V^{(k)} : v|_{\partial\Omega^{(k)}} = 0\}$. The linear system (3.5) can be rewritten as follows. Find $w = (w^{(1)}, \dots, w^{(K)}) \in W$ such that

$$\underbrace{\sum_{k=1}^K a^{(k)}(\mathcal{H}_h^{(k)} w^{(k)}, \mathcal{H}_h^{(k)} q^{(k)})}_{s(w, q) :=} = \underbrace{\sum_{k=1}^K \langle f, \mathcal{H}_h^{(k)} q^{(k)} \rangle}_{\langle g, q \rangle :=} \quad \text{for all } q = (q^{(1)}, \dots, q^{(K)}) \in W.$$

The next step is to enforce continuity between the patches. So, for any two patches $\Omega^{(k)}$ and $\Omega^{(\ell)}$ that share an edge $\Gamma^{(k, \ell)}$ and all functions $\phi_i^{(k)}$ and $\phi_j^{(\ell)}$ such that

$$\phi_i^{(k)}|_{\Gamma^{(k, \ell)}} = \phi_j^{(\ell)}|_{\Gamma^{(k, \ell)}} \neq 0,$$

we introduce a constraint of the form

$$w_{i^*}^{(k)} - w_{j^*}^{(\ell)} = 0, \quad (3.6)$$

where $i^* = i - N_{\Gamma}^{(k)}$ and $j^* = j - N_{\Gamma}^{(\ell)}$ are the corresponding indices. If we choose the vertex values as primal degrees of freedom (see also Algorithms A and C below), we do not introduce such constraints for the functions associated to the vertices, see Figure 1. If only the edge averages are chosen as primal degrees of freedom (see also Algorithm B below), we additionally introduce constraints of the form (3.6) for the corresponding degrees of freedom for the patches that share only a vertex. This means that we represent the vertex values in a fully redundant way, see Figure 2.

Say, the number of constraints of the form (3.6) is N_{Γ} . Then, we define a matrix $B \in \mathbb{R}^{N_{\Gamma} \times N}$ such that each of the constraints (3.6) constitutes one row of the linear system

$$B \underline{w} = 0.$$

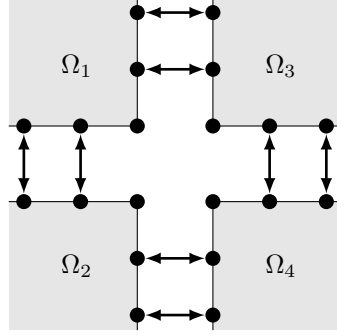
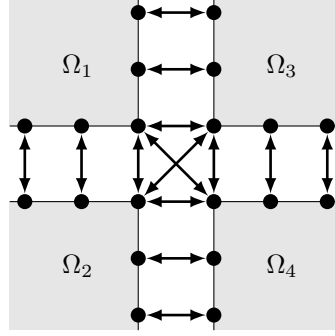
This means that each row of the matrix B has exactly two non-vanishing entries: one with value 1 and one with value -1 . We decompose the matrix B into a collection of patch-local matrices $B^{(1)}, \dots, B^{(K)}$ such that $B = (B^{(1)} \dots B^{(K)})$.

The original variational problem (2.7) is equivalent to the following problem. Find $(\underline{w}, \underline{\lambda})$ such that

$$\begin{pmatrix} S & B^{\top} \\ B & \end{pmatrix} \begin{pmatrix} \underline{w} \\ \underline{\lambda} \end{pmatrix} = \begin{pmatrix} \underline{g} \\ 0 \end{pmatrix}.$$

For patches $\Omega^{(k)}$ that do not contribute to the Dirichlet boundary of the physical domain Ω , the corresponding matrices $A^{(k)}$ and $S^{(k)}$ refer to Poisson problems with pure Neumann boundary conditions, which means that these matrices are singular. So, in general, the matrix S is singular as well.

To overcome this problem, primal degrees of freedom are introduced. Here, we have several possibilities:


 Figure 1: Omitting vertices
(Algorithms A and C)

 Figure 2: Fully redundant
(Algorithm B)

- *Algorithm A (Vertex values)*: The space \widetilde{W} is the subspace of functions where the vertex values agree, i.e.,

$$\widetilde{W} := \left\{ w \in W : \begin{array}{l} w^{(k)}(\mathbf{x}) = w^{(\ell)}(\mathbf{x}) \\ \text{for all common vertices } \mathbf{x} \text{ of all pairs of } \Omega^{(k)} \text{ and } \Omega^{(\ell)} \end{array} \right\}.$$

The subspace $\widetilde{W}_\Delta \subset \widetilde{W}$ satisfies these conditions homogeneously, i.e., $\widetilde{W}_\Delta := \prod_{k=1}^K \widetilde{W}_\Delta^{(k)}$ and

$$\widetilde{W}_\Delta^{(k)} := \{w \in W^{(k)} : w(\mathbf{x}) = 0 \text{ for all vertices } \mathbf{x} \text{ of all } \Omega^{(k)}\}.$$

- *Algorithm B (Edge averages)*: The space \widetilde{W} is the subspace of functions where the averages of the function values over the edges agree, i.e.,

$$\widetilde{W} := \left\{ w \in W : \int_{\Gamma^{(k,\ell)}} w^{(k)}(x) dx = \int_{\Gamma^{(k,\ell)}} w^{(\ell)}(x) dx \text{ for all edges } \Gamma^{(k,\ell)} \right\}.$$

The B-spline space \widetilde{W}_Δ satisfies these conditions homogeneously, i.e., $\widetilde{W}_\Delta := \prod_{k=1}^K \widetilde{W}_\Delta^{(k)}$ and

$$\widetilde{W}_\Delta^{(k)} := \left\{ w \in W^{(k)} : \int_{\Gamma^{(k,\ell)}} w^{(k)}(x) dx = 0 \text{ for all edges } \Gamma^{(k,\ell)} \right\}.$$

- *Algorithm C (Vertex values and edge averages)*: We combine the constraints from both cases. So, the spaces \widetilde{W} and $\widetilde{W}_\Delta^{(k)}$ and \widetilde{W}_Δ are the intersections of the corresponding spaces obtained by Algorithms A and B.

We introduce matrices $C^{(k)}$ representing the subspace $\widetilde{W}_\Delta^{(k)}$, i.e., we have

$$C^{(k)} \underline{w}^{(k)} = 0 \quad \Leftrightarrow \quad w^{(k)} \in \widetilde{W}_\Delta^{(k)}$$

for all $w^{(k)} \in W^{(k)}$ with vector representation $\underline{w}^{(k)}$. The matrix $C^{(k)}$ is chosen to have full rank, i.e., the number of rows coincides with the number of primal degrees

of freedom per patch. The matrix C is a block-diagonal matrix containing the blocks $C^{(1)}, \dots, C^{(K)}$.

For every choice of primal degrees of freedom, the space \widetilde{W}_Π is the S -orthogonal complement of \widetilde{W}_Δ in \widetilde{W} , i.e.,

$$\widetilde{W}_\Pi := \{w \in \widetilde{W} : s(w, q) = 0 \text{ for all } q \in \widetilde{W}_\Delta\}. \quad (3.7)$$

Let $\psi^{(1)}, \dots, \psi^{(N_\Pi)}$ be a basis of \widetilde{W}_Π . For the computation, one usually chooses a nodal basis, where the vertex values and/or the edge averages form the nodal values. The matrix Ψ represents the basis in terms of the basis for the space W , i.e.,

$$\Psi = \begin{pmatrix} \underline{\psi}^{(1)} \\ \vdots \\ \underline{\psi}^{(N_\Pi)} \end{pmatrix}. \quad (3.8)$$

Following Ref. 21, the following linear system is an equivalent rewriting of the original variational problem (2.7): Find $(\underline{w}_\Delta, \underline{\mu}, \underline{w}_\Pi, \underline{\lambda})$ such that

$$\begin{pmatrix} S & C^\top & & B^\top \\ C & & & \\ & \Psi^\top S \Psi & \Psi^\top B^\top & \\ B & B \Psi & & \end{pmatrix} \begin{pmatrix} \underline{w}_\Delta \\ \underline{\mu} \\ \underline{w}_\Pi \\ \underline{\lambda} \end{pmatrix} = \begin{pmatrix} \underline{g} \\ 0 \\ \Psi^\top \underline{g} \\ 0 \end{pmatrix},$$

where the solution for the original problem is obtained by $\underline{w} = \underline{w}_\Delta + \Psi \underline{w}_\Pi$. By reordering and by building the Schur complement, we obtain the following equivalent formulation

$$F \underline{\lambda} = \underline{d}, \quad (3.9)$$

where

$$F := \underbrace{(B \ 0 \ B \Psi) \begin{pmatrix} S & C^\top \\ C & \Psi^\top S \Psi \end{pmatrix}^{-1} \begin{pmatrix} I \\ 0 \\ \Psi^\top \end{pmatrix}}_{F_0} B^\top \quad \text{and} \quad \underline{d} := F_0 \underline{g}. \quad (3.10)$$

The idea of the IETI-DP method is to solve the linear system (3.9) with a preconditioned conjugate gradient (PCG) solver. As preconditioner, we use the scaled Dirichlet preconditioner M_{sD} , which is given by

$$M_{\text{sD}} := B D^{-1} S D^{-1} B^\top,$$

where $D \in \mathbb{R}^{N_\Gamma \times N_\Gamma}$ is a diagonal matrix defined based on the principle of multiplicity scaling. This means that each coefficient $d_{i,i}$ of D is assigned the number of Lagrange multipliers that act on the corresponding basis function, but at least 1, i.e.,

$$d_{i,i} := \max \left\{ 1, \sum_{j=1}^{N_\Gamma} b_{i,j}^2 \right\},$$

where $b_{i,j}$ are the coefficients of the matrix B .

For the realization of the proposed method, one has to perform the following steps.

- Compute the vectors $\underline{g}^{(k)}$ according to (3.3).
- Compute the matrix $\underline{\Psi}$ according to (3.7) and (3.8).
- Compute $S_{\Pi} := \Psi^{\top} S \Psi$
- Execute a preconditioned conjugate gradient (PCG) solver for computing $\underline{\lambda}$. This requires the computation of the residual and the application of the preconditioner. For the computation of the residual $\underline{\hat{w}} := F\underline{\lambda} - \underline{d}$, the following steps are applied:
 - Compute $\underline{\hat{q}} = ((\hat{q}^{(1)})^{\top} \dots (\hat{q}^{(K)})^{\top})^{\top} := B^{\top} \underline{\lambda} - \underline{g}$.
 - Solve the linear system

$$\begin{pmatrix} S^{(k)} & (C^{(k)})^{\top} \\ C^{(k)} & \end{pmatrix} \begin{pmatrix} \underline{\hat{w}}_{\Delta}^{(k)} \\ \underline{\hat{\mu}}^{(k)} \end{pmatrix} = \begin{pmatrix} \underline{\hat{q}}^{(k)} \\ 0 \end{pmatrix} \quad (3.11)$$

for all $k = 1, \dots, K$. If the vertex values are chosen as primal degrees of freedom (Algorithms A and C), it is possible to get an equivalent formulation by eliminating the degrees of freedom corresponding to the vertex values and the corresponding Lagrange multipliers.

- Solve the linear system

$$S_{\Pi} \underline{\hat{w}}_{\Pi} = \Psi^{\top} \underline{\hat{q}}, \quad (3.12)$$

which is usually expected to be small.

- The residual is given by

$$\underline{\hat{w}} := B \begin{pmatrix} \underline{\hat{w}}_{\Delta}^{(1)} \\ \vdots \\ \underline{\hat{w}}_{\Delta}^{(K)} \end{pmatrix} + B \Psi \underline{\hat{w}}_{\Pi}. \quad (3.13)$$

The computation of the preconditioned residual $\underline{\hat{p}} := M_{\text{sD}} \underline{\hat{w}}$ only requires matrix-vector multiplications.

- To obtain the solution vector \underline{u} , first \underline{w} is computed analogously to (3.11), (3.12) and (3.13) based on $\underline{q} = ((q^{(1)})^{\top} \dots (q^{(K)})^{\top})^{\top} := B^{\top} \underline{\lambda}$. Then, the solution vector \underline{u} is obtained by (3.4).

4. Condition number estimate

In this section, we prove the following condition number estimate.

Theorem 4.1. *Provided that the IETI-DP solver is set up as outlined in the pre-*

vious sections, the condition number of the preconditioned system satisfies

$$\kappa(M_{\text{sD}}F) \leq Cp \left(1 + \log p + \max_{k=1, \dots, K} \log \frac{H^{(k)}}{h^{(k)}} \right)^2,$$

where C only depends on the constants from the Assumptions 1, 3, and 4.

Note that, for a fixed choice of the spline degree p , this estimate behaves essentially the same as one would expect for standard low-order finite elements.

Notation 1. We use the notation $a \lesssim b$ if there is a constant $c > 0$ that only depends on the constants from the Assumptions 1, 3, and 4 such that $a \leq cb$. Moreover, we write $a \approx b$ if $a \lesssim b \lesssim a$.

Following the standard path, we develop an analysis in the $H^{1/2}$ -seminorm on the skeleton. For this, we need to know that the H^1 -seminorm of the discrete harmonic extension is bounded by the $H^{1/2}$ -seminorm on the boundary. Since such a result has not been worked out for spline spaces, we give an estimate in Section 4.1. In Section 4.2, we give a p -robust embedding statement and in Section 4.3, we give a lemma that allows to tear the $H^{1/2}$ -seminorm apart. In Section 4.4, we use these results to give a condition number bound.

4.1. Estimates for the discrete harmonic extension

The analysis of this subsection follows Ref. 22, where the same technique has been used to estimate the discrete harmonic extension in the context of Finite Element methods. In this subsection, we only consider one single patch at a time, i.e., k is fixed. Let

$$\mathcal{V} := S[p, \Xi^{(k,1)}] \otimes S[p, \Xi^{(k,2)}] \quad \text{and} \quad \mathcal{W} := \{v|_{\partial\widehat{\Omega}} : v \in \mathcal{V}\}$$

be the corresponding function space and its restriction to $\partial\widehat{\Omega}$, respectively. Note that $\widehat{V}^{(k)} = \{v \in \mathcal{V} : v|_{\widehat{\Gamma}^{(k)}} = 0\}$.

Let $Z^{(k,\delta)} = (\zeta_1^{(k,\delta)}, \dots, \zeta_{N_Z^{(k,\delta)}}^{(k,\delta)})$ be the vectors of breakpoints associated to the knot vectors $\Xi^{(k,\delta)}$, for $\delta = 1, 2$, respectively. Let \widehat{h} be the corresponding grid size, using Assumption 4, we have

$$\widehat{h} \lesssim \zeta_{i+1}^{(k,\delta)} - \zeta_i^{(k,\delta)} \leq \widehat{h}$$

for all $i = 1, \dots, N_Z^{(k,\delta)} - 1$ and all $\delta = 1, 2$. As a next step, we introduce a hierarchy of nested grids

$$Z^{(k,\delta,0)} := (0, 1), \quad Z^{(k,\delta,1)}, \quad \dots, \quad Z^{(k,\delta,L)} := Z^{(k,\delta)}$$

with $Z^{(k,\delta,\ell)} = (\zeta_1^{(k,\delta,\ell)}, \dots, \zeta_{N_Z^{(k,\delta,\ell)}}^{(k,\delta,\ell)})$ such that $\zeta_1^{(k,\delta,\ell)} = 0$, $\zeta_{N_Z^{(k,\delta,\ell)}}^{(k,\delta,\ell)} = 1$ and

$$\widehat{h}_\ell \lesssim \zeta_{i+1}^{(k,\delta,\ell)} - \zeta_i^{(k,\delta,\ell)} \leq \widehat{h}_\ell, \quad \text{where} \quad \widehat{h}_\ell := 4^{L-\ell} \widehat{h} \quad (4.1)$$

for all $i = 1, \dots, N_Z^{(k, \delta, \ell)} - 1$, all $\ell = 1, \dots, L$ and all $\delta = 1, 2$. The following Lemma guarantees the existence of such grids.

Lemma 4.1. *Let $Z := (\zeta_1, \dots, \zeta_{N_Z})$ with $\zeta_1 = 0$ and $\zeta_{N_Z} = 1$ be a given vector of breakpoints with grid size $h := \max_{i=1, \dots, N_Z-1} \zeta_{i+1} - \zeta_i$. For all $\tilde{h} \geq 3h$, there is a vector of breakpoints $\tilde{Z} := (\tilde{\zeta}_0, \dots, \tilde{\zeta}_{N_{\tilde{Z}}})$ such that*

$$\min\{1, \frac{1}{3}\tilde{h}\} \leq \tilde{\zeta}_{i+1} - \tilde{\zeta}_i \leq \tilde{h} \quad \text{for all } i = 1, \dots, N_{\tilde{Z}} - 1. \quad (4.2)$$

Proof. Let $\tilde{h} \geq 3h$ be arbitrary but fixed. Let $\hat{\zeta}_0 := 0$ and

$$\hat{\zeta}_i := \max\{\zeta_j : \zeta_j \leq \hat{\zeta}_{i-1} + \frac{2}{3}\tilde{h}\} \quad \text{for } i = 1, 2, \dots \quad (4.3)$$

Let $N_{\tilde{Z}}$ be the smallest index such that $\hat{\zeta}_{N_{\tilde{Z}}} = 1$. Define

$$\tilde{Z} := (\tilde{\zeta}_1, \dots, \tilde{\zeta}_{N_{\tilde{Z}}}) := \begin{cases} (\hat{\zeta}_1, \dots, \hat{\zeta}_{N_{\tilde{Z}}}) & \text{if } \hat{\zeta}_{N_{\tilde{Z}}} - \hat{\zeta}_{N_{\tilde{Z}}-1} \geq \min\{1, \frac{1}{3}\tilde{h}\} \\ (\hat{\zeta}_1, \dots, \hat{\zeta}_{N_{\tilde{Z}}-2}, \hat{\zeta}_{N_{\tilde{Z}}}) & \text{otherwise.} \end{cases} \quad (4.4)$$

Here, we take the minimum of 1 and $\frac{1}{3}\tilde{h}$ in the first case to make this definition to be correct also for the case $N_{\tilde{Z}} = 1$. From (4.3), we immediately obtain that $\hat{\zeta}_{i+1} - \hat{\zeta}_i \leq \frac{2}{3}\tilde{h}$ for all $i = 1, \dots, N_{\tilde{Z}}$. From this and the construction in (4.4), we obtain the upper bound in (4.2).

Observe that $\hat{\zeta}_{i+1} - \hat{\zeta}_i \geq \frac{2}{3}\tilde{h} - h \geq \frac{1}{3}\tilde{h}$ for all $i = 1, \dots, N_{\tilde{Z}} - 1$. This follows from the definition in (4.3) in combination with the fact that the grid size of the original grid is h . This immediately implies $\tilde{\zeta}_{i+1} - \tilde{\zeta}_i \geq \frac{1}{3}\tilde{h}$ for all $i = 1, \dots, N_{\tilde{Z}} - 1$. If $N_{\tilde{Z}} = N_Z - 1$, this finishes the proof. If $N_{\tilde{Z}} = N_Z$, we have by assumption that $\tilde{\zeta}_{N_{\tilde{Z}}} - \tilde{\zeta}_{N_{\tilde{Z}}-1} \geq \min\{1, \frac{1}{3}\tilde{h}\}$, which finishes the proof also in this case. \square

For each of this vectors of breakpoints, we introduce corresponding p -open knot vectors without repeated inner knots:

$$\Xi^{(k, \delta, \ell)} := \underbrace{(\zeta_1^{(k, \delta, \ell)}, \dots, \zeta_1^{(k, \delta, \ell)})}_{p+1 \text{ times}}, \zeta_2^{(k, \delta, \ell)}, \dots, \zeta_{N_Z^{(k, \delta, \ell)}-1}^{(k, \delta, \ell)}, \underbrace{(\zeta_{N_Z^{(k, \delta, \ell)}}^{(k, \delta, \ell)}, \dots, \zeta_{N_Z^{(k, \delta, \ell)}}^{(k, \delta, \ell)})}_{p+1 \text{ times}}.$$

Based on these grids, we introduce coarse-grid spline spaces of maximum smoothness by

$$\mathcal{W}^{(\ell)} := \left\{ w \in (S[p, \Xi^{(k, 1, \ell)}] \otimes S[p, \Xi^{(k, 2, \ell)}])|_{\partial\hat{\Omega}} : w \circ \gamma \in C^{p-1}(-\infty, \infty) \right\}$$

where $\gamma : (-\infty, \infty) \rightarrow \partial\hat{\Omega}$ is given by

$$\gamma(t) := \begin{cases} (0, t) & \text{if } t \in [0, 1) + 4\mathbb{Z} \\ (t-1, 1) & \text{if } t \in [1, 2) + 4\mathbb{Z} \\ (1, 3-t) & \text{if } t \in [2, 3) + 4\mathbb{Z} \\ (4-t, 0) & \text{if } t \in [3, 4) + 4\mathbb{Z} \end{cases}, \quad (4.5)$$

where \mathbb{Z} is the set of integers. Note that the condition $w \circ \gamma \in C^{p-1}(-\infty, \infty)$ introduces special smoothness requirements on the vertices of $\widehat{\Omega}$. For convenience, we define $\mathcal{W}^{(0)} := \{0\}$ and $\mathcal{W}^{(L+1)} := \mathcal{W}$ and observe that the spaces $\mathcal{W}^{(\ell)}$ are nested, i.e.,

$$\mathcal{W}^{(0)} \subset \mathcal{W}^{(1)} \subset \dots \subset \mathcal{W}^{(L)} \subset \mathcal{W}^{(L+1)}.$$

Since the spaces $\mathcal{W}^{(\ell)}$ for all $\ell = 1, \dots, L$ are periodic and of maximum smoothness, we have a robust inverse estimate. The space $\mathcal{W}^{(L+1)}$ is not a spline space of maximum smoothness, thus only a standard inverse estimate for piecewise polynomial functions can be used.

Lemma 4.2. *The following $H^1 - L_2$ and $H^2 - H^1$ -inverse estimates hold:*

- $|w|_{H^1(\partial\widehat{\Omega})}^2 \lesssim p^4 \widehat{h}_{L+1}^{-2} \|w\|_{L_2(\partial\widehat{\Omega})}^2$ for $w \in \mathcal{W}^{(L+1)}$.
- $|w|_{H^1(\partial\widehat{\Omega})}^2 \lesssim \widehat{h}_\ell^{-2} \|w\|_{L_2(\partial\widehat{\Omega})}^2$ for $w \in \mathcal{W}^{(\ell)}$ with $\ell = 1, \dots, L$.

Proof. This follows immediately from definition (4.5), the observation that $|w|_{H^1(\partial\widehat{\Omega})} = |w \circ \gamma|_{H^1(0,4)}$ and $\|w\|_{L_2(\partial\widehat{\Omega})} = \|w \circ \gamma\|_{L_2(0,4)}$, the inverse inequalities given in Theorem 4.76, eq. (4.6.5) in Ref. 30 (for $\ell = L + 1$) and Theorem 6.1 in Ref. 32 (for $\ell = 1, \dots, L$) and (4.1). \square

The following Lemma shows that there is also an $H^1 - H^{1/2}$ -inverse estimate.

Lemma 4.3. *The $H^1 - H^{1/2}$ -inverse estimate $|w|_{H^1(\partial\widehat{\Omega})}^2 \lesssim p^2 \widehat{h}_{L+1}^{-1} |w|_{H^{1/2}(\partial\widehat{\Omega})}^2$ holds for all $w \in \mathcal{W}^{(L+1)}$.*

Proof. Using the reiteration theorem and the fact that the fractional order Sobolev spaces coincide with the corresponding interpolation spaces, cf. Theorems 7.21 and 7.31 in Ref. 1, we obtain that H^1 is the interpolation between the Sobolev spaces $H^{1/2}$ and H^2 , i.e.,

$$\|w\|_{H^1(\partial\widehat{\Omega})} \approx \|w\|_{[H^{1/2}(\partial\widehat{\Omega}), H^2(\partial\widehat{\Omega})]_{1/3}}.$$

Thus, we obtain

$$\|w\|_{H^1(\partial\widehat{\Omega})}^3 \lesssim \|w\|_{H^{1/2}(\partial\widehat{\Omega})}^2 \|w\|_{H^2(\partial\widehat{\Omega})}.$$

Since the derivative of a spline is again a spline, we obtain analogously to the proof of Lemma 4.2 that $\|w\|_{H^2(\partial\widehat{\Omega})} \lesssim p^2 \widehat{h}_{L+1}^{-1} \|w\|_{H^1(\partial\widehat{\Omega})}$. This yields the estimate $\|w\|_{H^1(\partial\widehat{\Omega})}^2 \lesssim p^2 \widehat{h}_{L+1}^{-1} \|w\|_{H^{1/2}(\partial\widehat{\Omega})}^2$. The Poincaré inequality finishes the proof.

□

Let Q_ℓ be the L_2 -orthogonal projector into $\mathcal{W}^{(\ell)}$ and let Π_ℓ be the H^1 -orthogonal projector into $\mathcal{W}^{(\ell)}$, minimizing the distance in the norms $\|\cdot\|_{L_2(\partial\widehat{\Omega})}^2$ and $\|\cdot\|_{H^1(\partial\widehat{\Omega})}^2$, respectively. For these projectors, the following error estimates are satisfied.

Lemma 4.4. *The estimate $\|(I - Q_\ell)w\|_{L_2(\partial\widehat{\Omega})}^2 \lesssim \widehat{h}_\ell \|w\|_{H^{1/2}(\partial\widehat{\Omega})}^2$ holds for all functions $w \in H^1(\partial\widehat{\Omega})$ and all $\ell = 0, \dots, L + 1$.*

Proof. For $\ell = 0$, we have $Q_\ell = 0$ and $\widehat{h}_\ell = 1$. Thus, the statement is trivial. Now, assume that $\ell \geq 1$. Let $w \in H^1(\partial\widehat{\Omega})$ be arbitrary but fixed and let $v := w \circ \gamma$. Observe that

$$\|(I - Q_\ell)w\|_{L_2(\partial\widehat{\Omega})}^2 = \inf_{w_\ell \in \mathcal{W}^{(\ell)}} \|w - w_\ell\|_{L_2(\partial\widehat{\Omega})}^2 = \inf_{v_\ell \in \mathcal{V}^{(\ell)}} \|v - v_\ell\|_{L_2(0,4)}^2,$$

where γ is as in (4.5). The Theorem 4.1 in Ref. 26 and (4.1) yield

$$\|(I - Q_\ell)w\|_{L_2(\partial\widehat{\Omega})}^2 \lesssim \widehat{h}_\ell^2 |v|_{H^1(0,4)}^2.$$

For the choice $v_\ell := 0$, we have

$$\|(I - Q_\ell)w\|_{L_2(\partial\widehat{\Omega})}^2 \lesssim \|v\|_{L_2(0,4)}^2.$$

Using Hilbert space interpolation theory, cf. Theorems 7.23 and 7.31 in Ref. 1, we obtain

$$\|(I - Q_\ell)w\|_{L_2(\partial\widehat{\Omega})}^2 \lesssim \widehat{h}_\ell \|v\|_{H^{1/2}(0,4)}^2.$$

Using $\|v\|_{L_2(0,4)}^2 = \|w\|_{L_2(\partial\widehat{\Omega})}^2$ and

$$|v|_{H^{1/2}(0,4)}^2 = \int_0^4 \int_0^4 \frac{|v(s) - v(t)|^2}{|s - t|^2} ds dt \lesssim \int_0^4 \int_0^4 \frac{|v(s) - v(t)|^2}{|\gamma(s) - \gamma(t)|^2} ds dt = |w|_{H^{1/2}(\partial\widehat{\Omega})}^2,$$

we immediately obtain the desired result. □

Lemma 4.5. *$\|(I - \Pi_\ell)w\|_{L_2(\partial\widehat{\Omega})}^2 \lesssim \widehat{h}_\ell^2 |w|_{H^1(\partial\widehat{\Omega})}^2$ holds for all $w \in H^1(\partial\widehat{\Omega})$ and all $\ell = 0, \dots, L + 1$.*

Proof. Analogously to the proof of Lemma 4.4, one can derive from Theorem 4.1 in Ref. 26 and (4.1) that $|(I - \Pi_\ell)w|_{H^1(\partial\widehat{\Omega})}^2 \lesssim \widehat{h}_\ell^2 |w|_{H^2(\partial\widehat{\Omega})}^2$ holds for all $w \in H^1(\partial\widehat{\Omega})$. The desired result follows using a standard Aubin-Nitsche duality argument. □

The following Lemma shows that an L_2 -orthogonal decomposition almost realizes the minimum of arbitrary decompositions.

Lemma 4.6. *The estimate*

$$\sum_{\ell=1}^{L+1} \widehat{h}_\ell^{-2} \|(Q_\ell - Q_{\ell-1})w\|_{L_2(\partial\widehat{\Omega})}^2 \approx \inf_{w_\ell \in \mathcal{W}^{(\ell)}, w = \sum_{\ell=1}^{L+1} w_\ell} \sum_{\ell=1}^{L+1} \widehat{h}_\ell^{-2} \|w_\ell\|_{L_2(\partial\widehat{\Omega})}^2$$

holds for all $w \in H^1(\partial\widehat{\Omega})$.

Proof. Since $(Q_\ell - Q_{\ell-1})w \in \mathcal{W}^{(\ell)}$, we immediately obtain

$$\sum_{\ell=1}^{L+1} \widehat{h}_\ell^{-2} \|(Q_\ell - Q_{\ell-1})w\|_{L_2(\partial\widehat{\Omega})}^2 \geq \inf_{w_\ell \in \mathcal{W}^{(\ell)}, w = \sum_{\ell=1}^{L+1} w_\ell} \sum_{\ell=1}^{L+1} \widehat{h}_\ell^{-2} \|w_\ell\|_{L_2(\partial\widehat{\Omega})}^2.$$

For the proof of the other direction, let $w \in H^1(\partial\widehat{\Omega})$ be arbitrary but fixed. Consider any fixed representation of $w = \sum_{\ell=1}^{L+1} w_\ell$ with $w_\ell \in \mathcal{W}^{(\ell)}$. Observe that w_ℓ can be uniquely written as

$$w_\ell = \sum_{n=1}^{\ell} w_\ell^{(n)}, \quad \text{where} \quad w_\ell^{(n)} = (Q_n - Q_{n-1})w_\ell.$$

From $w = \sum_{\ell=1}^{L+1} w_\ell$, we immediately obtain $(Q_n - Q_{n-1})w = \sum_{\ell=n}^{L+1} w_\ell^{(n)}$. Using the triangle inequality and the Cauchy-Schwarz inequality, we obtain

$$\begin{aligned} \widehat{h}_n^{-2} \|(Q_n - Q_{n-1})w\|_{L_2(\partial\widehat{\Omega})}^2 &\leq \widehat{h}_n^{-2} \left(\sum_{\ell=n}^{L+1} \|w_\ell^{(n)}\|_{L_2(\partial\widehat{\Omega})} \right)^2 \\ &\leq \underbrace{\left(\sum_{\ell=n}^{L+1} \frac{\widehat{h}_\ell^2}{\widehat{h}_n^2} \right)}_{\lesssim 1} \left(\sum_{\ell=n}^{L+1} \widehat{h}_\ell^{-2} \|w_\ell^{(n)}\|_{L_2(\partial\widehat{\Omega})}^2 \right). \end{aligned}$$

By summing over n , we obtain using orthogonality

$$\sum_{n=1}^{L+1} \widehat{h}_n^{-2} \|(Q_n - Q_{n-1})w\|_{L_2(\partial\widehat{\Omega})}^2 \lesssim \sum_{n=1}^{L+1} \sum_{\ell=n}^{L+1} \widehat{h}_\ell^{-2} \|w_\ell^{(n)}\|_{L_2(\partial\widehat{\Omega})}^2 = \sum_{\ell=1}^{L+1} \widehat{h}_\ell^{-2} \|w_\ell\|_{L_2(\partial\widehat{\Omega})}^2,$$

which finishes the proof. \square

Lemma 4.7. *The estimate $\sum_{\ell=1}^{L+1} \widehat{h}_\ell^{-2} \|(Q_\ell - Q_{\ell-1})w\|_{L_2(\partial\widehat{\Omega})}^2 \lesssim \|w\|_{H^1(\partial\widehat{\Omega})}^2$ holds for all $w \in H^1(\partial\widehat{\Omega})$.*

Proof. Using Lemma 4.6 and by bounding the infimum by a particular decomposition, and using Lemma 4.5, $\Pi_0 = 0$ and $\widehat{h}_0 \approx 1$ we obtain

$$\begin{aligned} \sum_{\ell=1}^L \widehat{h}_\ell^{-2} \|(Q_\ell - Q_{\ell-1})w\|_{L_2(\partial\widehat{\Omega})}^2 &\approx \inf_{w_\ell \in \mathcal{W}^{(\ell)}, w = \sum_{\ell=1}^{L+1} w_\ell} \sum_{\ell=1}^{L+1} \widehat{h}_\ell^{-2} \|w_\ell\|_{L_2(\partial\widehat{\Omega})}^2 \\ &\leq \sum_{\ell=1}^{L+1} \widehat{h}_\ell^{-2} \|\Pi_\ell - \Pi_{\ell-1}\|_{L_2(\partial\widehat{\Omega})}^2 \lesssim \sum_{\ell=1}^{L+1} \|\Pi_\ell - \Pi_{\ell-1}\|_{H^1(\partial\widehat{\Omega})}^2 = \|w\|_{H^1(\partial\widehat{\Omega})}^2, \end{aligned}$$

which finishes the proof. \square

Lemma 4.8. *The estimate $\sum_{\ell=1}^{L+1} \widehat{h}_\ell^{-1} \|(Q_\ell - Q_{\ell-1})w\|_{L_2(\partial\widehat{\Omega})}^2 \lesssim \|w\|_{H^{1/2}(\partial\widehat{\Omega})}^2$ holds for all $w \in H^1(\partial\widehat{\Omega})$.*

Proof. Orthogonality and Lemma 4.7 yield

$$\|Qw\|_{L_2(\partial\widehat{\Omega})}^2 = \|w\|_{L_2(\partial\widehat{\Omega})}^2 \quad \text{and} \quad \|Qw\|_{L_h}^2 \lesssim \|w\|_{H^1(\partial\widehat{\Omega})}^2,$$

where

$$Q := \begin{pmatrix} Q_1 - Q_0 \\ Q_2 - Q_1 \\ \vdots \\ Q_{L+1} - Q_L \end{pmatrix}, \quad \text{and} \quad \|(w_1, \dots, w_{L+1})\|_{L_h}^2 := \sum_{\ell=1}^{L+1} h_\ell^{-2} \|w_\ell\|_{L_2(\partial\widehat{\Omega})}^2.$$

Using Hilbert space interpolation theory, cf. Theorems 7.23 and 7.31 in Ref. 1, we immediately obtain the desired result. \square

As a next step, we define extension operators that extend the solution from one edge into the interior. Let $\widehat{\Gamma}^{(\delta)} := \{\gamma(s) : s \in (\delta - 1, \delta)\}$ for $\delta = 1, 2, 3, 4$ be the four sides of $\widehat{\Omega}$. We define the extension operator $\mathcal{E}^{(\ell, \widehat{\Gamma}^{(1)})} : \mathcal{W} \rightarrow \mathcal{V}$ as follows:

$$(\mathcal{E}^{(\ell, \widehat{\Gamma}^{(1)})} w)(x, y) := w(0, y) \underbrace{\max\{0, x/\eta^{(\ell, \widehat{\Gamma}^{(1)})}\}^p}_{\theta^{(\ell, \widehat{\Gamma}^{(1)})}(x)},$$

where $\eta^{(\ell, \widehat{\Gamma}^{(1)})} := \max\{\zeta_i^{(k,2)} : \zeta_i^{(k,2)} \leq \widehat{h}_\ell\}$ is the largest breakpoint which is smaller than or equal to \widehat{h}_ℓ . Note that (4.1) implies that $\eta^{(\ell, \widehat{\Gamma}^{(1)})} > 0$ and that

$$\eta^{(\ell, \widehat{\Gamma}^{(1)})} \approx \widehat{h}_\ell. \quad (4.6)$$

Note that by construction

$$\begin{aligned} (\mathcal{E}^{(\ell, \widehat{\Gamma}^{(1)})} w)|_{\widehat{\Gamma}^{(1)}} &= w, & (\mathcal{E}^{(\ell, \widehat{\Gamma}^{(1)})} w)|_{\widehat{\Gamma}^{(2)}} &= w(0, 1) \theta^{(\ell, \widehat{\Gamma}^{(1)})}(x), \\ (\mathcal{E}^{(\ell, \widehat{\Gamma}^{(1)})} w)|_{\widehat{\Gamma}^{(3)}} &= 0, & (\mathcal{E}^{(\ell, \widehat{\Gamma}^{(1)})} w)|_{\widehat{\Gamma}^{(4)}} &= w(0, 0) \theta^{(\ell, \widehat{\Gamma}^{(1)})}(x) \end{aligned}$$

for all $w \in \mathcal{W}$. For the other edges we define the extension operator $\mathcal{E}^{(\ell, \widehat{\Gamma}^{(\delta)})}$ and the function $\theta^{(\ell, \widehat{\Gamma}^{(\delta)})}$ analogously.

Let $\widehat{\mathbf{x}}^{(\delta)} := \gamma(\delta - 1)$ for $\delta = 1, 2, 3, 4$ be the vertices of $\widehat{\Omega} = (0, 1)^2$. Consider the vertex $\widehat{\mathbf{x}}^{(1)} = (0, 0)^\top$. The adjacent edges are $\widehat{\Gamma}^{(1)}$ and $\widehat{\Gamma}^{(4)}$. We define a vertex extension operator $\mathcal{E}^{(\ell, \widehat{\mathbf{x}}^{(1)})} : \mathcal{W} \rightarrow \mathcal{V}$ as follows:

$$(\mathcal{E}^{(\ell, \widehat{\mathbf{x}}^{(1)})} w)(x, y) := w(0, 0) \theta^{(\ell, \widehat{\Gamma}^{(1)})}(x) \theta^{(\ell, \widehat{\Gamma}^{(4)})}(y).$$

For any other vertices $\widehat{\mathbf{x}}^{(\delta)}$, we define $\mathcal{E}^{(\ell, \widehat{\mathbf{x}}^{(\delta)})}$ analogously. Let $\mathcal{E} : \mathcal{W} \rightarrow \mathcal{V}$ be an overall extension operator, defined as follows:

$$\mathcal{E} := \sum_{\ell=1}^{L+1} \mathcal{E}^{(\ell)}(Q_\ell - Q_{\ell-1}), \quad \text{where} \quad \mathcal{E}^{(\ell)} := \sum_{\delta=1}^4 \mathcal{E}^{(\ell, \widehat{\Gamma}^{(\delta)})} - \sum_{\delta=1}^4 \mathcal{E}^{(\ell, \widehat{\mathbf{x}}^{(\delta)})}. \quad (4.7)$$

By construction, we have

$$(\mathcal{E}^{(\ell)} w)|_{\partial\widehat{\Omega}} = w \quad \text{and} \quad (\mathcal{E} w)|_{\partial\widehat{\Omega}} = \sum_{\ell=1}^{L+1} (Q_\ell - Q_{\ell-1})w = (Q_{L+1} - Q_0)w = w. \quad (4.8)$$

As a next step, we estimate the H^1 -seminorm of the extension. Here, we estimate the constituent parts of $\mathcal{E}^{(\ell)}$ separately.

Lemma 4.9. *The estimate*

$$\begin{aligned} (\mathcal{E}^{(\ell, \widehat{\Gamma}^{(\delta)})} w, \mathcal{E}^{(n, \widehat{\Gamma}^{(\delta)})} q)_{H^1(\widehat{\Omega})}^2 &\lesssim \sqrt{\frac{\min\{\widehat{h}_\ell, \widehat{h}_n\}}{\max\{\widehat{h}_\ell, \widehat{h}_n\}}} (p^{-1} \widehat{h}_\ell |w|_{H^1(\widehat{\Gamma}^{(\delta)})}^2 + p^{-1} \widehat{h}_n |q|_{H^1(\widehat{\Gamma}^{(\delta)})}^2 \\ &\quad + p \widehat{h}_\ell^{-1} \|w\|_{L_2(\widehat{\Gamma}^{(\delta)})}^2 + p \widehat{h}_n^{-1} \|q\|_{L_2(\widehat{\Gamma}^{(\delta)})}^2) \end{aligned}$$

holds for all $w \in \mathcal{W}^{(\ell)}$, $q \in \mathcal{W}^{(n)}$ with $\ell, n = 1, \dots, L+1$ and $\delta = 1, \dots, 4$.

Proof. Observe that the definition of the extension operator and the tensor-product structure of the domain immediately imply

$$\begin{aligned} &(\mathcal{E}^{(\ell, \widehat{\Gamma}^{(\delta)})} w, \mathcal{E}^{(n, \widehat{\Gamma}^{(\delta)})} q)_{H^1(\widehat{\Omega})} \\ &= (\theta^{(\ell, \widehat{\Gamma}^{(\delta)})}, \theta^{(n, \widehat{\Gamma}^{(\delta)})})_{L_2(0,1)}(w, q)_{H^1(\widehat{\Gamma}^{(\delta)})} + (\theta^{(\ell, \widehat{\Gamma}^{(\delta)})}, \theta^{(n, \widehat{\Gamma}^{(\delta)})})_{H^1(0,1)}(w, q)_{L_2(\widehat{\Gamma}^{(\delta)})}. \end{aligned}$$

By computing the corresponding integrals, we immediately obtain

$$\begin{aligned} &|(\mathcal{E}^{(\ell, \widehat{\Gamma}^{(\delta)})} w, \mathcal{E}^{(n, \widehat{\Gamma}^{(\delta)})} q)_{H^1(\widehat{\Omega})}| \\ &\quad \approx p^{-1} \min\{\widehat{h}_\ell, \widehat{h}_n\} |(w, q)_{H^1(\widehat{\Gamma}^{(\delta)})}| + p(\max\{\widehat{h}_\ell, \widehat{h}_n\})^{-1} |(w, q)_{L_2(\widehat{\Gamma}^{(\delta)})}|. \end{aligned} \quad (4.9)$$

Using the Cauchy-Schwarz inequality and the geometric-arithmetic mean inequality, we obtain

$$(w, q)_{H^1(\widehat{\Gamma}^{(\delta)})} \leq \sqrt{\frac{1}{\widehat{h}_\ell \widehat{h}_n}} (\widehat{h}_\ell |w|_{H^1(\widehat{\Gamma}^{(\delta)})}^2 + \widehat{h}_n |q|_{H^1(\widehat{\Gamma}^{(\delta)})}^2)$$

and

$$(w, q)_{L_2(\widehat{\Gamma}^{(\delta)})} \leq \sqrt{\widehat{h}_\ell \widehat{h}_n} (\widehat{h}_\ell^{-1} \|w\|_{L_2(\widehat{\Gamma}^{(\delta)})}^2 + \widehat{h}_n^{-1} \|q\|_{L_2(\widehat{\Gamma}^{(\delta)})}^2).$$

By plugging these estimates into (4.9), we obtain the desired result. \square

The following lemma allows to estimate the vertex extensions by the function values on one of the adjacent edges.

Lemma 4.10. *The estimate*

$$\begin{aligned} (\mathcal{E}^{(\ell, \widehat{\mathbf{x}}^{(\delta)})} w, \mathcal{E}^{(n, \widehat{\mathbf{x}}^{(\delta)})} q)_{H^1(\widehat{\Omega})}^2 &\lesssim \frac{\min\{\widehat{h}_\ell, \widehat{h}_n\}}{\max\{\widehat{h}_\ell, \widehat{h}_n\}} (p^{-1} \widehat{h}_\ell |w|_{H^1(\widehat{\Gamma}^{(\delta)})}^2 + p^{-1} \widehat{h}_n |q|_{H^1(\widehat{\Gamma}^{(\delta)})}^2 \\ &\quad + p \widehat{h}_\ell^{-1} \|w\|_{L_2(\widehat{\Gamma}^{(\delta)})}^2 + p \widehat{h}_n^{-1} \|q\|_{L_2(\widehat{\Gamma}^{(\delta)})}^2) \end{aligned}$$

holds for all $w \in \mathcal{W}^{(\ell)}$, $q \in \mathcal{W}^{(n)}$ with $\ell, n = 1, \dots, L+1$ and $\delta = 1, \dots, 4$.

Proof. Let δ be arbitrary but fixed and observe that the edges adjacent to $\widehat{\mathbf{x}}^{(\delta)}$ are $\widehat{\Gamma}^{(\delta-1)}$ and $\widehat{\Gamma}^{(\delta)}$, where we make use of $\widehat{\Gamma}^{(0)} := \widehat{\Gamma}^{(4)}$. Direct computations yield

$$\begin{aligned} &|(\mathcal{E}^{(\ell, \widehat{\mathbf{x}}^{(\delta)})} w, \mathcal{E}^{(n, \widehat{\mathbf{x}}^{(\delta)})} q)_{H^1(\widehat{\Omega})}| \\ &= \left((\theta^{(\ell, \widehat{\Gamma}^{(\delta-1)})}, \theta^{(n, \widehat{\Gamma}^{(\delta-1)})})_{L_2(0,1)} (\theta^{(\ell, \widehat{\Gamma}^{(\delta)})}, \theta^{(n, \widehat{\Gamma}^{(\delta)})})_{H^1(0,1)} \right. \\ &\quad \left. + (\theta^{(\ell, \widehat{\Gamma}^{(\delta-1)})}, \theta^{(n, \widehat{\Gamma}^{(\delta-1)})})_{H^1(0,1)} (\theta^{(\ell, \widehat{\Gamma}^{(\delta)})}, \theta^{(n, \widehat{\Gamma}^{(\delta)})})_{L_2(0,1)} \right) |w(\widehat{\mathbf{x}}^{(\delta)})| |q(\widehat{\mathbf{x}}^{(\delta)})| \\ &\approx \frac{\min\{\widehat{h}_\ell, \widehat{h}_n\}}{\max\{\widehat{h}_\ell, \widehat{h}_n\}} |w(\widehat{\mathbf{x}}^{(\delta)})| |q(\widehat{\mathbf{x}}^{(\delta)})| \leq \frac{\min\{\widehat{h}_\ell, \widehat{h}_n\}}{\max\{\widehat{h}_\ell, \widehat{h}_n\}} (|w(\widehat{\mathbf{x}}^{(\delta)})|^2 + |q(\widehat{\mathbf{x}}^{(\delta)})|^2) \end{aligned} \tag{4.10}$$

Using standard estimates, we obtain

$$|w(\widehat{\mathbf{x}}^{(\delta)})|^2 \lesssim \|w\|_{L_2(\widehat{\Gamma}^{(\delta)})} \|w\|_{H^1(\widehat{\Gamma}^{(\delta)})} \lesssim p \widehat{h}_\ell^{-1} \|w\|_{L_2(\widehat{\Gamma}^{(\delta)})}^2 + p^{-1} \widehat{h}_\ell |w|_{H^1(\widehat{\Gamma}^{(\delta)})}^2$$

and an analogous statement for q . By plugging these results into (4.10), we obtain the desired result. \square

Lemma 4.11. *The relation $\sum_{\ell=1}^{L+1} \frac{\min\{\widehat{h}_\ell, \widehat{h}_n\}}{\max\{\widehat{h}_\ell, \widehat{h}_n\}} \approx \sum_{\ell=1}^{L+1} \sqrt{\frac{\min\{\widehat{h}_\ell, \widehat{h}_n\}}{\max\{\widehat{h}_\ell, \widehat{h}_n\}}} \approx 1$ holds for all $n = 1, \dots, L+1$.*

Proof. By splitting the sum and using the summation formula for the geometric series, we obtain

$$\sum_{\ell=1}^{L+1} \frac{\min\{\widehat{h}_\ell, \widehat{h}_n\}}{\max\{\widehat{h}_\ell, \widehat{h}_n\}} = \sum_{\ell=1}^n \frac{\widehat{h}_n}{\widehat{h}_\ell} + \sum_{\ell=n+1}^{L+1} \frac{\widehat{h}_\ell}{\widehat{h}_n} = \sum_{\ell=1}^n 4^{\ell-n} + \sum_{\ell=n+1}^{L+1} 4^{n-\ell} \approx 1.$$

The proof of $\sum_{\ell=1}^{L+1} \sqrt{\frac{\min\{\widehat{h}_\ell, \widehat{h}_n\}}{\max\{\widehat{h}_\ell, \widehat{h}_n\}}} \approx 1$ can be done analogously. \square

Using these estimates, we are now able to show that the overall extension operator \mathcal{E} is bounded as follows.

Lemma 4.12. *The estimate $|\mathcal{E}w|_{H^1(\widehat{\Omega})}^2 \lesssim p |w|_{H^{1/2}(\partial\widehat{\Omega})}^2$ holds for all $w \in \mathcal{W}$.*

Proof. Let $w \in \mathcal{W}$ be arbitrary but fixed and define $w_\ell := (Q_\ell - Q_{\ell-1})w$ for $\ell = 1, \dots, L+1$. Observe that the definition and the triangle inequality yield

$$|w_{L+1}|_{H^1(\partial\widehat{\Omega})}^2 = |(I - Q_L)w|_{H^1(\partial\widehat{\Omega})}^2 \leq |(I - \Pi_L)w|_{H^1(\partial\widehat{\Omega})}^2 + |(\Pi_L - Q_L)w|_{H^1(\partial\widehat{\Omega})}^2,$$

where Π_L is the H^1 -orthogonal projection. Using stability and Lemma 4.3, we obtain $\|(I - \Pi_L)w\|_{H^1(\partial\widehat{\Omega})}^2 \lesssim p^2 \widehat{h}^{-1} \|w\|_{H^{1/2}(\partial\widehat{\Omega})}^2$. Using this and Lemma 4.2 and the triangle inequality, we obtain

$$|w_{L+1}|_{H^1(\partial\widehat{\Omega})}^2 \lesssim p^2 \widehat{h}^{-1} \|w\|_{H^{1/2}(\partial\widehat{\Omega})}^2 + \widehat{h}^{-2} \|(I - \Pi_L)w\|_{L_2(\partial\widehat{\Omega})}^2 + \widehat{h}^{-2} \|(I - Q_L)w\|_{L_2(\partial\widehat{\Omega})}^2.$$

Using Lemmas 4.4 and 4.5, we further obtain

$$|w_{L+1}|_{H^1(\partial\widehat{\Omega})}^2 \lesssim p^2 \widehat{h}^{-1} \|w\|_{H^{1/2}(\partial\widehat{\Omega})}^2. \quad (4.11)$$

Observe that (4.7), the triangle inequality yield, and Assumption 3

$$|\mathcal{E}w|_{H^1(\widehat{\Omega})}^2 \lesssim \sum_{\delta=1}^4 \left| \sum_{\ell=1}^{L+1} \mathcal{E}^{(\ell, \widehat{\Gamma}^{(\delta)})} w_\ell \right|_{H^1(\widehat{\Omega})}^2 + \sum_{\delta=1}^4 \left| \sum_{\ell=1}^{L+1} \mathcal{E}^{(\ell, \widehat{\mathbf{x}}^{(\delta)})} w_\ell \right|_{H^1(\widehat{\Omega})}^2. \quad (4.12)$$

Using Lemmas 4.9 and 4.11, we obtain

$$\begin{aligned} \left| \sum_{\ell=1}^{L+1} \mathcal{E}^{(\ell, \widehat{\Gamma}^{(\delta)})} w_\ell \right|_{H^1(\widehat{\Omega})}^2 &= \sum_{\ell=1}^{L+1} \sum_{n=1}^{L+1} \left(\mathcal{E}^{(\ell, \widehat{\Gamma}^{(\delta)})} w_\ell, \mathcal{E}^{(n, \widehat{\Gamma}^{(\delta)})} w_n \right)_{H^1(\widehat{\Omega})} \\ &\lesssim p^{-1} \sum_{\ell=1}^{L+1} \widehat{h}_\ell |w_\ell|_{H^1(\partial\widehat{\Omega})}^2 + p \sum_{\ell=1}^{L+1} \widehat{h}_\ell^{-1} \|w_\ell\|_{L_2(\partial\widehat{\Omega})}^2. \end{aligned}$$

Analogously, we obtain using Lemmas 4.10 and 4.11

$$\left| \sum_{\ell=1}^{L+1} \mathcal{E}^{(\ell, \widehat{\mathbf{x}}^{(\delta)})} w_\ell \right|_{H^1(\widehat{\Omega})}^2 \lesssim p^{-1} \sum_{\ell=1}^{L+1} \widehat{h}_\ell |w_\ell|_{H^1(\partial\widehat{\Omega})}^2 + p \sum_{\ell=1}^{L+1} \widehat{h}_\ell^{-1} \|w_\ell\|_{L_2(\partial\widehat{\Omega})}^2.$$

By plugging these two estimates into (4.12), we obtain

$$|\mathcal{E}w|_{H^1(\widehat{\Omega})}^2 \lesssim p^{-1} \sum_{\ell=1}^{L+1} \widehat{h}_\ell |w_\ell|_{H^1(\partial\widehat{\Omega})}^2 + p \sum_{\ell=1}^{L+1} \widehat{h}_\ell^{-1} \|w_\ell\|_{L_2(\partial\widehat{\Omega})}^2.$$

Using (4.11) and Lemma 4.2, we obtain further

$$\begin{aligned} |\mathcal{E}w|_{H^1(\widehat{\Omega})}^2 &\lesssim p \|w\|_{H^{1/2}(\partial\widehat{\Omega})}^2 + p^{-1} \sum_{\ell=1}^L \widehat{h}_\ell^{-1} \|w_\ell\|_{L_2(\partial\widehat{\Omega})}^2 + p \sum_{\ell=1}^{L+1} \widehat{h}_\ell^{-1} \|w_\ell\|_{L_2(\partial\widehat{\Omega})}^2 \\ &\lesssim p \|w\|_{H^{1/2}(\partial\widehat{\Omega})}^2 + p \sum_{\ell=1}^{L+1} \widehat{h}_\ell^{-1} \|w_\ell\|_{L_2(\partial\widehat{\Omega})}^2. \end{aligned}$$

Using $w_\ell = (Q_\ell - Q_{\ell-1})w$ and Lemma 4.8, we obtain further

$$|\mathcal{E}w|_{H^1(\widehat{\Omega})}^2 \lesssim p \|w\|_{H^{1/2}(\partial\widehat{\Omega})}^2.$$

A Poincaré type argument finishes the proof. \square

Since we have an estimate for the extension operator \mathcal{E} , we can also give a corresponding estimate for the discrete harmonic extension \mathcal{H}_h . Before we do this, we state a standard result on the equivalence of the norms between physical domain and parameter domain.

Lemma 4.13. *For all patches $k = 1, \dots, K$, we have*

- $|u|_{H^s(\Omega^{(k)})} \approx (H^{(k)})^{1-s} |u \circ G_k|_{H^s(\widehat{\Omega})}$ for all $u \in H^s(\Omega^{(k)})$ and $s \in \{0, 1\}$, and
- $|w|_{H^s(\partial\Omega^{(k)})} \approx (H^{(k)})^{1/2-s} |w \circ G_k|_{H^s(\partial\widehat{\Omega})}$ for all $w \in H^{1/2}(\partial\Omega^{(k)})$ and $s \in \{0, 1/2, 1\}$,

where we use the notation $H^0 := L_2$. The same holds if $\partial\Omega^{(k)}$ and $\partial\widehat{\Omega}$ are replaced by $\Gamma^{(k,\ell)}$ and $\widehat{\Gamma}^{(k,\ell)}$, respectively.

Proof. The results for $s \in \{0, 1\}$ directly follow from Assumption 1 and the chain rule for differentiation and the substitution rule, see, e.g, Lemma 3.5 in Ref. 3. The results for $s = 1/2$ are then obtained by Hilbert space interpolation theory, cf. Theorems 7.23 and 7.31 in Ref. 1. \square

Finally, we can state the main theorem of this section.

Theorem 4.2. *The estimate*

$$|w|_{H^{1/2}(\partial\Omega^{(k)})}^2 \lesssim |\mathcal{H}_h w|_{H^1(\Omega^{(k)})}^2 \lesssim p |w|_{H^{1/2}(\partial\Omega^{(k)})}^2$$

holds for all $w \in W^{(k)}$.

Proof. Let $\mathcal{H} : H^{1/2}(\partial\Omega^{(k)}) \rightarrow H^1(\Omega^{(k)})$ be the harmonic extension. Since the H^1 -seminorm of the harmonic extension is equivalent to the $H^{1/2}$ -seminorm on the boundary and since the harmonic extension minimizes over $H^1(\Omega^{(k)}) \supset V^{(k)}$, we have

$$|w|_{H^{1/2}(\partial\Omega^{(k)})}^2 \lesssim |\mathcal{H} w|_{H^1(\Omega^{(k)})}^2 \leq |\mathcal{H}_h w|_{H^1(\Omega^{(k)})}^2,$$

which shows the first part of the estimate. Since the discrete harmonic extension minimizes over $V^{(k)}$, we obtain using Lemma 4.13

$$|\mathcal{H}_h w|_{H^1(\Omega^{(k)})}^2 = \inf_{v \in V^{(k)} : v|_{\partial\Omega^{(k)}} = w} |v|_{H^1(\Omega^{(k)})}^2 \approx \inf_{v \in \widehat{V}^{(k)} : v|_{\partial\widehat{\Omega}} = w \circ G_k} |v|_{H^1(\widehat{\Omega})}^2.$$

Since the operator \mathcal{E} preserves the Dirichlet boundary conditions, we have that $\mathcal{E}(w \circ G_k) \in \widehat{V}^{(k)}$ and thus we obtain using Lemmas 4.12 and 4.13

$$|\mathcal{H}_h w|_{H^1(\widehat{\Omega})}^2 \leq |\mathcal{E}(w \circ G_k)|_{H^1(\widehat{\Omega})}^2 \lesssim p |w \circ G_k|_{H^{1/2}(\partial\widehat{\Omega})}^2 \approx p |w|_{H^{1/2}(\partial\Omega^{(k)})}^2,$$

which shows the second part of the estimate. \square

4.2. An embedding result

The following Lemma shows that we are able to bound the function values of a spline function from above using the H^1 -norm. If we consider the physical domain, the additional scaling factor is such that it can be eliminated using the Poincaré inequality.

Lemma 4.14. *The estimates*

$$\sup_{x \in \widehat{\Omega}} |\widehat{u}(x)|^2 \lesssim \Lambda \|\widehat{u}\|_{H^1(\widehat{\Omega})}^2 \quad \text{and} \quad \sup_{x \in \Omega^{(k)}} |u(x)|^2 \lesssim \Lambda (|u|_{H^1(\Omega^{(k)})}^2 + (H^{(k)})^{-2} \|u\|_{L_2(\Omega^{(k)})}^2),$$

where $\Lambda := 1 + \log p + \log \frac{H^{(k)}}{h^{(k)}}$, hold for all $\widehat{u} \in \widehat{V}^{(k)}$ and $u \in V^{(k)}$.

Proof.

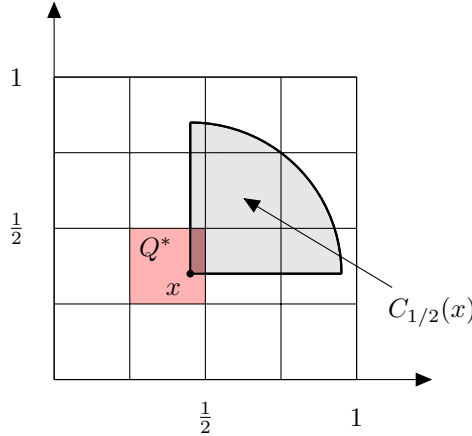


Figure 3: The chosen point x with its cone of radius $r = 1/2$

Let $\widehat{u} \in \widehat{V}^{(k)}$ and $x \in \widehat{\Omega}$ be arbitrary but fixed. We assume without loss of generality that

$$x \in [0, 1/2]^2. \quad (4.13)$$

The domain $[0, 1]^2$ is composed of elements $Q^{(i)} := [q_1^{(i)}, \bar{q}_1^{(i)}] \times [q_2^{(i)}, \bar{q}_2^{(i)}]$ on which the function u is polynomial. Let $Q^* := [q_1^*, \bar{q}_1^*] \times [q_2^*, \bar{q}_2^*]$ be the element containing x . Using Assumption 4, we obtain

$$\widehat{h}^{(k)} \approx \widehat{h} := \max_{\delta=1,2} \bar{q}_\delta^* - q_\delta^*. \quad (4.14)$$

For any $r > 0$, let $C_r(x)$ be the cone with vertex $x = (x_1, x_2)$, defined by

$$C_r(x) := \{\xi = (\xi_1, \xi_2) \in \mathbb{R}^2 : \|\xi - x\|_{\ell^2} \leq r, \xi_1 \geq x_1, \xi_2 \geq x_2\}.$$

The assumption (4.13) yields $C_{1/2}(x) \subset [0, 1]^2$. Consider the case that $C_\epsilon(x) \subset \overline{Q^*}$, where $\epsilon := 4^{-1}p^{-4}\widehat{h}^{3/2}$, first. Using the notation

$$v(r, \theta) := \widehat{u}(x_1 + r \sin \theta, x_2 + r \cos \theta)$$

and the fundamental theorem of calculus, we obtain

$$\widehat{u}(x) = v(0, \theta) = v(r, \theta) - \int_0^r \frac{\partial v}{\partial \rho}(\rho, \theta) d\rho$$

for all $r \in (0, 1/2]$ and all $\theta \in [0, \pi/2]$. By splitting up the integral and using Young's inequality, we obtain

$$\widehat{u}(x)^2 \lesssim v(r, \theta)^2 + \left(\int_0^\epsilon \frac{\partial v}{\partial \rho}(\rho, \theta) d\rho \right)^2 + \left(\int_\epsilon^r \frac{\partial v}{\partial \rho}(\rho, \theta) d\rho \right)^2 \quad (4.15)$$

for all $r \in (1/4, 1/2) \subset (\epsilon, 1/2)$. Using Theorem 4.76, eq. (4.6.2) and (4.6.1) in Ref. 30, we estimate the first integral as follows

$$\int_0^\epsilon \frac{\partial v}{\partial \rho}(\rho, \theta) d\rho \leq \epsilon \left\| \frac{\partial v}{\partial \rho} \right\|_{L^\infty((0, \epsilon) \times (0, \pi/2))} \leq \epsilon |\widehat{u}|_{W_\infty^1(Q^*)} \lesssim p^4 \widehat{h}^{-3/2} \epsilon \|\widehat{u}\|_{L_2(Q^*)}. \quad (4.16)$$

The second inequality is estimated using the Cauchy-Schwarz inequality as follows

$$\begin{aligned} \int_\epsilon^r \frac{\partial v}{\partial \rho}(\rho, \theta) d\rho &= \int_\epsilon^r \rho^{-1/2} \left(\frac{\partial v}{\partial \rho}(\rho, \theta) \rho^{1/2} \right) d\rho \\ &\leq (\log r - \log \epsilon)^{1/2} \left(\int_\epsilon^r \frac{\partial v}{\partial \rho}(\rho, \theta)^2 \rho d\rho \right)^{1/2}. \end{aligned}$$

By transforming the integral back to Cartesian coordinates, we obtain further

$$\int_\epsilon^r \frac{\partial v}{\partial \rho}(\rho, \theta) d\rho \leq (\log r - \log \epsilon)^{1/2} |\widehat{u}|_{H^1(C_r(x) \setminus C_\epsilon(x))} \lesssim (-\log \epsilon)^{1/2} |\widehat{u}|_{H^1(C_{1/2}(x) \setminus C_\epsilon(x))} \quad (4.17)$$

for all $r \in (1/4, 1/2)$. By plugging the estimates from (4.16) and (4.17) into (4.15), we obtain

$$\begin{aligned} \widehat{u}(x)^2 &\lesssim v(r, \theta)^2 + p^4 \widehat{h}^{-3/2} \epsilon \|\widehat{u}\|_{L_2(Q^*)}^2 - \log \epsilon |\widehat{u}|_{H^1(C_{1/2}(x) \setminus C_\epsilon(x))}^2 \\ &\lesssim v(r, \theta)^2 + (1 + \log p - \log \widehat{h}) \|\widehat{u}\|_{H^1(\widehat{\Omega})}^2 \end{aligned}$$

for all $r \in (1/4, 1/2)$ and all $\theta \in [0, \pi/2]$. By taking the integral over the domain $C_{1/2}(x) \setminus C_{1/4}(x)$, we obtain

$$\begin{aligned} \widehat{u}(x)^2 &\approx \int_{C_{1/2}(x) \setminus C_{1/4}(x)} \widehat{u}(x)^2 d\xi = \int_0^{\pi/2} \int_{1/4}^{1/2} r \widehat{u}(x)^2 dr d\theta \\ &\lesssim \int_0^{\pi/2} \int_{1/4}^{1/2} r v(r, \theta)^2 dr d\theta + (1 + \log p - \log \widehat{h}) \|\widehat{u}\|_{H^1(\widehat{\Omega})}^2 \int_0^{\pi/2} \int_{1/4}^{1/2} r dr d\theta \\ &\lesssim (1 + \log p - \log \widehat{h}) \|\widehat{u}\|_{H^1(\widehat{\Omega})}^2, \end{aligned}$$

which shows the desired result.

Now, we consider the case that $C_\epsilon \not\subset \overline{Q^*}$. In this case, we define

$$\tilde{x} := (\min\{x_1, \bar{q}_1^* - \epsilon\}, \min\{x_2, \bar{q}_2^* - \epsilon\})^\top$$

and observe that (4.14) yields that $\tilde{x} \in Q^*$. By construction, $C_\epsilon(\tilde{x}) \subset \overline{Q^*}$. Thus, using the arguments above, we obtain

$$\widehat{u}(\tilde{x})^2 \lesssim (1 + \log p - \log \widehat{h}) \|\widehat{u}\|_{H^1(\widehat{\Omega})}^2. \quad (4.18)$$

Moreover, we observe that $\|x - \tilde{x}\|_{\ell^2} \lesssim p^{-4} \widehat{h}^{3/2}$. Thus, we conclude using the fundamental theorem of calculus

$$(\widehat{u}(x) - \widehat{u}(\tilde{x}))^2 \leq \|x - \tilde{x}\|_{\ell^2} \|\widehat{u}\|_{W_\infty^1(Q^*)}^2 \lesssim p^{-4} \widehat{h}^{3/2} \|\widehat{u}\|_{W_\infty^1(Q^*)}^2.$$

Using Theorem 4.76, eq. (4.6.2) and (4.6.1) in Ref. 30, we further obtain

$$(\widehat{u}(x) - \widehat{u}(\tilde{x}))^2 \lesssim \|\widehat{u}\|_{L_2(Q^*)}^2.$$

By combining this result with (4.18), the triangle inequality and Young's inequality, we immediately obtain the first bound. The second statement follows directly using Lemma 4.13. \square

4.3. *The tearing lemma*

The variation of a function v over a domain T is defined via

$$|v|_{L_\infty^0(T)} := \inf_{c \in \mathbb{R}} \|v - c\|_{L_\infty(T)} = \frac{1}{2} (\text{ess.sup}_{x \in T} v(x) - \text{ess.inf}_{x \in T} v(x)),$$

where ess.sup and ess.inf are the essential supremum and infimum, respectively. Thus, we have obviously

$$|v|_{L_\infty^0(T_1 \cup T_2)} \leq |v|_{L_\infty^0(T_1)} + |v|_{L_\infty^0(T_2)} \quad \text{and} \quad \|v(x) - v\|_{L_\infty(T)} \leq 2|v|_{L_\infty(T)} \quad \text{for all } x \in T. \quad (4.19)$$

Lemma 4.15. *The estimates*

$$\begin{aligned} \sum_{\ell=1}^4 |\widehat{u}|_{H^{1/2}(\widehat{\Gamma}(\ell))}^2 &\leq |\widehat{u}|_{H^{1/2}(\partial\widehat{\Omega})}^2 \lesssim \sum_{\ell=1}^4 \left(|\widehat{u}|_{H^{1/2}(\widehat{\Gamma}(\ell))}^2 + \Lambda |\widehat{u}|_{L_\infty^0(\widehat{\Gamma}(\ell))}^2 \right), \\ \sum_{\ell=1}^4 |u|_{H^{1/2}(\Gamma^{(k,\ell)})}^2 &\leq |u|_{H^{1/2}(\partial\Omega^{(k)})}^2 \lesssim \sum_{\ell \in \mathcal{N}^{(k)}} \left(|u|_{H^{1/2}(\Gamma^{(k,\ell)})}^2 + \Lambda |u|_{L_\infty^0(\Gamma^{(k,\ell)})}^2 \right) \end{aligned}$$

hold for all $\widehat{u} \in \widehat{V}^{(k)}$ and $u \in V^{(k)}$, where $\Lambda := 1 + \log p + \log \frac{H^{(k)}}{h^{(k)}}$.

Proof. We begin with the first statement. Obviously,

$$|\widehat{u}|_{H^{1/2}(\partial\widehat{\Omega})}^2 = \sum_{\ell=1}^4 |\widehat{u}|_{H^{1/2}(\widehat{\Gamma}^{(\ell)})}^2 + \underbrace{\sum_{\ell=1}^4 \sum_{n \in \{1, \dots, 4\} \setminus \{\ell\}} \int_{\widehat{\Gamma}^{(\ell)}} \int_{\widehat{\Gamma}^{(n)}} \frac{|\widehat{u}(x) - \widehat{u}(y)|^2}{|x - y|^2} dx dy}_{\geq 0}, \quad (4.20)$$

which immediately shows the first side of the desired inequality. For the second side, we have to estimate the double integral. Let us consider a term with $\ell \neq n$. The edges $\widehat{\Gamma}^{(\ell)}$ are parameterized by the functions

$$\gamma^{(1)}(t) := (0, t), \quad \gamma^{(2)}(t) := (1 - t, 1), \quad \gamma^{(3)}(t) := (1, 1 - t), \quad \gamma^{(4)}(t) := (t, 0).$$

(Note that this parameterization is different than that of (4.5).) We define the functions $v := u \circ \gamma^{(\ell)}$ and $w := u \circ \gamma^{(n)}$. Simple calculations show that

$$|s + t|^2 \lesssim |\gamma^{(m)}(s) - \gamma^{(n)}(t)|^2$$

holds for all m and n with $m \neq n$. Thus, we obtain

$$\int_{\widehat{\Gamma}^{(m)}} \int_{\widehat{\Gamma}^{(n)}} \frac{|\widehat{u}(x) - \widehat{u}(y)|^2}{|x - y|^2} dx dy \lesssim \int_0^1 \int_0^1 \frac{|v(s) - w(t)|^2}{|s + t|^2} ds dt$$

By adding a productive zero and using the triangle inequality, we obtain

$$\begin{aligned} & \int_{\widehat{\Gamma}^{(m)}} \int_{\widehat{\Gamma}^{(n)}} \frac{|\widehat{u}(x) - \widehat{u}(y)|^2}{|x - y|^2} dx dy \\ & \lesssim \int_0^1 \int_0^1 \frac{|v(s) - v(0)|^2}{|s + t|^2} ds dt + \int_0^1 \int_0^1 \frac{|w(t) - w(0)|^2}{|s + t|^2} ds dt \\ & \approx \int_0^1 \frac{|v(s) - v(0)|^2}{s} ds + \int_0^1 \frac{|w(t) - w(0)|^2}{t} dt \end{aligned} \quad (4.21)$$

Consider the first of these summands. Let $\epsilon := p^{-4}(\widehat{h}^{(k)})^2$. By splitting up the integral, we obtain

$$\int_0^\epsilon \frac{|v(s) - v(0)|^2}{s} ds = \int_0^\epsilon \frac{|v(s) - v(0)|^2}{s} ds + \int_\epsilon^1 \frac{|v(s) - v(0)|^2}{s} ds. \quad (4.22)$$

Using the fundamental theorem of calculus and the Cauchy-Schwarz inequality, we obtain

$$\int_0^\epsilon \frac{|v(s) - v(0)|^2}{s} ds = \int_0^\epsilon \frac{\left(\int_0^s v'(z) dz\right)^2}{s} ds \leq \int_0^\epsilon \int_0^s v'(z)^2 dz ds \leq \epsilon |v|_{H^1(0,1)}^2.$$

Using a standard inverse estimate, cf. Theorem 4.76, eq. (4.6.5) in Ref. 30, we obtain further

$$\int_0^\epsilon \frac{|v(s) - v(0)|^2}{s} ds \lesssim \epsilon p^4 (\widehat{h}^{(k)})^{-2} \inf_{c \in \mathbb{R}} \|v - c\|_{L_2(0,1)}^2 \leq \epsilon p^4 (\widehat{h}^{(k)})^{-2} |v|_{L_\infty^0(0,1)}^2. \quad (4.23)$$

For the second integral in (4.22), we obtain using (4.19) that

$$\int_{\epsilon}^1 \frac{|v(s) - v(0)|^2}{s} ds \lesssim \int_{\epsilon}^1 \frac{1}{s} ds \|v(\cdot) - v(0)\|_{L_{\infty}(0,1)}^2 = -\log \epsilon |v|_{L_{\infty}^0(0,1)}^2. \quad (4.24)$$

The combination of (4.22), (4.23), (4.24), and the definition of ϵ yields

$$\int_0^1 \frac{|v(s) - v(0)|^2}{s} ds \lesssim \left(1 + \log p - \log \widehat{h}^{(k)}\right) |v|_{L_{\infty}^0(0,1)}^2.$$

Since we can estimate the second integral in (4.21) analogously, we have

$$\int_{\widehat{\Gamma}^{(m)}} \int_{\widehat{\Gamma}^{(n)}} \frac{|\widehat{u}(x) - \widehat{u}(y)|^2}{|x - y|^2} dx dy \lesssim \left(1 + \log p - \log \widehat{h}^{(k)}\right) \left(|\widehat{u}|_{L_{\infty}^0(\widehat{\Gamma}^{(m)})}^2 + |\widehat{u}|_{L_{\infty}^0(\widehat{\Gamma}^{(n)})}^2\right).$$

The combination of this estimate and (4.20) shows the first statement. The second statement is obtained using Lemma 4.13. \square

4.4. Condition number estimate

In the following, we prove Theorem 4.1. The idea of the proof is to use Theorem 22 in Ref. 21, which states that

$$\kappa(M_{\text{sD}} F) \leq \sup_{w \in \widetilde{W}} \frac{\|B_D^{\top} B w\|_S^2}{\|w\|_S^2}, \quad (4.25)$$

where \underline{w} is the coefficient vector associated to w .

Lemma 4.16. *Let $u = (u^{(1)}, \dots, u^{(K)}) \in \widetilde{W}$ with coefficient vector \underline{u} and let $w = (w^{(1)}, \dots, w^{(K)})$ with coefficient vector \underline{w} be such that*

$$\underline{w} = B_D^{\top} B \underline{u}.$$

Then, we have for each patch $\Omega^{(k)}$ and each edge $\Gamma^{(k,\ell)}$ connecting the vertices $\mathbf{x}^{(k,\ell,1)}$ and $\mathbf{x}^{(k,\ell,2)}$

$$\begin{aligned} |w^{(k)}|_{H^{1/2}(\Gamma^{(k,\ell)})}^2 &\lesssim |u^{(k)}|_{H^{1/2}(\Gamma^{(k,\ell)})}^2 + |u^{(\ell)}|_{H^{1/2}(\Gamma^{(k,\ell)})}^2 + \Delta^{(k,\ell,1)} + \Delta^{(k,\ell,2)}, \\ |w^{(k)}|_{L_{\infty}^0(\Gamma^{(k,\ell)})}^2 &\lesssim |u^{(k)}|_{L_{\infty}^0(\Gamma^{(k,\ell)})}^2 + |u^{(\ell)}|_{L_{\infty}^0(\Gamma^{(k,\ell)})}^2 + \Delta^{(k,\ell,1)} + \Delta^{(k,\ell,2)}, \end{aligned}$$

where $\Delta^{(k,\ell,i)} = 0$ for Algorithms A and C and

$$\Delta^{(k,\ell,i)} := \sum_{j \in \mathcal{P}(\mathbf{x}^{(k,\ell,i)})} |u^{(k)}(\mathbf{x}^{(k,\ell,i)}) - u^{(j)}(\mathbf{x}^{(k,\ell,i)})|^2$$

for Algorithm B.

Proof. First discuss the entries of the scaling matrix D , which was defined to be a diagonal matrix with coefficients $d_{i,i}$ for $i = 1, \dots, N_{\Gamma}$. $d_{i,i}$ is defined to be the number of Lagrange multipliers that act on the corresponding basis function. For each edge basis function, i.e., a basis function that is active on one edge and

vanishes on all vertices, we have only one Lagrange multiplier. Thus, $d_{i,i} = 1$ for the corresponding values of i .

For the vertex basis functions, we have to distinguish based on the choice of the primal degrees of freedom. For the Algorithms A and C, no Lagrange multiplier acts on the respective degree of freedom. Thus, we have $d_{i,i} = 1$. For Algorithm B, we have $d_{i,i} = |\mathcal{P}(\mathbf{x}^{(i)})| - 1$, where $\mathbf{x}^{(i)}$ is the corresponding vertex, since we use a fully redundant scheme.

Simple calculations yield for Algorithms A and C that

$$w^{(k)}|_{\Gamma^{(k,\ell)}} = u^{(k)}|_{\Gamma^{(k,\ell)}} - u^{(\ell)}|_{\Gamma^{(k,\ell)}} - \sum_{i=1}^2 \theta^{(k,\ell,i)} (u^{(k)}(\mathbf{x}^{(k,\ell,i)}) - u^{(\ell)}(\mathbf{x}^{(k,\ell,i)})).$$

where $\theta^{(k,\ell,i)}$ is the basis function in $\Phi^{(k)}$ such that $\theta^{(k,\ell,i)}(\mathbf{x}^{(k,\ell,i)}) = 1$. Since u satisfies the primal constraints, we have $u^{(k)}(\mathbf{x}^{(k,\ell,i)}) = u^{(\ell)}(\mathbf{x}^{(k,\ell,i)})$ and thus

$$w^{(k)}|_{\Gamma^{(k,\ell)}} = u^{(k)}|_{\Gamma^{(k,\ell)}} - u^{(\ell)}|_{\Gamma^{(k,\ell)}}.$$

Therefore, we have $|w^{(k)}|_{H^{1/2}(\Gamma^{(k,\ell)})}^2 \lesssim |u^{(k)}|_{H^{1/2}(\Gamma^{(k,\ell)})}^2 + |u^{(\ell)}|_{H^{1/2}(\Gamma^{(k,\ell)})}^2$ and $|w^{(k)}|_{L_\infty^0(\Gamma^{(k,\ell)})}^2 \lesssim |u^{(k)}|_{L_\infty^0(\Gamma^{(k,\ell)})}^2 + |u^{(\ell)}|_{L_\infty^0(\Gamma^{(k,\ell)})}^2$, which finishes the proof for the Algorithms A and C.

For Algorithm B, one obtains

$$\begin{aligned} w^{(k)}|_{\Gamma^{(k,\ell)}} &= u^{(k)}|_{\Gamma^{(k,\ell)}} - u^{(\ell)}|_{\Gamma^{(k,\ell)}} \\ &+ \sum_{i=1}^2 \frac{1}{|\mathcal{P}(\mathbf{x}^{(k,\ell,i)})| - 1} \theta^{(k,\ell,i)} \sum_{j \in \mathcal{P}(\mathbf{x}^{(k,\ell,i)}) \setminus \{\ell\}} (u^{(k)}(\mathbf{x}^{(k,\ell,i)}) - u^{(j)}(\mathbf{x}^{(k,\ell,i)})). \end{aligned}$$

Note that $\theta^{(k,\ell,i)}$ behaves like $\max\{0, 1 - |x - \mathbf{x}^{(k,\ell,i)}|/h^{(k)}\}^p$. So,

- $|\theta^{(k,\ell,i)}|_{H^{1/2}(\Gamma^{(k,\ell)})}^2 \lesssim |\theta^{(k,\ell,i)}|_{H^1(\Gamma^{(k,\ell)})} \|\theta^{(k,\ell,i)}\|_{L_2(\Gamma^{(k,\ell)})} \approx 1$, and
- $\|\theta^{(k,\ell,i)}\|_{L_\infty(\Gamma^{(k,\ell)})}^2 = 1$.

Assumption 3 yields $|\mathcal{P}(\mathbf{x}^{(k,\ell,i)})| \approx 1$. Thus, we immediately obtain the desired result also for Algorithm B. \square

The term $\Delta^{(k,\ell,i)}$ contains the differences $u^{(k)}(\mathbf{x}^{(k,\ell,i)}) - u^{(j)}(\mathbf{x}^{(k,\ell,i)})$ for any patch $\Omega^{(j)}$ that shares the corresponding vertex $\mathbf{x}^{(k,\ell,i)}$ with the patch $\Omega^{(k)}$. The following Lemma shows that these terms can be estimated from above with differences that only involve patches sharing an edge.

Lemma 4.17. *We have*

$$\Delta^{(k,\ell,i)} \lesssim \sum_{m \in \mathcal{P}(\mathbf{x}^{(k,\ell,i)})} \sum_{n \in \mathcal{P}(\mathbf{x}^{(k,\ell,i)}) \cap \mathcal{N}_\Gamma(m)} |u^{(m)}(\mathbf{x}^{(k,\ell,i)}) - u^{(n)}(\mathbf{x}^{(k,\ell,i)})|^2,$$

where $\Delta^{(k,\ell,i)}$ is as in Lemma 4.16.

Proof. The sum in $\Delta^{(k,\ell,i)}$ contains contributions of the form

$$|u^{(k)}(\mathbf{x}^{(k,\ell,i)}) - u^{(\ell)}(\mathbf{x}^{(k,\ell,i)})|^2.$$

If the patches $\Omega^{(k)}$ and $\Omega^{(\ell)}$ share an edge, we are done. Otherwise, there is a sequence $k =: n_0, n_1, \dots, n_j := \ell$ such that

- the patches $\Omega^{(n_i)}$ and $\Omega^{(n_{i-1})}$ share an edge, thus $n_i \in \mathcal{N}_\Gamma(n_{i-1})$ for all $i = 1, \dots, j$ and
- all patches $\Omega^{(n_i)}$ contain the vertex $\mathbf{x}^{(k,\ell,i)}$, thus $n_i \in \mathcal{P}(\mathbf{x}^{(k,\ell,i)})$ for all $i = 0, \dots, j$.

Thus, we have using Assumption 3

$$|u^{(k)}(\mathbf{x}^{(k,\ell,i)}) - u^{(\ell)}(\mathbf{x}^{(k,\ell,i)})|^2 \lesssim \sum_{i=1}^j |u^{(n_{i-1})}(\mathbf{x}^{(k,\ell,i)}) - u^{(n_i)}(\mathbf{x}^{(k,\ell,i)})|^2,$$

which finishes the proof since $n_{i-1} \in \mathcal{P}(\mathbf{x}^{(k,\ell,i)})$ and $n_i \in \mathcal{P}(\mathbf{x}^{(k,\ell,i)}) \cap \mathcal{N}_\Gamma(n_i)$. \square

Lemma 4.18. *Let $\Omega^{(k)}$ and $\Omega^{(\ell)}$ be two patches sharing the edge $\Gamma^{(k,\ell)}$. Assume that $\Gamma^{(k,\ell)}$ connects the two vertices $\mathbf{x}^{(k,\ell,1)}$ and $\mathbf{x}^{(k,\ell,2)}$. Provided that the integrals over the edge agree, i.e., $\int_{\Gamma^{(k,\ell)}} u^{(k)}(s) - u^{(\ell)}(s) ds = 0$, we have*

$$\sum_{i=1}^2 |u^{(k)}(\mathbf{x}^{(k,\ell,i)}) - u^{(\ell)}(\mathbf{x}^{(k,\ell,i)})|^2 \lesssim \Lambda \left(|\mathcal{H}_h^{(k)} u^{(k)}|_{H^1(\Omega^{(k)})}^2 + |\mathcal{H}_h^{(\ell)} u^{(\ell)}|_{H^1(\Omega^{(\ell)})}^2 \right)$$

for all $u = (u^{(1)}, \dots, u^{(K)}) \in V$, where $\Lambda := 1 + \log p + \max_{j=1, \dots, K} \log \frac{H^{(j)}}{h^{(j)}}$.

Proof. Let $\widehat{u}^{(k)} := u^{(k)} \circ G_k$ and $\widehat{u}^{(\ell)} := u^{(\ell)} \circ G_\ell$. By an unitary transformation (rotation, reflection), we obtain a representation such that the pre-image of the joint edge $\Gamma^{(k,\ell)}$ is $(0, 1) \times \{0\}$ and that the pre-image of $\mathbf{x}^{(k,\ell,1)}$ is 0. Thus, the assumption on the edge average reads as

$$\int_{(0,1) \times \{0\}} \widehat{u}^{(k)}(x) - \widehat{u}^{(\ell)}(x) dx = 0.$$

In the interior, consider the difference of the respective discrete harmonic extensions $\widehat{\mathcal{H}}_h^{(k)} \widehat{u}^{(k)} - \widehat{\mathcal{H}}_h^{(\ell)} \widehat{u}^{(\ell)}$. Lemma 4.14 yields

$$|u^{(k)}(\mathbf{x}^{(k,\ell,1)}) - u^{(\ell)}(\mathbf{x}^{(k,\ell,1)})|^2 = |\widehat{u}^{(k)}(0) - \widehat{u}^{(\ell)}(0)|^2 \lesssim \Lambda^{(k,\ell)} \|\widehat{\mathcal{H}}_h^{(k)} \widehat{u}^{(k)} - \widehat{\mathcal{H}}_h^{(\ell)} \widehat{u}^{(\ell)}\|_{H^1(\widehat{\Omega})}^2.$$

Using Theorem 1.24 in Ref. 25 (using the choice $\psi(u) := \int_{\widehat{\Gamma}} u(s) ds$), we obtain further

$$\begin{aligned} |u^{(k)}(\mathbf{x}^{(k,\ell,1)}) - u^{(\ell)}(\mathbf{x}^{(k,\ell,1)})|^2 &\lesssim \Lambda^{(k,\ell)} |\widehat{\mathcal{H}}_h^{(k)} \widehat{u}^{(k)} - \widehat{\mathcal{H}}_h^{(\ell)} \widehat{u}^{(\ell)}|_{H^1(\widehat{\Omega})}^2 \\ &\lesssim \Lambda^{(k,\ell)} (|\widehat{\mathcal{H}}_h^{(k)} \widehat{u}^{(k)}|_{H^1(\widehat{\Omega})}^2 + |\widehat{\mathcal{H}}_h^{(\ell)} \widehat{u}^{(\ell)}|_{H^1(\widehat{\Omega})}^2), \end{aligned}$$

which finishes the proof. \square

Using the last three Lemmas and Assumption 3, we immediately obtain

$$\begin{aligned}
 \sum_{k=1}^K \sum_{\ell \in \mathcal{N}_\Gamma(k)} |w^{(k)}|_{H^{1/2}(\Gamma^{(k,\ell)})}^2 &\lesssim \sum_{k=1}^K \sum_{\ell \in \mathcal{N}_\Gamma(k)} \left(|u^{(k)}|_{H^{1/2}(\Gamma^{(k,\ell)})}^2 + \Delta^{(k,\ell,1)} + \Delta^{(k,\ell,2)} \right) \\
 &\lesssim \sum_{k=1}^K \sum_{\ell \in \mathcal{N}_\Gamma(k)} \left(|u^{(k)}|_{H^{1/2}(\Gamma^{(k,\ell)})}^2 + \sum_{i=1}^2 |u^{(k)}(\mathbf{x}^{(k,\ell,i)}) - u^{(\ell)}(\mathbf{x}^{(k,\ell,i)})|^2 \right) \\
 &\lesssim \sum_{k=1}^K \sum_{\ell \in \mathcal{N}_\Gamma(k)} |u^{(k)}|_{H^{1/2}(\Gamma^{(k,\ell)})}^2 + \Lambda \sum_{k=1}^K |\mathcal{H}_h^{(k)} u^{(k)}|_{H^1(\Omega^{(k)})}^2,
 \end{aligned} \tag{4.26}$$

where Λ is as in Lemma 4.18, and an analogous estimate for the $L_\infty^0(\Gamma^{(k,\ell)})$ -seminorms. Finally, we are able to give a proof of the main theorem.

Proof. (of Theorem 4.1). Let $u = (u^{(1)}, \dots, u^{(K)})$ with coefficient vector \underline{u} be arbitrary but fixed and let $w = (w^{(1)}, \dots, w^{(K)})$ with coefficient vector \underline{w} be such that $\underline{w} = B_D^\top B \underline{u}$. Theorem 4.2 yields

$$\|B_D^\top B \underline{u}\|_S^2 = \|\underline{w}\|_S^2 = \sum_{k=1}^K |\mathcal{H}_h^{(k)} w^{(k)}|_{H^1(\Omega^{(k)})}^2 \lesssim p \sum_{k=1}^K |w^{(k)}|_{H^{1/2}(\partial\Omega^{(k)})}^2,$$

where $\mathcal{H}_h^{(k)}$ is the discrete harmonic extension into $H^1(\Omega^{(k)})$. Lemma 4.15 yields further

$$\|B_D^\top B \underline{u}\|_S^2 \lesssim p\Lambda \sum_{k=1}^K \sum_{\ell \in \mathcal{N}(k)} \left(|w^{(k)}|_{H^{1/2}(\Gamma^{(k,\ell)})}^2 + |w^{(k)}|_{L_\infty^0(\Gamma^{(k,\ell)})}^2 \right),$$

where Λ is as in Lemma 4.18. Using (4.26), we obtain

$$\begin{aligned}
 \|B_D^\top B \underline{u}\|_S^2 &\lesssim p\Lambda \sum_{k=1}^K \sum_{\ell \in \mathcal{N}_\Gamma(k)} \left(|u^{(k)}|_{H^{1/2}(\Gamma^{(k,\ell)})}^2 + |u^{(k)}|_{L_\infty^0(\Gamma^{(k,\ell)})}^2 \right) \\
 &\quad + p\Lambda^2 \sum_{k=1}^K |\mathcal{H}_h^{(k)} u^{(k)}|_{H^1(\Omega^{(k)})}^2.
 \end{aligned}$$

Lemma 4.15 and Theorem 4.2 yield

$$\|B_D^\top B \underline{u}\|_S^2 \lesssim \sum_{k=1}^K p\Lambda^2 |\mathcal{H}_h^{(k)} u^{(k)}|_{H^1(\Omega^{(k)})}^2 + p\Lambda |u^{(k)}|_{L_\infty^0(\partial\Omega^{(k)})}^2.$$

Lemma 4.14 yields

$$\|B_D^\top B \underline{u}\|_S^2 \lesssim p\Lambda^2 \sum_{k=1}^K \inf_{c \in \mathbb{R}} \left(|\mathcal{H}_h^{(k)} u^{(k)}|_{H^1(\Omega^{(k)})}^2 + (H^{(k)})^{-2} \|\mathcal{H}_h^{(k)}(u^{(k)} - c)\|_{L_2(\Omega^{(k)})}^2 \right).$$

A standard Poincaré inequality yields further

$$\|B_D^\top B \underline{u}\|_S^2 \lesssim p\Lambda^2 \sum_{k=1}^K |\mathcal{H}_h^{(k)} u^{(k)}|_{H^1(\Omega^{(k)})}^2 = p\Lambda^2 \|\underline{u}\|_S^2.$$

The combination of this estimate and (4.25) finishes the proof. \square

5. Numerical results

In this section, we give results from numerical experiments that illustrate the convergence theory presented in this paper. We consider the Poisson problem

$$\begin{aligned} -\Delta u(x, y) &= 2\pi^2 \sin(\pi x) \sin(\pi y) && \text{for } (x, y) \in \Omega \\ u &= 0 && \text{on } \partial\Omega, \end{aligned}$$

where we consider the two domains shown in Figure 4. The first domain is a circular ring consisting of 12 patches. Each patch is parameterized using a NURBS mapping of degree 2. The second domain is the Yeti-footprint, where we have decomposed the 12 patches of the standard representation into 84 patches to obtain a representation with inner vertices as well.

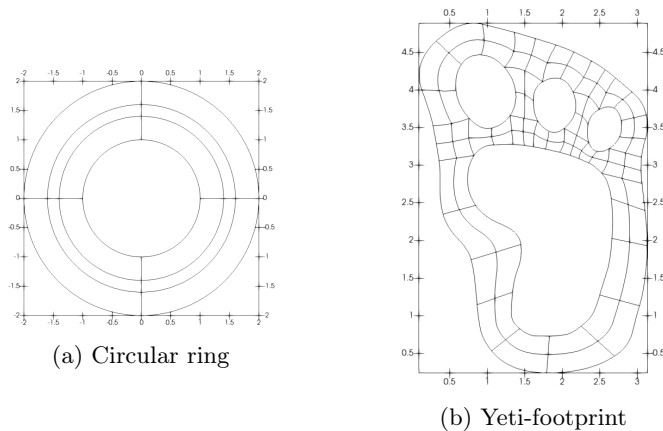


Figure 4: Computational domains and the decomposition into patches

The numerical experiments are based on the presented IETI-DP approach (Algorithms A, B and C), where the primal degrees of freedom for the corners are implemented by elimination of the corresponding degrees of freedom as suggested in Section 3. As patch-local discretization spaces $\widehat{V}^{(k)}$, we use spline spaces of maximum smoothness. The discretization of the triple ring on the coarsest grid level ($r = 0$) only consists for each of the patches only of global polynomials, i.e., there are no inner knots. The same holds for most of the patches of the Yeti-footprint,

$r \setminus p$	2		3		4		5		6		7		8	
	it	κ	it	κ	it	κ	it	κ	it	κ	it	κ	it	κ
2	10	5.96	11	6.46	12	6.79	12	7.29	13	7.53	14	7.99	15	8.16
3	11	6.69	12	7.30	12	7.81	13	8.28	13	8.64	14	9.03	15	9.32
4	12	7.67	12	8.41	13	9.02	13	9.53	14	9.98	14	10.38	14	10.73
5	12	8.81	13	9.68	14	10.38	14	10.96	14	11.47	15	11.91	15	12.31
6	14	10.14	14	11.13	15	11.91	15	12.57	15	13.13	15	13.63	15	14.07
7	15	11.65	15	12.75	16	13.62	17	14.35	16	14.97	17	15.52	17	16.00
8	15	13.33	17	14.56	17	15.51	17	16.31	17	16.99	17	17.58	17	18.11

Table 1: Iteration counts (it) and condition numbers κ ; Algorithm A; triple ring

$r \setminus p$	2		3		4		5		6		7		8	
	it	κ	it	κ	it	κ	it	κ	it	κ	it	κ	it	κ
2	14	15.43	13	15.69	13	15.56	15	16.32	15	16.33	16	17.01	16	17.04
3	15	18.32	15	18.51	15	18.93	16	19.45	17	19.81	17	20.30	18	20.63
4	16	21.52	16	21.77	16	22.24	16	22.73	16	23.19	17	23.63	18	24.02
5	17	24.92	17	25.17	18	25.67	17	26.18	18	26.67	18	27.12	18	27.54
6	18	28.47	18	28.73	19	29.25	19	29.78	19	30.28	19	30.75	21	31.19
7	19	32.16	19	32.43	21	32.97	22	33.52	21	34.05	22	34.54	21	34.99
8	21	36.01	22	36.28	22	36.84	22	37.42	23	37.96	23	38.47	24	38.95

Table 2: Iteration counts (it) and condition numbers κ ; Algorithm B; triple ring

while the discretization for the 20 patches that have a non-square shape have one inner knot that connects the midpoints of the two longer sides of the patch. The finer discretizations ($r = 1, 2, \dots$) are obtained by a uniform refinement such that the grid size behaves like $h \approx 2^{-r}$. The newly introduced knots are single knots, so splines of maximum smoothness are obtained in the interior of each patch. The local subproblems are solved using a standard direct solver. The overall problem is solved using a conjugate gradient solver, preconditioned with the scaled Dirichlet preconditioner. The starting value has been a vector with random entries in the interval $[-1, 1]$. The iteration has been stopped when the ℓ_2 -norm of the residual has been reduced by a factor 10^{-6} compared to the ℓ_2 -norm of the initial residual. For all numerical experiments, we present the required number of iterations (it) and the condition number (κ) of the preconditioned system $M_{\text{SD}}F$ as estimated by the conjugate gradient solver.

In the Tables 1, 2 and 3, we present the numerical experiments for the triple ring (Figure 4a). In Table 1, we provide the results for the case that only the vertex values are chosen as primal degrees of freedom (Algorithm A). Here, we observe that the condition number grows like $\log^2 H/h$ as predicted by the theory. The growth in the spline degree seems to be like \sqrt{p} or even slower, while the theory predicts $p \log^2 p$. In Table 2, we consider the case that only the averages of the function

r \ p	2		3		4		5		6		7		8	
	it	κ	it	κ	it	κ	it	κ	it	κ	it	κ	it	κ
2	9	2.18	9	2.33	10	2.55	10	2.68	11	2.91	11	3.01	12	3.24
3	10	2.46	10	2.74	10	3.00	11	3.21	11	3.42	12	3.58	12	3.77
4	10	2.92	11	3.29	11	3.59	11	3.85	12	4.07	12	4.26	12	4.44
5	11	3.47	11	3.93	12	4.27	12	4.55	13	4.80	13	5.01	13	5.20
6	12	4.16	12	4.64	13	5.01	13	5.32	13	5.59	14	5.82	14	6.03
7	13	4.89	14	5.42	14	5.82	15	6.16	15	6.44	15	6.69	16	6.91
8	14	5.69	15	6.25	15	6.69	16	7.05	16	7.35	16	7.61	16	7.84

Table 3: Iteration counts (it) and condition numbers κ ; Algorithm C; triple ring

r \ p	2		3		4		5		6		7		8	
	it	κ	it	κ	it	κ	it	κ	it	κ	it	κ	it	κ
1	12	3.02	14	4.06	15	4.72	17	5.60	18	6.13	20	6.88	22	7.33
2	14	4.64	16	5.76	17	6.64	18	7.49	20	8.15	21	8.86	22	9.39
3	16	6.71	19	8.03	20	9.11	20	10.04	21	10.83	23	11.57	24	12.20
4	20	9.22	22	10.77	22	12.03	23	13.09	23	14.01	24	14.84	26	15.57
5	22	12.17	23	13.96	25	15.40	25	16.60	26	17.64	27	18.59	27	19.39
6	25	15.58	26	17.59	26	19.20	27	20.55	28	21.72	28	22.73	29	23.65
7	27	19.42	28	21.69	28	23.48	29	24.94	31	26.26	30	27.35	32	28.79

Table 4: Iteration counts (it) and condition numbers κ ; Algorithm A; Yeti-footprint

values on the edges are chosen as primal degrees of freedom (Algorithm B). Here, the condition number seems to be larger than that of Algorithm A. Here, the growth of the condition number in the grid size looks like $\log^2 H/h$. The growth in the spline degree seems to be linear. Table 3 shows the results for the case that both kinds of primal degrees of freedom are combined (Algorithm C). As expected, the obtained condition numbers are smaller than those obtained with Algorithms A and B. We observe the same behavior as for Algorithm A.

In the Tables 4, 5 and 6, we present the numerical experiments for the Yeti-footprint (Figure 4b). Here, we observe in all cases that the growth of the condition number in the grid size is like $\log^2 H/h$ and the growth in the spline degree is like \sqrt{p} or slightly smaller. Opposite to the results for the triple ring, we observe that Algorithm B performs better than Algorithm A. Again, Algorithm C yields the smallest condition numbers, see Table 6.

6. Conclusions

In the paper, we have extended the known convergence analysis for IETI-DP methods such that it also covers the case that only edge-averages are used as primal

r \ p	2		3		4		5		6		7		8	
	it	κ	it	κ	it	κ	it	κ	it	κ	it	κ	it	κ
1	8	1.47	9	1.64	10	1.79	11	1.94	12	2.08	13	2.22	14	2.34
2	10	2.03	11	2.25	12	2.46	13	2.65	14	2.82	15	2.94	16	3.13
3	13	2.77	14	3.04	15	3.31	15	3.55	16	3.76	17	3.95	18	4.13
4	15	3.70	16	4.03	17	4.33	18	4.62	18	4.87	20	5.09	21	5.31
5	18	4.80	19	5.19	19	5.55	20	5.87	21	6.15	21	6.41	23	6.65
6	20	6.08	21	6.52	22	6.93	22	7.29	23	7.62	24	7.91	24	8.17
7	22	7.53	23	8.03	23	8.49	24	8.90	25	9.26	25	9.58	26	9.87

Table 5: Iteration counts (it) and condition numbers κ ; Algorithm B; Yeti-footprint

r \ p	2		3		4		5		6		7		8	
	it	κ	it	κ	it	κ	it	κ	it	κ	it	κ	it	κ
1	6	1.22	7	1.38	8	1.50	9	1.65	10	1.77	11	1.88	12	1.97
2	8	1.43	9	1.61	10	1.78	11	1.95	12	2.10	13	2.25	14	2.39
3	10	1.84	11	2.11	12	2.34	13	2.56	13	2.74	14	2.91	16	3.09
4	12	2.48	13	2.82	14	3.12	15	3.38	16	3.62	17	3.83	18	4.04
5	14	3.32	15	3.75	16	4.13	17	4.45	18	4.73	19	4.99	20	5.23
6	17	4.41	18	4.93	18	5.37	19	5.76	20	6.10	21	6.40	22	6.69
7	19	5.73	20	6.34	21	6.86	21	7.31	22	7.70	23	8.05	24	8.37

Table 6: Iteration counts (it) and condition numbers κ ; Algorithm C; Yeti-footprint

degrees of freedom. Moreover, we provide estimates for the condition number that are explicit both in the grid sizes and in the spline degree. The main ingredient for that analysis is an estimate for the discrete harmonic extension, cf. Section 4.1. The dependence of the condition number on the grid size is the same as in any other standard analysis. In the spline degree, the bound on the convergence number depends like $p \log^2 p$ on the spline degree p . Most numerical experiments indicate that the true growth might be smaller, but one of the numerical experiments has shown a growth similar to the upper bound from the convergence theory. Since in Iso-geometric Analysis, only moderate values of p are of interest, the corresponding dependence seems to be moderate.

Note that the condition number of the stiffness matrix $\kappa(A)$ grows exponentially in the spline degree. This means that the condition number of the preconditioned IETI-DP system grows only logarithmic in $\kappa(A)$, if the spline degree p is increased. The same is observed for the grid size, since $\kappa(A) \approx h^{-2}$ and the condition number of the preconditioned IETI-DP system grows (poly-)logarithmic in h^{-1} or, equivalently, in $\kappa(A)$.

The extension of the presented method to problems with varying diffusion co-

efficients seems to be straight-forward. Our analysis for Algorithm B also allows the extension of the IETI-DP solvers (and their analysis) to discretizations with non-matching patches, that are of interest for moving domains, like for the analysis of electrical motors.

Acknowledgments

The first author was supported by the Austrian Science Fund (FWF): S117 and W1214-04, while the second author has received support by the Austrian Science Fund (FWF): P31048. Moreover, the authors thank Ulrich Langer for fruitful discussions and help with the study of existing literature.

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