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Preprint no. 2015-05



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hp-discontinuous Galerkin method based on local higher order reconstruction

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Abstract

We present a new adaptive higher-order finite element method (hp-FEM) for the solution of boundary value problems formulated in terms of partial differential equations (PDEs). The method does not use any information about the solved problem which makes it robust and equation-independent. It employs a higher-order reconstruction scheme over local element patches which makes it faster and easier to parallelize compared to hp-adaptive methods that are based on the solution of a reference problem on a globally hp-refined mesh. The method can be used for the solution of linear as well as nonlinear problems discretized by conforming or non-conforming finite element methods, and it can be combined with arbitrary a posteriori error estimators.

Keywords: hp-adaptivity, error estimates, higher-order reconstruction, optimal re-meshing 2000~MSC: 65M50,~65M60,~65D05

1. Introduction

Adaptive methods are an efficient tool for the numerical solution of PDEs. Automatic mesh refinement or, more generally, an enhancement of the functional space where the approximate solution is sought, can significantly reduce the computational cost. A prominent place among adaptive methods has the hp-FEM which leads to unconditional exponential convergence [1–4].

Various approaches to automatic adaptivity include refining an element without increasing its polynomial degree (h-refinement), increasing the polynomial degree of an element without spatial subdivision (p-refinement), and performing refinements that combine spatial splitting of an element with various distributions of the polynomial degrees in subelements (genuine hp-refinement [5, 6]).

To achieve exponential convergence, large higher-order elements must be used where the solution is smooth, and at the same time small low-order elements must be used where the solution exhibits non-smooth features such as singularities or internal/boundary layers. Therefore it would seem that looking at the smoothness of the solution is the best way to design a hp-adaptive method. However, hp-adaptive strategies based on smoothness estimation usually can only decide between h- and p-refinements because they do not have enough information to select optimal genuine hp-refinements [7–9].

To take full advantage of genuine hp-refinements, one has to obtain a better information about the error – not only as an error estimate in the form of a number per element, but about its shape as a function that is defined inside an element. This can be done by solving a reference problem on a globally hp-refined mesh [5, 6, 10]. It leads to superior convergence rates in terms of degrees

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of freedom, but the reference problem tends to be huge and therefore the convergence is rather slow in terms of computing time.

In this paper we propose a novel method for automatic hp-adaptivity that is guided by the approximation error, but it removes the tedious computation of the global reference solution. Instead, it calculates a more accurate approximation on local element patches constructed for each element separately. With the aid of weighted least square reconstruction, we construct piecewise polynomial function which approximate the exact solution. The presented hp-adaptive strategy is based on the comparing of the higher-order reconstruction with the approximate solution u_h and its L^2 -projections to lower-degrees polynomial spaces. Then, for each candidate, we directly estimate the number of degrees of freedom necessary to achieve a local tolerance for each element and propose a new (better) polynomial approximation degree and a new (better) element size. Consequently, a new triangular grid is constructed.

Although we originally developed this technique for the discontinuous Galerkin method, it can be simply modified to other types of finite element approximations including conforming finite elements, mixed finite elements, etc. Moreover, since this approach is based on the reconstruction of the approximate solution, we can employ it for arbitrary (linear as well as non-linear) boundary value problems. Finally, its extension to 3D problems is straightforward.

The outline of the paper is as follows: In Section 2, we introduce the governing equations and their discretization by the discontinuous Galerkin method. In Section 3, we introduce the higher-order reconstruction technique and formulate the saturation assumption. Its validity is numerically demonstrated in Section 4. The main novelty of this paper is presented in Section 5, where we present the hp-adaptive strategy. In Section 6, we present three numerical examples that demonstrate the performance and robustness of the proposed method.

2. Problem description

2.1. Governing equations

We consider the nonlinear convection-diffusion problem

$$\nabla \cdot \boldsymbol{f}(u) - \nabla \cdot (\boldsymbol{K}(u)\nabla u) = g(x), \tag{1a}$$

$$u|_{\partial\Omega_D} = u_D,$$
 (1b)

$$K(u)\frac{\partial u}{\partial n}|_{\partial\Omega_N} = g_N,$$
 (1c)

where $u: \Omega \to \mathbb{R}$ is an unknown scalar function defined on $\Omega \in \mathbb{R}^2$. We assume that Ω is polygonal for simplicity. Moreover, $f(u) = (f_1(u), f_2(u)) : \mathbb{R} \to \mathbb{R}^2$ and $K(u) = \{K_{ij}(u)\}_{i,j=1}^2 : \mathbb{R} \to \mathbb{R}^{2\times 2}$ are nonlinear functions of their arguments, n is the unit outer normal to $\partial\Omega$ and $\emptyset \neq \partial\Omega_D \cup \partial\Omega_N = \partial\Omega$ are disjoint parts of the boundary of Ω . Symbols ∇ and ∇ · mean the gradient and divergence operators, respectively.

We assume that $f_s \in C^1(\mathbb{R})$, $f_s(0) = 0$, s = 1, 2, K is bounded and positively definite, $g \in L^2(\Omega)$, u_D is the trace of some $u^* \in H^1(\Omega) \cap L^{\infty}(\Omega)$ on $\partial \Omega_D$ and $g_N \in L^2(\partial \Omega_N)$. We use the standard notation for function spaces (see, e.g., [11]): $L^p(\Omega)$ denote the Lebesgue spaces, $W^{k,p}(\Omega)$, $H^k(\Omega) = W^{k,2}(\Omega)$ are the Sobolev spaces and $P^k(M)$ denotes the space of polynomial functions of degree $\leq k$ defined on the domain $M \subset \mathbb{R}^2$. Let us note that any function from $P^k(M)$ can be interpreted as a polynomial function defined on \mathbb{R}^2 restricted to M. By $\phi|_M$ we denote the restriction of a function ϕ on M.

In order to introduce the weak solution, we define the spaces

$$V := \{ v; \ v \in H^1(\Omega), \ v|_{\partial \Omega_D} = 0 \}, \quad W := \{ v; \ v \in H^1(\Omega), \ v - u^* \in V \}.$$
 (2)

We say that function u is the weak solution of (1), if the following conditions are satisfied

$$u \in W \cap L^{\infty}(\Omega),$$
 (3a)

$$\int_{\Omega} \left[\nabla \cdot \boldsymbol{f}(u) \, v + (\boldsymbol{K}(u) \nabla u) \cdot \nabla v \right] \, \mathrm{d}x = \int_{\Omega} g v \, \mathrm{d}x + \int_{\partial \Omega_N} u \, v \, \mathrm{d}S \qquad \forall v \in V. \tag{3b}$$

The assumption $u \in L^{\infty}(\Omega)$ in (3) guarantees the boundedness of functions f(u) and K(u) and therefore the existence of the integrals in (3a). This assumption can be weakened if functions f(u) and K(u) satisfy some growth conditions.

2.2. Discretization of the problem

Let \mathscr{T}_h (h > 0) be a partition of the closure $\overline{\Omega}$ of the domain Ω into a finite number of triangles K with mutually disjoint interiors. We call $\mathscr{T}_h = \{K\}_{K \in \mathscr{T}_h}$ a triangulation of Ω and for simplicity, we assume that \mathscr{T}_h satisfies the conforming properties from the finite element method, see, e.g, [12]. The diameter of $K \in \mathscr{T}_h$ is denoted by h_K .

By \mathscr{F}_h we denote the set of all open edges of all elements $K \in \mathscr{T}_h$. Further, the symbol \mathscr{F}_h^I stands for the set of all $\Gamma \in \mathscr{F}_h$ that are contained in Ω (inner edges). Moreover, we introduce notations \mathscr{F}_h^D and \mathscr{F}_h^N for the sets of all $\Gamma \in \mathscr{F}_h$ such that $\Gamma \subset \partial \Omega_D$ and $\Gamma \subset \partial \Omega_N$, respectively. In order to simplify the notation, we put $\mathscr{F}_h^{ID} = \mathscr{F}_h^I \cup \mathscr{F}_h^D$ and $\mathscr{F}_h^B = \mathscr{F}_h^D \cup \mathscr{F}_h^N$ (superscript B as boundary). Finally, for each $\Gamma \in \mathscr{F}_h$, we define a unit normal vector \mathbf{n}_{Γ} . We assume that for $\Gamma \in \mathscr{F}_h^B$ the vector \mathbf{n}_{Γ} has the same orientation as the outer normal of $\partial \Omega$. For \mathbf{n}_{Γ} , $\Gamma \in \mathscr{F}_h^I$, the orientation is arbitrary but fixed for each edge.

Over the triangulation \mathscr{T}_h we define the so-called broken Sobolev space $H^s(\Omega, \mathscr{T}_h) := \{v; v|_K \in H^s(K) \ \forall \ K \in \mathscr{T}_h\}, \ s \geq 0 \ \text{with the seminorm} \ |v|_{H^s(\Omega, \mathscr{T}_h)} := \left(\sum_{K \in \mathscr{T}_h} |v|_{H^s(K)}^2\right)^{1/2}, \ \text{where} \ |\cdot|_{H^s(K)} \ \text{denotes the seminorm of the Sobolev space} \ H^s(K), \ K \in \mathscr{T}_h.$

Moreover, to each $K \in \mathscr{T}_h$, we assign a positive integer p_K (=local polynomial degree). Then we define the set $p := \{p_K, K \in \mathscr{T}_h\}$ and the finite-dimensional subspace of $H^2(\Omega, \mathscr{T}_h)$ which consists of discontinuous piecewise polynomial functions associated with the vector p by

$$S_h^{\mathsf{p}} = \{ v; \ v \in L^2(\Omega), \ v|_K \in P_{p_K}(K) \ \forall K \in \mathcal{T}_h \}, \tag{4}$$

where $P_{p_K}(K)$ denotes the space of all polynomials on K of degree $\leq p_K$, $K \in \mathscr{T}_h$. The dimension of S_h^{p} is called the number of degrees of freedom (DOF) which satisfies

DOF := dim
$$S_h^{\mathsf{p}} = \sum_{K \in \mathcal{T}_h} (p_K + 1)(p_K + 2)/2.$$
 (5)

The pair $\{\mathscr{T}_h, \mathsf{p}\}$ is called the hp-mesh and we use the notation $\mathscr{T}_h^{\mathsf{p}} := \{\mathscr{T}_h, \mathsf{p}\}$. The hp-mesh $\mathscr{T}_h^{\mathsf{p}}$ uniquely corresponds to the space S_h^{p} and vice-versa.

For the purpose of the presented hp-adaptation method, we also define the space

$$S_h^{p+1} := \{ v_h \in L^2(\Omega); \ v_h|_K \in P^{p_K+1}(K) \ \forall K \in \mathcal{T}_h \}.$$
 (6)

Obviously, $S_h^{\mathsf{p}} \subset S_h^{\mathsf{p}+1} \subset H^2(\Omega, \mathscr{T}_h)$.

Furthermore, let $\Gamma \in \mathscr{F}_h^I$, $v \in H^2(\Omega, \mathscr{T}_h)$, by $v|_{\Gamma}^{(+)}$ and $v|_{\Gamma}^{(-)}$ we denote the trace of v on Γ from the direction and from the opposite direction of n_{Γ} , respectively. Moreover, by $\langle v \rangle_{\Gamma}$ and $\llbracket v \rrbracket_{\Gamma}$ we denote the mean value and the jump of v on Γ , respectively. Finally, for $\Gamma \in \mathscr{F}_h^B$, we put $\langle v \rangle_{\Gamma} = \llbracket v \rrbracket_{\Gamma} := v|_{\Gamma}^{(+)}$. In case that n_{Γ} , $\llbracket \cdot \rrbracket_{\Gamma}$ and $\langle \cdot \rangle_{\Gamma}$ are arguments of $\int_{\Gamma} \dots dS$, $\Gamma \in \mathscr{F}_h$, we omit the subscript Γ and write simply n, $\llbracket \cdot \rrbracket$ and $\langle \cdot \rangle_{\Gamma}$ respectively.

We discretize equation (1a) with the aid of the interior penalty Galerkin (IPG) variant of the discontinuous Galerkin (DG) method in the same way as in [13, 14]. For $u, v \in H^2(\Omega, \mathcal{T}_h)$ we define the form $a_h: H^2(\Omega, \mathcal{T}_h) \times H^2(\Omega, \mathcal{T}_h) \to \mathbb{R}$ by

$$a_{h}(u,v) := \sum_{\Gamma \in \mathscr{F}_{h}} \int_{\Gamma} H(u|_{\Gamma}^{(+)}, u|_{\Gamma}^{(-)}, \boldsymbol{n}) \llbracket v \rrbracket \, dS - \sum_{K \in \mathscr{T}_{h}} \int_{K} \boldsymbol{f}(u) \cdot \nabla v \, dx$$

$$+ \sum_{K \in \mathscr{T}_{h}} \int_{K} \boldsymbol{K}(u) \nabla u \cdot \nabla v \, dx + \sum_{\Gamma \in \mathscr{F}_{h}^{I}} \int_{\Gamma} \left(-\langle \boldsymbol{K}(u) \nabla u \rangle \cdot \boldsymbol{n} \llbracket v \rrbracket + g \langle \boldsymbol{K}(u) \nabla v \rangle \cdot \boldsymbol{n} \llbracket u \rrbracket + \sigma \llbracket u \rrbracket \, \llbracket v \rrbracket \right) dS$$

$$+ \sum_{\Gamma \in \mathscr{F}_{h}^{D}} \int_{\Gamma} \left(-\boldsymbol{K}(u) \nabla u \cdot \boldsymbol{n} \, v + g \boldsymbol{K}(u) \nabla v \cdot \boldsymbol{n} \, (u - u_{D}) + \sigma(u - u_{D}) \, v \right) dS - (g, v) - (g_{N}, v)_{N},$$

where g=-1,0 and 1 for SIPG, IIPG and NIPG variants of discontinuous Galerkin method, respectively, the penalty parameter σ is chosen by $\sigma|_{\Gamma}=\varepsilon C_W h_{\Gamma}^{-1}$, $\Gamma\in\mathscr{F}_h$, where $h_{\Gamma}=\mathrm{diam}(\Gamma)$, $\Gamma\in\mathscr{F}_h$, ε denotes the amount of diffusion ($\approx \mathbf{K}(\cdot)$) and $C_W>0$ is a suitable constant which guarantees the convergence of the method. The function H in (7) is the numerical flux, well-known from finite volume methods (see, e.g., [15], Section 3.2), which approximates convective flux by $\mathbf{f}(u)\cdot\mathbf{n}\approx H\left(u|_{\Gamma}^{(+)},u|_{\Gamma}^{(-)},\mathbf{n}\right)$ on an element edge. On $\partial\Omega_D$ the value $u|_{\Gamma}^{(-)}$ is taken from the boundary conditions (1b) and on $\partial\Omega_N$ the value $u_{\Gamma}^{(-)}$ is extrapolated from the interior of Ω . We shall assume that the numerical flux is conservative and consistent, i.e., $H(u,v,\mathbf{n})=-H(v,u,-\mathbf{n})$ and $H(u,u,\mathbf{n})=\mathbf{f}(u)\cdot\mathbf{n}$, respectively, see [13] for details.

We say that function $u_h \in S_h^{\mathsf{p}}$ is an approximate solution of (3), if

$$a_h(u_h, v_h) = 0 \qquad \forall v_h \in S_h^{\mathbf{p}}. \tag{8}$$

The problem (8) represents a system of nonlinear algebraic equations, which we solve by a damped Newton-like method, where the Jacobi matrix is replaced by the flux matrix that arises by a partial linearization of the form a_h . More details, including the used stopping criteria, can be found in [16].

The time-dependent variant of (1) was analysed in [17], see also [18], where quasilinear elliptic boundary value problems were considered. Hence, the approximate solution satisfies the following a priori error estimate

$$||u - u_h||_{\mathrm{DG}} \le C \sum_{K \in \mathcal{T}_h} \frac{h_K^{2\mu_K - 2}}{p_K^{2s_K - 3}} ||u||_{H^{\mu_K}(K)}^2, \tag{9}$$

where u and u_h are the exact and the approximate solution, respectively, C>0 is a constant independent of h and p_K , $K \in \mathscr{T}_h$, p_K is the polynomial approximation degree on K, s_k denotes the local Sobolev regularity of the exact solution on K (i.e., $u|_K \in H^{s_K}(K)$, $K \in \mathscr{T}_h$), $\mu_K = \min(p_K + 1, s_K)$ and the DG-norm is defined by

$$||v||_{\mathrm{DG}} := \left(|v|_{H^{1}(\Omega,\mathscr{T}_{h})}^{2} + ||v||_{\mathrm{J}}^{2}\right)^{1/2}, \quad \text{where} \quad ||v||_{\mathrm{J}}^{2} := \sum_{\Gamma \in \mathscr{F}_{h}^{I}} \int_{\Gamma} \sigma \llbracket v \rrbracket^{2} \, \mathrm{d}S + \sum_{\Gamma \in \mathscr{F}_{h}^{D}} \int_{\Gamma} \sigma v^{2} \, \mathrm{d}S.$$

$$\tag{10}$$

In the following we going to estimate the error in the broken H^1 -seminorm, i.e., $|u-u_h|_{H^1(\Omega,\mathcal{T}_h)}$. It will be more rigorous to estimate the error in the DG-norm but since $[\![u_h-u]\!]|_{\Gamma}=[\![u_h]\!]|_{\Gamma}$, $\Gamma\in\mathcal{F}_h^I$ and $u|_{\Gamma}=u_D$, $\Gamma\in\mathcal{F}_h^D$, we have

$$||u - u_h||_{\mathrm{DG}}^2 = |u - u_h|_{H^1(\Omega, \mathscr{T}_h)}^2 + ||u - u_h||_{\mathrm{J}}^2 = |u - u_h|_{H^1(\Omega, \mathscr{T}_h)}^2 + \sum_{\Gamma \in \mathscr{F}_h^D} \int_{\Gamma} \sigma(u_D - u_h)^2 \, \mathrm{d}S.$$

The last term can be evaluated directly.

3. Error estimates based on a higher-order reconstruction

In [19–21], authors compute a reference solution $u^{\text{ref}} \in S_{h/2}^{p+1}$ on a globally refined mesh with an increased polynomial degree of elements, to guide automatic hp-adaptivity. The reference solution is defined by

$$a_{\frac{h}{2}}(u^{\text{ref}}, v_h) = 0 \qquad \forall v_h \in S_{h/2}^{\mathsf{p}+1},$$
 (11)

where $a_{\frac{h}{2}}$ is defined by (7) on the triangular mesh $\mathcal{T}_{h/2}$, which arises by the refinement of \mathcal{T}_h , where each $K \in \mathcal{T}_h$ is split onto 4 sub-triangles by connecting the middles of its edges, and

$$S_{h/2}^{\mathsf{p}+1} := \{ v_h \in L^2(\Omega); \ v_h|_{K'} \in P^{p_K+1}(K') \ \forall K' \in \mathscr{T}_{h/2} \}. \tag{12}$$

Then the *error indicator* is defined by

$$\tilde{\eta}_K := \|u_h - u^{\text{ref}}\|_K \qquad \forall K \in \mathcal{T}_h. \tag{13}$$

However, the computation of the reference solution by the solution of (11) is time consuming.

Our idea is to obtain an analogue to the reference solution without the necessity to solve the large global problem (11). We employ a *high-order piecewise polynomial* reconstruction for local element patches.

3.1. Higher-order reconstruction

Let $u_h \in S_h^{\mathsf{p}}$ be the approximate solution given by (8). Our aim is to construct for each $K \in \mathcal{T}_h$ a function $\tilde{u}_K \in P^{p_K+1}(K)$ such that it approximate the exact solution on K better than u_h , i.e.,

$$|\tilde{u}_K - u|_K| \ll |u_h|_K - u|_K|, \qquad K \in \mathcal{T}_h. \tag{14}$$

In order to construct \tilde{u}_K , we use a weighted least square approximation from the elements sharing at least a vertex with K. Let $K \in \mathcal{T}_h$, we denote by \mathcal{D}_K the patch

$$\mathscr{D}_K := K \cup \{ K' \in \mathscr{T}_h; \ K' \text{ share at least a vertex with } K \}$$
 (15)

and $w_{K'} > 0$, $K' \in \mathcal{D}_K$ be the weights. Then we define a function $\tilde{U}_K \in P^{p_K+1}(\mathcal{D}_K)$ by

$$\tilde{U}_K := \arg \min_{U_h \in P^{p_K + 1}(\mathscr{D}_K)} \sum_{K' \in \mathscr{D}_K} w_{K'} \|U_h - u_h\|_{H^1(K')}^2.$$
(16)

The existence and uniqueness of \tilde{U}_K follows from the fact that $P^{p_K+1}(\mathcal{D}_K)$ is a complete finite-dimensional space. Examples of patches \mathcal{D}_K corresponding to interior and boundary elements are shown in Fig. 1.

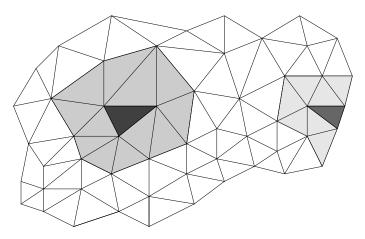


Figure 1: Examples of patches \mathscr{D}_K corresponding to interior and boundary elements.

3.2. Higher-order reconstruction

Let φ_i , $i=1,\ldots,d_K$ be basis function of $P^{p_K+1}(K)$ which can be extended as polynomials to $P^{p_K+1}(\mathscr{D}_K)$. Obviously, $d_K:=\dim P^{p_K+1}(\mathscr{D}_K)=(p_K+2)(p_K+3)/2$. Then problem (16) is equivalent to the linear algebraic system

$$\mathbb{A}\boldsymbol{x} = \boldsymbol{b}, \qquad \mathbb{A} = \{A_{i,j}\}_{i,j=1}^{d_K}, \quad A_{i,j} = \sum_{K' \in \mathscr{D}_K} w_{K'}(\varphi_i, \varphi_j)_{1,K'}, \ i, j = 1, \dots, d_K,$$

$$\boldsymbol{b} = \{b_i\}_{i=1}^{d_K}, \quad b_i = \sum_{K' \in \mathscr{D}_K} w_{K'}(u_h, \varphi_i)_{1,K'}, \ i, j = 1, \dots, d_K,$$

$$(17)$$

where $(\cdot,\cdot)_{1,K'}$ denotes the scalar product generating the H^1 -norm on $K' \in \mathcal{D}_K$ defined by

$$(v, w)_{1,K'} = \int_{K'} (v w + \nabla v \cdot \nabla w) \, \mathrm{d}x.$$

Let $\boldsymbol{x} = (x_1, \dots, x_{d_K})^T$ be the solution of (17) then $\tilde{U}_K = \sum_{i=1}^{d_K} x_i \varphi_i$. Finally, we put

$$\tilde{u}_K := \tilde{U}_K|_K,\tag{18}$$

where \tilde{U}_K is given by (16). Concerning the weights $w_{K'}$, $K' \in \mathcal{D}_K$, we use the values

$$w_{K'} := \begin{cases} 1 & \text{if } K' = K \text{ or } K \text{ and } K' \text{ are neighbours,} \\ \epsilon & \text{otherwise,} \end{cases}$$
 (19)

where $\epsilon > 0$ is a small parameter. In the computations presented here, we use the value $\epsilon = 0.05$. Let us note that if $p_{K'} = p \ \forall K' \in \mathscr{D}_K$, then the numerical experiments shows that it is possible to use $\epsilon = 0$. However, when polynomial degrees are varying on \mathscr{D}_K it is better to add the influence from the triangles sharing only a vertex. Otherwise, the reconstruction is not stable. However, numerical experiments show that the results are not too sensitive to the choice of ε , e.g., the values $\epsilon = 0.02$ or $\epsilon = 0.1$ give similar results.

Finally, we define the higher order reconstruction of $u_h \in S_h^{\mathsf{p}}$ by an element-wise compositions of the restrictions of \tilde{u}_K on $K \in \mathcal{T}_h$, i.e.,

$$\tilde{u}_h \in S_h^{\mathsf{p}+1} : \quad \tilde{u}_h|_K := \tilde{u}_K \quad \forall K \in \mathscr{T}_h,$$
 (20)

where \tilde{u}_K is given by (18).

3.3. Saturation assumption

By (20), we defined higher-order polynomial reconstruction \tilde{u}_h which should better approximate the exact solution than the approximate one u_h . This property is formulated as the *saturation* assumption, which will be verified numerically in Section 4 by two examples.

Assumption 1. Let $u \in V$ be the exact solution of (3) and $u_h \in S_h^p$ be the approximate solution given by (8). Let $\tilde{u}_h \in S_h^{p+1}$ be the higher-order reconstruction defined by (20). Then we assume that

$$e_h := u - u_h \approx \tilde{u}_h - u_h =: \mathcal{E}_h \quad in \ \Omega,$$
 (21)

where e_h denotes the discretization error and \mathcal{E}_h its approximation by the higher-order reconstruction.

The quantity \mathcal{E}_h is used for the definition of the presented hp-adaptation strategy. Moreover, in this paper, we employ (21) also for the estimation of the error. Therefore, we define the global and $element\ error\ indicators$ by

$$\eta := |\mathscr{E}_h|_{H^1(\Omega,\mathscr{T}_h)} \quad \text{and} \quad \eta_K := |\mathscr{E}_h|_{H^1(K)} \quad \forall K \in \mathscr{T}_h.$$
(22)

Obviously, $\eta^2 = \sum_{K \in \mathcal{I}_h} \eta_K^2$. However, generally, η and η_K can be replaced by another error estimators or indicators appearing in the literature.

4. Verification of assumption (21)

In this section we numerically verify the saturation assumption (21) by two (linear and nonlinear) convection-diffusion equations: The first one gives a regular solution with two weak boundary layers and the second one contains a corner singularity. For both problems, we carried out computations using P^p , $p=1,\ldots,6$ polynomial approximation on uniform triangular meshes with mesh spacing h=1/8, h=1/16, h=1/32 and h=1/64.

For each computation, we evaluate the error $e_h = u - u_h$ and its estimate $\mathcal{E}_h = \tilde{u}_h - u_h$ in the broken H^1 -seminorm, and the *effectivity index* given by

$$i^{\text{eff}} := \frac{|\mathscr{E}_h|_{H^1(\Omega,\mathscr{T}_h)}}{|e_h|_{H^1(\Omega,\mathscr{T}_h)}}.$$
(23)

We also plot the local distribution of the error $|e_h|_{H^1(K)}$, $K \in \mathcal{T}_h$ and its estimator $|\mathcal{E}_h|_{H^1(K)}$, $K \in \mathcal{T}_h$ for selected cases. Moreover, we evaluate the *experimental order of convergence* (EOC) with respect to h according to the formulae $|e_h| \approx c h^{\text{EOC}}$.

4.1. Regular problem

We consider the scalar linear convection-diffusion equation (similarly as in [22], [23])

$$-\varepsilon \triangle u - \frac{\partial u}{\partial x_1} - \frac{\partial u}{\partial x_2} = g \quad \text{in } \Omega = (0, 1)^2, \tag{24}$$

where $\varepsilon = 0.1$ is a constant diffusion coefficient. We prescribe a Dirichlet boundary condition on $\partial\Omega$ and the source term g such that the exact solution has the form

$$u(x_1, x_2) = \left(c_1 + c_2(1 - x_1) + e^{-x_1/\varepsilon}\right) \left(c_1 + c_2(1 - x_2) + e^{-x_2/\varepsilon}\right)$$
(25)

with $c_1 = -e^{-1/\varepsilon}$, $c_2 = -1 - c_1$. The solution contains two weak boundary layers along $x_1 = 0$ and $x_2 = 0$, whose width is proportional to ε .

Table 1 shows the achieved results. We observe that except the P_6 -approximation (caused probably by the limits of the finite precision arithmetic), the index i^{eff} is not far from 1. Moreover, we detect something like asymptotic exactness. Furthermore, Fig. 2 shows the distribution of the computational error in the broken H^1 -seminorm and its estimate in the computational domain for selected computations. We observe that $|e_h|_{H^1(K)}$ and $|\mathscr{E}_h|_{H^1(K)}$ have similar distribution for $K \in \mathscr{T}_h$. Therefore, \mathscr{E}_h approximate e_h well also locally.

h	p	$ e_h _{H^1(\Omega,\mathscr{T}_h)}$	EOC	$ \mathscr{E}_h _{H^1(\Omega,\mathscr{T}_h)}$	EOC	$i^{\mathrm{eff}}h$	p	$ e_h _{H^1(\Omega,\mathscr{T}_h)}$	EOC	$ \mathcal{E}_h _{H^1(\Omega,\mathcal{T}_h)}$	EOC	$i^{ m eff}$
1/8	1	5.04E-01	_	3.58E-01	-	0.17/18	4	2.29E-03	_	4.99E-03	-	2.18
1/16	1	2.69E-01	0.91	2.25E-01	0.67	018/46	4	1.64E-04	3.81	3.75E-04	3.73	2.29
1/32	1	1.37E-01	0.97	1.27E-01	0.82	01932	4	1.05E-05	3.95	2.02E-05	4.22	1.91
1/64	1	6.89E-02	0.99	6.78E-02	0.91	01984	4	6.62E-07	3.99	9.58E-07	4.40	1.45
1/8	2	9.46E-02	_	8.14E-02	-	0.868	5	2.78E-04	_	1.22E-03	_	4.40
1/16	2	2.62E-02	1.85	2.44E-02	1.74	01936	5	9.97E-06	4.80	4.95E-05	4.63	4.97
1/32	2	6.76E-03	1.96	6.56E-03	1.90	019/32	5	3.21E-07	4.96	1.31E-06	5.24	4.08
1/64	2	1.71E-03	1.99	1.69E-03	1.96	019964	5	1.01E-08	5.00	2.78E-08	5.56	2.76
1/8	3	1.56E-02	_	1.89E-02	-	1.2/18	6	2.83E-05	_	2.86E-04	_	10.09
1/16	3	2.21E-03	2.82	2.67E-03	2.83	112/16	6	5.09E-07	5.80	6.20E-06	5.53	12.17
1/32	3	2.85E-04	2.95	3.19E-04	3.06	111/22	6	8.20E-09	5.96	8.30E-08	6.22	10.13
1/64	3	3.58E-05	2.99	3.75E-05	3.09	110/54	6	1.28E-10	6.00	8.59E-10	6.59	6.69

Table 1: Regular problem (24) – (25): the computational errors in the broken H^1 -seminorm with their estimates and the effectivity indexes.

4.2. Singular problem

We consider the scalar nonlinear convection-diffusion equation

$$-\nabla \cdot (\boldsymbol{K}(u)\nabla u) - \frac{\partial u^2}{\partial x_1} - \frac{\partial u^2}{\partial x_2} = g \quad \text{in} \quad \omega := (0,1)^2,$$

$$u = u_D \quad \text{on} \quad \partial\Omega,$$
(26)

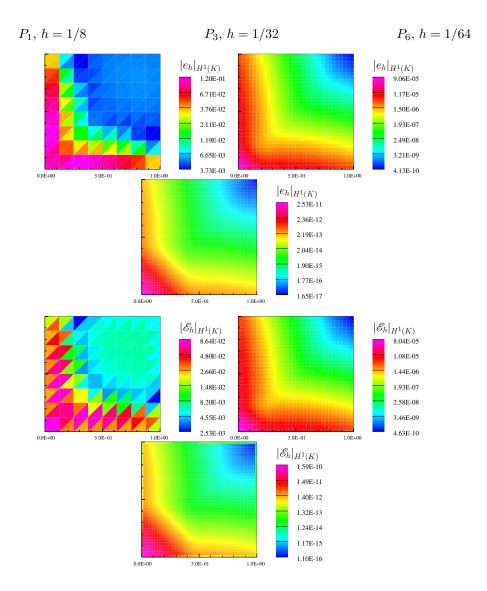


Figure 2: Regular problem (24) – (25): local distribution of the computational error $||e_h||_{H^1(K)}$, $K \in \mathscr{T}_h$ (top) and its estimate $|\mathscr{E}_h|_{H^1(K)}$, $K \in \mathscr{T}_h$ (bottom) for the selected computations.

where K(u) is the nonsymmetric matrix given by

$$\mathbf{K}(u) = \varepsilon \begin{pmatrix} 2 + \arctan(u) & (2 - \arctan(u))/4 \\ 0 & (4 + \arctan(u))/2 \end{pmatrix}. \tag{27}$$

The parameter $\varepsilon > 0$ plays a role of an amount of diffusion and we put $\varepsilon = 10^{-3}$. We prescribe a Dirichlet boundary condition u_D on $\partial\Omega$ and set the source term g such that the exact solution is

$$u(x_1, x_2) = (x_1^2 + x_2^2)^{\alpha/2} x_1 x_2 (1 - x_1)(1 - x_2), \quad \alpha \in \mathbb{R}.$$
 (28)

We put $\alpha = -3/2$. It is possible to show (see [24]) that $u \in H^{\kappa}(\omega)$, $\kappa \in (0, 3/2)$. where $H^{\kappa}(\omega)$ denotes the Sobolev-Slobodetskii space of functions with "non-integer derivatives".

Table 2 shows the achieved results. We observe reasonable values of the indexes i_{L^2} and i_{H^1} , they indicate h-independence and a weak p-dependence. Moreover, Fig. 3 shows the distribution of the computational error in the H^1 -seminorm and its estimate in the computational domain for selected computations. Again, the distributions of $|e_h|_{H^1(K)}$ and $|\mathscr{E}_h|_{H^1(K)}$ are similar for $K \in \mathscr{T}_h$.

Some exception is the case P_6 and h = 1/64, when the error is many time underestimate in regions far from the singularity. However, the errors arising in these regions contribute negligibly to the total error.

h	p	$ e_h _{H^1(\Omega,\mathscr{T}_h)}$	EOC	$ \mathscr{E}_h _{H^1(\Omega,\mathscr{T}_h)}$	EOC	$i^{\mathrm{eff}}\!\hbar$	p	$ e_h _{H^1(\Omega,\mathscr{T}_h)}$	EOC	$ \mathscr{E}_h _{H^1(\Omega,\mathscr{T}_h)}$	EOC	$i^{ m eff}$
1/8	1	3.93E-01	_	2.93E-01	-	0.17/58	4	1.29E-01	_	2.40E-01	_	1.85
1/16	1	2.82E-01	0.48	2.20E-01	0.42	01786	4	9.24E-02	0.49	1.73E-01	0.47	1.87
1/32	1	2.00E-01	0.49	1.60E-01	0.46	018/32	4	6.58E-02	0.49	1.24E-01	0.48	1.89
1/64	1	1.42E-01	0.50	1.14E-01	0.48	018/64	4	4.67E-02	0.49	8.84E-02	0.49	1.89
1/8	2	1.86E-01	_	2.37E-01	-	1.12/78	5	1.23E-01	_	2.30E-01	_	1.87
1/16	2	1.32E-01	0.50	1.72E-01	0.47	113/0.6	5	8.77E-02	0.49	1.66E-01	0.47	1.89
1/32	2	9.38E-02	0.49	1.23E-01	0.48	113/132	5	6.25E-02	0.49	1.19E-01	0.48	1.90
1/64	2	6.65E-02	0.50	8.78E-02	0.49	11 3/B 4	5	4.44E-02	0.49	8.47E-02	0.49	1.91
1/8	3	1.68E-01	_	2.50E-01	-	1.14,98	6	1.27E-01	_	2.30E-01	_	1.81
1/16	3	1.20E-01	0.49	1.82E-01	0.46	115/26	6	9.10E-02	0.48	1.66E-01	0.47	1.82
1/32	3	8.50E-02	0.49	1.30E-01	0.48	11 5/3 2	6	6.49E-02	0.49	1.19E-01	0.48	1.83
1/64	3	6.04E-02	0.49	9.31E-02	0.49	115/64	6	4.62E-02	0.49	8.46E-02	0.49	1.83

Table 2: Singular problem (26) – (28): the computational errors in the broken H^1 -seminorm with their estimates and the effectivity indexes.

5. hp-adaptive strategy

5.1. Isotropic mesh adaptation

The ultimate goal of the computation is to achieve the prescribed error tolerance with the smallest possible number of degrees of freedom (DOF), cf. (5). More precisely, we need to minimize DOF of the hp-mesh $\mathcal{T}_h^{\mathsf{p}}$ such that

$$\eta \le \omega,$$
(29)

where η is a posteriori error estimate given by (22) (or any other estimator) and $\omega > 0$ is the given tolerance. In order to fulfill (29), we require

$$\eta_K \le \omega_K := \omega \sqrt{1/\# \mathscr{T}_h} \quad \forall K \in \mathscr{T}_h,$$
(30)

where η_K is given by (22) and $\# \mathscr{T}_h$ denotes the number of elements of \mathscr{T}_h . Obviously, due to (22),

$$\eta_K \le \omega_K \ \forall K \in \mathscr{T}_h \Longrightarrow \eta \le \omega,$$

hence, the local condition (30) is stronger than the global one (29). However, the great advantage of (30) is the possibility to adapt the whole mesh at once (and not only the elements with the highest error estimates) which saves the computational time. Let us note that in order to achieve the faster convergence in practical computations, we set $\omega_K := C_F \omega \sqrt{1/\# \mathcal{T}_h}$ where $C_F = 1/2$ is the security factor.

Based on the higher-order reconstruction $\tilde{u}_h \in S_h^{p+1}$ introduced in Section 3, we define the adaptive process which modifies the given hp-mesh and creates a new (better) one. Particularly, let h_K be the diameter of $K \in \mathcal{T}_h$ and p_K be the corresponding polynomial approximation degree. Then we set the better degree $p_K^{\text{new}} \in \{p_K - 1, p_K, p_K + 1\}$ and the better diameter $h_K^{\text{new}} > 0$ in such a way that a local tolerance condition (30) will be satisfied with the smallest possible number of degrees of freedom.

Finally, when p_K^{new} and h_K^{new} are set for each $K \in \mathcal{T}_h$, we employ the framework of the isotropic mesh adaptation, which is a simplification of technique developed in [25, 26]. We define an isotropic Riemann metric (= Euclidean metric scaled by $1/h_K^{\text{new}}$). Then, with the aid of inhouse code ANGENER [27], we create new triangulation $\mathcal{T}_h^{\text{new}}$ which is not nested in the relation with \mathcal{T}_h . Moreover, from p_K^{new} , $K \in \mathcal{T}_h$, we define piecewise constant function on $\mathcal{T}_h^{\text{new}}$, which is interpolated into the space of integer-valued piece-wise constant functions on $\mathcal{T}_h^{\text{new}}$.

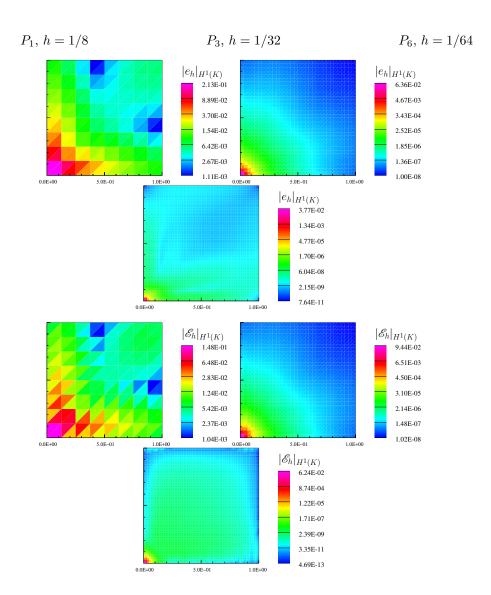


Figure 3: Singular problem (26) – (28): local distribution of the computational error $\|e_h\|_{H^1(K)}$, $K \in \mathscr{T}_h$ (top) and its estimate $|\mathscr{E}_h|_{H^1(K)}$, $K \in \mathscr{T}_h$ (bottom) for the selected computations.

5.2. Main idea of the hp-adaptation

In agreement with (20),

$$\tilde{u}_K = (\tilde{u}_h)|_K, \qquad K \in \mathcal{T}_h$$
 (31)

denotes the restriction of the higher order reconstruction to the mesh elements. We define the value

$$\vartheta_{K,0} := |\tilde{u}_K - u_h|_{H^1(K)}, \qquad K \in \mathscr{T}_h. \tag{32}$$

Let us note that due to (22), we have $\vartheta_{K,0} = \eta_K$. However, we use this duplicity in the notation since η_K represents the estimation of the error (which can be replaced by any other indicator) whereas $\vartheta_{K,0}$ is employed for a decision about p-adaptation.

Obviously, the quantity $\vartheta_{K,0}$ approximate the discretization error achieved on K having the diameter h_K and the polynomial approximation degree p_K . Now we are interested in the following question:

(Q1) What happens if we locally decrease the polynomial approximation degree p_K to $p_K - \ell$, $\ell = 1.2$?

It is natural to expect that the decrease of the polynomial approximation degree leads to an increase of the computational error as well as its estimate $\vartheta_{K,0}$. (Obviously, the number of degrees of freedom on K decreases.) Therefore, in order to keep the local error estimate $\vartheta_{K,0}$, it is necessary to refine the element K into several *sub-elements*, see Fig. 4.

Hence, we replace question (Q1) by a new one which is more quantitative:

(Q2) Let $\vartheta_{K,0}$ be the estimate of the discretization error achieved on K with the polynomial approximation degree p_K . If we replace p_K by $p_K - \ell$, $\ell = 1, 2$, onto how many sub-elements we need to split K such that the corresponding error will be again $\vartheta_{K,0}$?

In order to find the answer to question (Q2), we define, similarly as in (6), the space of piecewise-polynomial functions

$$S_h^{\mathsf{p}-\ell} := \{ v_h \in L^2(\Omega); \ v_h|_K \in P^{p_K-\ell}(K) \ \forall K \in \mathcal{T}_h \}, \quad \ell = 1, 2,$$
 (33)

where $P^{p_K-\ell}(K)$ denotes the space of polynomials of degree $\leq p_K - \ell$ on K (for simplicity we assume that $p_K - \ell \geq 0$). Obviously, $S_h^{\mathsf{p}-\ell} \subset S_h^{\mathsf{p}}$ for $\ell = 1, 2$. Moreover, we denote by $\Pi_h^{\mathsf{p}-\ell}$, $\ell = 1, 2$ the L^2 -projection from S_h^{p} to $S_h^{\mathsf{p}-\ell}$, i.e.,

$$\Pi_h^{\mathsf{p}-\ell} u_h \in S_h^{\mathsf{p}-\ell} : \quad \left(\Pi_h^{\mathsf{p}-\ell} u_h, v_h\right)_{0,\Omega} = (u_h, v_h)_{0,\Omega} \quad \forall v_h \in S_h^{\mathsf{p}-\ell}, \qquad \ell = 1, 2$$
 (34)

Let us note that if we consider a hierarchical basis of S_h^{p} then the projection (34) is straightforward, we simply remove the basis coefficients corresponding to the basis functions of degree greater than $p_K - \ell$ for each $K \in \mathscr{T}_h$.

Now, for $K \in \mathcal{T}_h$, we define two additional quantities

$$\vartheta_{K,\ell} := \left| \tilde{u}_K - \Pi_h^{\mathsf{p}-\ell} u_h \right|_{H^1(K)}, \qquad \ell = 1, 2, \ K \in \mathscr{T}_h, \tag{35}$$

which approximate the local discretization errors in the H^1 -seminorm for the used polynomial approximation degrees $p_K - \ell$ on K. Their evaluation is very simple since \tilde{u}_K is already available due to (20) and (31).

Usually, the quantities $\vartheta_{K,\ell}$, $\ell = 0, 1, 2$ satisfy the inequalities

$$\vartheta_{K,0} \le \vartheta_{K,1} \le \vartheta_{K,2}, \quad K \in \mathscr{T}_h,$$
 (36)

which corresponds to the fact that a higher degree polynomial approximation gives at most the same error as a lower degree one.

Let us assume that the exact solution is sufficiently regular, the non-regular case is discussed later. In agreement with (9), we have

$$\vartheta_{K,\ell} \approx C h_K^{p_K - \ell}, \quad \ell = 0, 1, 2,$$
 (37)

where C is a unknown constant independent of h_K . In order to find the answer to question (Q2), we seek the *diameter* of *sub-elements* of K, denoted by $h_{K,\ell}$, such that the resulting error will by equal to $\vartheta_{K,0}$, i.e.,

$$\vartheta_{K,0} \approx C(h_{K,\ell})^{p_K - \ell}, \quad \ell = 0, 1, 2.$$
 (38)

The above together with (37) gives the ratios

$$q_{\ell} := \frac{h_K}{h_{K,\ell}} = \left(\frac{\vartheta_{K,\ell}}{\vartheta_{K,0}}\right)^{\frac{1}{p_K - \ell}}, \quad \ell = 0, 1, 2.$$

$$(39)$$

Obviously, $q_0 = 1$. Therefore, (theoretically), if we split K onto q_ℓ^2 sub-triangles then the error estimate $\vartheta_{K,\ell}$ reduces to $\vartheta_{K,0}$, see Fig. 4 for an illustration. Generally, the ratios q_1 and q_2 are non-integer but it does not exhibit any obstacle in the framework of the isotropic mesh adaptation.

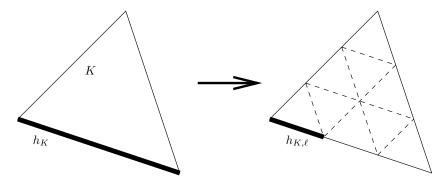


Figure 4: Illustration of the element splitting for $q_{\ell} = 3$ in (39).

If the element K is split onto q_ℓ^2 sub-elements and on each of them the polynomial approximation degree $p_K - \ell$ is used, then corresponding number of degrees of freedom is equal to

$$d_{K,\ell} = q_{\ell}^2 (p_K - \ell + 1)(p_K - \ell + 2)/2, \quad \ell = 0, 1, 2.$$
(40)

Therefore, we have three candidates, $\ell=0,1,2$, which achieve theoretically the level $\vartheta_{K,0}$ with $d_{K,\ell}$ degrees of freedom. Hence, we choose this one which has minimal value $d_{K,\ell}$, i.e., let $\bar{\ell}=\arg\min_{\ell\in\{0,1,2\}}d_{K,\ell}$, then

$$\bar{\ell} = 0 \Rightarrow p_K^{\text{new}} := p_K + 1,
\bar{\ell} = 1 \Rightarrow p_K^{\text{new}} := p_K,
\bar{\ell} = 2 \Rightarrow p_K^{\text{new}} := p_K - 1.$$
(41)

Finally, we consider the case when the exact solution is not regular, particularly when the local polynomial approximation degree is greater than the local Sobolev regularity. Let $u|K \in H^{s_K}(K)$ such that $s_K < p_K$. In this case, instead of (36), the quantities $\vartheta_{K,\ell}$, $\ell = 0, 1, 2$ satisfy the relations

$$\vartheta_{K,0} \approx \vartheta_{K,1} \approx \vartheta_{K,2},\tag{42}$$

which is in agreement with (9) and it corresponds to the observation than an increase of the polynomial approximation degree does not bring any essential increase of the accuracy when the solution is not regular. Therefore, relations (39) and (42) imply $q_1 \approx 1$ and $q_2 \approx 1$. Consequently, relation (40) gives $d_{K,2} < d_{K,1} < d_{K,0}$ and according (41), the algorithm automatically decreases p_K .

5.3. hp-adaptive algorithm

end for

Based on the consideration formulated in Section 5.2, we formulate the hp-adaptive algorithm. Let $\omega > 0$ be the prescribed tolerance and ω_K be the local error tolerance defined by (30). The presented algorithm is based on the following two steps:

- (S1) according (41), we set the better polynomial degree,
- (S2) we propose new (more optimal) size of K according

$$h_K^{\text{new}} = h_K/\rho, \quad \text{where } \rho := (\eta_K/\omega_K)^{1/p_K}.$$
 (43)

The formulas (43) follow from relations $\eta_K \approx C h_K^{p_K}$ and $\omega_K \approx C (h_K^{\rm new})^{p_K}$, cf. (37). In order to avoid a strong refinement in one level of mesh adaptation, we restrict $\rho \leq 5$. Furthermore, the algorithm contains two branches depending if the local tolerance condition (30) is satisfied or not. If (30) is violated and if we increase p_K according (41) then we do not change element size according to (43) since then we avoid a needless h-refinement at the beginning of mesh adaptations. Hence, we define the following algorithm:

Algorithm 1 setting of h_K^{new} and p_K^{new} for $K \in \mathcal{T}_h$

```
let u_h \in S_h^{\mathsf{p}} be given
for all K \in \mathcal{T}_h do
   let h_K = \operatorname{diam}(K) and p_K be the corresponding polynomial approximation degree,
   compute the higher-order reconstruction \tilde{u}_K by (16) and (18),
   set \eta_K by (22) (or by another error estimate),
   set \omega_K by (30),
   set \vartheta_{K,\ell}, \ell = 0, 1, 2 by (35),
   set d_{K,\ell}, \ell = 0, 1, 2 by (40),
   set \bar{\ell} for which d_{K,\ell} is the minimal
   if \eta_K > \omega_K then
      if \bar{\ell} = 0 then
          p_K^{\text{new}} := p_K + 1,
                                     \rho := 1
       else if \bar{\ell} = 1 then
      \begin{array}{c} p_K^{\rm new} := p_K \ , \qquad \rho := (\eta_K/\omega_K)^{1/p_K} \\ \text{else if} \ \ \bar{\ell} = 2 \ \ \text{then} \end{array}
          p_K^{\mathrm{new}} := p_K - 1, \qquad 
ho := (\eta_K/\omega_K)^{1/(p_K - 1)}
       end if
   else
      if \ell = 0 then
          p_K^{\text{new}} := p_K + 1, \rho := (\eta_K/\omega_K)^{1/(p_K + 1)}
       else if \bar{\ell} = 1 then
          p_K^{\text{new}} := p_K, \qquad \rho := (\eta_K/\omega_K)^{1/p_K}
       else if \bar{\ell} = 2 then
          p_{K}^{\text{new}} := p_K - 1, \qquad \rho := (\eta_K/\omega_K)^{1/(p_K - 1)}
       end if
   end if
   h_K^{\text{new}} := \frac{h_K}{\min(\rho, 5)}
```

This algorithm works locally for each $K \in \mathcal{T}_h$, hence it can be simply parallelized. However, even for one computer core, the total computational time of the mesh adaptation including the higher-order reconstruction and the creation of the new hp-grid is shorter than the computational time necessary for the assembling and solving the nonlinear algebraic system (8). Moreover, the presented algorithm does not contain any empirical constant.

The use of Algorithm 1 is obvious: We start with an initial hp-mesh, solve the approximate problem (8) and by Algorithm 1 we generate a new hp-mesh where we solve problem (8) again, The combination of the problem solution and mesh adaptation is repeated until the condition (29) is achieved.

6. Numerical examples

In this section we demonstrate the computational performance of the proposed hp-adaptive algorithm. We present three examples which are (except the first one) modifications of problems from [28] where a collection of 2D elliptic problems for testing adaptive grid refinement algorithms was published. Our aim is to demonstrate the exponential rate of the convergence of the error with respect to DOF and also to show a reasonable ability of the error estimator (22) to approximate the error. Moreover, the selection of the examples (Laplace problem, linear convection-diffusion equations, quasilinear elliptic equation) indicate the robustness of this approach.

For each case, we present the convergence of the adaptive process, namely the values of triangles $\#\mathscr{T}_h$ of the generated grids, the corresponding DOF given by (5), the computational errors in the broken H^1 -seminorm, their estimates $|\mathscr{E}_h|_{H^1(\Omega,\mathscr{T}_h)}$ and the effectivity index i^{eff} given by (23).

Moreover, we investigate the experimental order of convergence (EOC) with respect to DOF according to the formulae $e_h \approx c \, \text{DOF}^{-\text{EOC}}$. In some situation, the EOC may be negative, namely when the algorithm decreases DOF as well es the error and its estimates. This is in fact the advantage of our approach.

6.1. Case 1: re-entrant corner singularity

We consider the Poisson problem

$$-\Delta u = 0 \quad \text{in } \Omega := (-1, 1)^2 \setminus \{(x_1, x_2); -2x_1 < x_2 < 0\},$$

$$u = u_D \quad \text{on } \partial\Omega,$$
(44)

where u_D is chosen such that the exact solution is

$$u(r,\varphi) = r^{2/3}\sin(2\varphi/3). \tag{45}$$

Here (r,φ) are the polar coordinates. This problems has a corner singularity at the origin.

We applied Algorithm 1 with $\omega = 10^{-4}$. Table 3 shows the convergence of the adaptive process. We observe that Algorithm 1 gives exponential order of convergence, which that the decrease of the error is faster than any linear decrease in logarithmic scale, see the corresponding figure of data from Table 3. Moreover, the effectivity index i^{eff} is very close to 1 which means that the used "naive" error estimator η defined by (22) approximate error very well.

Furthermore, Fig. 5 shows the hp-grids with several details for selected levels of adaptations. We observed a strong h-refinement in a small neighbourhood of the corner, which is necessary for a decrease of the error under the given tolerance.

6.2. Case 2: Wave front

We consider the convection diffusion equation

$$-2\varepsilon \frac{\partial^2 u}{\partial x_1^2} - \frac{\varepsilon}{2} \frac{\partial^2 u}{\partial x_2^2} - x_2 \frac{\partial u}{\partial x_1} + x_1 \frac{\partial u}{\partial x_2} = g \quad \text{in } \Omega := (-1, 1)^2, \tag{46}$$

$$u = u_D \quad \text{on } \partial\Omega,$$
 (47)

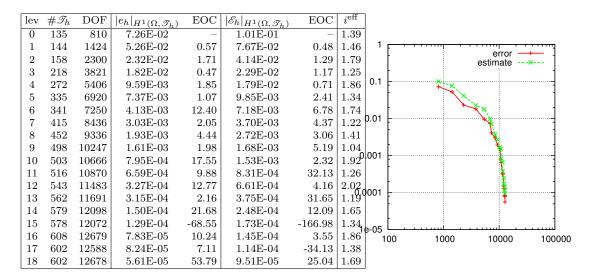


Table 3: Case 1, hp-adaptive computation: the errors in the broken H^1 -seminorm, their estimates and corresponding EOC and i^{eff} , numerical values (left) and the corresponding plot (right).

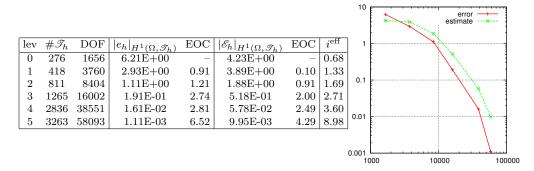


Table 4: Case 2, hp-adaptive computation: the errors in the broken H^1 -seminorm, their estimates and corresponding EOC and i^{eff} , numerical values (left) and the corresponding plot (right).

where the right-hand side g and the boundary condition u_D are prescribed such that the exact solution is

$$u(x_1, x_2) = \tan^{-1}(m(r - r_0)), \tag{48}$$

where $r = (x_1^2 + x_2^2)^{1/2}$, m = 50 and $r_0 = 0.25$. Moreover, we put The solution that has a steep wave front in the interior of the domain. Due to \tan^{-1} function, there is a also mild singularity at the center of the circle.

We applied Algorithm (1) with $\omega=10^{-2}$. Table 4 shows the exponential convergence of the adaptive process, the EOC is increasing with respect to levels of adaptation. On the other hand, the effectivity index i^{eff} is increasing for decreasing error. This is caused by the fact that patches \mathscr{D}_K , $K\in\mathscr{T}_h$ are much larger in comparison with the width of the interior layer and then the higher-order reconstruction is smeared. Moreover, Fig. 6 shows the hp-grids with several details for selected levels of adaptations. We observe a strong h-refinement along the steep wave front and also around the singularity at the center of the circle. On the other hand, the effectivity index i_{H^1} is increasing.

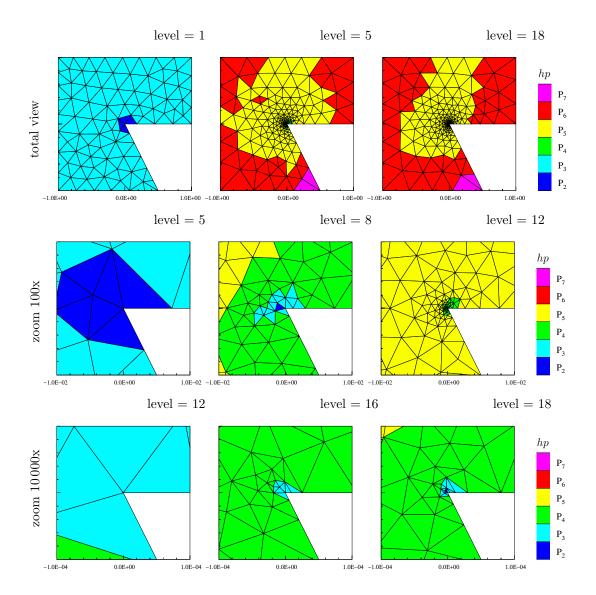


Figure 5: Case 1: hp-grids with details around origin for selected levels of adaptations.

6.3. Case 3: Interior line singularity

We consider the scalar nonlinear diffusion equation

$$-\nabla \cdot (\boldsymbol{K}(u)\nabla u) = g \quad \text{in } \Omega := (-1,1)^2,$$

$$u = u_D \quad \text{on } \partial\Omega,$$
(49)

where K(u) is given by (27) with $\varepsilon = 1$. We prescribe a Dirichlet boundary condition u_D on $\partial\Omega$ and set the source term g such that the exact solution is

$$u(x_1, x_2) = \begin{cases} \cos(\pi x_2/2) & \text{for } x_1 \le \beta(x_2 - 1), \\ \cos(\pi x_2/2) + (x_1 - \beta(x_2 - 1))^{\alpha} & \text{for } x_1 > \beta(x_2 - 1), \end{cases}$$
(50)

where we put $\alpha = 2$ and $\beta = 0.6$. The solution satisfies $u \in H^{\alpha+1/2-\epsilon}(\Omega)$ and posses a weak singularity along the line $x_1 - \beta(x_2 - 1) = 0$. This line singularity is difficult to capture since it is very weak. However, without a sufficient refinement along this line, it is not possible to decrease the computational error under the given tolerance.

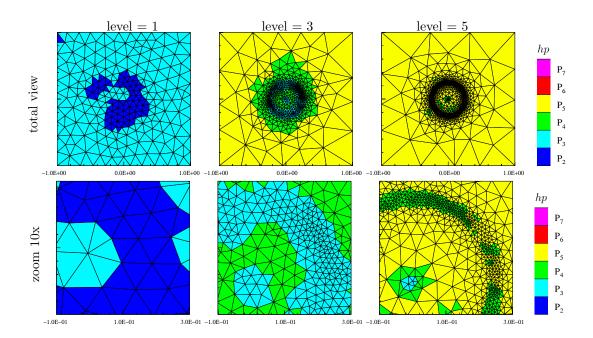


Figure 6: Case 2: hp-grids with details containing the origin singularity and the part of the wave front for selected levels of adaptations.

									error + estimatex
lev	$\#\mathscr{T}_h$	DOF	$ e_h _{H^1(\Omega,\mathscr{T}_h)}$	EOC	$ \mathscr{E}_h _{H^1(\Omega,\mathscr{T}_h)}$	EOC	i^{eff}		
0	586	3516	7.06E-03	_	8.30E-03	_	1.17	0.001	
1	589	5890	2.99E-03	1.67	5.15E-03	0.92	1.72		<u> </u>
2	764	10480	1.28E-03	1.48	2.05E-03	1.60	1.61		\\\ \ \
3	792	10992	5.05E-04	19.43	8.54E-04	18.39	1.69		
4	1344	17021	1.82E-04	2.33	3.26E-04	2.20	1.79	0.0001	
5	2051	25172	7.93E-05	2.12	1.41E-04	2.14	1.78		\\\\\\\\\\\\\\\\\\\\\\\\\\\\\\\\\\\\\\
6	2774	32959	4.31E-05	2.26	7.57E-05	2.31	1.75		† †
									1
								1e-05	
								10	00 10000 100000

Table 5: Case 3, hp-adaptive computation: the errors in the broken H^1 -seminorm, their estimates and corresponding EOC and i^{eff} , numerical values (left) and the corresponding plot (right).

We applied Algorithm (1) with $\omega=10^{-4}$. Table 5 shows the convergence of the adaptive process, however, it is not exponential but the given tolerance is achieved within few adaptive levels. On the other hand, the effectivity index $i^{\rm eff}$ indicates an accurate estimation of the error. Moreover, Fig. 7 shows the hp-grids with several details for selected levels of adaptations. We observe a h-refinement along the line singularity, which is spread only to few elements in the direction perpendicular to this line.

7. Conclusion

We presented a new hp-adaptive technique for the numerical solution of boundary value problems. The method is based on higher-order reconstruction over local element patches which makes it faster and easy to parallelize. The presented algorithm is free of user-defined parameters, it can be generalized for any numerical method using polynomial approximation, and combined with any a posteriori error estimator. Several numerical experiments were performed to demonstrate the outstanding performance and robustness of the presented method.

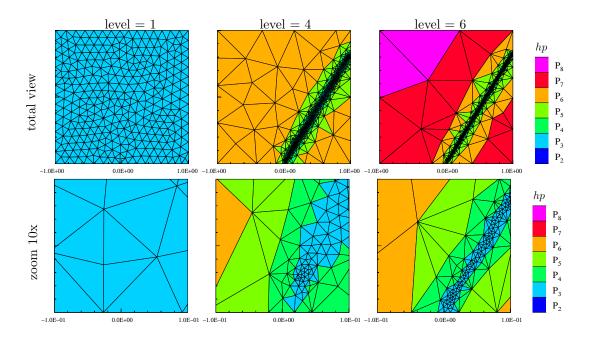


Figure 7: Case 3: hp-grids with details containing detail along the interior layer for selected levels of adaptations.

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