ADAPTIVE VERTEX-CENTERED FINITE VOLUME METHODS
FOR GENERAL SECOND-ORDER LINEAR ELLIPTIC PDES

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ABSTRACT. We prove optimal convergence rates for the discretization of a general second-order linear elliptic PDE with an adaptive vertex-centered finite volume scheme. While our prior work Erath and Praetorius [SIAM J. Numer. Anal., 54 (2016), pp. 2228–2255] was restricted to symmetric problems, the present analysis also covers non-symmetric problems and hence the important case of present convection.

1. Introduction

In this work we consider a general second-order linear elliptic PDE and approximate the solution with an adaptive vertex-centered finite volume method (FVM). Finite volume methods are well established in fluid mechanics, since they naturally preserve numerical flux conservation.

1.1. Model problem. Let $\Omega \subset \mathbb{R}^d$, $d = 2, 3$, be a bounded Lipschitz domain with polygonal boundary $\Gamma := \partial \Omega$. As model problem, we consider the following stationary diffusion problem: Given $f \in L^2(\Omega)$, find $u \in H^1(\Omega)$ such that

$$
\text{div}(-A\nabla u + bu) + cu = f \quad \text{in } \Omega \quad \text{and} \quad u = 0 \quad \text{on } \Gamma.
$$

We suppose that the diffusion matrix $A = A(x) \in \mathbb{R}^{d \times d}$ is bounded, symmetric, and uniformly positive definite, i.e., there exist constants $\lambda_{\text{min}}, \lambda_{\text{max}} > 0$ such that

$$
\lambda_{\text{min}} |v|^2 \leq v^T A(x) v \leq \lambda_{\text{max}} |v|^2 \quad \text{for all } v \in \mathbb{R}^d \text{ and almost all } x \in \Omega.
$$

Let $T_0$ be a given initial triangulation of $\Omega$; see Section 2.2 below. For convergence of FVM and well-posedness of the residual error estimator, we additionally require that $A(x)$ is piecewise Lipschitz continuous, i.e.,

$$
A \in W^{1,\infty}(T)^{d \times d} \quad \text{for all } T \in T_0.
$$

We suppose that the lower-order terms satisfy the assumption

$$
b \in W^{1,\infty}(\Omega)^d \quad \text{and} \quad c \in L^\infty(\Omega) \quad \text{with} \quad \frac{1}{2} \text{div } b + c \geq 0 \quad \text{almost everywhere on } \Omega.
$$

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With \((\phi, \psi)_{\Omega} = \int_{\Omega} \phi(x) \psi(x) \, dx\) being the \(L^2\)-scalar product, the weak formulation of the model problem (1) reads as follows: Find \(u \in H^1_0(\Omega)\) such that

\[
A(u, w) := (A \nabla u - bu, \nabla w)_{\Omega} + (cu, w)_{\Omega} = (f, w)_{\Omega} \quad \text{for all } w \in H^1_0(\Omega).
\]

According to our assumptions (2)–(4), the bilinear form \(A(\cdot, \cdot)\) is continuous and elliptic on \(H^1_0(\Omega)\). Existence and uniqueness of the solution \(u \in H^1_0(\Omega)\) of (5) thus follow from the Lax-Milgram theorem. Moreover, the operator induced quasi norm \(||\cdot||\) satisfies that

\[
C_{\text{ell}} \|v\|_{H^1(\Omega)}^2 \leq \|v\|^2 := A(v, v) \leq C_{\text{cont}} \|v\|_{H^1(\Omega)}^2 \quad \text{for all } v \in H^1_0(\Omega),
\]

where \(C_{\text{ell}} > 0\) depends only on \(\lambda_{\min}\) and \(\Omega\), whereas \(C_{\text{cont}} > 0\) depends only on \(\lambda_{\max}\), \(\|b\|_{L^\infty(\Omega)}\) and \(\|c\|_{L^\infty(\Omega)}\).

### 1.2. Adaptive FVM.

In the past 20 years, there have been major contributions to the mathematical understanding of adaptive mesh refinement algorithms, mainly in the context of the finite element method (FEM). While the seminal works [Dör96, MNS00, BDD04, Ste07, CKNS08] were restricted to symmetric operators, the recent works [MN05, CN12, FFP14, BHP17] proved convergence of adaptive FEM with optimal algebraic rates for general second order linear elliptic PDEs. The work [CFPP14] gives an exhausted overview of the developments and it gains, in an abstract framework, a general recipe to prove optimal adaptive convergence rates of adaptive mesh refining algorithms. Basically, the numerical discretization scheme, the \textit{a posteriori} error estimator and the adaptive algorithm have to fulfill four criteria (called \textit{axioms} in [CFPP14]), namely, \textit{stability on non-refined elements}, \textit{reduction on refined elements}, \textit{general quasi-orthogonality}, and \textit{discrete reliability}. Building upon these findings, our recent work [EP16] gave the first proof of convergence of adaptive FVM with optimal algebraic rates for a symmetric model problem (1) with \(b = 0\) and \(c = 0\).

### 1.3. Contributions and outline.

In this work, we are in particular interested in the non-symmetric model problem with \(b \neq 0\) in (1). The proofs of \textit{stability on non-refined elements}, \textit{reduction on refined elements}, and \textit{discrete reliability} follow basically the proofs in [EP16]; see Section 3.3. Thus, the major contribution of the present work is the proof of the \textit{general quasi-orthogonality} property for the non-symmetric problem, which is satisfied under some mild regularity assumptions on the dual problem. Similar assumptions are required in [MN05, CN12] to prove convergence for an adaptive FEM procedure. Moreover, we note that [MN05, CN12] require slightly more restrictions on the model data (namely, \(\text{div}(b) = 0\)) and on the mesh-refinement (the so-called \textit{interior node property}) for proving quasi-orthogonality which are avoided in the present analysis.

At this point, we note that [FFP14, BHP17] improve the FEM result of [MN05, CN12] by a different approach. Instead of the duality argument, the analysis exploits the \textit{a priori} convergence of FEM solutions (which follows from the classical Céa lemma) by splitting the operator into a symmetric and elliptic part and a compact perturbation. In particular, there is no duality argument applied and, therefore, no additional regularity assumption is required. However, it seems to be difficult to transfer the analysis of [FFP14, BHP17] to FVM due to the lack of the Céa lemma.

We also mention that unlike the FEM literature, a direct proof of the \textit{general quasi-orthogonality} is not available for FVM due to the lack of Galerkin orthogonality. Instead, the FVM work [EP16] first proves linear convergence which relies on a quasi-Galerkin orthogonality [EP16, Lemma 11] for FVM. Unfortunately, this auxiliary result does not hold for non-symmetric problems.
Hence, to handle the non-symmetric case, the missing Galerkin orthogonality and the lack of an optimal $L^2$ estimate for FVM seem to be the bottlenecks. To overcome these difficulties, we first estimate the FVM error in the bilinear form by oscillations in Lemma 9. Then we provide a new $L^2$-type estimate in Lemma 10 which depends on the regularity of the corresponding dual problem plus oscillations. These two results provide the key arguments to prove a quasi-Galerkin orthogonality in Proposition 8. Unlike the literature, this estimate also includes a mesh-size weighted estimator term. With the aid of the previous results, we show linear convergence in Theorem 12, where the proof relies on the previous results. Finally, optimal algebraic convergence rates are guaranteed by Theorem 15 which follows directly from the literature.

We remark that the proposed Algorithm 7 additionally marks oscillations to overcome the lack of classical Galerkin orthogonality. Note that this is not required for adaptive FEM. However, since FVM is not a best approximation method, the proposed approach appears to be rather natural. In practice, however, this additional marking is negligible; see also [EP16, Remark 12]. Overall, the present work seems to be the first which proves convergence with optimal rates of an adaptive FVM algorithm for the solution of general second-order linear elliptic PDEs.

2. Preliminaries

This section introduces the notation, the discrete scheme, as well as the residual a posteriori error estimator. In particular, we fix our notation used throughout this work.

2.1. General notation. Throughout, $n$ denotes the unit normal vector to the boundary pointing outward the respective domain. In the following, we mark the mesh dependency of quantities by appropriate indices, e.g., $u_i$ is the solution on the triangulation $\mathcal{T}_i$. Furthermore, $\lesssim$ abbreviates $\leq$ up to some (generic) multiplicative constant which is clear from the context.

2.2. Triangulations. The FVM relies on two partitions of $\Omega$, the primal mesh $\mathcal{T}_x$ and the associated dual mesh $\mathcal{T}_x^*$. The primal mesh $\mathcal{T}_x$ is a regular triangulation of $\Omega$ into non-degenerate closed triangles/tetrahedra $T \in \mathcal{T}_x$, where the possible discontinuities of the coefficient matrix $A$ are aligned with $\mathcal{T}_x$. Define the local mesh-size function

$$ h_x \in L^\infty(\Omega), \quad h_x|_T := h_T := |T|^{1/d} \quad \text{for all } T \in \mathcal{T}_x. $$

Let $\text{diam}(T)$ be the Euclidean diameter of $T$. Suppose that $\mathcal{T}_x$ is $\sigma$-shape regular, i.e.,

$$ \max_{T \in \mathcal{T}_x} \frac{\text{diam}(T)}{|T|^{1/d}} \leq \sigma < \infty. $$

Note that this implies $h_T \leq \text{diam}(T) \leq \sigma h_T$. Let $\mathcal{N}_x$ (or $\mathcal{N}_x^\Omega$) denote the set of all (or all interior) nodes. Let $\mathcal{F}_x$ (or $\mathcal{F}_x^\Omega$) denote the set of all (or all interior) facets. For $T \in \mathcal{T}_x$, let $\mathcal{F}_T := \{ F \in \mathcal{F}_x : F \subseteq \partial T \}$ be the set of facets of $T$. Moreover,

$$ \omega_x(T) := \bigcup \{ T' \in \mathcal{T}_x : T \cap T' \neq \emptyset \} \subseteq \overline{\Omega} $$

denotes the element patch of $T$ in $\mathcal{T}_x$.

The associated dual mesh $\mathcal{T}_x^*$ is obtained as follows: For $d = 2$, connect the center of gravity of an element $T \in \mathcal{T}_x$ with the midpoint of an edge of $\partial T$. These lines define the non-degenerate closed polygons $V_i \in \mathcal{T}_x^*$; see Figure 1(a). For $d = 3$, we first connect the center of gravity of $T \in \mathcal{T}_x$ with each center of gravity of the four faces of $F \in \mathcal{F}_T$ by straight lines. Then, as in the 2D case, we connect each center of gravity of $F \in \mathcal{F}_T$ to
Figure 1. Construction of the dual mesh $T^\ast_x$ from the primal mesh $T_x$ in 2D (left) and 2D newest vertex bisection (right). Each triangle has a reference edge (indicated by the double line). If edges are marked for refinement (indicated by dots), the resulting configurations are shown.

2.3. Discrete spaces. For a partition $M$ of $\Omega$ and $p \in \mathbb{N}_0$, let

$\mathcal{P}^p(M) := \{v : \Omega \to \mathbb{R} : \forall M \in M \ v|_M \text{ is polynomial of degree } \leq p\}$

be the space of $M$-piecewise polynomials of degree $p$. With this at hand, let

$S^1(T_x) := \mathcal{P}^1(T_x) \cap H_0^1(\Omega) = \{v_x \in C(\Omega) : \forall T \in T_x \ v_x|_T \text{ is affine}\}$

Then the discrete ansatz space

$S_0^1(T_x) := S^1(T_x) \cap H_0^1(\Omega) = \{v_x \in S^1(T_x) : v_x|_\Gamma = 0\}$

consists of all $T_x$-piecewise affine and globally continuous functions which are zero on $\Gamma$. By convention, the discrete test space

$\mathcal{P}_0^0(T^\ast_x) := \{v_x^\ast \in \mathcal{P}^0(T^\ast_x) : v_x^\ast|_\Gamma = 0\}$

consists all $T^\ast_x$-piecewise constant functions which are zero on all $V \in T^\ast_x$ with $\partial V \cap \Gamma \neq \emptyset$.

2.4. Mesh-refinements. For local mesh-refinement, we employ newest vertex bisection (NVB); see [Ste08, KPP13] and Figure 1(b). Below, we use the following notation: First, $T' := \text{refine}(T, M)$ denotes the coarsest conforming triangulation generated by NVB from a conforming triangulation $T$ such that all marked elements $M \subseteq T$ have been refined, i.e., $M \subseteq T \setminus T'$. Second, we simply write $T' \in \text{refine}(T)$, if $T'$ is an arbitrary refinement of $T$, i.e., there exists a finite number of refinements steps $j = 1, \ldots, n$ such that $T' = T'_n$ can be generated from $T = T'_0$ with marked elements $M'_j \subseteq T'_j$ and $T'_j = \text{refine}(T'_{j-1}, M'_j)$. Note that NVB guarantees that there exist only finitely many shapes of triangles and patches in $T' \in \text{refine}(T)$. These shapes are determined by $T$. In particular, the meshes $T' \in \text{refine}(T)$ are uniformly $\sigma$-shape regular (8), where $\sigma$ depends only on $T$. 

(a) $T_x$ (triangles) and $T^\ast_x$ (grey boxes). 
(b) Refinement of a triangle by newest vertex bisection (NVB).
2.5. Vertex-centered finite volume method (FVM). The FVM approximates the solution \( u \in H^1_0(\Omega) \) of (5) by some \( u_\tau \in S_0^1(\mathcal{T}_\tau) \). The scheme is based on the balance equation over \( \mathcal{T}_\tau \) and reads in variational form as follows: Find \( u_\tau \in S_0^1(\mathcal{T}_\tau) \) such that

\[
(14) \quad A_\tau(u_\tau, w_\tau^*) = (f, w_\tau^*)_\Omega = \sum_{a_i \in N^a_\tau} w_\tau^*|_{a_i} \int_{V_i} f \, dx \quad \text{for all } w_\tau^* \in P_0^0(\mathcal{T}_\tau^*).
\]

For all \( v_\tau \in S_0^1(\mathcal{T}_\tau) \) and all \( w_\tau^* \in P_0^0(\mathcal{T}_\tau^*) \), the bilinear form reads

\[
A_\tau(v_\tau, w_\tau^*) := \sum_{a_i \in N^a_\tau} w_\tau^*|_{a_i} \left( \int_{\partial V_i} (-A \nabla v_\tau \cdot n \, ds + \int_{V_i} cv_\tau \, dx) \right).
\]

To recall that the FVM is well-posed on sufficiently fine triangulations \( \mathcal{T}_\tau \), we require the following interpolation operator; see, e.g., \([\text{Era12, EP16}]\).

**Lemma 1.** With \( \chi^*_i \in P_0^0(\mathcal{T}_\tau^*) \) being the characteristic function of \( V_i \in \mathcal{T}_\tau^* \), define

\[
I^*_\tau : C(\overline{\Omega}) \to P_0^0(\mathcal{T}_\tau^*), \quad I^*_\tau v := \sum_{a_i \in N^a_\tau} v(a_i) \chi^*_i.
\]

Then, for all \( T \in \mathcal{T}_\tau \), \( F \in \mathcal{F}_T \), and \( v_\tau \in S_1^1(\mathcal{T}_\tau) \), it holds that

\[
(15) \quad \int_T (v_\tau - I^*_\tau v_\tau) \, dx = 0 = \int_F (v_\tau - I^*_\tau v_\tau) \, ds,
\]

\[
(16) \quad \|v_\tau - I^*_\tau v_\tau\|_{L^2(T)} \leq h_T \|\nabla v_\tau\|_{L^2(T)}(T),
\]

\[
(17) \quad \|v_\tau - I^*_\tau v_\tau\|_{L^2(F)} \leq C h_T^{1/2} \|\nabla v_\tau\|_{L^2(T)}.
\]

In particular, it holds that \( I^*_\tau v_\tau \in P_0^0(\mathcal{T}_\tau^*) \) for all \( v_\tau \in S_0^1(\mathcal{T}_\tau) \). The constant \( C > 0 \) depends only on \( \sigma \)-shape regularity of \( \mathcal{T}_\tau \). \( \square \)

The following lemma is a key observation for the FVM analysis. For Lipschitz continuous \( A \), the proof is found in \([\text{ELL02, Era12}]\). We note that the result transfers directly to the present situation \([\text{EP16, EP17}]\), where \( A \) satisfies (2)–(3) and \( b \neq 0 \) and \( c \neq 0 \).

**Lemma 2.** There exists \( C_{\text{bil}} > 0 \) such that for all \( v_\tau, w_\tau \in S_0^1(\mathcal{T}_\tau) \)

\[
(18) \quad |A(v_\tau, w_\tau) - A_\tau(v_\tau, I^*_\tau w_\tau)| \leq C_{\text{bil}} \sum_{T \in \mathcal{T}_\tau} h_T \|v_\tau\|_{H^1(\Omega)} \|w_\tau\|_{H^1(\Omega)}.
\]

Moreover, let \( \mathcal{T}_\tau \) be sufficiently fine such that \( C_{\text{ell}} - C_{\text{bil}} \|h\|_{L^\infty(\Omega)} > 0 \), where \( C_{\text{ell}} > 0 \) is the ellipticity constant from (6). Then there exists \( C_{\text{stab}} > 0 \) such that

\[
(19) \quad A_\tau(v_\tau, I^*_\tau v_\tau) \geq C_{\text{stab}} \|v_\tau\|^2_{H^1(\Omega)} \quad \text{for all } v_\tau \in S_0^1(\mathcal{T}_\tau).
\]

In particular, the FVM system (14) admits a unique solution \( u_\tau \in S_0^1(\mathcal{T}_\tau) \). The constants \( C_{\text{bil}} \) and \( C_{\text{stab}} \) depend only on the data assumptions (2)–(4), the \( \sigma \)-shape regularity of \( \mathcal{T}_\tau \), and \( \Omega \). \( \square \)

2.6. Weighted-residual a posteriori error estimator. For all \( v_\tau \in S_0^1(\mathcal{T}_\tau) \), we define the volume residual \( R_\tau \) and the normal jump \( J_\tau \) by

\[
(20) \quad R_\tau(v_\tau)|_T := f - \nabla v_\tau \cdot (-A \nabla v_\tau + b v_\tau) - c v_\tau \quad \text{for all } T \in \mathcal{T}_\tau,
\]

\[
(21) \quad J_\tau(v_\tau)|_F := [A \nabla v_\tau]_F \quad \text{for all } F \in \mathcal{F}_\tau.
\]
Here, $\text{div}_x$ denotes the $T_x$-piecewise divergence operator, and the normal jump reads 
$[g]_F := (g|_T - g|_{T'}) \cdot n$, where $g|_T$ denotes the trace of $g$ from $T$ onto $F$ and $n$ points from $T$ to $T'$. Let $\Pi_x$ be the edgewise or elementwise integral mean operator, i.e.,

$$(\Pi_x)_\tau = \frac{1}{|\tau|} \int_{\tau} v \, dx \quad \text{for all } \tau \in T_x \cup F_x \text{ and all } v \in L^2(\tau).$$

For all $T \in T_x$, we define the local error indicators and oscillations by

$$\eta_x(T, v_x)^2 := h_T^2 \| R_x(v_x) \|_{L^2(T)}^2 + h_T \| J_x(v_x) \|_{L^2(\partial T \setminus \Gamma)}^2,$$

$$\text{osc}_x(T, v_x)^2 := h_T^2 \| (1 - \Pi_x)(R_x(v_x)) \|_{L^2(T)}^2 + h_T \| (1 - \Pi_x)J_x(v_x) \|_{L^2(\partial T \setminus \Gamma)}^2.$$ 

Then the error estimator $\eta_x$ and the oscillations $\text{osc}_x$ are defined by

$$\eta_x(v_x)^2 := \sum_{T \in T_x} \eta_x(T, v_x)^2 \quad \text{and} \quad \text{osc}_x(v_x)^2 := \sum_{T \in T_x} \text{osc}_x(T, v_x)^2.$$ 

To abbreviate notation, we write $\eta_x := \eta_x(u_x)$ and $\text{osc}_x := \text{osc}_x(u_x)$. The following proposition is proved, e.g., in [CLT05, Era13];

**Proposition 3** (reliability and efficiency). The residual error estimator $\eta_x$ satisfies

$$C_{\text{rel}}^{-1} \| u - u_x \|_{H^1(\Omega)} \leq \eta_x^2 \leq C_{\text{eff}} (\| u - u_x \|_{H^1(\Omega)} + \text{osc}_x^2),$$

where $C_{\text{rel}}, C_{\text{eff}} > 0$ depend only on the $\sigma$-shape regularity of $T_x$, the data assumptions (2)-(4), and $\Omega$. \hfill $\square$

Note that a robust variant of this estimator with respect to an energy norm is found and analyzed in [Era13, Theorem 4.9, Theorem 6.3, and Remark 6.1], where we additionally require the assumption $\| \text{div} \, b + \ell \|_{L^\infty(\Omega)} \leq C (\| \text{div} \, b + \ell \|_{L^\infty(\Omega)} + \text{osc}_x)$, with $C > 0$. One of the key ingredients of Proposition 3 is (25) of the following lemma which will be employed below. The proof of the orthogonality relation (25) is well-known and found, e.g., in [CLT05, Era10, Era13], whereas (26) is proved in [EP16, Lemma 16] for arbitrary refinement of meshes and can easily be transferred to the present model problem (1).

**Lemma 4**. Let $T_0 \in \text{refine}(T_0)$ and $T_x \in \text{refine}(T_0)$. Suppose that the discrete solutions $u_x \in S_0^h(T_x)$ or $u_x \in S_0^h(T_x)$ exist. Then there holds the $L^2$-orthogonality

$$\sum_{T \in T_0} (R_0(u_0), v_x^*)_T - \sum_{F \in F^0} (J_0(u_0), v_x^*)_F = 0 \quad \text{for all } v_x^* \in \mathcal{P}_0^0(T^*)$$

as well as the discrete defect identity

$$\sum_{T \in T_0} (R_0(u_0), v_x^*)_T - \sum_{F \in F^0} (J_0(u_0), v_x^*)_F = \mathcal{A}_x(u_x - u_0, v_x^*) \quad \text{for all } v_x^* \in \mathcal{P}_0^0(T^*_x). \hfill \square$$

**2.7. Comparison result and a priori error estimate.** The following proposition states that the FVM error estimator is equivalent to the optimal total error (i.e., error plus oscillations) and so improves Proposition 3. The result is first proved in [EP16] for $b = 0$ and $c = 0$ and generalized to the present model problem in [EP17].

**Proposition 5.** Let $T_x$ be sufficiently fine such that $C_{\text{ell}} - C_{\text{bal}} \| h_x \|_{L^\infty(\Omega)} > 0$ with $C_{\text{ell}}$ and $C_{\text{bal}}$ from (6) and (18), respectively. Then it holds that

$$C_{\text{ell}}^{-1} \eta_x \leq \min_{v_x \in S_0^h(T_x)} \left( \| u - u_x \|_{H^1(\Omega)} + \text{osc}_x(v_x) \right) \leq \| u - u_x \|_{H^1(\Omega)} + \text{osc}_x \leq C_1 \eta_x.$$
Moreover, if $u_x^{\text{FEM}} \in S_0^1(T_x)$ denotes the FEM solution of $A(u_x^{\text{FEM}}, w_x) = (f, w_x)_\Omega$ for all $w_x \in S_0^1(T_x)$, it holds that
\[
C_2^{-1} \bigl(\|u - u_x\|_{H^1(\Omega)} + \text{osc}_x\bigr) \leq \|u - u_x^{\text{FEM}}\|_{H^1(\Omega)} + \text{osc}_x(u_x^{\text{FEM}}) \leq C_2 \bigl(\|u - u_x\|_{H^1(\Omega)} + \text{osc}_x\bigr).
\]
The constants $C_1, C_2 > 0$ depend only on $\Omega$, the $\sigma$-shape regularity of $T_x$, and the data assumptions (2)–(4).

As a direct consequence of Proposition 5, one obtains the following convergence result and a priori estimate which confirms first-order convergence of FVM; see again [EP16, EP17]. Note that the statement even holds for $u \in H^1_0(\Omega)$, whereas in the literature standard FVM analysis usually requires, e.g., $u \in H^{1+\varepsilon}(\Omega)$ for some $\varepsilon > 0$.

**Corollary 6.** Let $\{T_x\}$ be a family of sufficiently fine and uniformly $\sigma$-shape regular triangulations. Let $u \in H^1_0(\Omega)$ be the solution of (5). Then there holds convergence
\[
\|u - u_x\|_{H^1(\Omega)} + \text{osc}_x \to 0 \quad \text{as} \quad \|h_x\|_{L^\infty(\Omega)} \to 0.
\]
Moreover, additional regularity $u \in H^1_0(\Omega) \cap H^2(\Omega)$ implies first-order convergence
\[
\|u - u_x\|_{H^1(\Omega)} + \text{osc}_x = O(\|h_x\|_{L^\infty(\Omega)}).
\]

3. Adaptive FVM

In this section, we apply an adaptive mesh-refining algorithm for FVM. We combine ideas from [MN05] and [EP16] to prove that adaptive FVM leads to linear convergence with optimal algebraic rates for the error estimator (and hence for the total error; see Proposition 5).

3.1. Adaptive algorithm. As in [EP16], we employ the following adaptive algorithm:

**Algorithm 7.** Input: Let $0 < \theta' \leq \theta \leq 1$ and $C_{\text{mark}}, C'_{\text{mark}} \geq 1$. Let $T_0$ be a conforming triangulation of $\Omega$ which resolves possible discontinuities of $A$.

Loop: For $\ell = 0, 1, 2, \ldots$, iterate the following steps (i)–(v):

(i) **Solve:** Compute the discrete solution $u_{\ell} \in S_0^1(T_\ell)$ from (14).

(ii) **Estimate:** Compute $\eta(T, u_{\ell})$ and $\text{osc}_{\ell}(T, u_{\ell})$ from (22) for all $T \in T_\ell$.

(iii) **Mark I:** Find $M^0_\ell \subseteq T_\ell$ of up to the multiplicative constant $C_{\text{mark}} \geq 1$ minimal cardinality which satisfies the Dörfler marking criterion
\[
(28) \quad \theta \sum_{T \in T_\ell} \eta(T, u_{\ell})^2 \leq \sum_{T \in M^0_\ell} \eta(T, u_{\ell})^2.
\]

(iv) **Mark II:** Find $M_\ell \subseteq T_\ell$ of up to the multiplicative constant $C'_{\text{mark}} \geq 1$ minimal cardinality which satisfies $M^0_\ell \subseteq M_\ell$ as well as the Dörfler marking criterion
\[
(29) \quad \theta' \sum_{T \in T_\ell} \text{osc}_{\ell}(T, u_{\ell})^2 \leq \sum_{T \in M_\ell} \text{osc}_{\ell}(T, u_{\ell})^2.
\]

(v) **Refine:** Generate new triangulation $T_{\ell+1} := \text{refine}(T_\ell, M_\ell)$ by refinement of all marked elements.

Output: Adaptively refined triangulations $T_\ell$, corresponding discrete solutions $u_{\ell}$, estimators $\eta_{\ell}$, and data oscillations $\text{osc}_{\ell}$ for $\ell \geq 0$. 


Due to the lack of standard Galerkin orthogonality (see Section 3.2), we additionally have to mark the oscillations (29). In practice, however, this marking is negligible, since $\theta'$ can be chosen arbitrary small; see [EP16, Remark 7] for more details.

### 3.2. Quasi-Galerkin orthogonality.

Given $g \in L^2(\Omega)$, we consider the dual problem: Find $\phi \in H^1_0(\Omega)$ such that

$$A(v, \phi) = (g, v)_\Omega \quad \text{for all } v \in H^1_0(\Omega).$$

(30)

The Lax-Milgram theorem proves existence and uniqueness of $\phi \in H^1_0(\Omega)$. Let $0 < s \leq 1$. We suppose that the dual problem (30) is $H^{1+s}$-regular, i.e., there exists a constant $C_{\text{dual}} > 0$ such that for all $g \in L^2(\Omega)$, the solution of (30) satisfies

$$\phi \in H^1_0(\Omega) \cap H^{1+s}(\Omega) \quad \text{with } \|\phi\|_{H^{1+s}(\Omega)} \leq C_{\text{dual}} \|g\|_{L^2(\Omega)}.$$  

(31)

We refer to [Gri85] for a discussion on this regularity assumption. The main result of this section is the following quasi-Galerkin orthogonality with respect to the operator-induced quasi norm from (6). The proof is postponed to the end of this section.

**Proposition 8.** Let $0 < s \leq 1$ and suppose that the dual problem (30) is $H^{1+s}$-regular (31). Let $T_0 \in \text{refine}(T_0)$ and $T_x \in \text{refine}(T_0)$. Then there exists $C_{\text{gal}} > 0$ such that

$$\|u - u_x\|_0^2 \leq \|u - u_0\|_0^2 - \frac{1}{2} \|u_x - u_0\|_0^2 + C_{\text{gal}} \|h_x\|_{L^\infty(\Omega)}^2 \|\phi\|_{L^2(\Omega)}^2 + C_{\text{gal}} \|\phi\|_{H^{1+s}(\Omega)}^2.$$  

(32)

The constant $C_{\text{gal}} > 0$ depends only on $C_{\text{dual}}, C_{\text{osc}}, C_{\text{rel}}, C_{\text{ell}}, C_{\text{cont}}, \text{diam}(\Omega)$, and $\|b\|_{W^{1,\infty}(\Omega)}$ as well as on $\sigma$-shape regularity and all possible shapes of element patches in $T_x$.

For the FVM error, the classical Galerkin orthogonality fails, i.e., $A(u - u_x, v_x) \neq 0$ for some $v_x \in S^1_0(T_x)$. However, there holds the following estimate; see, e.g., [EP16].

**Lemma 9.** The FVM error $u - u_x$ satisfies that

$$|A(u - u_x, v_x)| \leq C_{\text{osc}} \|v_x\|_{H^{1+s}(\Omega)} \quad \text{for all } v_x \in S^1_0(T_x).$$  

(33)

The constant $C_{\text{osc}} > 0$ depends only on $\sigma$-shape regularity of $T_x$.

**Proof.** Standard calculations (see, e.g., [Era13, Theorem 4.9]) show that

$$A(u - u_x, v_x) = \sum_{T \in T_x} \int_T R_x(u_x) v_x \, dx + \sum_{F \in \mathcal{F}_x} \int_F J_x(u_x) v_x \, ds.$$  

Together with (25) for $v^*_x = T^*_x v_x \in P_0^0(T^*_x)$, this leads to

$$A(u - u_x, v_x) = \sum_{T \in T_x} \int_T R_x(u_x) (v_x - v^*_x) \, dx + \sum_{F \in \mathcal{F}_x} \int_F J_x(u_x) (v_x - v^*_x) \, ds.$$  

We apply (15) for the involved integrals and obtain that

$$A(u - u_x, v_x) = \sum_{T \in T_x} \int_T (R_x(u_x) - \Pi_x R_x(u_x)) (v_x - v^*_x) \, dx + \sum_{F \in \mathcal{F}_x} \int_F (J_x(u_x) - \Pi_x J_x(u_x)) (v_x - v^*_x) \, ds.$$  

The Cauchy-Schwarz inequality and (16)--(17) conclude the proof. \hfill \Box
Lemma 10. Let $0 < s \leq 1$ and suppose that the dual problem (30) is $H^{1+s}$-regular (31). Then the FVM error satisfies

$$C_{aux}^{-1} \| u - u_x \|^2_{L^2(\Omega)} \leq \| h_x \|^2_{L^\infty(\Omega)} \| u - u_x \|^2_{H^1(\Omega)} + \text{osc}_x^2.$$  

The constant $C_{aux} > 0$ depends only on the $\sigma$-shape regularity of $T_x$, diam($\Omega$), $C_{cont}$, and $C_{dual}$ as well as on all possible shapes of element patches in $T_x$.

Proof. The proof is split into two steps.

**Step 1.** Let $I_x : H^1(\Omega) \to S^1(T_x)$ be the Scott-Zhang projector [SZ90]. Recall the following properties of $I_x$ for all $v \in H^1(\Omega)$ and $v_x \in S^1(T_x)$, and all $T \in T_x$:

- $I_x$ has a local projection property, i.e., $(I_x v)|_T = v_x|_T$ if $v|_{\omega_x(T)} = v_x|_{\omega_x(T)}$;
- $I_x$ preserves discrete boundary data, i.e., $v|_{\Gamma} = v_x|_{\Gamma}$ implies that $(I_x v)|_T = v|_T$;
- $I_x$ is locally $H^1$-stable, i.e., $\| \nabla I_x v \|_{L^2(\Omega)} \leq C_{sz} \| \nabla v \|_{H^1(\omega_x(T))}$;
- $I_x$ has a local approximation property, i.e., $\| v - I_x v \|_{L^2(\Omega)} \leq C_{sz} h_T \| \nabla v \|_{H^1(\omega_x(T))}$.

The constant $C_{sz} > 0$ depends only on $\sigma$-shape regularity of $T_x$. In particular,

$$\| v - I_x v \|_{H^1(\Omega)} \lesssim \| v \|_{H^1(\Omega)}$$

for all $v \in H^1(\Omega)$, where the hidden constant depends only on $C_{sz}$ and diam($\Omega$). With the local projection property of $I_x$, we may apply the Bramble-Hilbert lemma. For $v \in H^2(\Omega)$, scaling arguments then prove that

$$\| v - I_x v \|_{H^1(\Omega)} \lesssim \text{diam}(\omega_x(T)) \| v \|_{H^1(\omega_x(T))}$$

for all $T \in T_x$, where the hidden constant depends only on the shape of $\omega_x(T)$ and on the operator norm of $A := 1 - I_x$ (and hence on diam($\Omega$) and $C_{sz}$). Altogether, this proves the operator norm estimates

$$\| A := 1 - I_x : H^{1+t}(\Omega) \to H^1(\Omega) \| \leq C \| h_x \|_{L^\infty(\Omega)}$$

for $t \in \{0, 1\}$, where $C > 0$ depends only on $C_{sz}$, diam($\Omega$), and all possible shapes of element patches in $T_x$. Interpolation arguments [BL76] conclude that (36) holds for all $0 \leq t \leq 1$. For $t = s$, this proves that

$$\| v - I_x v \|_{H^1(\Omega)} \leq C \| h_x \|_{L^\infty(\Omega)} \| v \|_{H^{1+s}(\Omega)}$$

for all $v \in H^{1+s}(\Omega)$.

**Step 2.*** With $g = v = u - u_x$ in (30), it holds that

$$\| u - u_x \|^2_{L^2(\Omega)} = A(u - u_x, \phi) = A(u - u_x, \phi - I_x \phi) + A(u - u_x, I_x \phi).$$

Since we suppose $\phi \in H^{1+s}(\Omega)$, the first summand is bounded by (37). This yields that

$$\| u - u_x \|^2_{L^2(\Omega)} \leq A(u - u_x, I_x \phi) \lesssim \| u - u_x \|_{H^1(\Omega)} \| \phi - I_x \phi \|_{H^1(\Omega)} \lesssim \| h_x \|_{L^\infty(\Omega)} \| u - u_x \|_{H^1(\Omega)} \| \phi \|_{H^{1+s}(\Omega)},$$

where the hidden constants depend only on $C_{cont}$, $C_{sz}$, and diam($\Omega$). The second summand is bounded by (33) and $H^1$-stability of $I_x$. This yields that

$$\| u - u_x, I_x \phi \| \lesssim \text{osc}_x \| I_x \phi \|_{H^1(\Omega)} \lesssim \text{osc}_x \| \phi \|_{H^1(\Omega)} \lesssim \text{osc}_x \| \phi \|_{H^{1+s}(\Omega)},$$

where the hidden constant depends only on $C_{osc}$, $C_{sz}$ and diam($\Omega$). Combining the latter three estimates with $H^{1+s}$-regularity (31), we prove that

$$\| u - u_x \|^2_{L^2(\Omega)} \lesssim (\| h_x \|_{L^\infty(\Omega)} \| u - u_x \|_{H^1(\Omega)} + \text{osc}_x) \| \phi \|_{H^{1+s}(\Omega)} \lesssim (\| h_x \|_{L^\infty(\Omega)} \| u - u_x \|_{H^1(\Omega)} + \text{osc}_x) \| u - u_x \|_{L^2(\Omega)},$$

where the hidden constant depends additionally on $C_{dual}$. This concludes the proof. □
Proof of Proposition 8. Recall that \( A(v, w) = (A \nabla v, \nabla w)_\Omega - (b v, \nabla w)_\Omega \) and thus \( A(v, w) = (A \nabla w, \nabla v)_\Omega - (b w, \nabla v)_\Omega + (c w, v)_\Omega \). For \( v, w \in H^1_0(\Omega) \), integration by parts proves that

\[-(b w, \nabla v)_\Omega = (b \cdot \nabla w, v)_\Omega + (\text{div}(b) w, v)_\Omega\]

and hence

\[ A(v, w) + A(w, v) = 2A(v, w) + 2(v, b \cdot \nabla w)_\Omega + (\text{div}(b) v, w)_\Omega. \]

By definition of \( ||| \cdot ||| \), this proves that

\[ |||v + w|||^2 = |||v|||^2 + |||w|||^2 + A(v, w) + A(w, v) \]

\[ = |||v|||^2 + |||w|||^2 + 2A(v, w) + 2(v, b \cdot \nabla w)_\Omega + (\text{div}(b) v, w)_\Omega. \]

This leads to

\[ |||v|||^2 = |||v + w|||^2 - |||w|||^2 - 2A(v, w) - 2(v, b \cdot \nabla w)_\Omega - (\text{div}(b) v, w)_\Omega. \]

With \( C_1 := C_{\text{ell}}^{-1} (2||b||_{L^\infty(\Omega)} + ||\text{div}(b)||_{L^\infty(\Omega)})^2 \), the Young inequality \( ab \leq \frac{1}{4} a^2 + b^2 \) and norm equivalence (6) prove that

\[ -2(v, b \cdot \nabla w)_\Omega - (\text{div}(b) v, w)_\Omega \leq ||v||_{L^2(\Omega)} ||w||_{H^1(\Omega)} (2||b||_{L^\infty(\Omega)} + ||\text{div}(b)||_{L^\infty(\Omega)}) \]

\[ \leq \frac{1}{4} ||w||^2 + C_1 ||v||^2_{L^2(\Omega)}. \]

Choose \( v = u - u_\times \) as well as \( w = u_\times - u_\circ \). So far, we have shown that

\[ ||u - u_\times||^2 \leq ||u - u_\circ||^2 - \frac{3}{4} ||u_\times - u_\circ||^2 - 2A(u - u_\times, u_\times - u_\circ) + C_1 ||u - u_\times||^2_{L^2(\Omega)}. \]

We apply (33), norm equivalence (6), and the Young inequality \( 2ab \leq \frac{1}{4} a^2 + 4b^2 \) to see that

\[ -2A(u - u_\times, u_\times - u_\circ) \leq 2C_{\text{osc}} ||u_\times - u_\circ||_{H^1(\Omega)} \text{osc}_{\times} \]

\[ \leq 2C_{\text{osc}} C_{\text{ell}}^{-1/2} ||u_\times - u_\circ||_{\text{osc}_{\times}} \leq \frac{1}{4} ||u_\times - u_\circ||^2 + 4C_{\text{osc}}^2 C_{\text{ell}}^{-1} \text{osc}_{\times}^2. \]

Next, Lemma 10 and reliability (24) lead to

\[ C_{\text{aux}}^{-1} ||u - u_\times||^2_{L^2(\Omega)} \leq ||h_\times||_{L^\infty(\Omega)}^2 ||u - u_\times||_{H^1(\Omega)}^2 + \text{osc}_{\times}^2 \leq C_{\text{rel}} ||h_\times||_{L^\infty(\Omega)}^2 \| \eta_\times \|^2_{L^2(\Omega)} + \text{osc}_{\times}^2. \]

Combining the latter three estimates, we prove that

\[ ||u - u_\times||^2 \leq ||u - u_\circ||^2 - \frac{1}{2} ||u_\times - u_\circ||^2 \]

\[ + C_4 C_{\text{aux}} C_{\text{rel}} ||h_\times||_{L^\infty(\Omega)}^2 \| \eta_\times \|^2_{L^2(\Omega)} + (4C_{\text{osc}}^2 C_{\text{ell}}^{-1} + C_4 C_{\text{aux}}) \text{osc}_{\times}^2. \]

Choosing \( C_{\text{gal}} = \max\{C_4 C_{\text{aux}} C_{\text{rel}}, 4C_{\text{osc}}^2 C_{\text{ell}}^{-1} + C_4 C_{\text{aux}}\} \), we conclude the proof. \( \square \)

3.3. Linear convergence and general quasi-orthogonality. The following properties (A1)–(A2) of the estimator and (B1)–(B2) of the oscillations are some key observations to prove linear convergence of Algorithm 7. The proof is based on scaling arguments and can be found in literature, e.g., [CKNS08, Section 3.1] for (A1)–(A2) and [EP16, Section 3.3] for (B1)–(B2). Their proofs apply almost verbatim to the present non-symmetric problem with \( b \neq 0 \). Therefore, the details are left to the reader.
Lemma 11. There exist constants $0 < q < 1$ and $C > 0$ such that for all $\mathcal{T}_0 \in \text{refine}(\mathcal{T}_0)$, all $\mathcal{T}_x \in \text{refine}(\mathcal{T}_0)$, and all $v_x \in S_0^1(\mathcal{T}_x)$, $v_0 \in S_0^1(\mathcal{T}_0)$, it holds that (stability of estimator on non-refined elements)

\[
\tag{A1} \left( \sum_{T \in \mathcal{T}_x \cap \mathcal{T}_0} \eta_x(T, v_x)^2 \right)^{1/2} - \left( \sum_{T \in \mathcal{T}_x \cap \mathcal{T}_0} \eta_0(T, v_0)^2 \right)^{1/2} \leq C \| v_x - v_0 \|_{H^1(\Omega)},
\]

(reduction of estimator on refined elements)

\[
\tag{A2} \sum_{T \in \mathcal{T}_x \setminus \mathcal{T}_0} \eta_x(T, v_x)^2 \leq q \sum_{T \in \mathcal{T}_x \setminus \mathcal{T}_x} \eta_0(T, v_0)^2 + C \| v_x - v_0 \|_{H^1(\Omega)}^2,
\]

(stability of oscillations on non-refined elements)

\[
\tag{B1} \left( \sum_{T \in \mathcal{T}_x \cap \mathcal{T}_0} \text{osc}_x(T, v_x)^2 \right)^{1/2} - \left( \sum_{T \in \mathcal{T}_x \cap \mathcal{T}_0} \text{osc}_0(T, v_0)^2 \right)^{1/2} \leq C \| h_x \|_{L^\infty(\Omega)} \| v_x - v_0 \|_{H^1(\Omega)},
\]

(reduction of oscillations on refined elements)

\[
\tag{B2} \sum_{T \in \mathcal{T}_x \setminus \mathcal{T}_0} \text{osc}_x(T, v_x)^2 \leq q \sum_{T \in \mathcal{T}_x \setminus \mathcal{T}_x} \text{osc}_0(T, v_0)^2 + C \| h_x \|_{L^\infty(\Omega)} \| v_x - v_0 \|_{H^1(\Omega)}^2.
\]

The constants $0 < q < 1$ and $C > 0$ depend only on the $\sigma$-shape regularity (8) and on the data assumptions (2)–(4).

Theorem 12 (linear convergence). Let $0 < \theta' \leq \theta \leq 1$. There exists $H > 0$ such that the following statement is valid provided that $\| h_0 \|_{L^\infty(\Omega)} \leq H$ and that the dual problem (30) is $H^{1+s}$-regular (31) for some $0 < s \leq 1$: There exist $C_{\text{lin}} > 0$ and $0 < q_{\text{lin}} < 1$ such that Algorithm 7 guarantees linear convergence in the sense of

\[
\tag{38} \eta_{\ell+n}^2 \leq C_{\text{lin}} q_{\text{lin}}^2 \eta_{\ell}^2 \quad \text{for all } \ell, n \in \mathbb{N}_0.
\]

The constant $H$ depends only on the $\sigma$-shape regularity (8), on the data assumptions (2)–(4), $C_{\text{gal}}$, $\theta$, and $\theta'$, whereas $C_{\text{lin}}$ and $q_{\text{lin}}$ additionally depend on $C_{\text{cont}}$ and $C_{\text{rel}}$.

Proof. The proof is split into three steps.

**Step 1.** There exist constants $C > 0$ and $0 < q < 1$ which depend only on $0 < \theta \leq 1$, $C_{\text{ell}}$, and the constants in (A1)–(A2), such that

\[
\tag{39} \eta_{\ell+1}^2 \leq q \eta_\ell^2 + C \| u_{\ell+1} - u_\ell \|^2 \quad \text{for all } \ell \in \mathbb{N}_0.
\]

Furthermore, there exist constants $C > 0$ and $0 < q < 1$ which depend only on $0 < \theta' \leq 1$, $C_{\text{ell}}$, and the constants in (B1)–(B2), such that

\[
\tag{40} \text{osc}_{\ell+1}^2 \leq q \text{osc}_\ell^2 + C \| h_{\ell+1} \|_{L^\infty(\Omega)} \| u_{\ell+1} - u_\ell \|^2 \quad \text{for all } \ell \in \mathbb{N}_0.
\]

The proofs of (39) and (40) rely only on (A1)–(A2) with the Dörfler marking (28) and (B1)–(B2) with marking (29), respectively. For details, we refer, e.g., to [EP16, Proposition 10 (step 1 and step 2)].

**Step 2.** Without loss of generality, we may assume that the constants $C > 0$ and $0 < q < 1$ in (39)–(40) are the same. With free parameters $\gamma, \mu > 0$, we define

\[
\Delta_x := \| u - u_x \|^2 + \gamma \eta_x^2 + \mu \text{osc}_x^2.
\]

We employ the quasi-Galerkin orthogonality (32) and obtain that

\[
\Delta_{\ell+1} \leq \| u - u_\ell \|^2 + [\gamma + C_{\text{gal}} \| h_{\ell+1} \|_{L^\infty(\Omega)}^2] \eta_{\ell+1}^2 + [\mu + C_{\text{gal}}] \text{osc}_{\ell+1}^2 - \frac{1}{2} \| u_{\ell+1} - u_\ell \|^2.
\]
Using (39)–(40), we further derive that
\[
\Delta_{\ell+1} \leq \|u - u_{\ell}\|^2 + [\gamma + C_{\text{gal}} \|h_{\ell+1}\|_{L^\infty(\Omega)}^2] q \eta_\ell^2 + [\mu + C_{\text{gal}}] q \text{osc}_\ell^2
\]
\[
- \left(\frac{1}{2} - C [\gamma + C_{\text{gal}} h_{\ell+1}^2] - C [\gamma h_{\ell+1}^2] \left[\mu + C_{\text{gal}}\right]\right) \|u_{\ell+1} - u_{\ell}\|^2.
\]

Let \( H > 0 \) be a free parameter and suppose that \( \|h_0\|_{L^\infty(\Omega)} \leq H \). We estimate \( \|h_{\ell+1}\|_{L^\infty(\Omega)} \leq H \). Norm equivalence (6) and reliability (24) prove that
\[
\|u - u_{\ell}\|^2 \leq C_{\text{cont}} \|u - u_{\ell}\|_{H^1(\Omega)}^2 \leq C_{\text{cont}} C_{\text{rel}} \eta_\ell^2.
\]

Let \( \varepsilon > 0 \) be a free parameter. Combining the last two estimates, we see that
\[
\Delta_{\ell+1} \leq (1 - \varepsilon) \|u - u_{\ell}\|^2 + [\gamma [(1 + \gamma^{-1} C_{\text{gal}} H^2)q + \gamma^{-1} \varepsilon C_{\text{cont}} C_{\text{rel}}] \eta_\ell^2 + \mu [1 + \mu^{-1} C_{\text{gal}}] q \text{osc}_\ell^2
\]
\[
- \left(\frac{1}{2} - C [\gamma + C_{\text{gal}} H^2] - C H^2 [\mu + C_{\text{gal}}]\right) \|u_{\ell+1} - u_{\ell}\|^2.
\]

**Step 3.** It only remains to fix the four free parameters \( \gamma, \mu, \varepsilon, \) and \( H \).
- Choose \( \gamma > 0 \) sufficiently small such that \( \gamma C < 1/2 \).
- Choose \( \mu > 0 \) sufficiently large such that \( q_{\text{osc}} := [1 + \mu^{-1} C_{\text{gal}}] q < 1 \).
- Choose \( H \) sufficiently small such that
  - \( C [\gamma + C_{\text{gal}} H^2] + C H^2 [\mu + C_{\text{gal}}] < 1/2 \),
  - \( (1 + \gamma^{-1} C_{\text{gal}} H^2)q < 1 \).
- Choose \( 0 < \varepsilon < 1 \) such that \( q_{\text{est}} := [(1 + \gamma^{-1} C_{\text{gal}} H^2)q + \gamma^{-1} \varepsilon C_{\text{cont}} C_{\text{rel}}] < 1 \).

With \( q_{\text{lin}} := \max\{1 - \varepsilon, q_{\text{est}}, q_{\text{osc}}\} \), we then obtain that
\[
\Delta_{\ell+1} \leq (1 - \varepsilon) \|u - u_{\ell}\|^2 + [\gamma [(1 + \gamma^{-1} C_{\text{gal}} H^2)q + \gamma^{-1} \varepsilon C_{\text{cont}} C_{\text{rel}}] \eta_\ell^2 + \mu [1 + \mu^{-1} C_{\text{gal}}] q \text{osc}_\ell^2
\]
\[
\leq \max\{1 - \varepsilon, q_{\text{est}}, q_{\text{osc}}\} \Delta_\ell = q_{\text{lin}} \Delta_\ell.
\]

Induction on \( n \), norm equivalence (6), reliability (24), and \( \text{osc}_\ell^2 \leq \eta_\ell^2 \) prove that
\[
\gamma \eta_{\ell+n}^2 \leq \Delta_{\ell+n} \leq q_{\text{lin}}^n \Delta_\ell \leq q_{\text{lin}}^n (C_{\text{rel}} C_{\text{cont}} + \gamma + \mu) \eta_\ell^2 \quad \text{for all } \ell, n \in \mathbb{N}_0
\]
This concludes linear convergence (38) with \( C_{\text{lin}} = (C_{\text{rel}} C_{\text{cont}} + \gamma + \mu) \gamma^{-1} \).

From the linear convergence (38), we immediately obtain the so-called general quasi-orthogonality; see, e.g., [CFPP14, Proposition 4.11] or [EP16, Proposition 10 (step 5)].

**Corollary 13 (general quasi-orthogonality).** Let \( (u_\ell) \) be the sequence of solutions of Algorithm 7. Then there exists \( C > 0 \) such that
\[
(A3) \quad \sum_{k=\ell}^\infty \|u_{k+1} - u_k\|_{H^1(\Omega)}^2 \leq C \eta_\ell^2 \quad \text{for all } \ell \in \mathbb{N}_0.
\]

The constant \( C > 0 \) has the same dependencies as \( C_{\text{lin}} \) from (38).

### 3.4. Optimal algebraic convergence rates.
In order to prove optimal convergence rates of Algorithm 7, we need one further property of the error estimator, namely the so-called discrete reliability (A4). The proof of the following lemma follows as for the symmetric case in [EP16, Proposition 15]. While the proof is thus omitted, we note that the main difficulties over the well-known FEM proof [CKNS08] arise in the handling of the piecewise constant test space on \( T_x \) and \( T_\sigma \), respectively, and the fact that these test spaces are not nested.
Lemma 14 (discrete reliability). There exists a constant $C > 0$ such that for all $T_{0} \in \text{refine}(T_{0})$ and all $T_{x} \in \text{refine}(T_{0})$, it holds that

\[ (A4) \quad \|u_{x} - u_{c}\|_{H^{1}(\Omega)}^{2} \leq C \left( \sum_{T \in T_{x}} h_{T}^{2} \|u_{x} - u_{c}\|_{H^{1}(T)}^{2} + \sum_{T \in R_{0}} \eta_{0}(T, u_{c})^{2} \right), \]

where $R_{0} := \{ T \in T_{0} : \exists T' \in T_{0} \setminus T_{x} \text{ with } T \cap T' \neq \emptyset \}$ consists of all refined elements $T_{0} \setminus T_{x}$ plus one additional layer of neighboring elements. The constant $C > 0$ depends only on the $\sigma$-shape regularity (8), the data assumptions (2)–(4), and $\Omega$. Note that for a sufficiently fine initial mesh $T_{0}$, e.g., $C \|h_{0}\|_{L^{\infty}(\Omega)}^{2} \leq 1/2$, (A4) leads to discrete reliability as stated in [CFPP14]. \hfill $\Box$

Let $T := \text{refine}(T_{0})$ be the set of all possible triangulations obtained by NVB. For $N \geq 0$, let $T_{N} := \{ T_{x} \in T : \#T_{x} - \#T_{0} \leq N \}$. For $s > 0$, define

\[ (41) \quad \|u\|_{A_{s}} := \sup_{N \in \mathbb{N}_{0}} \inf_{T_{N} \in T_{N}} (N + 1)^{s} \eta_{x}. \]

Note that $\|u\|_{A_{s}} < \infty$ implies an algebraic decay $\eta_{x} = O((\#T_{x})^{-s})$ along the optimal sequence of meshes (which minimize the error estimator). Optimal convergence of the adaptive algorithm thus means that for all $s > 0$ with $\|u\|_{A_{s}} < \infty$, the adaptive algorithm leads to $\eta_{x} = O((\#T_{x})^{-s})$. The work [CFPP14, Theorem 4.1] proves in a general framework the following Theorem 15, if the adaptive algorithm applied to a numerical scheme and a corresponding estimator satisfies (A1), (A2), (A3), and (A4).

Theorem 15 (optimal algebraic convergence rates). Suppose that the dual problem (30) is $H^{1+s}$-regular (31) for some $0 < s \leq 1$. Let the initial mesh $T_{0}$ be sufficiently fine, i.e., there exists a constant $H > 0$ such that $\|h_{0}\|_{L^{\infty}(\Omega)} \leq H$. Finally, suppose that there is a constant $C_{\text{MNS}} \geq 1$ such that $\#M_{t} \leq C_{\text{MNS}}\#M_{0}^{t}$ for all $t \in \mathbb{N}_{0}$. Then there exists a bound $0 < \theta_{\text{opt}} \leq 1$ such that for all $0 < \theta < \theta_{\text{opt}}$ and all $s > 0$ with $\|u\|_{A_{s}} < \infty$, there exists a constant $C_{\text{opt}} > 0$ such that

\[ (42) \quad \eta_{t} \leq C_{\text{opt}} (\#T_{t} - \#T_{0})^{-s} \quad \text{for all } t \in \mathbb{N}. \]

The constant $\theta_{\text{opt}}$ depends only on $\Omega$, $H$, uniform $\sigma$-shape regularity of the triangulations $T_{x} \in \text{refine}(T_{0})$, and the data assumptions (2)–(4). The constant $C_{\text{opt}}$ additionally depends on $s$, the constant $q_{\text{lin}}$ from (38), the use of NVB, and on $C_{\text{MNS}}$. \hfill $\Box$

Remark 16. A direct consequence of the assumption $\#M_{t} \leq C_{\text{MNS}}\#M_{0}^{t}$ in Theorem 15 is that data oscillation marking (29) is negligible with respect to the overall number of marked elements [EP16, Remark 7]. In practice, (28) already implies (29) since $\theta' > 0$ can be chosen arbitrarily small. Furthermore, efficiency (24) is not required to show (38) and (42) but guarantees (optimal) linear convergence also for the FVM error.

4. Numerical examples

In extension of our theory, we consider the model problem (1) with inhomogeneous Dirichlet boundary conditions. For all experiments in 2D, we run Algorithm 7 with $\theta = 1 = \theta'$ and $\theta = 0.5 = \theta'$ for uniform mesh-refinement and adaptive mesh-refinement, respectively.

4.1. Experiment with smooth solution. On the square $\Omega = (-1,1)^2$, we prescribe the exact solution $u(x_{1}, x_{2}) = (1 - 10x_{1}^{2} - 10x_{2}^{2})e^{-5(x_{1}^{2} + x_{2}^{2})}$ with $x = (x_{1}, x_{2}) \in \mathbb{R}^{2}$. We choose the diffusion matrix

$$
A = \begin{pmatrix}
10 + \cos x_{1} & 9 x_{1} x_{2} \\
9 x_{1} x_{2} & 10 + \sin x_{2}
\end{pmatrix},
$$

13
the velocity \( b = (\sin x_1, \cos x_2)^T \) and the reaction \( c = 1 \). Note that (2) holds with \( \lambda_{\min} = 0.82293 \) and \( \lambda_{\max} = 10.84096 \), and (4) with \( \frac{1}{2} \text{div} b + c > 0 \). The right-hand side \( f \) is calculated appropriately. The uniform initial mesh \( T_0 \) consists of the 16 triangles.

In Figure 2(a) we see an adaptively generated mesh after 16 refinements. Figure 2(b) plots the smooth solution on the mesh \( T_{16} \). Both, uniform and adaptive mesh-refinement,
Table 1. Experiment with smooth solution from Section 4.1: We compute \( \tilde{C}_{MNS} := \frac{\mathcal{M}_\ell}{\mathcal{M}_\ell^0} \leq 1.3 \). Hence, the additional assumption in Theorem 15 is experimentally verified. Furthermore, we compute \( \tilde{\theta} := \frac{\operatorname{osc}_\ell(M_{\ell}^0)^2}{\operatorname{osc}_\ell^2} \geq 0.2 \) with \( \operatorname{osc}_\ell(M_{\ell}^0)^2 := \sum_{T \in \mathcal{M}_\ell^0} \operatorname{osc}_\ell(T, u_{\ell})^2 \), i.e., the choice \( \theta = 0.5, \tilde{\theta} = 0.2 \) would guarantee \( \mathcal{M}_\ell = \mathcal{M}_\ell^0 \) in Algorithm 7.

| \( \ell \) | \#\( T_\ell \) | \#\( \mathcal{M}_\ell \) | \#\( \mathcal{M}_\ell^0 \) | \( \frac{\operatorname{osc}_\ell(M_{\ell}^0)^2}{\operatorname{osc}_\ell^2} \) | \( \ell \) | \#\( T_\ell \) | \#\( \mathcal{M}_\ell \) | \#\( \mathcal{M}_\ell^0 \) | \( \frac{\operatorname{osc}_\ell(M_{\ell}^0)^2}{\operatorname{osc}_\ell^2} \) |
|---|---|---|---|---|---|---|---|---|---|
| 0 | 16 | 1,000 | 0.631 | 12 | 944 | 1,027 | 0.431 | 24 | 52,422 | 1,000 | 0.540 |
| 1 | 22 | 1,000 | 0.615 | 13 | 1,338 | 1,025 | 0.400 | 25 | 72,454 | 1,007 | 0.404 |
| 2 | 28 | 1,000 | 0.704 | 14 | 1,910 | 1,018 | 0.387 | 26 | 98,232 | 1,000 | 0.508 |
| 3 | 32 | 1,000 | 0.769 | 15 | 2,748 | 1,026 | 0.374 | 27 | 135,172 | 1,004 | 0.446 |
| 4 | 40 | 1,214 | 0.338 | 16 | 3,842 | 1,015 | 0.358 | 28 | 184,142 | 1,000 | 0.606 |
| 5 | 78 | 1,111 | 0.446 | 17 | 5,430 | 1,003 | 0.449 | 29 | 251,896 | 1,002 | 0.475 |
| 6 | 112 | 1,133 | 0.292 | 18 | 7,438 | 1,013 | 0.359 | 30 | 342,148 | 1,001 | 0.488 |
| 7 | 156 | 1,119 | 0.410 | 19 | 10,590 | 1,003 | 0.445 | 31 | 461,674 | 1,000 | 0.617 |
| 8 | 216 | 1,062 | 0.394 | 20 | 14,478 | 1,019 | 0.323 | 32 | 635,266 | 1,004 | 0.416 |
| 9 | 331 | 1,198 | 0.264 | 21 | 20,286 | 1,004 | 0.430 | 33 | 852,730 | 1,000 | 0.664 |
| 10 | 460 | 1,014 | 0.472 | 22 | 27,558 | 1,004 | 0.457 | 34 | 1,172,122 | 1,002 | 0.464 |
| 11 | 660 | 1,049 | 0.371 | 23 | 38,450 | 1,010 | 0.324 | | | | |

Figure 4. Experiment with generic singularity in the reentrant corner \((0, 0)\) from Section 4.2: Adaptively generated mesh \( T_{16} \) from a uniform initial triangulation \( T_0 \) with 12 elements, and discrete FVM solution calculated on \( T_{16} \).

lead to the optimal convergence order \( \mathcal{O}(N^{-1/2}) \) with respect to the number \( N \) of elements since \( u \) is smooth; see Figure 3. The oscillations are of higher order and decrease with \( \mathcal{O}(N^{-1}) \).

Table 1 shows the experimental validation of the additional assumption in Theorem 15, i.e., marking for the data oscillations is negligible; see also Remark 16.

4.2. Experiment with generic singularity. On the L-shaped domain \( \Omega = (-1, 1)^2 \setminus ([0, 1] \times [-1, 0]) \) we consider the exact solution \( u(x_1, x_2) = r^{2/3} \sin(2\varphi/3) \) in polar coordinates \( r \in \mathbb{R}_+^+, \varphi \in [0, 2\pi], \) and \( (x_1, x_2) = r(\cos \varphi, \sin \varphi) \). It is well known that \( u \) has a generic singularity at the reentrant corner \((0, 0)\), which leads to \( u \in H^{1+2/3-\varepsilon}(\Omega) \).
Figure 5. Experiment with generic singularity from Section 4.2: Error $E_\ell = \|u - u_\ell\|_{H^1(\Omega)}$, weighted-residual error estimator $\eta_\ell$, and data oscillations $\text{osc}_\ell$ for uniform and adaptive mesh-refinement.

| $\ell$ | $\#T_\ell$ | $\#M_\ell$ | $\#M_\ell^{\eta}$ | $\text{osc}(M_\ell^{\eta})^2$ | $\text{osc}\ell$ | $\eta_\ell$ | $\text{osc}_\ell$ |
|-------|-------------|-------------|-------------------|-------------------|---------------|-----------|-----------|
| 0     | 12          | 1.667       | 0.135             |                   |               |           |           |
| 1     | 18          | 1.750       | 0.086             |                   |               |           |           |
| 2     | 29          | 1.600       | 0.027             |                   |               |           |           |
| 3     | 40          | 1.375       | 0.057             |                   |               |           |           |
| 4     | 56          | 1.400       | 0.252             |                   |               |           |           |
| 5     | 74          | 1.667       | 0.079             |                   |               |           |           |
| 6     | 114         | 1.286       | 0.148             |                   |               |           |           |
| 7     | 153         | 1.188       | 0.243             |                   |               |           |           |
| 8     | 212         | 1.111       | 0.256             |                   |               |           |           |
| 9     | 284         | 1.065       | 0.390             |                   |               |           |           |
| 10    | 380         | 1.194       | 0.168             |                   |               |           |           |
| 11    | 539         | 1.068       | 0.328             |                   |               |           |           |

Table 2. Experiment with smooth solution from Section 4.2: We compute $C_{\text{MNS}} := \#M_\ell^{0}/\#M_\ell^{\eta} \leq 1.8$. Hence, the additional assumption in Theorem 15 is experimentally verified. Furthermore, we compute $\bar{\theta} := \text{osc}_\ell(M_\ell^{\eta})^2/\text{osc}_\ell^2 \geq 0.02$ with $\text{osc}_\ell(M_\ell^{\eta})^2 := \sum_{T \in M_\ell^{\eta}} \text{osc}_\ell(T, u_\ell)^2$,

i.e., the choice $\theta = 0.5$, $\theta' = 0.02$ would guarantee $M_\ell = M_\ell^\eta$ in Algorithm 7.

for all $\varepsilon > 0$. We choose the diffusion matrix

$$A = \begin{pmatrix} 5 + (x_1^2 + x_2^2) \cos x_1 & (x_1^2 + x_2^2)^2/5 + (x_1^2 + x_2^2) \sin x_2 \\ (x_1^2 + x_2^2)^2 & 5 + (x_1^2 + x_2^2) \sin x_1 \end{pmatrix}$$
Figure 6. Convection dominated experiment from Section 4.3: The discrete FVM solution on an uniformly generated mesh $T_8$ and adaptively generated mesh $T_{14}$. The algorithm starts with a uniform initial triangulation $T_0$ with 32 elements.

Figure 7. Convection dominated experiment from Section 4.3: Adaptively generated meshes $T_{14}$ and $T_{20}$ from a uniform initial triangulation $T_0$ with 32 elements.

so that (2) holds with $\lambda_{\text{min}} = 0.46689$ and $\lambda_{\text{max}} = 5.14751$, and $b = (1, 1)^T$ and $c = 1$ so that (4) holds with $\frac{1}{2}\text{div} b + c = 1$. The right-hand side $f$ is calculated appropriately. The uniform initial mesh $T_0$ consists of 12 triangles. An adaptively generated mesh after 16 refinements and a plot of the discrete solution are shown in Figure 4.

We observe the expected suboptimal convergence order of $O(N^{-1/3})$ for uniform mesh-refinement. We regain the optimal convergence order of $O(N^{-1/2})$ for adaptive mesh-refinement; see Figure 5. As in the experiment of Section 4.1, the oscillations are of higher order $O(N^{-1})$. We refer to Table 2 for the experimental validation of the additional assumption in Theorem 15 that marking for the data oscillations is negligible.

4.3. Convection dominated experiment. The final example is taken from [MN05]. On the square $\Omega = (0, 1)^2$, we fix the diffusion $A = 10^{-3}I$ and the convection velocity $b = (x_2, 1/2 - x_1)^T$. The reaction and right-hand side are $c = f = 0$. Thus, (2) holds with $\lambda_{\text{min}} = \lambda_{\text{max}} = 10^{-3}$ and (4) with $\frac{1}{2}\text{div} b + c = 0$. On the Dirichlet boundary $\Gamma$, we
Figure 8. Convection dominated experiment from Section 4.3: Weighted-residual error estimator \( \eta_\ell \) and data oscillations \( \text{osc}_\ell \) for uniform and adaptive mesh-refinement.

prescribe the continuous piecewise linear function by

\[
u(x_1, x_2)|_\Gamma = \begin{cases} 
1 & \text{on } \{0.2005 \leq x_1 \leq 0.4995, x_2 = 0\}, \\
0 & \text{on } \Gamma \setminus \{0.2 \leq x_1 \leq 0.5; x_2 = 0\} \\
\text{linear} & \text{on } \{0.2 \leq x_1 \leq 0.2005 \text{ or } 0.4995 \leq x_1 \leq 0.5; x_2 = 0\},
\end{cases}
\]

The model has a moderate convection dominance with respect to the diffusion and simulates the transport of a pulse from \( \Gamma \) to the interior and back to \( \Gamma \). For this example, we do not know the analytical solution. The uniform initial mesh \( \mathcal{T}_0 \) consists of 32 triangles. In Figure 6(a), we see the solution with strong oscillations on an uniformly generated mesh with 8192 elements. The oscillations are due to the convection dominance. For the next refinement step (16384 elements, not plotted), however, the oscillations disappear since the shock region at the boundary is refined enough. Our adaptive Algorithm 7, which also has a mandatory oscillation marking, provides a stable solution on a mesh with only 779 elements; see Figure 6(b). In Figure 7, we plot adaptively generated meshes after 14 and 20 mesh-refinements. We see a strong refinement in the shock region. A similar observation can be found in [MN05]. We remark that this strategy only works for this moderate convection dominated problem. For \( A = 10^{-8}I \), we cannot see any stabilization effects by Algorithm 7 (not displayed). Hence, only a stabilization of the numerical scheme, e.g., FVM with upwinding, would avoid these instabilities. However, the analysis of such schemes is beyond the scope of this work. We observe the above stabilization effects also in the convergence plot of the estimator; see Figure 8. Note that the estimator for adaptive mesh-refinement is faster in the asymptotic convergence than the estimator for uniform mesh-refinement. Additionally, the convergence rate for the estimator is suboptimal for uniform mesh-refinement. For adaptive mesh-refinement, we
Table 3. Experimental results on marking strategy for the convection dominated experiment from Section 4.3: We compute $\tilde{C}_{\text{MNS}} := \frac{\#M_\ell}{\#\tilde{M}_\ell} \leq 3$ and see that the additional assumption in Theorem 15 is experimentally verified. In addition, we compute $\tilde{\theta}' := \frac{\text{osc}_c(\tilde{M}_\ell)^2}{\text{osc}_c^2} \geq 0.03$ with $\text{osc}_c(\tilde{M}_\ell)^2 := \sum_{T \in \tilde{M}_\ell} \text{osc}_c(T, u_\ell)^2$, i.e., the choice $\theta = 0.5$, $\theta' = 0.03$ would guarantee $\tilde{M}_\ell = M_\ell$ in Algorithm 7.

regain the optimal convergence order of $O(N^{-1/2})$; see Figure 8. As in the previous experiments, the oscillations are of higher order. In Table 3, we also see that the oscillation marking for this convection dominated problem is for more refinement steps dominant than for the previous problems; see also the discussion in [MN05].

5. Conclusions

In this work, we have proved linear convergence of an adaptive vertex-centered finite volume method with generically optimal algebraic rates to the solution of a general second-order linear elliptic PDE. Besides marking based on the local contributions of the a posteriori error estimator, we additionally had to mark the oscillations to overcome the lack of a classical Galerkin orthogonality property. In case of dominating convection, finite volume methods provide a natural upwind stabilization. Although there exist estimators also for these upwind discretizations [Era13], we were not able to provide a rigorous convergence result for the related adaptive mesh-refinement strategy. Note that the upwind direction and thus the corresponding error indicator contributions are defined over the boundary of the control volumes of the dual mesh. As mentioned above, the dual meshes are not nested even for a sequence of locally refined triangulations. This makes it difficult to show (A1)–(A2) and (B1)–(B2). We stress that the other error indicator contributions are defined over the elements of the primal mesh and can hence be treated by the developed techniques.

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