

Maximum norm *a posteriori* error estimates for convection–diffusion problems

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We prove residual-type *a posteriori* error estimates in the maximum norm for a linear scalar elliptic convection–diffusion problem that may be singularly perturbed. Similar error analysis in the energy norm by Verfürth indicates that a dual norm of the convective derivative of the error must be added to the natural energy norm in order for the natural residual estimator to be reliable and efficient. We show that the situation is similar for the maximum norm. In particular, we define a mesh-dependent weighted seminorm of the convective error, which functions as a maximum-norm counterpart to the dual norm used in the energy norm setting. The total error is then defined as the sum of this seminorm, the maximum norm of the error and data oscillation. The natural maximum norm residual error estimator is shown to be equivalent to this total error notion, with constant independent of singular perturbation parameters. These estimates are proved under the assumption that certain natural estimates hold for the Green’s function for the problem at hand. Numerical experiments confirm that our estimators effectively capture the maximum-norm error behavior for singularly perturbed problems, and can effectively drive adaptive refinement in order to capture layer phenomena.

Keywords: *a posteriori* error estimate; maximum norm; singular perturbation; convection–diffusion.

1. Introduction

Our goal is to prove residual-type *a posteriori* error estimates in the maximum norm for singularly perturbed convection–diffusion equations of the form

$$Lu := -\varepsilon \Delta u + \operatorname{div}(\mathbf{a}u) + bu = f \quad \text{in } \Omega, \quad u = 0 \quad \text{on } \partial\Omega. \quad (1.1)$$

Here, $0 < \varepsilon \leq 1$, Ω is a polyhedral domain in \mathbb{R}^n , $n = 2, 3$, and we assume that $\mathbf{a} = (a_1, \dots, a_n)$, b and f are sufficiently smooth on $\bar{\Omega}$, and that $|\mathbf{a}| > 0$, $b \geq 0$ and $b + \frac{1}{2} \operatorname{div} \mathbf{a} \geq 0$ in $\bar{\Omega}$. We make additional assumptions on the Green’s function of L , which agree with the sharp bounds on the Green’s function for a particular case of (1.1), rigorously proved in Franz & Kopteva (2011a,b, 2012, 2022). We emphasize that our estimates are robust with respect to the singular perturbation parameter ε , up to logarithmic terms that typically arise in the context of maximum norm estimates for finite element methods.

A guiding principle of *a posteriori* error estimation is that estimators should be reliable and efficient (i.e., provide *a posteriori* upper and lower bounds) for the error notion under consideration. For symmetric elliptic problems, residual-type error estimators of the type, we consider here are well known to be reliable and efficient for energy and a number of L_p -type norms, up to a data oscillation term that is heuristically of higher order and measures the distance of the right-hand side f to a piecewise polynomial space. By contrast, for convection–diffusion problems, it is known that standard residual estimators reliably bound the error in the energy norm defined by $\|\mathbf{v}\|^2 = \varepsilon \|\nabla \mathbf{v}\|_{L_2(\Omega)}^2 + \|(b + \frac{1}{2} \operatorname{div} \mathbf{a})^{1/2} \mathbf{v}\|_{L_2(\Omega)}^2$, but are not efficient. To be more precise (Verfürth, 2005), the natural energy-norm residual estimator then is reliable and efficient up to data oscillation for the error notion $\|u - u_h\| + \|\mathbf{a} \cdot \nabla(u - u_h)\|_*$, where $\|\varphi\|_* := \sup_{v \in H_0^1(\Omega) \setminus \{0\}} \frac{(\varphi, v)}{\|\mathbf{v}\|}$, that is for the sum of the energy norm and a dual norm of the convective derivative of the error (also to be referred to as the convective error). The dual convective norm is weaker and thus may be asymptotically negligible, but can also dominate the error as measured in the energy norm until layers in the solution are sufficiently refined. This framework is explained in Tobiska & Verfürth (2015), where residual-type *a posteriori* error estimates are proved for several stabilized finite element methods for equations of the type (1.1) (see also Verfürth, 2005, 2013). Philosophically, we closely follow this work, but while considering the maximum norm rather than an energy norm.

We begin by proving residual-type estimates in the maximum norm for standard Galerkin finite element methods without stabilization. Such methods are not necessarily practically relevant in the convection-dominated regime, but will allow us to establish a suitable theoretical framework before adding stabilization terms. Next, in order to obtain a reliable and efficient estimator, we add an elementwise-weighted measure of the convective derivative of the error to our error notion. This seminorm of the convective error has a similar purpose to the one defined for the energy norm in Tobiska & Verfürth (2015), and is similarly independent of the stabilization terms. However, in contrast to that work, our seminorm is mesh-dependent and is adapted to the particular case of the maximum norm. A precise definition is given below, but this seminorm behaves similarly to the quantity $\|\min\{1, \ell_h \varepsilon^{-1} h_T^2\} \mathbf{a} \cdot \nabla(u - u_h)\|_{L_\infty(\Omega)}$ under sufficiently restrictive assumptions. Here, u_h is the finite element solution, ℓ_h is a logarithmic factor, depending on ε and the minimum mesh diameter, and h_T is the local mesh size. This quantity typically dominates the original target error notion $\|u - u_h\|_{L_\infty(\Omega)}$ when $h_T > \varepsilon$, but becomes relatively negligible when $h_T \ll \varepsilon$. This error structure is similar to that observed in the energy norm case. After considering unstabilized finite element methods, we consider the effects of several stabilization schemes on our *a posteriori* estimates, also following the similar analysis for energy norms outlined in Tobiska & Verfürth (2015).

As in recent works on maximum-norm *a posteriori* error estimation (Demlow & Georgoulis, 2012; Demlow & Kopteva, 2016), we shall rely on the continuous Green’s function for the adjoint problem to (1.1) in order to represent the error. Estimates for the Green’s function in various norms are essential to proving sharp *a posteriori* estimates. The estimates that we need have been established under relatively restrictive assumptions on Ω and the streamline direction \mathbf{a} in Franz & Kopteva (2012). Extension to more general cases appears to be technically quite challenging, so we shall prove our results under the assumption that the Green’s function behaves as in Franz & Kopteva (2012), without giving a complete theoretical picture of the situations for which this assumption is valid.

Finally, we are not aware of previous works on pointwise or maximum-norm *a posteriori* error estimation for finite element methods for convection-dominated convection–diffusion equations. Maximum-norm *a posteriori* estimates for finite difference methods for one-dimensional convection–diffusion scalar problems and systems are contained in Kopteva (2001); Linss (2009, 2010). Well-known *a priori* maximum-norm analyses of streamline diffusion finite element methods are given in Johnson *et al.* (1987); Niijima (1990) for regions away from layers. There is also a considerable literature on

ε -uniform maximum-norm error bounds for finite difference methods on *a-priori*-chosen layer-adapted meshes; see, e.g., Kopteva & O'Riordan (2010), (Roos *et al.*, 2008, §III.2) and the references therein. Thus, there is longstanding interest in controlling pointwise errors in the presence of layer phenomena.

An outline of the paper is as follows. In Section 2, we give preliminaries, including the definition of the unstabilized finite element method and discussion of Green's functions. In Section 3, we prove *a posteriori* error estimates that are reliable and efficient for the sum of the maximum norm, and an appropriate mesh-dependent seminorm of the error in the convective derivative, up to a data oscillation term. In Section 4, we consider a few different stabilization schemes and prove that our estimates remain valid for them under suitable assumptions. Finally, in Section 5, we present numerical experiments that illustrate the behavior and performance of our estimators in the context of uniform and adaptive mesh refinement.

2. Preliminaries

In this section, we discuss the analytical framework for our results and also give some finite element tools.

2.1 Notation

We write $\alpha \simeq \beta$ when $\alpha \lesssim \beta$ and $\alpha \gtrsim \beta$, and $\alpha \lesssim \beta$ when $\alpha \leq C\beta$ with a generic constant C , depending on Ω , \mathbf{a} , b and f , but not on ε or the diameters of mesh elements. For $\mathcal{D} \subset \Omega$, $1 \leq p \leq \infty$ and $k \geq 0$, let $\|\cdot\|_{p;\mathcal{D}} = \|\cdot\|_{L_p(\mathcal{D})}$ and $|\cdot|_{k,p;\mathcal{D}} = |\cdot|_{W_p^k(\mathcal{D})}$, where $|\cdot|_{W_p^k(\mathcal{D})}$ is the standard Sobolev seminorm with integrability index p and smoothness index k , while $\langle \cdot, \cdot \rangle$ is the $L_2(\Omega)$ inner product.

2.2 Green's functions

As is standard in the literature on maximum-norm error bounds in FEM, we employ a Green's function in order to represent the error pointwise. Let $G = G(x, \cdot)$ be the Green's function associated with (1.1). For each fixed $x \in \Omega$, it satisfies

$$L^*G = -\varepsilon \Delta G - \mathbf{a} \cdot \nabla G + bG = \delta(x - \cdot) \quad \text{in } \Omega, \quad G = 0 \quad \text{on } \partial\Omega, \quad (2.1)$$

where $\delta(\cdot)$ is the n -dimensional Dirac δ -distribution.

Therefore, the unique solution u of (1.1) allows the representation

$$u(x) = \langle G(x, \cdot), f(\cdot) \rangle. \quad (2.2)$$

Similarly, any sufficiently smooth v allows the representation

$$v(x) = \varepsilon \langle \nabla v, \nabla G(x, \cdot) \rangle + \langle \operatorname{div}(\mathbf{a}v) + bv, G(x, \cdot) \rangle. \quad (2.3)$$

Setting $v := u_h$ in (2.3) and then subtracting (2.2), we immediately arrive at the error representation

$$(u_h - u)(x) = \varepsilon \langle \nabla u_h, \nabla G(x, \cdot) \rangle + \langle \operatorname{div}(\mathbf{a}u_h) + bu_h - f, G(x, \cdot) \rangle. \quad (2.4)$$

Similar to Demlow & Kopteva (2016), we shall rely on a number of bounds on the Green's function in which the dependence on the singular perturbation parameter ε is shown explicitly. It is worth noting

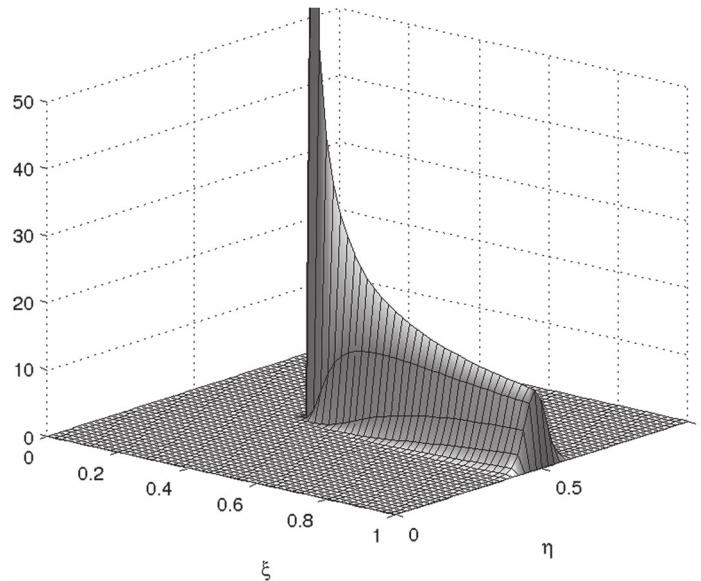


FIG. 1. Typical anisotropic behaviour of the Green's function $G(x, y; \xi, \eta)$ for problem (1.1) in $\Omega = (0, 1)^2$ with $\mathbf{a} = [-1, 0]$, $b = 0$, $(x, y) = (\frac{1}{3}, \frac{1}{2})$ and $\varepsilon = 10^{-3}$.

that Demlow & Kopteva (2016) addressed singularly perturbed equations of reaction–diffusion type, for which the Green's function in the unbounded domain is (almost) radially symmetric and exponentially decaying away from the singular point. By contrast, the Green's function for the convection–diffusion problem (1.1) exhibits a much more complex *anisotropic* structure (see Fig. 1).

We shall require the following bounds on the Green's function G from (2.1):

$$\|G(x, \cdot)\|_{1, \Omega} + \varepsilon^{1/2}|G(x, \cdot)|_{1, 1, \Omega} \lesssim 1, \quad (2.5a)$$

$$|G(x, \cdot)|_{1, 1; B(x, \rho) \cap \Omega} \lesssim \varepsilon^{-1} \rho, \quad (2.5b)$$

$$|G(x, \cdot)|_{2, 1; \Omega \setminus B(x, \rho)} \lesssim \varepsilon^{-1} (\ln(2 + \varepsilon/\rho) + |\ln \varepsilon|), \quad (2.5c)$$

as well as, occasionally,

$$\|\mathbf{a} \cdot \nabla G(x, \cdot)\|_{1, \Omega} \lesssim 1 + |\ln \varepsilon|. \quad (2.6)$$

Here $x \in \Omega$ and $\rho > 0$ are arbitrary, while $B(x, \rho)$ denotes the ball of radius ρ centered at x . Note that (2.6) follows from (2.5). Indeed, $\|\mathbf{a} \cdot \nabla G\|_{1; B(x, \varepsilon) \cap \Omega} \lesssim 1$ follows from (2.5b), while $\|\mathbf{a} \cdot \nabla G\|_{1; \Omega \setminus B(x, \varepsilon)}$ is easily bounded using (2.5c) and (2.5a) combined with the differential equation from (2.1).

In order to gain additional insight into the scaling in the bounds given in (2.5) and (2.6), note that the fundamental solution on \mathbb{R}^3 with $\mathbf{a} = [-a_1, 0, 0]$ is given by $G_{\mathbb{R}^3}(\mathbf{x}, \xi) = \frac{1}{4\pi\varepsilon} \frac{\exp(\frac{1}{2}a_1(\xi_1 - x_1 - r)/\varepsilon)}{r}$, where $r = \sqrt{(\xi_1 - x_1)^2 + (\xi_2 - x_2)^2 + (\xi_3 - x_3)^2}$. The scalings observed above may be directly computed from this function. The free-space fundamental solution for convection-dominated problems in two

space dimensions may also be written down explicitly and scalings computed from it, but the expression is more complex. Note also that the bounds (2.5) are isotropic, while a sharper bound for the convective derivative is given in (2.6). This reflects the anisotropic nature of the Green's function, which can be explicitly seen in the above expression for $G_{\mathbb{R}^3}$.

Note that, although the bounds (2.5) appear as an assumption for our results below, they were rigorously proved in Franz & Kopteva (2011a, 2012, 2022), and are also shown to be sharp in Franz & Kopteva (2011b) for a particular case of $|\mathbf{a}| = |a_1|$ in rectangular and cubic domains. We hypothesize that similar results would hold on more complicated domains such as those with reentrant corners, since similar results hold for standard elliptic and singularly perturbed reaction–diffusion problems on nonconvex as well as convex domains (Demlow & Kopteva, 2016). However, the proof techniques used for convection–diffusion problems are different than in these other cases. Extension to more complex domains would be technically challenging, and the form of the results is not completely clear. We also comment on the assumption $|\mathbf{a}| > 0$ made following (1.1). This condition is needed in the proofs of (2.5) given in Franz & Kopteva (2011a, 2012, 2022). These proofs are substantially different than those given for scaled Green's function estimates for reaction–diffusion problems in Demlow & Kopteva (2016), and it is not clear how to bridge the gap between these different techniques in order to approach the case of a convection coefficient \mathbf{a} which sometimes vanishes. However, the scaling obtained in (2.5) for convection–diffusion problems is very similar to that observed for reaction–diffusion problems, so it seems likely that (2.5) are also valid under a weaker assumption that $|\mathbf{a}| + b > 0$.

2.3 Finite element space

Let \mathcal{T} be a shape-regular and conforming simplicial partition of Ω , with \mathcal{E} denoting the set of all interior $(n-1)$ -dimensional element faces. Let the finite element space $S_h \subset H_0^1(\Omega)$ be the set of functions that are continuous on Ω , equal to 0 on $\partial\Omega$ and polynomials of degree at most r on each $T \in \mathcal{T}$, where $r \geq 1$ is a fixed polynomial degree.

Similarly to Demlow & Georgoulis (2012); Demlow & Kopteva (2016), we shall employ the Scott–Zhang interpolant, denoted G_h , of the Green's function $G(\cdot) := G(x, \cdot)$ from (2.1) (where $x \in \Omega$ remains fixed). We let G_h lie in the space of continuous piecewise-linear functions with respect to \mathcal{T} . Then $G_h \in S_h$ for any $r \geq 1$, and it satisfies the local stability and approximation property

$$|G - G_h|_{k,1;T} \lesssim h_T^{j-k} |G|_{j,1;\omega_T} \quad \forall T \in \mathcal{T}, \quad 0 \leq k \leq j \leq 2, \quad (2.7)$$

whenever the right-hand side of (2.7) is defined. Here, h_T is the diameter of element T , while ω_T denotes the standard patch of elements in \mathcal{T} touching T (including T).

3. A posteriori error estimation in the conforming case

3.1 Finite element method and error indicators

Introduce the standard bilinear form associated with (1.1):

$$\mathcal{B}(u, v) := \varepsilon \langle \nabla u, \nabla v \rangle + \langle \operatorname{div}(\mathbf{a}u) + bu, v \rangle. \quad (3.1)$$

The standard conforming finite element method is then given by:

$$\text{Find } u_h \in S_h : \quad \mathcal{B}(u_h, v_h) = \langle f, v_h \rangle \quad \forall v_h \in S_h. \quad (3.2)$$

We emphasize that this basic finite element method is not generally practical in the singularly perturbed case $\varepsilon \ll 1$ (as unstabilized finite element solutions typically exhibit non-physical oscillations unless h_T/ε is sufficiently small). We first study this unstabilized method mainly in order to understand the structure of the error. Stabilized schemes that are more practically relevant for singularly perturbed problems are considered below.

We shall use an *a posteriori* error indicator defined $\forall T \in \mathcal{T}$ by

$$\eta_\infty(T) := \alpha_T \|\varepsilon \Delta u_h - \operatorname{div}(\mathbf{a} u_h) - b u_h + f\|_{\infty;T} + \beta_T \|\llbracket \nabla u_h \rrbracket\|_{\infty;\partial T \setminus \partial \Omega}, \quad (3.3a)$$

$$\alpha_T := \min\{1, \ell_h \varepsilon^{-1} h_T^2\}, \quad \beta_T := \min\{\varepsilon^{1/2}, \ell_h h_T\} \quad \text{in } T \in \mathcal{T}, \quad (3.3b)$$

where the definition of the logarithmic factor $\ell_h := 1 + \ln(2 + \varepsilon h^{-1}) + |\ln \varepsilon|$, with $h := \min_{T \in \mathcal{T}} h_T$, is motivated by the logarithmic factors in (2.5). Here, we also use the standard notation $\llbracket \nabla u_h \rrbracket := \nabla u_h^+ \cdot n^+ + \nabla u_h^- \cdot n^-$ on a face shared by two elements $T^+, T^- \in \mathcal{T}$, with their respective outward normal unit vectors n^+ and n^- .

Following the analytical techniques used in Demlow & Kopteva (2016) to prove similar maximum-norm *a posteriori* error estimates for singularly perturbed reaction–diffusion problems, we shall derive (see Lemma 1) an *a posteriori* upper bound of the form

$$\|u - u_h\|_{\infty;\Omega} \lesssim \max_{T \in \mathcal{T}} \eta_\infty(T). \quad (3.4)$$

In the reaction–diffusion case, it was also possible to prove the corresponding lower *a posteriori* bounds (efficiency estimates) (Demlow & Kopteva, 2016). However, when the convection term $\operatorname{div}(\mathbf{a} u)$ is present in the equation, one cannot expect to prove a standard lower *a posteriori* bound of the form $\eta_\infty(T) \lesssim \|u - u_h\|_{\infty;\omega_T} + \text{osc}$, where osc is a data oscillation term. Numerical experiments outlined below confirm that such a standard lower bound indeed does not hold. Instead, our efficiency analysis below reveals that the error bound in (3.4) holds true for the error in a stronger norm, with a certain seminorm of $\mathbf{a} \cdot \nabla(u - u_h)$ added in the left-hand side. Furthermore, we shall show that the latter version of (3.4), with the error measured in this new stronger norm, is efficient.

3.2 Reliability

LEMMA 1 Under assumptions (2.5) on G , the error of the computed solution u_h from (3.2) satisfies (3.4) with the *a posteriori* error indicators $\eta_\infty(T)$ as defined in (3.3).

Proof. To estimate the error at any fixed $x \in \Omega$, with slight abuse of notation, let $G(\cdot) := G(x, \cdot)$ be the Green’s function from (2.1). Recall the error representation (2.4) and subtract (3.2) with $v_h := G_h \in S_h$. Then, with the notation $g := G - G_h$, one gets

$$(u_h - u)(x) = \varepsilon \langle \nabla u_h, \nabla g \rangle + \langle \operatorname{div}(\mathbf{a} u_h) + b u_h - f, g \rangle. \quad (3.5)$$

Next, a standard integration by parts in each $T \in \mathcal{T}$ yields

$$(u_h - u)(x) = \langle -\varepsilon \Delta_h u_h + \operatorname{div}(\mathbf{a} u_h) + b u_h - f, g \rangle + \frac{1}{2} \sum_{T \in \mathcal{T}} \int_{\partial T \setminus \partial \Omega} \varepsilon \llbracket \nabla u_h \rrbracket g. \quad (3.6)$$

A comparison of (3.6) and the desired estimate (3.3), (3.4) shows that it suffices to prove that

$$I + II := \|\alpha_T^{-1} g\|_{1;\Omega} + \sum_{T \in \mathcal{T}} \varepsilon \beta_T^{-1} \|g\|_{1;\partial T} \lesssim 1. \quad (3.7)$$

When estimating I and II , we shall, to a degree, follow Demlow & Kopteva (2016, §3.2). A special treatment will be required for the elements in $\mathcal{T}_0 := \{T \in \mathcal{T} : \omega_T \cap B(x, ch_{T_0}) \neq \emptyset\}$, where $T_0 \ni x$. By shape regularity, we may choose $c > 0$ sufficiently small so that the number of these elements is uniformly bounded, and thus, by shape regularity, $h_T \simeq h_{T_0} \forall T \in \mathcal{T}_0$.

For I , by (3.3b), note that $\alpha_T^{-1} \simeq 1 + \ell_h^{-1} \varepsilon h_T^{-2}$, while for $g = G - G_h$, in view of (2.7), one has $\|g\|_{1;T} \lesssim \min\{\|G\|_{1;\omega_T}, h_T \|\nabla G\|_{1;\omega_T}, h_T^2 |G|_{2,1;\omega_T}\}$. Hence,

$$I = \|\alpha_T^{-1} g\|_{1;\Omega} \lesssim \|G\|_{1;\Omega} + \ell_h^{-1} \varepsilon \left(|G|_{2,1;\Omega \setminus B(x, ch_{T_0})} + \sum_{T \in \mathcal{T}_0} h_T^{-1} \|\nabla G\|_{1;\omega_T} \right) \lesssim 1, \quad (3.8)$$

where α_T is understood as an elementwise-defined piecewise-constant weight. Here, we used the bound (2.5a) for $\|G\|_{1;\Omega}$, and then (2.5c) for $|G|_{2,1;\Omega \setminus B(x, ch_{T_0})}$. For each $T \in \mathcal{T}_0$, we employed (2.5b) with the ball $B(x, \bar{C}h_T) \supset \omega_T$, with a sufficiently large constant \bar{C} depending only on the shape regularity of \mathcal{T} .

For II , we employ a scaled trace theorem in the form $\|g\|_{1;\partial T} \lesssim \|\nabla g\|_{1;T} + h_T^{-1} \|g\|_{1;T}$. Combining this with (2.7) yields $\|g\|_{1;\partial T} \lesssim \min\{\|\nabla G\|_{1;\omega_T}, h_T |G|_{2,1;\omega_T}\}$. Note also that $\beta_T^{-1} \simeq \varepsilon^{-1/2} + \ell_h^{-1} h_T^{-1}$, so $\varepsilon \beta_T^{-1} \simeq \varepsilon^{1/2} + \varepsilon \ell_h^{-1} h_T^{-1}$. Combining these observations, one gets

$$II \lesssim \varepsilon^{1/2} \|\nabla G\|_{1;\Omega} + \varepsilon \ell_h^{-1} \left(|G|_{2,1;\Omega \setminus B(x, ch_{T_0})} + \sum_{T \in \mathcal{T}_0} h_T^{-1} \|\nabla G\|_{1;\omega_T} \right) \lesssim 1.$$

Compared with the above estimation of I , we additionally used $\varepsilon^{1/2} \|\nabla G\|_{1;\Omega} \lesssim 1$, in view of (2.5a). This completes the proof of (3.7). \square

REMARK 1 (Nonhomogeneous Dirichlet and Neumann boundary conditions). If $u = g$ and $u_h = g_h$ on $\partial\Omega$, then (3.4) holds with an additional term $\|g - g_h\|_{L_\infty(\partial\Omega)}$ added to the right-hand side assuming sufficient regularity of g . In particular, denoting $e = u - u_h$, let $e = e_{int} + e_\partial$ with $e_\partial = g - g_h$ on $\partial\Omega$, satisfying $\varepsilon \langle \nabla e_\partial, \nabla v \rangle + (\operatorname{div}(\mathbf{a}e_\partial) + b, v) = 0$, $v \in H_0^1(\Omega)$. Then $\|e_\partial\|_{\infty;\Omega} \leq \|g - g_h\|_{\infty;\Omega}$ by the weak maximum principle, and one can show in a manner similar to above that $\|e_{int}\|_{\infty;\Omega} \lesssim \max_{T \in \mathcal{T}} \eta_\infty(T)$. The case of Neumann boundary conditions is less clear. We are unaware of *a posteriori* maximum norm bounds in the literature for Neumann boundary conditions even for symmetric or non-singularly perturbed problems. Some initial groundwork for the present singularly perturbed convection–diffusion case is contained in Franz & Kopteva (2022), where Green’s functions estimates are proved assuming homogeneous Neumann conditions along the characteristic boundaries.

3.3 Efficiency of the volume residual

We start with the volume residual term in (3.3). The following notation will be used. In Ω , let $e := u - u_h$, and then define the residual

$$R_h := -\varepsilon \Delta u_h + \operatorname{div}(\mathbf{a}u_h) + bu_h - f \quad \text{in any } T \in \mathcal{T}. \quad (3.9)$$

Additionally, for any $T \in \mathcal{T}$, let $R_{h,T}$ be the L_2 projection of R_h over T onto the space of polynomials $\mathbb{P}^{r-1}(T)$ of degree $r-1$. Also, using the barycentric coordinates $\{\lambda_i\}_{i=1}^{n+1}$ associated with T , define a standard bubble function $b_T := \prod_{i=1}^{n+1} \lambda_i^2$.

LEMMA 2 For any $u_h \in S_h$ in any $T \in \mathcal{T}$, one has

$$\alpha_T \|R_h\|_{\infty;T} \lesssim \ell_h \|e\|_{\infty;T} + \alpha_T \sup_{\psi \in \mathbb{P}^{r-1}(T): \|\psi\|_{1;T}=1} \left| \int_T b_T \psi \mathbf{a} \cdot \nabla e \right| + \alpha_T \|R_h - R_{h,T}\|_{\infty;T}. \quad (3.10)$$

Proof. For the bubble function $b_T = \prod_{i=1}^{n+1} \lambda_i^2$, standard arguments show that $\|R_{h,T}\|_{\infty;T} \|R_{h,T}\|_{1;T} \lesssim \|R_{h,T}\|_{2;T}^2 \lesssim \int_T b_T R_{h,T}^2$ (Verfürth, 2013, §1.3.4). Next, for the function $w_T := b_T R_{h,T} \|R_{h,T}\|_{1;T}^{-1} \in \mathbb{P}^{2n+r+1}(T)$, note that both w_T and ∇w_T vanish on ∂T , while $h_T^2 \|\Delta w_T\|_{1;T} \lesssim \|w_T\|_{1;T} \lesssim 1$. Additionally, one gets $\|R_{h,T}\|_{\infty;T} \lesssim \int_T w_T R_{h,T}$, which yields

$$\|R_h\|_{\infty;T} \lesssim \int_T w_T R_h + \|R_h - R_{h,T}\|_{\infty;T}. \quad (3.11)$$

Here, $R_h = \varepsilon \Delta e - \operatorname{div}(\mathbf{a}e) - be$ in terms of $e = u - u_h$, so integrating by parts twice the term with Δ , one gets

$$\begin{aligned} \int_T w_T R_h &= \int_T w_T (\varepsilon \Delta e - \operatorname{div}(\mathbf{a}e) - be) \\ &= \int_T (\varepsilon \Delta w_T - (\operatorname{div} \mathbf{a} + b) w_T) e - \int_T w_T \mathbf{a} \cdot \nabla e. \end{aligned} \quad (3.12)$$

Note that $\|\varepsilon \Delta w_T - (\operatorname{div} \mathbf{a} + b) w_T\|_{1;T} \lesssim \varepsilon h_T^{-2} + 1 \lesssim \alpha_T^{-1} \ell_h$. Also, with $\psi_T := R_{h,T} / \|R_{h,T}\|_{1;T}$, one has $w_T = b_T \psi_T$ in the final term of (3.12), where $\psi_T \in \mathbb{P}^{r-1}$ and $\|\psi_T\|_{1;T} = 1$. It remains to combine these two observations with (3.11) and (3.12), both multiplied by α_T . \square

REMARK 2 Instead of arguing as in (3.12), one could integrate the convection term by parts in order to obtain $-\int_T w_T \operatorname{div}(\mathbf{a}e) = \int_T e \mathbf{a} \cdot \nabla w_T \lesssim \|\nabla w_T\|_{1;T} \|e\|_{\infty;T} \lesssim h_T^{-1} \|e\|_{\infty;T}$. Note that $h_T^{-1} \alpha_T = \min\{h_T^{-1}, \ell_h h_T \varepsilon^{-1}\}$, which is not bounded by ℓ_h until $h_T \varepsilon^{-1} \leq 1$. This argument does not yield a suitable efficiency result, thus the need for the additional term in (3.10).

It is convenient to denote the seminorm of $\mathbf{a} \cdot \nabla e$ present in (3.10) by

$$|\mathbf{a} \cdot \nabla e|_{*,T} := \alpha_T \sup_{\psi \in \mathbb{P}^{r-1}(T): \|\psi\|_{1;T}=1} \left| \int_T b_T \psi \mathbf{a} \cdot \nabla e \right| \quad \forall T \in \mathcal{T}. \quad (3.13)$$

COROLLARY 1 For any $u_h \in S_h$ in any $T \in \mathcal{T}$, one has

$$\alpha_T \|R_h\|_{\infty;T} \lesssim \ell_h \|e\|_{\infty;T} + |\mathbf{a} \cdot \nabla e|_{*,T} + \alpha_T \|R_h - R_{h,T}\|_{\infty;T}, \quad (3.14)$$

$$|\mathbf{a} \cdot \nabla e|_{*,T} \lesssim \ell_h \|e\|_{\infty;T} + \alpha_T \|R_h\|_{\infty;T}. \quad (3.15)$$

Proof. The first assertion (3.14) is equivalent to (3.10). To prove the second, note that (3.12), as well as most estimates in the proof of Lemma 2 that involve $w_T = b_T \psi_T$, also hold true for a more general $w_T := b_T \psi$ with an arbitrary $\psi \in \mathbb{P}^{r-1}(T)$ such that $\|\psi\|_{1;T} = 1$. The only exceptions are (3.11) and the related bound $\|R_{h,T}\|_{\infty;T} \lesssim \int_T w_T R_{h,T}$, which are no longer true. Instead, we shall now employ $|\int_T w_T R_h| \lesssim \|R_h\|_{\infty;T}$. Hence, from (3.12) with this more general w_T , one gets the following version of (3.10):

$$\alpha_T \left| \int_T b_T \psi \mathbf{a} \cdot \nabla e \right| \lesssim \ell_h \|e\|_{\infty;T} + \alpha_T \left| \int_T w_T R_h \right|,$$

and then the desired (3.15). \square

REMARK 3 One can extend the above Lemma 2 and Corollary 1 for a slightly simpler bubble function $b_T = \prod_{i=1}^{n+1} \lambda_i$ (instead of $b_T = \prod_{i=1}^{n+1} \lambda_i^2$). Note that in this case, (3.12) will additionally involve $\int_{\partial T} \varepsilon e \nabla w_h \cdot n$, so the proof will be slightly more involved.

3.4 Efficiency of the jump residual

Now we turn to the jump residual term in (3.3), for which following Demlow & Kopteva (2016), we modify the standard edge residual efficiency proof by employing subscale mesh elements when h_T does not resolve $\sqrt{\varepsilon}$.

The following notation will be used. For an interior face $E \in \mathcal{E}$, shared by T^+ and T^- in \mathcal{T} , construct two, not necessarily shape-regular, sub-simplices $T_E^\pm \subset T^\pm$ that share the entire face E and satisfy

$$|T_E^\pm| = \hat{h}_E |E|, \quad \hat{h}_E := \min\{\varepsilon^{1/2}, |T^+||E|^{-1}, |T^-||E|^{-1}\} \simeq \min\{\varepsilon^{1/2}, h_{T^\pm}\}.$$

To be more specific, unless $T_E^\pm = T^\pm$, one may impose that the vertex of T_E^\pm opposite to E lies on the corresponding median line in T^\pm . The simplices T_E^\pm do not necessarily satisfy either a minimum or maximum angle condition, so it is necessary to take extra care in our arguments below at a couple of points.

Next let $\{\lambda_i^\pm\}_{i=1}^{n+1}$ be the barycentric coordinates in T_E^\pm . Assume that $\lambda_{n+1}^\pm|_E \equiv 0$ so that $\{\lambda_i^\pm\}_{i=1}^n$ are the barycentric coordinates associated with the vertices of E . For the tangential gradient along E , we have $|\nabla_E \lambda_i^\pm| \lesssim |\text{diam } E|^{-1} \lesssim \hat{h}_E^{-1}$, whereas in the direction perpendicular to E the height of the triangle is $2\hat{h}_E$ and so $|\partial_{E^\perp} \lambda_i^\pm| \lesssim \hat{h}_E^{-1}$ with constant, depending on the shape regularity of T^\pm . Finally, define a standard face bubble $b_E := \prod_{i=1}^n (\lambda_i^\pm)^2$ on T_E^\pm , and the seminorm

$$|\mathbf{a} \cdot \nabla e|_{*,E} := \sup_{\varphi \in \mathbb{P}^{r-1}(E): \|\varphi\|_{1;E}=1} \hat{h}_E^{-1} \left| \int_{T_E^+ \cup T_E^-} \alpha_T b_E \varphi \mathbf{a} \cdot \nabla e \right|. \quad (3.16)$$

Here, inside the volume integral, $\varphi \in \mathbb{P}^{r-1}(E)$ is understood as extended to \mathbb{R}^n such that it remains constant in the direction normal to E .

LEMMA 3 Let $u_h \in S_h$. For any interior face $E = \partial T^+ \cap \partial T^-$, shared by T^+ and T^- in \mathcal{T} , with $\beta_E := \min\{\varepsilon^{1/2}, \ell_h h_{T^+}, \ell_h h_{T^-}\} \simeq \beta_{T^\pm}$, one has

$$\beta_E \|\llbracket \nabla u_h \rrbracket\|_{\infty;E} \lesssim \ell_h \|e\|_{\infty;T^+ \cup T^-} + |\mathbf{a} \cdot \nabla e|_{*,E} + \|\alpha_T R_h\|_{\infty;T^+ \cup T^-}. \quad (3.17)$$

Proof. Set $J_h := \llbracket \nabla u_h \rrbracket \in \mathbb{P}^{r-1}(E)$, for which standard arguments show that $\|J_h\|_{\infty;E} \|J_h\|_{1;E} \lesssim \|J_h\|_{2;E}^2 \lesssim \int_E b_E J_h^2$ (Verfürth, 2013, §1.3.4). So using the function $w_E := b_E J_h \|J_h\|_{1;E}^{-1} \in \mathbb{P}^{2n+r-1}(E)$, one gets $\|J_h\|_{\infty;E} \lesssim \int_E J_h w_E$. Note that J_h , as a polynomial function on E , can be extended to the entire $(n-1)$ -dimensional plane that contains E . Next, extend it to \mathbb{R}^n by letting J_h remain constant in the direction normal to E . This extends w_E to $T_E^+ \cup T_E^-$, with both w_E and ∇w_E vanishing on $\partial T_E^\pm \setminus E$. Note also that (as elaborated in Remark 4 below)

$$\hat{h}_E^2 \|\Delta w_E\|_{1;T_E^\pm} \lesssim \|w_E\|_{1;T_E^\pm} \lesssim \hat{h}_E, \quad \hat{h}_E \|\llbracket \nabla w_E \rrbracket\|_{1;E} \lesssim 1. \quad (3.18)$$

A version of (3.12) (taking into account, when carrying out the integration by parts twice, that w_E and ∇w_E do not vanish on E) yields

$$\begin{aligned} \int_{T_E^+ \cup T_E^-} w_E R_h &= \int_{T_E^+ \cup T_E^-} (\varepsilon \Delta w_E - (\operatorname{div} \mathbf{a} + b) w_E) e - \int_{T_E^+ \cup T_E^-} w_E \mathbf{a} \cdot \nabla e \\ &\quad - \int_E \varepsilon (J_h w_E + \llbracket \nabla w_E \rrbracket e), \end{aligned} \quad (3.19)$$

where we used $\llbracket \nabla e \rrbracket = -\llbracket \nabla u_h \rrbracket = -J_h$ on E . Next, recalling that $\|J_h\|_{\infty;E} \lesssim \int_E J_h w_E$, and also (3.18), one gets

$$\varepsilon \|J_h\|_{\infty;E} \lesssim (\varepsilon \hat{h}_E^{-1} + \hat{h}_E) \|e\|_{\infty;T^+ \cup T^-} + \left| \int_{T_E^+ \cup T_E^-} w_E \mathbf{a} \cdot \nabla e \right| + \hat{h}_E \|R_h\|_{\infty;T^+ \cup T^-}.$$

Multiply this by $\varepsilon^{-1} \beta_E$ and note that, in view of $\hat{h}_E \leq \varepsilon^{1/2}$, one has $\varepsilon \hat{h}_E^{-1} + \hat{h}_E \leq 2\varepsilon \hat{h}_E^{-1}$, so $\varepsilon^{-1} \beta_E (\varepsilon \hat{h}_E^{-1} + \hat{h}_E) \lesssim \beta_E \hat{h}_E^{-1} \lesssim \ell_h$. Also, for each $T = T^\pm$, with $\alpha_T = \min\{1, \ell_h \varepsilon^{-1} h_T^2\}$, note that $\beta_E \hat{h}_E \lesssim \min\{\varepsilon, \ell_h h_T^2\} = \varepsilon \alpha_T$ yields $\varepsilon^{-1} \beta_E \lesssim \hat{h}_E^{-1} \alpha_T$. So

$$\beta_E \|J_h\|_{\infty;E} \lesssim \ell_h \|e\|_{\infty;T^+ \cup T^-} + \hat{h}_E^{-1} \left| \int_{T_E^+ \cup T_E^-} \alpha_T w_E \mathbf{a} \cdot \nabla e \right| + \|\alpha_T R_h\|_{\infty;T^+ \cup T^-}.$$

It remains to note that $w_E = b_E \varphi_E$, where $\varphi_E := J_h \|J_h\|_{1;E}^{-1}$ satisfies $\|\varphi_E\|_{1;E} = 1$. \square

COROLLARY 2 Under the conditions of Lemma 3, one has

$$\beta_E \|\llbracket \nabla u_h \rrbracket\|_{\infty;E} \lesssim \ell_h \|e\|_{\infty;T^+ \cup T^-} + |\mathbf{a} \cdot \nabla e|_{*,E} + \|\alpha_T R_h\|_{\infty;T^+ \cup T^-}, \quad (3.20)$$

$$|\mathbf{a} \cdot \nabla e|_{*,E} \lesssim \ell_h \|e\|_{\infty;T^+ \cup T^-} + \beta_E \|\llbracket \nabla u_h \rrbracket\|_{\infty;E} + \|\alpha_T R_h\|_{\infty;T^+ \cup T^-}. \quad (3.21)$$

Proof. The first assertion (3.20) is equivalent to (3.17). For the second (similarly to the proof of Corollary 1), we note that (3.19), as well as most evaluations in the proof of Lemma 3, holds true for a more general $w_E := b_E \varphi$ with an arbitrary $\varphi \in \mathbb{P}^{r-1}(E)$ subject to $\|\varphi\|_{1;E} = 1$. The main exception is $\|J_h\|_{\infty;E} \lesssim \int_E J_h w_E$, which is no longer true, and instead of which we now employ $|\int_E J_h w_E| \lesssim \|J_h\|_{\infty;E}$, which yields (3.21). \square

REMARK 4 ((3.18) on anisotropic elements). Consider a less standard (3.18) for T_E^+ (as T_E^- is similar). If $\hat{h}_E = |T^+| |E|^{-1}$, then $T_E^+ = T^+$ is shape-regular with $\hat{h}_E \simeq h_{T^+}$, so (3.18) is standard. Otherwise, i.e., if $\hat{h}_E < |T^+| |E|^{-1}$, set $\theta := \hat{h}_E / \{|T^+| |E|^{-1}\} \simeq \hat{h}_E / h_{T^+}$ and define an affine transformation from the shape-regular T^+ to the anisotropic T_E^+ by

$$\hat{x} = x - (P_0 - \hat{P}_0) \lambda_0(x).$$

Here, P_0 and \hat{P}_0 are the vertices in T^+ and T_E^+ opposite to E , while λ_0 is the barycentric coordinate in T^+ associated with P_0 . As $|T_E^+| = \theta |T^+|$, the Jacobian determinant of this transformation equals θ . Additionally, in the shape-regular T^+ , one has $|\nabla \lambda_0| \simeq h_{T^+}^{-1}$ and $|P_0 - \hat{P}_0| \lesssim h_{T^+}$, so all elements in the transformation matrix are $\lesssim 1$. Now, by Cramer's rule, all entries of the inverse transformation matrix are $\lesssim \theta^{-1}$, so for a generic v , one gets $|\nabla_{\hat{x}} v| \lesssim \theta^{-1} |\nabla_x v|$, which yields (3.18) after application of a standard inverse inequality for shape-regular elements.

3.5 Seminorms of the convective derivative and overall efficiency result

Combining the definition of $\eta_\infty(T)$ in (3.3) with (3.14) and (3.20) yields in summary that

$$\eta_\infty(T) \lesssim \ell_h \|e\|_{\infty; \omega_T} + \max_{T' \subset \omega_T} |\mathbf{a} \cdot \nabla e|_{*, T'} + \max_{E \subset \partial T} |\mathbf{a} \cdot \nabla e|_{*, E} + \text{osc}(\alpha_T R_h, \omega_T), \quad (3.22)$$

where

$$\text{osc}(\alpha_T R_h, \omega_T) := \max_{T' \subset \omega_T} \|\alpha_{T'}(R_h - R_{h,T'})\|_{\infty; T'}. \quad (3.23)$$

Recall that here we used the elementwise seminorm definitions (3.13) and (3.16):

$$\begin{aligned} |\mathbf{a} \cdot \nabla e|_{*, T} &:= \alpha_T \sup_{\psi \in \mathbb{P}^{r-1}(T): \|\psi\|_{1;T}=1} \left| \int_T b_T \psi \mathbf{a} \cdot \nabla e \right|, \\ |\mathbf{a} \cdot \nabla e|_{*, E} &:= \sup_{\varphi \in \mathbb{P}^{r-1}(E): \|\varphi\|_{1;E}=1} \hat{h}_E^{-1} \left| \int_{T_E^+ \cup T_E^-} \alpha_T b_E \varphi \mathbf{a} \cdot \nabla e \right|. \end{aligned}$$

These are seminorms because it is possible that $\mathbf{a} \cdot \nabla e \neq 0$ is orthogonal to $\mathbb{P}^{r+2n+1} \ni b_T \psi, b_E \varphi$ over the relevant volumes. We now define the following related global seminorm of the convective derivative:

$$|\mathbf{a} \cdot \nabla e|_* := \max_{T \subset \mathcal{T}} |\mathbf{a} \cdot \nabla e|_{*, T} + \max_{E \subset \mathcal{E}} |\mathbf{a} \cdot \nabla e|_{*, E}. \quad (3.24)$$

From (3.22) and (3.23), one immediately gets the global estimate

$$\max_{T \in \mathcal{T}} \eta_\infty(T) \lesssim \ell_h \|e\|_{\infty; \Omega} + |\mathbf{a} \cdot \nabla e|_* + \text{osc}(\alpha_T R_h, \Omega).$$

On the other hand, combining (3.4) with (3.15) and (3.21), one gets $|\mathbf{a} \cdot \nabla e|_* \lesssim \ell_h \|e\|_{\infty; \Omega} + \max_{T \in \mathcal{T}} \eta_\infty(T)$, and then

$$\|e\|_{\infty; \Omega} + \ell_h^{-1} |\mathbf{a} \cdot \nabla e|_* \lesssim \max_{T \in \mathcal{T}} \eta_\infty(T).$$

Combining these relationships with $\text{osc}(\alpha_T R_h, T) \leq \eta_\infty(T)$, we have proved the following for the standard conforming finite element method.

THEOREM 1 Under assumptions (2.5) on G , the error of the computed solution u_h from (3.2) satisfies

$$\begin{aligned} \|u - u_h\|_{\infty; \Omega} + \ell_h^{-1} |\mathbf{a} \cdot \nabla(u - u_h)|_* + \text{osc}(\alpha_T R_h, \Omega) &\lesssim \max_{T \in \mathcal{T}} \eta_\infty(T) \\ &\lesssim \ell_h \|u - u_h\|_{\infty; \Omega} + |\mathbf{a} \cdot \nabla(u - u_h)|_* + \text{osc}(\alpha_T R_h, \Omega). \end{aligned} \quad (3.25)$$

Thus, while our original intention was to bound $\|u - u_h\|_{\infty; \Omega}$, the estimator that we have naturally derived is *not* efficient for this norm. Deriving an upper bound for η_∞ instead requires inclusion of the $|\cdot|_*$ seminorm. Recall also that a similar situation is observed when bounding energy norms in singularly perturbed convection–diffusion problems, where a dual norm of the convective derivative plays a similar role in the analysis (Tobiska & Verfürth, 2015). Our numerical experiments below also highlight the importance of the $|\cdot|_*$ seminorm in *a posteriori* analysis for convection–diffusion problems.

We finish this section with a further discussion of the $|\cdot|_*$ seminorm, as its definition involves multiple terms and its meaning may not be intuitively clear at first glance. Consider first the simpler norm

$$|\mathbf{a} \cdot \nabla e|_{**} := \|\alpha_T \mathbf{a} \cdot \nabla e\|_{\infty; \Omega}, \quad (3.26)$$

of the convective derivative, where $\alpha_T = \min\{1, \ell_h \varepsilon^{-1} h_T^2\} \leq 1$ is understood as an elementwise-defined piecewise-constant weight. The above definitions easily yield that $|\mathbf{a} \cdot \nabla v|_* \lesssim |\mathbf{a} \cdot \nabla v|_{**}$. Thus, our estimator measures the error in a mesh-dependent norm that lies between $\|e\|_{\infty; \Omega}$ and $\|e\|_{\infty; \Omega} + \|\alpha_T \mathbf{a} \cdot \nabla e\|_{\infty; \Omega}$. The $|\cdot|_{**}$ norm is still mesh-dependent, but gives a more transparent measure of the convective derivative of the error. It is not generally true that $|\mathbf{a} \cdot \nabla v|_{**} \lesssim |\mathbf{a} \cdot \nabla v|_*$. It is straightforward to instead prove that $\alpha_T \|P_{r-1}(\mathbf{a} \cdot \nabla e)\|_{\infty; T} \lesssim |\mathbf{a} \cdot \nabla e|_{*, T}$ and thus that

$$|\mathbf{a} \cdot \nabla e|_{**} \lesssim \max_{T \in \mathcal{T}} |\mathbf{a} \cdot \nabla e|_{*, T} + \max_{T \in \mathcal{T}} \alpha_T \|\mathbf{a} \cdot \nabla e - P_{r-1}(\mathbf{a} \cdot \nabla e)\|_{\infty; T}, \quad (3.27)$$

where P_{r-1} is the L_2 projection onto the elementwise polynomials of degree $r-1$. Thus, the $**$ -seminorm bounds the $**$ -norm only up to an oscillation term. In fact, $|\mathbf{a} \cdot \nabla u|_{**}$ and hence $|\mathbf{a} \cdot \nabla e|_{**}$ may even be unbounded in cases where $|\mathbf{a} \cdot \nabla u|_*$ and $|\mathbf{a} \cdot \nabla e|_*$ are finite. In particular, $|\mathbf{a} \cdot \nabla u|_{**} < \infty$ requires $\mathbf{a} \cdot \nabla u \in L_\infty$, whereas $|\mathbf{a} \cdot \nabla u|_* < \infty$ requires only $\mathbf{a} \cdot \nabla u \in L_1$. The latter is true, but not the former, for example, when Ω is a nonconvex polygonal domain and thus ∇u is unbounded at reentrant corners. In the latter case, a minor modification of $|\cdot|_{**}$ by inclusion of an appropriate local bubble again results

in a finite quantity. It is also possible to define other norms between $|\cdot|_*$ and $|\cdot|_{**}$. In doing so, there appears to be a tradeoff between simplicity and transparency on the one hand and fidelity to actual estimator behavior on the other, with no obvious choice doing an outstanding job at both of these tasks.

In spite of its drawbacks, the $**$ norm is easily understood as an elementwise-weighted norm of the error, tracks the $*$ -seminorm closely in some situations and provides clear insight into the convergence behavior of the $*$ seminorm. To illustrate the latter point, assume momentarily for simplicity that the mesh is quasi-uniform with diameter h , ignore logarithmic factors and assume that $h^2 \lesssim \varepsilon$. Then $|\mathbf{a} \cdot \nabla e|_{**} \simeq \frac{h^2}{\varepsilon} \|\mathbf{a} \cdot \nabla e\|_{\infty; \Omega}$. Now, as $h \rightarrow 0$ and $h \lesssim \varepsilon$, we get $|\mathbf{a} \cdot \nabla e|_{**} \ll h \|\mathbf{a} \cdot \nabla e\|_{\infty; \Omega}$. Typically, the maximum norm converges with one power of h faster than the W_∞^1 norm, so we may roughly expect $|\mathbf{a} \cdot \nabla e|_{**}$ to be equivalent to $\|e\|_{\infty; \Omega}$ when $h \simeq \varepsilon$, and for the latter to dominate the former as $h/\varepsilon \rightarrow 0$.

Alternatively, we may integrate by parts (3.13) and (3.16) to obtain

$$\begin{aligned} |\mathbf{a} \cdot \nabla e|_* &\lesssim \max_{T \in \mathcal{T}} \left(\alpha_T h_T^{-1} + \alpha_T \min\{\varepsilon^{1/2}, h_T\}^{-1} \right) \|e\|_{\infty; T} \\ &\lesssim \ell_h \varepsilon^{-1} \max_{T \in \mathcal{T}} \left(\min\{\varepsilon^{1/2}, h_T\} \right) \|e\|_{\infty; T}. \end{aligned} \quad (3.28)$$

Here, we also used $\alpha_T \leq \ell_h \varepsilon^{-1} \min\{\varepsilon, h_T^2\}$. We thus see that up to log factors, $|\mathbf{a} \cdot \nabla e|_* \lesssim \|e\|_{\infty; \Omega}$ when $h_T \leq \varepsilon$ (as the latter also implies $h_T \leq \varepsilon^{1/2}$). In addition, $\frac{|\mathbf{a} \cdot \nabla e|_*}{\|e\|_{\infty; \Omega}} \rightarrow 0$ as $\max_{T \in \mathcal{T}} h_T \rightarrow 0$. In other words, the extended norm in (3.25) is dominated by the maximum norm over areas of Ω where the local mesh size resolves ε . If the problem is not singularly perturbed (i.e., $\varepsilon \simeq 1$), then this heuristic is valid on any mesh.

Both of the preceding analyses indicate that $|\mathbf{a} \cdot \nabla e|_*$ may play an important role in understanding the behavior of the maximum-norm error estimator $\max_{T \in \mathcal{T}} \eta_\infty(T)$ when $h_T \gg \varepsilon$, but diminishes in importance relative to $\|u - u_h\|_{\infty; \Omega}$ and $\max_{T \in \mathcal{T}} \eta_\infty(T)$ as h_T resolves ε . We verify this behavior in our numerical experiments below, and additionally provide a computational comparison between $|\cdot|_*$ and $|\cdot|_{**}$.

4. A posteriori error estimation for stabilized methods

In this section, we explore stabilization schemes and their effects on the above *a posteriori* error estimates. Stabilized methods frequently have the form: find $u_h \in S_h$ such that

$$\mathcal{B}(u_h, v_h) + S_{\mathcal{T}}(u_h, v_h) = \langle f, v_h \rangle \quad \forall v_h \in S_h, \quad (4.1)$$

where the stabilization term is described using $S_{\mathcal{T}} : S_h \times S_h \rightarrow \mathbb{R}$.

In order to develop *a posteriori* error estimates for stabilized schemes of the form (4.1), we imitate the proof of Lemma 1 and use the Green's function $G(\cdot) := G(x, \cdot)$ and its interpolant $G_h \in S_h$. We again recall the error representation (2.4) and subtract (4.1) with $v_h := G_h \in S_h$. Then, with the notation $g := G - G_h$, one gets

$$(u_h - u)(x) = \varepsilon \langle \nabla u_h, \nabla g \rangle + \langle \operatorname{div}(\mathbf{a} u_h) + b u_h - f, g \rangle - S_{\mathcal{T}}(u_h, G_h),$$

i.e., compared with (3.5), we have an additional term $S_{\mathcal{T}}(u_h, G_h)$. Lemma 1 for this case then reads as

$$\|u - u_h\|_{\infty; \Omega} \lesssim \max_{T \in \mathcal{T}} \eta_\infty(T) + \sup_{x \in \Omega} |S_{\mathcal{T}}(u_h, G_h)|. \quad (4.2)$$

Bounds for the last term depend on the stabilization method; we explore some options below. We also note that there is a large literature on stabilization methods that we do not explore here. We shall somewhat follow [Tobiska & Verfürth \(2015\)](#) in our presentation (where a similar analysis is carried out for the energy norm) and refer to that work for more discussion. Hence, our exploration of this topic is cursory and focused only on effects on maximum-norm *a posteriori* error estimation. In particular, we establish that the *a posteriori* error estimation framework described above remains valid for a number of stabilized methods.

4.1 Streamline diffusion method

The streamline diffusion method is a residual-based method introduced in [Hughes & Brooks \(1979\)](#) (see also [Roos et al., 2008](#), §III.3.2.1, and the references therein). Here the stabilization term has the form

$$S_{\mathcal{T}}(u_h, v_h) = \sum_{T \in \mathcal{T}} \delta_T \int_T R_h \mathbf{a} \cdot \nabla v_h, \quad (4.3)$$

where $R_h = -\varepsilon \Delta u_h + \operatorname{div}(\mathbf{a} u_h) + b u_h - f$ is the elementwise residual from (3.9). Here, $\delta_T \geq 0$ is a user-chosen parameter. Note one standard choice ([Brooks & Hughes, 1982](#); [John & Knobloch, 2007](#))

$$\delta_T = h_T a_T^{-1} \xi\left(\frac{1}{2} Pe_T\right), \quad \xi(s) := \coth(s) - s^{-1} \simeq \min\{1, s\}, \quad a_T := \|\mathbf{a}\|_{\infty; T}, \quad (4.4)$$

where we used the local Péclet number $Pe_T := \varepsilon^{-1} a_T h_T$. Note also that the above δ_T , as well as many other standard choices, satisfies the hypothesis of Corollary 3 below.

LEMMA 4 Suppose G satisfies (2.5), and $G_h \in S_h$ is its interpolant from (2.7). Then, for (4.3), one gets

$$|S_{\mathcal{T}}(u_h, G_h)| \lesssim \max_{T \in \mathcal{T}} \left\{ \gamma_T \delta_T \|R_h\|_{\infty; T} \right\}, \quad \gamma_T := \min \left\{ a_T h_T^{-1}, \ell_h (1 + a_T \varepsilon^{-1} h_T) \right\}. \quad (4.5)$$

COROLLARY 3 Suppose that u_h satisfies (4.1), (4.3) with $\delta_T \lesssim h_T \min\{a_T^{-1}, \varepsilon^{-1} h_T\} \forall T \in \mathcal{T}$. Then, under the conditions of Lemma 4, one has $|S_{\mathcal{T}}(u_h, G_h)| \lesssim \max_{T \in \mathcal{T}} \eta_{\infty}(T)$ for any $x \in \Omega$, and, hence, the error bound (3.4).

Proof. In view of (4.2), it suffices to establish the desired bound on $S_{\mathcal{T}}(u_h, G_h)$. For the latter, a comparison of (4.5) with (3.3) shows that it suffices to prove that $\gamma_T \delta_T \lesssim \alpha_T = \min\{1, \ell_h \varepsilon^{-1} h_T^2\}$. From $\gamma_T \lesssim a_T h_T^{-1}$ combined with $\delta_T \lesssim h_T a_T^{-1}$, one immediately gets $\gamma_T \delta_T \lesssim 1$, so it remains to prove that we also have $\gamma_T \delta_T \lesssim \ell_h \varepsilon^{-1} h_T^2$. The latter follows by combining $\gamma_T \lesssim \ell_h (1 + a_T \varepsilon^{-1} h_T) \simeq \ell_h \max\{1, a_T \varepsilon^{-1} h_T\}$ with $\delta_T \lesssim h_T a_T^{-1} \min\{1, a_T \varepsilon^{-1} h_T\}$ (in view of $\min\{1, s\} \max\{1, s\} = s \forall s$). \square

Proof of Lemma 4. A comparison of the desired bound (4.5) with (4.3) shows that it suffices to prove that

$$I^* := \sum_{T \in \mathcal{T}} I_T^* := \sum_{T \in \mathcal{T}} \gamma_T^{-1} \|\mathbf{a} \cdot \nabla G_h\|_{1; T} \lesssim 1.$$

Note that here $\gamma_T^{-1} \simeq h_T a_T^{-1} + \ell_h^{-1} \min\{1, \varepsilon h_T^{-1} a_T^{-1}\}$. Hence, a calculation using $G_h = G - (G - G_h)$ leads to

$$I_T^* \lesssim h_T \|a_T^{-1} \mathbf{a} \cdot \nabla G_h\|_{1;T} + \ell_h^{-1} \|\mathbf{a} \cdot \nabla G\|_{1;T} + \ell_h^{-1} \varepsilon h_T^{-1} \|a_T^{-1} \mathbf{a} \cdot \nabla (G - G_h)\|_{1;T}.$$

Here $|a_T^{-1} \mathbf{a}| \lesssim 1$. Additionally, for the first term, an inverse inequality applied elementwise yields $h_T \|\nabla G_h\|_{1;T} \lesssim \|G_h\|_{1;T} \lesssim \|G\|_{1;\omega_T}$, where we also used (2.7). For the final term, (2.7) implies $\|\nabla(G - G_h)\|_{1;T} \lesssim \min\{\|\nabla G\|_{1;\omega_T}, h_T \|D^2 G\|_{1;\omega_T}\}$. Combining these observations, one now gets

$$I^* \lesssim \|G\|_{1;\Omega} + \ell_h^{-1} \|\mathbf{a} \cdot \nabla G\|_{1;\Omega} + \ell_h^{-1} \varepsilon \left(|G|_{2,1;\Omega \setminus B(x, ch_{T_0})} + \sum_{T \in \mathcal{T}_0} h_T^{-1} \|\nabla G\|_{1;\omega_T} \right),$$

where we again used $T_0 \ni x$ and $\mathcal{T}_0 = \{T \in \mathcal{T} : \omega_T \cap B(x, ch_{T_0}) \neq \emptyset\}$. Most ingredients of the right-hand side have been estimated in (3.8). The remaining $\ell_h^{-1} \|\mathbf{a} \cdot \nabla G\|_{1;\Omega}$ is estimated using (2.6), which yields the desired bound $I^* \lesssim 1$. (Note that (2.6) follows from (2.5).)

4.2 Continuous interior penalty stabilization

We next let u_h satisfy (4.1) with the stabilizing term (Douglas & Dupont, 1976; Burman & Hansbo, 2004) (see also Roos *et al.*, 2008, §III.3.3.2, Tobiska & Verfürth, 2015, §2.2.4 and the references therein)

$$S_{\mathcal{T}}(u_h, v_h) = \sum_{E \in \mathcal{E}} \tau_E \int_E [\![\mathbf{a} \cdot \nabla u_h]\!] [\![\mathbf{a} \cdot \nabla v_h]\!]. \quad (4.6)$$

Here, we used the standard notation $[\![\cdot]\!]$, which, for a generic scalar function v , is defined by $[\![v]\!] := [\![vn_E]\!]$ on any $E \in \mathcal{E}$ using any fixed normal unit vector n_E to E . A user-chosen parameter τ_E typically satisfies

$$\tau_E \lesssim h_E^2. \quad (4.7)$$

Following the analysis in Tobiska & Verfürth (2015, Lemma 2.6), we restrict our consideration to the case of \mathbb{P}^1 elements and, thus, get the following result.

LEMMA 5 Suppose that u_h satisfies (4.1), (4.6), (4.7) with the space S_h of \mathbb{P}^1 elements, and that the functions \mathbf{a} , $\operatorname{div} \mathbf{a}$, b and f in (1.1) are continuous in Ω . Suppose also that G satisfies (2.5), and $G_h \in S_h$ is its interpolant from (2.7). Then, for (4.6), one gets $|S_{\mathcal{T}}(u_h, G_h)| \lesssim \max_{T \in \mathcal{T}} \eta_{\infty}(T)$ for any $x \in \Omega$, and, hence, the error bound (3.4).

Proof. In view of (4.2), it suffices to establish the desired bound on $S_{\mathcal{T}}(u_h, G_h)$. As we consider the case of \mathbb{P}^1 elements, $\varepsilon \Delta u_h = 0$ elementwise for any $u_h \in S_h$. Hence, the residual becomes $R_h = \operatorname{div}(\mathbf{a} u_h) + b u_h - f$, while, in view of the continuity of \mathbf{a} , $\operatorname{div} \mathbf{a}$, b and f , one then gets $[\![\mathbf{a} \cdot \nabla u_h]\!] = [\![R_h]\!]$. Hence, (4.6) leads to

$$|S_{\mathcal{T}}(u_h, G_h)| \lesssim I^{**} \max_{T \in \mathcal{T}} \left\{ \alpha_T \|R_h\|_{\infty;T} \right\}, \quad I^{**} := \sum_{T \in \mathcal{T}} \alpha_T^{-1} h_T^2 \|[\![\mathbf{a} \cdot \nabla G_h]\!]\|_{1;\partial T},$$

where we also used that $h_E \simeq h_T$ for any shape-regular T sharing a face E . As $\alpha_T \|R_h\|_{\infty;T} \leq \eta_\infty(T)$ (in view of (3.3)), it remains to show that $I^{**} \lesssim 1$.

For the latter, first, note that an inverse inequality yields $h_T^2 \|\llbracket \nabla G_h \rrbracket\|_{1;\partial T} \lesssim h_T^j |G_h|_{j,1;\omega_T}$ for $j = 0, 1$, while (2.7) implies $|G_h|_{j,1;T} \lesssim |G|_{j,1;\omega_T}$. On the other hand, $\llbracket \nabla G_h \rrbracket = -\llbracket \nabla g \rrbracket$, where $g = G - G_h$, so a standard scaled trace inequality yields $\|\llbracket \nabla g \rrbracket\|_{1;\partial T} \lesssim \|D^2 g\|_{1;\omega_T} + h_T^{-1} \|\nabla g\|_{1;\omega_T}$, for which (2.7) gives $\|D^2 g\|_{1;T} + h_T^{-1} \|\nabla g\|_{1;T} \lesssim \|G\|_{2,1;\omega_T}$. Combining these observations, one gets

$$h_T^2 \|\llbracket \mathbf{a} \cdot \nabla G_h \rrbracket\|_{1;\partial T} \lesssim \min \left\{ \|G\|_{1;\omega'_T}, h_T |G|_{1,1;\omega'_T}, h_T^2 |G|_{2,1;\omega'_T} \right\}, \quad (4.8)$$

where ω'_T denotes the patch of elements in \mathcal{T} touching ω_T (including those in ω_T). Finally, combining the definition of I^{**} with (4.8) and $\alpha_T^{-1} \simeq 1 + \ell_h^{-1} \varepsilon h_T^{-2}$, one gets

$$I^{**} \lesssim \|G\|_{1,\Omega} + \ell_h^{-1} \varepsilon \left(|G|_{2,1;\Omega \setminus B(x, ch_{T_0})} + \sum_{T \in \mathcal{T}'_0} h_T^{-1} \|\nabla G\|_{1;\omega'_T} \right),$$

where $\mathcal{T}'_0 := \{T \in \mathcal{T} : \omega'_T \cap B(x, ch_{T_0}) \neq \emptyset\}$, with $T_0 \ni x$. Now, the desired bound $I^{**} \lesssim 1$ is obtained similarly to (3.8). \square

4.3 Local projection stabilization

In this section, we shall discuss local projection stabilization methods (and, very briefly, somewhat related subgrid-scale schemes). We shall see that for such methods one can choose the Green's function interpolant $G_h \in S_h$ such that $S_{\mathcal{T}}(u_h, G_h) = 0$ in (4.2), which immediately yields the *a posteriori* error (3.4) and, hence, a more general (3.25).

We shall mainly focus on local projection stabilization methods of the form (4.1)—see, e.g. (Roos *et al.*, 2008, §III.3.3.1), (Tobiska & Verfürth, 2015, §2.2.2) for further details and the references therein—with the stabilizing term

$$S_{\mathcal{T}}(u_h, v_h) = \sum_{M \in \mathcal{M}} \delta_M \int_M \kappa_h(\bar{\mathbf{a}}_M \cdot \nabla u_h) \kappa_h(\bar{\mathbf{a}}_M \cdot \nabla v_h). \quad (4.9)$$

This stabilizing term uses fluctuations of the convective derivatives computed using the fluctuation operator $\kappa_h := I - \pi_h$, where π_h is a projection onto an appropriate discontinuous finite element space related to an auxiliary partition \mathcal{M} of Ω . The approximation $\bar{\mathbf{a}}_M$ of \mathbf{a} is assumed constant in each $M \in \mathcal{M}$. A user-chosen parameter δ_M in (4.9) typically satisfies

$$\delta_M \simeq h_M \|\mathbf{a}\|_{\infty;M}^{-1}.$$

Both $\mathcal{M} = \mathcal{T}$ (one-level approach) and \mathcal{T} generated by a single-level refinement of each element in \mathcal{M} (two-level approach) have been introduced in the literature (Becker & Braack, 2001, 2004; Matthies *et al.*, 2007). This can be implemented in various ways. To be more precise, it will be convenient to denote by S_h^r the set of functions that are continuous on Ω , equal to 0 on $\partial\Omega$, and polynomials of degree at most r on each $T \in \mathcal{T}$, where $r \geq 1$ is a fixed polynomial degree. An analogous set of functions

relative to the partition \mathcal{M} will be denoted S_H^r . With this notation, we assume that the finite element space S_h and the fluctuation operator κ_h satisfy

$$S_h^1 \subseteq S_h \subseteq S_h^r \quad \text{for some } r \geq 1, \quad \kappa_h(\nabla v_H) = 0 \quad \forall v_H \in S_H^1. \quad (4.10)$$

The above assumption on κ_h is satisfied if $\pi_h v$ in each $M \in \mathcal{M}$ is defined as the L_2 projection of v onto $\mathbb{P}^q(M)$ for some $q \geq 0$. In the two-level case, one may set $S_h := S_h^r$ and $q := r - 1$. In the one-level case $\mathcal{M} = \mathcal{T}$, for some $q \geq 0$, the space S_h^{q+1} can be enriched by appropriate bubble basis functions, so one gets (4.10) with $r := q + n + 1$. For example, if $q = 0$, one may employ $S_h = S_h^1 \oplus \text{span}\{b_T : T \in \mathcal{T}\}$, where $b_T = \prod_{i=1}^{n+1} \lambda_i \in \mathbb{P}^{n+1}(T)$ is the bubble function associated with T , so (4.10) is satisfied with $r = n + 1$.

For both one-level and two-level approaches under the general assumption (4.10), we get the following result.

LEMMA 6 Suppose that u_h satisfies (4.1), (4.9) under assumption (4.10). Then, under the conditions (2.5) on G , the error of the computed solution u_h satisfies (3.25), i.e., Theorem 1 remains valid for this method.

Proof. In view of (4.10), one has $u_h \in S_h^r$, so the efficiency results of §§3.3–3.4 apply immediately. Hence, to establish (3.25), it suffices to prove (3.4).

Next, construct the interpolant $G_h \in S_H^1 \subseteq S_h^1$ of G exactly as described in §2.3, only relative to the partition \mathcal{M} . Then the bounds (2.7) hold true, only with ω_T now denoting the patch of elements in \mathcal{M} touching $M \supseteq T$ (which also includes this M). With this tweak in the notation, the estimates in §3.2 remain valid, so Lemma 1 for this case again reads as (4.2). Finally, $G_h \in S_H^1$ combined with (4.10) yields $S_{\mathcal{T}}(u_h, G_h) = 0$, which, combined with (4.2), gives the desired (3.4). \square

Note that Lemma 6 also remains valid for a version of (4.9) with the fluctuations of the full gradient, i.e., with $\kappa_h(\bar{\mathbf{a}}_M \cdot \nabla \cdots)$ replaced by $\kappa_h(\nabla \cdots)$ (as in this case we again enjoy $S_{\mathcal{T}}(u_h, G_h) = 0$ for $G_h \in S_H^1$).

In addition, the above argument may be applied to subgrid-scale methods, in which gradients of fluctuations are used instead of fluctuations of gradients as in (4.9) for local projection methods (Guermond, 1999, 2001); see also, e.g. (Matthies *et al.*, 2007, §5), (Roos *et al.*, 2008, §IV.4.5), (Tobiska & Verfürth, 2015, §2.2.3). For example, one may replace the terms of type $\kappa_h(\bar{\mathbf{a}}_M \cdot \nabla \cdots)$ in (4.9) by $\mathbf{a}_M \cdot \nabla(\tilde{\kappa}_h \cdots)$ (or the full-gradient version $\nabla(\tilde{\kappa}_h \cdots)$). A typical fluctuation operator $\tilde{\kappa}_h$ satisfies $\ker \tilde{\kappa}_h \supseteq S_H^1$ (so $\tilde{\kappa}_h G_h = 0$), in which case we again get Lemma 6.

4.4 Concluding remarks on stabilized methods

Above, we have established that the introduction of a variety of stabilization techniques does not affect the ability to bound $\|u - u_h\|_{L_\infty(\Omega)}$ using our residual estimator, although the form of the proof depends on the particular stabilization technique. The rest of our arguments concerning the seminorm $|\mathbf{a} \cdot \nabla e|_*$ and efficiency of our estimators are generally not affected by the introduction of stabilization, since they do not use Galerkin orthogonality in their proof. The only exception comes in the choice of r used to define oscillation and $|\cdot|_*$, which, as noted in the preceding subsection, may require a little bit of care when employing projection methods with bubble functions in the definition of S_h . We summarize these findings below.

THEOREM 2 Consider streamline diffusion stabilization under the assumptions of Corollary 3, continuous interior penalty stabilization under the assumptions of Lemma 5, or local projection stabilization under the assumptions of Lemma 6. Under these conditions and the assumptions of Theorem 1, the conclusions of Theorem 1 remain valid. That is, the estimator $\max_{T \in \mathcal{T}} \eta_\infty(T)$ is reliable and efficient for the error notion $\|u - u_h\|_{\infty; \Omega} + |\mathbf{a} \cdot \nabla(u - u_h)|_* + \text{osc}(\alpha_T R_h, \Omega)$ up to factors of ℓ_h .

5. Numerical experiments

In this section, we present numerical experiments that illustrate the practical behavior of the error estimators and indicators defined above. All computations were carried out in MATLAB using suitably modified routines from the adaptive finite element library iFEM Chen (2009).

In all cases, we took Ω to be the unit square $(0, 1) \times (0, 1)$ and the coefficients $\mathbf{a} = [0, 1]$ and $b = 1$. We computed using affine Lagrange elements and either uniform or adaptive refinement. In the case of adaptive refinement, we used a modified maximum strategy in the standard `solve` → `estimate` → `mark` → `refine` loop. Let $\eta_\infty = \max_{T \in \mathcal{T}} \eta_\infty(T)$ be the overall error estimator, and fix a (small) positive integer K_{\max} . An element $T \in \mathcal{T}$ is bisected K_{\max} times if $\eta_\infty(T) \geq 0.5\eta_\infty$, $K_{\max} - 1$ times if $0.5\eta_\infty > \eta_\infty(T) \geq 0.25\eta_\infty$, etc. K_{\max} can be varied based on the degree of singular perturbation, with $K_{\max} = 4$ being used in the experiments below. This scheme helped to prevent too few elements being refined at each iteration of the adaptive procedure, and thus too many iterations from occurring. It also aided in more efficient resolution of boundary and interior layers since elements with large indicators are subdivided multiple times in each adaptive step.

Unknown constants appear in our error estimators and must be fixed. In our experiments, we chose the definition

$$\begin{aligned} \eta_\infty(T) = \min & \left[1, 0.0125\ell_h \frac{h_T^2}{\varepsilon} \right] \|R_T\|_{L_\infty(T)} \\ & + \min \left[\sqrt{\varepsilon}, 0.03\ell_h h_T \right] \|[\nabla u_h]\|_{L_\infty(\partial T)}. \end{aligned}$$

Experiments were conducted using either an unstabilized scheme or streamline diffusion stabilization with the parameter chosen as in (4.4).

5.1 Experiment 1: smooth solution

In this experiment, we consider a simple smooth solution

$$u_1(x, y) = \sin(\pi x) \sin(\pi y).$$

No stabilization was used. Our goal here is to illustrate and compare the convergence orders of the *a posteriori* error estimator $\max_{T \in \mathcal{T}} \eta_\infty(T)$, the target error norm $\|u - u_h\|_{L_\infty(\Omega)}$ and the convective error $|\mathbf{a} \cdot \nabla(u - u_h)|_*$. We thus take a uniform series of mesh refinements, and carry out convergence studies with $\varepsilon = 3 \times 10^{-3}$ and $\varepsilon = 10^{-5}$. Results are displayed in Fig. 2. In both cases, we observe that $\|u - u_h\|_{L_\infty(\Omega)}$ converges with order $DOF^{-1} = O(h^2)$, and with the same order of magnitude observed in each case. When $\varepsilon = 3 \times 10^{-3}$, we observe a preasymptotic regime in which the error estimator and convective error $|\mathbf{a} \cdot \nabla(u - u_h)|_*$ both converge with order $DOF^{-3/2} = h^3$. As h sufficiently resolves ε , the estimator instead tracks the error $\|u - u_h\|_{L_\infty(\Omega)}$ with order DOF^{-1} , while the convective error

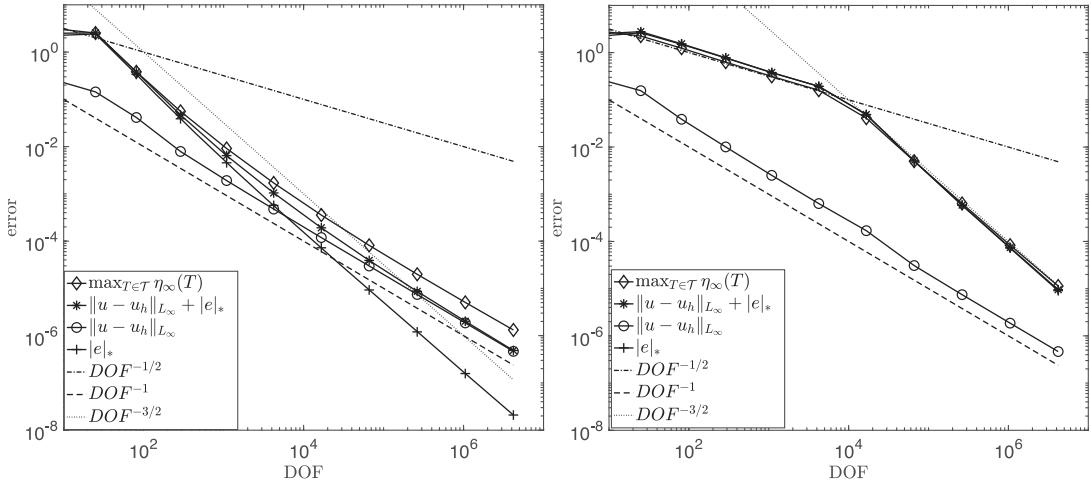


FIG. 2. Smooth solution u_1 and uniform refinement, standard Galerkin method without stabilization, $\varepsilon = 3 \times 10^{-3}$ (left) and $\varepsilon = 10^{-5}$ (right).

measured in the $*$ -seminorm continues to decrease with order $DOF^{-3/2}$. In all regimes, this illustrates that $\max_{T \in \mathcal{T}} \eta_\infty(T) \simeq \|u - u_h\|_{L_\infty(\Omega)} + |\mathbf{a} \cdot \nabla(u - u_h)|_*$ (up to data oscillation, which here is bounded by the latter two terms). In addition, it highlights two different dominant convergence rates for the estimator. A third initial regime of $O(DOF^{-1/2}) = O(h)$ convergence for the estimator and $|\mathbf{a} \cdot \nabla(u - u_h)|_*$ is illustrated in the right plot in Fig. 2, where $\varepsilon = 10^{-5}$.

Understanding these convergence regimes is easiest when considering the $**$ -seminorm $|\mathbf{a} \cdot \nabla(u - u_h)|_{**} = \sup_{T \in \mathcal{T}} \min\left(1, 0.0125\ell_h \frac{h_T^2}{\varepsilon}\right) \|\mathbf{a} \cdot \nabla e\|_{\infty;T}$, which closely tracks the $*$ seminorm in this case. We may expect $\|\mathbf{a} \cdot \nabla(u - u_h)\|_{L_\infty(T)} = O(h) = O(DOF^{-1/2})$. Initially, $\min(1, \ell_h \frac{h_T^2}{\varepsilon}) = 1$, so $|\mathbf{a} \cdot \nabla(u - u_h)|_{**} = O(h)$ also. When $h_T^2 \ell_h < \sqrt{\varepsilon}$, then we have $|\mathbf{a} \cdot \nabla(u - u_h)|_{**} \lesssim \frac{h^3 \ell_h}{\varepsilon}$, leading to the increased rate of convergence observed in Fig. 2.

REMARK 5 In Fig. 2, we observe that $\|u - u_h\|_{L_\infty(\Omega)}$ converges with optimal rate $O(DOF^{-1})$ from essentially the first mesh refinement. This implies that oscillations often associated with unstabilized solution of singularly perturbed problems are not present here. This was confirmed by viewing plots of the discrete solution. Lack of instability in the discrete solution may be due to symmetries in the test solution, which in 1D examples has been observed to lead to similar unexpectedly good results for unstabilized methods (Kopteva, 1993, Chap. 4). Although numerical stability is uncharacteristically good for this example, it is nonetheless useful as it allows for clear exposition of the properties of our estimator relative to the target error notion $\|u - u_h\|_{\infty; \Omega}$ and seminorm $|\mathbf{a} \cdot \nabla(u - u_h)|_*$.

5.2 Experiment 2: outflow boundary layer

In this subsection, we consider the outflow boundary layer solution

$$u_2(x, y) = x(1 - x) \left[y - \frac{e^{-(1-y)/\varepsilon} - e^{-1/\varepsilon}}{1 - e^{-1/\varepsilon}} \right].$$

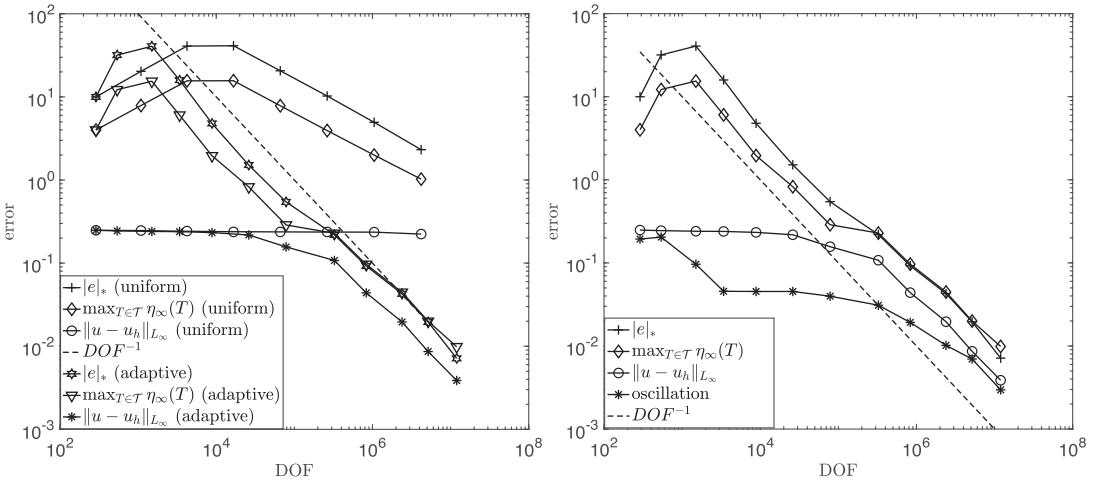


FIG. 3. Outflow boundary layer solution u_2 with $\varepsilon = 10^{-5}$ and streamline diffusion stabilization. Uniform and adaptive refinement (left), and with oscillation displayed (right).

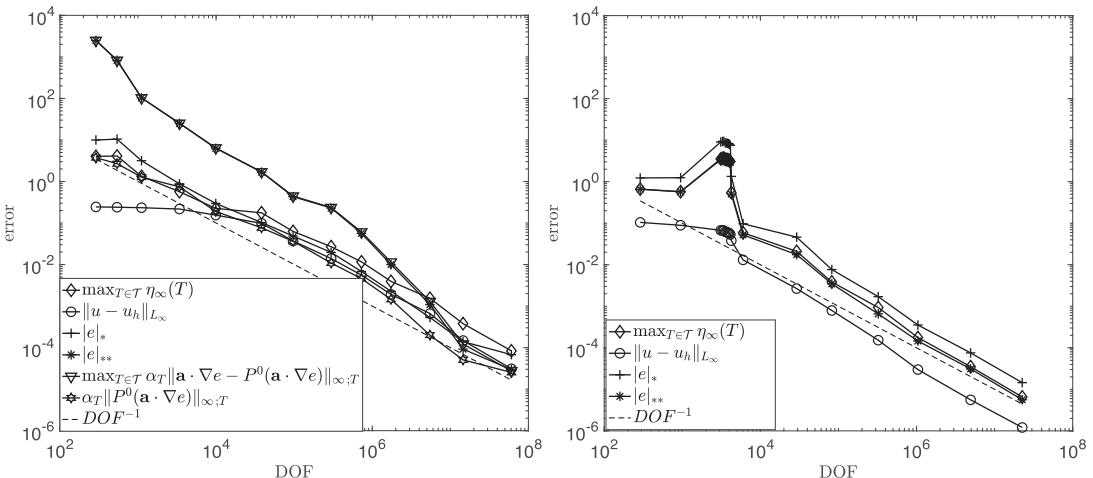


FIG. 4. Outflow boundary layer solution u_2 with $\varepsilon = 10^{-4}$ (left); interior layer solution u_3 with $\varepsilon = 10^{-5}$ (right), streamline diffusion stabilization with adaptive refinement.

This solution exhibits a strong layer at $y = 1$ of width $O(\varepsilon)$ and corresponding maximum solution gradient size $O(\varepsilon^{-1})$. In this experiment and in all of the displayed adaptive experiments involving layers, we employed streamline diffusion stabilization.

In the left plot in Fig. 3, we illustrate the advantage of adaptive versus uniform refinement with $\varepsilon = 10^{-5}$. Uniform refinement leads to essentially no decrease in the target norm $\|u - u_h\|_{L_\infty(\Omega)}$, but some decrease in the $*$ -seminorm and estimator. Adaptive refinement yields little initial decrease in the maximum error, but optimal $O(DOF^{-1})$ decrease begins with a little over 10^5 degrees of freedom. The $*$ -seminorm and estimator decrease with order DOF^{-1} for most of the convergence history.

In the right plot of Fig. 3, we include data oscillation in the plot in order to illustrate that the $|\cdot|_*$ -seminorm is an essential part of the error notion measured by $\max_{T \in \mathcal{T}} \eta_{\infty}(T)$. In particular, we see that $\max_{T \in \mathcal{T}} \eta_{\infty}(T) \simeq |\mathbf{a} \cdot \nabla e|_* \gg \|u - u_h\|_{\infty, \Omega} + \max_{T \in \mathcal{T}_h} \alpha_T \|R_h - R_{h,T}\|_{\infty, T}$ for much of the convergence history, confirming that the estimator is reliable, but not efficient when measuring only the sum of the maximum error and data oscillation.

5.3 Experiment 3: interior layer, $|\cdot|_*$ versus $|\cdot|_{**}$ error measure

Here, in addition to the outflow boundary layer problem considered above, we also consider the simple interior layer solution (Ern & Stephansen, 2008)

$$u_3(x, y) = 2x(1-x)y(1-y) \left(1 - \tanh[(.5 - x)/\sqrt{\varepsilon}]\right).$$

Note that the interior layer solution possesses milder layer behavior than does the outflow layer.

We shall now use these two tests—see Fig. 4—not only to illustrate the performance of our estimator, but to also follow up on the theoretical comparison of $|\cdot|_*$ versus $|\cdot|_{**}$ in Section 3.5. Recall the definitions (3.24) and (3.26) of the $*$ and $**$ -(semi)norms, and that the $*$ -seminorm accurately reflects estimator behavior, but may not be intuitive while the $**$ norm has a more concrete form, but may overestimate the $*$ -seminorm. In some cases of interest, the $*$ and $**$ (semi) norms of the convective error are nonetheless very close in size. To illustrate this fact and the inequality (3.27), we compare behavior of the outflow boundary layer problem and the simple interior layer solution. In the left plot in Fig. 4, we take $\varepsilon = 10^{-4}$ in the outflow boundary layer problem. This yields $|\mathbf{a} \cdot \nabla(u - u_h)|_{**} \sim \varepsilon^{-1}$, while $|\mathbf{a} \cdot \nabla(u - u_h)|_*$ is much smaller. We see that the quantity $\max_{T \in \mathcal{T}} \alpha_T \|P^0(\mathbf{a} \cdot \nabla)\|_{\infty, T}$ closely tracks $|\mathbf{a} \cdot \nabla e|_*$, while $\max_{T \in \mathcal{T}} \alpha_T \|\mathbf{a} \cdot \nabla e - P^0(\mathbf{a} \cdot \nabla e)\|_{\infty, T}$ closely tracks $|\mathbf{a} \cdot \nabla e|_{**}$; cf. (3.27). Also, heavy refinement is required before oscillation in $\nabla(u - u_h)$ is resolved, and the $*$ and $**$ quantities become more or less equivalent. In the right plot, we consider the interior layer problem with $\varepsilon = 10^{-5}$. Here, the $*$ and $**$ (semi)norms of the convective error are essentially equivalent throughout the convergence history. These experiments confirm that, while the $**$ seminorm lends some intuition to the error behavior, the $*$ norm most accurately captures the error dynamics.

We also emphasize that in all of our experiments, the estimator $\max_{T \in \mathcal{T}} \eta_{\infty}(T)$ in fact closely tracked the total error $\|u - u_h\|_{L_{\infty}(\Omega)} + |\mathbf{a} \cdot \nabla(u - u_h)|_*$ as predicted. Elementwise, oscillation of the residual R_h appeared not to dominate the other terms in the total error in all cases.

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Data availability

The data underlying this article will be shared on reasonable request to the corresponding author.

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