

ON THE CONVERGENCE OF ADAPTIVE NON-CONFORMING FINITE ELEMENT METHODS

C. ORTNER & D. PRAETORIUS

ABSTRACT. We formulate and analyze an adaptive non-conforming finite-element method for the solution of convex variational problems. The class of minimization problems we admit, includes highly singular problems for which no Euler–Lagrange equation (or inequality) is available. As a consequence, our arguments only use the structure of the energy functional. We are nevertheless able to prove convergence of an adaptive algorithm, using even refinement indicators that are not reliable error indicators.

1. INTRODUCTION

We present a new convergence proof for adaptive non-conforming finite element methods which is applicable to a wide class of convex variational problems.

For fixed $n \geq 2$ and $m \geq 1$, let Ω be a bounded polygonal domain in \mathbb{R}^n and let $W : \mathbb{R}^{m \times n} \rightarrow [0, +\infty]$ be a convex stored energy density which satisfies a p -growth condition from below, i.e.,

$$W(\xi) \geq C(|\xi|^p - 1) \quad \text{for all } \xi \in \mathbb{R}^{m \times n} \text{ and some } p > 1. \quad (1)$$

For a fixed dead load $f \in L^{p'}(\Omega)$, where $1/p' + 1/p = 1$, we define the energy functional $\mathcal{J} : W^{1,p}(\Omega)^m \rightarrow \mathbb{R} \cup \{+\infty\}$ by

$$\mathcal{J}(v) = \int_{\Omega} (W(\nabla v) - f \cdot v) \, dx. \quad (2)$$

Given some $g \in W^{1,p}(\Omega)^m$ with $\mathcal{J}(g) < +\infty$ and sets $\Gamma^{(i)} \subset \partial\Omega$, $i = 1, \dots, m$, with positive surface measure $|\Gamma^{(i)}| > 0$, we define the *admissible set*

$$\mathcal{A} = \{v \in W^{1,p}(\Omega)^m : v^{(i)} = g^{(i)} \text{ on } \Gamma^{(i)}, i = 1, \dots, m\} \neq \emptyset; \quad (3)$$

here and throughout we use superscripts to denote components of a vector-valued function. We note that admissible functions satisfy the Poincaré-type inequality

$$\|v - g\|_{L^p(\Omega)} \leq C_p \|\nabla v - \nabla g\|_{L^p(\Omega)} \quad \text{for all } v \in \mathcal{A}, \quad (4)$$

where C_p is some fixed constant.

Date: April 29, 2009.

2000 Mathematics Subject Classification. 49J45, 65N30, 49M25.

Key words and phrases. Calculus of Variations, Crouzeix–Raviart FEM, Convergence of AFEM, Lavrentiev Phenomenon.

The authors acknowledge the kind support and generous hospitality of the Hausdorff Centre for Mathematics (University of Bonn) where this work was initiated.

In this paper, we analyze the numerical solution of the minimization problem

$$u \in \operatorname{argmin} \mathcal{J}(\mathcal{A}). \quad (5)$$

by means of an adaptive non-conforming finite element method.

Under the conditions we imposed, the functional \mathcal{J} may be non-differentiable at a solution, even if W itself is smooth [3], and hence the associated Euler–Lagrange equation is unavailable to us. In this situation, the method of choice for the analysis of (5) (and its discretization) is the *direct method of the calculus of variations* [15], whose application immediately gives the following basic existence result. Since it is helpful to understand our subsequent analysis, we give a brief outline of the proof. We refer to Dacorogna’s monograph [15] (in particular Theorem 4.1) for further details.

Proposition 1. *There exists at least one solution of (5).*

Proof. For an admissible function $v \in \mathcal{A}$, the Poincaré inequality (4) implies

$$\|v\|_{L^p(\Omega)} \lesssim \|\nabla(v - g)\|_{L^p(\Omega)} + \|g\|_{L^p(\Omega)} \lesssim \|\nabla v\|_{L^p(\Omega)} + \|g\|_{W^{1,p}(\Omega)},$$

where $a \lesssim b$ abbreviates $a \leq Cb$ with a generic constant $C > 0$. From the p -growth condition (1) and Young’s inequality, we infer

$$\mathcal{J}(v) \gtrsim \|\nabla v\|_{L^p(\Omega)}^p - \|f\|_{L^{p'}(\Omega)} \|v\|_{L^p(\Omega)} - |\Omega| \gtrsim \|\nabla v\|_{L^p(\Omega)}^p - C(f, g, \Omega)$$

with an additive constant $C(f, g, \Omega) \geq 0$. Applying the Poincaré-type inequality (4) again, we find that \mathcal{J} is coercive, i.e.,

$$\|v\|_{W^{1,p}(\Omega)}^p \lesssim \mathcal{J}(v) + 1 \quad \text{for all } v \in \mathcal{A}. \quad (6)$$

Suppose now that $(u_\ell)_{\ell \in \mathbb{N}} \subset \mathcal{A}$ is a minimizing sequence for \mathcal{J} , i.e., $\mathcal{J}(u_\ell) \rightarrow \inf \mathcal{J}(\mathcal{A})$ as $\ell \rightarrow \infty$. By (6) and reflexivity of $W^{1,p}(\Omega)^m$, we may assume that $(u_\ell)_{\ell \in \mathbb{N}}$ converges weakly to some limit u in $W^{1,p}(\Omega)^m$. Since \mathcal{A} is convex and closed, it is weakly closed, and hence it follows that $u \in \mathcal{A}$. Finally, convexity and non-negativity of W imply the weak lower semicontinuity of \mathcal{J} , i.e.,

$$\mathcal{J}(u) \leq \liminf_{\ell \rightarrow \infty} \mathcal{J}(u_\ell),$$

cf. [15, Theorem 3.4]. Thus, we conclude that $\mathcal{J}(u) = \inf \mathcal{J}(\mathcal{A})$, i.e., that $u \in \operatorname{argmin} \mathcal{J}(\mathcal{A})$. \square

The class of problems which is included in our analysis is surprisingly general. It includes not only standard variational model problems such as the Dirichlet problem [10, 13, 18, 21], or the p -Laplacian [17, 24], or convex problems with control of stresses [4, 9] (all of these works require uniform p -growth from below *and* above), but is in fact suited for any convex functional of the type (2). Even our p -growth condition from below can be relaxed to some extent [23]. For simplicity we have restricted our presentation to Dirichlet constraints, but this constraint is easily lifted as well (cf. Section 5).

One of the most important features of our analysis is that we do not require \mathcal{J} to be continuous in the strong topology of $W^{1,p}(\Omega)^m$ and hence, using ideas from [23], we are even able to overcome the *Lavrentiev gap phenomenon*. If $g \in W^{1,\infty}(\Omega)^m$, then

$\mathcal{A}_\infty := \mathcal{A} \cap W^{1,\infty}(\Omega)^m$ is non-empty, and we say that (5) exhibits the Lavrentiev gap phenomenon if

$$\inf \mathcal{J}(\mathcal{A}_\infty) > \min \mathcal{J}(\mathcal{A}). \quad (7)$$

Foss, Hrusa and Mizel [19] have shown that this effect can indeed occur under the conditions we have posed. Note that, since typical conforming finite element functions are Lipschitz continuous, non-conformity of the numerical method is essential for treating problems in that class. Furthermore, since the Lavrentiev effect is closely linked to singularities in the solution of nonlinear variational problems, adaptive solution techniques are particularly important. We refer to [3, 12, 23] for overviews of this fascinating subdiscipline of the calculus of variations.

To the best of our knowledge, three classes of convergence proofs for adaptive finite element methods for linear problems exist to date. The first idea [18, 21] used the so-called inner node property from which a lower bound on the discrete error in terms of the estimator can be derived. This implies, up to oscillation terms, the reduction of the error at each step of the adaptive algorithm. A further step was taken in [22] by circumventing the inner node property at a given step but still requiring that it should be obtained after a fixed number of refinements. The first proof which completely circumvented the use of lower bounds but only requires reliability of the error estimator, is given in [13]. Extensions of this convergence analysis to the non-linear Laplacian, or more general convex variational problems can be found in [4, 9, 17, 24]. These works require the use of some additional structure of the problems but are still heavily based on the analysis for the Dirichlet problem.

Proofs of the convergence of adaptive non-conforming and mixed finite element methods can be found, for example, in [10, 11]. The analysis contained therein is largely adapted from the conforming case, the crucial modification being a control of non-conformity which leads to so-called quasi-Galerkin orthogonality relations.

In particular, all previous proofs require the use of Euler–Lagrange equations, and of reliable *a posteriori* error estimates, both of which are not available for our problem class. We therefore have to use a completely different approach. Our convergence argument is based on the direct method of the calculus of variations, and is strictly tailored to non-conforming finite element methods. It does not apply in an obvious way to the conforming case. In addition to the lack of Euler–Lagrange equations in our problem, it is also interesting to note that we obtain convergence of our adaptive algorithm even though the refinement indicators we use do not provide reliable error bounds.

The first part of our convergence proof (Section 3.1) is motivated by [23] and, to the best of our knowledge, uses techniques which have not previously been employed in the context of adaptive finite element methods. The second part (Theorem 8) is closely related to, and in fact inspired by, the proof given in [22].

The paper is organized as follows. In Section 2, we fix the notation and state some auxiliary results. In Section 3.1, we provide sufficient conditions for the convergence of the Crouzeix–Raviart finite element method. This analysis motivates the definition of convergence indicators which we discuss in some detail in Section 3.2. Finally,

in Section 3.3 we formulate an adaptive algorithm and prove its convergence. To conclude, we present several numerical experiments in Section 4.

2. PRELIMINARIES

2.1. Function Spaces. Let A be an open subset of \mathbb{R}^n . We use $L^p(A)$ and $W^{1,p}(A)$ to denote the standard Lebesgue and Sobolev spaces and equip them with their usual norms. The space of distributions is denoted by $\mathcal{D}'(A)$ [2]. The distributional gradient operator is denoted D , while the weak gradient operator is denoted ∇ . The space of k -times continuously differentiable functions with compact support in A is denoted $C_0^k(A)$.

2.2. Triangulation of Ω . For every $\ell \in \mathbb{N}$, we assume that \mathcal{T}_ℓ is a regular triangulation of $\bar{\Omega}$ into closed simplices $T \in \mathcal{T}_\ell$. In particular, \mathcal{T}_ℓ has no hanging nodes in 2D, no hanging nodes or edges in 3D, and so forth. Let \mathcal{E}_ℓ denote the collection of $n - 1$ dimensional faces of elements and $\mathcal{E}_\ell^{\text{int}}$ the collection of interior faces. Let \mathcal{N}_ℓ denote the set of vertices and $\mathcal{N}_\ell^{\text{nc}}$ denote the set of barycenters of the faces. We assume throughout that, up to surface measure zero, the sets $\Gamma^{(i)}$ are the union of faces $\mathcal{E}_\ell^{d,i} \subset \mathcal{E}_\ell$, $i = 1, \dots, m$, and we set $\mathcal{E}_\ell^{\text{int},i} = \mathcal{E}_\ell \setminus \mathcal{E}_\ell^{d,i}$.

For each element $T \in \mathcal{T}_\ell$, we set $h_T = \text{diam}(T)$. For each face $E \in \mathcal{E}$, we set $h_E = \text{diam}(E)$. The mesh-size function $h_\ell : \Omega \rightarrow \mathbb{R}_{>0}$ is then almost everywhere defined by $h_\ell(x) = h_T$ for x in the interior of an element $T \in \mathcal{T}_\ell$ and $h_\ell(x) = h_E$ for x in the relative interior of a face $E \in \mathcal{E}_\ell$.

The shape regularity constant $\sigma(\mathcal{T}_\ell)$ is the smallest number $C > 0$ such that

$$C^{-1} h_T^n \leq |T| \leq h_T^n, \quad C^{-1} h_E^{n-1} \leq |E| \leq h_E^{n-1} \quad \text{and} \quad h_E \leq h_T \leq C h_E$$

for all elements $T \in \mathcal{T}_\ell$ and faces $E \in \mathcal{E}_\ell$ with $E \subset \partial T$. A family $(\mathcal{T}_\ell)_{\ell \in \mathbb{N}}$ of regular triangulations is uniformly shape-regular, if $\sup_{\ell \in \mathbb{N}} \sigma(\mathcal{T}_\ell) < \infty$.

2.3. Finite Element Spaces. We briefly describe the Crouzeix–Raviart finite element space used to discretize (5); see [5, 7, 14] for further details.

The space of all \mathcal{T}_ℓ -piecewise affine functions is denoted by

$$P_1(\mathcal{T}_\ell) = \{v \in L^1(\Omega) : v|_T \text{ is affine for all } T \in \mathcal{T}_\ell\}.$$

$\text{CR}(\mathcal{T}_\ell)$ denotes the Crouzeix–Raviart finite element space,

$$\text{CR}(\mathcal{T}_\ell) = \{v_\ell \in P_1(\mathcal{T}) : v_\ell \text{ is continuous at all face barycentres } z \in \mathcal{N}_\ell^{\text{nc}}\}.$$

For each barycenter $z \in \mathcal{N}_\ell^{\text{nc}}$, let $E_z \in \mathcal{E}_\ell$ be the unique face which contains z . The interpolation operator $\Pi_\ell : W^{1,1}(\Omega)^m \rightarrow \text{CR}(\mathcal{T}_\ell)^m$ is defined via [14]

$$\Pi_\ell v(z) = |E_z|^{-1} \int_{E_z} v \, ds \quad \text{for all } z \in \mathcal{N}_\ell^{\text{nc}}$$

and thus satisfies $\int_E \Pi_\ell v \, ds = \int_E v \, ds$ for all faces $E \in \mathcal{E}_\ell$. We summarize its most important properties for our purpose in the following lemma. It is worth noting that

neither the stability of Π_ℓ , nor the interpolation error estimate depend on the shape regularity $\sigma(\mathcal{T}_\ell)$.

Lemma 2 (Properties of Crouzeix–Raviart Interpolant). *Let $v \in W^{1,p}(\Omega)^m$ and $T \in \mathcal{T}_\ell$, then Π_ℓ has the first-order approximation property*

$$\|v - \Pi_\ell v\|_{L^p(T)} \leq C_{\text{apx}} h_T \|\nabla v\|_{L^p(T)} \quad (8)$$

with $C_{\text{apx}} = 1 + 2/n \leq 2$. Furthermore, it satisfies the mean value property

$$|T|^{-1} \int_T \nabla v \, dx = |T|^{-1} \int_T \nabla(\Pi_\ell v) \, dx = \nabla(\Pi_\ell v)|_T, \quad (9)$$

which implies the stability estimate

$$\|\nabla(\Pi_\ell v)\|_{L^p(T)} \leq \|\nabla v\|_{L^p(T)}. \quad (10)$$

Proof. Since the outer unit normal ν to T is constant on each of its faces, integration by parts in T yields

$$\int_T \frac{\partial}{\partial x_j} v \, dx = \int_{\partial T} v \nu_j \, ds = \int_{\partial T} (\Pi_\ell v) \nu_j \, ds = \int_T \frac{\partial}{\partial x_j} (\Pi_\ell v) \, dx,$$

which proves (9). In particular, we have

$$\|\nabla(\Pi_\ell v)\|_{L^p(T)} = |T|^{1/p-1} \left| \int_T \nabla v \, dx \right| \leq \|\nabla v\|_{L^p(T)},$$

which establishes (10).

Next, we recall the well-known trace identity

$$\frac{1}{|E|} \int_E w \, ds = \frac{1}{|T|} \int_T w \, dx + \frac{1}{n|T|} \int_T (x - z) \cdot \nabla w \, dx \quad \text{for all } w \in W^{1,1}(T),$$

where $z \in T \cap \mathcal{N}$ is the vertex opposite to E , i.e. $T = \text{conv}(E \cup \{z\})$. By definition, $w := v - \Pi_\ell v$ satisfies $\int_E w \, ds = 0$. Therefore, the integral mean $w_T := |T|^{-1} \int_T w \, dx$ can be estimated by

$$|w_T| = \frac{1}{|T|} \left| \int_T w \, dx \right| \leq \frac{h_T}{n|T|} \|\nabla w\|_{L^1(T)} \leq \frac{h_T |T|^{1/p'-1}}{n} \|\nabla w\|_{L^p(T)},$$

whence $\|w_T\|_{L^p(T)} \leq (h_T/n) \|\nabla w\|_{L^p(T)}$. Next, we use the Poincaré inequality $\|w - w_T\|_{L^p(T)} \leq (h_T/2) \|\nabla w\|_{L^p(T)}$ on the convex domain T [1]. This gives

$$\|w\|_{L^p(T)} \leq \|w - w_T\|_{L^p(T)} + \|w_T\|_{L^p(T)} \leq \left(\frac{1}{2} + \frac{1}{n}\right) h_T \|\nabla w\|_{L^p(T)}.$$

Hence, we can deduce (8) from (10) and a triangle inequality. \square

Since $v_\ell \in \text{CR}(\mathcal{T}_\ell)^m$ may be discontinuous across faces $E \in \mathcal{E}_\ell$, v_ℓ is not weakly differentiable; nevertheless we use ∇v_ℓ to denote its \mathcal{T}_ℓ -elementwise gradient. We also require a notation for the jumps across interior faces. For $E = T^+ \cap T^- \in \mathcal{E}_\ell^{\text{int}}$, we

fix the labelling of the elements T^\pm and we let v_ℓ^\pm denote the traces from T^\pm , and $\nu = \nu_E$ the outer unit normals to T^+ . We define the jump across E by

$$[v_\ell]^{(i)} = v_\ell^+ - v_\ell^-.$$

For a boundary face $E \subset \partial\Omega$, $E = \partial\Omega \cap T$, we let $\nu = \nu_E$ be the outer unit normal to Ω , and we define

$$[v_\ell]_i = \begin{cases} v_\ell^{(i)} - g^{(i)}, & \text{if } E \subset \Gamma^{(i)} \\ 0, & \text{otherwise.} \end{cases}$$

With this notation, the distributional gradient reads

$$\langle Dv_\ell, \varphi \rangle = - \int_\Omega v_\ell \cdot \operatorname{div} \varphi \, dx = \int_\Omega \nabla v_\ell : \varphi \, dx - \int_{\cup \mathcal{E}_\ell^{\text{int}}} ([v_\ell] \otimes \nu) : \varphi \, ds, \quad (11)$$

for $v_\ell \in P_1(\mathcal{T}_\ell)^m$ and $\varphi \in C_0^1(\Omega)^{m \times n}$. The representation formula (11) can be verified by integration by parts on each element. The symbol \otimes denotes the tensor product $a \otimes b \in \mathbb{R}^{m \times n}$, i.e., $(a \otimes b)_{ij} = a_i b_j$.

3. ADAPTIVE SOLUTION

Let $(\mathcal{T}_\ell)_{\ell \in \mathbb{N}}$ be a uniformly shape-regular family of triangulations of Ω , which will subsequently be generated by an adaptive algorithm. We extend the definition of the energy functional \mathcal{J} to the Crouzeix–Raviart finite element space $\text{CR}(\mathcal{T}_\ell)^m$ by setting

$$\mathcal{J}(v_\ell) = \int_\Omega [W(\nabla v_\ell) - f \cdot v_\ell] \, dx.$$

We stress, however, that ∇v_ℓ now denotes the \mathcal{T}_ℓ -piecewise gradient.

Since $\text{CR}(\mathcal{T}_\ell)$ is not a subspace of $W^{1,p}(\Omega)$ we need to take care in defining the set of discrete admissible functions. A natural definition is to impose the Dirichlet condition on the face barycenters,

$$\mathcal{A}_\ell = \{v_\ell \in \text{CR}(\mathcal{T}_\ell)^m : v_\ell^{(i)}(z) = \Pi_\ell g^{(i)}(z) \text{ for } z \in \Gamma^{(i)} \cap \mathcal{N}_\ell^{\text{nc}}, i = 1, \dots, m\}. \quad (12)$$

We note that $\Pi_\ell v \in \mathcal{A}_\ell$, for each $v \in \mathcal{A}$, and hence, \mathcal{A}_ℓ is sufficiently rich to be a ‘good’ approximation of \mathcal{A} .

The Crouzeix–Raviart finite element discretization of (5) is to find a minimizer

$$u_\ell \in \operatorname{argmin} \mathcal{J}(\mathcal{A}_\ell). \quad (13)$$

As in the continuous case, we have the following result.

Proposition 3. *There exists at least one solution to (13).*

Proof. We simply adapt the proof of Proposition 1. As before, \mathcal{A}_ℓ is convex and closed, and \mathcal{J} is (weakly) lower semicontinuous on the finite dimensional space $\text{CR}(\mathcal{T}_\ell)^m$. It only remains to prove the coercivity of \mathcal{J} in \mathcal{A} : a broken Poincaré inequality [8, Corollary 4.3] (for the case $p = 2$ see also [6]) provides

$$\|v_\ell\|_{L^p(\Omega)} \leq \|v_\ell - \Pi_\ell g\|_{L^p(\Omega)} + \|\Pi_\ell g\|_{L^p(\Omega)} \lesssim \|\nabla(v_\ell - \Pi_\ell g)\|_{L^p(\Omega)} + \|\Pi_\ell g\|_{L^p(\Omega)},$$

so that stability of Π_ℓ yields

$$\|v_\ell\|_{L^p(\Omega)} \lesssim \|\nabla v_\ell\|_{L^p(\Omega)} + \|g\|_{W^{1,p}(\Omega)}.$$

With the same arguments as in the proof of Proposition 1, we obtain the discrete analogue of (6),

$$\|v_\ell\|_{L^p(\Omega)}^p + \|\nabla v_\ell\|_{L^p(\Omega)}^p \lesssim \mathcal{J}(v_\ell) + 1 \quad \text{for all } v_\ell \in \mathcal{A}_\ell. \quad (14)$$

Arguing as above, we conclude the proof. \square

3.1. Sufficient Conditions for Convergence. In this section, we derive conditions under which a sequence $(u_\ell)_{\ell \in \mathbb{N}}$ of discrete solutions converges to a solution of (5). Lemma 4 is the main observation which led to the convergence theorem for uniformly refined meshes [23, Equation (31)] and will again play a prominent role here. Lemma 5 is a refinement of [23, Lemma 8] which allows us to adapt the convergence argument to adaptively refined meshes.

Lemma 4 (Upper Bound). *For every $v \in W^{1,p}(\Omega)^m$, it holds that*

$$\mathcal{J}(\Pi_\ell v) \leq \mathcal{J}(v) + C_{\text{apx}} \|h_\ell f\|_{L^{p'}(\Omega)} \|\nabla v\|_{L^p(\Omega)}. \quad (15)$$

Proof. Jensen's inequality yields $W(|T|^{-1} \int_T \nabla v \, dx) \leq |T|^{-1} \int_T W(\nabla v) \, dx$. From the mean value property (9), we infer

$$\int_T W(\nabla \Pi_\ell v) \, dx \leq \int_T W(\nabla v) \, dx,$$

and hence,

$$\mathcal{J}(\Pi_\ell v) \leq \mathcal{J}(v) + \int_\Omega f \cdot (v - \Pi_\ell v) \, dx.$$

Elementwise application of the approximation error estimate (8) results in (15). \square

Remark 1. The upper bound (15) should be expected to be suboptimal. For example, for the standard Dirichlet problem (where $W(F) = |F|^2$), the energy error is formally of order $O(h^2)$ while the estimate in (15) is only of order $O(h)$. One reason for this is that (15) provides an upper bound for *any* v and not only for the energy minimum. However, since we cannot assume any differentiability properties on \mathcal{J} (this is due to the lack of a growth condition on W from above), there is little hope to recover an $O(h^2)$ estimate. \square

Lemma 5 (Compactness of Sublevel Sets). *Let $v_\ell \in \mathcal{A}_\ell$, $\ell \in \mathbb{N}$, be a sequence satisfying $\sup_{\ell \in \mathbb{N}} \mathcal{J}(v_\ell) < \infty$, and assume that*

$$\|h_\ell[v_\ell]\|_{L^1(\cup \mathcal{E}_\ell)} \xrightarrow{\ell \rightarrow \infty} 0. \quad (16)$$

(We stress that the skeleton $\bigcup \mathcal{E}_\ell$ includes $\partial\Omega$.) Then, there exists a subsequence $(v_{\ell_k})_{k \in \mathbb{N}}$ and a limit $v \in \mathcal{A}$ such that

$$\begin{aligned} v_{\ell_k} &\rightharpoonup v && \text{weakly in } L^p(\Omega)^m, \\ \nabla v_{\ell_k} &\rightharpoonup \nabla v && \text{weakly in } L^p(\Omega)^{m \times n}. \end{aligned} \tag{17}$$

Proof. As in the proof of Proposition 3 (cf. (14)) boundedness of the energy gives

$$\sup_{\ell \in \mathbb{N}} (\|v_\ell\|_{L^p(\Omega)} + \|\nabla v_\ell\|_{L^p(\Omega)}) < \infty.$$

Since $L^p(\Omega)$ is reflexive, we may assume without loss of generality that v_ℓ as well as ∇v_ℓ are weakly convergent with limits $v \in L^p(\Omega)^m$ and $F \in L^p(\Omega)^{m \times n}$. We now aim to show that v is weakly differentiable with $\nabla v = F$. To this end, we fix $\varphi \in C_0^\infty(\Omega)^{m \times n}$, and recall the representation formula (11) to obtain

$$\langle Dv_\ell, \varphi \rangle = \int_\Omega \nabla v_\ell : \varphi \, dx - \int_{\bigcup \mathcal{E}_\ell} ([v_\ell] \otimes \nu) : \varphi \, ds.$$

For the first term, weak convergence of ∇v_ℓ to F implies

$$\int_\Omega \nabla v_\ell : \varphi \, dx \xrightarrow{\ell \rightarrow \infty} \int_\Omega F : \varphi \, dx.$$

For each face $E \in \mathcal{E}_\ell$, let $\varphi_E := |E|^{-1} \int_E \varphi \, ds$ denote the integral mean of φ over E . Using the fact that $\int_E [v_\ell] \, ds = 0$ for all interior faces $E \in \mathcal{E}_\ell^{\text{int}}$, we estimate the second term by

$$\begin{aligned} \left| \int_{\bigcup \mathcal{E}_\ell} ([v_\ell] \otimes \nu) : \varphi \, ds \right| &= \left| \sum_{E \in \mathcal{E}_\ell} \int_E ([v_\ell] \otimes \nu) : (\varphi - \varphi_E) \, ds \right| \\ &\leq \sum_{E \in \mathcal{E}_\ell} h_E \int_E |[v_\ell]| \, ds \|\nabla \varphi\|_{L^\infty(\Omega)} \\ &= \int_{\bigcup \mathcal{E}_\ell} h_\ell |[v_\ell]| \, ds \|\nabla \varphi\|_{L^\infty(\Omega)}. \end{aligned}$$

By hypothesis (16) and the definition of the distributional gradient, we obtain

$$\int_\Omega F : \varphi \, dx = \lim_{\ell \rightarrow \infty} \langle Dv_\ell, \varphi \rangle = - \lim_{\ell \rightarrow \infty} \int_\Omega v_\ell \cdot \operatorname{div} \varphi \, dx = - \int_\Omega v \cdot \operatorname{div} \varphi \, dx = \langle Dv, \varphi \rangle,$$

which proves $v \in W^{1,p}(\Omega)^m$ with $\nabla v = Dv = F$.

It remains to show that $v|_{\Gamma^{(i)}} = g|_{\Gamma^{(i)}}$. Here it is crucial that (16) includes the condition that $\|h_\ell(v_\ell - g)^{(i)}\|_{L^1(\Gamma^{(i)})} \rightarrow 0$. The result then follows upon combining the arguments from [23, Lemma 8] with the generalization presented above. \square

Theorem 6 (Convergence of Discrete Minimizers). *Suppose that a sequence $u_\ell \in \operatorname{argmin} \mathcal{J}(\mathcal{A}_\ell)$ of discrete minimizers satisfies*

$$\|h_\ell f\|_{L^{p'}(\Omega)} + \|h_\ell[u_\ell]\|_{L^1(\bigcup \mathcal{E}_\ell)} \xrightarrow{\ell \rightarrow \infty} 0. \tag{18}$$

Then there exists a subsequence $(u_{\ell_k})_{k \in \mathbb{N}}$, and $u \in \operatorname{argmin} \mathcal{J}(\mathcal{A})$ such that

$$\begin{aligned} u_{\ell_k} &\rightharpoonup u && \text{weakly in } L^p(\Omega)^m, \\ \nabla u_{\ell_k} &\rightharpoonup \nabla u && \text{weakly in } L^p(\Omega)^{m \times n}, \text{ and} \\ \mathcal{J}(u_{\ell_k}) &\rightarrow \mathcal{J}(u) = \inf \mathcal{J}(\mathcal{A}). \end{aligned}$$

Moreover, unique solvability (i.e., $\#\operatorname{argmin} \mathcal{J}(\mathcal{A}) = 1$) implies weak convergence $u_{\ell} \rightharpoonup u$ of the entire sequence. Finally, if W is strictly convex, it even holds that

$$\nabla u_{\ell} \rightarrow \nabla u \quad \text{strongly in } L^p(\Omega)^{m \times n}.$$

Proof. Fix an arbitrary $v \in \mathcal{A}$ with finite energy. From Lemma 4, we infer

$$\mathcal{J}(u_{\ell}) \leq \mathcal{J}(\Pi_{\ell} v) \leq \mathcal{J}(v) + C_{\text{apx}} \|h_{\ell} f\|_{L^{p'}(\Omega)} \|\nabla v\|_{L^p(\Omega)}.$$

Thus, the sequence $(u_{\ell})_{\ell \in \mathbb{N}}$ has uniformly bounded energy, and hence, Lemma 5 provides a weakly convergent subsequence $(u_{\ell_k})_{k \in \mathbb{N}}$ with limit $u \in \mathcal{A}$. Since W is convex, \mathcal{J} is lower semicontinuous along the sequence u_{ℓ_k} [15, Theorem 3.4]. This gives

$$\mathcal{J}(u) \leq \liminf_{k \rightarrow \infty} \mathcal{J}(u_{\ell_k}) \leq \limsup_{k \rightarrow \infty} \mathcal{J}(u_{\ell_k}) \leq \limsup_{\ell \rightarrow \infty} \mathcal{J}(u_{\ell}) \leq \mathcal{J}(v). \quad (19)$$

Since $v \in \mathcal{A}$ was arbitrary, we deduce $u \in \operatorname{argmin} \mathcal{J}(\mathcal{A})$. Moreover, the choice $v = u$ yields equality in the latter estimate, and hence $\mathcal{J}(u) = \lim_k \mathcal{J}(u_{\ell_k}) = \inf \mathcal{J}(\mathcal{A})$.

The convergence $\mathcal{J}(u) = \lim_{\ell} \mathcal{J}(u_{\ell})$ follows from the fact that $\mathcal{J}(u) = \inf \mathcal{J}(\mathcal{A})$, i.e., that the limit is independent of the subsequence. Namely, if $(\tilde{u}_{\ell})_{\ell \in \mathbb{N}}$ is an arbitrary subsequence of $(u_{\ell})_{\ell \in \mathbb{N}}$ for which $(\mathcal{J}(\tilde{u}_{\ell}))_{\ell \in \mathbb{N}}$ is convergent, then the preceding arguments shows that $\lim_{\ell} \mathcal{J}(\tilde{u}_{\ell}) = \inf \mathcal{J}(\mathcal{A})$. In particular, $\liminf_{\ell} \mathcal{J}(u_{\ell}) = \limsup_{\ell} \mathcal{J}(u_{\ell}) = \inf \mathcal{J}(\mathcal{A})$.

If the minimizer $u \in \operatorname{argmin} \mathcal{J}(\mathcal{A})$ is unique, we can use the same kind of uniqueness argument to show that the entire sequence $(u_{\ell})_{\ell \in \mathbb{N}}$ converges weakly to u : More precisely, the preceding argument shows that any subsequence $(\tilde{u}_{\ell})_{\ell \in \mathbb{N}}$ of $(u_{\ell})_{\ell \in \mathbb{N}}$ has a weakly convergent subsequence $(\tilde{u}_{\ell_k})_{k \in \mathbb{N}}$, whose limit is the unique minimizer $u \in \mathcal{A}$. Consequently, the whole sequence $(u_{\ell})_{\ell \in \mathbb{N}}$ converges weakly to u .

Finally, if W is strictly convex, a result of Visintin [25] shows that weak convergence together with convergence of the energy implies strong convergence. \square

3.2. Refinement Indicators. The analysis of the previous section has demonstrated that condition (18) is sufficient in order to obtain convergence of the CR-FEM. It is therefore natural to use the quantities featured therein to steer the mesh refinement. Since the origin of the two quantities, $\|h_{\ell} f\|_{L^{p'}}$ and $\|h_{\ell}[u_{\ell}]\|_{L^1}$ is somewhat unusual (in particular, they do not arise from an *a posteriori* error estimate), we make a remark on their origin and interpretation.

Remark 2. The two quantities $\|h_{\ell} f\|_{L^{p'}}$ and $\|h_{\ell}[u_{\ell}]\|_{L^1}$ in (18) are closely linked to two conditions known in the calculus of variations as the *limsup condition* and the *liminf condition* of Γ -convergence [16] (or, simply, the *upper bound* and the *lower*

bound) and which guarantee convergence of minimizers of a sequence of minimization problems to a minimizer of the correct limit problem.

The main step in the Γ -convergence argument is (19) in the proof of Theorem 6. Note that guaranteeing $\|h_\ell f\|_{L^{p'}} \rightarrow 0$ provides the last inequality in (19) (the *limsup condition*), while guaranteeing $\|h_\ell[u_\ell]\|_{L^1} \rightarrow 0$ establishes weak convergence of the broken gradients which, together with convexity of W , provides the first inequality in (19) (the *liminf condition*).

Thus, our convergence indicators are not linked to any *a posteriori* error estimate in the usual sense, but arise from the use of the direct method of the calculus of variations (or Γ -convergence) in the weak convergence argument of Theorem 6. \square

In what follows, we discuss straightforward modifications of the convergence indicators which are more suitable for steering an adaptive algorithm, but for which our theory still applies. In order to associate the quantity $\|h_\ell f\|_{L^{p'}(\Omega)}$ to faces $E \in \mathcal{E}_\ell$, it is natural to define a related convergence indicator as

$$\eta_\ell = \sum_{E \in \mathcal{E}_\ell} \eta_\ell(E) = \sum_{E \in \mathcal{E}_\ell} h_E^{p'} \|f\|_{L^{p'}(\omega_E)}, \quad (20)$$

where $\omega_E = \bigcup\{T \in \mathcal{T}_\ell : E \subset T\}$ denotes the patch of elements adjacent to E .

Next, applying Hölder's inequality on each face E shows

$$\begin{aligned} \sum_{E \in \mathcal{E}_\ell} \int_E h_E |[u_\ell]| \, ds &\leq \sum_{E \in \mathcal{E}_\ell} \left(\int_E h_E \, ds \right)^{1/p'} \left(\int_E h_E |[u_\ell]|^p \, ds \right)^{1/p} \\ &\lesssim |\Omega|^{1/p'} \left(\sum_{E \in \mathcal{E}_\ell} h_E \int_E |[u_\ell]|^p \, ds \right)^{1/p}. \end{aligned} \quad (21)$$

Thus, a straightforward generalization of the indicator $\|h_\ell[u_\ell]\|_{L^1(\cup \mathcal{E}_\ell)}$, is given by

$$\mu_\ell^{(0)} = \sum_{E \in \mathcal{E}_\ell} \mu_\ell^{(0)}(E) = \sum_{E \in \mathcal{E}_\ell} h_E \| [u_\ell] \|_{L^p(E)}^p. \quad (22)$$

In many situations, this quantity is a bad candidate for steering the mesh refinement. To see this, let us consider the Dirichlet problem

$$-\Delta u = f \text{ in } \Omega \quad \text{with homogeneous boundary conditions} \quad u = 0 \text{ on } \partial\Omega, \quad (23)$$

where $W(F) = \frac{1}{2}|F|^2$, $p = 2$, $\Gamma^{(i)} = \partial\Omega$, and $g = 0$. For this problem, the “natural” error indicator is given by

$$\varrho_\ell^2 = \sum_{E \in \mathcal{E}_\ell \cap \Omega} h_E \int_E |[\nabla u_\ell]|^2 \, ds + \sum_{E \in \mathcal{E}_\ell \cap \partial\Omega} h_E \int_E |\partial_\tau u_\ell|^2 \, ds + \sum_{E \in \mathcal{E}_\ell} \eta_\ell(E),$$

where $\eta_\ell(E)$ is defined in (20) and where $\partial_\tau u_\ell$ denotes the tangential part of the gradient. This indicator is reliable and efficient (up to data oscillations) in the sense that

$$C_{\text{rel}}^{-1} \|u - u_\ell\|_{W^{1,2}(\Omega)} \leq \varrho_\ell \leq C_{\text{eff}} (\|u - u_\ell\|_{W^{1,2}(\Omega)} + \|h_\ell(f - \Pi_\ell f)\|_{L^2(\Omega)}),$$

and it leads to a convergent adaptive algorithm [10]. Furthermore, using the fact that normal jumps can be estimated above by the $\eta_\ell(E)$ terms (cf. [10, Theorem 3.5]) and

that the tangential jump of the gradient can be estimated by the tangential jump of the function, it follows that this indicator is equivalent to

$$\tilde{\varrho}_\ell^2 = \sum_{E \in \mathcal{E}_\ell} \left\{ h_E^{-1} \int_E |[u_\ell]|^2 ds + \eta_\ell(E)^2 \right\} = \sum_{E \in \mathcal{E}_\ell} \left\{ h_E^{-1} \|[u_\ell]\|_{L^2(E)}^2 + \eta_\ell(E)^2 \right\}$$

This argument suggests that the indicator $\mu_\ell^{(0)}$ from (22) is not suitable, since it uses the “wrong” scaling of the mesh size. It therefore appears natural to us to use the generalization

$$\mu_\ell^{(1)} = \sum_{E \in \mathcal{E}_\ell} h_E^{1-p} \int_E |[u_\ell]|^p ds = \sum_{E \in \mathcal{E}_\ell} h_E^{1-p} \|[u_\ell]\|_{L^p(E)}^p \quad (24)$$

as a convergence indicator. Simple scaling arguments show why this is, in fact, the correct generalization (cf. [8] and Lemma 7). We will now define a further refinement indicator which can be thought of as an interpolation between $\mu_\ell^{(0)}$ and $\mu_\ell^{(1)}$: for some fixed parameter $\alpha \in [0, 1]$, let

$$\mu_\ell^{(\alpha)} = \sum_{E \in \mathcal{E}_\ell} \mu_\ell^{(\alpha)}(E) = \sum_{E \in \mathcal{E}_\ell} h_E^{1-\alpha p} \|[u_\ell]\|_{L^p(E)}^p. \quad (25)$$

Although we have initially motivated the definition of $\mu_\ell^{(\alpha)}$ through the error estimator for the Dirichlet problem, we can give an alternative interpretation. On the Dirichlet boundary, $\mu_\ell^{(\alpha)}(E)$ weakly imposes the Dirichlet condition, while in the interior, it can be thought of as a measure of the local regularity of ∇u_ℓ . In this sense, it seems a reasonable indicator which is independent of the problem solved.

We conclude this discussion with two simple observations. The first allows us to replace the term $\|h_\ell[u_\ell]\|_{L^1(\cup \mathcal{E}_\ell)}$ in Theorem 6 by $\mu_\ell^{(\alpha)}$, while the second is intended to simplify the subsequent analysis.

Lemma 7. *Suppose that $0 \leq \alpha \leq 1$, $\ell \in \mathbb{N}$, and $u_\ell \in \mathcal{A}_\ell$, then*

$$\|h_\ell[u_\ell]\|_{L^1(\cup \mathcal{E}_\ell)}^p \leq C_\mu \mu_\ell^{(\alpha)}. \quad (26)$$

Furthermore, we have the bounds

$$\mu_\ell^{(\alpha)}(E) \leq C'_\mu \|h_\ell^{1-\alpha} \nabla u_\ell\|_{L^p(\omega_E)}^p \quad \text{for all interior faces } E \in \mathcal{E}_\ell^{\text{int}}, \quad (27)$$

$$\mu_\ell^{(\alpha)}(E) \leq C''_\mu (\|h_\ell^{1-\alpha} \nabla u_\ell\|_{L^p(\omega_E)}^p + \|h_\ell^{1-\alpha} \nabla g\|_{L^p(\omega_E)}^p) \quad \text{for all } E \in \mathcal{E}_\ell \setminus \mathcal{E}_\ell^{\text{int}}. \quad (28)$$

The constants C_μ , C'_μ , and C''_μ depend on the shape regularity $\sigma(\mathcal{T}_\ell)$, and C_μ additionally on $|\Omega|$ and on $\text{diam}(\Omega)$.

Proof. The first bound follows from (21). To prove (27), let $T^\pm \in \mathcal{T}_\ell$ denote the unique elements with $E = T^+ \cap T^- \in \mathcal{E}_\ell^{\text{int}}$ and $\omega_E = T^+ \cup T^-$. If z_E denotes the barycentre of E , then $[u_\ell](z_E) = 0$ yields

$$|[u_\ell]| \leq h_E |\nabla[u_\ell]| \leq h_E (|\nabla u_\ell|_{T^+} + |\nabla u_\ell|_{T^-}) \quad \text{pointwise on } E.$$

Shape regularity of \mathcal{T}_ℓ gives $|T^\pm| h_E^{(1-\alpha)p} \approx |E| h_E^p h_E^{1-\alpha p}$ and results in

$$\mu_\ell^{(\alpha)}(E) \lesssim h_E^{1-\alpha p} |E| h_E^p (|\nabla u_\ell|_{T^+}|^p + |\nabla u_\ell|_{T^-}|^p) \approx h_E^{(1-\alpha)p} \|\nabla u_\ell\|_{L^p(\omega_E)}^p.$$

To prove (28) note that $[u_\ell]^{(i)} = 0$ on E if $E \cap \Gamma^{(i)} = \emptyset$. We may therefore assume, without loss of generality, that $\Gamma^{(i)} = \partial\Omega$ for all $i = 1, \dots, m$, and hence $[u_\ell] = g - u_\ell$ on $\partial\Omega$. The trace inequality reads

$$h_E \|g - u_\ell\|_{L^p(E)}^p \lesssim \|g - u_\ell\|_{L^p(T)}^p + h_E^p \|\nabla(g - u_\ell)\|_{L^p(T)}^p.$$

Note that $\int_E (g - u_\ell) ds = 0$ by definition of $\mathcal{A}_\ell \ni u_\ell$, and recall that in the proof of Lemma 2 it was sufficient to have mean zero on one single face to obtain the first-order approximation property. This provides $\|g - u_\ell\|_{L^p(T)} \lesssim h_T \|\nabla(g - u_\ell)\|_{L^p(T)}$, which gives

$$\mu_\ell^{(\alpha)}(E) = h_E^{1-\alpha p} \|[u_\ell]\|_{L^p(E)}^p = h_E^{1-\alpha p} \|g - u_\ell\|_{L^p(E)}^p \lesssim h_E^{p-\alpha p} \|\nabla(g - u_\ell)\|_{L^p(T)}^p$$

and immediately implies (28). \square

3.3. Adaptive Strategy. We are now in a position to formulate an adaptive mesh refinement strategy for the solution of (13). In what follows, $\alpha \in [0, 1]$ is an arbitrary but fixed parameter of the algorithm.

Algorithm 1. Input: *Marking parameters* $\theta \in (0, 1]$, $\alpha \in [0, 1]$; *Initial mesh* \mathcal{T}_0 . *Set* $\ell = 0$.

- (a) *Compute a discrete minimizer* $u_\ell \in \operatorname{argmin} \mathcal{J}(\mathcal{A}_\ell)$.
- (b) *Compute refinement indicators* η_ℓ and $\mu_\ell^{(\alpha)}$ from (20) and (25), respectively.
- (c) *Generate a set of marked faces* $\mathcal{M}_\ell \subseteq \mathcal{E}_\ell$ such that

$$\sum_{E \in \mathcal{M}_\ell} (\eta_\ell(E) + \mu_\ell^{(\alpha)}(E)) \geq \theta(\eta_\ell + \mu_\ell^{(\alpha)}) \quad (29)$$

- (d) *Generate a regular triangulation* $\mathcal{T}_{\ell+1}$, where at least the marked faces $E \in \mathcal{M}_\ell$ are refined.
- (e) *Increase* $\ell \mapsto \ell + 1$ and go to (a). \square

A marking strategy satisfying (29) is often called *Dörfler marking*. It was a crucial ingredient in the first convergence proofs of the adaptive finite element method and has been identified to also play an important role in obtaining optimal convergence rates [13], in which case the cardinality of the set \mathcal{M}_ℓ should be minimal. Generically, the value $\theta = 1$ corresponds to uniform refinement, whereas small θ leads to highly adapted meshes.

We use newest vertex bisection in step (4) to ensure that the sequence of triangulations $(\mathcal{T}_\ell)_{\ell \in \mathbb{N}}$ generated by Algorithm 1 is uniformly shape-regular. However, besides the uniform shape-regularity, the following convergence result only requires that marked faces are reduced by a uniform factor $\kappa \in (0, 1)$, i.e.,

$$h_{E'} \leq \kappa h_E \quad \text{for all } E' \in \mathcal{E}_{\ell+1} \text{ with } E' \subset E \in \mathcal{M}_\ell. \quad (30)$$

Thus, the precise refinement rule in step (d) is fairly arbitrary.

Theorem 8 (Convergence of the Adaptive Algorithm). *Suppose that the sequence $(\mathcal{T}_\ell)_{\ell \in \mathbb{N}}$ generated by Algorithm 1 is uniformly shape-regular, that it satisfies (30), and assume in addition that $0 \leq \alpha < 1$. Then the refinement indicators converge to zero, i.e.,*

$$\eta_\ell + \mu_\ell^{(\alpha)} \xrightarrow{\ell \rightarrow \infty} 0, \quad (31)$$

and the sequence $(u_\ell)_{\ell \in \mathbb{N}}$ of discrete minimizers satisfies the conditions of Theorem 6.

Proof. This proof is largely inspired by [22].

As in the proof of Theorem 6, it follows that $\mathcal{J}(u_\ell)$ and hence $\|u_\ell\|_{L^p(\Omega)} + \|\nabla u_\ell\|_{L^p(\Omega)}$ are bounded sequences.

To abbreviate the notation, we now drop the superscript (α) in $\mu_\ell^{(\alpha)}$, and we write

$$\eta_\ell(\mathcal{S}_\ell) := \sum_{E \in \mathcal{S}_\ell} \eta_\ell(E) \quad \text{and} \quad \mu_\ell(\mathcal{S}_\ell) := \sum_{E \in \mathcal{S}_\ell} \mu_\ell(E) \quad \text{for all } \mathcal{S}_\ell \subseteq \mathcal{E}_\ell.$$

We consider the set of all faces resp. all elements which are eventually not refined, i.e.,

$$\tilde{\mathcal{E}} := \bigcap_{k \geq 0} \bigcup_{\ell \geq k} \mathcal{E}_\ell \quad \text{and} \quad \tilde{\mathcal{T}} := \bigcap_{k \geq 0} \bigcup_{\ell \geq k} \mathcal{T}_\ell.$$

It is evident that $T \in \tilde{\mathcal{T}}$ if, and only if, all faces of T belong to $\tilde{\mathcal{E}}$. For the proof of (31), we split the indicators into

$$\eta_\ell + \mu_\ell = (\eta_\ell(\mathcal{E}_\ell \setminus \tilde{\mathcal{E}}) + \mu_\ell(\mathcal{E}_\ell \setminus \tilde{\mathcal{E}})) + (\eta_\ell(\mathcal{E}_\ell \cap \tilde{\mathcal{E}}) + \mu_\ell(\mathcal{E}_\ell \cap \tilde{\mathcal{E}})).$$

Step 1: In the first step, we will prove that

$$\eta_\ell(\mathcal{E}_\ell \setminus \tilde{\mathcal{E}}) + \mu_\ell(\mathcal{E}_\ell \setminus \tilde{\mathcal{E}}) \xrightarrow{\ell \rightarrow \infty} 0. \quad (32)$$

Recall that $\omega_E := \bigcup \{T \in \mathcal{T}_\ell : E \subset \partial T\}$. Setting $\tilde{\Omega}_\ell = \bigcup \{\omega_E : E \in \mathcal{E}_\ell \setminus \tilde{\mathcal{E}}\}$, we first claim that $\chi_{\tilde{\Omega}_\ell} h_\ell \xrightarrow{\ell \rightarrow \infty} 0$ a.e. in Ω . To see this, fix $x \in \Omega \setminus (\bigcup_\ell \bigcup \mathcal{E}_\ell)$ outside of the skeletons of all \mathcal{T}_ℓ which form a null-set. For each ℓ , there is a unique element $T_\ell \in \mathcal{T}_\ell$ with $x \in T_\ell$. If $\lim_\ell h_{T_\ell} = 0$, we conclude $\lim_\ell (\chi_{\tilde{\Omega}_\ell} h_\ell)(x) = 0$. Otherwise, T_ℓ is only refined finitely many times, i.e., there holds $T_\ell = T_{\ell_0}$ for some $\ell_0 \in \mathbb{N}$ and all $\ell \geq \ell_0$, i.e., $T_\ell \in \tilde{\mathcal{T}}$ and therefore its faces belong to $\tilde{\mathcal{E}}$. Consequently, $x \notin \tilde{\Omega}_\ell$ for all $\ell \geq \ell_0$, and hence $(\chi_{\tilde{\Omega}_\ell} h_\ell)(x) = 0$ for $\ell \geq \ell_0$. We have therefore shown that

$$\lim_{\ell \rightarrow \infty} \chi_{\tilde{\Omega}_\ell} h_\ell = 0 \quad \text{pointwise a.e. in } \Omega.$$

During mesh-refinement, the local mesh-size h_ℓ is pointwise decreasing. Consequently, the dominated convergence theorem yields

$$\chi_{\tilde{\Omega}_\ell} h_\ell^\beta \psi \xrightarrow{\ell \rightarrow \infty} 0 \quad \text{strongly in } L^q(\Omega), \quad (33)$$

for all $\beta > 0$, and $\psi \in L^q(\Omega)$. With $\beta = 1$ and $\psi = f$, we infer

$$\eta_\ell(\mathcal{E}_\ell \setminus \tilde{\mathcal{E}}) = \sum_{E \in \mathcal{E}_\ell \setminus \tilde{\mathcal{E}}} \|h_\ell f\|_{L^{p'}(\omega_E)}^{p'} \lesssim \|h_\ell f\|_{L^{p'}(\tilde{\Omega}_\ell)}^{p'} = \|\chi_{\tilde{\Omega}_\ell} h_\ell f\|_{L^{p'}(\Omega)}^{p'} \xrightarrow{\ell \rightarrow \infty} 0.$$

Before we prove convergence of $\mu_\ell(\mathcal{E}_\ell \setminus \tilde{\mathcal{E}})$ to zero, it is instructive to consider the refinement indicator $\|h_\ell[u_\ell]\|_{L^1(\cup \mathcal{E}_\ell)}$ first. Using the facts that $h_E \approx h_\ell$ in ω_E , and that $[u_\ell](z_E) = 0$, we can estimate

$$\int_E h_E |[u_\ell]| \, ds \lesssim h_E^2 \int_E |[\nabla u_\ell]| \, ds \lesssim \int_{\omega_E} h_\ell |\nabla u_\ell| \, dx.$$

Summing over $E \in \mathcal{E}_\ell \setminus \tilde{\mathcal{E}}$ and using Hölder's inequality, we obtain

$$\|h_\ell[u_\ell]\|_{L^1(\cup(\mathcal{E}_\ell \setminus \tilde{\mathcal{E}}))} \lesssim \sum_{E \in \mathcal{E}_\ell \setminus \tilde{\mathcal{E}}} \|h_\ell \nabla u_\ell\|_{L^1(\omega_E)} \lesssim \|h_\ell \nabla u_\ell\|_{L^1(\tilde{\Omega}_\ell)} \leq \|h_\ell\|_{L^{p'}(\tilde{\Omega}_\ell)} \|\nabla u_\ell\|_{L^p(\tilde{\Omega}_\ell)}.$$

According to (33), with $\beta = 1$ and $\psi = 1$, the upper bound tends to zero as $\ell \rightarrow \infty$.

If we attempt to use the same idea for proving that $\mu_\ell(\mathcal{E}_\ell \setminus \tilde{\mathcal{E}}) \rightarrow 0$ then, using (27) and (28), we first obtain the following bound:

$$\mu_\ell(\mathcal{E}_\ell \setminus \tilde{\mathcal{E}}) \lesssim \|h_\ell^{(1-\alpha)} \nabla u_\ell\|_{L^p(\tilde{\Omega}_\ell)}^p + \|h_\ell^{(1-\alpha)} \nabla g\|_{L^p(\tilde{\Omega}_\ell)}^p. \quad (34)$$

Using (33), with $\beta = 1 - \alpha > 0$ and $\psi = \nabla g$, we immediately find that the second integral on the right-hand side of (34) converges to zero. It thus only remains to treat the first term on the right-hand side. Unfortunately, we control ∇u_ℓ only in $L^p(\Omega)^{m \times n}$. Therefore, we cannot immediately use Hölder's inequality as before to verify that the first term on the right-hand side of (34) tends to zero. Since we don't know whether ∇u_ℓ converges pointwise a.e., we have no hope of using Fatou's lemma either. Instead, we make use of the additional flexibility provided by the condition $\alpha < 1$ to estimate

$$\begin{aligned} \|h_\ell^{(1-\alpha)} \nabla u_\ell\|_{L^p(\tilde{\Omega}_\ell)}^p &= \int_{\tilde{\Omega}_\ell} h_\ell^{(1-\alpha)p} |\nabla u_\ell|^p \, dx \leq \sum_{\substack{T \in \mathcal{T}_\ell \\ T \subseteq \tilde{\Omega}_\ell}} h_T^{n+(1-\alpha)p} |\nabla u_\ell|_T|^p \\ &\leq \left(\sum_{\substack{T \in \mathcal{T}_\ell \\ T \subseteq \tilde{\Omega}_\ell}} h_T^{(n+(1-\alpha)p)q/p} |\nabla u_\ell|_T|^q \right)^{p/q} \\ &\lesssim \left(\sum_{\substack{T \in \mathcal{T}_\ell \\ T \subseteq \tilde{\Omega}_\ell}} h_T^{nq/p+(1-\alpha)q-n} \int_T |\nabla u_\ell|^q \right)^{p/q}, \end{aligned}$$

where $1 \leq q < p$. In the second estimate above we used the bound $\|\cdot\|_{\ell^p} \leq \|\cdot\|_{\ell^q}$. Setting $\beta = n(q/p - 1) + (1 - \alpha)q$, we obtain

$$\|h_\ell^{(1-\alpha)} \nabla u_\ell\|_{L^p(\tilde{\Omega}_\ell)}^p \lesssim \left(\int_{\tilde{\Omega}_\ell} h_\ell^\beta |\nabla u_\ell|^q \, dx \right)^{p/q} \leq \|h_\ell^\beta \chi_{\tilde{\Omega}_\ell}\|_{L^{p/(p-q)}}^{p/q} \|\nabla u_\ell\|_{L^p(\Omega)}^p.$$

by use of Hölder's inequality. For this bound to tend to zero, we require that $\beta > 0$ which can be achieved by choosing q sufficiently close to p and using the fact that $\alpha < 1$. Thus, we have successfully established (32).

Step 2: In the second step, we use the properties of our marking strategy to conclude the proof of (31). Namely, observe that at step ℓ any marked face will be

refined during this step, i.e.,

$$\mathcal{M}_\ell \subset \mathcal{E}_\ell \setminus \tilde{\mathcal{E}}.$$

Therefore, (29) and *Step 1* imply that

$$\theta(\eta_\ell + \mu_\ell) \leq \eta_\ell(\mathcal{M}_\ell) + \mu_\ell(\mathcal{M}_\ell) \leq \eta_\ell(\mathcal{E}_\ell \setminus \tilde{\mathcal{E}}) + \mu_\ell(\mathcal{E}_\ell \setminus \tilde{\mathcal{E}}) \rightarrow 0,$$

as $\ell \rightarrow \infty$, which concludes the proof. \square

Remark 3. We have already remarked in Section 3.2 that, for $\alpha < 1$, our refinement indicators are not reliable error indicators, even for a simple Dirichlet problem with homogeneous boundary conditions. Furthermore, our refinement indicators have no information about the free boundary which gives further indication to its “incompleteness”. We found it therefore somewhat surprising that we were able to prove convergence of our adaptive strategy.

We note also that our proof does not extend in an obvious way to conforming finite element methods, where the upper bound (15) is false even for quadratic functionals. \square

Remark 4. We have only proven convergence of our adaptive algorithm for the case $\alpha < 1$. It does not appear straightforward to include the case $\alpha = 1$ as well. The analysis in [22] shows that obtaining strong convergence of the sequence $(u_\ell)_{\ell \in \mathbb{N}}$ to some \tilde{u} *a priori* (instead of merely weak convergence) is the key. However, this appears difficult for problems of the generality which we consider here.

In practice, one may safely ignore this fact and choose $\alpha = 1$, possibly implementing a safeguard strategy which changes α if it should become apparent that $\eta_\ell + \mu_\ell \not\rightarrow 0$. \square

4. NUMERICAL EXPERIMENTS

We have implemented Algorithm 1 for two two-dimensional model problems: the Laplace problem with Dirichlet and with Neumann boundary conditions as well as the example of Foss, Hrusa, and Mizel [19] which exhibits a Lavrentiev gap. Before we present the computational experiments, we briefly outline the details of our implementation.

- (a) The solution of the optimization problem is achieved by a damped Newton method if it is nonlinear and a direct solver if it is linear.
- (b) We have found that Dörfler’s marking strategy with a minimal set \mathcal{M}_ℓ yields very slow mesh growth for the highly nonlinear and singular problems which we consider here. Therefore, our strategy is to mark a fixed fraction of edges (with largest indicators) for refinement, i.e.,

$$\#\mathcal{M}_\ell \geq \theta \#\mathcal{E}_\ell \quad \text{with} \quad \min_{E \in \mathcal{M}_\ell} (\eta_\ell(E) + \mu_\ell^{(\alpha)}(E)) \geq \max_{E \in \mathcal{E}_\ell \setminus \mathcal{M}_\ell} (\eta_\ell(E) + \mu_\ell^{(\alpha)}(E)).$$

Note that then,

$$\sum_{E \in \mathcal{M}_\ell} (\eta_\ell(E) + \mu_\ell^{(\alpha)}(E)) \geq \frac{\theta}{1 + \theta} \sum_{E \in \mathcal{E}_\ell} (\eta_\ell(E) + \mu_\ell^{(\alpha)}(E)),$$

- so that Dörfler marking (29) still holds with θ replaced by $\theta/(1+\theta)$. We usually chose $\theta = 0.25$ which roughly doubles the number of elements at each iteration.
- (c) The mesh refinement is achieved via *newest vertex bisection* which halves every marked edge and which preserves shape regularity.
 - (d) We terminate the algorithm when a prescribed number of elements is attained.

For the computations in Section 4.1, we estimate the error for the energy by comparing it to a conforming computation. If $f \equiv 0$, then

$$\mathcal{J}(u_\ell) \leq \inf \mathcal{J}(\mathcal{A}) \leq \mathcal{J}(\bar{u}) \quad \text{for all } \bar{u} \in \mathcal{A}.$$

If f is non-zero then the above estimate depends on an unknown quantity, namely $\|\nabla u\|_{L^p}$ where $u \in \operatorname{argmin} \mathcal{J}(\mathcal{A})$. However, we have observed that even in that case, $\mathcal{J}(u_\ell)$ is monotonically increasing towards the energy of the exact solution. Therefore, we compute a conforming \bar{u} using a standard adaptive P_1 -finite element method [9] and take

$$\inf \mathcal{J}(\mathcal{A}) - \mathcal{J}(u_\ell) \leq \mathcal{J}(\bar{u}) - \mathcal{J}(u_\ell)$$

as a slightly heuristic energy error estimate.

4.1. Linear Laplacian. We begin our experiments with the Laplace equation on the slit domain $\Omega = (-1, 1)^2 \setminus [0, 1] \times \{0\}$. It is equivalently formulated by setting $W(F) = \frac{1}{2}|F|^2$ with $m = 1$ and $n = 2$. First, we consider the pure Dirichlet problem

$$-\Delta u = 1 \text{ in } \Omega \quad \text{with homogeneous boundary conditions} \quad u = 0 \text{ on } \partial\Omega,$$

where in the energy formulation

$$\Gamma^{(1)} = \partial\Omega, \quad f = 1, \quad g = 0. \tag{35}$$

In order to investigate the effect of a Neumann boundary, we also consider the mixed boundary value problem

$$-\Delta u = 0 \text{ in } \Omega \quad \text{with} \quad \partial u / \partial n = 0 \text{ on } \Gamma_N \quad \text{and} \quad u = g \text{ on } \Gamma_D,$$

where $|\Gamma_D \cap \Gamma_N| = 0$ and $\partial\Omega = \Gamma_D \cup \Gamma_N$. We choose

$$\Gamma^{(1)} = \Gamma_D = \partial\Omega \cap \{x_1 = 1\}, \quad f = 0, \quad g(1, x_2) = \operatorname{sign}(x_2). \tag{36}$$

The convergence rates for problems (35) and (36) are shown in Figures 1 and 2, respectively. As expected, we observe that the accuracy improves as α approaches 1.0. In the Dirichlet problem, the convergence rate for $\alpha = 1.0$ and for $\alpha = 0.9$ can barely be distinguished. What is surprising though is that, for the Neumann problem, the value of α does not seem to affect the convergence rate at all. We have no explanation for this effect but we note that we will also observe it in our second model problem.

4.2. Lavrentiev Phenomenon. For our second numerical experiment, we use an example which exhibits the Lavrentiev gap phenomenon — the focus of our investigation. To this end, we slightly modify the example of Foss, Hrusa and Mizel [19].

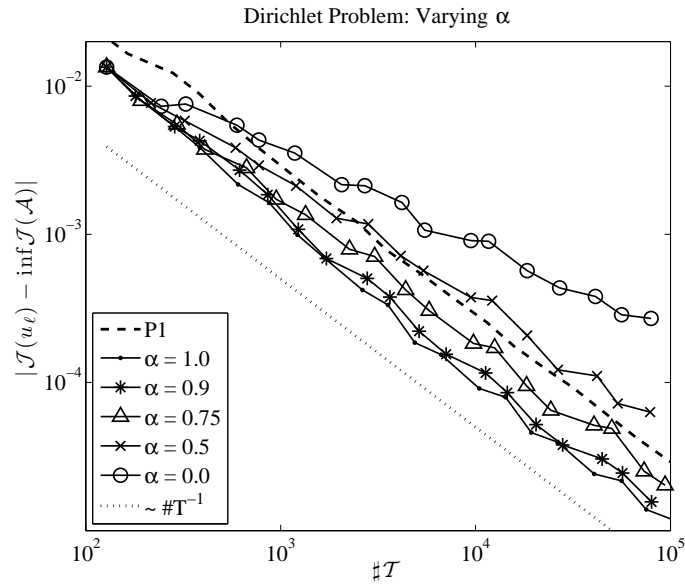


FIGURE 1. Convergence rates for Algorithm 1 applied to problem (35). As $\alpha \nearrow 1$, the convergence rate approaches the optimal rate $\#\mathcal{T}^{-1}$.

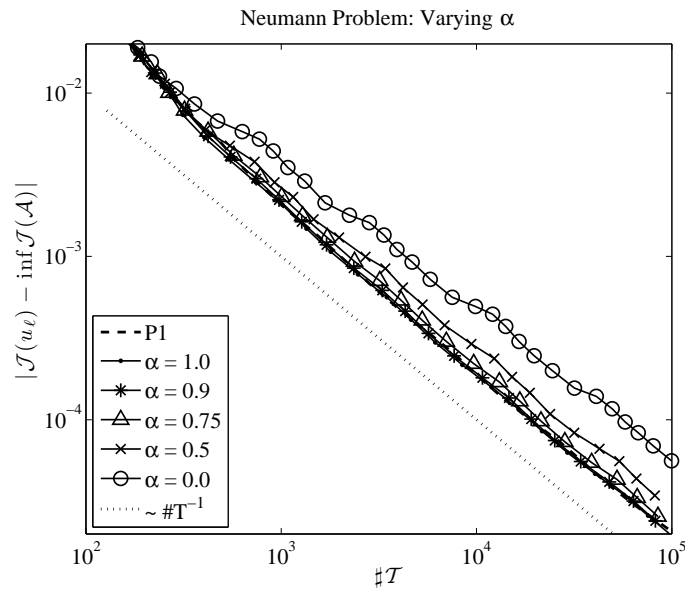


FIGURE 2. Convergence rates for Algorithm 1 applied to problem (36). Contrary to intuition, the convergence rate seems to be optimal, independent of the parameter α .

Let $m = n = 2$, let the domain be the half disk $\Omega = \{|x| < 1, x_2 > 0\}$, and let $\Gamma^{(1)} = (-1, 0) \times \{0\} \cup \{|x| = 1, x_2 > 0\}$ and $\Gamma^{(2)} = (0, 1) \times \{0\} \cup \{|x| = 1, x_2 > 0\}$.

Furthermore, $f \equiv 0$, $g^{(i)} = 0$ on $\{x_2 = 0\}$, and $g = (\cos \frac{\theta}{2}, \sin \frac{\theta}{2})$ on $\{|x| = 1\}$ in polar coordinates (r, θ) . Thus, admissible functions are deformations of the half disk Ω into the quarter disk $\{|x| < 1, x_1 > 0, x_2 > 0\}$. Suppose, for the moment, that the stored energy density is given by

$$W(F) = (|F|^2 - 2 \det F)^4.$$

Convexity of W follows immediately from the fact that $F \mapsto (|F|^2 - 2 \det F)$ is a non-negative quadratic form. However, the associated energy functional is not coercive. Nevertheless, Foss, Hrusa, and Mizel showed in [19] that it exhibits the Lavrentiev gap phenomenon. The global minimum of \mathcal{J} in \mathcal{A} is the function

$$\bar{u} = r^{1/2} \left(\cos \frac{\theta}{2}, \sin \frac{\theta}{2} \right),$$

for which $\mathcal{J}(\bar{u}) = 0$. It is furthermore easy to verify that \bar{u} also minimizes the Dirichlet integral for the same boundary conditions. Consequently, for $p = 2$, \bar{u} is also the global minimizer of

$$\mathcal{J}_p(v) = \int_{\Omega} W_p(\nabla v) \, dx, \tag{37}$$

in \mathcal{A} , where

$$W_p(F) = (|F|^2 - 2 \det F)^4 + \frac{1}{p} (|F_1|^p + |F_2|^p).$$

Since we know the solution for $p = 2$ explicitly, we can explicitly compute the energy minimum,

$$\inf \mathcal{J}_2(\mathcal{A}) = \mathcal{J}_2(\bar{u}) = \pi/4.$$

In Figure 3, we plot the convergence rate for the minimization problem, for varying α . We observe the same effect as for the Neumann problem in Section 4.1: surprisingly, the convergence rate seems to be independent of the parameter α . While it is encouraging that the convergence rate for the energy appears to be linear, despite the fact that we are solving a highly non-linear and singular problem (note that \mathcal{J}_2 is not even continuous in the strong topology of $W^{1,2}(\Omega)^2$), we strongly suspect that this is related to the particularly simple structure of W_2 and the fact that $\nabla \bar{u}$ minimizes the first term of $W_2(\nabla \bar{u})$ pointwise.

4.3. Verification of Lavrentiev gaps. In our final experiment, we demonstrate how one could verify whether a given minimization problem exhibits a Lavrentiev gap. We consider the energy functional \mathcal{J}_p from (37). For the parameters $p = 2, 3, 4, 6$, we apply Algorithm 1 with the minimization problem $\operatorname{argmin} \mathcal{J}_p(\mathcal{A})$ and obtain discrete solutions u_ℓ . In addition, we compute an adaptive P_1 -solution \bar{u}_ℓ for the same problem (though possibly on different meshes) and we plot the difference in energy $\mathcal{J}_p(\bar{u}_\ell) - \mathcal{J}_p(u_\ell)$. The theory in [19] would lead us to expect (but except for the case $p = 2$ this is not at all clear) that, for $p = 2, 3$ a Lavrentiev gap occurs, while for $p = 4, 6$ no gap occurs. The computations which we show in Figure 4 agree with this prediction, except possibly in the case $p = 4$, where they suggest that no Lavrentiev gap may, in fact, be present.

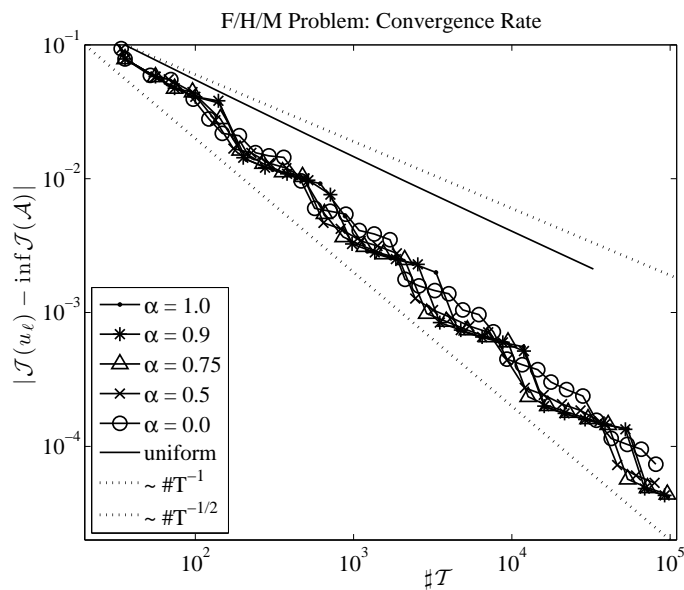


FIGURE 3. Convergence rates for Algorithm 1 applied to the minimization problem $u \in \operatorname{argmin} \mathcal{J}_2(\mathcal{A})$, with varying marking parameter α .

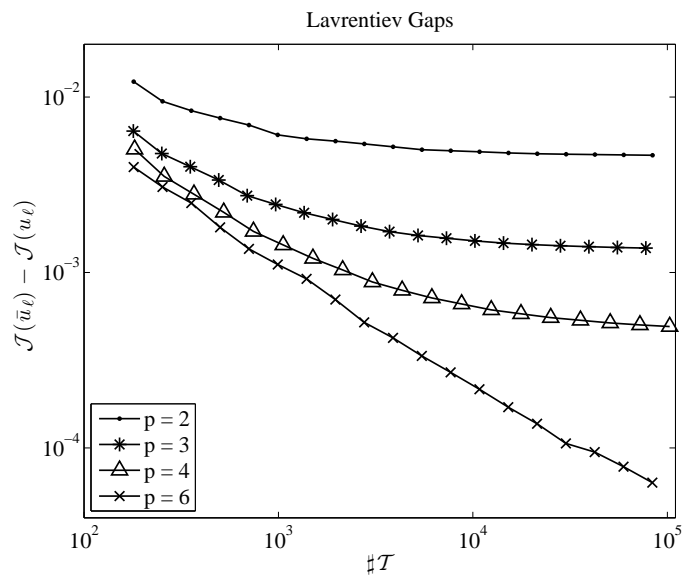


FIGURE 4. Adaptive computation of the Lavrentiev gap $\inf \mathcal{J}_p(\mathcal{A}_\infty) - \inf \mathcal{J}_p(\mathcal{A})$ for $p = 2, 3, 4, 6$. Contrary to intuition, for $p = 4$, the computation suggests that a Lavrentiev gap is present.

5. CONCLUSION

We have presented an adaptive finite element algorithm for the solution of convex variational problems. Despite the fact that the refinement indicators are *not* reliable error indicators in any classical sense, we have succeeded in proving convergence of our adaptive scheme. To conclude, we briefly mention some possible generalizations of our analysis.

In order for the minimization problem (5) to be well-posed (or rather, for the direct method technique to apply), it is necessary that \mathcal{J} is coercive in \mathcal{A} which, in particular, requires that elements of \mathcal{A} satisfy a Poincaré-type inequality. Similarly, we require a broken Poincaré-type inequality for the discrete admissible set \mathcal{A}_ℓ in order to be able to extract weakly convergent subsequences. The entire analysis applies whenever such a broken Poincaré-type inequality is available for \mathcal{A}_ℓ . It is therefore straightforward to generalize the results, for example, to problems involving pointwise constraints on the function (e.g., an obstacle problem) or on the gradient (e.g., problems arising in plasticity).

A second important generalization is to allow W to depend on x and on u . It is not at all clear in which generality this can be achieved. Mild dependencies such as piecewise constant dependence on x are easily included in the analysis, however, a strong coupling of (x, u) to ∇u must be avoided. This follows immediately upon considering Manià's functional [20]

$$\mathcal{J}(u) = \int_0^1 u_x^6 (u^3 - x)^2 dx,$$

which is to be minimized subject to $u(0) = 0, u(1) = 1$. Since the CR-FEM reduces to the P_1 -FEM in one dimension, and since Manià's example exhibits a Lavrentiev gap, it follows that a sufficiently strong coupling of (x, u) to ∇u destroys the convergence of the method.

Finally, the generalization to polyconvex or even quasiconvex W is even more difficult. Here, both the upper bound (15) and the lower bound (19), i.e., the weak lower-semicontinuity of \mathcal{J} along the sequence u_ℓ , are completely open.

REFERENCES

- [1] G. Acosta and R. G. Durán. An optimal Poincaré inequality in L^1 for convex domains. *Proc. Amer. Math. Soc.*, 132(1):195–202, 2004.
- [2] R.A. Adams and J.F. Fournier. *Sobolev Spaces*. Elsevier Science, 2003.
- [3] J. M. Ball. Singularities and computation of minimizers for variational problems. In *Foundations of computational mathematics (Oxford, 1999)*, volume 284 of *London Math. Soc. Lecture Note Ser.*, pages 1–20. Cambridge Univ. Press, Cambridge, 2001.
- [4] S. Bartels and C. Carstensen. A convergent adaptive finite element method for an optimal design problem. *Numer. Math.*, 108(3):359–385, 2008.
- [5] D. Braess. *Finite Elements*. Cambridge University Press, Cambridge, third edition, 2007. Theory, fast solvers, and applications in elasticity theory, Translated from the German by Larry L. Schumaker.
- [6] S. C. Brenner. Poincaré-Friedrichs inequalities for piecewise H^1 functions. *SIAM J. Numer. Anal.*, 41(1):306–324, 2003.

- [7] F. Brezzi and M. Fortin. *Mixed and hybrid finite element methods*, volume 15 of *Springer Series in Computational Mathematics*. Springer-Verlag, New York, 1991.
- [8] A. Buffa and C. Ortner. Compact embeddings of broken Sobolev Spaces and applications. To appear in *IMA J. Numer. Anal.* (doi:10.1093/imanum/drn038).
- [9] C. Carstensen. Convergence of an adaptive FEM for a class of degenerate convex minimization problems. *IMA J. Numer. Anal.*, 28(3):423–439, 2008.
- [10] C. Carstensen and R. H. W. Hoppe. Convergence analysis of an adaptive nonconforming finite element method. *Numer. Math.*, 103(2):251–266, 2006.
- [11] C. Carstensen and R. H. W. Hoppe. Error reduction and convergence for an adaptive mixed finite element method. *Math. Comp.*, 75(255):1033–1042, 2006.
- [12] C. Carstensen and C. Ortner. Computation of the Lavrentiev phenomenon. OxMoS Preprint No. 17 (2009).
- [13] J. M. Cascon, C. Kreuzer, R. H. Nochetto, and K. G. Siebert. Quasi-optimal convergence rate for an adaptive finite element method. *SIAM J. Numer. Anal.*, 46(5):2524–2550, 2008.
- [14] M. Crouzeix and P.-A. Raviart. Conforming and nonconforming finite element methods for solving the stationary Stokes equations. I. *Rev. Française Automat. Informat. Recherche Opérationnelle Sér. Rouge*, 7(R-3):33–75, 1973.
- [15] B. Dacorogna. *Direct methods in the calculus of variations*, volume 78 of *Applied Mathematical Sciences*. Springer-Verlag, Berlin, 1989.
- [16] G. Dal Maso. *An introduction to Γ -convergence*. Progress in Nonlinear Differential Equations and their Applications, 8. Birkhäuser Boston Inc., Boston, MA, 1993.
- [17] L. Diening and C. Kreuzer. Linear convergence of an adaptive finite element method for the p -Laplacian equation. *SIAM J. Numer. Anal.*, 46(2):614–638, 2008.
- [18] W. Dörfler. A convergent adaptive algorithm for Poisson’s equation. *SIAM J. Numer. Anal.*, 33(3):1106–1124, 1996.
- [19] M. Foss, W. J. Hrusa, and V. J. Mizel. The Lavrentiev gap phenomenon in nonlinear elasticity. *Arch. Ration. Mech. Anal.*, 167(4):337–365, 2003.
- [20] B. Maniá. Sopra un esempio di Lavrentieff. *Boll. Un. Mat. Ital.*, 13:147–153, 1934.
- [21] P. Morin, R. H. Nochetto, and K. G. Siebert. Data oscillation and convergence of adaptive FEM. *SIAM J. Numer. Anal.*, 38(2):466–488, 2000.
- [22] P. Morin, K. G. Siebert, and A. Veeger. A basic convergence result for conforming adaptive finite elements. *Math. Models Methods Appl. Sci.*, 18(5):707–737, 2008.
- [23] C. Ortner. A non-conforming finite element method for convex minimization problems. Submitted for publication.
- [24] A. Veeger. Convergent adaptive finite elements for the nonlinear Laplacian. *Numer. Math.*, 92(4):743–770, 2002.
- [25] A. Visintin. Strong convergence results related to strict convexity. *Comm. Partial Differential Equations*, 9(5):439–466, 1984.