# CONVERGENCE OF ADAPTIVE FEM FOR A CLASS OF DEGENERATE CONVEX MINIMIZATION PROBLEMS

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ABSTRACT. A class of degenerate convex minimization problems allows for some adaptive finite element method (AFEM) to compute strongly converging stress approximations. The algorithm AFEM consists of successive loops of the form

$$\mathtt{SOLVE} o \mathtt{ESTIMATE} o \mathtt{MARK} o \mathtt{REFINE}$$

and employs the bulk criterion. The convergence in  $L^{p'}(\Omega; \mathbb{R}^{m \times n})$  relies on new sharp strict convexity estimates of degenerate convex minimization problems with

$$\mathcal{J}(v) := \int_{\Omega} W(Dv) \, dx - \int_{\Omega} fv \, dx \quad \text{for } v \in V := W_0^{1,p}(\Omega; \mathbb{R}^m).$$

The class of minimization problems includes strong convex problems and allows applications in an optimal design task, Hencky elastoplasticity, or relaxation of 2-well problems allowing for microstructures.

## 1. Class of Convex Minimization Problems

This section specifies a class of  $C^1$  energy densities  $W: \mathbb{R}^{m \times n} \to \mathbb{R}$  characterized by (H1)-(H2) for some constants  $1 , and <math>0 \le s < \infty$  with

$$\max\{(1+s/r)/(1-1/r), 2n/(n+2)\} \le p,$$

through the two-sided growth condition

(H1) 
$$|F|^p - 1 \lesssim W(F) \lesssim 1 + |F|^p \text{ for all } F \in \mathbb{R}^{m \times n}$$

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and the *convexity control* 

(H2) 
$$(1+|A|^s+|B|^s)^{-1}|DW(A) - DW(B)|^r$$
  $\lesssim W(B) - W(A) - DW(A) : (B-A) \text{ for all } A, B \in \mathbb{R}^{m \times n}.$ 

Here and throughout "·" denotes the scalar product in  $\mathbb{R}^m$ , "·" denotes the scalar product in  $\mathbb{R}^{m \times n}$ , and the expression " $\lesssim$ " abbreviates an inequality up to some multiplicative generic constant, i.e.,  $A \lesssim B$  means  $A \leq cB$  with some generic constant c > 0, which is independent of the arguments A, B, F in (H1)-(H2) (but may depend on W and on the aspect ratio of finite element triangulations).

Finally, t := 1 + s/p and the Hölder conjugate p' of p satisfy

$$1 < p' \le r/t < \infty$$
, and  $1/p + 1/p' = 1$ 

and where r/t and r/(r-t) are conjugate exponents, i.e., t/r+(r-t)/r=1.

Section 3 exposes a list of examples with (H1)-(H2). The two-sided growth control (H1) is standard in the form of

$$|F|^p \lesssim W(F) + 1$$
 and  $W(F) \lesssim 1 + |F|^p$ .

By adding a constant to W(F), it could be replaced even by

$$|F|^p \lesssim W(F) \lesssim 1 + |F|^p$$
.

The convexity control (H2) implies the monotonicity condition

(H3) 
$$(1+|A|^s+|B|^s)^{-1}|DW(A)-DW(B)|$$

$$\lesssim (DW(A)-DW(B)): (A-B) \text{ for all } A,B \in \mathbb{R}^{m\times n}$$

from [10, 11]. Under some conditions, (H2) is in fact equivalent to (H3) [15, 16].

Given such energy density  $W: \mathbb{R}^{m \times n} \to \mathbb{R}$  and a bounded Lipschitz domain  $\Omega \subset \mathbb{R}^n$ , n = 2, 3, and some right-hand side  $f \in L^{p'}(\Omega; \mathbb{R}^m)$ , define  $\mathcal{J}: V \to \mathbb{R}$  by

$$(1.1) \ \mathcal{J}(v) := \int_{\Omega} W(Dv) \, dx - \int_{\Omega} f \cdot v \, dx \quad \text{for } v \in V := W_0^{1,p}(\Omega; \mathbb{R}^m).$$

Throughout this paper, Dv(x) denotes the  $m \times n$  functional matrix of V at x and we adapt standard notation on Lebesgue and Sobolev spaces, e.g.,  $W_0^{1,p}(\Omega)$  denotes the subset of functions in  $W^{1,p}(\Omega)$  with trace zero on the boundary  $\partial\Omega$  of  $\Omega$ .

The minimization problem reads: Seek minimizers in  $\mathcal{J}$  in V, written

(1.2) 
$$u \in \arg\min_{v \in V} \mathcal{J}(v).$$

The existence of minimizers u or  $u_{\ell}$  of (1.1) in V or some closed subspace  $V_{\ell}$  of V is guaranteed under (H1)-(H2) while, in general, their uniqueness fails. However, the respective exact and discrete stress

$$\sigma := DW(Du)$$
 and  $\sigma_{\ell} = DW(Du_{\ell}) \in L^{r/t}(\Omega; \mathbb{R}^{m \times n})$ 

is unique [11], i.e.,  $\sigma$  and  $\sigma_{\ell}$  do not depend on the choice of u and  $u_{\ell}$  amongst the set of exact and discrete minimizers. The smoothness of  $\sigma \in W^{1,p}_{loc}(\Omega; \mathbb{R}^{m \times n})$  has been analysed in [10, 16], while the smoothness of u is open (recall that u may be non-unique). Therefore the a priori error estimate (valid for any choice of  $u \in \operatorname{argmin} J$ )

$$\|\sigma - \sigma_{\ell}\|_{L^{q}(\Omega; \mathbb{R}^{m \times n})} \lesssim \min_{v_{\ell} \in V_{\ell}} \|u - v_{\ell}\|_{V},$$

although it may be regarded as quasi-optimal convergent, has its limitations. The a posteriori error estimates for  $\|\sigma - \sigma_{\ell}\|_{L^{q}(\Omega; \mathbb{R}^{m \times n})}$  known from the literature even face some reliability-efficiency gap [9], cf. Section 2 and Remark 2.1 below. Surprisingly, this does not prevent the design of convergent adaptive mesh-refining algorithms.

## 2. AFEM

This section describes the adaptive mesh-refining strategy, proposed in this paper and states the main result.

2.1. **Outline.** Given an initial coarse mesh  $\mathcal{T}_0$ , an adaptive finite element method (AFEM) successively generates a sequence of meshes  $\mathcal{T}_1, \mathcal{T}_2, \ldots$  and associated discrete subspaces

$$(2.1) V_0 \stackrel{\subset}{\neq} V_1 \stackrel{\subset}{\neq} \cdots \stackrel{\subset}{\neq} V_{\ell} \stackrel{\subset}{\neq} V_{\ell+1} \stackrel{\subset}{\neq} \cdots \stackrel{\subset}{\neq} V$$

with discrete problems  $(P_0)$ ,  $(P_1)$ ,  $(P_2)$ , ... and discrete solutions  $u_0$ ,  $u_1, u_2, \ldots$  and discrete stresses  $\sigma_0, \sigma_1, \sigma_2, \ldots$  steered by refinement rules and indicators. A typical loop from  $V_{\ell}$  to  $V_{\ell+1}$  (at the frozen level  $\ell$ ) consists of the steps

(2.2) SOLVE 
$$\rightarrow$$
 ESTIMATE  $\rightarrow$  MARK  $\rightarrow$  REFINE explained in the following Subsections.

2.2. **Input.** Input a shape-regular triangulation  $\mathcal{T}_0$  of  $\Omega \subset \mathbb{R}^n$  into closed triangles (if n=2) or closed tetrahedra (if n=3) with associated first-order finite element space  $V_0$ ; suppose each element domain in  $\mathcal{T}_0$  (and furthermore in  $\mathcal{T}_1, \mathcal{T}_2, \ldots$ ) has at least one vertex in the interior of  $\Omega$ , put level  $\ell := 0$ .

A triangulation  $\mathcal{T}_{\ell}$  is regular if two distinct closed-element domains are either disjoint or their intersection is one common vertex, one common

edge (or, if n=3 possibly one common face). For simplicity, all triangulations in the paper will be regular. Those common faces are called sides  $\mathcal{E}_{\ell}$ , if n=3. For n=2,  $\mathcal{E}_{\ell}$  are the interior edges.

2.3. SOLVE. Given the triangulation  $\mathcal{T}_{\ell}$  with set of interior sides  $\mathcal{E}_{\ell}$  and interior nodes  $\mathcal{K}_{\ell}$ , the piecewise affine space  $\mathcal{P}_{1}(\mathcal{T}_{\ell})$  reads

$$\mathcal{P}_1(\mathcal{T}_\ell; \mathbb{R}^m) := \left\{ v \in L^{\infty}(\Omega; \mathbb{R}^m) : \forall T \in \mathcal{T}_\ell, \ v|_T \in \mathcal{P}_1(T; \mathbb{R}^m) \right\};$$

$$\mathcal{P}_1(T; \mathbb{R}^m) := \left\{ v \in C^{\infty}(T; \mathbb{R}^m) : \exists A \in \mathbb{R}^{m \times n} \exists b \in \mathbb{R}^m \right.$$

$$\forall x \in T : v(x) = Ax + b \right\}.$$

The discrete space  $V_{\ell} := V \cap \mathcal{P}_1(\mathcal{T}_{\ell}; \mathbb{R}^m)$  is the first-order finite element space and allows for a nodal basis  $(\varphi_z : z \in \mathcal{K}_{\ell})$ . Then the step SOLVE reads: Solve the nonlinear discrete problem

(2.3) 
$$u_{\ell} \in \arg\min_{v_{\ell} \in V_{\ell}} \mathcal{J}(v_{\ell}) \text{ and set } \sigma_{\ell} := DW(Du_{\ell}).$$

The  $\mathbb{R}^{m\times n}$ -valued stress  $\sigma_{\ell}$  is piecewise constant with respect to  $\mathcal{T}_{\ell}$ .

2.4. ESTIMATE. Given any interior side  $E \in \mathcal{E}_{\ell}$  with measure |E|, and normal unit vector  $\nu_E$ , compute the jump

$$J_E := [\sigma_\ell]_E \, \nu_E \in \mathbb{R}^m$$

of the discrete normal stresses  $\sigma_{\ell}\nu_{E}$  over E, where

$$[\sigma_{\ell}]_{E}(x) := \lim_{T_{+}\ni a\to x} \sigma_{\ell}(a) - \lim_{T_{-}\ni b\to x} \sigma_{\ell}(b)$$

for all  $x \in E = \partial T_+ \cap \partial T_-$ , and by convention,  $\nu_E$  is exterior to  $T_+$ . Then define

(2.4) 
$$\eta_{\ell} := \left(\sum_{E \in \mathcal{E}_{\ell}} \eta_{E}^{p'}\right)^{1/p'}$$
 with  $\eta_{E} := h_{E}^{1/p'} |E|^{1/p'} |J_{E}|$  for  $E \in \mathcal{E}_{\ell}$ .

It is essentially known from [9, 11] that  $\eta_{\ell}$  is a reliable a posteriori error estimator in the sense that

(2.5) 
$$\|\sigma - \sigma_{\ell}\|_{L^{r/t}(\Omega; \mathbb{R}^{m \times n})}^{r} \lesssim \eta_{\ell} + \operatorname{osc}_{\ell},$$

cf. Lemma 4.2 below. Here and throughout,  $\operatorname{osc}_{\ell}$  denotes data oscillations. Given any connected open nonvoid  $\omega \subset \Omega$ , let

(2.6) 
$$\operatorname{osc}(f,\omega)^{p'} := \operatorname{diam}(\omega)^{p'} \|f - f_{\omega}\|_{L^{p'}(\omega)}^{p'} \text{ with } f_{\omega} := |\omega|^{-1} \int_{\omega} f \, dx,$$

the integral mean of f over  $\omega$ . For each node z in the triangulation  $\mathcal{T}_{\ell}$  with nodal basis function  $\varphi_z \in V_{\ell}$ , let  $\omega_z := \{x \in \Omega : \varphi(x) > 0\}$  denote

the patch of z. Then, recall  $\mathcal{K}_{\ell}$  denotes the set of all interior nodes,

(2.7) 
$$\operatorname{osc}_{\ell}^{p'} := \sum_{z \in \mathcal{K}_{\ell}} \operatorname{osc}(f, \omega_z)^{p'}.$$

Since  $\operatorname{osc}_{\ell}$  depends on the given data and explicitly on  $\mathcal{T}_{\ell}$ , it can easily be made arbitrarily small by additional refinement steps. This data oscillation control allows for  $\lim_{\ell\to\infty}\operatorname{osc}_{\ell}=0$ ; cf. [17, 22] for algorithmic details.

Remark 2.1. The upper bound in (2.5) is not sharp, the estimator  $\eta_{\ell}$  is not efficient, because of r > 1. This is called reliability-efficiency gap [9].

2.5. MARK. Select a subset  $\mathcal{M}_{\ell}$  of  $\mathcal{E}_{\ell}$  in the current triangulation  $\mathcal{T}_{\ell}$  with

(2.8) 
$$\eta_{\ell}^{p'} \lesssim \sum_{E \in \mathcal{M}_{\ell}} \eta_{E}^{p'}.$$

Given a parameter  $0 < \Theta < 1$  the selection condition (2.8) results from choosing sufficiently many sides E with bigger  $\eta_E$  in  $\mathcal{M}_{\ell}$  such that the bulk criterion [13, 17, 18, 22] holds:

$$\Theta \, \eta_{\ell}^{p'} \le \sum_{E \in \mathcal{M}_{\ell}} \eta_{E}^{p'}.$$

This is easily arranged with some greedy algorithm.

2.6. REFINE. Refine the triangulation  $\mathcal{T}_{\ell}$  and design a refined shaperegular triangulation  $\mathcal{T}_{\ell+1}$  such that each interior side  $E = \partial T_+ \cap \partial T_- \in \mathcal{M}_{\ell}$  is refined in  $\mathcal{T}_{\ell+1}$ , for  $T_+, T_- \in \mathcal{T}_{\ell}$  and  $T_+ \cup T_-$  includes at least one new node on E and at least one new node in the interior of either  $T_+$ 

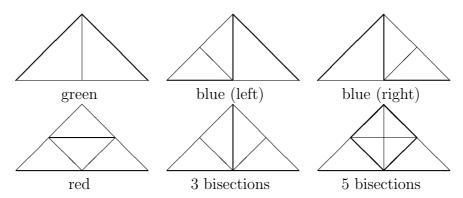


FIGURE 2.1. Possible refinements of a triangle in REFINE of AFEM. The 5 bisections allow for an interior node property.

- or  $T_{-}$ . For n=2 the inner node property is easily depicted with 5 bisections as in Figure 2.1. More details on the shape-regular refinement strategies can be found in [6].
- 2.7. **Output.** The AFEM computes a sequence of discrete stresses  $\sigma_0, \sigma_1, \sigma_2, \ldots$  in  $L^{p'}(\Omega; \mathbb{R}^{m \times n})$  as approximations to  $\sigma := DW(Du)$ . The main result of this paper is the strong convergence of the stresses.

Theorem 2.1 (Convergence Theorem). Suppose (H1)-(H2) and

$$\lim_{\ell \to \infty} \operatorname{osc}_{\ell} = 0.$$

Then the sequence of stress fields  $\sigma_0, \sigma_1, \sigma_2, \ldots$  converges strongly towards the exact stress field  $\sigma$  in  $L^{r/t}(\Omega; \mathbb{R}^{m \times n})$ .

The technical proof is postponed to Section 4, after the motivating list of examples in Section 3.

### 3. Examples and Applications

This section briefly summarizes a few applications with explicit proofs of (H1)-(H2) and hence with a convergent AFEM.

3.1. Uniformly Convex Minimization. Uniformly convex  $C^1$  function  $W: \mathbb{R}^{m \times n} \to \mathbb{R}$  with globally Lipschitz continuous derivative DW, i.e., for all  $A, B \in \mathbb{R}^{m \times n}$  there holds

$$|A - B|^2 \lesssim DW(A) : (A - B) - W(A) + W(B)$$
$$|DW(A) - DW(B)| \lesssim |A - B|.$$

This implies (H1)-(H2) with p=2=r and s=0 and, thus, the class (i) is included in class (ii). Simple examples are  $W(F)=\varphi(|\text{sym }F|)|F|^2$  for proper  $\mathcal{C}^2$  functions  $\varphi$  (cf., e.g., [23, Sections 62.3, 62.8-9] and [15, Exercise 1.7 on page 21]).

- 3.2. **Nonlinear Laplacian.** The *p*-Laplacian satisfies (H1)-(H2) for any  $2 \le p < \infty$  and r = 2, s = p 2.
- **Lemma 3.1.** Given  $1 \leq p < \infty$  define the function  $W : \mathbb{R}^{m \times n} \to \mathbb{R}$  by  $W(A) := |A|^p/p$ . Then there exist a constant  $c_1 = c(p)$  such that for all  $A, B \in \mathbb{R}^{m \times n}$  there holds

$$|DW(A) - DW(B)|^2 \le c_1(|A|^{p-2} + |B|^{p-2}) \times (W(B) - W(A) - DW(A; B - A)).$$

Proof. Given  $A, B \in \mathbb{R}^{m \times n}$  with  $A \neq B$  set a := |A| and b := |B|. A quick check verifies that the assertion holds for either a = 0 or b = 0 with  $c_1 = \max\{p, q\}$ . It is therefore assumed that ab > 0 in the sequel and c := A : B/(ab). Then  $0 < t := b/a < \infty$ . The left- and right-hand side of the assertion vanish for a = b and c = +1. This situation is therefore excluded in the sequel. Then,

$$W(B) - W(A) - DW(A; B - A) = b^{p}/p - a^{p}/p - a^{p-1}(cb - a)$$
$$= b^{p}/p + a^{p}/q - a^{p-1}bc$$

is strictly positive (non-negativity immediately follows from Young's inequality and  $-1 \le c \le 1$ ). Since

$$|DW(A) - DW(B)|^2 = a^{2(p-1)} + b^{2(p-1)} - 2ca^{p-1}b^{p-1}.$$

The quotient of the left- and the right-hand side of the assertion reads

$$\frac{a^{2(p-1)} + b^{2(p-1)} - 2ca^{p-1}b^{p-1}}{(a^{p-2} + b^{p-2})(b^p/p + a^p/q - a^{p-1}bc)} = \frac{1 + t^{2(p-1)} - 2ct^{p-1}}{(1 + t^{p-2})(t^p/p + 1/q - ct)}$$
$$=: f(t, c).$$

A direct calculation verifies that  $\partial f/\partial c$  as a function of c has one sign (which depends on t and p) and hence is monotone increasing or decreasing. Therefore

$$\max_{-1 < c < 1} f(t, c) = \max\{f(t, 1), f(t, -1)\}\$$

and the assertion reads  $f(t,1) \leq c_1$  and  $f(t,-1) \leq c_1$  for all  $0 < t < \infty$ . The case c = +1 is the crucial one because  $t^p/p + 1/q - t$  vanishes for t = 1. Hospital's rule yields f(1,1) = 0. Since f(0,1) = q and  $\lim_{t\to\infty} f(t,1) = p$ , one deduces from continuity of f(t,1) in t that

$$\sup_{0 < t < \infty} f(t, 1) =: c_1 < \infty.$$

The analysis for c = -1 is simpler and hence omitted.

3.3. Optimal Design Problem. Let  $0 < t_1 < t_2$  and  $0 < \mu_2 < \mu_1$  be positive real numbers with  $t_1\mu_1 = t_2\mu_2$  and consider a convex  $C^1$  function  $\psi : [0, \infty) \to \mathbb{R}$  with  $\psi(0) = 0$  and

$$\psi'(t) := \begin{cases} \mu_1 t & \text{for } 0 \le t \le t_1, \\ t_1 \mu_1 = t_2 \mu_2 & \text{for } t_1 \le t \le t_2, \\ \mu_2 t & \text{for } t_2 \le t. \end{cases}$$

The energy density  $W(A) := \psi(|A|)$ ,  $A \in \mathbb{R}^n$ , results from a relaxation process [14]. It satisfies (H1)-(H2) with p = r = 2 and s = 0. Details can be found in [2].

3.4. **Scalar 2-Well Problem.** The scalar convexified 2-well energy density W results from a relaxation in nonconvex minimization problems allowing for microstructures [11]. It satisfies (H1)-(H2) with p=4 and r=2=s.

**Proposition 3.2.** Given distinct  $F_1$  and  $F_2$  in  $\mathbb{R}^n$  set  $A := (F_2 - F_1)/2 \neq 0$  and  $B := (F_1 + F_2)/2$  where  $(\cdot)_+ := \max\{0, \cdot\}$  and  $(\cdot)_+^2 := \max\{0, \cdot\}^2$ . For any  $F \in \mathbb{R}^n$  let

$$W(F) := (|F - B|^2 - |A|^2)_{\perp}^2 + 4(|A|^2|F - B|^2 - (A \cdot (F - B))^2).$$

Then for any  $F, G \in \mathbb{R}^n$  with  $\xi := (|F - B|^2 - |A|^2)_+$  and  $\eta := (|G - B|^2 - |A|^2)_+$  there holds

$$|DW(G) - DW(F)|^{2}$$

$$\leq 32(|A|^{2} + \xi + \eta)(W(G) - W(F) - DW(F) \cdot (G - F)).$$

The proof of Proposition 3.2 is based on two lemmas.

**Lemma 3.3.** Given  $A, B \in \mathbb{R}^n$  let  $W(F) := (|F - B|^2 - |A|^2)_+^2$ . For any F and G in  $\mathbb{R}^n$  let

$$\xi := (|F - B|^2 - |A|^2)_+$$
 and  $\eta := (|G - B|^2 - |A|^2)_+$ 

Then there holds

$$|DW(F) - DW(G)|^2$$
  
 $\leq 32(|A|^2 + \xi + \eta)(W(G) - W(F) - DW(F) \cdot (G - F)).$ 

Proof. Let U := F - B, V := G - B, a := |A| and notice that  $DW(F) = 4\xi U$  and  $DW(G) = 4\eta V$ . In the first case suppose that both,  $\xi = |U|^2 - a^2$  and  $\eta = |V|^2 - a^2$ , are positive. Utilizing

$$DW(F) - DW(G) = 4(\xi U - \eta V) = 4\xi(U - V) + 4(\xi - \eta)V$$

one obtains

$$1/32 |DW(F) - DW(G)|^2 \le \xi^2 |U - V|^2 + (\xi - \eta)^2 |V|^2.$$

Since  $|V|^2 = \eta + a^2$  this proves

$$(3.1) \ 1/32 |DW(F) - DW(G)|^2 \le (a^2 + \xi + \eta)(\xi |U - V|^2 + (\xi - \eta)^2).$$

On the other hand, the preceeding situation allows the direct calculation of

$$W(G) - W(F) - DW(F) \cdot (F - G)$$

$$= \eta^2 - \xi^2 + 4\xi U \cdot (U - V)$$

$$= \eta^2 - \xi^2 + 2\xi (|U|^2 - |V|^2) + 2\xi |U - V|^2$$

$$= 2\xi |U - V|^2 + (\xi - \eta)^2.$$

The combination with (3.1) shows the assertion in the present first case of positive  $\xi$  and  $\eta$ . For  $\xi = 0 < \eta = |V|^2 - a^2$  the assertion reads

$$16\eta^2 |V|^2 \le 32(a^2 + \eta) \,\eta^2$$

which follows immediately from  $|V|^2 \le (a^2 + \eta)$ . In the remaining case  $\eta = a < \xi = |U|^2 - a^2$ , whence  $|V| \le a < |U|$ , the assertion reads

$$16\xi^2|U|^2 \le 32(a^2 + \xi)(4\xi U \cdot (U - V) - \xi^2).$$

This is equivalent to

$$\xi^2 |U|^2 \le 2(a^2 + \xi)(\xi^2 + 2\xi(a^2 - |V|^2) + 2\xi|U - V|^2)$$

and hence follows from  $|U|^2 = a^2 + \xi$  and  $0 \le a^2 - |V|^2$ .

**Lemma 3.4.** Let S be a symmetric and positive semidefinite real  $n \times n$  matrix with spectral radius  $\varrho(S)$  and pseudo inverse  $S^+$  and induced seminorm  $|\cdot|_{S^+}$ , i.e.,

$$|F|_{S^+} := (F \cdot S^+ F)^{1/2}$$
 for all  $F \in \mathbb{R}^n$ .

Then the function  $W: \mathbb{R}^n \to \mathbb{R}$  defined by

$$W(F) := 1/2 F \cdot SF \quad for \ F \in \mathbb{R}^n$$

satisfies

$$\begin{split} \varrho(S)^{-1}|DW(F) - DW(G)|^2 &\leq |DW(F) - DW(G)|_{S^+}^2 \\ &= (F-G) \cdot S(F-G) \\ &= 2(W(G) - W(F) - (SF) \cdot (G-F)). \end{split}$$

*Proof.* Since S is symmetric,  $S = SS^+S$ , and so DW(F) = SF satisfies

$$|S(F-G)|^2 \le \varrho(S)|S^{1/2}(F-G)|^2 = \varrho(S)|S(F-G)|_{S^+}^2$$

The remaining identity results from

$$1/2(F-G) \cdot S(F-G) = W(G) - W(F) + F \cdot S(F-G).$$

Proof of Proposition 3.2. Notice that W(F) is the sum of the two energy densities of the aforegoing lemmas. Indeed, let  $A^0 := A/|A|$  and define the symmetric and positive semidefinite matrix  $S := 1 - A^0 \otimes A^0$ . Then

$$4(|A|^2|F-B|^2 - (A \cdot (F-B))^2) = 4|A|^2|F-B|_S^2.$$

Observe the upper bound of S

$$|DW(G) - DW(F)|^2 \le 32|\xi U - \eta V|^2 + 32|A|^4|U - V|_S^2$$

is estimated in Lemma 3.3 and Lemma 3.4, respectively. This concludes the proof.  $\hfill\Box$ 

3.5. **Vectorial 2-Well Problem.** Given two distinct wells  $E_1$  and  $E_2$  in  $\mathbb{R}^{n \times n}_{\text{sym}}$  with minimal energies  $W_1^0$  and  $W_2^0$  in  $\mathbb{R}$ , we consider the quadratic elastic energies

$$W_j(E) := 1/2(E - E_j) : \mathbb{C}(E - E_j) + W_j^0$$
 for all  $E \in \mathbb{R}_{\text{sym}}^{n \times n}$ .

Energy minimization leads to an optimal choice of the configuration of the two phases, and so the strain energy density  $\tilde{W}$  is modelled by the minimum

$$\tilde{W}(E) = \min\{W_1(E), W_2(E)\}$$
 for all  $E \in \mathbb{R}_{\text{sym}}^{n \times n}$ .

The two wells (transformation strains) are said to be *compatible* if

(3.2) 
$$E_1 = E_2 + \frac{1}{2}(a \otimes b + b \otimes a) \text{ for some } a, b \in \mathbb{R}^n.$$

Then the constant  $\gamma = 1/2|E_2 - E_1|_{\mathbb{C}}^2$  and the quasiconvexification W of  $\tilde{W} = \{W_1, W_2\}$  [14] is given by

$$W(E) = \begin{cases} W_2(E) & \text{if } W_2(E) + \gamma \leq W_1(E), \\ \frac{1}{2}(W_2(E) + W_1(E)) - \frac{1}{4\gamma}(W_2(E) - W_1(E))^2 - \frac{\gamma}{4} \\ & \text{if } |W_2(E) - W_1(E)| \leq \gamma, \\ W_1(E) & \text{if } W_1(E) + \gamma \leq W_2(E). \end{cases}$$

The convex W satisfies (H1)-(H2) with p = 2 = r and s = 0.

**Proposition 3.5.** In the compatible case (3.2) there holds, for all  $A, B \in \mathbb{R}^{n \times n}_{svm}$ ,

$$||1/2||DW(A) - DW(B)||_{\mathbb{C}^{-1}}^2 \le W(B) - W(A) - DW(A) : (B - A).$$

*Proof.* A translation of the argument in W allows us to assume, without loss of generality, that  $E_1 + E_2 = 0$ . For  $E \in \mathbb{R}^{n \times n}_{sym}$ , let

$$\varphi(E) := \gamma^{-1}(W_2(E) - W_1(E)), 
\psi(E) := \max\{-1, \min\{1, \varphi(E)\}\}.$$

As in [12] one deduces, for  $E \in \mathbb{R}^{n \times n}_{\text{sym}}$  and  $\gamma \varphi(E) = 2(\mathbb{C}E_1) : E + W_2^0 - W_1^0$ ,

$$DW(E) = \mathbb{C}E - \psi(E)\mathbb{C}E_1$$

and observes that  $\psi(E) = \varphi(E)$  for  $E \in \mathbb{R}_{\text{sym}}^{n \times n}$  with  $-1 \le \varphi(E) \le 1$ . The proof of the proposition starts with the discussion of

(3.3) 
$$\gamma/2(\psi(B) - \psi(A))(\psi(A) - \varphi(A)) \ge 0.$$

In fact,  $\psi(A) \neq \varphi(A)$  implies either  $\psi(A) = 1 < \varphi(A)$  [notice  $\psi(B) - 1 \leq 0$ ] or  $\psi(A) = -1 > \varphi(A)$  [notice  $\psi(B) + 1 \geq 0$ ] and in each case (3.3) follows. Algebraic manipulations will show in the sequel that (3.3) is equivalent to the assertion. Abbreviate  $\sigma := DW(A)$  and  $\tau := DW(B)$  to compute the left-hand side of the assertion, namely

$$1/2 |\sigma - \tau|_{\mathbb{C}^{-1}}^2 = 1/2 (\tau - \sigma) : \mathbb{C}^{-1} (\tau + \sigma) + (\sigma - \tau) : \mathbb{C}^{-1} \sigma.$$

With  $\mathbb{C}^{-1}(\sigma - \tau) = A - B - \psi(A)E_1 + \psi(B)E_1$ , this reads

$$\sigma: (A - B) - \frac{1}{2} |\sigma - \tau|_{\mathbb{C}^{-1}}^{2}$$

$$= (\psi(A) - \psi(B)) E_{1} : \sigma - \frac{1}{2} |\tau|_{C^{-1}}^{2} + \frac{1}{2} |\sigma|_{\mathbb{C}^{-1}}^{2}.$$

The definition of  $\sigma$  and  $\tau$  and  $\gamma/2 = |E_1|_{\mathbb{C}}^2$  show

$$\frac{1}{2} |\sigma|_{\mathbb{C}^{-1}}^{2} - \frac{1}{2} |\tau|_{\mathbb{C}^{-1}}^{2} = \frac{1}{2} |A|_{\mathbb{C}}^{2} - \frac{1}{2} |B|_{\mathbb{C}}^{2} + \frac{\gamma}{4} (\psi(A)^{2} - \psi(B)^{2}) 
- \psi(A)A : \mathbb{C}E_{1} + \psi(B)B : \mathbb{C}E_{1}.$$

It is a lengthy but direct verification that W(E),  $E \in \mathbb{R}_{\text{sym}}^{n \times n}$ , can be written as

$$W(E) = \frac{1}{2}E : \mathbb{C}E + \frac{1}{2}(W_1^0 + W_2^0) + \frac{\gamma}{4}\psi(E)(\psi(E) - 2\varphi(E)).$$

The combination of the preceding three identities [the last applied to E = A and E = B] shows

$$W(B) - W(A) + \sigma : (A - B) - \frac{1}{2} |\sigma - \tau|_{\mathbb{C}^{-1}}^{2}$$

$$= (\psi(A) - \psi(B))(E_{1} : \mathbb{C}A - \psi(A)\gamma/2)$$

$$- \psi(A)A : \mathbb{C}E_{1} + \psi(B)B : \mathbb{C}E_{1}$$

$$+ \gamma/2 \varphi(A)\psi(A) - \gamma/2 \varphi(B)\psi(B)$$

$$= -\gamma/2 \psi(A)^{2} + \gamma/2 \psi(A)\psi(B) - \psi(B)E_{1} : \mathbb{C}(A - B)$$

$$+ \gamma/2 \varphi(A)\psi(A) - \gamma/2 \varphi(B)\psi(B).$$

Since  $E_1: \mathbb{C}(A-B) = \gamma/2(\varphi(A)-\varphi(B))$  shows that the preceding expression equals the left-hand side of (3.3).

Remark 3.1. The immediate corollary (H3) of Proposition 3.5 is known from [10, 12] and fundamental for error analysis and regularity.

3.6. Hencky elastoplasticity with hardening. One time step within an elastoplastic evolution problem leads to Hencky's model. For various hardening laws and von-Mises yield conditions, an elimination of internal variables [1] leads to the energy function

(3.4) 
$$W(E) := \frac{1}{2}E : \mathbb{C}E - \frac{1}{4\mu} \max\{0, |\det \mathbb{C}E| - \sigma_y\}^2/(1+\eta)$$

for  $E \in \mathbb{R}^{n \times n}_{\text{sym}}$ . Here we adopt notation of the previous section and  $\mathbb{C}$  is the fourth-order elasticity tensor,  $\sigma_y > 0$  is the yield stress, and  $\eta > 0$  is the modulus of hardening. The model of perfect plasticity corresponds to  $\eta = 0$  [21]. For  $\eta > 0$  there holds (H1)-(H2) for p = 2 = r and s = 0.

**Proposition 3.6.** For all  $A, B \in \mathbb{R}_{\text{sym}}^{n \times n}$  there holds

$$|1/2|DW(A) - DW(B)|_{\mathbb{C}^{-1}}^2 \le W(B) - W(A) - DW(A) : (B - A).$$

Proof. Set  $\psi(x) := 1 - \max\{0, 1 - \sigma_y/(2\mu x)\}/(1 + \eta)$  to define the continuous and monotone decreasing function  $\psi : [0, \infty) \to (\eta/(1 + \eta), 1]$  which satisfies

$$DW(E) = (\lambda + 2\mu/n)\operatorname{tr}(E) \mathbf{1} + 2\mu\psi(|\operatorname{dev} E|)\operatorname{dev} E \quad \text{for all } E \in \mathbb{R}^{n \times n}_{\operatorname{sym}}$$

Given  $A, B \in \mathbb{R}_{\text{sym}}^{n \times n}$ , the following abbreviations will be used throughout the remaining part of the proof:

$$\sigma := DW(A),$$
  $a := |\operatorname{dev} A|,$   $\alpha := \psi(a),$   $\tau := DW(B),$   $b := |\operatorname{dev} B|,$   $\beta := \psi(b).$ 

Then the assertion reads

$$\delta := W(B) - W(A) + \sigma : (A - B) - 1/2 |\sigma - \tau|_{\mathbb{C}^{-1}}^2 \ge 0.$$

In the first three steps one computes  $\delta$ . The aforementioned formulae for DW(A) and DW(B) and elementary calculations with the third formula of Binomi yield in step one that

$$\sigma: \mathbb{C}^{-1}(\sigma - \tau) - 1/2|\sigma - \tau|_{\mathbb{C}^{-1}}^{2}$$

$$= 1/2 |\sigma|_{\mathbb{C}^{-1}}^{2} - 1/2 |\tau|_{\mathbb{C}^{-1}}^{2}$$

$$= (\lambda/2 + \mu/n)(\operatorname{tr}(A)^{2} - \operatorname{tr}(B)^{2}) + \mu(\alpha^{2}a^{2} - \beta^{2}b^{2}).$$

Step two employs the definition of  $\psi$  to rewrite the energy as

$$W(E) = 1/2 |E|_{\mathbb{C}}^{2} - (1+\eta)\mu (1 - \psi(|\operatorname{dev} E|))^{2} |\operatorname{dev} E|^{2},$$

for all  $E \in \mathbb{R}_{\text{sym}}^{n \times n}$ . Step three employs the above formulae for  $\sigma$  and  $\tau$  to estimate

$$\sigma: (A-B) - \sigma: \mathbb{C}^{-1}(\sigma - \tau) = 2\mu \,\alpha \,\mathrm{dev}\, A: ((1-\alpha)\,\mathrm{dev}\, A - (1-\beta)\,\mathrm{dev}\, B).$$

The Cauchy inequality, leads to

$$\sigma: (A-B) - \sigma: \mathbb{C}^{-1}(\sigma-\tau) \ge 2\mu \,\alpha(1-\alpha)a^2 - 2\mu \,\alpha(1-\beta)ab.$$

The left-hand sides considered in the first three steps add up to  $\delta$  and so lead to a lower bound of  $\delta$ . Elementary manipulations with this

lower bound in step four of the proof yield the estimate

$$\begin{split} \delta/\mu & \geq \alpha^2 a^2 - \beta^2 b^2 + b^2 - a^2 + (1+\eta)(1-\alpha)^2 a^2 - (1+\eta)(1-\beta)^2 b^2 \\ & + 2\alpha(1-\alpha)a^2 - 2\alpha(1-\beta)ab \\ & = \eta(1-\alpha)^2 a^2 - \eta(1-\beta)^2 b^2 + 2(1-\beta)b(\beta b - \alpha a) \\ & = \eta \Big( (1-\alpha)a - (1-\beta)b \Big)^2 \\ & + 2(1-\beta)b \Big( (1+\eta)(\beta b - \alpha a) - \eta(b-a) \Big). \end{split}$$

Step five concerns the function  $g(x) := x\psi(x)$  which satisfies g'(x) = 1 and  $g'(x) = \eta/(1+\eta)$  for  $2\mu x < \sigma_y$  and  $\sigma_y < 2\mu x$ , respectively. For  $a \le b$ , this and the fundamental theorem of calculus show

(3.5) 
$$\eta(b-a) \le (1+\eta) \int_a^b g'(x) dx = (1+\eta)(\beta b - \alpha a).$$

This concludes the proof of  $\delta \geq 0$  in this case. In the case b < a, the above lower bound of  $\delta$  shows  $\delta \geq 0$  if  $\beta = 1$ . Hence it remains to consider b < a and  $\beta < 1$  which implies  $\sigma_y < 2\mu b$  and so  $g'(x) = \eta/(1+\eta)$  for all b < x < a. This yields equality in (3.5) and so proves  $\delta \geq 0$ .

Remark 3.2. Although (H2) holds for  $\eta = 0$  as well, the linear growth condition yields a different functional analytical setting in  $BD(\Omega)$  [21].

## 4. Proof of Convergence

This section provides a proof of Theorem 2.1 on the convergence of the stress fields in  $L^{r/t}(\Omega; \mathbb{R}^{m \times n})$ . Throughout this section, the focus is on the energy difference

$$\delta_{\ell} := \mathcal{J}(u_{\ell}) - \mathcal{J}(u) \ge 0.$$

Due to (2.1), the sequence  $(\delta_{\ell})_{\ell}$  is monotone decreasing, and hence convergent to some limit  $\delta \geq 0$ . It is essential to prove  $\delta = 0$ , which is not known in the beginning of the proof.

## Lemma 4.1. There holds

$$\|\sigma_{\ell+1} - \sigma_{\ell}\|_{L^{r/t}(\Omega;\mathbb{R}^{m \times n})}^r \lesssim \delta_{\ell} - \delta_{\ell+1}.$$

*Proof.* The two-sided growth conditions in (H1) lead in [11] to the boundedness of discrete minimizers in  $W^{1,p}$  and show

(4.1) 
$$\int_{\Omega} (1 + |Du_{\ell}|^{s} + |Du_{\ell+1}|^{s})^{p/s} dx \lesssim 1.$$

Since  $\sigma_{\ell+1}$  satisfies the discrete Euler-Lagrange equations, there holds

$$\int_{\Omega} \sigma_{\ell+1} : D(u_{\ell} - u_{\ell+1}) \, dx = \int_{\Omega} f \cdot (u_{\ell} - u_{\ell+1}) \, dx.$$

Therefore,

$$\delta_{\ell} - \delta_{\ell+1} = \int_{\Omega} \left( W(Du_{\ell}) - W(Du_{\ell+1}) - f \cdot (u_{\ell} - u_{\ell+1}) \right) dx$$
$$= \int_{\Omega} \left( W(Du_{\ell}) - W(Du_{\ell+1}) - \sigma_{\ell+1} : D(u_{\ell} - u_{\ell+1}) \right) dx.$$

An application of (H2) with  $A = Du_{\ell+1}(x)$  and  $B = Du_{\ell}(x)$  leads to an estimate for all x in  $\Omega$ . The integral of those inequalities reads

$$\int_{\Omega} (1 + |Du_{\ell}|^{s} + |Du_{\ell+1}|^{s})^{-1} |\sigma_{\ell} - \sigma_{\ell+1}|^{r} dx$$

$$\lesssim \int_{\Omega} (W(Du_{\ell}) - W(Du_{\ell+1}) - \sigma_{\ell+1} : D(u_{\ell} - u_{\ell+1})) dx$$

$$= \delta_{\ell} - \delta_{\ell+1}.$$

The Hölder inequality with t and t' = 1 + p/s, 1/t + 1/t' = 1, plus (4.1) with t'/t = p/s lead to

$$\|\sigma_{\ell+1} - \sigma_{\ell}\|_{L^{r/t}(\Omega;\mathbb{R}^{m \times n})}^{r/t} = \int_{\Omega} (1 + |Du_{\ell}|^{s} + |Du_{\ell+1}|^{s})^{-1/t} |\sigma_{\ell} - \sigma_{\ell+1}|^{r/t} \times (1 + |Du_{\ell}|^{s} + |Du_{\ell+1}|^{s})^{1/t} dx$$
$$\lesssim \left( \int_{\Omega} (1 + |Du_{\ell}|^{s} + |Du_{\ell+1}|^{s})^{-1} |\sigma_{\ell} - \sigma_{\ell+1}|^{r} dx \right)^{1/t}.$$

The combination of this estimate with (4.2) proves the lemma.

Lemma 4.2. There holds (2.5), namely

$$\|\sigma - \sigma_{\ell}\|_{L^{r/t}(\Omega; \mathbb{R}^{m \times n})}^{r} \lesssim \eta_{\ell} + \operatorname{osc}_{\ell}.$$

Proof. In slightly different notation, it is proven in [11] that

It is known since [19, 20] that the volume contribution  $||h_{\mathcal{I}_{\ell}}f||_{L^{p'}(\Omega)}$  can be controlled by  $\eta_{\ell} + \operatorname{osc}_{\ell}$  and so (4.3) leads to the assertion; cf. [9] for one particular case. The main arguments are recalled here for convenient reading. A triangle inequality yields, for each free node z, that

$$(4.4) ||f||_{L^{p'}(\omega_z)} \le ||f - f_{\omega_z}||_{L^{p'}(\omega_z)} + |f_{\omega_z}||\omega_z|^{1/p'}.$$

The integral mean equals

(4.5) 
$$f_{\omega_z}|\omega_z| \approx \int_{\Omega} \varphi_z f_{\omega_z} dx = \int_{\Omega} \varphi_z (f - f_{\omega_z}) dx + \int_{\Omega} \varphi_z f dx.$$

The combination of (4.4)-(4.5) plus a Hölder inequality shows

$$(4.6) ||f||_{L^{p'}(\omega_z)} \lesssim ||f - f_{\omega_z}||_{L^{p'}(\omega_z)} + |\omega_z|^{-1/p} \left| \int_{\Omega} \varphi_z f \, dx \right|.$$

On the other hand, the discrete Euler-Lagrange equations show for the j-th component  $f_j$  of f and the components  $\sigma_{\ell,j} := (\sigma_{\ell,j_1}, \ldots, \sigma_{\ell,j_n})$  of  $\sigma_{\ell}$ , that

(4.7) 
$$\int_{\Omega} \varphi_z f_j \, dx = \int_{\Omega} \sigma_{\ell,j} \cdot \nabla \varphi_z \, dx = \sum_{E \in \mathcal{E}} \int_{E} \left( [\sigma_{\ell,j}] \cdot \nu_E \right) \varphi_z \, ds$$

with an elementwise integration by parts. Let  $\mathcal{E}(z) := \{E \in \mathcal{E} : z \in E\}$  denote the set of sides which contribute in (4.7). Then for all  $j = 1, 2, \ldots, m$  components in (4.7) it follows that

$$(4.8) \qquad \left| \int_{\Omega_z} f \varphi_z \, dx \right| \leq \left( \sum_{E \in \mathcal{E}(z)} \eta_E^{p'} \right)^{1/p'} \left( \sum_{E \in \mathcal{E}(z)} h_E^{-p/p'} \| \varphi_z \|_{L^p(E)}^p \right)^{1/p}.$$

Since the last factor in (4.8) is proportional to  $h_z^{n/p}$  <sup>-1</sup> for  $h_z = \text{diam}(\omega_z)$ , (4.7)-(4.8) yield

(4.9) 
$$|\omega_z|^{-p'/p} \left| \int_{\Omega} f \varphi_z \, dx \right|^{p'} \lesssim h_z^{-p'} \sum_{E \in \mathcal{E}(z)} \eta_E^{p'}.$$

Since  $\mathcal{E}(z)$ , for free nodes  $z \in \mathcal{K}$ , have a finite overlap, the combination of (4.6) and (4.9) shows

$$\|h_{\mathcal{I}_{\ell}}f\|_{L^{p'}(\Omega)}^{p'} \approx \sum_{z \in \mathcal{K}} h_z^{p'} \|f\|_{L^{p'}(\omega_z)}^{p'} \lesssim \operatorname{osc}_{\ell}(f)^{p'} + \eta_{\ell}.$$

This and (4.3) proof the assertion.

Remark 4.1. The condition that each element has at least one vertex, which is a free node, leads to  $\Omega = \bigcup_{z \in \mathcal{K}} \omega_z$  in the proof of Lemma 4.2. This can be generalised by enlarging  $\omega_z$  to  $\Omega_z$  by some elements near the boundary. We refer to [5, 4, 7, 8] for details.

**Lemma 4.3.** For any  $E \in \mathcal{M}_{\ell}$  with  $E = \partial T_{+} \cup \partial T_{-}$  for  $T_{+}, T_{-} \in \mathcal{T}_{\ell}$  and  $\omega_{E} = \operatorname{int}(T_{+} \cup T_{-})$  there holds

$$\eta_E \lesssim \|\sigma_{\ell+1} - \sigma_{\ell}\|_{L^{p'}(\omega_E; \mathbb{R}^{m \times n})} + \|f - f_{\omega_E}\|_{L^{p'}(\omega_E; \mathbb{R}^m)}.$$

*Proof.* REFINE allows for nodal basis functions  $\varphi_E$  of a new node  $\operatorname{mid}(E)$  in E and  $\psi_E$  of a new node  $\operatorname{mid}(\omega_E)$  in either  $T_+$  or  $T_-$ , with respect to the finer triangulation  $\mathcal{T}_{\ell+1}$  and  $E, T_+, T_-$  from  $\mathcal{T}_{\ell}$ . Then, there exists some linear combination

$$V_E := \alpha \varphi_E + \beta \psi_E \in V_{\ell+1} \cap W_0^{1,p}(\omega_E; \mathbb{R}^m)$$

with the following conditions

$$\int_E v_E \, ds = |E|, \ \int_{\omega_E} v_E \, dx = 0, \ \|v_E\|_V \approx h_E^{-1} \, |\omega_E|^{1/p}.$$

The construction of such  $V_E$  is the same as in linear problems [3, 13, 17, 18, 22] and hence the remaining details are neglected and the subsequent outline is kept brief. Since  $J_E$  is constant along E

$$|E|J_E = \int_E ([\sigma_\ell]\nu_E) \cdot v_E \, ds = \int_{\omega_E} \sigma_\ell : Dv_E \, dx.$$

Since  $v_E \in V_{\ell+1}$  and  $\sigma_{\ell+1}$  satisfy the discrete Euler-Lagrange equations,

$$\int_{\omega_E} \sigma_\ell : Dv_E \, dx = \int_{\omega_E} (\sigma_\ell - \sigma_{\ell+1}) : Dv_E \, dx + \int_{\omega_E} (f - f_{\omega_E}) \cdot v_E \, dx$$

with the constant integral mean  $f_{\omega_E}$  of f over  $\omega_E$ . The combination of the above identity with Friedrichs inequality  $||v_E||_{L^p(\omega_E;\mathbb{R}^m)} \lesssim h_E||v_E||_V$  proves

$$\eta_E = h_E^{1/p'} |E|^{1/p'} |J_E| \lesssim h_E^{1/p'} |E|^{1/p} (\|\sigma_\ell - \sigma_{\ell+1}\|_{L^{p'}(\omega_E; \mathbb{R}^{m \times n})} + h_{\omega_E} \|f - f_{\omega_E}\|_{L^{p'}(\omega_E; \mathbb{R}^m)}) \|v_E\|_V. \square$$

Proof of Theorem 2.1. Notice that the patches have a finite overlap and

$$\sum_{E \in \mathcal{E}_{\ell}} h_E^{p'} \| f - f_{\omega_E} \|_{L^{p'}(\omega_E; \mathbb{R}^m)} \lesssim \operatorname{osc}_{\ell}^{p'}.$$

Hence Lemma 4.3 leads to

$$\sum_{E \in \mathcal{M}} \eta_E^{p'} \lesssim \|\sigma_{\ell+1} - \sigma_{\ell}\|_{L^{p'}(\Omega; \mathbb{R}^{m \times n})}^{p'} + \operatorname{osc}_{\ell}^{p'}.$$

This, (2.8) in MARK and Lemma 4.2 show

Since  $(\delta_{\ell}) \to \delta$ , the right-hand side in Lemma 4.1 converges to zero, i.e.,

$$\lim_{\ell \to \infty} \|\sigma_{\ell+1} - \sigma_{\ell}\|_{L^{r/t}(\Omega; \mathbb{R}^{m \times n})} = 0.$$

Since  $p' \leq r/t$  and  $|\Omega| \lesssim 1$ , the right-hand side in (4.10) tends to zero as  $\ell \to \infty$ . This proves the claimed strong convergence

$$\lim_{\ell \to \infty} \|\sigma - \sigma_{\ell}\|_{L^{r/t}(\Omega; \mathbb{R}^{m \times n})} = 0. \quad \Box$$

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