# Hamburger Beiträge zur Angewandten Mathematik

## Identification of matrix parameters in elliptic PDEs

Klaus Deckelnick and Michael Hinze

Nr. 2011-05 February 2011

### Identification of matrix parameters in elliptic PDEs

Klaus Deckelnick\*& Michael Hinze<sup>†</sup>

**Abstract:** In the present work we treat the inverse problem of identifying the matrix-valued diffusion coefficient of an elliptic PDE from measurements with the help of techniques from PDE constrained optimization. We prove existence of solutions using the concept of H–convergence and employ variational discretization for the discrete approximation of solutions. Using a discrete version of H–convergence we are able to establish the strong convergence of the discrete solutions. Finally we present some numerical results.

#### Mathematics Subject Classification (2000): 49J20, 49K20, 35B37

**Keywords:** Parameter identification, elliptic optimal control problem, control constraints, H– convergence, variational discretization.

#### 1 Introduction

In this work we consider the inverse problem of identifying the diffusion matrix A = A(x) in an elliptic PDE

$$-\operatorname{div} (A(x)\nabla y) = g \text{ in } \Omega, \ y = 0 \text{ on } \partial\Omega$$

$$(1.1)$$

from measurements of data. Here,  $\Omega \subset \mathbb{R}^n$  is a bounded open set with a Lipschitz boundary. Furthermore, we assume that  $A(x) = (a_{ij}(x))_{i,j=1}^n$  satisfies  $a_{ij} \in L^{\infty}(\Omega)$  and that there exists a > 0 such that  $\sum_{i,j=1}^n a_{ij}(x)\xi_i\xi_j \ge a|\xi|^2$  for all  $\xi \in \mathbb{R}^n$  and a.a.  $x \in \Omega$ . Given  $g \in H^{-1}(\Omega)$ , the boundary value problem (1.1) then has a unique weak solution  $y \in H_0^1(\Omega)$  in the sense that

$$\int_{\Omega} A\nabla y \cdot \nabla v dx = \langle g, v \rangle \quad \text{for all } v \in H_0^1(\Omega), \tag{1.2}$$

where  $\langle \cdot, \cdot \rangle$  denotes the duality pairing between  $H^{-1}(\Omega)$  and  $H^{1}_{0}(\Omega)$ . Furthermore,

$$\|y\|_{H^1_0} \le C \|g\|_{H^{-1}},\tag{1.3}$$

with a constant C which only depends on a. We shall denote this solution by y = T(A, g) in order to also emphasize its dependence on A.

<sup>\*</sup>Institut für Analysis und Numerik, Otto–von–Guericke–Universität Magdeburg, Universitätsplatz 2, 39106 Magdeburg, Germany

<sup>&</sup>lt;sup>†</sup>Schwerpunkt Optimierung und Approximation, Universität Hamburg, Bundesstraße 55, 20146 Hamburg, Germany.

In what follows we assume that measurements  $(z^{(i)}, f^{(i)}) \in Z \times H^{-1}(\Omega), 1 \leq i \leq N$  $(Z = L^2(\Omega) \text{ or } Z = H_0^1(\Omega))$  are available, from which we would like to reconstruct the diffusion matrix A. To do so, we employ a least squares approach together with a Tikhonov regularization, i.e. we consider

(P) 
$$\min_{A \in \mathcal{M}} J(A) := \frac{1}{2} \sum_{i=1}^{N} \|y^{(i)} - z^{(i)}\|_{Z}^{2} + \frac{\gamma}{2} \|A\|^{2} \text{ s.t. } y^{(i)} = T(A, f^{(i)}), 1 \le i \le N.$$
(1.4)

Here,  $\gamma > 0$  and we use the symbol  $\|\cdot\|$  for the  $L^2$  norm on spaces of scalar, vector or matrixvalued functions, while the admissible set  $\mathcal{M}$  will be specified in Section 2. Our choice, motivated by the concept of H-convergence, guarantees the existence of a minimum of J. By discretizing (1.1) with the help of linear finite elements we obtain an approximation  $J_h$  of J. Our main result, Theorem 3.4, says that each sequence of minimizers  $(A_h)_{h>0}$  of  $J_h$  has a subsequence that converges strongly in  $L^2$  to a minimum of J. In order to establish this result we shall adapt a discrete version of H-convergence, introduced by Eymard and Gallouët in [3] for finite volume schemes, to our setting. The above convergence result justifies the use of  $J_h$  in solving the identification problem. In practice we employ a projected steepest descent algorithm for minimizing  $J_h$ , see Section 4.

Let us review related work which is concerned with the identification of matrix-valued parameters in elliptic PDEs. In [1], Alt, Hoffmann and Sprekels obtain a reconstructed matrix by investigating the long time behaviour of a suitable dynamical system, see also [6]. In [10], Kohn and Lowe introduce a variational method that is based on a convex functional involving the variables y and  $A\nabla y$  and investigate its stability properties. Stability results for the reconstruction of matrices of the form  $A = \nabla p \otimes \nabla y$  can be found in [8]. In [13], Rannacher and Vexler prove a-priori estimates for a matrix-identification problem in which a finite number of unknown parameters is estimated from finitely many pointwise observations.

A lot of work has been devoted to the parameter estimation problem for a scalar diffusion coefficient. Identifiability results can e.g. be found in [2], [14] and [16]. A survey of numerical methods for parameter estimation problems can be found in [11]. Error estimates for a least squares approach have been obtained by Falk in [4] and more recently by Wang and Zou [17] for a functional involving a Tikhonov regularization. That paper also contains a long list of further contributions. Let us finally note that the concept of H–convergence has recently been used by Leugering and Stingl in [12] in order to treat problems in material design, in particular to identify strain tensors from displacements in linear elasticity.

#### 2 Existence of a minimum

Let us denote by  $S_n$  the set of all symmetric  $n \times n$  matrices endowed with the inner product  $A \cdot B = \text{trace}(AB)$ . We consider the subset

$$K := \{ A \in \mathcal{S}_n \mid a \le \lambda_i(A) \le b, i = 1, \dots, n \}$$

where  $0 < a < b < \infty$  are given constants and  $\lambda_1(A), \ldots, \lambda_n(A)$  denote the eigenvalues of A. Since K is a convex and closed subset of  $S_n$  we may introduce the orthogonal

projection  $P_K : S_n \to K$  for which we can derive a formula as follows: given  $A \in S_n$ , let S be an orthogonal matrix such that  $SAS^t = \operatorname{diag}(\lambda_1(A), \ldots, \lambda_n(A)) =: D$ . If we let  $\tilde{D} = \operatorname{diag}(P_{[a,b]}(\lambda_1(A)), \ldots, P_{[a,b]}(\lambda_n(A)))$ , where  $P_{[a,b]}(x) := \max\{a, \min\{x, b\}\}, x \in \mathbb{R}$ , then clearly  $S^t \tilde{D}S \in K$  and we have for every  $B \in K$ 

$$(A - S^{t}\tilde{D}S) \cdot (B - S^{t}\tilde{D}S) = (D - \tilde{D}) \cdot (SBS^{t} - \tilde{D}) = \sum_{i=1}^{n} (\lambda_{i}(A) - P_{[a,b]}(\lambda_{i}(A)))(\tilde{b}_{ii} - P_{[a,b]}(\lambda_{i}(A))),$$

where  $\tilde{B} = SBS^t \in K$ . Hence  $\tilde{b}_{ii} \in [a, b], i = 1, ..., n$  which immediately yields

$$(A - S^t \tilde{D}S) \cdot (B - S^t \tilde{D}S) \le 0, \quad \text{for all } B \in K$$

and therefore  $P_K(A) = S^t \operatorname{diag} (P_{[a,b]}(\lambda_1(A)), \dots, P_{[a,b]}(\lambda_n(A))) S$ . Next, let us introduce the set

$$\mathcal{M} := \{ A \in L^{\infty}(\Omega)^{n,n} \, | \, A(x) \in K \text{ a.e. in } \Omega \}.$$

In proving the existence to problem (P) the following compactness result is crucial, see e.g. [15].

**Theorem 2.1.** Let  $(A_k)_{k\in\mathbb{N}}$  be a sequence in  $\mathcal{M}$ . Then there exists a subsequence  $(A_{k'})_{k'\in\mathbb{N}}$ and an element  $A \in \mathcal{M}$  such that for every  $g \in H^{-1}(\Omega)$ 

$$T(A_{k'},g) \rightharpoonup T(A,g) \text{ in } H^1_0(\Omega) \text{ and } A_{k'} \nabla T(A_{k'},g) \rightharpoonup A \nabla T(A,g) \text{ in } L^2(\Omega)^n.$$
 (2.1)

The sequence  $(A_{k'})_{k' \in \mathcal{M}}$  is then said to be H-convergent to A and one writes  $A_{k'} \xrightarrow{H} A$ .

**Lemma 2.2.** Suppose that  $(A_k)_{k\in\mathbb{N}}$  is a sequence in  $\mathcal{M}$  with  $A_k \xrightarrow{H} A$  and  $A_k \xrightarrow{*} A_0$  in  $L^{\infty}(\Omega)^{n,n}$ . Then  $A(x) \leq A_0(x)$  a.e. in  $\Omega$  and

$$||A||^{2} \le ||A_{0}||^{2} \le \liminf_{k \to \infty} ||A_{k}||^{2}.$$
(2.2)

*Proof.* The proof of Corollary 3.3 below will include an argument which shows in a similar setting that  $A \leq A_0$  a.e. in  $\Omega$ . Furthermore, from the Courant–Fischer minmax theorem we infer that  $\lambda_i(A(x)) \leq \lambda_i(A_0(x)), i = 1, ..., n$  and hence taking into account that  $\lambda_i(A(x)) \geq 0$ 

$$|A(x)|^2 = \sum_{i=1}^n \lambda_i (A(x))^2 \le \sum_{i=1}^n \lambda_i (A_0(x))^2 = |A_0(x)|^2 \quad \text{ a.e. in } \Omega.$$

Integration over  $\Omega$  together with the weak lower semicontinuity of the  $L^2$ -norm then implies (2.2).

We are now in position to establish the existence of a solution to the minimization problem (1.4).

**Theorem 2.3.** Problem (P) has a solution  $A \in \mathcal{M}$ .

Proof. Let  $(A_k)_{k\in\mathbb{N}} \subseteq \mathcal{M}$  be a minimizing sequence for problem (P) so that  $J(A_k) \searrow \inf_{A \in \mathcal{M}} J(A)$  as  $k \to \infty$ . Combining Theorem 2.1 with the fact that  $(A_k)_{k\in\mathbb{N}}$  is bounded in  $L^{\infty}(\Omega)^{n,n}$  we deduce that there exist  $A \in \mathcal{M}, A_0 \in L^{\infty}(\Omega)^{n,n}$  such that  $A_{k'} \xrightarrow{H} A$  and  $A_{k'} \xrightarrow{\simeq} A_0$  in  $L^{\infty}(\Omega)^{n,n}$  for some suitable subsequence. Letting  $y_{k'}^{(i)} = T(A_{k'}, f^{(i)}), y^{(i)} = T(A, f^{(i)}), 1 \leq i \leq N$ , we therefore have  $y_{k'}^{(i)} \rightharpoonup y^{(i)}$  in  $H_0^1(\Omega)$ . Hence,

$$\begin{split} J(A) &= \frac{1}{2} \sum_{i=1}^{N} \|y^{(i)} - z^{(i)}\|_{Z}^{2} + \frac{\gamma}{2} \|A\|^{2} \leq \liminf_{k' \to \infty} \frac{1}{2} \sum_{i=1}^{N} \|y^{(i)}_{k'} - z^{(i)}\|_{Z}^{2} + \frac{\gamma}{2} \liminf_{k' \to \infty} \|A_{k'}\|^{2} \\ &\leq \liminf_{k' \to \infty} J(A_{k'}) = \inf_{A \in \mathcal{M}} J(A), \end{split}$$

where we also used (2.2).

Let us next derive a suitable form of the necessary first order optimality conditions for a solution of (P). To begin, it is not difficult to verify that J is Fréchet differentiable on  $\mathcal{M}$  with

$$J'(A)H = \sum_{i=1}^{N} (y^{(i)} - z^{(i)}, w^{(i)})_{Z} + \gamma(A, H)_{L^{2}}, \quad H \in L^{\infty}(\Omega)^{n, n}$$
(2.3)

where  $y^{(i)} = T(A, f^{(i)})$  and  $w^{(i)} = D_A T(A, f^{(i)}) H \in H_0^1(\Omega), 1 \leq i \leq N$  is the partial derivative of T with respect to A in direction H which is given as the unique solution of

$$\int_{\Omega} A\nabla w^{(i)} \cdot \nabla v dx = -\int_{\Omega} H\nabla y^{(i)} \cdot \nabla v dx \quad \text{for all } v \in H_0^1(\Omega).$$
(2.4)

In order to rewrite (2.3) we introduce the functions  $p^{(i)} \in H_0^1(\Omega)$ , i = 1, ..., N as the unique solutions of the following adjoint problems:

$$\int_{\Omega} A\nabla v \cdot \nabla p^{(i)} dx = (y^{(i)} - z^{(i)}, v)_Z \quad \text{for all } v \in H^1_0(\Omega).$$
(2.5)

Abbreviating  $(a \otimes b)_{kl} = \frac{1}{2}(a_k b_l + a_l b_k), k, l = 1, \dots, n$  for  $a, b \in \mathbb{R}^n$  we then have

$$J'(A)H = \int_{\Omega} \left( -\sum_{i=1}^{N} \nabla y^{(i)} \otimes \nabla p^{(i)} + \gamma A \right) \cdot H dx, \quad H \in L^{\infty}(\Omega)^{n,n}.$$
(2.6)

Note that the above integral exists since  $\nabla y^{(i)} \otimes \nabla p^{(i)} \in L^1(\Omega)^{n,n}$ . In conclusion

**Theorem 2.4.** Let  $A \in \mathcal{M}$  be a solution of (P). Then for every  $\lambda > 0$ 

$$A(x) = P_K \left( A - \lambda \left( \gamma A - \sum_{i=1}^N \nabla y^{(i)}(x) \otimes \nabla p^{(i)}(x) \right) \right) \quad a.e. \text{ in } \Omega$$

*Proof.* The optimality of A implies that  $J'(A)(\tilde{A} - A) \ge 0$  for all  $\tilde{A} \in \mathcal{M}$  which can be rewritten with the help of (2.6) as follows:

$$\int_{\Omega} \left( A - \lambda \left( \gamma A - \sum_{i=1}^{N} \nabla y^{(i)}(x) \otimes \nabla p^{(i)}(x) \right) - A \right) \cdot \left( \tilde{A} - A \right) dx \le 0 \quad \text{ for all } \tilde{A} \in \mathcal{M}.$$

A localization argument shows that A(x) is the orthogonal projection of

$$A(x) - \lambda \left( \gamma A(x) - \sum_{i=1}^{N} \nabla y^{(i)}(x) \otimes \nabla p^{(i)}(x) \right)$$

onto K a.e. in  $\Omega$  which implies the result.

Let us note that the particular choice  $\lambda = \frac{1}{\gamma}$  gives

$$A(x) = P_K\left(\frac{1}{\gamma}\left(\sum_{i=1}^N \nabla y^{(i)}(x) \otimes \nabla p^{(i)}(x)\right)\right) \text{ a.e. in } \Omega.$$
(2.7)

#### 3 Finite element discretization

Let  $\mathcal{T}_h$  be a regular triangulation of  $\Omega$  with maximum mesh size  $h := \max_{T \in \mathcal{T}_h} \operatorname{diam}(T)$  and suppose that  $\overline{\Omega}$  is the union of the elements of  $\mathcal{T}_h$ ; boundary elements are allowed to have one curved face. We define the space of linear finite elements,

$$X_h := \{ v_h \in H^1_0(\Omega) \, | \, v_h \text{ is a linear polynomial on each } T \in \mathcal{T}_h \}.$$

It is well known that there exists an interpolation operator  $\Pi_h: H_0^1(\Omega) \to X_h$  such that

$$\Pi_h w \to w \text{ in } H^1(\Omega) \text{ as } h \to 0 \quad \text{ for every } w \in H^1_0(\Omega).$$
(3.1)

For given  $A \in \mathcal{M}$  and  $g \in H^{-1}(\Omega)$ , the problem

$$\int_{\Omega} A\nabla y_h \cdot \nabla v_h dx = \langle g, v_h \rangle \quad \text{ for all } v_h \in X_h$$

has a unique solution  $y_h = T_h(A, g) \in X_h$ . Furthermore, a standard argument yields the error bound

$$\|y - y_h\|_{H_0^1} \le \frac{b}{a} \inf_{v_h \in X_h} \|y - v_h\|_{H_0^1}, \quad \text{where } y = T(A, g).$$
(3.2)

In order to set up an approximation of (P) we use variational discretization as in [5] and consider

$$(P_h) \quad \min_{A \in \mathcal{M}} J_h(A) := \frac{1}{2} \sum_{i=1}^N \|y_h^{(i)} - z^{(i)}\|^2 + \frac{\gamma}{2} \|A\|^2 \text{ s.t.} y_h^{(i)} = T_h(A, f^{(i)}), 1 \le i \le N.$$
(3.3)

Similar arguments as in Section 2 show that  $J_h$  is Fréchet differentiable and that for  $A \in \mathcal{M}$ 

$$J_{h}'(A)H = \int_{\Omega} \left( -\sum_{i=1}^{N} \nabla y_{h}^{(i)} \otimes \nabla p_{h}^{(i)} + \gamma A \right) \cdot Hdx, \quad H \in L^{\infty}(\Omega)^{n,n}.$$
(3.4)

Since dim $X_h < \infty$  it is straightforward to see that  $(P_h)$  has a solution  $A_h \in \mathcal{M}$ . Furthermore, every solution  $A_h$  of  $(P_h)$  satisfies

$$A_h(x) = P_K\left(\frac{1}{\gamma}\sum_{i=1}^N \nabla y_h^{(i)}(x) \otimes \nabla p_h^{(i)}(x)\right) \text{ a.e. in } \Omega,$$
(3.5)

cf. (2.7). Here,  $y_h^{(i)} = T_h(A_h, f^{(i)})$  and  $p_h^{(i)} \in X_h$  are the solutions of the adjoint problems

$$\int_{\Omega} A_h \nabla v_h \cdot \nabla p_h^{(i)} dx = (y_h^{(i)} - z^{(i)}, v_h)_Z \quad \text{for all } v_h \in X_h, 1 \le i \le N.$$
(3.6)

**Remark 3.1.** Let us note that in view of (3.5)  $A_h$  is piecewise constant so that a discretization of the set  $\mathcal{M}$  is not required. Variational discretization automatically yields solutions to (3.3) which allow a finite-dimensional representation.

In order to investigate the convergence of the approximate solutions we shall employ a discrete version of Theorem 2.1.

**Theorem 3.2.** Let  $(A_h)_{h>0}$  be a sequence in  $\mathcal{M}$ . Then there exists a subsequence  $(A_{h'})_{h'>0}$ and  $A \in \mathcal{M}$  such that for every  $g \in H^{-1}(\Omega)$ 

$$T_{h'}(A_{h'},g) \rightharpoonup T(A,g) \text{ in } H^1_0(\Omega) \text{ and } A_{h'} \nabla T_{h'}(A_{h'},g) \rightharpoonup A \nabla T(A,g) \text{ in } L^2(\Omega)^n.$$
 (3.7)

We then say that the sequence  $(A_{h'})_{h' \in \mathcal{M}}$  Hd-converges to A and write  $A_{h'} \stackrel{Hd}{\rightarrow} A$ .

*Proof.* The line of argument follows the corresponding proof in the continuous case (see [15]) and a similar result for a finite volume scheme, see [3]. We therefore only sketch the main steps.

Step 1: One first shows that there exists a subsequence, for ease of notation again denoted by  $(A_h)_{h>0}$ , and continuous linear operators  $S: H^{-1}(\Omega) \to H^1_0(\Omega), R: H^{-1}(\Omega) \to L^2(\Omega)^n$ such that for every  $g \in H^{-1}(\Omega)$ 

$$T_h(A_h, g) \rightharpoonup S(g) \text{ in } H^1_0(\Omega), \quad A_h \nabla T_h(A_h, g) \rightharpoonup R(g) \text{ in } L^2(\Omega)^n \quad \text{as } h \to 0.$$
 (3.8)

Step 2: We show that S is invertible. For  $g \in H^{-1}(\Omega)$  denote by  $w \in H^1_0(\Omega), w_h \in X_h$  the solutions of

$$\int_{\Omega} \nabla w \cdot \nabla v dx = \langle g, v \rangle, \quad v \in H_0^1(\Omega), \qquad \int_{\Omega} \nabla w_h \cdot \nabla v_h dx = \langle g, v_h \rangle, \quad v_h \in X_h.$$

Clearly,  $||w||_{H_0^1} = ||g||_{H^{-1}}$  and  $w_h \to w$  in  $H_0^1(\Omega)$  in view of (3.1). Setting  $y_h = T_h(A_h, g)$  we have in addition that

$$\int_{\Omega} A_h \nabla y_h \cdot \nabla v_h dx = \langle g, v_h \rangle = \int_{\Omega} \nabla w_h \cdot \nabla v_h dx, \quad v_h \in X_h,$$

from which we infer that  $||w_h||_{H_0^1} \leq b||y_h||_{H_0^1}$  recalling the definition of  $\mathcal{M}$ . Combining this bound with (3.8) and using again the properties of  $\mathcal{M}$  we deduce that

$$\|g\|_{H^{-1}}^{2} = \|w\|_{H^{1}_{0}}^{2} = \lim_{h \to 0} \|w_{h}\|_{H^{1}_{0}}^{2} \le b^{2} \liminf_{h \to 0} \|y_{h}\|_{H^{1}_{0}}^{2}$$

$$\le \frac{b^{2}}{a} \liminf_{h \to 0} \int_{\Omega} A_{h} \nabla y_{h} \cdot \nabla y_{h} dx = \frac{b^{2}}{a} \liminf_{h \to 0} \langle g, y_{h} \rangle = \frac{b^{2}}{a} \langle g, S(g) \rangle,$$

$$(3.9)$$

which implies that S is invertible.

Step 3: Let  $C: H_0^1(\Omega) \to L^2(\Omega)^n$  be defined by  $Cv := RS^{-1}v$ . For a given  $g \in H^{-1}(\Omega)$  the function  $y_h = T_h(A_h, g)$  satisfies by definition

$$\int_{\Omega} A_h \nabla y_h \cdot \nabla v_h dx = \langle g, v_h \rangle \quad \text{for all } v_h \in X_h.$$
(3.10)

Sending  $h \to 0$  and taking into account (3.8) and (3.1) we infer

$$\int_{\Omega} Cy \cdot \nabla v dx = \langle g, v \rangle \quad \text{for all } v \in H_0^1(\Omega), \quad \text{where } y = S(g).$$
(3.11)

Next, let  $g, \tilde{g} \in H^{-1}(\Omega)$  be arbitrary and define  $y = S(g), \tilde{y} = S(\tilde{g})$  as well as  $y_h = T_h(A_h, g), \tilde{y}_h = T_h(A_h, \tilde{g})$ . Recalling (3.10) we have for every  $\varphi \in C_0^{\infty}(\Omega)$ 

$$\int_{\Omega} \varphi A_h \nabla y_h \cdot \nabla \tilde{y}_h dx = \int_{\Omega} A_h \nabla y_h \cdot \nabla r_h dx + \langle g, \varphi \tilde{y}_h \rangle - \langle g, r_h \rangle - \int_{\Omega} A_h \nabla y_h \cdot \nabla \varphi \, \tilde{y}_h dx,$$

where we have abbreviated  $r_h = \varphi \tilde{y}_h - I_h(\varphi \tilde{y}_h)$  and  $I_h$  denotes the Lagrange interpolation operator. A standard interpolation estimate implies

$$\|\varphi \tilde{y}_{h} - I_{h}(\varphi \tilde{y}_{h})\|_{H^{1}(T)} \le Ch \|D^{2}(\varphi \tilde{y}_{h})\|_{L^{2}(T)} \le Ch \|\varphi\|_{H^{2,\infty}(T)} \|\tilde{y}_{h}\|_{H^{1}(T)}, \quad T \in \mathcal{T}_{h},$$

so that  $r_h \to 0$  in  $H_0^1(\Omega)$  as  $h \to 0$  since  $\|\tilde{y}_h\|_{H^1} \leq C$ . Observing in addition that  $A_h \nabla y_h \rightharpoonup C y$ in  $L^2(\Omega)^n$ ,  $\varphi \tilde{y}_h \rightharpoonup \varphi \tilde{y}$  in  $H_0^1(\Omega)$  and  $\tilde{y}_h \to \tilde{y}$  in  $L^2(\Omega)$  we obtain

$$\lim_{h \to 0} \int_{\Omega} \varphi A_h \nabla y_h \cdot \nabla \tilde{y}_h dx = \langle g, \varphi \tilde{y} \rangle - \int_{\Omega} C y \cdot \nabla \varphi \; \tilde{y} dx,$$

which, combined with (3.11), yields

$$\lim_{h \to 0} \int_{\Omega} \varphi A_h \nabla y_h \cdot \nabla \tilde{y}_h dx = \int_{\Omega} \varphi C y \cdot \nabla \tilde{y} dx.$$
(3.12)

Similarly as in [3, Proof of Theorem 2] one now deduces from (3.11) and (3.12) that there exists  $A \in \mathcal{M}$  such that

$$(Cy)(x) = A(x)\nabla y(x)$$
 a.e. in  $\Omega$ . (3.13)

This completes the proof of the theorem.

**Corollary 3.3.** Suppose that  $(A_h)_{h>0}$  is a sequence in  $\mathcal{M}$  with  $A_h \xrightarrow{Hd} A$  and  $A_h \xrightarrow{*} A_0$  in  $L^{\infty}(\Omega)^{n,n}$ . Then  $A \leq A_0$  a.e. in  $\Omega$  and  $||A||^2 \leq ||A_0||^2 \leq \liminf_{h\to 0} ||A_h||^2$ .

Proof. We use the same notation as in the proof of Theorem 3.2. By Step 2 above there exists for every  $y \in H_0^1(\Omega)$  an element  $g \in H^{-1}(\Omega)$  such that y = S(g). Defining  $y_h = T_h(A_h, g)$ there holds  $y_h \to y$  in  $H_0^1(\Omega)$ ,  $A_h \nabla y_h \to A \nabla y$  in  $L^2(\Omega)^n$ . Furthermore, we have for any  $\varphi \in C_0^{\infty}(\Omega), \varphi \ge 0$  that

$$0 \leq \int_{\Omega} \varphi A_h \nabla (y_h - y) \cdot \nabla (y_h - y) dx$$
  
= 
$$\int_{\Omega} \varphi A_h \nabla y_h \cdot \nabla y_h dx - 2 \int_{\Omega} \varphi A_h \nabla y_h \cdot \nabla y dx + \int_{\Omega} \varphi A_h \nabla y \cdot \nabla y dx$$

Recalling (3.12), (3.13) and the fact that  $A_h \stackrel{*}{\rightharpoonup} A_0$  in  $L^{\infty}(\Omega)^{n,n}$  we obtain upon sending  $h \to 0$ 

$$0 \leq -\int_{\Omega} \varphi A \nabla y \cdot \nabla y dx + \int_{\Omega} \varphi A_0 \nabla y \cdot \nabla y dx,$$

from which we infer that  $A\nabla y \cdot \nabla y \leq A_0 \nabla y \cdot \nabla y$  a.e. in  $\Omega$ . Since  $y \in H_0^1(\Omega)$  is arbitrary we deduce that  $A \leq A_0$  a.e. in  $\Omega$ . The remaining estimate is obtained in the same way as in the proof of Lemma 2.2.

We are now in position to prove a convergence result for a sequence  $(A_h)_{h>0}$  of solutions of  $(P_h)$ .

**Theorem 3.4.** Let  $A_h \in \mathcal{M}$  be a solution of  $(P_h)$ . Then there exists a subsequence  $(A_{h'})_{h'>0}$ and  $A \in \mathcal{M}$  such that  $A_{h'} \to A$  in  $L^2(\Omega)^{n,n}$ ,  $T_{h'}(A_{h'}, f^{(i)}) \to T(A, f^{(i)})$  in  $Z, 1 \leq i \leq N$  and A is a solution of (P).

Proof. In view of Theorem 3.2 and Corollary 3.3 there exists a subsequence, again denoted by  $(A_h)_{h>0}$ , and  $A \in \mathcal{M}$  such that  $A_h \stackrel{Hd}{\to} A$  and  $A_h \stackrel{*}{\to} A_0$  in  $L^{\infty}(\Omega)^{n,n}$  with  $A \leq A_0$ a.e. in  $\Omega$ . Let  $y_h^{(i)} = T_h(A_h, f^{(i)}), y^{(i)} = T(A, f^{(i)}), 1 \leq i \leq N$ . Then  $y_h^{(i)} \to y^{(i)}$  in  $H_0^1(\Omega)$ ,  $A_h \nabla y_h^{(i)} \to A \nabla y^{(i)}$  in  $L^2(\Omega)^n$ , so that we may deduce similarly as in the proof of Theorem 2.3 that  $J(A) \leq \liminf_{h \to 0} J_h(A_h)$ . Next, Theorem 2.3 implies that (P) has a solution  $\overline{A} \in \mathcal{M}$ . Then we have

$$J(\bar{A}) \le J(A) \le \liminf_{h \to 0} J_h(A_h) \le \limsup_{h \to 0} J_h(A_h) \le \limsup_{h \to 0} J_h(\bar{A}) = J(\bar{A}),$$

where the last equality follows from (3.2) and (3.1). We deduce that

$$\lim_{h \to 0} J_h(A_h) = J(A) = J(\bar{A}), \tag{3.14}$$

in particular, A is a minimum of J. Furthermore, we have

$$\begin{split} \frac{1}{2} \sum_{i=1}^{N} \|y_{h}^{(i)} - y^{(i)}\|_{Z}^{2} + \frac{\gamma}{2} \|A_{h} - A\|^{2} &= \frac{1}{2} \sum_{i=1}^{N} \|(y_{h}^{(i)} - z^{(i)}) - (y^{(i)} - z^{(i)})\|_{Z}^{2} + \frac{\gamma}{2} \|A_{h} - A\|^{2} \\ &= \frac{1}{2} \sum_{i=1}^{N} \|y_{h}^{(i)} - z^{(i)}\|_{Z}^{2} - \sum_{i=1}^{N} (y_{h}^{(i)} - z^{(i)}, y^{(i)} - z^{(i)})_{Z} + \frac{1}{2} \sum_{i=1}^{N} \|y^{(i)} - z^{(i)}\|_{Z}^{2} \\ &+ \frac{\gamma}{2} \|A_{h}\|^{2} - \gamma(A_{h}, A) + \frac{\gamma}{2} \|A\|^{2} \\ &= J_{h}(A_{h}) + J(A) - \sum_{i=1}^{N} (y_{h}^{(i)} - z^{(i)}, y^{(i)} - z^{(i)})_{Z} - \gamma(A_{h}, A) \\ &\to 2J(A) - \sum_{i=1}^{N} \|y^{(i)} - z^{(i)}\|_{Z}^{2} - \gamma(A_{0}, A) \\ &\leq 2J(A) - \sum_{i=1}^{N} \|y^{(i)} - z^{(i)}\|_{Z}^{2} - \gamma\|A\|^{2} = 0, \end{split}$$

where we have used (3.14) and the fact that  $A \leq A_0$  a.e. in  $\Omega$ . The theorem is proved.

#### 4 Numerical examples

Let  $\Omega := (-1,1)^2 \subset \mathbb{R}^2$ . We consider a finite element approximation with piecewise linear, continuous functions defined on a triangulation containing 512 triangles, constructed with

the POIMESH environment of MATLAB. We take N = 1 with data (z, f) given by  $z = I_h y$ where

$$y(x_1, x_2) = (1 - x_1^2)(1 - x_2^2)$$
 and  $f(x_1, x_2) = (1 - x_2^2)(6x_1^2 + 2) + 2(1 - x_1^2)$ .

Note that y is the solution of (1.1) when

$$A(x_1, x_2) = \begin{bmatrix} 1 + x_1^2 & 0 \\ 0 & 1 \end{bmatrix}.$$

In the definition of K we have chosen a = 0.5 and b = 10. The discrete problem (3.3) is solved using the projected steepest descent method with Armijo step size rule, see e.g. [9]. In view of Remark 3.1 it is sufficient to iterate within the class of matrices in  $\mathcal{M}$  that are piecewise constant. Given such an A the new iterate is computed according to

$$A^{+} = A(\tau) \text{ with } \tau = \max_{l \in \mathbb{N}} \{\beta^{l}; J_{h}(A(\beta^{l})) - J_{h}(A) \leq -\frac{\sigma}{\beta^{l}} \|A(\beta^{l}) - A\|^{2} \}$$

where  $\beta \in (0, 1)$  and

$$A(\tau)_{|T} := P_K \Big( A_{|T} + \tau \big( \nabla y_{h|T} \otimes \nabla p_{h|T} - \gamma A_{|T} \big) \Big), \quad T \in \mathcal{T}_h.$$

Here,  $y_h = T_h(A, f)$  and  $p_h$  is the solution of the adjoint problem (3.6). In our calculations we chose  $\gamma = 0.001$ ,  $\sigma = 10^{-4}$ ,  $\beta = 0.5$  and as initial matrix

$$A^0 := \left[ \begin{array}{cc} 2 & -1 \\ -1 & 2 \end{array} \right].$$

The iteration was stopped if  $||A^+ - A(1)|| \leq \tau_a + \tau_r ||A^0 - A^0(1)||$  or the maximum number of 5000 iterations was reached. For  $\tau_a = 10^{-3}$  and  $\tau_r = 10^{-2}$  we have  $||A^0 - A^0(1)|| =$  $7.94 \times 10^{-2}$ ,  $J_h(A^0) = 2.18 \times 10^{-1}$  and the algorithm terminated after 400 iterations with  $\tilde{A}$ and  $\tilde{y}_h = T_h(\tilde{A}, f)$  such that

$$\|\tilde{y}_h - z\| = 1.02 \times 10^{-2}, \quad \|A - \tilde{A}\| = 2.05 \text{ and } J_h(\tilde{A}) = 2.77 \times 10^{-2}.$$

Note that we cannot expect the difference  $A-\tilde{A}$  to become small since the diffusion matrix will not be determined uniquely by just one set of data. Performing 5000 iterations we obtained  $\tilde{A}$  and  $\tilde{y}_h$  such that

$$\|\tilde{y}_h - z\| = 8.22 \times 10^{-3}, \quad \|A - \tilde{A}\| = 1.53 \text{ and } J_h(\tilde{A}) = 2.32 \times 10^{-2}.$$

Fig. 1 from left to right shows  $\tilde{y}_h, z$  and  $\tilde{y}_h - z$  after 400 iterations.

By combining the projected gradient method with a homotopy in the parameter  $\gamma$  we were also able to treat the case  $\gamma = 0$ . We started with  $\gamma = 1$  and reduced  $\gamma$  by a factor of 0.8 after every ten iterations. Using the same notation as above we obtained after 5000 iterations

$$\|\tilde{y}_h - z\| = 9.61 \times 10^{-4}, \quad \|A - \tilde{A}\| = 1.40$$

and the corresponding results are displayed in Fig. 2. One observes that the difference between  $\tilde{y}_h$  and z is comparatively large in regions where  $\nabla y$  is small which is in agreement with classical results on the identifiability of scalar diffusion coefficients, see e.g. [14].



Figure 1: Numerical solution, desired state, and error  $\tilde{y}_h - z$  for  $\gamma = 1. \times 10^{-3}$  after the stopping criterion of the projected steepest descent method is met.



Figure 2: Numerical solution, desired state, error  $\tilde{y}_h - z$  for  $\gamma = 0$  after 5000 iterations of the steepest descent method.

#### Acknowledgements

The presentation of the numerical results is partly based on a MATLAB code developed by Ronny Hoffmann in his diploma thesis [7] written under the supervision of the second author. The authors gratefully acknowledge the support of the DFG Priority Program 1253 entitled *Optimization With Partial Differential Equations*.

#### References

- Alt, H.W., Hoffmann, K.H., Sprekels, J.: A numerical procedure to solve certain identification problems, Intern. Ser. Numer. Math. 68, 11–43 (1984).
- [2] Chicone, C., Gerlach, J.: A note on the identifiability of distributed parameters in elliptic equations, SIAM J. Math. Anal. 18, 1378–1384 (1987).
- [3] Eymard, R., Gallouët, T.: H-convergence and numerical schemes for elliptic problems, Siam J. Numer. Anal. 41, 539–562 (2003).
- [4] Falk, R.S.: Error estimates for the numerical identification of a variable coefficient, Math. Comput. 40, 537–546 (1983).
- [5] Hinze, M.: A variational discretization concept in control constrained optimization: the linear-quadratic case, Comput. Optim. Appl. 30, 45–61 (2005).

- [6] Hoffmann, K.H., Sprekels, J.: On the identification of coefficients of elliptic problems by asymptotic regularization. Numer. Funct. Anal. Optim. 7, 157-177 (1984/85).
- [7] Hoffmann, R.: Entwicklung numerischer Methoden zur Schätzung matrixwertiger verteilter Parameter bei elliptischen Differentialgleichungen, Diploma thesis, TU Dresden, 2005.
- [8] Hsiao, G.C., Sprekels, J.: A stability result for distributed parameter identification in bilinear systems, Math. Meth. Appl. Sciences 10, 447–456 (1988).
- [9] Kelley, C.T.: Iterative Methods for Optimization. SIAM, 1999.
- [10] Kohn, R.V., Lowe, B.D.: A variational method for parameter identification, RAIRO Modél. Math. Anal. Numér. 22, 119–158 (1988).
- [11] Kunisch, K.: Numerical methods for parameter estimation problems, Inverse problems in diffusion processes (Lake St. Wolfgang, 1994), 199-216, SIAM, Philadelphia, PA, 1995.
- [12] Leugering, G., Stingl, M.: PDE-constrained optimization for advanced materials. In Constrained Optimization and Optimal Control for Partial Differential Equations. Birkhäuser, G. Leugering et al. Eds., 2010.
- [13] Rannacher, R., Vexler, B.: A priori estimates for the finite element discretization of elliptic parameter identification problems with pointwise measurements, SIAM J. Cont. Optim. 44, 1844–1863 (2005).
- [14] Richter, G.R.: An inverse problem for the steady state diffusion equation, SIAM J. Appl. Math. 41, 210-221 (1981).
- [15] Tartar, L.: Estimation of homegenized coefficients. Topics in the mathematical modelling of composite materials, page 9-20, Andrej Charkaev, Robert Kohn Eds., 1997.
- [16] Vainikko, G., Kunisch, K.: Identifiability of the transmissivity coefficient in an elliptic boundary value problem, Z. Anal. Anwendungen 12, 327-341 (1993).
- [17] Wang, L., Zou, J.: Error estimates of finite element methods for parameter identification problems in elliptic and parabolic systems, Discrete Contin. Dyn. Syst. Ser. B 14, 1641-1670 (2010).