Admittivity imaging from multi-frequency micro-electrical impedance tomography^{*}

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Abstract

The aim of this paper is to propose an optimal control optimization algorithm for reconstructing admittivity distributions (i.e., both conductivity and permittivity) from multi-frequency micro-electrical impedance tomography. A convergent and stable optimization scheme is shown to be obtainable from multi-frequency data. The results of this paper have potential applicability in cancer imaging, cell culturing and differentiation, food sciences, and biotechnology.

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1 Introduction

This paper aims at proposing and analyzing an optimal control approach for imaging the admittivity distributions of biological tissues.

Biological tissues possess characteristic distributions of permittivity and conductivity [33]. Conductivity can be regarded as a measure of the ability to transport

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charge throughout material's volume under an applied electric field, while permittivity is a measure of the ability of the dipoles within a material to rotate (or of the material to store charge) under an applied external field. At low frequencies, biological tissues behave like a conductor, but capacitive effects become important at higher frequencies due to the membranous structures [38, 40]. The electric behavior of a biological tissue under the influence of an electric field at frequency ω can be characterized by its frequency-dependent admittivity $\sigma + i\omega\varepsilon$, where σ and ε are respectively its conductivity and permittivity. As recently shown in [7, 40], spectroscopic admittivity imaging can provide information about the microscopic structure of a medium, from which physiological or pathological conditions of tissue can be derived, because the admittivity of biological tissue varies with its composition, membrane characteristics, intra-and extra-cellular fluids, and other factors.

In this paper, we consider the imaging of admittivity distributions of biological tissues from multi-frequency micro-electrical impedance data. Micro-electrical impedance tomography [25, 31] can be used to reconstruct a high resolution admittivity distribution from internal measurements of electrical potential at multiple frequencies. The technique uses planar arrays of micro-electrodes to nondestructively sense thin layers of biological samples [14, 25, 27, 36, 44]. It has potential applications in cell electrofusion and electroporation, cell culturing, cell differentiation and drug screening; see [12, 25, 28, 29, 30, 34, 35, 37, 47]. It is capable of high-resolution imaging. Other methods of electrical tissue property imaging using internal data are investigated in [3, 4, 5, 6, 11, 16, 41, 42, 46]. Resolution and stability enhancements are achieved from internal measurements [8, 9, 10].

To solve the admittivity imaging problem from multi-frequency micro-electrical data, we propose an optimal control optimization algorithm and rigorously prove its stability and convergence properties. Internal potential measurements at a single frequency are known to be insufficient for reconstructing the admittivity distribution. An initial guess is constructed by solving a boundary value problem. Note that the method of characteristics [23] can not be used to solve the transport equation satisfied by the logarithm of the admittivity, because the electrical potential is a complex valued function. It is unlikely that a direct (noniterative) method can be designed for solving the admittivity imaging problem. As far as we know, the approach in this paper and the analysis of its convergence and stability have not been reported elsewhere. The convergence result is among the very few algorithms for reconstructing the electrical properties of tissue from internal data.

To formulate mathematically the imaging problem, we consider a medium of conductivity σ and permittivity ε occupying Ω , \mathcal{C}^2 -domain of \mathbb{R}^2 . (Hereafter, the medium is simply called Ω .) The problem of micro–electrical impedance tomography is to reconstruct σ and ε from the vector of potential $u_{\omega}, \omega \in (\underline{\omega}, \overline{\omega})$, the solution of

$$\begin{cases} \nabla \cdot (\sigma + i\omega\varepsilon)\nabla u_{\omega} = 0 & \text{in } \Omega, \\ u_{\omega} = \varphi & \text{on } \partial\Omega, \end{cases}$$
(1.1)

where $\varphi = (\varphi_1, \varphi_2) \in H^{1/2}(\partial \Omega)^2$. It is proved in this paper that the above inverse problem is stably solvable with a good choice of boundary datum φ ; that is, φ belongs to what we will refer to as the proper set of boundary measurements; see [5, 41, 43].

The paper is organized as follows. First, in section 2 we review some useful regularity results for elliptic systems of partial differential equations. In section 3 we introduce the set of proper boundary measurements. Section 4 is devoted to the reconstruction method. We prove that the minimization functional is Fréchet differentiable and we compute its derivative. Then we construct an initial guess and prove the convergence of a minimizing sequence. The paper ends with a short discussion. In the appendix, we prove the convergence of Landweber sequences with cutoff functions.

2 Preliminaries on regularities

Let $\Omega' = \{x \in \Omega : \operatorname{dist}(x, \partial \Omega) > c_0\}$ for a small constant $c_0 > 0$. We assume that σ and ε are constant and known in $\Omega \setminus \Omega'$. In the following, we let σ_* and ε_* , the true conductivity and permittivity of Ω , belong to the convex subset of $H^2(\Omega)^2$ given by

$$\mathcal{S} = \{ (\sigma, \varepsilon) := (\sigma_0, \varepsilon_0) + (\eta_1, \eta_2) | (q_1, q_2) \in \mathcal{S} \},\$$

where the positive constants σ_0 and ε_0 are respectively the conductivity and permittivity in $\Omega \setminus \Omega'$ and

$$\mathcal{S} = \{ (\eta_1, \eta_2) \in H_0^2(\Omega)^2 | c_1 < \eta_1 + \sigma_0 < c_2, c_1 < \eta_2 + \epsilon_0 < c_2, \text{ supp } \eta_j \subset \Omega', \\ \|\eta_j\|_{H^2(\Omega)} \le c_3 \|\eta_j\|_{H^1(\Omega)}, \|\eta_j\|_{H^1(\Omega)} \le c_4 \text{ for } j = 1, 2 \}$$

$$(2.1)$$

with c_1, c_2, c_4 and c_4 being positive constants and supp denoting the support. In other words, we can write $\widetilde{\mathcal{S}} = (\sigma_0, \varepsilon_0) + \mathcal{S}$. Here, the condition of $\|\eta_j\|_{H^2(\Omega)} \leq c_3 \|\eta_j\|_{H^1(\Omega)}$ is used to exclude any micro-local oscillation on the admittivity distribution.

Introducing an open subset of \mathbb{C}

$$\mathcal{O} := \left\{ o \in \mathbb{C} | \Im m \, o < \frac{c_1}{2c_2} \right\},\tag{2.2}$$

we first establish a useful lemma, which is a direct consequence of standard regularity results.

Lemma 2.1. Let $(\sigma, \varepsilon) \in \widetilde{S}$, $\omega \in \mathcal{O}$, and $f \in L^p(\Omega)$ for $2 \leq p < \infty$. If $v \in H^1(\Omega)$ satisfy

$$\nabla \cdot (\sigma + i\omega\varepsilon)\nabla v = f \quad in \quad \Omega, \tag{2.3}$$

then $v \in W^{2,p}(\Omega')$ and

$$\|v\|_{W^{2,p}(\Omega')} \leq C (\|v\|_{L^{p}(\Omega)} + \|f\|_{L^{p}(\Omega)}),$$
(2.4)

where C depends only on $c_i, i = 0, ..., 4$, p, and Ω . Moreover, if v = 0 on $\partial \Omega$, then

$$\|v\|_{W^{2,p}(\Omega)} \leq C (\|v\|_{L^{p}(\Omega)} + \|f\|_{L^{p}(\Omega)}).$$
(2.5)

Proof. From the standard regularity estimate, we have

$$\|v\|_{H^{2}(\Omega')} \leq C (\|f\|_{L^{2}(\Omega)^{2}} + \|v\|_{L^{2}(\Omega)}).$$
(2.6)

The first equation in (2.1) can be rewritten as

$$\Delta v = -\nabla v^T \frac{\nabla(\sigma + i\omega\varepsilon)}{\sigma + i\omega\varepsilon} + \frac{f}{\sigma + i\omega\varepsilon},$$
(2.7)

where T denotes the transpose. Since supp $\nabla(\sigma + i\omega\varepsilon) \subset \Omega'$, we have

$$\begin{aligned} \|\nabla v^{T} \frac{\nabla(\sigma + i\omega\varepsilon)}{\sigma + i\omega\varepsilon}\|_{L^{p}(\Omega)} &= \|\nabla v^{T} \frac{\nabla(\sigma + i\omega\varepsilon)}{\sigma + i\omega\varepsilon}\|_{L^{p}(\Omega')} \\ &\leq C \|\nabla v^{T}\|_{L^{2p}(\Omega')^{2}} \|\frac{\nabla(\sigma + i\omega\varepsilon)}{\sigma + i\omega\varepsilon}\|_{L^{2p}(\Omega')^{2}} \\ &\leq C \|v\|_{H^{2}(\Omega')} \|\sigma + i\omega\varepsilon\|_{H^{2}(\Omega')} \\ &\leq C \left(\|v\|_{L^{2}(\Omega')} + \|f\|_{L^{2}(\Omega)}\right) \|\sigma + i\omega\varepsilon\|_{H^{2}(\Omega')}.\end{aligned}$$

Here, Schwartz inequality was used for the second inequality; Sobolev embedding for the third inequality; and the last inequality comes from (2.6). Hence, the right side of (2.7) is in $L^{p}(\Omega)$. Now, we apply the standard $W^{2,p}$ -estimate for Poisson's equation (2.7) to get

$$\|v\|_{W^{2,p}(\Omega')} \leq C \left(\|v\|_{L^{p}(\Omega)^{2}} + \|f\|_{L^{p}(\Omega)} \right).$$

3 Sets of proper boundary conditions

The main purpose of this section is to choose "good" boundary datum φ in (1.1) so that the measurements of the corresponding vector potential u_{ω} are helpful in our reconstruction algorithm. Such a set of good functions, henceforth coined as *a* set of proper boundary conditions, is defined as follows.

Definition 3.1. Let $\varphi \in H^{1/2}(\partial \Omega)^2$. We say that φ is a proper set of boundary conditions if and only if the 2×2 matrix ∇u_{σ} is invertible in Ω for all $\sigma \in \sigma_0 + S$ where the vector u_{σ} denotes the solution of the boundary value problem

$$\left\{ \begin{array}{ll} \nabla\cdot\sigma\nabla u=0 & \quad in\ \Omega,\\ u=\varphi & \quad on\ \partial\Omega. \end{array} \right.$$

The existence of a set of proper boundary conditions was proved in [2, 13, 39]. The following proposition is the main result of this section.

Proposition 3.1. For all $(\sigma, \varepsilon) \in \widetilde{S}$, we denote by u_{ω} the solution of (1.1) with φ being a proper set of boundary conditions. There exist N > 1 open pairwise disjoint open subsets B_1, B_2, \dots, B_N of Ω , and N frequencies $\omega_1, \dots, \omega_N \in (\underline{\omega}, \overline{\omega})$ such that

- (i) $\overline{\Omega'} \subset \bigcup_{j=1}^N \overline{B}_j \subset \Omega;$
- (ii) The matrix $A_{\omega_i}(x) = \nabla u_{\omega}$ is invertible for all $x \in B_j$.

In [1], G. Alberti has proved the result when the dependence of coefficients on the frequency is different from that in our context. The key of his arguments is the fact that u_{ω} is analytic with respect to ω . Fortunately, his technique is still applicable to (1.1). We present the proof here for the completeness' sake.

Lemma 3.1. Let \mathcal{O} be defined by (2.2). The map

$$L: \mathcal{O} \to H^2_{loc}(\Omega)^2,$$
$$\omega \mapsto u_{\omega},$$

where u_{ω} is the solution to (1.1), is analytic. Moreover, the derivative of L at ω_0 is given by the solution of

$$\begin{cases} \nabla \cdot (\sigma + i\omega_0 \varepsilon) \nabla w = -\nabla \cdot i\varepsilon \nabla L(\omega_0) & in \ \Omega, \\ w = 0 & on \ \partial \Omega \end{cases}$$
(3.1)

for all $\omega_0 \in \mathcal{O}$.

Proof. The quotient

$$z := \frac{L(\omega) - L(\omega_0)}{\omega - \omega_0}$$

solves

$$\begin{cases} \nabla \cdot (\sigma + i\omega\varepsilon)\nabla z = -i\nabla \cdot \varepsilon\nabla L(\omega_0) & \text{in } \Omega, \\ z = 0 & \text{on } \partial\Omega. \end{cases}$$
(3.2)

Since $\nabla \cdot \varepsilon \nabla L(\omega_0) = 0$ in $\Omega \setminus \overline{\Omega'}$ and $\nabla \cdot \varepsilon \nabla L(\omega_0)$ is in $L^2(\Omega')$ (see Lemma 2.1), we can use Lemma 2.1 again to get

$$||z||_{H^{2}(\Omega)} \leq C ||L(\omega_{0})||_{H^{2}(\Omega)}$$
(3.3)

for some positive constant C.

On the other hand, the difference between z and w satisfies

$$\begin{cases} \nabla \cdot (\sigma + i\omega_0 \varepsilon) \nabla (z - w) &= -\nabla \cdot i(\omega - \omega_0) \varepsilon \nabla z & \text{in } \Omega, \\ z - w &= 0 & \text{on } \partial \Omega, \end{cases}$$
(3.4)

where w is defined by (3.1). Applying Lemma 2.1 one more time allows us to obtain

$$||z - w||_{H^2(\Omega)} \le C |\omega - \omega_0| ||\nabla z||_{H^2(\Omega)}.$$

This, together with (3.3), completes the proof of this lemma.

We are now in position to prove Proposition 3.1.

Proof of Proposition 3.1. Let $\Omega'' = \{x \in \Omega : \operatorname{dist}(x, \partial\Omega) > c_0/2\}$, so that $\Omega' \subset \subset \Omega'' \subset \subset \Omega$. From Lemma 2.1, $u_\omega \in W^{2,p}(\Omega'')$ for any p > 2. Hence, it follows from Sobolev embedding that $u_\omega \in \mathcal{C}^{1,\alpha}(\overline{\Omega''})$ for some $\alpha \in (0, 1)$. Thus we can consider u_ω and ∇u_ω pointwisely. We employ the ideas in [1] to prove the proposition. Since det : $\mathcal{C}(\overline{\Omega''})^{2\times 2} \to \mathcal{C}(\overline{\Omega''})$ is multilinear and bounded and

$$\begin{array}{ccc} \mathcal{O} & \to & \mathcal{C}^{1,\alpha}(\overline{\Omega''})^2 \\ \omega & \mapsto & u_{\omega} \end{array}$$

is analytic thanks to Lemma 3.1. Moreover,

$$\mathcal{O} \quad \to \quad \mathcal{C}^{0,\alpha}(\overline{\Omega''}) \\ \omega \quad \mapsto \quad \det(\nabla u_\omega)$$

is also analytic. For $x \in \Omega$, if det $A_{\omega}(x) = 0$ for every $\omega \in [\underline{\omega}, \overline{\omega}]$ then for all $\omega \in \mathcal{O}$, det $A_{\omega}(x) = 0$ by the analytic continuation theorem. In particular, det $A_0(x) = 0$ which conflicts with the choice of proper boundary conditions. Hence, we can find $\omega_x \in (\underline{\omega}, \overline{\omega})$ such that $|\det A_{\omega_x}(x)| > 0$. Moreover, since the map $|\det A_{\omega_x}(\cdot)|$ is continuous, it is strictly positive in the ball $B_{r_x}(x)$, centered at x and of radius $r_x > 0$. Noting that $\bigcup_{x \in \Omega'} B_{r_x}(x)$ covers Ω' , we can use the compactness of $\overline{\Omega'}$ in \mathbb{R}^2 to complete the proof.

From now on, a proper set of boundary conditions φ has been chosen. However, in practice, one might not know the values of frequencies and the set B_1, \dots, B_N . We thus suggest to measure the data u_{ω} for all $\omega \in (\underline{\omega}, \overline{\omega})$. The following corollary of Proposition 3.1 will be useful for the sequel. **Corollary 3.1.** If φ is a proper set of boundary conditions then we can find $\lambda > 0$ such that

$$\int_{\underline{\omega}}^{\overline{\omega}} |\det \nabla u_{\omega}(x)| dx > \lambda,$$

where $u_{\omega}(x)$ is the solution of (1.1).

4 The reconstruction method

4.1 Optimization scheme

Let the function $U_{\omega} = F[\sigma_*, \varepsilon_*; \omega]$ represent the measurement of the solution vector with σ_* and ε_* being the true distributions.

Consider

$$\begin{aligned} F: \quad \widetilde{\mathcal{S}} \times (\underline{\omega}, \overline{\omega}) &\to \quad H^2(\Omega)^2 \\ (\sigma, \varepsilon; \omega) &\mapsto \quad u_\omega - U_\omega, \end{aligned}$$

where again u_{ω} is the solution to (1.1) with a proper set of boundary conditions φ . Here \widetilde{S} is considered as a subset of the Hilbert space $H^2(\Omega)^2$. Note that F is well-defined thanks to Lemma 2.1.

To reconstruct σ and ε , we minimize the discrepancy functional

$$J[\sigma,\varepsilon] = \frac{1}{2} \int_{\underline{\omega}}^{\overline{\omega}} \|F[\sigma,\varepsilon;\omega]\|_{H^1(\Omega)}^2 d\omega$$

for $(\sigma, \varepsilon) \in \widetilde{\mathcal{S}}$.

We first investigate the differentiability of F with respect to the pair of admittivity (σ, ε) . For doing so, we need one more notation. Let $A : B = \sum_{i,j} a_{ij} b_{ij}$ for two matrices $A = (a_{ij})$ and $B = (b_{ij})$. Let \langle , \rangle_{H^s} denote the $H^s(\Omega)^2$ -scalar product for s = 1, 2. The following lemma holds.

Lemma 4.1. (i) The map F is Fréchet differentiable in $(\sigma, \varepsilon) \in \widetilde{S}$. For all $(h,k) \in S$, $DF[\sigma, \varepsilon; \omega](h,k)$ is given by the solution of

$$\begin{cases} \nabla \cdot (\sigma + i\omega\varepsilon)\nabla v_{\omega} = -\nabla \cdot (h + i\omega k)\nabla u_{\omega} & \text{in } \Omega, \\ v_{\omega} = 0 & \text{on } \partial\Omega. \end{cases}$$
(4.1)

Moreover, DF is Lipschitz continuous with respect to (σ, ε) .

(ii) J is Fréchet differentiable in $(\sigma, \varepsilon) \in \widetilde{S}$. Moreover, for all $(h, k) \in S$,

$$DJ[\sigma,\varepsilon](h,k) = \Re e \int_{\underline{\omega}}^{\overline{\omega}} \langle DF[\sigma,\varepsilon;\omega](h,k), F[\sigma,\varepsilon;\omega] \rangle_{H^{1}},$$

$$= \Re e \int_{\underline{\omega}}^{\overline{\omega}} \langle (h,k), DF[\sigma,\varepsilon;\omega]^{*}(F[\sigma,\varepsilon;\omega]) \rangle_{H^{2}},$$
(4.2)

where $DF[\sigma, \varepsilon; \omega]^*$ is the adjoint of $DF[\sigma, \varepsilon; \omega]$.

(iii) Furthermore, for all $(h, k) \in \mathcal{S}$,

$$DJ[\sigma,\varepsilon](h,k) = \Re e \int_{\underline{\omega}}^{\overline{\omega}} \int_{\Omega} (h+i\omega k) \nabla u_{\omega} : \nabla p_{\omega} \, d\omega, \qquad (4.3)$$

where $p_{\omega} \in H^2(\Omega)$ is the solution to the adjoint problem

$$\begin{cases} \nabla \cdot (\sigma + i\omega\varepsilon)\nabla p_{\omega} = \overline{F(\sigma,\varepsilon;\omega)} - \Delta \overline{F(\sigma,\varepsilon;\omega)} & in \ \Omega, \\ p_{\omega} = 0 & on \ \partial\Omega. \end{cases}$$
(4.4)

Proof. Take $(h, k) \in \mathcal{S}$ such that $(\sigma + h, \varepsilon + k)$ still belongs to $\widetilde{\mathcal{S}}$. Define

$$w_{h,k} = F[\sigma + h, \varepsilon + k; \omega] - F[\sigma, \varepsilon; \omega] \in H_0^1(\Omega)^2.$$

We have

$$\nabla \cdot (\sigma + h + i\omega(\varepsilon + k))\nabla w_{h,k} = -\nabla \cdot (\sigma + h + i\omega(\varepsilon + k))\nabla (F[\sigma, \varepsilon; \omega] + U_{\omega})$$
$$= \nabla \cdot (h + i\omega k)\nabla (F[\omega, \sigma, \varepsilon] + U_{\omega}).$$

Using Sobolev embedding and Lemma 2.1, we have

$$\begin{aligned} \|w_{h,k}\|_{H^{2}(\Omega)^{2}} &\leq C \|\nabla \cdot (h+i\omega k)\nabla (F[\sigma,\varepsilon;\omega]+U_{\omega})\|_{L^{2}(\Omega')^{2}} \\ &\leq C \Big(\|h+i\omega k\|_{L^{\infty}(\Omega')}\|F[\sigma,\varepsilon;\omega]+U_{\omega}\|_{H^{2}(\Omega')^{2}} \\ &\quad +\|\nabla (h+i\omega k)\|_{L^{4}(\Omega')^{2}}\|\nabla (F[\sigma,\varepsilon;\omega]+U_{\omega})\|_{L^{4}(\Omega')^{2\times 2}}\Big) \\ &\leq C \left(\|h\|_{H^{2}(\Omega)}+\|k\|_{H^{2}(\Omega)}\right) \left(\|F[\omega,\sigma,\varepsilon]\|_{H^{2}(\Omega')^{2}}+\|U_{\omega}\|_{H^{2}(\Omega')^{2}}\right). \end{aligned}$$

$$(4.5)$$

The function $w_{h,k} - v_{\omega} \in H_0^1(\Omega)$ and satisfies

$$\nabla \cdot (\sigma + i\omega\varepsilon)\nabla (w_{h,k} - v_{\omega}) = -\nabla \cdot (h + i\omega k)\nabla w_{h,k}.$$

Thus, again by repeating the estimates as in (4.5), we get

$$\begin{aligned} \|w_{h,k} - v_{\omega}\|_{H^{2}(\Omega)^{2}} &\leq C \left(\|h\|_{H^{2}(\Omega)} + \|k\|_{H^{2}(\Omega)}\right) \|w_{h,k}\|_{H^{2}(\Omega')^{2}} \\ &\leq C \left(\|h\|_{H^{2}(\Omega)} + \|k\|_{H^{2}(\Omega)}\right)^{2} \left(\|F[\omega,\sigma,\varepsilon]\|_{H^{2}(\Omega')^{2}} + \|U_{\omega}\|_{H^{2}(\Omega')^{2}}\right). \end{aligned}$$

Item (i) has been then proved. Moreover, it is easy to see that DF is Lipschitz continuous with respect to (σ, ε) . In fact, let (σ, ε) and (σ', ε') be in \widetilde{S} . Let (h, k) be in \mathcal{S} . Then, $DF[\sigma, \varepsilon; \omega](h, k) - DF[\sigma', \varepsilon'; \omega](h, k)$ is solution to the following equation:

$$\begin{cases} \nabla \cdot (\sigma + i\omega\varepsilon)\nabla \left(DF[\sigma,\varepsilon;\omega](h,k) - DF[\sigma',\varepsilon';\omega](h,k)\right) = \\ -\nabla \cdot (h + i\omega k)\nabla (F[\sigma,\varepsilon;\omega] - F[\sigma',\varepsilon';\omega]) \\ -\nabla \cdot (\sigma - \sigma' + i\omega(\varepsilon - \varepsilon'))\nabla DF[\sigma',\varepsilon';\omega](h,k) & \text{in }\Omega, \\ DF[\sigma,\varepsilon;\omega](h,k) - DF[\sigma',\varepsilon';\omega](h,k) = 0 & \text{on }\partial\Omega. \end{cases}$$

Therefore, applying similar estimate as in (4.5), we have

$$\begin{aligned} \|(DF[\sigma,\varepsilon;\omega] - DF[\sigma',\varepsilon';\omega])(h,k)\|_{H^{2}(\Omega)^{2}} \\ &\leq C\left(\|h + i\omega k\|_{H^{2}(\Omega)}\|F[\sigma,\varepsilon;\omega] - F[\sigma',\varepsilon';\omega])\|_{H^{2}(\Omega')^{2}} \\ &+ \|\sigma - \sigma' + i\omega(\varepsilon - \varepsilon')\|_{H^{2}(\Omega)}\|DF[\sigma',\varepsilon';\omega](h,k)\|_{H^{2}(\Omega)^{2}}\right). \end{aligned}$$

$$(4.6)$$

Since $F[\sigma, \varepsilon; \omega] - F[\sigma', \varepsilon'; \omega]$ satisfies

$$\nabla \cdot (\sigma + i\omega\varepsilon)\nabla (F[\sigma,\varepsilon;\omega] - F[\sigma',\varepsilon';\omega]) = -\nabla \cdot (\sigma - \sigma' + i\omega(\varepsilon - \varepsilon'))\nabla (F[\sigma',\varepsilon';\omega] + U_{\omega}),$$

we apply a similar estimate as in (4.5) to get Lipschitz continuity of F:

$$\|F[\sigma,\varepsilon;\omega] - F[\sigma',\varepsilon';\omega])\|_{H^{2}(\Omega')} \leq C\|\sigma - \sigma' + i\omega(\varepsilon - \varepsilon')\|_{H^{2}(\Omega)} \times \left(\|F[\sigma',\varepsilon';\omega]\|_{H^{2}(\Omega')^{2}} + \|U_{\omega}\|_{H^{2}(\Omega')^{2}}\right).$$

$$(4.7)$$

Noting that $DF[\sigma, \varepsilon; \omega](h, k)$ is the solution of (4.1), we also have

$$\|DF[\sigma',\varepsilon';\omega](h,k)\|_{H^{2}(\Omega')} \leq C\|h+i\omega k\|_{H^{2}(\Omega)}\|\left(F[\sigma',\varepsilon';\omega]\|_{H^{2}(\Omega')^{2}}+\|U_{\omega}\|_{H^{2}(\Omega')^{2}}\right).$$
(4.8)

Hence, combining estimates (4.6)-(4.8), we have

$$\begin{split} \|DF[\sigma,\varepsilon;\omega] - DF[\sigma',\varepsilon';\omega]\|_{\mathcal{L}(H^{2}(\Omega),H^{2}(\Omega))} &\leq C \|\sigma - \sigma' + i\omega(\varepsilon - \varepsilon')\|_{H^{2}(\Omega)} \\ &\times \left(\|F[\sigma',\varepsilon';\omega]\|_{H^{2}(\Omega')^{2}} + \|U_{\omega}\|_{H^{2}(\Omega')^{2}}\right). \end{split}$$

Item (ii) can be easily proved by using arguments similar to those used above. Item (iii) follows by integration by parts. $\hfill \Box$

We can now apply the gradient descent method to minimize the discrepancy functional J. We compute the iterates

$$(\sigma_{n+1}, \varepsilon_{n+1}) = T[\sigma_n, \varepsilon_n] - \mu D J[T[\sigma_n, \varepsilon_n]], \qquad (4.9)$$

where $\mu > 0$ is the step size and T[f] is any approximation of the Hilbert projection from $H^2(\Omega)^2$ onto $\overline{\widetilde{S}}$ with $\overline{\widetilde{S}}$ being the closure of \widetilde{S} (in the H^2 -norm). The derivative $DJ[T[\sigma_n, \varepsilon_n]]$ is given by

$$DJ[T[\sigma_n, \varepsilon_n]] = (-\Re e \,\nabla u_\omega : \nabla p_\omega, \omega \Im m \,\nabla u_\omega : \nabla p_\omega),$$

where u_{ω} and p_{ω} are respectively the solutions to (1.1) and (4.4) with $(\sigma, \varepsilon) = T[\sigma_n, \varepsilon_n]$.

The presence of T is necessary because $(\sigma_n, \varepsilon_n)$ might not be in $\widetilde{\mathcal{S}}$.

Using (iv) we can show that the optimal control algorithm (4.9) is nothing else than the following Landweber scheme [24, 20] given by

$$(\sigma_{n+1}, \varepsilon_{n+1}) = T[\sigma_n, \varepsilon_n] - \mu \int_{\underline{\omega}}^{\overline{\omega}} DF^*[T[\sigma_n, \varepsilon_n]; \omega](F[T[\sigma_n, \varepsilon_n]; \omega]) \, d\omega.$$
(4.10)

4.2 Initial guess

To initialize the previous optimal control algorithm, we need to construct an initial guess for the electrical property distributions σ and ϵ .

Consider the solution u_{ω} to (1.1). For all $x \in \Omega$,

$$\Delta u_{\omega} + \frac{\nabla u_{\omega}^T \nabla (\sigma + i\omega\varepsilon)}{\sigma + i\omega\varepsilon} = 0.$$

It follows that

$$A_{\omega}^{T} \frac{\nabla(\sigma + i\omega\varepsilon)}{\sigma + i\omega\varepsilon} = -\nabla \cdot A_{\omega}, \qquad (4.11)$$

where

$$A_{\omega} = \nabla u_{\omega}$$

Equation (4.11) gives us several ways to reconstruct both σ and ε . We suggest to define the map $\gamma_{\omega} = \log(\sigma + i\omega\varepsilon)$, whose imaginary part is chosen in $[0, \frac{\pi}{2})$, and solve

$$\begin{cases} \Delta \gamma_{\omega} = \nabla \cdot \left(-(\overline{A_{\omega}} A_{\omega}^{T})^{\dagger} \overline{A_{\omega}} \nabla \cdot A_{\omega} \right) & \text{in } \Omega, \\ \gamma_{\omega} = \log(\sigma_{0} + i\omega\varepsilon_{0}) & \text{on } \partial\Omega, \end{cases}$$
(4.12)

where \dagger denotes the pseudo-inverse. The knowledge of γ_{ω} implies those of σ and ε . We denote by σ_I and ε_I the obtained functions by averaging γ_{ω} over ω :

$$\sigma_I + i \frac{(\overline{\omega}) + \underline{\omega}}{2} \varepsilon_I = \frac{1}{\overline{\omega} - \underline{\omega}} \int_{\underline{\omega}}^{\overline{\omega}} e^{\gamma_\omega} d\omega,$$

where γ_{ω} is given by (4.12). We use σ_I and ε_I as the initial guess for our desired coefficients.

4.3 Convergence of the minimizing sequence

The following theorem holds.

Theorem 4.1. For all $(h, k) \in S$, we have the following estimate:

$$\int_{\underline{\omega}}^{\overline{\omega}} \|DF[\sigma,\varepsilon;\omega](h,k)\|_{H^1(\Omega)^2} d\omega \ge C \|(h,k)\|_{H^2(\Omega)^2}$$
(4.13)

for some positive constant C.

Proof. Assume to the contrary that (4.13) is not true. That means we can find h_n and k_n in S such that

$$||h_n||_{H^2(\Omega)} + ||k_n||_{H^2(\Omega)} = 1$$

and

$$\int_{\underline{\omega}}^{\overline{\omega}} \|DF[\sigma,\varepsilon;\omega](h_n,k_n)\|_{H^1(\Omega)} d\omega \to 0$$

as $n \to \infty$. By compactness, up to extracting a subsequence, we can assume that

$$(h_n, k_n) \rightharpoonup (h, k) \quad \text{in } H_0^1(\Omega)^2.$$
 (4.14)

Denote by u_{ω} the vector $F[\sigma, \varepsilon; \omega]$ and v_{ω}^n the vector $DF[\sigma, \varepsilon; \omega](h_n, k_n)$. We have

$$v^n_\omega \to 0$$
 in $H^1_0(\Omega)$

for all $\omega \in (\underline{\omega}, \overline{\omega})$.

Recall $N, B_1, \dots, B_N, \omega_1, \dots, \omega_N$, as in Proposition 3.1. Fixing $j \in \{1, \dots, N\}$, we have

$$-\nabla \cdot (\sigma + i\omega_j \varepsilon) \nabla v_{\omega_j}^n = \nabla \cdot (h_n + i\omega_j k_n) \nabla u_{\omega_j}$$

= $(\sigma + i\omega_j \varepsilon) \nabla u_{\omega_j}^T \nabla \frac{h_n + i\omega_j k_n}{\sigma + i\omega_j \varepsilon}$

in B_j . Equivalently,

$$\nabla u_{\omega_j}^T \nabla \frac{h_n + i\omega_j k_n}{\sigma + i\omega_j \varepsilon} = -\nabla \log(\sigma + i\omega_j \varepsilon) \cdot \nabla v_{\omega_j}^n - \Delta v_{\omega_j}^n.$$

Note that the left-hand side of the equation above tends to 0 in $H^{-1}(\Omega)$, so is $\nabla \frac{h_n + i\omega_j k_n}{\sigma + i\omega_j \varepsilon}$ in $L^2(B_j)$. By using Poincaré's inequality and the fact that $\overline{\Omega'} \subset \bigcup_{j=1}^N \overline{B}_j$, we arrive at h = k = 0. Since $(h_n, k_n) \in \mathcal{S}$, $\|h_n\|_{H^2(\Omega)} + \|k_n\|_{H^2(\Omega)} \to 0$, which contradicts the assumption.

Note that as a direct consequence of Theorem 4.1, it follows that

$$\left(\int_{\underline{\omega}}^{\overline{\omega}} \|DF[\sigma,\varepsilon;\omega](h,k)\|_{H^1(\Omega)^2}^2 d\omega\right)^{\frac{1}{2}} \ge C\|(h,k)\|_{H^2(\Omega)^2}$$
(4.15)

for some positive constant C. Hence, Theorem 4.1 and Proposition A.1 yield our main result in this paper.

Theorem 4.2. The sequence defined in (4.10) converges to the true admittivity $(\sigma_*, \varepsilon_*)$ of Ω in the following sense: there is $\eta > 0$ such that if $||T[\sigma_I, \varepsilon_I] - (\sigma_*, \varepsilon_*)||_{H^2(\Omega)^2} < \eta$, then

$$\lim_{n \to +\infty} \|\varepsilon_n - \varepsilon_*\|_{H^2(\Omega)} + \|\sigma_n - \sigma_*\|_{H^2(\Omega)} = 0.$$

5 Concluding remarks

In this paper we have proposed for the first time an optimal control algorithm for admittivity imaging from multi-frequency micro-electrical data. We have proved its convergence and its local stability. Our approach in this paper can be extended to elastography and can be used to image both shear modulus and viscosity tissue properties from internal displacement measurements. Another interesting problem is to image tissues with anisotropic impedance distribution from micro-electrical data.

A The convergence of the Landweber sequence with a Hilbert projection

This appendix follows from [20]; see also [21]. It proves the convergence of the Landweber scheme with a Hilbert projection.

We are interested in solving the equation

$$F[x_*;\omega] = 0 \quad \text{for all } \omega \in (\underline{\omega}, \overline{\omega}).$$
 (A.1)

It is natural to minimize

$$J[x] = \frac{1}{2} \int_{\underline{\omega}}^{\overline{\omega}} \|F[x;\omega]\|_{Y}^{2} d\omega, \qquad (A.2)$$

with $x \in K$. Assume that $F[\cdot; \omega]$ is Fréchet differentiable. So is J. The derivative of J is given by

$$DJ[x](h) = \int_{\underline{\omega}}^{\overline{\omega}} \langle DF[x;\omega](h), F[x;\omega] \rangle_Y d\omega$$

= $\int_{\underline{\omega}}^{\overline{\omega}} \langle h, DF[x;\omega]^*(F[x;\omega]) \rangle_X d\omega,$

where the superscript * indicates the dual map. The iteration sequence due to the descent gradient method is given by

$$x_{n+1} = T[x_n] - \mu \int_{\underline{\omega}}^{\overline{\omega}} DF[T[x_n]; \omega]^* (F[T[x_n]; \omega]) \, d\omega.$$
(A.3)

Here, μ is a small number and $T[x] \in K$ is an approximation of the Hilbert projection of X onto \overline{K}

$$P: X \ni x \mapsto \operatorname{argmin}\{\|x - a\| : a \in \overline{K}\}.$$
 (A.4)

Without loss of generality, we can assume that

$$||T[x_n] - P[x_n]||_X \le 2^{-n}, \quad n \ge 1.$$

The presence of T in (A.3) is necessary because x_n might not be in K and $F[x_n]$ might not be well-defined. The map T above also increases the rate of convergence of (x_n) to x_* due to

$$||T[x_n] - x_*||_X \le ||x_n - x_*||_X + 2^{-n}, \quad n \ge 1.$$
(A.5)

The following proposition holds.

Proposition A.1. Assume that $DF[x; \omega]$ is Lipschitz continuous and that, for all $x, h \in K$,

$$\int_{\underline{\omega}}^{\overline{\omega}} \|DF[x;\omega](h)\|_{Y}^{2} d\omega \ge c \|h\|_{X}^{2}.$$
(A.6)

Then the sequence defined in (A.3) converges to x_* provided that x_0 is a "good" initial guess for x_* and μ is sufficiently small.

Proof. Since $DF[x;\omega]$ is Lipschitz continuous, for all x such that $||x - x_*||_X < \eta$ with η being a small positive number, we have

$$\int_{\underline{\omega}}^{\overline{\omega}} \|F[x;\omega] - F[x_*;\omega] - DF[x;\omega](x-x_*)\|_Y^2 d\omega$$

$$\leq C\eta^2 \|x-x_*\|_X^2$$

$$\leq C\eta^2 \int_{\underline{\omega}}^{\overline{\omega}} \|F[x;\omega] - F[x_*;\omega]\|_Y^2 d\omega \qquad (A.7)$$

for some positive constant C. Note that we have used here (A.6) and the mean-value theorem for the second inequality above.

For all $n \ge 1$, let

$$\epsilon_n[\omega] = F[T[x_n]; \omega].$$

We have

$$\begin{split} \|x_{n+1} - x_*\|_X^2 - \|x_n - x_*\|_X^2 - 2^{-n} \\ &\leq \|x_{n+1} - x_*\|_X^2 - \|T[x_n] - x_*\|_X^2 \\ &= 2\langle x_{n+1} - T[x_n], T[x_n] - x_*\rangle_X + \|x_{n+1} - T[x_n]\|_X^2 \\ &\leq 2\mu \int_{\underline{\omega}}^{\overline{\omega}} \langle -DF[T[x_n]; \omega]^* \epsilon_n[\omega], T[x_n] - x_*\rangle_X d\omega \\ &+ \int_{\underline{\omega}}^{\overline{\omega}} \langle \mu \epsilon_n[\omega], \mu DF[T[x_n]; \omega] DF[T[x_n]; \omega]^*(\epsilon_n[\omega])\rangle_Y d\omega \\ &= \int_{\underline{\omega}}^{\overline{\omega}} \langle \epsilon_n[\omega], 2\mu \epsilon_n[\omega] - 2\mu DF[T[x_n]; \omega](T[x_n] - x_*)\rangle_Y d\omega - \mu \int_{\underline{\omega}}^{\overline{\omega}} \|\epsilon_n[\omega]\|_Y^2 d\omega \\ &+ \int_{\underline{\omega}}^{\overline{\omega}} \langle \sqrt{\mu} \epsilon_n[\omega], (-I + \mu DF[T[x_n]; \omega] DF[T[x_n]; \omega]^*))(\sqrt{\mu} \epsilon_n[\omega])\rangle_Y d\omega \\ &\leq 2\mu \left(\int_{\underline{\omega}}^{\overline{\omega}} \|\epsilon_n[\omega]\|_Y^2 d\omega\right)^{\frac{1}{2}} \left(\int_{\underline{\omega}}^{\overline{\omega}} \|\epsilon_n[\omega] - DF[T[x_n]; \omega](T[x_n] - x_*)\|_Y^2 d\omega\right)^{\frac{1}{2}} \\ &- \mu \int_{\underline{\omega}}^{\overline{\omega}} \|\epsilon_n[\omega]\|_Y^2 d\omega + \int_{\underline{\omega}}^{\overline{\omega}} \langle \sqrt{\mu} \epsilon_n[\omega], (-I + \mu DF[T[x_n]; \omega] DF[T[x_n]; \omega] DF[T[x_n]; \omega]) P[T[x_n]; \omega] DF[T[x_n]; \omega]) \rangle_Y d\omega \\ &\leq 2\mu \left(\int_{\underline{\omega}}^{\overline{\omega}} \|\epsilon_n[\omega]\|_Y^2 d\omega + \int_{\underline{\omega}}^{\overline{\omega}} \langle \sqrt{\mu} \epsilon_n[\omega], (-I + \mu DF[T[x_n]; \omega] DF[T[x_n]; \omega]) DF[T[x_n]; \omega] DF[T[x_n]; \omega] DF[T[x_n]; \omega] DF[T[x_n]; \omega] \rangle \right) d\omega \\ &\leq \mu (2\sqrt{C}\eta - 1) \int_{\underline{\omega}}^{\overline{\omega}} \|\epsilon_n[\omega]\|_Y^2 d\omega. \end{split}$$

Here, we have used (A.7) for the last inequality. It follows that

$$||x_{n+1} - x_*||_X^2 + \mu(1 - 2\sqrt{C\eta}) \int_{\underline{\omega}}^{\overline{\omega}} ||\epsilon_n||_Y^2 d\omega - 2^{-n} \le ||x_n - x_*||_X^2,$$

and therefore,

$$\sum_{n=1}^{\infty} \int_{\underline{\omega}}^{\overline{\omega}} \|F[T[x_n]; \omega]\|_Y^2 d\omega \le \frac{\|x_0 - x_*\|_X^2}{\mu(1 - 2\sqrt{C}\eta)} + 1.$$

We now obtain the convergence of (x_n) to x_* using again the mean-value theorem and condition (A.6):

$$c\|T[x_n] - x_*\|_X^2 \le \int_{\underline{\omega}}^{\overline{\omega}} \|DF[\tilde{x}_n;\omega](T[x_n] - x_*)\|_Y^2 d\omega = \int_{\underline{\omega}}^{\overline{\omega}} \|F[T[x_n];\omega] - F[x_*;\omega]\|_Y^2 d\omega \to 0$$

for some $\tilde{x}_n = tT[x_n] + (1-t)x_*, t \in (0,1).$

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