Solution of the Forward Problem in Magnetic-Field Tomography (MFT) Based on Magnetoencephalography (MEG)

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This paper presents the methodology and some of the results of accurate solution of the forward problem in magnetic-field tomography based on magnetoencephalography for brain imaging. The solution is based on modeling and computation of magnetic-field distribution in and around the head produced by distributed 2-D cortical and 3-D volume lead current sources. The 3-D finite-element model of the brain incorporates realistic geometry based on accurate magnetic resonance imaging data and inhomogeneous conductivity properties. The model allows arbitrary placement of line, surface, and volume current sources. This gives flexibility in the source current approximation in terms of size, orientation, placement, and spatial distribution.

Index Terms—Biomedical imaging, brain modeling, forward problem, magnetoencephalography.

I. INTRODUCTION

T
HE term magnetic-field tomography (MFT) refers to a relatively new imaging modality which involves localization and subsequent imaging of active areas in the brain by measuring the extremely weak neuromagnetic fields (10–100 fT) produced by neural currents in these areas associated with cognitive processing (magnetoencephalogram). This approach, called the magnetoencephalography (MEG) technique (recording of magnetic fields produced by electrical activity in the brain), is the only truly noninvasive method which could provide information about functional brain activity. The MFT, based on MEG data, would provide images of the brain “at work” and, as such, could have major implications for neurology and neuropsychiatry, in general, and new instrumentation for diagnosis in particular. Compared to other imaging modalities [e.g., computed tomography (CT), positron emission tomography (PET), single photon emission computed tomography (SPECT), magnetic resonance imaging (MRI)], the MEG technique is the only imaging modality that combines high temporal with high spatial resolution. Similar to any other tomographic technique, MFT involves the solution of two distinct problems: 1) the forward problem of calculating the magnetic-field distribution from known generators (sources) in the brain and 2) the inverse problem of localizing and imaging the generators by using MEG data measured around the head, and the data obtained from the forward solution. Besides, an accurate solution of the forward problem has implications for design, configuration, and placement of superconducting quantum interference device (SQUID) sensors, used to measure the neuromagnetic fields around the head, and which constitute the sensing subsystem of the MFT system. Thus, the successful solution of the inverse problem and, hence, the effectiveness of the MFT as a whole is very much dependent upon the accurate solution of the forward problem. This paper presents an accurate solution of the forward problem by using realistic brain geometry and inhomogeneous material properties coupled with various realistic source current approximations. Following from some of the previous works reported in this area, for example, in [1]–[3], the 3-D finite-element (FE) model of the brain incorporates considerable flexibility in the source-current approximation in terms of size, orientation, placement, and spatial distribution.

II. FINITE-ELEMENT MODELING OF MAGNETIC FIELDS IN MAGNETOENCEPHALOGRAPHY

A. Mathematical Model

The mathematical model of magnetic fields produced by bioelectric current sources in the brain is based on a set of quasistatic Maxwell’s equations which lead to appropriate Poisson’s equation. In doing so, it is assumed that the permeability of brain matter is the same as that of free space (μ = μ0). The quasistatic nature of the field is justified by the fact that bioelectrical activities that give rise to magnetic fields are predominantly of low frequency (from below 100 Hz to less than 1 kHz). This, together with the material properties of brain matter (e.g., conductivity σ and permittivity ε) suggest that in calculating the electric-field intensity E and magnetic flux density B vectors, the time derivative terms ∂E/∂t and ∂B/∂t in Maxwell’s equations can be ignored [4]. This leads to the following set of Maxwell’s equations:

\[ \nabla \times \mathbf{B} = \mu_0 \mathbf{J} \]  
\[ \nabla \cdot \mathbf{B} = 0 \]  
\[ \nabla \times \mathbf{E} = 0, \]  

In (1), the total current density

\[ \mathbf{J} = \mathbf{J}_p + \mathbf{J}_v \]  

where \( \mathbf{J}_p \) is the primary “excitation” current (or impressed current if at the cellular level) produced by electromotive force (emf) in the conducting brain tissue. The volume current \( \mathbf{J}_v = \)
\(\sigma E\) is attributed to the effect of the macroscopic electric field \(E\) on charge carriers [4]. It has been shown that for a realistic brain model with inhomogeneous conductivity distribution, the magnetic field from this volume current can be comparable with that from the primary current source (e.g., dipole) [5]. Thus, the total current density becomes

\[
J = J_p + \sigma E = J_p - \sigma \nabla \varphi
\]  

(5)

where \(\varphi\) is the electric scalar potential. The previous equations lead to the following Poisson’s equation for the quasistatic magnetic field, the solution of which constitutes the solution of the forward problem in MFT based on MEG [4]:

\[
\nabla \cdot (\sigma \nabla \varphi) = \nabla \cdot J_p \text{ in } \Omega = \Omega(x, y, z),
\]  

(6)

Under appropriate boundary conditions, (6) is solved for the unknown potential distribution \(\varphi = \varphi(x, y, z)\) by the finite-element method (FEM). The magnetic field \(B(\mathbf{r})\) at a given point \(\mathbf{r}\) in the problem domain \(\Omega\) is then found by using

\[
B(\mathbf{r}) = B_{0}(\mathbf{r}) - \frac{\mu_0}{4\pi} \sum_{i=1}^{m} \sigma_i \int_{C_i} \nabla \times \frac{\mathbf{R}}{\mathbf{r}^3} \, dl' 
\]  

(7)

\[
B_{0}(\mathbf{r}) = \frac{\mu_0}{4\pi} \int_{C_p} \mathbf{J}_p(\mathbf{r}') \times \frac{\mathbf{R}}{\mathbf{r}^3} \, dl'
\]  

(8)

where the source conductor consists of \(G_i\) piecewise homogeneous parts, \(B_0(\mathbf{r})\) is the magnetic field produced only by the primary current \(J_p\), \(\mathbf{R} = \mathbf{r} - \mathbf{r}'\) and \(\mathbf{r}'\) relates to the source regions.

B. Realization of the 3-D Finite-Element Model of the Brain

Considering the implications of the accuracy and quality of the solution of forward problem on the solution of the inverse problem in MEG, considerable emphasis was placed in this paper on the creation of an accurate realistic geometric and realistic material-property brain model. It is believed that this constitutes the strength of this work compared to those carried out previously.

The accurate 3-D model was constructed from 150 cross-sectional MRI slices of the brain (which are just 2 mm apart). For each slice, special editing tools were used to accurately detect edges (isolines) of white and gray matters and filter the isolines. These slices were then stacked vertically and lofted before external bounding surfaces of matters were created by using nonuniform rational B-spline (NURBS). Typically, this would, for example, create more than 30000 NURBS surfaces for the white matter which very accurately represents its intricate geometric features. These also provide additional flexibility in the FE models to easily take differences in brain geometry into account that may be encountered in realistic geometry brain models. Fig. 1 shows a schematic representation of the process described before.

The complex polysurfaces [shown in Fig. 2(a) and (b)] obtained before by NURBS surfacing are further smoothed and, if required, the number of surfaces is reduced by merging several surfaces. This is then followed by connecting these surfaces to relevant solid objects. The gray and white matters with 3-D polygonal meshes are then combined together by a solid modeler to obtain the 3-D solid model of the brain shown in Fig. 2(c). It has been found that the large number of polysurfaces of which this solid model is made can pose substantial problems in exporting these data (solid model) to an EM solver which may have limitations in manipulating such models. It is, therefore, recommended that the number of these surfaces should be minimized without compromising the accuracy of geometric representation.

The solid model of the brain obtained before is appropriately discretized, taking into account the complex nature of the brain geometry, solution accuracy needed, and the available computer
hardware for field solution. The resulting 3-D FE model of the brain used in this paper is shown in Fig. 3. A cubic air region, placed sufficiently far away from the brain itself, was used to represent the total problem domain [Fig. 3(a)]. The zero boundary condition was used on all of the surfaces of this cube. For the 3-D FE mesh shown in Fig. 3(b) and (c), tetrahedral FE elements with good aspect ratio and regularity were used to represent the brain geometry and physical properties as accurately as was practicable (defined by a tradeoff between the number of elements and accuracy of topological representation).

Table I shows some of the main parameters for a typical model used in this paper. The seemingly large number of elements (more than 1.35 m) shown is typically required to take various internal and surface geometric features of a human brain accurately into account.

TABLE I

III. Some Results and Discussions

Fig. 4 shows some of the modeling results which were particularly used to validate the accuracy and flexibility of the realistic geometry and realistic material-property brain model developed in this paper. An imaginary “sensor surface” was used...
Fig. 4. Some of the initial results. (a) 3-D brain model with imaginary “sensor surface” above the head on which some of the modeling results were plotted for comparison and validation purposes. (b) Magnetic-field distribution for a current loop placed vertically and (c) horizontally near the center of the brain. Above the head [Fig. 4(a)] on which some of the modeling results were plotted and compared with known analytical solutions for given source configurations. Two examples of such solutions are shown in Fig. 4(b) and (c). These figures show the magnetic-field distributions for a circular current loop placed near the center of the brain. The particular choice of a circular current loop (up to 10 mm in diameter carrying various currents, say 1 A for validation with known analytical solutions only) is attributable to the fact that it represents a line source with a localized surface topology. The magnitude and the distribution of the field are in good agreement with the known solution for such configurations. These and other modeling results confirm the accuracy and, to some extent, the developed flexibility of the brain model.

IV. Conclusion

It is believed that this paper presents the most accurate 3-D realistic FE model of the brain developed so far for solving the forward problem in MFT brain imaging based on MEG. With more than 1.3-m finite elements (extendable up to 5 m), this model provides flexibility in representing the complex internal features and surface topology of the human brain as well as its material inhomogeneity. It can also accommodate, with relative ease, any primary source conductor configurations in terms of its geometry and material properties. To solve the forward problem in MFT, in this paper, only the main body of the brain, in which the flow of biomagnetic currents takes place during cognitive processing, was taken into account. Hence, the skull and other “nonbrain” matters within the skull were justifiably ignored for modeling purposes since they have no influence on the magnetic-field distribution at the very low frequencies involved.

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