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# Simulation of a synchronized methodology for MR-based electromechanical property imaging during transcranial electrical stimulation

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**Introduction:** Recent investigations into the biomechanics of the brain have unveiled alteration in tissue stiffness triggered by external stimuli. For instance, visual stimulation effects can be measured in elasticity images of the cortex generated by functional magnetic resonance elastography (MRE). Such a mechanical characterization method combined with non-invasive brain stimulation (NIBS), a technique that seeks to selectively modulate particular parts of the brain using weak electrical currents, has the potential to influence research on various neurological disorders. In this *in silico* study, we aimed to elucidate individual and interdependent aspects related to a synchronized biomechanical imaging and non-invasive brain stimulation methodology. Magnetic resonance electrical impedance tomography (MREIT) was incorporated to the pipeline, providing a promising way of evaluating NIBS-induced electrical current patterns in the brain while leveraging MRE and transcranial alternating current stimulation (tACS) experimental settings.

**Methods:** A mouse head model was assembled using open-access atlases to include five anatomical structures: skin/subcutaneous tissue, skull, cerebrospinal fluid (CSF), brain white and grey matters. MRE, tACS, and MREIT experiments were simulated using Comsol Multiphysics with Matlab Livelink. Synthetic MRE and MREIT data were processed using the subzone non-linear inversion and harmonic Bz algorithm, respectively, to reconstruct images of the distributed complex shear modulus and electrical conductivity.

**Results and Discussion:** Lorentz body forces arising from simultaneous MRE and tACS elicited elastic waves of negligible amplitude compared with the extrinsic actuation levels reported in the literature, which allowed accurate reconstructions of the complex shear modulus. Qualitative electrical conductivity maps retrieved by MREIT accurately delineated anatomical regions of the brain model and could be used to recover reasonably accurate

distributions of tACS-induced currents. This multi-physics approach has potential for translation to human brain imaging, and may provide more possibilities for the characterization of brain function together than in isolation.

#### KEYWORDS

MRE, MREIT, tACS, finite elements, simulation, inverse problem

## 1 Introduction

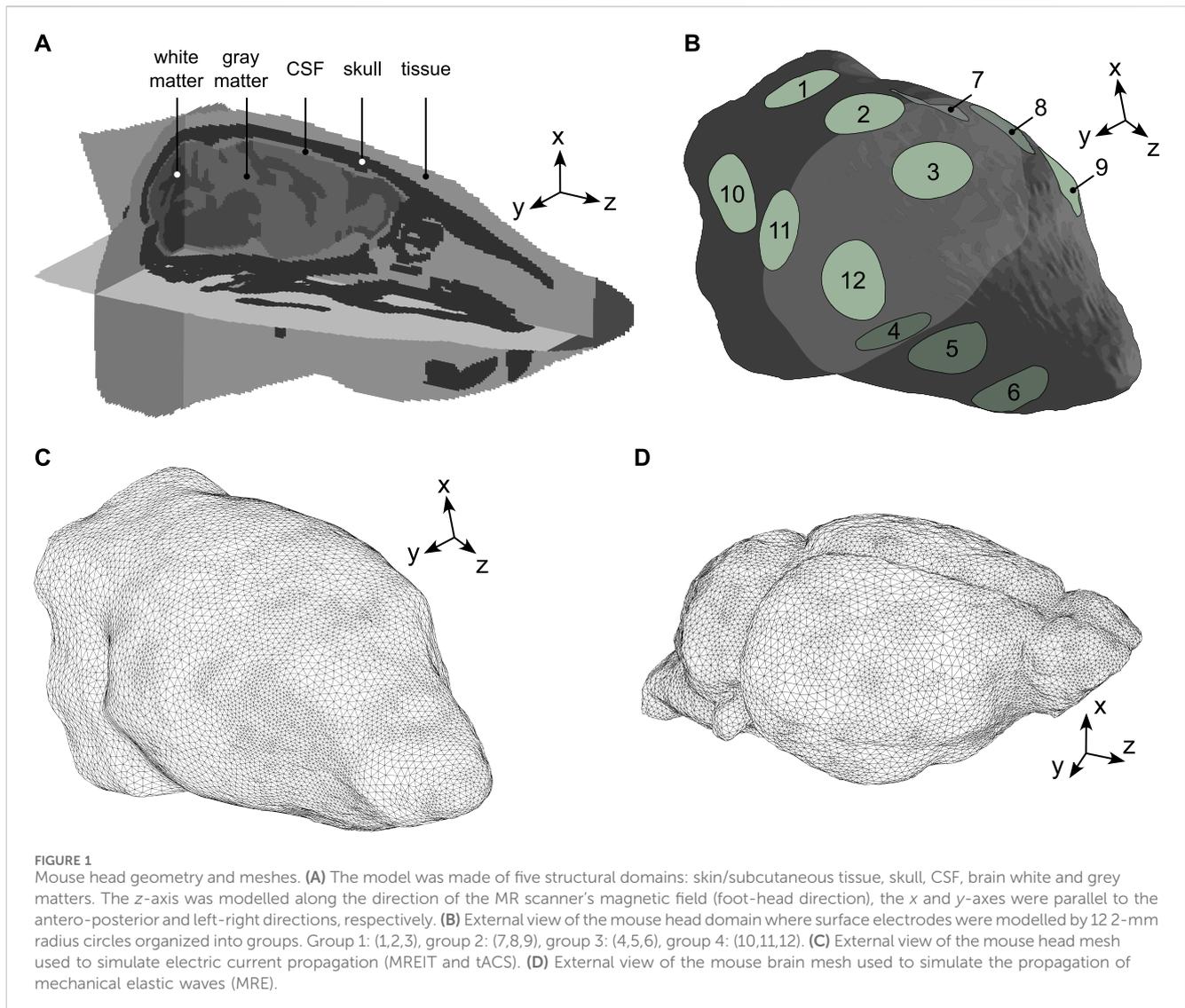
Application of non-invasive brain stimulation (NIBS) in neuroscience research has drawn significant attention for investigating psychiatric and neurological disorders, as well as the neuronal mechanisms underlying behavior and cognition. NIBS modulates specific cerebral regions by inducing electric fields within the brain. These fields are generated either through magnetic coils in the case of transcranial magnetic stimulation, or via direct placement of electrodes on the scalp. Since the inception of NIBS, significant research has focused on the intricate interplay between brain function and behavior. As such, various experimental strategies have been devised to investigate individual responses to electrical stimulation and evaluate the physiological impact of brain stimulation sites on vision [1], audition [2], motor function [3–5], somatosensation [6], language [7, 8], attention [9, 10], memory [11, 12], reasoning [13, 14], decision making [15–17], and social behavior [18–20], as reviewed by Polania *et al.* [4]. In addition to modulating brain function, recent research indicates that non-invasive brain stimulation may also have implications for treating neurological disorders. Notably, alterations in cognitive functions observed in individuals with Alzheimer's disease subsequent to a non-invasive brain stimulation session lend credence to the hypothesis that neuroplasticity can be modulated as a strategy for symptom reduction [21–25]. Likewise, NIBS was observed to enhance motor functions in cases of Parkinson's disease, for which no cure currently exists [26–30]. The effectiveness of NIBS treatments on depressive disorders has also been investigated. Despite encouraging results, evidence of beneficial impacts remains sparse and more research is needed [31, 32].

Transcranial magnetic stimulation (TMS) and transcranial electrical stimulation (TES) are the main techniques used to modulate cerebral activity [4]. TMS is based on remote electromagnetic induction of weak electric current loops in the conductive brain tissue using a distant coil supplied with short electric pulses. The temporal profile of the stimulation can be shaped by applying a single or multiple pulses at adjustable repetition rates. Due to the focused nature of the induced electric field, TMS tends to provide more localized stimulations than TES, but has a lower penetration depth, which restricts the stimulation site to shallow cortical regions. In comparison with TMS, TES delivers weak electrical currents through electrodes directly placed on the scalp. The temporal profile of the electrical stimulation determines the TES variant: transcranial direct current stimulation (tDCS), transcranial alternating current stimulation (tACS), and transcranial random noise stimulation (tRNS). In all cases, electrode design and positioning govern the size and location of the stimulated area [33], which is often confined to the cortex as the electric field disperses in deeper tissues. Transcranial temporal interference electrical stimulation (tTIS) is a recent technology that seeks to overcome this limitation by applying two alternating currents oscillating at nearly identical frequencies resulting in constructive and destructive electric field

interferences. The interference pattern has a narrow spatial support allowing the recruitment of interior neurons while preventing stimulation of the surrounding area. This method has been demonstrated in mice [34] and further evaluated by simulating electric field distributions in numerical models of the human brain derived from Magnetic Resonance Imaging (MRI) [35]. Applications to hippocampal activity modulation in humans are currently investigated [36]. Finally, transcranial focused ultrasounds have been presented as an emerging technique that complements TES and TMS by its higher spatial resolution and ability to target the deep brain [37].

Overall, published NIBS guidelines underline the importance of combining neuroimaging and computational modelling to optimize NIBS protocols, achieve finer stimulation spatial selectivity, and interpret results [4]. However, the subject-specific electrical conductivity distribution needed to evaluate the stimulation current patterns in the brain is challenging to measure. Magnetic resonance electrical impedance tomography (MREIT) aims at providing such estimates of the ohmic conductivity through the analysis of the magnetic flux density induced in the imaged tissue as an electric flux passes [38–41]. While electric property characterization and computational modelling do not supersede physiological validation, they are seen as tools that may help better understand the brain's response to NIBS and identify areas that might have been unintentionally activated in the process.

In addition to NIBS, changes in the biomechanics of the brain provide insights into local functional activity [42]. Such measurements are performed with magnetic resonance elastography (MRE) [43], which images non-invasively dynamic tissue deformations in organs [44]. The motion fields used to reconstruct the mechanical model parameters, often the shear modulus, can be induced intrinsically by the cardiac pressure pulse [45, 46] or extrinsically using mechanical surface actuators [47]. Extrinsic MRE (exMRE) offers the possibility of studying the brain's dynamic response across a range of mechanical actuation frequencies, which helps stabilize numerically the elastography inversion [48] or investigate viscoelastic dispersive effects [49]. For instance, 50-Hz exMRE was used to establish a standard-space atlas of the viscoelastic properties of the human brain [50]. On the lower end of the actuation frequency spectrum, intrinsic MRE operates at the physiological rate of heart pulsation and eliminates the need for external actuators [51]. The tissue mechanical response at such low frequencies, around 1 Hz, is better described by poroelastic models, where the fluid and solid components of the biological medium are represented by separate phases [45, 52, 53]. Intrinsic MRE notably demonstrated sensitivity to stiffness variations in distinct regions of the visual cortex during visual stimulation [42], establishing a proxy to study in physiological operating conditions the biomechanical response of brain tissue to neuronal activity. In similar functional MRE studies, contrasts observed in images of biomechanical features correlated with specific tasks (motor or cognitive) participants were requested to perform in the protocol. For instance, memory performance was



related to tissue biomechanics (shear stiffness [54] and damping ratio [55–57] in the subfields of the hippocampus). Likewise, motor function performances correlated with damping ratio variations in the hippocampus for aerobic exercise tasks [57] and with global brain shear stiffness variations for dynamic balance related tasks [58]. Recently, functional MRE of the mouse brain was used to capture the neuronal activity triggered by fast electrical hind limb stimulation, suggesting rapid changes in brain elasticity [59].

Intrinsic neuronal currents arising from natural neuronal activity manifest a Lorentz force upon exposure to the magnetic field of the magnetic resonance (MR) system. This observation led researchers to develop Lorentz effect imaging (LEI) for visualizing brain activity [60]. The fundamental assumption is that the small distance moved by electrically-active tissue, driven by the Lorentz force, results in a phase shift in the MR signal that can be used for precise localization of neuronal activity. Initial results of brain activity detection using LEI presented by Truong et al. [61–63] were further analyzed by Roth et al. [64–66], who suggested that the size of the Lorentz force-induced displacements would be order of magnitudes too weak to be detected by MRI.

Overall, the growing number of research endeavours aimed at illuminating brain function using MRE and non-invasive brain stimulation has examined biomechanical and bioelectrical properties distinctly. In this study, we anticipate that the individual facets of NIBS, ohmic conductivity, and viscoelasticity mapping hold promise to merge into a valuable multi-physics MR-based tool for neurological disorder characterization. Using simulations in a numerical model of a mouse head, we review and elucidate the underlying principles of the methodology.

## 2 Methods

### 2.1 Mouse head model

A well-established pipeline was used to design our numerical simulations. This pipeline involves developing a geometry from anatomical image segmentation and assigning physical properties to the corresponding segments [67–70]. A mouse head model was constructed from open-access CT and MRI mouse scans and consisted

of five anatomical structures: skin/subcutaneous tissue, skull, cerebrospinal fluid (CSF), brain white and grey matters. Head contours were segmented from the Digimouse atlas [71], the skull from a surgical atlas [72], and white and grey matter domains from a brain template [73]. Individual segments were then registered to the Digimouse coordinate system and the CSF was represented by the space between the brain and the skull. Handling of mouse scans and domain registration were performed using the medical image processing application ITK-SNAP [74] and exported in NiFTI format to Matlab (Matlab R2019b, The Mathworks, Massachusetts, United States) for a segment-based definition of material physical properties (see Sections 2.2, 2.4 for details). The final mouse head model had a resolution of  $120 \mu\text{m}^3$  and is shown in Figure 1A.

## 2.2 MREIT

Typical brain MREIT pipelines induce a direct low-amplitude electric current density in the head,  $\mathbf{J}(\mathbf{r})$ , and images the resulting magnetic flux density's  $z$  component,  $B_z(\mathbf{r})$ . The internal current density is generated by delivering an external current to the scalp through surface electrodes operated by a waveform generator synchronized with the MRI pulse sequence. Four electrodes are usually used and placed to ensure a suitable current coverage as well as generate at least two linearly independent current distributions in the imaged domain. The Biot-Savart law relates the tomographic direct current density in the imaged slices to the induced magnetic flux density's  $z$  component:

$$B_z(\mathbf{r}) = \frac{\mu_0}{4\pi} \int_{\Omega} \frac{J_x(\mathbf{r}')(y - y') - J_y(\mathbf{r}')(x - x')}{|\mathbf{r} - \mathbf{r}'|^3} d\mathbf{r}', \quad (1)$$

where  $\mu_0$  is the magnetic permeability of the free space, which is assumed to be the same as that of the material. The volume integral in Eq. 1 can be expressed as a convolution in the physical space [38, 75]:

$$B_z(\mathbf{r}) = \mu_0(G_y * J_x - G_x * J_y)(\mathbf{r}), \quad (2)$$

where  $G_x = \frac{1}{4\pi} \frac{x}{|\mathbf{r}|^3}$  and  $G_y = \frac{1}{4\pi} \frac{y}{|\mathbf{r}|^3}$  are scalar arrays, and  $\mathbf{r}$  is the vector position, which has its origin at the origin of the coordinate system. The convolution in Eq. 2 turns into a product in the  $k$ -space:

$$B_z(\mathbf{r}) = \mu_0 \mathcal{F}^{-1} \{ \mathcal{F}\{G_y\} \cdot \mathcal{F}\{J_x\} - \mathcal{F}\{G_x\} \cdot \mathcal{F}\{J_y\} \}(\mathbf{r}), \quad (3)$$

where  $\mathcal{F}$  denotes the Fourier transform. The  $k$ -space formulation allows a faster evaluation of  $B_z(\mathbf{r})$  from current densities than the evaluation of the volume integral. The current density distribution is given by the Ohm's law:

$$\mathbf{J} = \sigma \mathbf{E} = -\sigma \nabla V, \quad (4)$$

where  $\sigma$  is the electric conductivity and  $V$  is the electric potential, which is a solution to following boundary value problem:

$$\begin{cases} \nabla \cdot (\sigma \nabla V) = 0 & \text{in } \Omega, \\ I = \int_{\varepsilon^+} \sigma \frac{\partial V}{\partial \mathbf{n}} ds = - \int_{\varepsilon^-} \sigma \frac{\partial V}{\partial \mathbf{n}} ds, & \nabla V \times \mathbf{n} = 0 \text{ on } \varepsilon^+ \cup \varepsilon^-, \\ \sigma \frac{\partial V}{\partial \mathbf{n}} = 0 & \text{on } \partial\Omega \setminus \varepsilon^+ \cup \varepsilon^-, \end{cases} \quad (5)$$

where  $\partial\Omega$  represents the surface bounding the computational domain  $\Omega$ ,  $\varepsilon^+$  and  $\varepsilon^-$  represent the load and grounded electrodes, and  $\mathbf{n}$  is the unit vector normal to  $\partial\Omega$ .

Current densities in MREIT experiments were simulated in the whole mouse head by solving the forward problem described by Eqs 4, 5 using the electric current module of Comsol Multiphysics with Matlab Livelink (Comsol Inc. Stockholm, Sweden). The induced magnetic flux density was computed with Matlab via Eq. 3. The computational domain was defined in Comsol by lofting the mouse head transverse  $(x, y)$  contours along the  $z$ -axis. Internal domains were modelled by distributing material properties across the geometry. Electric conductivity values were set to  $0.465 \text{ S.m}^{-1}$  in the skin,  $0.01 \text{ S.m}^{-1}$  in the skull,  $1.654 \text{ S.m}^{-1}$  in the CSF,  $0.126 \text{ S.m}^{-1}$  in the white matter, and  $0.275 \text{ S.m}^{-1}$  in the gray matter [76]. Relative permittivity and permeability were defined globally and set to a homogeneous value of 1 across the entire domain. Then, a set of 12 2-mm radius circular surface electrodes was designed using Comsol's Computer Aided Design (CAD) tools. Electrodes were organized into four groups of three units evenly spaced along the  $z$ -axis on opposing sides of the mouse scalp, and placed to cover to whole brain region. An external view of the mouse head domain with the electrodes is shown in Figure 1B. The computational mesh, shown in Figure 1C, contained 1,283,014 tetrahedral elements with minimum and maximum target sizes of  $2.4 \times 10^{-4} \text{ m}$  and  $5.0 \times 10^{-4} \text{ m}$ , respectively. Current and magnetic flux densities were computed for six MREIT numerical experiments, where different combinations of electrode groups were used. Each experiment involved one electrode group for current delivery (1 mA in total) and one grounded electrode group. The six combinations were (group 1, group 3), (group 2, group 4), (group 4, group 1), (group 3, group 4), (group 2, group 3), and (group 2, group 1), where the first group of each couple indicates the loading electrodes and the second group the grounded electrodes.

The inverse problem of MREIT was solved using the *iterative harmonic  $B_z$*  method [77, 78]. This technique is based on the observations that  $B_z$  is sensitive to conductivity contrasts in directions perpendicular to the tomographic current densities and insensitive in parallel directions. For direct currents, the magnetic flux density, the conductivity of the medium, and the electric potential are related through [79]:

$$\nabla^2 \mathbf{B} = \mu_0 \nabla \sigma \times \nabla V. \quad (6)$$

The  $z$  component of Eq. 6 is then given by:

$$\nabla^2 B_z = \mu_0 \left( \frac{\partial \sigma}{\partial x} \frac{\partial V}{\partial y} - \frac{\partial \sigma}{\partial y} \frac{\partial V}{\partial x} \right). \quad (7)$$

Considering the six MREIT experiments, Eq. 7 can be expressed in matrix form as:

$$\underbrace{\begin{bmatrix} \frac{\partial V_1}{\partial y} & -\frac{\partial V_1}{\partial x} \\ \vdots & \vdots \\ \frac{\partial V_6}{\partial y} & -\frac{\partial V_6}{\partial x} \end{bmatrix}}_{\mathbf{u}} \underbrace{\begin{Bmatrix} \frac{\partial \sigma}{\partial x} \\ \frac{\partial \sigma}{\partial y} \end{Bmatrix}}_{\mathbf{s}} = \frac{1}{\mu_0} \underbrace{\begin{Bmatrix} \nabla^2 B_{z,1} \\ \vdots \\ \nabla^2 B_{z,6} \end{Bmatrix}}_{\mathbf{b}}. \quad (8)$$

The following iterative reconstruction procedure is defined from Eq. 8:

$$\mathbf{U}_k \mathbf{s}_{k+1} = \mathbf{b}, \quad (9)$$

which allows evaluating the  $x$  and  $y$  components of the conductivity gradient. The actual conductivity distribution is finally recovered

from its gradient in Eq. 9 through the following Poisson's equation [77]:

$$\begin{cases} \tilde{\nabla}^2 \sigma_{k+1} &= \tilde{\nabla} \cdot \mathbf{s}_{k+1} & \text{in } \Omega_t, \\ \sigma_{k+1} &= \sigma_e & \text{on } \partial\Omega_t, \end{cases} \quad (10)$$

where  $\tilde{\nabla}$  indicates 2D in-plane derivatives in the slice  $\Omega_t$  and  $\sigma_e$  represents the conductivity on the external boundary of the domain. The harmonic  $B_z$  method uses an initial conductivity distribution  $\sigma_0$  to start the reconstruction procedure and evaluate the matrix termed  $\mathbf{U}$  in Eq. 8. A homogeneous distribution  $\sigma_0 = \sigma_e = 0.465 \text{ S.m}^{-1}$  was used and corresponded to the conductivity on the head domain boundary indicated in Eq. 10, which was assumed to be experimentally measurable. The inversion was conducted over ten iterations. For meshing, size constraints were imposed to enforce refinement in specific regions. Target sizes were set to  $2.4 \times 10^{-4} \text{ m}$  for elements in the vicinity of expected material discontinuities, based on simulated  $B_z$  data, and to  $1.0 \times 10^{-3} \text{ m}$  for elements away from material discontinuities, where low  $B_z$  variations were observed. The mesh contained 940,023 tetrahedral elements.

## 2.3 tACS

For tACS simulations, the same mouse head model as in the MREIT forward model was used. Two numerical experiments were conducted using two two-electrode combinations. This time, electrodes were used in isolation rather than in groups and the combinations were (electrode 2, electrode 11) and (electrode 11, electrode 5), where the first member of each electrode pair indicates the loading electrode and the second member the grounded electrode. Electrode 2 was placed above the cortex, electrode 11 was facing the caudate putamen, and electrode 5 was on the neck, opposing electrode 11. Electrode combinations were chosen with the aim of generating Lorentz body forces (BF) of maximum amplitude. This was achieved by selecting loading and grounded electrodes with the same  $z$  coordinate, thus resulting in current densities with main directions lying in  $(x, y)$  planes and producing Lorentz BF,  $\mathbf{f}_L$ , with main directions perpendicular to both the current density,  $\mathbf{J}$ , and the MR scanner's magnetic flux density,  $\mathbf{B}_0$ , according to:

$$\mathbf{f}_L = \mathbf{J} \times \mathbf{B}_0. \quad (11)$$

The alternating current frequency was set to 900 Hz. In practice, tACS has been operated across a range of frequencies from less than 1 Hz to a few kilohertz [80–82]. Here, the specific value of 900 Hz was chosen to match a typical actuation frequency in mouse brain MRE experiments [83], which allowed evaluating the impact of tACS induced motion, through Lorentz BF, on the MRE reconstruction procedure. Electric conductivities were the same as in the MREIT simulations and were assumed frequency independent in the investigated frequency range [84].

## 2.4 MRE

In MRE, the mechanical properties of soft solids are reconstructed from the analysis of mechanical elastic waves induced in the imaged tissue and tracked with MRI [85]. The elastic waves are low frequency (up to about 100 Hz in humans and to about 1.5 kHz in mice) and are

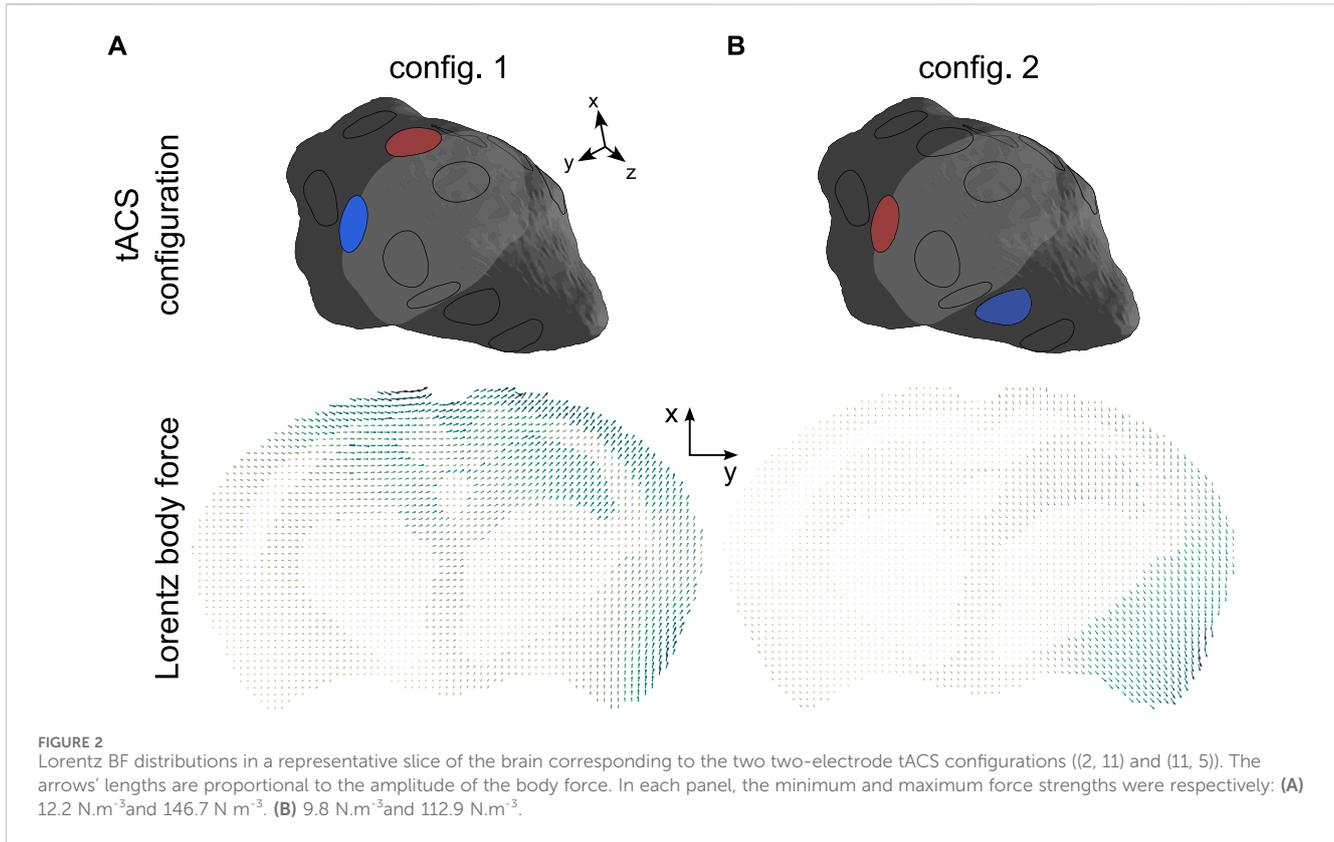
induced by external actuators or are naturally present in the body from heart pulsations. Assuming that soft brain tissues are heterogeneous, viscoelastic, nearly incompressible, and isotropic materials, the response of the tissue undergoing a harmonic actuation at an angular frequency  $\omega$  is described by the following boundary value problem with displacement variables:

$$\begin{cases} \nabla \cdot (\mu \nabla \mathbf{u}) + \nabla (\mu \nabla \cdot \mathbf{u} - p) + \mathbf{f}_b = -\rho \omega^2 \mathbf{u} & \text{in } \Omega, \\ K \text{tr}(\boldsymbol{\varepsilon}) = -p & \text{in } \Omega, \\ \mathbf{u} = \mathbf{u}_e & \text{on } \Gamma_u, \\ \mathbf{n} \cdot \boldsymbol{\sigma}_e = \mathbf{f}_e & \text{on } \Gamma_\sigma, \end{cases} \quad (12)$$

where  $\mu$  is the complex shear modulus (Pa),  $\mathbf{u} = [u, v, w]^T$  is the complex displacement vector (m),  $p$  is the pressure field (Pa),  $\mathbf{f}_b$  is the body force density ( $\text{N.m}^{-3}$ ),  $\rho$  is the mass density ( $\text{kg.m}^{-3}$ ),  $\omega$  is the actuation angular frequency ( $\text{rad.s}^{-1}$ ),  $\Omega$  is the brain domain,  $K$  is the bulk modulus (Pa),  $\boldsymbol{\varepsilon}$  is the strain tensor,  $\mathbf{u}_0$  is the displacement vector on the domain's boundary  $\Gamma_u$ ,  $\mathbf{n}$  is the unit vector normal to the surface boundary  $\Gamma_\sigma$ ,  $\boldsymbol{\sigma}_e$  is the stress tensor, and  $\mathbf{f}_e$  is the traction force vector. In standard MRE experiments, the scanned tissue is usually free of body forces and the corresponding  $\mathbf{f}_b$  term in Eq. 12 is neglected in the formulation of the inverse problem of elastography. When MRE and tACS are conducted simultaneously, Lorentz BF develop in the brain, as described by Eq. 11, and the assumption of a negligible source term must be evaluated.

The solid mechanics module of Comsol Multiphysics was used to conduct the MRE simulations in the brain domain of the head model detailed in Section 2.1. First, a stereolithography-format (STL) brain mesh was generated with Matlab using the iso2mesh toolbox [86, 87] and imported in Comsol to create the geometry object. Then, internal domains were modelled by distributing the mechanical properties across the brain geometry. Shear modulus values were set to  $5397 + i1836 \text{ Pa}$  in the white matter and  $4997 + i1426 \text{ Pa}$  in the grey matter domains [83]. A homogeneous density and incompressible bulk modulus were defined and set to  $1,000 \text{ kg.m}^{-3}$  and  $2.2 \times 10^9 \text{ Pa}$  [88], respectively. A heterogeneous mesh was generated with mesh size gradients to resolve the material discontinuities. A target size of  $2.2 \times 10^{-4} \text{ m}$  at material boundaries and an overall maximum size of  $3.0 \times 10^{-4} \text{ m}$  were imposed. An external view of the final mesh, containing 400,888 tetrahedral elements, is shown in Figure 1D. In a first numerical experiment, a displacement field was simulated in the brain from the application of Dirichlet boundary conditions (BC) ( $5.0 \times 10^{-6} \text{ m}$  displacements along the  $x$ -axis) to mimic a typical mouse brain MRE experiment, free of body forces [89]. In a second numerical experiment, Lorentz BF-induced displacement fields were simulated using the tACS current distributions described in Section 2.3 and a magnetic flux density of 11.7 T oriented along the  $z$ -axis of the scanner. Such a high field value was used to depict a situation in which Lorentz body force effects are maximised. The Lorentz BF distributions corresponding to the two two-electrode configurations of Section 2.3 were computed according to Eq. 11 with  $\|\mathbf{B}_0\| = 11.7 \text{ T}$ . In Eq. 12, this translates to  $\mathbf{f}_b = \mathbf{f}_L$ . In both experiments, the actuation frequency (through prescribed displacements or Lorentz forces) was 900 Hz [83].

The inverse problem of elastography was addressed using a specialized non-linear inversion (NLI) technique [90, 91]. NLI identifies mechanical property distributions, the shear modulus in the present work, that is a best-fit of the measured displacements  $\mathbf{u}_m$  through Eq. 12, with  $\mathbf{f}_b = 0$ . Finite element solutions of this equation,  $\mathbf{u}_e$ , are computed using sets of shear modulus distributions,



successively enhanced with property updates across 100 global iterations in order to minimize the mismatch between  $\mathbf{u}_c$  and  $\mathbf{u}_m$  through the following objective function:

$$\Phi = \frac{1}{2} (\mathbf{u}_c(\boldsymbol{\theta}) - \mathbf{u}_m)^H (\mathbf{u}_c(\boldsymbol{\theta}) - \mathbf{u}_m), \quad (13)$$

where  $\boldsymbol{\theta}$  is the vector containing the inferred mechanical properties and  $H$  is the complex-conjugate transpose. The most likely approximation of the true shear modulus is the distribution that minimizes the objective function (Eq. 13). The particularity of this NLI formulation is the decomposition of the total imaged domain into overlapping *subzones* processed individually and in parallel. Thus, the inverse problem consisting in identifying relevant property updates is solved at the subzone level, which mitigates the computational cost associated to the 3D problem. In this numerical study, the measured displacements  $\mathbf{u}_m$  were the displacements simulated with Comsol in the mouse head model. In the inverse problem, the solution to the boundary value system in Eq. 12 (with  $\mathbf{f}_b = 0$ ) was computed on a 27-node hexahedral finite-element mesh. Displacement and property meshes had the same resolution of 120  $\mu\text{m}^3$ . Subzones had an isotropic size of 2.1 mm with a 20% overlap. Property updates in each subzone were computed using the conjugate gradient method (CG) and a global property distribution in the total domain was generated by assembling the subzones at the end of each global iteration. Gaussian smoothing with a kernel size of 84  $\mu\text{m}$  was applied to the property distribution at the subzone assembly step in order to stabilize the reconstruction process. The initial shear modulus distribution used to start the reconstruction process was set to 6000 +  $i$ 2000 Pa. This initial modulus was chosen to lie outside the range of viscoelasticity values used in the mechanical model of the brain.

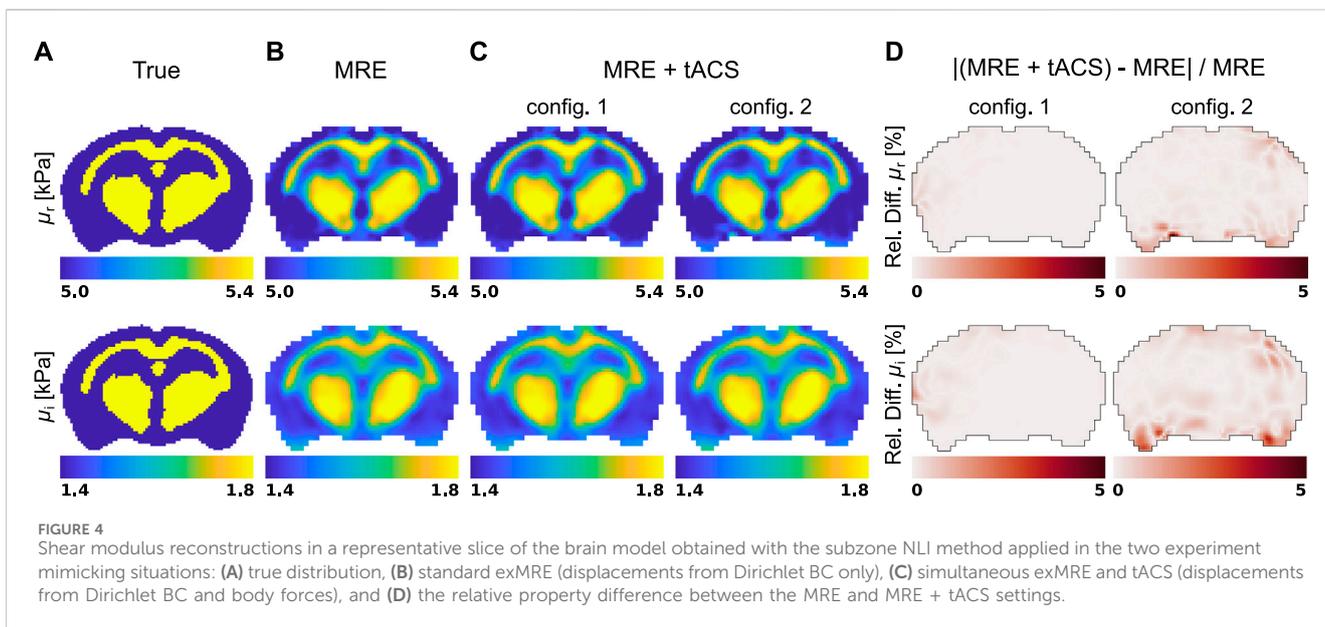
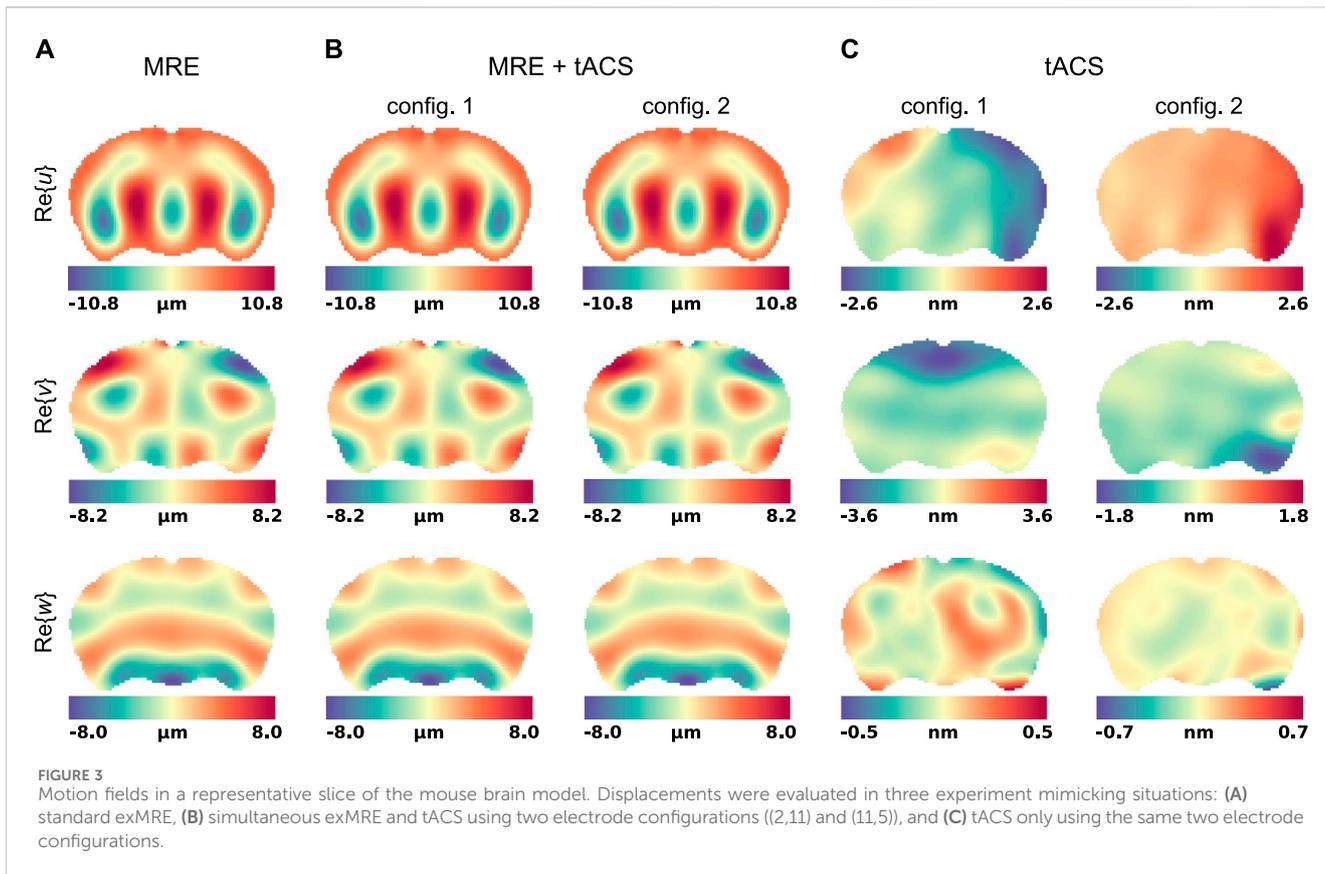
Three displacement data types were processed with NLI. First, the body force free displacement field resulting from prescribed displacements on boundaries and mimicking standard exMRE. Second, the superposition of the Lorentz BF induced displacements and the body force free displacements, mimicking simultaneous exMRE and tACS experiments. Third, Lorentz BF induced displacements alone, mimicking a tACS experiment.

### 3 Results

Two MR-assisted tACS sessions were considered, representing the successive use of two distinct electrode sets (configurations 1 and 2). The corresponding distributions of the tACS-induced Lorentz body forces, simulated by solving Eqs 5 and 11 in the two two-electrode configurations, are displayed in Figure 2.

When MRE and tACS are performed simultaneously, the Lorentz BF shown in Figure 2 induce motions that superimpose with the displacement generated by the surface actuators and contribute to the total field encoded by the MR system's gradients. The real part of the finite element displacement fields, solution to Eq. 12 in the exMRE, simultaneous exMRE and tACS, as well as tACS only experiments, are presented in Figure 3.

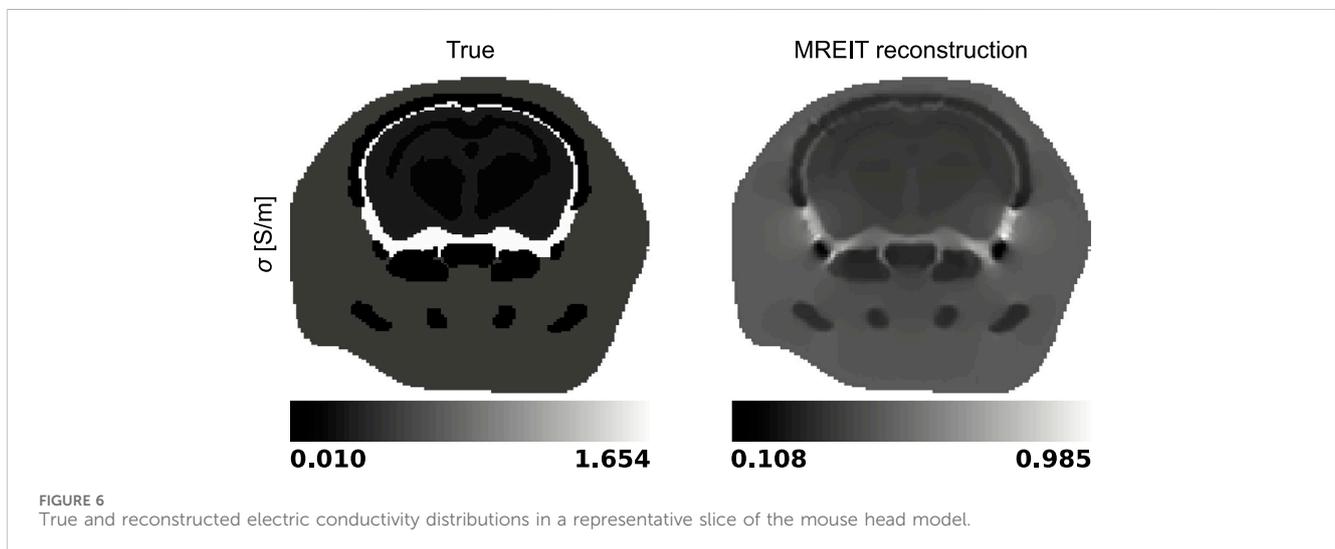
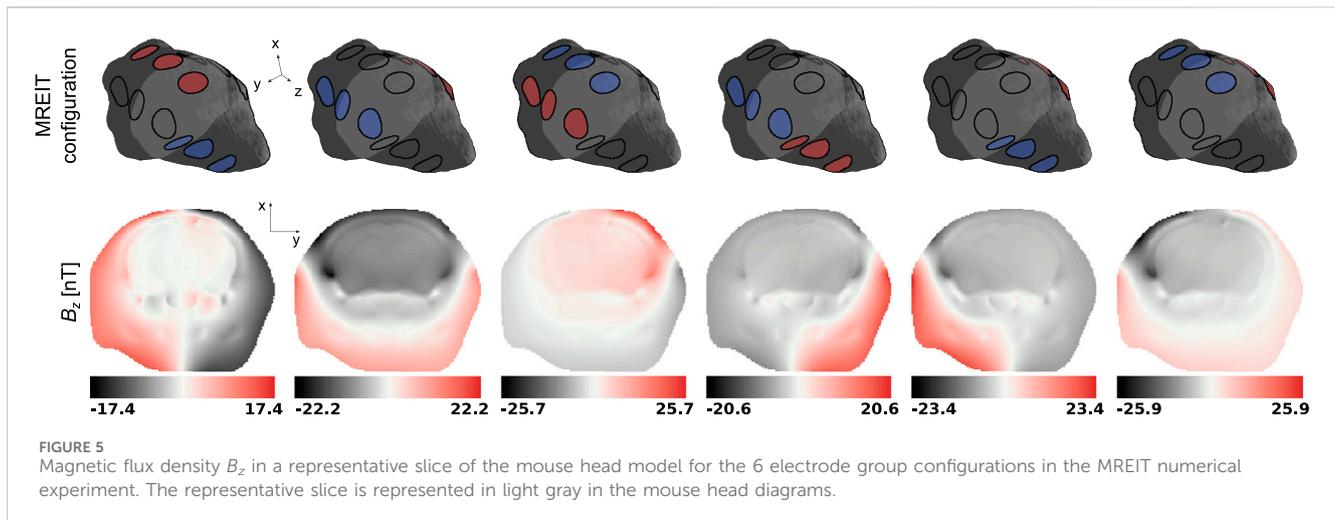
The five displacement fields considered in Figure 3 were used to reconstruct viscoelasticity images of the brain with the subzone NLI algorithm. Figure 4 shows the corresponding storage and loss moduli,  $\mu_r$  and  $\mu_i$ , respectively, that form the viscoelastic complex shear modulus  $\mu = \mu_r + i\mu_i$  in the MRE and MRE + tACS experiments, as well as the relative property differences observed between the two settings.



MR electrical impedance tomography was introduced in the simulation pipeline to reconstruct the electrical conductivity distribution,  $\sigma(x, y, z)$ , from images of the magnetic flux density,  $B_z(x, y, z)$ , induced by weak direct currents applied to the mouse scalp. Images of  $B_z$  corresponding to the six MREIT series were

computed using the  $k$ -space formulation of the Biot-Savart law, given by Eq. 3, and are shown in Figure 5, along with the corresponding electrode configurations.

Then, the six  $B_z$  distributions and electric fields evaluated at each iteration of the harmonic  $B_z$  algorithm were combined into a global



system of equations, Eq. 7, where the electric conductivity was the unknown to be solved for. The true and reconstructed conductivity distributions are shown in a representative slice in Figure 6.

Finally, the three components of the electric current densities corresponding to the two electrode configurations and evaluated using the true and reconstructed conductivity distributions are presented in Figure 7. The last row of each panel indicates the relative reconstruction error.

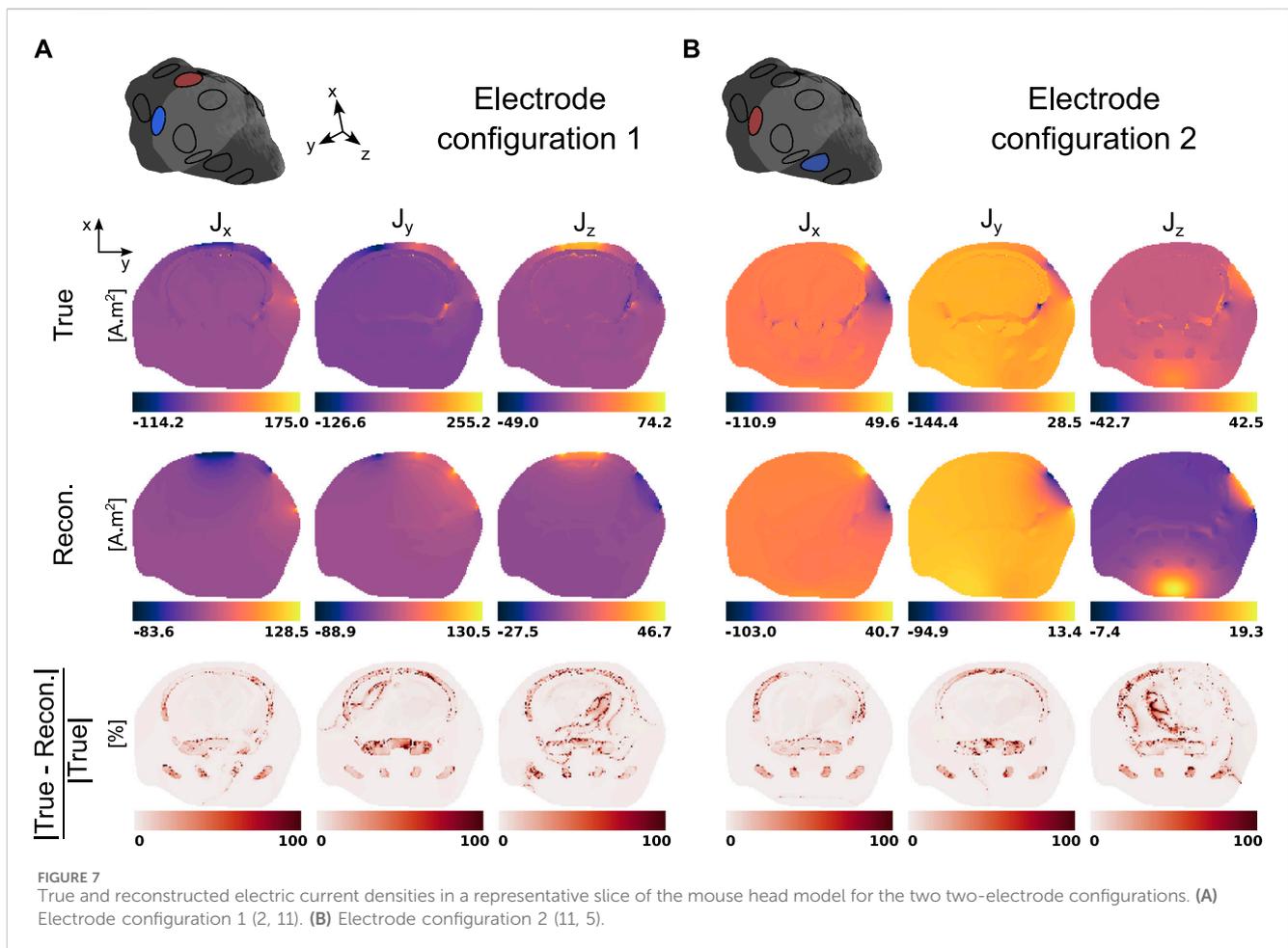
## 4 Discussion

We detailed a methodology incorporating NIBS, MRE, and MREIT that shows potential in neuroscience to advance the study of complex cerebral electromechanical interactions [59, 92]. MRE non-invasively probes the biomechanics of the brain, which is an undisputed biomarker of tissue health [93–96], and a proxy for brain activity characterization according to recent research [42, 54, 59]. NIBS has historically attempted to reduce symptoms linked to neurological disorders [97, 98], and has increasingly used head conductivity models to map resulting electrical current patterns

[99], which MREIT has the potential to provide [38, 100]. This work reviewed each technique and simulated the pipeline as we envision it in an experimental settings.

The amplitudes and relative orientations of the NIBS currents and the magnetic field of the MR system govern the size of the Lorentz BF shown in Figure 2. In our simulations, a typical current intensity of 1 mA was used, in agreement with a previous TES study [34]. The unusual choice of 11.7 T for modelling the MR's magnetic flux density was made with the aim of investigating the effects of maximised Lorentz BF on the elastography pipeline. This high field scenario is challenging experimentally (MRE [101, 102], MREIT [103, 104]) and presents limited benefits for our methodology compared with more accessible lower field systems used for human (1.5, 3, 7 T) and small animal imaging (7 T). The absence of Lorentz BF-induced effects in the elastography reconstructions at 11.7 T suggests that MRE and NIBS imaging may be performed in humans at lower field strengths, as will be discussed below.

The displacements generated in an MRE acquisition are represented in Figure 3 and show the respective responses of the brain model to surface actuation and Lorentz BF. Although the interior patterns of displacements induced by pure Dirichlet BC also depend on the biomechanics of the material, care was taken to model realistic



conditions (displacement field [59, 105, 106] and viscoelasticity values [83]). Figure 3 showed that pure Dirichlet BC-induced motions dominate the Lorentz BF contribution, which consistently reproduces the order of magnitude predicted by Roth et al. (5–13-nm motion amplitude) [64, 66]. Under such conditions, the NLI method applied to displacement fields dominated by Dirichlet BCs (Dirichlet BC and Dirichlet BC + Lorentz forces) achieved quantitative reconstructions for both storage and loss moduli, as shown in Figures 4A, B, although the storage modulus dominated the loss modulus in all the anatomical domains, similarly to [107]. In these images, the blurred contours at material interfaces occurred because of the intrinsic limitation of elastography reconstructions when the size of the domain to identify is smaller than half of the induced wavelength [108]. From detection and reconstruction perspectives, these results suggest that exMRE and tACS may be performed simultaneously. In intrinsic MRE, the relatively large motion amplitudes due to the cardiac pressure pulse [109] indicate that combined MRE + tACS imaging would also be possible using intrinsic activation [51], although care would need to be taken regarding the tissue's response rate to electrical stimulation relative to the cardiac frequency and MR motion encoding periods. Resolving the Lorentz BF-induced motion would be an elegant approach to MRE + tACS, however the low amplitude of the displacements shown in Figure 3C would likely be obscured from detection within tACS safety limits using current MR systems and sequences, especially at more available 1.5 T, 3 T, and 7 T systems, where weaker magnetic fields produce weaker Lorentz BF. In

addition, the NLI reconstructions based on Lorentz BF-induced motions were unsuccessful due to the data-model mismatch arising from the presence of unaccounted body forces ( $\mathbf{f}_b = 0$  in Eq. 12).

MREIT originally used the harmonic  $B_z$  algorithm to reconstruct electrical conductivity distributions. This approach has demonstrated a good ability to differentiate structures with impedance contrasts in the presence of data with high signal-to-noise ratios. Figures 5, 6 confirmed this observation and showed recovered distributions of the electric conductivity consistent with the material structure. However, the harmonic  $B_z$  method failed to achieve quantitative conductivity estimates, which has also been observed in other published works [110, 111]. The potential of electrical conductivity as a biomarker of pathological tissues has been investigated [112–116], but thus far, MREIT has mostly developed into a tool used for studying the brain's response to electrical stimulation using modelling. The difficulty in accurately measuring the weak signals generated by MREIT currents explains the limited application of conductivity mapping using MR, although improvements in pulse sequence design have recently been reported [41]. Figure 7 illustrates a typical objective of MREIT consisting in evaluating current density distributions in the imaged brain using the reconstructed electrical conductivity. Higher error levels in the current density reconstructions are found at the interface and within segments of lower conductivity, which reflects the limitations of the harmonic  $B_z$  algorithm in such regions. The additional errors visible

in  $J_z$  images, the  $z$  component of the current density, are likely due to the tomographic nature of the reconstruction process.

In practice, NIBS, MRE, and MREIT may be performed through repeated scans within a single imaging session using an all-in-one experimental settings. The acoustic waves in MRE are usually transmitted through a vibrating device placed below or in front of the mouse [105, 106], and electric currents are administered using adjustable scalp electrodes. A challenge in exMRE + tACS experiments may arise from the stability of the setup during acquisitions. Electrical wires as well as head electrodes are experiencing oscillating Lorentz forces and should thus be firmly attached to a fixed head holder. The tACS currents were assumed to have a negligible effect on the MRE acquisition as MRE sequences often involve bipolar scans where the displacement field is sampled twice with opposite MEG polarities, thus canceling out persistent electric current related phase artifacts [117]. Finally, the eddy currents resulting from the oscillation of the conductive brain tissue in a magnetic field are second order effects and were not considered in this work.

## 5 Limitations

Our study incorporates simplifications and limitations, which warrant further discussion and contextualization. First, five main anatomical domains were integrated in the model to keep the simulation reasonably convenient to manipulate. Although a more detailed representation of structural constituents would better approximate real tissues and further challenge the reconstruction aspects, the proposed model was sufficiently detailed to tackle the study's specific inquiries. Then, electrodes were modelled as simple disks on the surface of the mouse head to which a normal current density was applied. Whereas this setting does not physically model a metal electrode in contact with skin tissues, it does approximate the MREIT experimental condition where recessed electrodes are used in order to avoid radio-frequency shielding issues related to metal objects placed nearby the imaged field of view. Finally, the same frequency was used for both the tACS current and the exMRE actuation. This characteristic was not crucial in our simulations since the contribution of the tACS-induced motion to the total displacement field were negligible. In ultimate experimental investigations aimed at segregating these contributions, it is worth noting that our simulated settings would be similar to a mono-frequency actuation with a full-waveform encoding scheme where the frequency of the MEG is specifically chosen to match that of the actuation, leading to the acquisition of both Lorentz BF and external actuator induced displacements at 900 Hz. If the mechanical actuator and the tACS system were operating at different frequencies, each motion contribution may be investigated separately by selecting the MEG frequency accordingly.

## 6 Conclusion

Numerical simulations of MR elastography and MR electrical impedance tomography in the context of MR-assisted transcranial electrical stimulation of the mouse brain were presented. Results

suggest that simultaneous tACS and exMRE would be experimentally viable. Optimization of NIBS necessitates precise evaluation of the tACS current flow patterns and subject-dependent electrical conductivity models, which the added MREIT session provides although improvements in the reconstruction method are needed. The benefits of multi-physics (and multi-modality) imaging approaches have been addressed previously in the context of breast cancer, where MRE, electrical impedance tomography, microwave-imaging, and near-infrared spectroscopy were seen as complementary to each other within a global diagnostic and computational framework [118]. The unification of electrical stimulation along with the monitoring of mechanical and electrical brain responses may provide greater neuronal activity mapping and brain function characterization possibilities together than in isolation.

## Data availability statement

The original contributions presented in the study are included in the article, further inquiries can be directed to the corresponding authors.

## Author contributions

GF: Conceptualization, Formal Analysis, Methodology, Software, Validation, Writing—original draft. EV: Software, Supervision, Writing—review and editing. GG: Conceptualization, Writing—review and editing. GC: Funding acquisition, Resources, Supervision, Writing—review and editing.

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## Conflict of interest

GG was employed by Philips Healthcare Canada.

The remaining authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

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