



Computational and Phantom-Based Feasibility Study of 3D dcNCI With Ultra-Low-Field MRI

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Specialty section:

This article was submitted to
Medical Physics and Imaging,
a section of the journal
Frontiers in Physics

Received: 29 December 2020

Accepted: 18 March 2021

Published: 26 April 2021

Citation:

Höfner N, Storm J-H, Hömmen P,
Cassarà AM and Körber R (2021)
Computational and Phantom-Based
Feasibility Study of 3D dcNCI With
Ultra-Low-Field MRI.
Front. Phys. 9:647376.
doi: 10.3389/fphy.2021.647376

The possibility to directly and non-invasively localize neuronal activities in the human brain, as for instance by performing neuronal current imaging (NCI) via magnetic resonance imaging (MRI), would be a breakthrough in neuroscience. In order to assess the feasibility of 3-dimensional (3D) NCI, comprehensive computational and physical phantom experiments using low-noise ultra-low-field (ULF) MRI technology were performed using two different source models within spherical phantoms. The source models, consisting of a single dipole and an extended dipole grid, were calibrated enabling the quantitative emulation of a long-lasting neuronal activity by the application of known current waveforms. The dcNCI experiments were also simulated by solving the Bloch equations using the calculated internal magnetic field distributions of the phantoms and idealized MRI fields. The simulations were then validated by physical phantom experiments using a moderate polarization field of 17 mT. A focal activity with an equivalent current dipole of about 150 nAm and a physiologically relevant depth of 35 mm could be resolved with an isotropic voxel size of 25 mm. The simulation tool enabled the optimization of the imaging parameters for sustained neuronal activities in order to predict maximum sensitivity.

Keywords: ultra-low-field magnetic resonance imaging, neuronal current imaging, current dipole phantom, MEG, simulation

1. INTRODUCTION

The realization of ultra-low-field (ULF) magnetic resonance imaging (MRI), i.e., MRI in the μ T-regime, was possible when new, very sensitive superconducting quantum interference devices (SQUID) sensors could be used to detect nuclear spin precession [1, 2], and, within a few years, several ULF scanners have been developed [3–7]. As the magnetization of the examined volume scales with the surrounding magnetic field, ULF MRI has to cope with a lower signal-to-noise ratio (SNR) compared to high-field MRI—typically employing fields on the order of 1 T and above. Despite different measures to mitigate this low SNR, e.g., pre-polarization, only comparably large voxel sizes in the mm-range could be reached until now in *in vivo* applications. However, the possibility of using ULF MRI for functional imaging is promising motivating the development effort albeit with some compromises regarding spatial resolution. Consequently, imaging of neuronal magnetic fields via their interaction with the spin population, coined neuronal current imaging (NCI), was soon investigated by attempting to directly detect the influence of biomagnetic signals on the phase and relaxation of the nuclear spin precession in the ultra-low-field regime [8].

The successful implementation of NCI would be of particular significance, as currently available non-invasive methods for imaging brain function, such as functional MRI (fMRI), magnetoencephalography (MEG), or electroencephalography (EEG) suffer from either poor temporal resolution or error-prone spatial localization accuracy [9]. The possibility of NCI via high-field MRI was studied intensively applying different approaches [10–12]. Unfortunately, the results were either inconclusive due to a lack of sensitivity of the MRI signal to evoked neuronal activity [13] or irreproducible because of the dominating blood oxygen level-dependent (BOLD) effect [14]. In contrast, owing to the low magnetic field, ULF MRI has the advantage that the BOLD effect becomes negligible [15].

Several phantom studies have explored possible implementations of NCI using ULF MRI, and have highlighted the necessary technical and methodical improvements [16–19]. These studies suggest two different approaches. The so-called *AC* or *resonant mechanism* proposes the use of the magnetic field generated by the neuronal currents itself to manipulate the spin dynamics. This condition would be best met by tuning the Larmor frequency to match the main frequencies of the neuronal activity (resonant condition) [16–18, 20] and the MR signal would be produced exclusively by the proton spins in close proximity to the source. The second approach, called the *DC mechanism*, aims at measuring the minuscule phase perturbation generated by the very weak neuronal magnetic field [16–18]. One possible implementation is based on the acquisition of MR images in the presence and absence of evoked activity. A difference image would then reveal the effect of the neuronal magnetic field. The effectiveness of the two approaches has been demonstrated in multiple phantom studies; however, a fundamental lack of sensitivity has to be overcome for *in vivo* applications. Körber et al. [18] showed that for sufficiently long-lasting neuronal activities, the *DC mechanism* could feature larger contrast-to-noise ratio (CNR) than the *AC mechanism*. As different simulation studies on NCI using high-field MRI

showed, the specific spatial and temporal characteristics of the neuronal currents affect the measurable phase and relaxation changes [20–25].

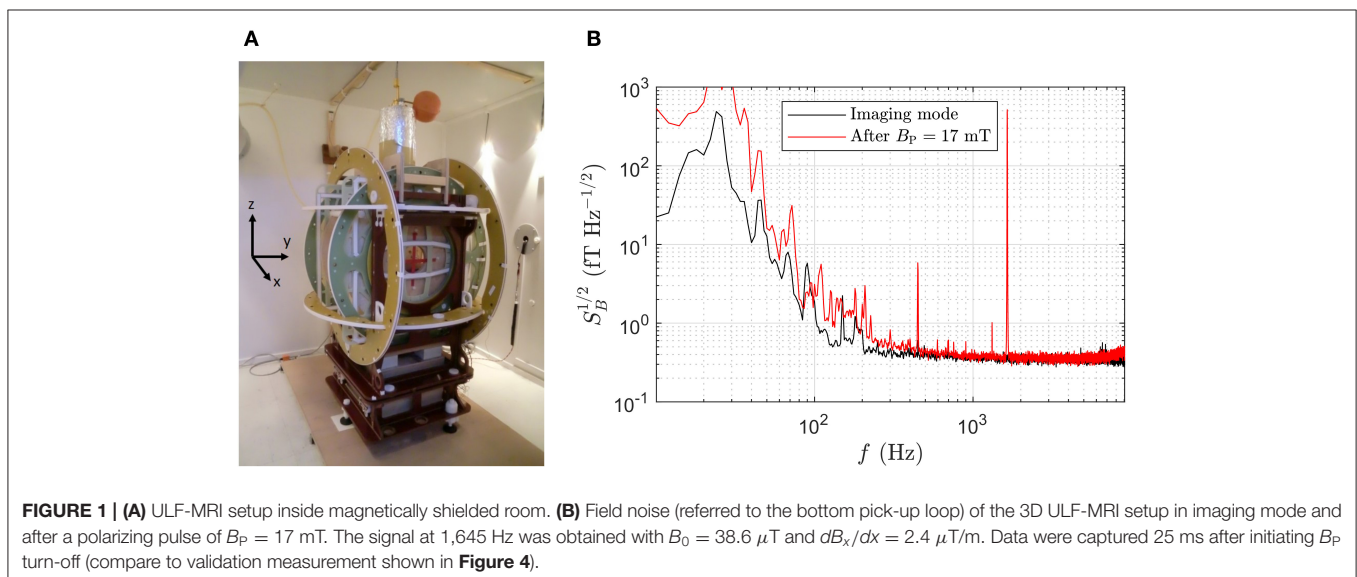
Motivated by this, and expanding on our earlier phantom experiments [17, 18, 26], we developed a computational framework that can be used to simulate 3-dimensional (3D) dcNCI sequences based on ULF MRI. It is centered on our low-noise ULF-MRI hardware and on a well-characterized long-lasting, monophasic neuronal activity which can be evoked by median nerve stimulation. We strove for current dipole phantoms, which could be manufactured with a well-defined geometry facilitating the simulation of their internal and external fields. Measuring their far-field distribution using a multi-channel SQUID system allowed the determination of their equivalent current dipole (ECD) strength for a given applied current. The simulated internal field distributions served as input to an analytic MR solver to characterize the dcNCI signature of the phantoms for multiple configurations and to optimize the imaging sequence for the existing experimental 3D ULF-MRI setup. The validated framework was then used to assess the feasibility and requirements toward *in vivo* dcNCI.

2. MATERIALS AND METHODS

2.1. 3D Ultra-Low-Field MRI Hardware

The experimental phantom study was performed using low-noise ULF-MRI hardware [27] at the Physikalisch-Technische Bundesanstalt (PTB), Berlin, as shown in **Figure 1A**. The system deploys resistive room temperature MRI coils and is located inside a moderately magnetically shielded room (MSR) which features a residual field of <1.5 nT after degaussing [28].

Helmholtz coils generate the homogeneous detection field \mathbf{B}_0 along x and the $\pi/2$ -pulse ($\mathbf{B}_{\pi/2}$) along the y -direction, while a Maxwell coil provides the frequency encoding gradient $G_F = dB_x/dx$ and two orthogonal biplanar gradient coils generate the phase encoding gradients $G_{Ph} = dB_x/dy$ and dB_x/dz . A



self-shielded coil provides the homogeneous polarizing field B_P along x while simultaneously reducing transient effects from the MSR. The coils, with the parameters given in **Table 1**, are powered by low-noise current sources placed outside the MSR that are disconnected during data acquisition, if possible.

For signal detection we use our single-channel ultra-low-noise SQUID system that is operated inside the low intrinsic noise dewar LINOD2 [29]. A SQUID current sensor is inductively coupled to a superconducting pick-up coil designed as an axial second-order gradiometer with an overall baseline of 125 mm and a diameter of 45 mm. Since a longitudinal gradient is applied for frequency encoding, this configuration and a very low-noise current source with $470 \text{ pA Hz}^{-1/2}$ driving the G_F coil (active during data acquisition) are necessary to reduce noise contributions to a negligible level. A transverse gradient can be used to relax these requirements. The gradiometer dimensions also maximize the SNR with respect to the sustained neuronal source, which was estimated to have a distance of about 50 mm (source-scalp distance 35 mm, see below) to the lower pick-up coil of the gradiometer [30]. When the single-channel SQUID system is operated within the 3D ULF-MRI setup in imaging mode, i.e., with B_0 and G_F coils powered, the total field noise (referred to the bottom pick-up loop) is about $380 \text{ aT Hz}^{-1/2}$ for frequencies above 1 kHz and does not increase after application of the polarizing pulse (see **Figure 1B**).

TABLE 1 | Coil parameters of the PTB ULF-MRI system.

Coil	B/I or G/I
$B_0(x)$, $\mu\text{T/A}$	48
$B_{\pi/2}(y)$, $\mu\text{T/A}$	4.8
$B_P(x)$, mT/A	1.0
$G_F(dB_x/dx)$, $\mu\text{T/(Am)}$	242
$G_{Ph}(dB_x/dy)$, $\mu\text{T/(Am)}$	65.8
$G_{Ph}(dB_x/dz)$, $\mu\text{T/(Am)}$	61.4

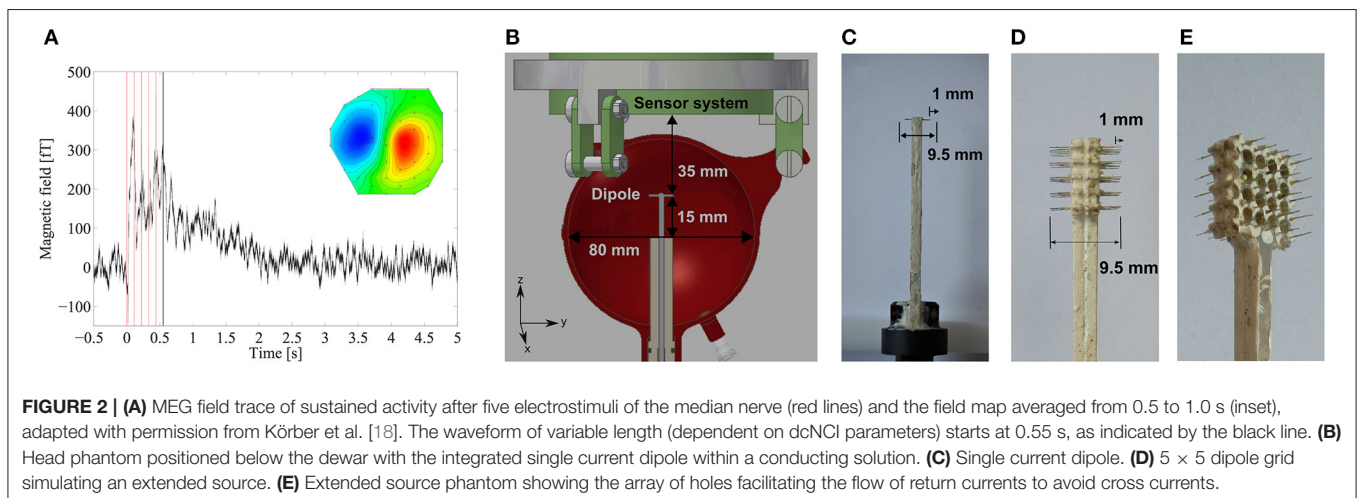
B/I and G/I denote the field-to-current and gradient-to-current ratio, respectively.

2.2. Current Dipole Phantoms

Current dipole phantoms mimic the underlying focal neuronal activity, which is a sustained, monophasic response following electrostimulation of the median nerve. Its characterization was carried out in previous MEG studies, as shown in **Figure 2A** [18]. The activity is located at a depth of about 35 mm to the scalp surface within the somatosensory cortex SII with a maximum ECD_{max} of about 50 nAm. A time-stable dipolar field indicates a focal activity which decays over about 1 s after stimulation. The derived waveform of variable length, depending on the dcNCI parameters, starts after the last stimulus which excludes artifacts in a potential *in vivo* experiment.

In **Figure 2B**, the phantom positioned below the ULF-MRI SQUID system is shown with the dipole at a depth of 35 mm corresponding to the depth of the physiological model. The NCI phantom with an 80 mm diameter is filled with a CuSO_4 and NaCl solution featuring the conductivity $\sigma_{\text{sol}} = 0.325 \text{ S/m}$ and the ULF-MR parameters of brain tissue with $T_1 = T_2 = 100 \text{ ms}$ [31]. A gap of 10 mm between the phantom and the dewar simulates the skull which has a negligible MR signal.

The scalable current dipole concept is based on commercially available printed-circuit-boards (PCBs), allowing a precise fabrication of various arrangements and numbers of current dipoles. The electrodes are made from insulated Pt wire (diameter $125 \mu\text{m}$) with a total length of 9.5 mm. At both wire-ends 1 mm of the insulation is removed. Two phantoms were fabricated: a single current dipole consisting of two electrodes pointing apart, as shown in **Figure 2C**, and a dipole grid with 5×5 current dipoles arranged in a square lattice with a dipole spacing of 2 mm, as shown in **Figures 2D,E**. If the grid dipoles were mounted on a solid, continuous plate, a slightly higher current would flow in the outside electrodes, since the shorter return path through the conducting medium leads to a lower resistance. Therefore, the mounting plate of the grid was assembled with holes to minimize this effect. Simulations described below also confirmed that this suppresses cross currents between electrodes on each side of the grid. The minute additional potential drop from the base to the tip of the electrode due to the current is practically identical for all individual electrodes.



Currents of up to $80 \mu\text{A}$ were applied using a home-built, battery-powered, voltage-controlled current source enabling a faithful reproduction of arbitrary waveforms. The phantoms were calibrated by measuring the external magnetic field map for a known drive current with 97 sensors of the 304-channel SQUID system of PTB [32]. Those were close to the phantom and more distant sensors were not used, since they would contain mostly noise and hence negligible signal information. Performing an equivalent current dipole reconstruction allowed the determination of the effective dipole length.

2.3. 3D Ultra-Low-Field MRI Pulse-Sequence

The 3D ULF-MRI sequence was a gradient echo sequence and is displayed in **Figure 3**. The polarizing field B_p of 17 mT,

applied for 500 ms (t_p) including a 150 ms ramp, is turned off adiabatically over 15 ms (t_{off}) until it is aligned with the permanent detection field B_0 of $38.6 \mu\text{T}$. A $\pi/2$ -pulse of length 2.43 ms (t_{RF}) initiates the precession of the magnetization M . Then, one frequency encoding gradient G_F (dB_x/dx) and two phase encoding gradients G_{Ph} (dB_x/dy), G_{Ph} (dB_x/dz) are applied for the spatial encoding of the three dimensions (k -space filling). For anatomical imaging, t_{ph} should be as short as possible to minimize signal loss due to relaxation. A t_{ph} of 30 ms was used, as in our previous *in vivo* imaging [27]. Detecting up to four consecutive echoes leads to an overall measurement time of about 800 ms, including the polarization period. A pause of 2 s between the individual measurements avoids overheating of the uncooled polarizing coil. **Table 2** summarizes the imaging parameters.

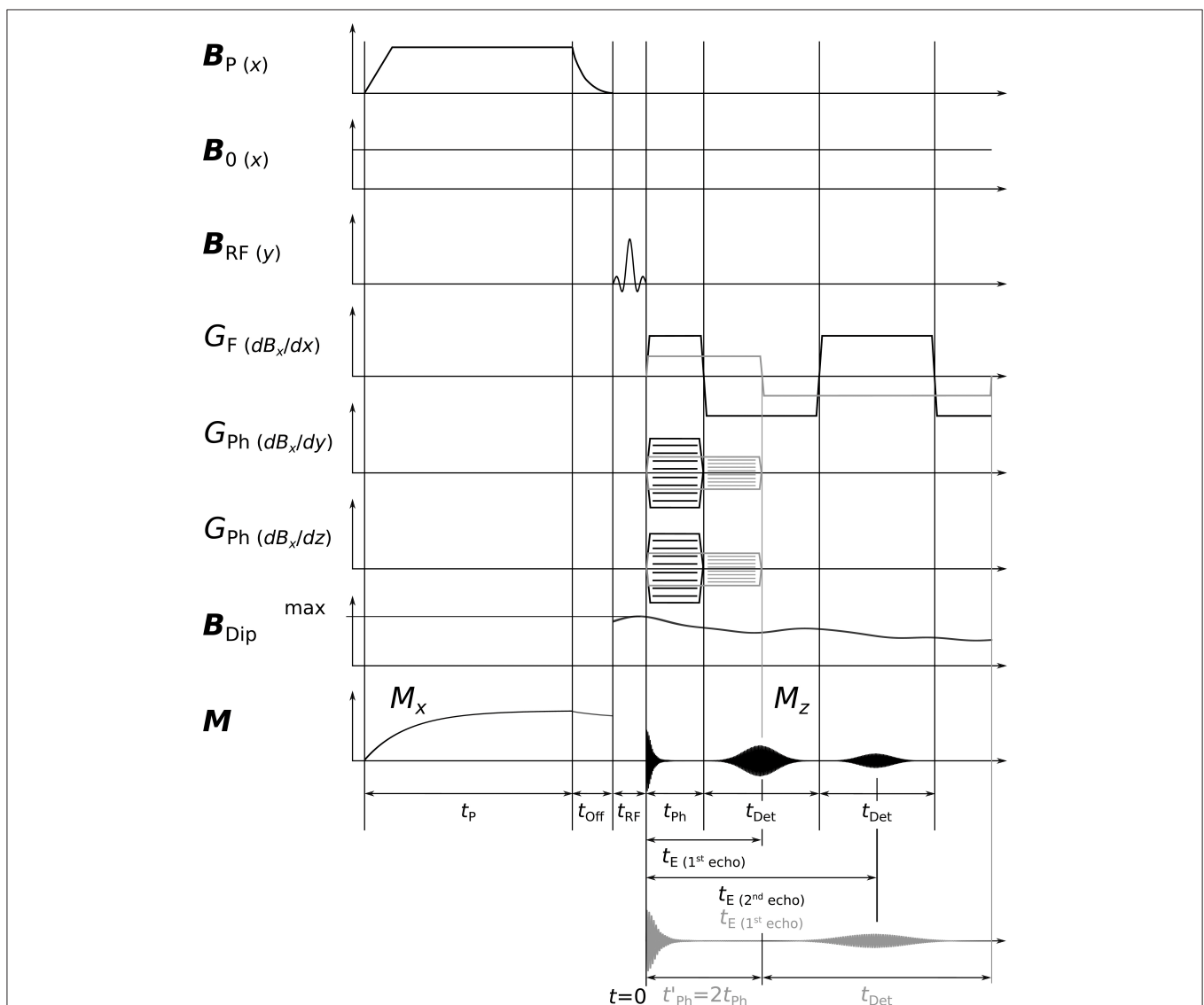


FIGURE 3 | Schematic 3D ULF-MRI sequence. After adiabatic turn-off of B_p and a subsequent $\pi/2$ -pulse, phase encoding gradients G_{Ph} and a frequency encoding gradient G_F are utilized and echoes are generated by reversing G_F . The dipole field B_{Dip} is applied permanently during the entire encoding period. Exemplary echoes are shown for phase times t_{ph} and $t_{\text{ph}}' = 2t_{\text{ph}}$ (gray color). In the latter, the gradient strengths are halved giving the same k -space coverage. Time periods t are differently scaled.

TABLE 2 | Imaging parameters for a phase time $t_{\text{Ph}} = 30$ ms and sampling frequency of 20 kHz resulting in isotropic voxel sizes.

Voxel size (mm)	Phase steps $N_y \times N_z$	FOV _{y,z} (cm)	G _F ($\mu\text{T/m}$)	G _{Ph,max} ($\mu\text{T/m}$)
5	20 × 20	10	78.2	74.7
10	20 × 20	20	39.1	37.2
15	10 × 10	15	26.1	23.5
20	10 × 10	20	19.6	17.6
25	10 × 10	25	15.6	14.1

All voxel sizes were used for the simulation and 25 mm for the experimental study. FOV_{y,z} denotes the field-of-view in the y or z-direction, respectively.

In order to guarantee maximum phase accumulation, and therefore the largest detectable signal, the dipoles are oriented perpendicularly to \mathbf{B}_0 . In this way, \mathbf{B}_{Dip} are parallel and anti-parallel to \mathbf{B}_0 just above and below the dipole, respectively. The applied dipole field, as shown in **Figure 3**, was derived from the measured temporal amplitude profile of a somatosensory evoked long-lasting activity. \mathbf{B}_{Dip} affects the precession during the spatial encoding and the readout of the echo signal causing phase and frequency shifts. The expectedly small frequency changes in the millihertz range [33] are visualized using a difference image technique. During one measurement the dipole field \mathbf{B}_{Dip} is active (on). A second measurement serves as a reference in which the dipole is not driven and $B_{\text{Dip}} = 0$ (off). Subtracting the two data sets (on–off) in the time domain reveals the affected ^1H -spins in the difference image. An additional reference measurement (off′) allows screening for artifacts by calculating a difference image (off–off′) which should not show a residual above the noise level.

The measurement of the very small B_{Dip} of several hundred pT imposes high demands on the 3D ULF-MRI setup, requiring a high SNR, as well as a high stability of the applied magnetic fields. In order to reduce the influence of magnetic field drifts due to current drifts in the detection field coil or changes of the background field within the magnetically shielded room, we alternate the order of the measurements with and without the applied dipole field. Nevertheless, low-frequency drifts mainly from the current source driving the detection coil lead to measured equivalent frequency shifts of up to ± 27 mHz for successive measurements. This causes artifacts in the difference image, masking the influence of the dipole field. To correct for these, the current through the detection coil was simultaneously measured by acquiring the voltage across a precise sense resistor, $R_S = 1\Omega$, for subsequent phase correction. At a Larmor frequency of 1,645 Hz the maximum frequency shifts correspond to ~ 16 ppm (parts per million), which translate to a current change of $13 \mu\text{A}$ in the detection coil or $13 \mu\text{V}$ over the sense resistor. After a ground isolation stage, this voltage is acquired with a 24-bit digitizer. The noise of $65 \text{ nV Hz}^{-1/2}$ is dominated by the isolation stage and would result in $5.8 \mu\text{V}_{\text{rms}}$ given the 8 kHz measurement bandwidth. Therefore, digital low-pass filtering with $f_c = 50$ Hz is applied to allow sub- μV resolution.

The number of measurements per setting, i.e., dipole on or dipole off, was 400 so that for voxel sizes of 15, 20, and 25

mm averaging could be implemented to increase sensitivity. The overall measurement time amounted to ~ 40 min (on and off) and to ~ 60 min if the additional reference off′ was taken, respectively.

2.4. Data Post-Processing

The magnetic flux changes during the data acquisition t_{Det} result from the induced echo signal $M_z(t)$ (see **Figure 3**), as well as interfering field changes of various sources. Starting the data analysis, we apply first a fitting routine and subtract low-frequency transients. These result from transient responses of the shielded room, background field changes or small vibrational movements of the sensor system within the gradient field. A truncation to remove filter artifacts leads to a $\sim 10\%$ larger voxel size in the x -direction.

The Hilbert-Transformation is applied to the real-valued measured data to derive the analytical signal $s_a(t_n)$, with t_n denoting a discrete step of time t . This is then demodulated using two synthetic waveforms c_1 and $c_2(t_n)$ according to:

$$s(t_n) = s_a(t_n) \cdot c_1 \cdot c_2(t_n) \quad (1)$$

$$c_1 = \exp(-i\Delta\phi_0) = \exp\left(-i2\pi \frac{t_{\text{Ph}}}{N} \sum_{k=1}^N \Delta f_L(t_k)\right) \quad (2)$$

$$c_2(t) = \exp(-i\Delta\phi(t)) = \exp\left(-i2\pi \int_{t_{\text{Ph}}}^t \Delta f_L(\tau) d\tau\right) \quad (3)$$

$$c_2(t_n) \approx \exp\left(-i2\pi \frac{t_n - t_{\text{Ph}}}{2n} \sum_{k=1}^{n-1} (\Delta f_L(t_k) + \Delta f_L(t_{k+1}))\right), \quad (4)$$

where c_1 adjusts the initial phase, accumulated during the phase encoding period t_{Ph} , and, $c_2(t_n)$ takes time varying phase changes during data acquisition into account, and, f_L is the Larmor frequency. A further phase correction is applied, in order to adjust the boundary between two slices to the position of the current dipole in the phantom, thus avoiding signal loss due to signal cancellation within one voxel reducing the difference signal amplitude.

Subsequently, performing the 3D Fourier transform of the time-domain difference signal $\Delta s = s_{\text{on}} - s_{\text{off}}$ and calculating its amplitude image ΔS , the highest possible CNR and reduced influence of the detection field instabilities can be achieved. For further analysis, the voxel with the largest amplitude ΔS_{max} is identified and the maximum CNR calculated according to:

$$\text{CNR} = \frac{\Delta S_{\text{max}}}{\text{SD}}. \quad (5)$$

Here, the image noise is given by the standard deviation SD of the complex Gaussian noise $\mathcal{N}(0, \text{SD}^2)$. It is obtained using $\text{SD} = \langle \Delta S_N \rangle / \sqrt{(\pi/2)}$ where $\langle \Delta S_N \rangle$ is the mean of the Rayleigh distributed amplitude noise of the difference image ΔS . For the simulations, $\langle \Delta S_N \rangle$ can be evaluated using the entire (off–off′) images. For the experiments, it is evaluated over the central

x -region of the difference images to take effects of residual artifacts into account (see **Appendix** for details).

2.5. Computational Models of the Phantoms

The magnetic field internal and external to the phantom is generated by the sum of the primary current in the dipole electrodes and the return volume currents within the solution. It was calculated using commercial software (COMSOL Multiphysics) in the quasi-static regime after the geometry of the phantom and the electric sources were carefully digitized. In the quasi-static approximation, the electric field \mathbf{E} satisfies $\nabla \times \mathbf{B} = 0$ both inside and outside the phantom volume and one can use the electrostatic potential V leading to the Laplace equation $\Delta V = 0$. Since the currents can only flow parallel to surfaces other than the electrodes, the boundary condition for the inner phantom surface, including the PCB mounting, is $E_{\perp} = 0$. For the calculation, the potential was set to $+V$ on one electrode base and to 0 on the counter electrode base. As $\sigma_{\text{Pt}} = 9.43 \times 10^6 \text{ S/m} \gg \sigma_{\text{Sol}}$, \mathbf{E} is essentially perpendicular to the uninsulated Pt conductor surface.

The volume current density \mathbf{J} was obtained by solving $\mathbf{J} = \sigma \mathbf{E}$ and used to determine the magnetic flux density \mathbf{B} per unit current using $\nabla \times \mathbf{B}/\mu_0 = \mathbf{J}$. The magnetic field inside the phantom mimics the neuromagnetic field in the proximity of the source. After normalization by unit current and multiplication by the shape of the long-lasting MEG-derived activity, it served as the input to the Bloch equation solver to investigate the impact of the source parameters on the dcNCI signal.

The magnetic field outside the phantom was calculated at all the coordinates of the 304 channel SQUID system of PTB [32] for comparison with experimental measurements and calibration purposes. In the model, the phantom was placed with its center 70 mm below the bottom of the sensor array. The simulated data allowed the estimation of the ECD of the phantoms which was compared to the ECD of the built phantoms, as estimated from direct measurements.

2.6. Analytical MR Solver for 3D dcNCI

A Matlab-based (The MathWorks, Inc.) NMR solver was created to execute virtual NCI, ULF-NMR, and MRI experiments. For arbitrary time varying fields $\mathbf{B}(t)$, the solution of the Bloch equation can usually only be obtained by numerical methods. However, if the spatial direction of the total magnetic field \mathbf{B} experienced by the sample is constant, the problem simplifies immensely and one can determine the evolution of the magnetization for any time dependent field using an analytical expression.

Since the dipole phantoms are central in the MRI coil system, we assume ideal MRI fields solely along the x -direction and neglect concomitant gradients. In this case, the instantaneous Larmor frequency, determined by $B_x(\mathbf{r}, t)$, can be evaluated directly provided accumulative phase adjustment is taken into account. With $t_0 = 0$ and \mathbf{M}_0 assumed to be constant throughout the volume due to the homogeneous \mathbf{B}_p , the precession of the complex magnetization $\mathbf{M}(\mathbf{r}, t_n) = M_y(\mathbf{r}, t_n) +$

$iM_z(\mathbf{r}, t_n)$ of a volume element dV and a given time step t_n is calculated as:

$$\mathbf{M}(\mathbf{r}, t_n) = \mathbf{M}_0 \exp \left(i\gamma B_x(\mathbf{r}, t_{n-1})t_n - i\gamma \sum_{k=2}^n \left(B_x(\mathbf{r}, t_{k-1}) - B_x(\mathbf{r}, t_{k-2}) \right) t_{k-1} \right) \exp(-t_n/T_2), \quad (6)$$

where γ is the gyromagnetic ratio of the proton. In case of the dipole field $\mathbf{B}_{\text{Dip}}(\mathbf{r}, t)$, only the x -component parallel to the much larger imaging fields is considered. The solver was verified against a numerical solution of the Bloch equation and found to be in very good agreement with differences below the parts-per-million level.

For the phantom volume, anisotropic discretization was implemented with a minimum spacing of 0.1 mm close to the dipoles. The time-domain signal $s(t_n)$ of the SQUID output is then computed according to:

$$s(t_n) = \int \mathbf{C}(\mathbf{r})^T \mathbf{M}(\mathbf{r}, t_n) dV, \quad (7)$$

where the coupling field of the sensor $\mathbf{C}(\mathbf{r})$ was obtained using the principle of reciprocity. A validation via an NMR experiment with an 80 mm diameter spherical sample of distilled water found excellent agreement both in the amplitude and the shape of the free induction decay (FID), but also small frequency changes in the measured data, as shown in **Figure 4**. By evaluating the phase difference $\Delta\phi$ between the simulated and measured FID (see **Appendix**), we determined the corresponding field drift ΔB_x within the first 50 ms to ~ 15 nT and identify transient fields following the fast turn-off of the x -directional B_p of 17 mT as the

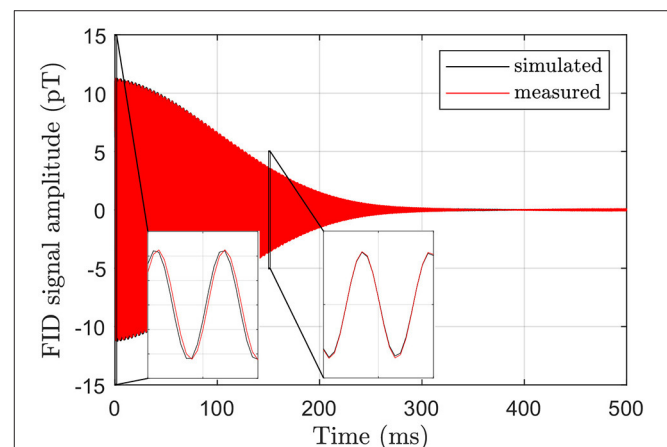


FIGURE 4 | Comparison of simulated and measured FID signals for dephasing in the static field $B_0 = 38.6 \mu\text{T}$ and the gradient $\partial B_x/\partial x = 2.4 \mu\text{T/m}$. The insets show close ups from 0 to 1 ms and from 150 to 151 ms, respectively, and reveal a small field drift of ~ 15 nT in the measured FID compared to the simulation which assumes a constant $B_x(\mathbf{r})$.

likely origin. However, we found that this transient response and additional frequency shifts are very reproducible and removed by the subtraction method described in section 2.4.

Phantom dcNCI experiments were then simulated for dipole depths of 15 and 35 mm representing a shallow and a deep cortical source, respectively. The voxel sizes were isotropic ranging from 5 to 25 mm.

3. RESULTS

3.1. Simulated Magnetic Field Distributions

We first present the results of the FEM calculations of the internal and external fields. In **Figure 5**, the magnetic field generated by the single dipole and the dipole grid within the phantom are shown for an applied current of $5 \mu\text{A}$. Compared to the single dipole, the maximum field per unit current produced by the extended dipole is significantly smaller due to cancellation effects within the array. The simulations also show that cross currents in the grid phantom are effectively suppressed by the holes in the mounting plate facilitating the current flow across the grid.

3.2. Calibration of Physical Phantoms

The calibration of the phantoms was carried out using the simulated and measured field distributions B_z/I outside the phantoms, as illustrated in **Figure 6**. The field generated by the extended dipole per unit current is 5% larger compared to the single dipole. The ECD reconstruction assumes a point-like dipole within a homogeneous conducting sphere. In case of the simulated data, the center of the sphere was set to -70 mm (with respect to the bottom of the sensor array) and the position of the dipole to 15 mm above it, as given by the geometry of the computational models.

For the measured maps, five fitting parameters were used: the coordinates of the sphere center relative to the bottom plane of the sensor array and the ECD components in x - and y -direction (zero z -component was imposed, as the alignment was defined by the phantom geometry). The estimated uncertainty of the ECD's vertical position of 0.5 mm was taken into account by performing source reconstructions for fixed distances in the

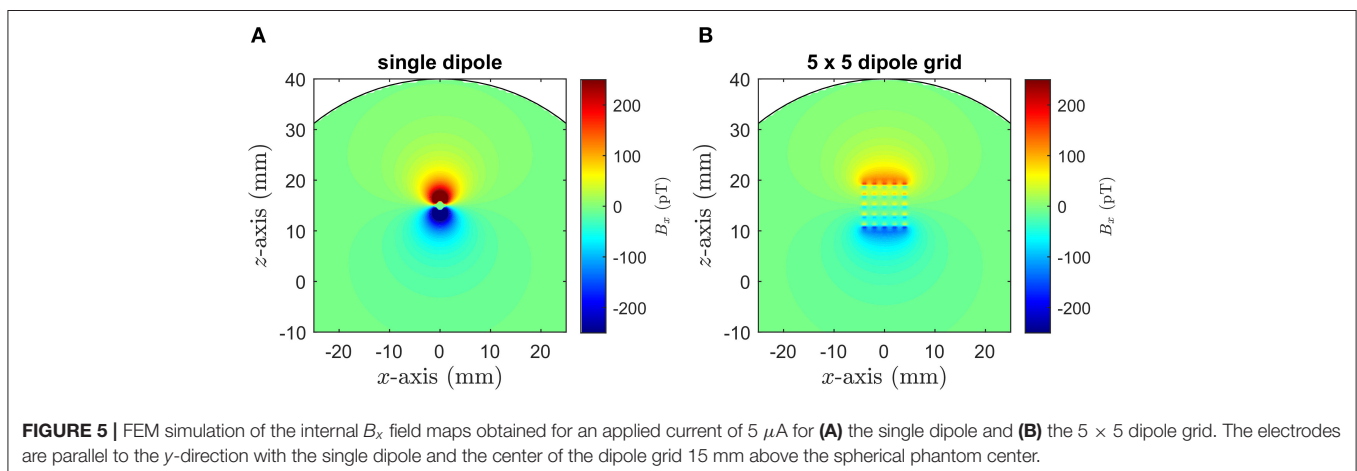
range of $(15 \pm 0.5) \text{ mm}$. The resulting effective dipole lengths (ECD normalized by applied current) are shown in **Table 3**. The uncertainties are derived from the standard deviation of the fit parameters combined with the uncertainty of the dipole distance to the sphere center, as shown in **Figure 6F**.

The simulated effective lengths of the dipoles were found to be about 10% smaller than the physical dimension of the dipole electrodes. This is in line with expectations, as the current in the electrodes flows into the solution as soon as it reaches the uninsulated element of the Pt wire. Also, the effective length of the extended dipole is slightly larger than the single dipole both in the simulation and the measurement. The disagreement of about 30% compared to the actual values arises from an insufficient electric insulation of the current feeds on the PCB immersed in the aqueous solution. Due to this shunting, some of the current does not flow over the electrodes resulting in a reduced effective length of the physical dipoles.

3.3. Simulations and Experiments of 3D dcNCI

We next present the results of the simulated noise-free dcNCI experiments considering 35 mm deep dipoles. In **Figure 7A**, the amplitude image together with the difference images are shown for the single dipole and the dipole grid for an ECD_{max} of 41 and 43 nAm, respectively. Also shown are experiments for a voxel size of 25 mm in **Figure 7B**, which were obtained for a dipole current of $80 \mu\text{A}$ corresponding to an ECD_{max} of 418 and 497 nAm for the single dipole and the dipole grid, respectively. The images have been phase adjusted to obtain the maximum difference signal ΔS_{max} . While the original location of the single dipole and the dipole grid center was $z = 0$ for all voxel sizes, this adjustment shifts the dipole positions toward the voxel edge along $+z$. In this way, signal loss due to cancellation effects within one voxel is minimized.

The maximum amplitude of the difference image ΔS_{max} is about 1,000 times smaller than the maximum amplitude image itself in agreement with previous NMR phantom studies [18]. As expected from the simulations of the phantom's internal field, ΔS_{max} of the single dipole is about a factor of 1.7 larger



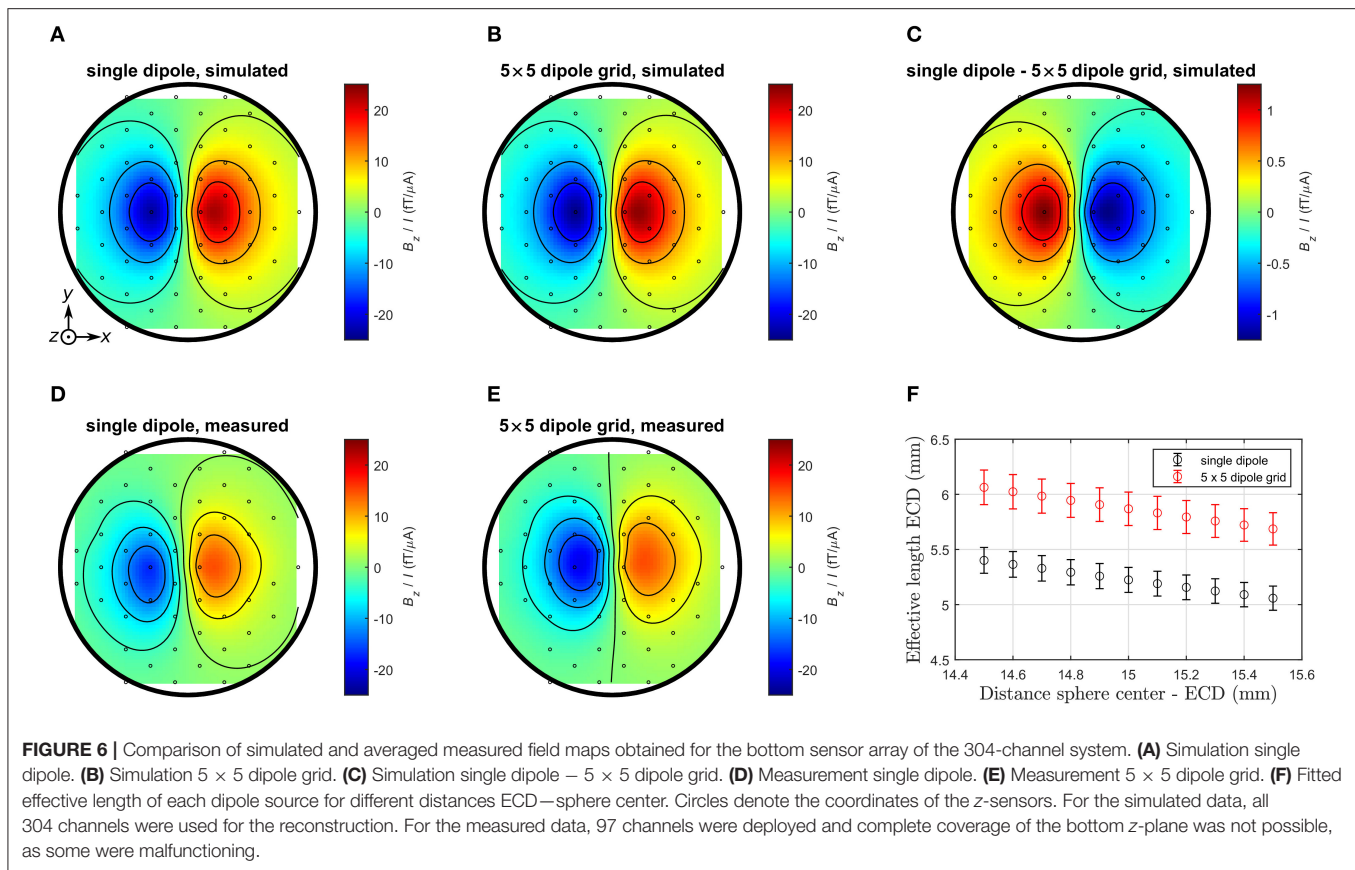


TABLE 3 | Simulated and measured effective dipole lengths of the two phantoms.

Phantom	Effective length simulated (mm)	Effective length measured (mm)
Single dipole	8.25	5.22 ± 0.30
5×5 dipole grid	8.70	5.87 ± 0.34

The physical length of an individual current dipole is 9.5 mm with 1 mm insulation removed at each electrode.

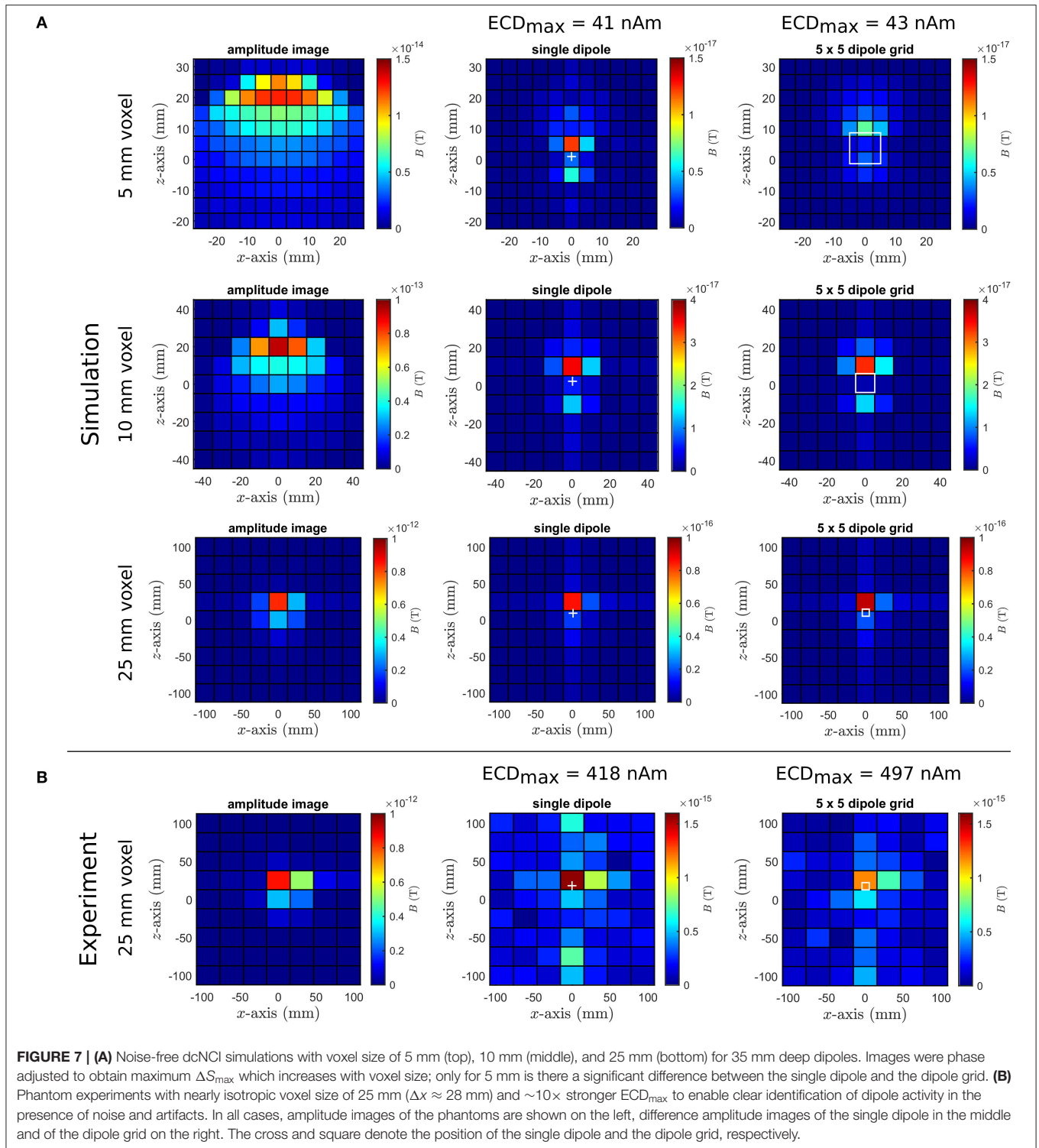
compared to the dipole grid for the 5 mm voxel size. The effect is most pronounced around the dipole where B_{Dip} is largest and the difference image appears smoother and more extended for the dipole grid. This effect disappears for voxel sizes larger than 10 mm and very similar ΔS_{max} are determined. Clearly, larger voxel sizes cannot reflect the detailed structure of the internal fields between the single dipole and the dipole grid. The phantom experiments for 25 mm voxel size reasonably reproduce the simulations with a twice as large expected ΔS_{max} for the single dipole. Good agreement is found for the dipole grid in this particular measurement. However, also visible are residual artifacts that could not be removed by the post-processing procedure (see **Appendix** for details).

The dependency of ΔS_{max} on the voxel size for dipole depths of 15 and 35 mm, as determined by the simulations, is displayed in **Figure 8A**. ΔS_{max} decreases with larger depth according to

the sensitivity profile of the sensor system. In addition, ΔS_{max} increases linearly with voxel size up to about 15 mm and then levels off, which is particularly clear for the 35 mm deep dipoles. Also included is the image noise which reflects the sensitivity of the MRI setup for the image sequence using a phase encoding time of 30 ms. **Figure 8A** suggests that a 15 mm deep dipole with a maximum strength of 50 nAm could be theoretically detected by the low-noise MRI setup using the defined imaging sequence for voxel sizes >20 mm. In contrast, a 35 mm deep dipole would be unresolvable even for the largest voxel sizes, and we note that there is no significant difference between the single and the extended dipole source. Only for voxel sizes comparable to the physical dimensions of the dipolar source does the single dipole show a larger ΔS_{max} compared to the dipole grid, as already mentioned before.

Figure 8B shows the dependence of ΔS_{max} on the applied ECD_{max} for dipoles at a fixed depth of 35 mm and a voxel size of 25 mm. The simulations predict, that the amplitude of the difference signal decreases linearly with a decreasing current dipole strength in line with previous NMR phantom studies [18]. For the experimental data, ΔS_{max} and the SD are factor of ~ 2 larger than predicted by the simulations. We attribute both effects to the aforementioned residual artifacts and discuss this in more detail later.

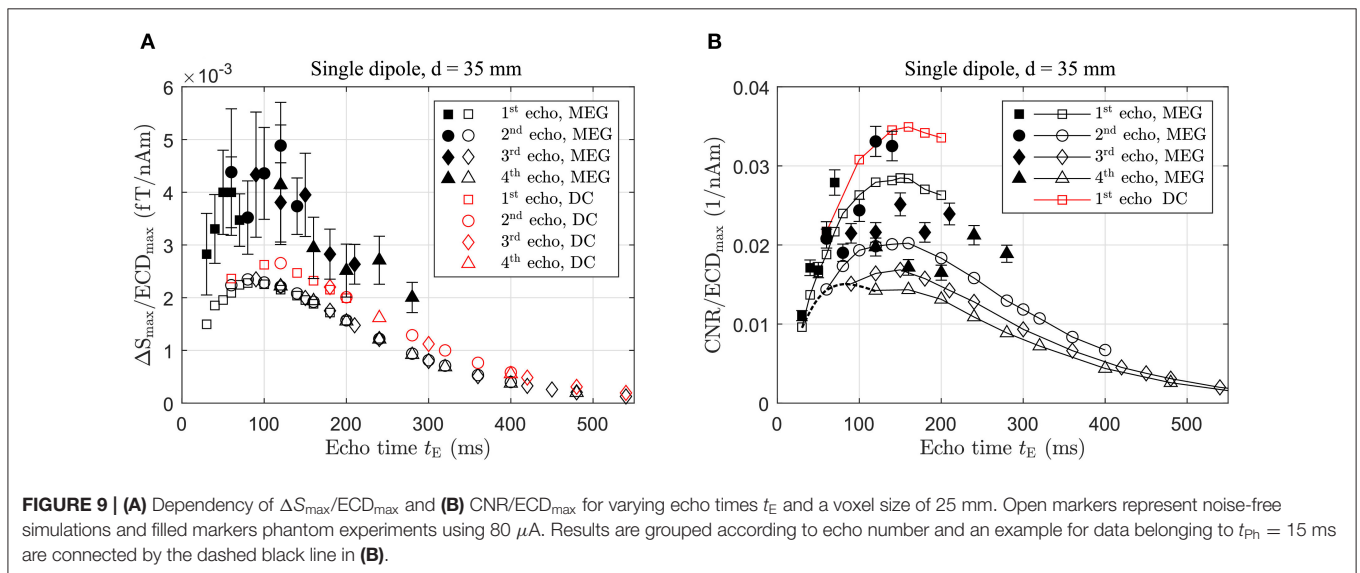
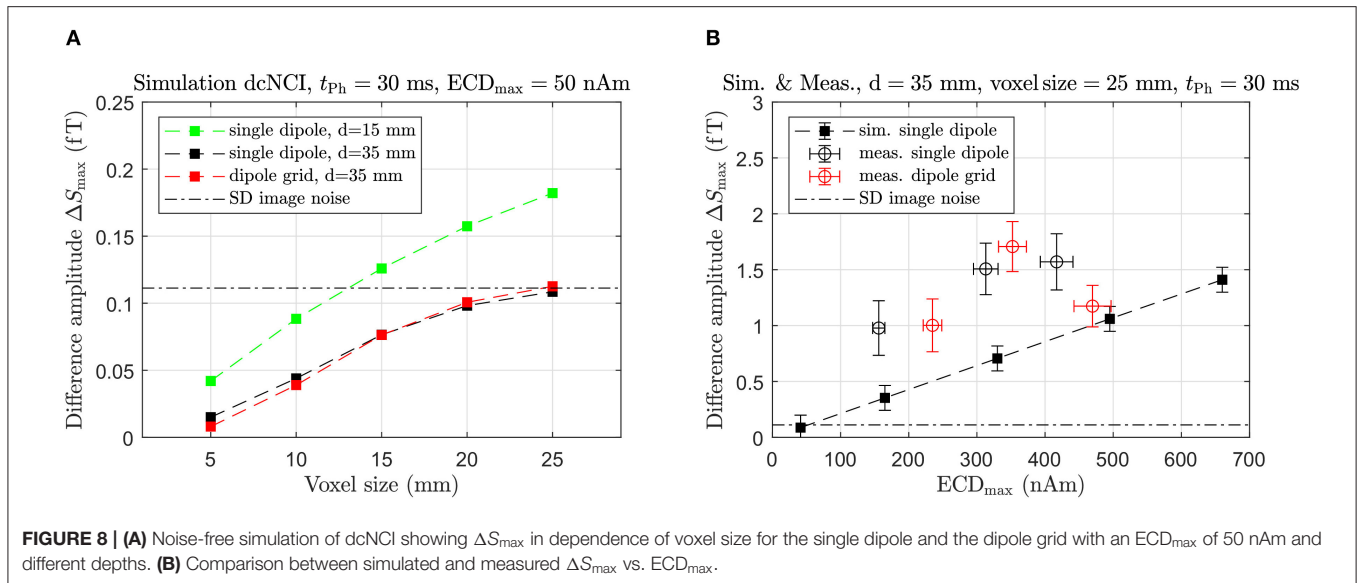
The results of the simulations and measurements indicate that a 35 mm deep dipole with an ECD_{max}



of about 150 nAm could be resolved using voxel sizes not smaller than 25 mm. Therefore, the detection of a realistic current dipole strength of 50 nAm requires an improvement of the experimental sensitivity by at least a factor of 3.

3.4. Optimization of 3D dcNCI Sequence

In order to identify optimal dcNCI sequence parameters for maximum CNR, multiple images were simulated varying the phase time t_{ph} from 15 to 100 ms corresponding to echo times t_E of 30 to 200 ms for the 1st echo. The gradients G_F and $G_{Ph,max}$



were adjusted accordingly to obtain a voxel size of 25 mm, as illustrated in **Figure 3**, while the number of phase encoding steps was fixed with $N_y = N_z = 10$ retaining the $FOV_{y,z}$ of 25 cm. Since the sampling frequency was kept constant and t_{Det} depends on t_{ph} , the FOV_x varied, but was in the meter-range in all cases.

An optimum sequence is expected, as two competing mechanisms occur. A longer t_E allows B_{Dip} to cause larger frequency shifts, but, on the other hand, spin relaxation leads to a signal decrease. As B_{Dip} is not constant and derived from an actual MEG measurement, comprehensive simulations are required to determine an optimal t_{ph} . For comparison, the optimization was also carried out for a true DC activity, which may be elicited for instance by auditory stimulation [34]. We reiterate that this is in contrast to conventional imaging where t_{ph} should be as short as possible to obtain maximum image SNR. We

also limit this optimization to an inverted frequency encoding gradient G_F during echo formation so that $t_{Det} = 2t_{ph}$ as this is implemented in our hardware. Application of a different G_F during this period would allow one to vary t_{Det} independently from t_{ph} .

In **Figure 9**, the dependencies of ΔS_{\max} and the CNR, both normalized by ECD_{\max} , on the echo time t_E are illustrated. The normalization allows the comparison of the predictions between the realistic sustained MEG activity and a constant DC signal. Since four echoes were simulated for each phase time, different t_{ph} result on occasion in an identical t_E for different echoes. As shown in **Figure 3**, for example, the 2nd echo of a sequence with t_{ph} is formed at the same time as the 1st echo when $t'_{ph} = 2t_{ph}$ resulting in equal echo peak amplitudes. Hence, the values for ΔS_{\max} collapse for separate echoes provided the echo times are

the same. The DC signal results in a larger normalized ΔS_{\max} and the maximum values are observed for an echo time of ~ 90 and 120 ms for the MEG and the DC signal, respectively.

The experimental results were obtained with a dipole current of $80 \mu\text{A}$. Again, they show values for $\Delta S_{\max}/\text{ECD}_{\max}$ about a factor of 2 larger than the predictions; however, they also clearly demonstrate the maximum at $t_E \sim 100$ ms in line with the simulations.

The CNR, on the other hand, shows an optimum at $t_E \sim 150$ ms both for the sustained MEG and the constant DC signal, respectively (see **Figure 9B**). In explaining this result, we refer again also to **Figure 3**. In our sequence we always have $t_{\text{Det}} = 2t_{\text{ph}}$ and for a given echo, an increase in t_{ph} , and correspondingly t_E , leads to a longer acquisition time t_{Det} . For the 1st echo, for instance, this gives $t_{\text{Det}} = t_E$. Since the sampling frequency is kept constant, the noise is then distributed over more frequency bins resulting in $\mathcal{N} \propto t_E^{-1/2}$. Beyond t_E resulting in maximum ΔS_{\max} , the image noise decreases more strongly compared to ΔS_{\max} pushing the optimum for CNR to longer echo times. The maxima are rather broad and, as $\Delta S_{\max}/\text{ECD}_{\max}$ is smaller for the MEG-derived signal, its CNR is correspondingly decreased. The maximum CNR for each echo occurs at the same echo time but decreases for consecutive echoes as t_{Det} gets successively shorter and consequently the noise larger, as discussed above. This is not shown for the true DC signal which exhibits the same behavior. For large t_E , spin relaxation dominates causing ΔS_{\max} and CNR to approach zero eventually.

The enhancement in CNR compared to the experimental setup when using a t_E of 60 ms is a factor of 1.5 for the MEG signal improving the detection limit from ~ 150 to 100 nAm. A true DC activation in comparison to the MEG-derived activity results in an even better improvement of 1.85 provided it has the same ECD_{\max} . The phantom experiments confirm this picture and, within the accuracy of the measurement, the maximum in CNR at $t_E \sim 150$ ms is also observed. It is worth pointing out that increasing t_E from 60 to 150 ms has only a secondary effect on the total measurement time, since this is dominated by the polarizing time t_p , in particular if only the 1st echo is acquired.

4. DISCUSSION

4.1. Experimental Validation

A comparison between simulated and measured 3D dcNCI phantom experiments show an approximately 2-fold larger ΔS_{\max} for the experimental case. In the simulation model, we used independently determined quantities, e.g., SQUID conversion factor, and the actual geometry of the phantom within idealized MRI fields without any freely adjustable parameter. The very good agreement in the validation measurement of the solver (see **Figure 4**) confirms the accuracy of the computational model. However, with this in mind, the generally larger effect in the 3D dcNCI phantom experiments deserves a closer inspection. It is possibly related to the increased uncertainty due to residual artifacts, which—as pointed out in the **Appendix**—are likely due to field drifts of environmental origin within the moderately shielded MSR that are undetectable by the sensing circuit. This

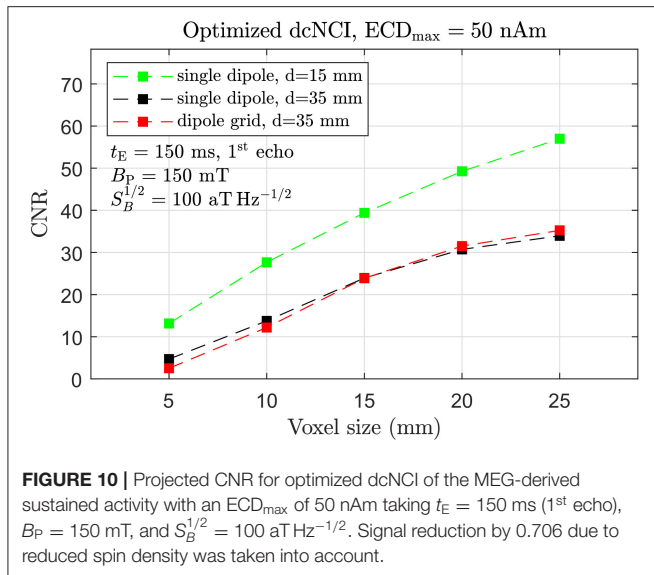
might introduce a bias toward an overestimation of ΔS_{\max} . Clearly, the residual artifacts after subtraction represent an experimental issue and a more strongly shielded environment may provide a possible solution. As a further comment, we consider an erroneous determination of the ECD_{\max} of the phantoms unlikely, since the observed shunting during the calibration proved to be stable in time. Further studies are required to determine the origin of the larger experimental ΔS_{\max} . Nevertheless, we consider the agreement adequate and the results of the simulations form the basis for further discussions below.

4.2. Theoretical Sensitivity Limits of 3D dcNCI

As we have shown, the theoretical CNR of 3D dcNCI using voxel sizes in the cm-range and an optimized sequence applied to our low-noise ULF-MRI setup is only slightly larger than unity. In addition, compared to the phantoms, the water content of brain tissue is lower and heterogeneous ranging from 70.6% for white matter, 84.3% for gray matter and 97.5% for cerebrospinal fluid (CSF) [35]. Partial volume effects may then be present as large voxel sizes are likely to include multiple tissues. Consequently, one can expect a reduction in CNR in an *in vivo* experiment which depends on the size and the location of each voxel and for the present discussion, we assume a worst case value of 70.6%. However, improvements are possible on the instrumental side. We performed the study for a moderate B_p of 17 mT, the maximum achievable field of our setup. A much larger B_p of up to 150 mT has been demonstrated in *in vivo* imaging of the human brain [3] corresponding to an increase in M_0 and consequently in ΔS_{\max} of a factor 9.

With respect to the noise performance, the application of a large B_p has been shown to lead to an excess low-frequency noise due to flux trapping in the superconducting pick-up coil [36]. However, it may be avoided by operation at a sufficiently large Larmor frequency [37], a suitable B_p ramp down [38] or rapid thermal cycling of the pick-up coil [39]. Consequently, the noise performance is ultimately limited by Johnson noise of the human body which was recently measured at 55 aT Hz^{-1/2} [40]. Improvements in SQUID performance, e.g., by use of sub-micron-sized Josephson junctions, lead us then to the conclusion that a noise level of about 100 aT Hz^{-1/2}, although being quite challenging, is nevertheless feasible [41, 42]. Of course, this discussion requires instrumental factors, such as noise generated by the field and gradient supplies to be negligible, as should be the case for our ULF-MRI setup. In addition, the occurrence of artifacts will have to be addressed. The projected CNR for the MEG-derived sustained activity with $\text{ECD}_{\max} = 50$ nAm and all these factors taken into account is shown in **Figure 10**.

An overall improvement of ~ 35 in CNR for the optimized dcNCI setup appears possible. With CNR ~ 10 , a 15 mm deep, shallow focal cortical source should be well resolvable with a voxel size of ~ 5 mm. For deeper sources, such as the exemplary somatosensory evoked sustained activity, the voxel size for CNR ~ 10 is about 10 mm. Voxel sizes in the range of 5 mm



result again in CNR close to unity and represent in our opinion the theoretical limit of 3D dcNCI.

4.3. Toward *in vivo* 3D dcNCI

The development of a validated computational framework, able to execute virtual dcNCI experiments, was motivated by multiple needs. The identification of the required technological improvements of ULF-MR hardware and the identification of optimized dcNCI sequence parameters being the most prominent. In particular, a sound knowledge of the impact of the neuronal source spatial distribution, timing and orientation on the MR signals permits the identification of optimal strategies for detection. Hardware measures, such as increase of B_P and the reduction of system noise, are rather obvious means to improve the CNR for the realization of dcNCI.

In assessing the relevance of the experimental and computational phantom study for *in vivo* dcNCI, several issues are worth discussing. dcNCI using ULF MRI, as presented here, is limited to evoked and sustained, ideally monophasic, neuronal activities. In addition, the emphasis is on focal activation which can be approximated, at least from an MEG point of view, by a dipolar source and parameterized by an ECD. In this case, the simulation studies show that for voxel sizes larger than 10 mm there is no significant difference between the single dipole and the extended dipole grid in sensitivity even though the internal fields are markedly different. This suggests, that the detailed spatial structure of the neuronal field in an *in vivo* experiment cannot be resolved by 3D dcNCI for voxel sizes larger than 10 mm and voxels in the low mm-range are needed to possibly achieve this.

The dependence on sustained activities of dcNCI has also bearing on the temporal resolution which is of the order of the echo time t_E , the time during which the long-lasting activity is present. For the somatosensory evoked and a true DC activation, this amounts to 120 ms for the optimized sequence which is

significantly longer than in MEG but superior to fMRI. Note that this comparison relates to the time needed to acquire one line in k -space and not the entire image, which is much longer. This emphasizes again the limitation to evoked activities that can be elicited reliably. By reducing t_E , an improvement of the temporal resolution at the expense of CNR is possible.

A further point which deserves attention is the requirement of parallel alignment between B_{Dip} and B_0 in order to result in a maximum CNR. However, this condition is less restrictive than it might first appear. As long as $B_0 \gg B_{Dip}$, which is certainly fulfilled even for dcNCI using ULF MRI, the component of B_{Dip} parallel to B_0 is of significance. As this scales as $\cos \phi$, where ϕ is the angle between B_0 and B_{Dip} , even a substantial misalignment by $\phi = 30^\circ$, for example, results in a reduction of the CNR by only $\sim 13\%$. As a corollary we note that a radial dipole, although silent in MEG, will show the full effect in dcNCI as long as the dipole is perpendicular B_0 .

Finally, the above estimation of the sensitivity limit was based on the assumption of a homogeneous distribution of white-matter-like tissue within the MRI voxel. However, as already mentioned, single voxel signals are expected to be largely influenced by partial volume effects due to the presence of multiple tissues with different proton densities and relaxation times, such as white matter, CSF, bone, dura, and blood vessels. This and many other concomitant factors, related to the complex anatomical and neuroelectric complexity of the human head, have not been investigated in this work. This could be done replacing simplified phantom models with high-resolution anatomical human head models, such as the ones of the Virtual Population (IT'IS Foundation, Zurich, Switzerland) [43] or the MIDA head model [44], combined with realistic electrophysiological models of neuronal networks.

5. CONCLUSION

In conclusion, we illustrated the elements of a validated computational framework allowing virtual experiments with the aim to assess the feasibility of 3D dcNCI. The simulations provide a controllable basis which allows the evaluation for a best-case scenario. To this end, we considered idealized MRI fields and magnetic field distributions generated by current dipole phantoms mimicking neuronal activities. The source models consisted of a single dipole and an extended dipole grid with well-defined phantom geometry enabling accurate FEM simulations of the internal and the external fields. The latter were validated with MEG-type measurements, which served as a calibration of the fabricated phantoms. An MR solver based on an analytical solution to the Bloch equation was developed and used to simulate the dcNCI experiment based on our low-noise 3D ULF-MRI setup. The framework was verified via phantom experiments and allowed the assessment of the detection limit. This experimental part was equally important, since it highlights the technical challenges which need to be addressed.

We found that with our current technology and an optimized dcNCI sequence minimal voxel sizes of 20 mm are required to detect a 35 mm dipole deep dipole with an ECD of about

100 nAm, which is a factor of 2 larger than the physiological value. In addition, we used this tool to project a possible 35-fold increase in CNR due to hardware improvements. The framework should be combined with field simulations of a realistic neuronal network embedded inside a cortical structure. This is highly desirable, as it would ultimately allow the optimization of *in vivo* dcNCI based on ULF MRI which remains a formidable challenge.

DATA AVAILABILITY STATEMENT

The raw data supporting the conclusions of this article will be made available by the authors, without undue reservation.

AUTHOR CONTRIBUTIONS

NH, AC, and RK contributed to the conception of the study. J-HS developed the dipole phantoms and performed the FEM

simulations. NH performed the dcNCI phantom experiments, analyzed the dcNCI data, and wrote the first draft of the manuscript. RK developed the MR solver and revised the manuscript based on the annotations of the co-authors. PH contributed to system performance. All authors contributed to manuscript revision, read, and approved the submitted version.

FUNDING

This project has received funding from the DFG under Grant KO 5321/1-1, the SNF under project No 200021E-166809, and from the European Union's Horizon 2020 research and innovation programme under grant agreement No 686865.

SUPPLEMENTARY MATERIAL

The Supplementary Material for this article can be found online at: <https://www.frontiersin.org/articles/10.3389/fphy.2021.647376/full#supplementary-material>

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Conflict of Interest: The 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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