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## Original Article

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# High-Resolution Numerical Simulation of Respiration-Induced Dynamic $B_0$ Shift in the Head in High-Field MRI

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**Purpose:** To demonstrate the high-resolution numerical simulation of the respiration-induced dynamic  $B_0$  shift in the head using generalized susceptibility voxel convolution (gSVC).

**Materials and Methods:** Previous dynamic  $B_0$  simulation research has been limited to low-resolution numerical models due to the large computational demands of conventional Fourier-based  $B_0$  calculation methods. Here, we show that a recently-proposed gSVC method can simulate dynamic  $B_0$  maps from a realistic breathing human body model with high spatiotemporal resolution in a time-efficient manner. For a human body model, we used the Extended Cardiac And Torso (XCAT) phantom originally developed for computed tomography. The spatial resolution (voxel size) was kept isotropic and varied from 1 to 10 mm. We calculated  $B_0$  maps in the brain of the model at 10 equally spaced points in a respiration cycle and analyzed the spatial gradients of each of them. The results were compared with experimental measurements in the literature.

**Results:** The simulation predicted a maximum temporal variation of the  $B_0$  shift in the brain of about 7 Hz at 7T. The magnitudes of the respiration-induced  $B_0$  gradient in the  $x$  (right/left),  $y$  (anterior/posterior), and  $z$  (head/feet) directions determined by volumetric linear fitting, were < 0.01 Hz/cm, 0.18 Hz/cm, and 0.26 Hz/cm, respectively. These compared favorably with previous reports. We found that simulation voxel sizes greater than 5 mm can produce unreliable results.

**Conclusion:** We have presented an efficient simulation framework for respiration-induced  $B_0$  variation in the head. The method can be used to predict  $B_0$  shifts with high spatiotemporal resolution under different breathing conditions and aid in the design of dynamic  $B_0$  compensation strategies.

**Keywords:** Dynamic  $B_0$  shift; Dynamic shim; Respiration; Susceptibility; Brain; 7T

## INTRODUCTION

In high-field MRI, tissue susceptibility-induced static field ( $B_0$ ) shift can cause significant imaging artifacts and spectroscopic errors. While numerous methods to mitigate the  $B_0$  shift have been proposed (1, 2), dynamically varying  $B_0$  shift remains particularly difficult to compensate for. Such changes can be caused by hardware drift (3), subject motion (4), and motion of foreign magnetic objects (5). In particular, the respiration-induced  $B_0$  shift in the head has been the focus of several experimental studies

due to its importance in functional MRI (6, 7) at ultra-high magnetic fields. Experimentally, the dynamic  $B_0$  shift can be measured using gradient-echo phase imaging (8), magnetic field probes (9), or from post-processing of multi-channel images (10). While providing valuable real-life  $B_0$  changes, experimental measurements of respiration effects have typically been limited to low spatial and temporal resolutions. Moreover, such studies have usually been conducted with only a small number of volunteers, representing a limited set of anatomies and breathing conditions.

In this respect, accurate numerical simulation of a respiration-induced  $B_0$  shift could offer an opportunity to investigate different factors affecting  $B_0$ , with flexibility and high spatiotemporal resolutions. Marques and Bowtell (11) reported a simulation study using a susceptibility-to- $B_0$  calculation algorithm based on the  $k$ -space-discretized Fourier space convolution. While fast and popular, the algorithm is known to require large zero-filled buffers in the computational grid to avoid aliasing artifacts (5, 12). Consequently, the study by Marques and Bowtell (11) used a large (7.6 mm isotropic) voxel, partly undermining the advantages of simulation studies.

In this work, we demonstrate dynamic, 1 mm-isotropic-voxel numerical simulation of respiration-induced  $B_0$  shift in the head using a detailed 4D human model. We used a recently published rapid  $B_0$  calculation algorithm (generalized susceptibility voxel convolution, gSVC) (12) to achieve high computational speeds while eliminating aliasing artifacts. We obtained  $B_0$  shift gradients in the head that were consistent with the experimental data in the literature. We propose that the method presented provides an efficient platform to investigate dynamic  $B_0$  shift in high-field neuro MRI.

## MATERIALS AND METHODS

### Human Body Model

The human body model, incorporating realistic organ motion from breathing, was generated from the Extended Cardiac And Torso (XCAT) model, which was originally used for computed tomography and positron emission tomography for radiation dose and image quality calculations (13–15). The model has also been used in MRI simulations for its utility in multimodal and interventional imaging research (16, 17). XCAT provides a set of adjustable parameters to define the anatomy and dynamics of the human body, including gender, scaling of the head and lung

sizes, segmented tissue properties (in our case, interpreted as magnetic susceptibility  $\chi$ ), turning on/off cardiac and respiratory motions, and depth of breathing. Notably, XCAT lets users change the voxel size to create body models with different voxel resolutions. In our work, we used the following parameters: gender = male, scaling = default, voxel size = 1 to 10 mm (varied, isotropic), susceptibility = 0 (air and lung)/-8.5 (blood)/-11 (bone)/-9 (other tissues) ppm, cardiac motion = off, maximum diaphragm motion = 2.0 cm, and maximum anterior/posterior expansion = 1.2 cm (nominally "normal" breathing). The model parameters were saved in a text file, which was read during the XCAT model generation process executed from within the Matlab (Mathworks, Natick, MA, USA) environment. No user-defined interpolation processes were needed to create models with different voxel sizes.

The respiration cycle was set to be 5 sec, starting from a fully exhaled position, designated as frame 1. A total of 10 frames within the cycle were simulated. The XCAT model is capable of generating voxelated whole-body breathing models at specified time points. From this model, we used the upper torso, from below the diaphragm (lumbar vertebrae 2) to the neck, as the time-varying susceptibility region. The  $B_0$  shift was then calculated in the brain mask of the first frame. Table 1 shows the first and last (axial) slice indices corresponding to the grids containing the torso portion and the head portion of the body model, at each voxel size. Figure 1 shows the susceptibility model and the brain mask for two frames (1, fully exhaled and 5, fully-inhaled) at two voxel sizes (1 and 7 mm).

Once XCAT generated the desired 4D model in a set of binary files, a custom Matlab code read the data for subsequent processing and  $B_0$  calculations. For all calculations, we used Matlab version R2017b on a 64 bit Windows PC with Intel Xeon CPU and 64 GB of RAM.

### Susceptibility-Induced $B_0$ Calculation

In our problem, the  $B_0$  change in the head was caused by susceptibility variation in the upper torso. The problem is a case where the susceptibility source grid (upper torso) is physically separated from the  $B_0$  target grid (head). In such a case, use of a conventional  $k$ -space-discretized  $B_0$  calculation algorithm applied to an all-encompassing grid containing both source and target regions is computationally inefficient (18). In the conventional method, the problem of converting a susceptibility map into a  $B_0$  change map is treated as that of image transformation, in which an input "image" (susceptibility) is convoluted

with a point dipolar field kernel  $K(\vec{r} - \vec{r}')$  to yield an output ( $B_0$  map). This method became popular (11) partly because of the existence of a simple expression for the kernel in the Fourier space,

$$K(\vec{k}) = \frac{1}{3} - \frac{k_z^2}{k^2} \quad [1]$$

which makes Fourier-domain convolution calculation easy and elegant. However, Eq. [1] is strictly valid when the spatial-domain kernel  $K(\vec{r} - \vec{r}')$  is defined in an infinite space. When discretizing Eq. [1] in the  $k$ -space, as is conventionally done, one implicitly assumes that the convolution is calculated in a finite-sized spatial domain. This results in the susceptibility map being replicated periodically outside the domain, leading to aliasing artifacts in the output  $B_0$  map (5). Practically, this is mitigated by adding zero-filled buffers around the computational grid (containing both susceptibility and target  $B_0$  regions) to sufficiently push away the susceptibility replicas. Proposals have been made to reduce the necessary computational grid size (and the time) by discretizing the kernel  $K$ , not in the  $k$ -space but in the real space, and also by separating the susceptibility grid from the target grid (18). However, these methods were limited to cases where the two grids have

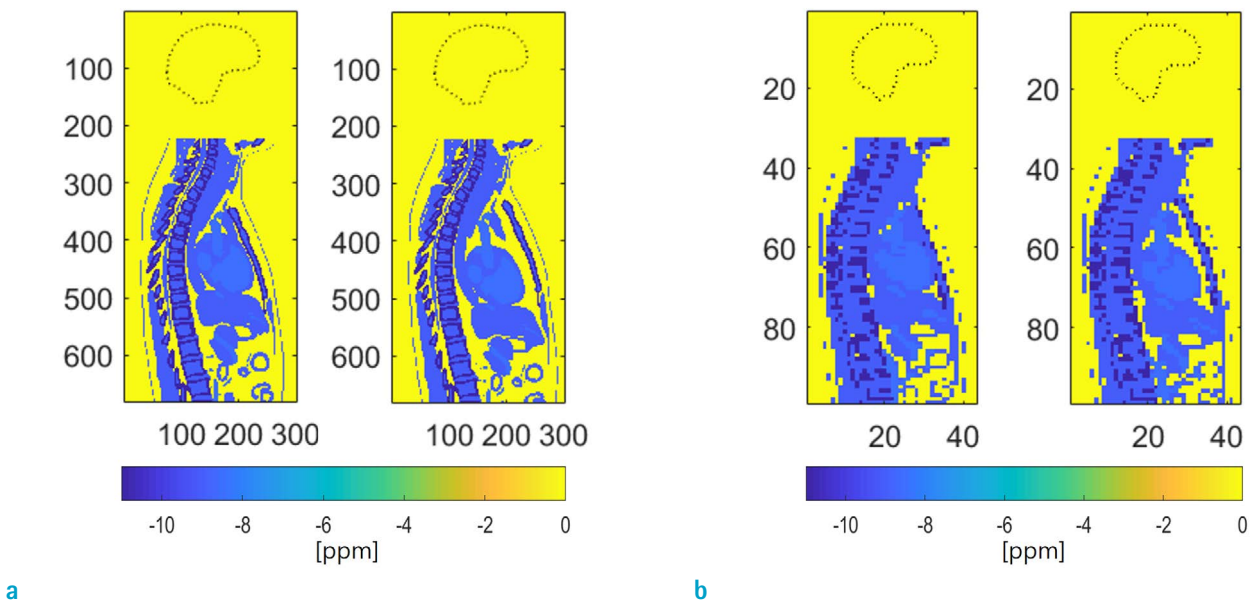
the same size and the voxels are isotropic, as explained by Lee et al. (12).

In this work, we used the gSVC method (12) to calculate the  $B_0$  change in the head from the voxelated susceptibility source in the upper torso at each time point in the breathing cycle. The method requires a modified dipolar field kernel that incorporates an analytical solution of the susceptibility-induced  $B_0$  field from a uniformly magnetized voxel (cubical in our case) instead of a point dipole. The kernel in the real space is given by

$$-\frac{1}{4\pi} \sum_{i=1}^2 \sum_{j=1}^2 \sum_{k=1}^2 (-1)^{i+j+k} \arctan \left( \frac{x_i y_j}{z_k r_{ijk}} \right) + c \cdot \frac{1}{3} \quad [2]$$

where  $x_i, y_j, z_k$  define the eight corners of the voxel, and  $r_{ijk} \equiv \sqrt{x_i^2 + y_j^2 + z_k^2}$ . In the second term,  $c = 1$  if the target point is within the voxel and  $c = 0$ , otherwise.

Eq. [2] is more complicated than the point dipole kernel but is readily calculable in Matlab. This approach eliminates aliasing artifacts without the need for grid-increasing buffers. Furthermore, in gSVC, the convolution in the real-space is explicitly formulated for separated input (susceptibility) and output ( $B_0$ ) grids, with generally different sizes. This results in the total computational grid having size  $N=N_s+N_T-1$  in all



**Fig. 1.** Sagittal cross sections of the brain mask (dotted line) and the susceptibility model at voxel sizes of 0.1 cm (a) and 0.7 cm (b). Tick mark labels indicate voxels. For each voxel size, frames 1 (left) and 5 (right) are displayed out of the 10 frames in the breathing cycle simulated by the 4D XCAT model. Color bar at the bottom indicates susceptibility in ppm (0 = air and lung, -8.5 = blood, -9 = other tissue, -11 = bone).

3 Cartesian dimensions, where  $N_s$  and  $N_t$  denote the source and target grid sizes, respectively. Once the kernel Eq. [2] is computed for the relevant grid of size  $N$ , such a kernel can be re-used in a dynamic  $B_0$  calculation, where discrete convolution between the (time-dependent) susceptibility and the (fixed) kernel is done for all time points (12).

In order to examine the effect of coarse-graining the susceptibility model, we computed the dynamic  $B_0$  fields at voxel sizes ranging from 1 to 10 mm. For each voxel size, we computed the  $B_0$  in the brain (with the same voxel size as the susceptibility model) and compared the results. The computational times for the dipolar field kernel calculation (called  $t_1$ ) and dynamic  $B_0$  update (called  $t_2$ ) were recorded at each voxel size.

**Analysis**

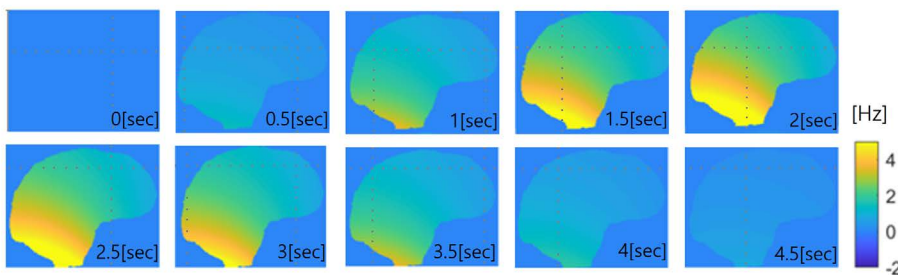
In order to focus our attention on the dynamic effects of breathing, the  $B_0$  map of the first frame was subtracted from the subsequent frames. In what follows, we assume that the symbol  $B_0$  denotes only this difference, namely the dynamic component of the field offset. The respiration-induced  $B_0$  in the head is expected to vary slowly in space, dominated by low-order spatial harmonics. Based on this, and in order to compare our results with previous experimental findings (8-10), we computed the linear gradients of the dynamic  $B_0$  shift maps along three directions:  $G_x = dB_0/dx$ ,  $G_y = dB_0/dy$ ,  $G_z = dB_0/dz$ ,  $x, y, z$  is not subscript. (Make it "dx") and  $G_z = dB_0/dz$ ,  $z$  is

not subscript. (Make it "dz") where  $x, y, z$  were coordinates in the left-to-right, posterior-to-anterior, and feet-to-head directions, respectively, for a patient in the head-first supine position. The gradients were obtained two different ways. First, the  $B_0$  values on each slice normal to a given gradient direction ( $x, y, z$ ) were averaged. The 2D averaged  $B_0$  thus obtained was then plotted against the coordinate in the gradient direction and fitted with a straight line. The slope of the line represented the gradient of  $B_0$ . This was the method used by Van de Moortele et al. (8). Second, we fitted the simulated 3D  $B_0$  maps to a volumetric linear function of the form  $f(x, y, z) = xG_x + yG_y + zG_z + const$  to obtain three gradients. The fitting in this case was done by minimizing the mean square of the residuals using the Matlab function 'fminsearch'.

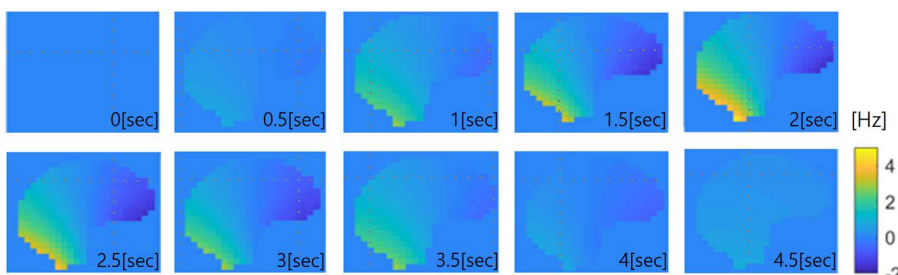
**RESULTS**

**Magnitude of Temporal Variation**

Figure 2 shows the temporal variation of the  $B_0$  shift shown on the midsagittal plane at voxel sizes 0.1 cm and 0.7 cm. All shifts are shown relative to the first frame. Therefore, the  $B_0$  maps show the change in  $B_0$  when inhalation occurs and the lung expands. Figure 2a shows that for the highest-resolution simulation (0.1 cm, presumed most reliable), inhalation generally shifts  $B_0$



a

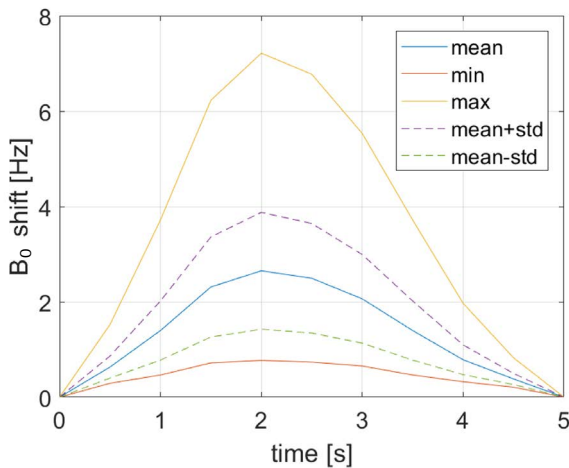


b

**Fig. 2.** Midsagittal view of the respiration-induced  $B_0$  shift in the brain at 7T at 10 time points in a breathing cycle. Time order is from left-to-right. The  $B_0$  calculation algorithm gSVC was used. (a) and (b) correspond to voxel sizes of 0.1 cm and 0.7 cm, respectively. Note significant differences between  $B_0$  shifts computed at 0.7 cm vs. 0.1 cm resolution.

upward. This can be understood as the result of the lung filling with more air, which is relatively more paramagnetic than the tissue. Figure 2b shows that the low-resolution simulation yields  $B_0$  shifts that are substantially different from the high-resolution one. From Figure 1, we believe that this was mainly due to the poor representation of the body organs at this voxel size. For example, partial volume errors could lead to overestimation of the total volume of the tissue in the torso at the inhaled position. This could cause a decrease in the  $B_0$  shift in the brain because of the negative susceptibility of the (overestimated) tissue.

Figure 3 shows the time dependence of the minimum,



**Fig. 3.** Time dependence of  $B_0$  shift statistics in the brain during breathing computed at 0.1 cm voxel size. Maximum  $B_0$  shift of 7.2 Hz is observed at 2 s (5<sup>th</sup> frame).

maximum, and mean values of the  $B_0$  shift in the brain at the 0.1 cm voxel size. Also indicated are one standard deviation of  $B_0$  at each time point. Over the entire brain, the mean  $B_0$  shift peaked at 2.7 Hz at the fully-inhaled position (frame 5). This is consistent with the experimental data by Van de Moortele et al. (8), who reported an overall brain  $B_0$  shift due to respiration of 1.45 to 4 Hz at 7T. The maximum local excursion of  $B_0$ , in contrast, was 7.2 Hz in our model, which occurred at frame 5 at the bottom of the brain. This value, relative to the main magnetic field of 7T, compares favorably with a recent measurement (10) reporting about 2.8 Hz  $B_0$  shift at a similar location at 3T.

### Spatial Gradient

Spatial gradients of the respiration-induced  $B_0$  shift can inform dynamic shimming in high-field brain imaging. Figure 4 shows the slice-averaged  $B_0$  values and their fitted line in three directions for the fully-inhaled (frame 5) temporal position. Compared to Table 1 in the study by Van de Moortele et al. (8), our results for  $G_x$ ,  $G_z$  are within the range observed experimentally ( $G_x = +0.0163$  Hz/cm in our case, ranged between -0.035 and +0.025 in (8);  $G_z = -0.365$  Hz/cm in our case, ranged between -0.380 and -0.155 Hz/cm in (8)). In contrast, we found higher simulated respiration-induced  $G_y$  (-0.236 Hz/cm) compared to their data (8) (-0.093 to -0.032). This discrepancy can be because, in that study, the measurement was made with a surface coil positioned on the posterior side of the head and may not have covered the frontal brain sufficiently.

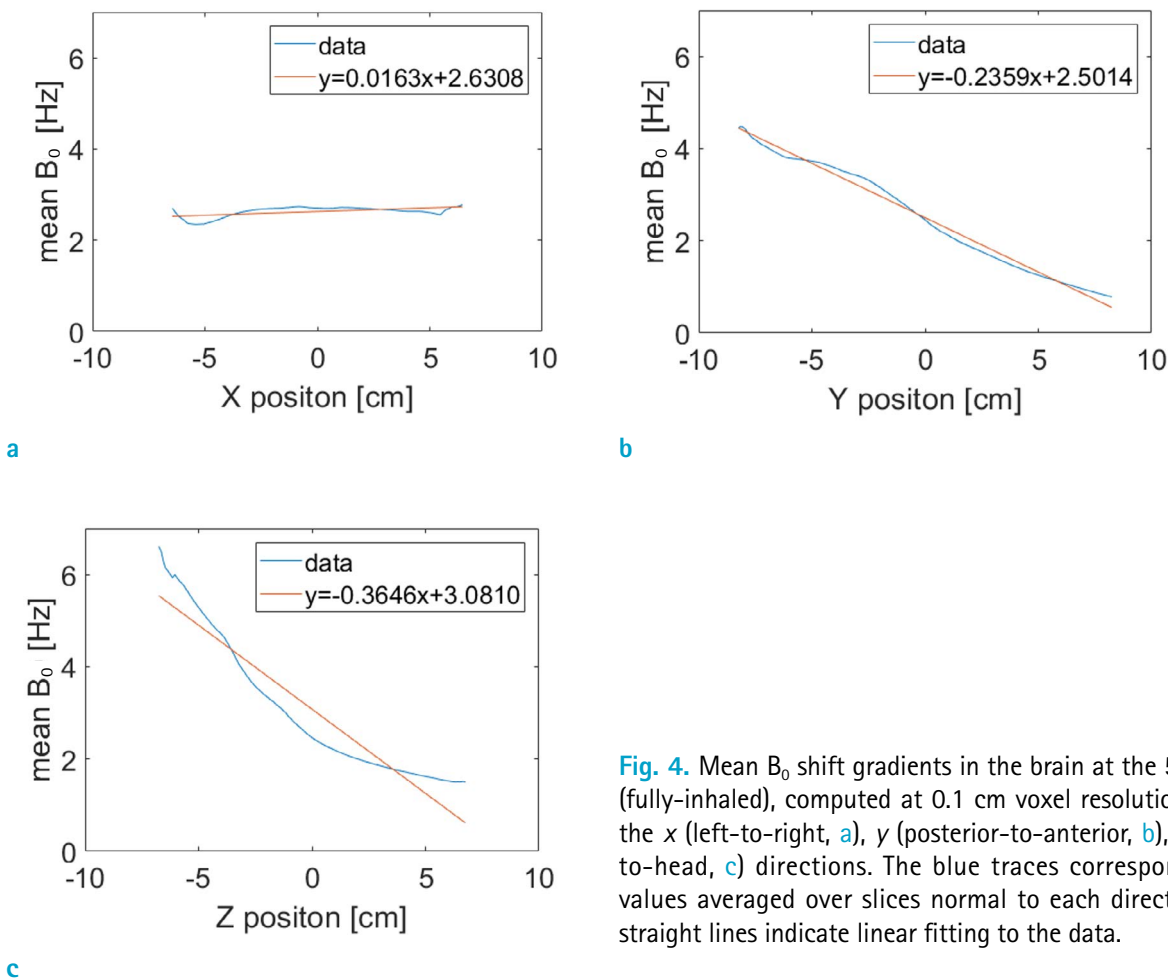
**Table 1.** Susceptibility source (torso) and  $B_0$  calculation target (head) positions, computational times, and respiration-induced  $B_0$  gradients at different human body model resolutions. The computational times are divided between the dipolar field kernel calculation ( $t_1$ ) and dynamic  $B_0$  update ( $t_2$ )

Voxel size (cm)		0.1	0.2	0.3	0.4	0.5	0.6	0.7	0.8	0.9	1.0
Torso slice indices	Start slice*	1100	550	367	275	220	183	157	138	122	110
	End slice	1560	780	520	390	312	260	223	195	173	156
Head slice indices	Start slice	1620	810	540	405	324	270	231	203	180	162
	End slice	1780	890	593	445	356	297	254	223	198	178
$t_1$ (s)		12.045	1.808	0.471	0.241	0.127	0.0725	0.0451	0.0269	0.0224	0.0181
$t_2$ (s)		77.398	11.874	2.529	1.346	0.604	0.4022	0.1872	0.1315	0.1086	0.0809
$G_x$ (Hz/cm)**		0.0076	0.0139	0.0053	0.0231	0.0105	-0.0269	-0.0226	-0.0297	0.0190	0.0620
$G_y$ (Hz/cm)		-0.1824	-0.1951	-0.2027	-0.2339	-0.2127	-0.2856	-0.2705	-0.2738	-0.2701	-0.0532
$G_z$ (Hz/cm)		-0.2619	-0.2413	-0.2133	-0.1819	-0.1555	-0.1383	-0.1325	-0.0538	-0.0491	-0.0004

\* Slice index rises in the foot-to-head direction.

\*\*  $B_0$  gradients were calculated by linear function fitting in the x (left to right), y (posterior to anterior), and z (foot to head) directions.





**Fig. 4.** Mean  $B_0$  shift gradients in the brain at the 5<sup>th</sup> frame (fully-inhaled), computed at 0.1 cm voxel resolution, along the x (left-to-right, a), y (posterior-to-anterior, b), z (foot-to-head, c) directions. The blue traces correspond to  $B_0$  values averaged over slices normal to each direction. The straight lines indicate linear fitting to the data.

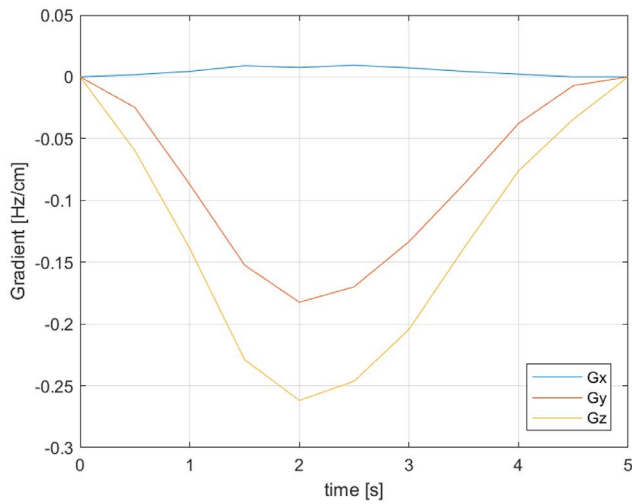
Figure 5 shows the temporal variation of the spatial gradients of the respiration-induced  $B_0$ , determined by volumetric linear fitting. Note that the gradient values at 2 s (frame 5) are not the same as the slopes of the fitted lines in Figure 4, because of the different fitting methods. The peak gradients (at frame 5) in Figure 5 are +0.008, -0.182, and -0.262 Hz/cm for x, y, z, respectively. The magnitude of these values can be compared to those of Vannesjo et al. (9), who also used volumetric fitting. In Figure 2 of that study, the linear dynamic gradient terms are displayed as the field at 10 cm from the isocenter. Also, the x and y direction definitions were swapped compared to ours. Given these facts, the data in that study are approximately in agreement with our data, showing about 0.2 Hz/cm variation for anterior-posterior and feet-head directions and negligible gradient in the left-right direction. The feet-head gradient ( $G_z$ ) variation was slightly larger than the anterior-posterior gradient ( $G_x$ ) in the study by Vannesjo et al. (9), while the

two gradient variations had the same sign, which is also consistent with our result.

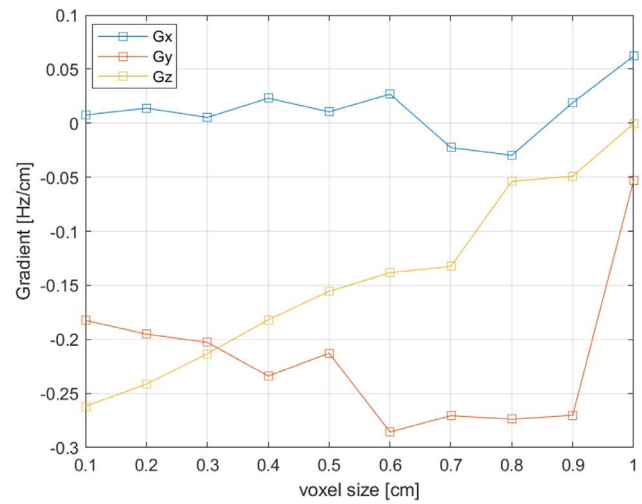
### Voxel Size Dependence

Table 1 shows that the computational time decreased dramatically as the voxel size increased. Both dipolar field kernel calculation ( $t_1$ ) and dynamic  $B_0$  update ( $t_2$ ) times scaled approximately inversely with the voxel volume (plot not shown). Applying gSVC to the physically separated source (torso) and target (head) grids allowed time-efficient computation of the  $B_0$  shift at high (0.1 cm) isotropic resolution, taking about 90 seconds ( $t_1 + t_2$ ) for the 10 frames.

While reduced spatial resolution greatly speeds up the computation, Figure 6 shows the increased problem (lack of reliability) associated with simulation at large voxel sizes. At voxel sizes larger than 0.3 cm, it was observed that the  $G_y$  magnitude surpassed that of  $G_z$ , which was not consistent with the literature (8, 9). Furthermore, for voxel sizes greater



**Fig. 5.** Plot of  $B_0$  shift gradients as a function of time at 0.1 cm voxel resolution. The gradient values were obtained by volumetric linear fitting at each time point. The maximum gradient amplitudes in the  $y, z$  directions are observed at 2 s (5<sup>th</sup> frame). Gradient in the  $x$ -direction shows little change.



**Fig. 6.** Plot of  $B_0$  shift gradients at frame 5 relative to frame 1 as a function of the voxel size. The gradient values become increasingly more erratic as the voxel size increases, indicating reduced reliability at large voxel sizes (larger than about 0.5 cm).

than 0.5 cm, the  $B_0$  gradient values became more erratic, departing significantly from the high-resolution cases.

## DISCUSSION

We have utilized a high-resolution 4D human body model and a computationally efficient  $B_0$  calculation method (gSVC) to simulate the respiration-induced  $B_0$  shift in the brain. Compared to previous simulation work (11), we implemented up to  $7.6^3 = 439$  times smaller voxel volumes, thereby capturing details of organ movements associated with respiration. The human body model used allows for the variation of many anatomical and physiological parameters which can be used for future investigations involving different breathing conditions.

We found that respiration with a moderate breathing depth (diaphragm movement of 2.0 cm) caused an average  $B_0$  shift in the head of up to about 7.2 Hz at 7T. The spatial distribution of the induced  $B_0$  shift showed mainly a  $z$ -directional gradient, accompanied by a significant  $y$ -directional gradient in our model. These are in agreement with a previous report (9) based on direct field-probe measurements around the head.

We also examined the accuracy of the results as a function of the model spatial resolution. The results showed that, as the voxel size increased, the spatial gradient of the

$B_0$  shift substantially changed for voxels larger than 0.5 cm. This indicates that at such a large voxel size, the breathing anatomy is probably not adequately captured to produce a reliable  $B_0$  prediction in the head. While our data are limited to a particular human body model, the observed voxel size dependence provides a guideline for future investigations of respiration effects using different susceptibility models and computational methods.

The present work had several limitations. First, the susceptibility model was simplified. In particular, susceptibility changes in the airway accompanying oxygen concentration change were ignored. We believe that this is justified based on the study (19) which showed that the respiration-induced  $B_0$  shift was mainly due to geometry and not air susceptibility changes. Second, we ignored cardiac motion and the associated blood flow. The cardiac motion effect was simulated by Lee et al. (20) and was shown to be much smaller than what we observed here. The effect of blood flow is expected to produce  $B_0$  perturbation close to the large vessels and might not significantly affect the global  $B_0$  gradient, as was calculated here for the whole brain. Finally, we used only a single specific dynamic human body model (using a specific set of parameters within XCAT). This work was primarily aimed at demonstrating a new numerical simulation platform for calculating the respiration-induced  $B_0$  shift, and we defer studies on different body models and parameters to future

research. Any findings regarding dependence on the model parameters (such as body mass index) should ultimately be confirmed experimentally.

In conclusion, we showed that the respiration-induced  $B_0$  shift in the head can be computed with high spatiotemporal resolution utilizing a detailed dynamic human body model and an efficient  $B_0$  calculation algorithm. We have greatly improved the fidelity of the human body model used in the simulation compared to a previous study and confirmed aspects of previous experimental observations. Answering specific questions regarding parametric dependence can be pursued in future research to better guide dynamic shim strategies for ultra-high field neuro MRI.

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