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Correcting gradient chain induced fat quantification errors in radial multi-echo Dixon imaging using a gradient modulation transfer function

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Purpose: Multi-echo Stack-of-stars (SoS) radial k-space trajectories with golden angle ordering are becoming popular for free-breathing abdominal Dixon imaging and proton density fat fraction (PDFF) mapping. Gradient chain imperfections including eddy currents and system delays are known to affect the image quality of radial imaging and to confound the estimation of PDFF mapping. This work proposes a retrospective trajectory correction method based on a simple gradient modulation transfer function (GMTF) measurement to predict and correct gradient chain induced k-space trajectory errors.

Methods: The GMTF was measured using the standard hardware of a 3 Tesla scanner on a phantom using the thin slice method and was applied to a 3D radial SoS Dixon imaging sequence. The impact of the GMTF-based correction on image reconstruction and PDFF quantification was investigated using numerical simulations and validated on experimental phantom data as well as on *in vivo* leg and liver data of healthy volunteers.

Results: Correcting the k-space trajectories with the measured GMTF during image reconstruction reduced PDFF quantification errors for phantom and *in vivo* acquisitions. A Bland-Altman comparison of the measured PDFF phantom and reference data confirmed that the GMTF correction narrowed down the limits of agreement (LoA) from $1.3\% \pm 8.1\%$ (uncorrected) to $1.9\% \pm 5.4\%$ (GMTF-corrected) over the full PDFF range (0%–100%) and from $-0.26\% \pm 2.8\%$ (uncorrected) to $0.12\% \pm 1.5\%$ (GMTF-corrected) within the 0%-50% PDFF range. Liver PDFF estimation was improved by reducing the standard deviation of the mean liver PDFF and the bias of the mean liver PDFF for all subjects.

Conclusion: The proposed GMTF-based k-space trajectory correction is a fast alternative method for avoiding PDFF quantitation errors caused by gradient-system induced k-space trajectory errors in 3D radial multi-echo gradient-echo acquisitions.

KEYWORDS

eddy currents, gradient delays, k-space trajectory distortions, radial, stack-of-stars, proton density fat fraction (PDFF), Dixon imaging, magnetic resonance imaging (MRI)

1 Introduction

Chemical shift encoding based water-fat separation techniques have been emerging for efficient fat suppression and fat quantification across organs [1, 2]. By acquiring the MR signal at different echo times, the so-called Dixon techniques encode water and fat in the time domain. The water and fat images can be estimated by fitting a signal model that accounts for confounding factors such as the T₁ bias [3, 4], the T₂* decay [5–7], the spectral complexity of fat [8], the noise bias [3] and eddy currents [9, 10] in a way that consequently a quantitative proton-density fat fraction (PDFF) map can be calculated. In abdominal MRI, the liver PDFF is an established biomarker for assessment of hepatic steatosis [11, 12] and is being used also beyond routine diagnostic imaging in lifestyle intervention studies [13].

While abdominal PDFF mapping is performed conventionally with Cartesian k-space trajectories [6–10], abdominal Dixon acquisitions can suffer from high motion sensitivity. They are therefore typically performed during a breath-hold, which limits achievable volumetric coverage, spatial resolution, or signal-to-noise ratio. This complicates the application of Dixon methods in subjects unable to suspend respiration, such as sick, elderly, or pediatric patients. Radial acquisitions are more robust to motion than Cartesian acquisitions, since they sample the k-space center continuously [14]. Radial stack-of-stars acquisitions enable the continuous sampling of the k-space center in 3D imaging and have been combined with Dixon imaging [15]. There have been multiple recent works proposing free-breathing motion-averaged or motion-resolved multi-echo radial SoS acquisitions to quantify PDFF in a free-breathing mode [16–21].

Due to varying readout directions, radial acquisitions are more sensitive to phase errors induced by gradient chain imperfections than their Cartesian counterpart. The gradient chain connects the input of the three physical x-, y- and z-waveforms to the gradient fields in the scanner bore and consists of pre-emphasis for eddy current compensation, the three gradient amplifiers, and respective gradient coils. Gradient system imperfections, more precisely eddy currents, gradient system delays and vibrations, or an inadequate pre-emphasis, cause k-space trajectory errors in radial acquisitions, resulting in deteriorated image quality [22-24]. It is already known from Cartesian acquisitions that phase errors induced by eddy currents cause a bias in abdominal fat quantification when using complex-based [25] chemical shift encoding based water-fat separation techniques [26-28]. Thus, when radial k-space trajectories are used for chemical shift encoding based water-fat separation, gradient chain imperfections are expected to affect the image quality of the multi-echo images and to induce bias in the PDFF quantification.

Eddy current correction techniques either use specialized NMR field probes to measure the gradients directly [29–31], perform separate calibration scans [32–35] or estimate the errors directly from the measured raw data without any additional calibration scans [36–39]. It had also been shown that eddy currents can be corrected by measuring the transfer function that describes the gradient system characteristics. In the context of radial water–fat imaging, most previous methods relied on the acquisition of calibration lines for each gradient axis with opposite polarity to determine the gradient timing errors [15–21]. However, the acquisition of such

calibration lines prolongs the acquisition time and must be repeated for each scan. It has been also demonstrated that a data-driven correction can estimate k-space shifts from radial raw data [40] and be applied in the context of PDFF mapping using a radial SoS acquisition [41]. A more general approach to correcting the errors induced by the gradient system imperfections can be achieved with the use of a gradient impulse response function (GIRF) [32, 42] or its Fourier transform, the gradient modulation transfer function (GMTF), under the assumption that the gradient system behaves linearly and time invariant. The GMTF measurement enables full characterization of the scanner gradient system chain in frequency space. Once the GMTF is known, arbitrary gradient waveforms from any pulse sequence can be corrected. The GMTF itself can be measured either with special NMR field probes [29, 30, 42, 43], or using the thin slice method in simple phantoms [32, 33], [44-46]. It has been shown that a GMTF-based trajectory correction for gradient imperfections improves image quality in non-Cartesian spiral acquisitions [32, 47]. However, it has not been applied to abdominal radial SoS Dixon imaging yet.

The purpose of this work is to propose a correction for gradient chain imperfections in abdominal radial SoS-based PDFF mapping based on a GMTF measurement, using the thin slice method without the need of any additional hardware. The impact of the GMTF correction on the PDFF quantification is investigated in simulations and validated in phantom and *in vivo* scans.

2 Materials and methods

2.1 GMTF theory and measurement

Each linear time invariant system can be fully characterized by its impulse response function, which describes the time-domain response of the system to an idealized point impulse [48]. The Fourier transform of this impulse response function yields the modulation transfer function. For most applications the MRI gradient chain can be treated as a linear time invariant system and as such it can be described by the GIRF h(t) [49]. Convolving the input gradient of the MR system with the GIRF yields the real gradient as it is played out by the scanner [50]:

$$g_{real}(t) = \int_{-\infty}^{+\infty} h(t-\tau) \cdot g_{input}(\tau) d\tau.$$

Written in the frequency domain the convolution becomes a multiplication:

$$G_{real}(\omega) = H(\omega) \cdot G_{input}(\omega).$$

where $H(\omega)$ is the gradient modulation transfer function, which is the GIRF's Fourier transform [45].

The measurement of the GMTF/GIRF itself is based on a thin slice method by Rahmer et al [45] which was simplified by Kronthaler et al [50] for the application in radial imaging. The thin slice method is based on the work by Duyn et al [33], where it was shown that the excitation of a thin slice, generating a 1D virtual probe, is sufficient to measure the behavior of the gradient system without any additional hardware. Applying a test gradient along the direction of the slice selection of the 1D probe allows the measurement of the response to the test gradient. Neglecting B_0 drift, through-slice dephasing and assuming homogeneous fields in *y* and *z* direction, the following signal model describes such a 1D probe GIRF measurement in the *x*-direction, with the distance x_0 of the excited slice from the isocenter,

$$s(x_0,t) = \rho_{slice} e^{-i\phi(x_0,t)}$$

i.e.,

$$\phi(x_{0},t) = \Delta\omega(x_{0})t + \sum_{j}k_{j}(t)b_{j}(x_{0}) + \phi_{c}(x_{0},t) + \phi_{other}(x_{0},t).$$

The evolution of $\phi(x_0, t)$ is governed by the frequency offset $\Delta\omega(x_0)$ resulting from static off-resonance, the linear response to the applied test gradient $\sum k_j(t)b_j(x_0)$, the phase due to concomitant gradients $\phi_c(x_0, t)$, and the phase $\phi_{other}(x_0, t)$ accumulated due to any other fields that do not result from the applied gradient, e.g., slice selection eddy currents. The linear response to the applied test gradient can be described by a sum over spatial basis functions b_j with coefficients k_j [51]. Using spherical harmonics as spatial basis functions up to j = 15 with $b_j(\mathbf{r}) = 1, x, y, z, 6xy, 3yz, \frac{1}{2}(3z^2 - \mathbf{r}^2), 3xz, 3(x^2 - y^2), 15y(3x^2 - y^2), 30xyz, \frac{3}{2}(5z^2 - \mathbf{r}^2), \frac{1}{2}z(5z^2 - \mathbf{r}^2), \frac{3}{2}x(5z^2 - \mathbf{r}^2), 15(x^2 - y^2)z, 15x(x^2 - 3y^2)$ the first terms of the sum are given by

$$b_0(\mathbf{r}) = 1 \leftrightarrow k_0(t) = \gamma \int_0^t \Delta B_0(t') dt'$$

$$b_1(\mathbf{r}) = x \leftrightarrow k_1(t) = k_x(t)$$

$$b_2(\mathbf{r}) = y \leftrightarrow k_2(t) = k_y(t)$$

$$b_3(\mathbf{r}) = z \leftrightarrow k_3(t) = k_z(t).$$

The spatial 0th order term is related to temporal variations in the B₀ field and the first order terms to temporal variations of the field gradients. k_1 , k_2 and k_3 can be identified as the k-space coordinates k_x , k_y and k_z . A 1D measurement in the *x*-direction corresponds to $\mathbf{r}_0 = (x_0, 0, 0)^T$.

To get rid of confounding signal contributions (ϕ_c and ϕ_{other}), not related to the test gradient, the measurement is performed twice with positive $\phi_+(x_0,t)$ and negative $\phi_-(x_0,t)$ polarity of the test gradient, respectively [44]. The difference of these measurements yields the linear response related only to the test gradient. The phase difference is given by

$$\Delta \phi(x_{0},t) = \phi_{+}(x_{0},t) - \phi_{-}(x_{0},t) = 2 \sum_{j=0,1,\dots} k_{j}(t) b_{j}(x_{0}).$$

In order to measure the spatial variation of the system response, several slices at different off-center locations are excited. The measured phase is then expanded into polynomials up to the second order [46] given by

$$\Delta\phi(x,t) \approx \Delta\phi_0(t) + k_x(t)x + k_{2x}(t)x^2.$$

The real gradient applied along the *x*-axis can be then calculated using the first-order component of the fitted phase.

$$g_{real,x}\left(t\right) = \frac{1}{\gamma}\frac{d}{dt}k_{x}\left(t\right)$$

The GMTF is calculated in the Fourier domain as follows [42, 52].

$$H_{x}(\omega) = \frac{G_{real,x}(\omega)}{G_{input,x}(\omega)}$$

Rahmer et al. extended the GMTF measurement to 3D by adding a phase encoding gradient. They reported that a 1D GMTF measurement is sufficient since second order components were often negligible. Thus, a simplified measurement technique was used in this work to estimate the first order response by measuring the signal from 4 parallel offcenter slices per gradient axis [50].

A spherical phantom (diameter 166 mm, volume 2 liters), containing CuSO₄-doped water (relaxation time values at 3 Tesla: $T_1\approx 280$ ms and a $T_2\approx 240$ ms) was used to measure the GMTF. Four slices of 1.5 mm thickness were excited with an excitation pulse with length 1.6 ms, a maximum amplitude of 5.45 μT and a flip angle of 45° . The four slices were located at distances of -26.25 mm, -8.75 mm, 8.75 mm, and 26.25 mm from the isocenter. A chirp waveform was chosen as the input test gradient with a frequency range of 0.1--10.0 kHz with an acquisition window of 80 ms (Figure 1). The measurement was performed separately along each gradient axis.

2.2 Pulse sequence and image reconstruction

The performance of the GMTF correction method was demonstrated for PDFF mapping using a radial SoS acquisition employing a multi-echo readout (Figure 2A) on a 3 T system (Elition X, Philips Healthcare, Best, the Netherlands). The spokes in the k_x - k_y -plane were acquired in a pseudo golden angle ordering [53], which provides a homogeneous distribution of the acquired spokes in k-space while still relying on a golden angle increment of 111.25°. Radial spokes with the same azimuthal angle were acquired for all k_z increments prior to azimuthal angle rotation. The Cartesian k_z -dimension was aligned with the B₀-direction. SENSE acceleration was also applied in the z-dimension for scan time reduction [54].

The GMTF correction method was implemented as a retrospective correction of the trajectory during the image reconstruction process [43, 47, 50, 52]. Before applying the GMTF correction to any logical gradients, the gradients were transformed to the physical gradient system by linear transformations. After applying the GMTF, the inverse transformation was used to calculate the corrected logical gradients. To obtain the exact k-space positions of the acquired data points, the real gradient waveform was predicted by convolving the input gradient waveform with the measured GIRF in the time domain. This convolution operation was performed as a multiplication of the Fourier-transformed input gradient waveform of a whole repetition time with the GMTF, thus providing the corrected gradient waveform of all six echoes (Figure 2B). Integration of this gradient waveform yielded the GMTF-corrected trajectory and a gridding-based non-Cartesian image reconstruction was performed subsequently. The data was gridded in the k_x-k_v-plane with the corresponding trajectories using a Kaiser-Bessel kernel of width 4 k-space sampling steps and a



gridding oversampling factor of 1.25. The gridding operation, the Fourier transform in 3D and the SENSE unfolding in the third Cartesian-sampled z-dimension were performed using an image reconstruction toolbox (ReconFrame, Gyrotools, Switzerland). The same toolbox was also used to access all necessary information about the gradient objects for each sequence from the raw data to reconstruct the gradient waveform, which will be used during the GMTF correction.

An additional data-driven correction method for gradient delays was implemented to compare it with the performance of the proposed GMTF correction. The intrinsic design of the SoS trajectory allows for a data-driven correction of gradient delay induced k-space shifts. Spokes of approximately opposite readout polarity within the kx-ky-plane were used to estimate these k-space shifts [24, 40]. Given that a constant shift in k-space can be represented as a linear phase term in image space, according to the Fourier shift theorem, the k-space shift was determined by fitting the linear phase in image space [28]: Therefore, the signal of each spoke was correlated in image space with the signal of the corresponding spoke with opposite readout direction to estimate a linear phase offset. Following that, the estimated linear phase offset was applied to each spoke in image space. It must be noted that this spoke alignment method corrects the k-space signal by shifting the signal along the frequency encoding direction while not changing the k-space trajectory. The spoke alignment method can be thus simply performed prior to the gridding routine.

2.3 PDFF simulation

A simulation was performed to investigate the impact of gradient chain induced trajectory errors on PDFF quantification. Therefore, the correction of a six-echo gradient-echo acquisition in a water-only spherical numerical phantom was investigated. K-space data was simulated from a numerical water phantom via a NUFFT operation based on the GMTF-corrected trajectory to generate the reference data. Gradient-chain system errors were modeled by performing a NUFFT reconstruction of this data with the uncorrected ideal trajectory which does not account for any gradient imperfections. A reference reconstruction was performed with the GMTF-corrected trajectory and both reconstructed data sets were passed through a water-fat separation pipeline to generate PDFF maps. All NUFFT operations were performed via the NUFFT implementation from BART [55] and all sequence parameters and the trajectories were selected based on the water phantom measurements (for a detailed list of parameters, please refer to the subsection "Phantom measurements" below). A T2*- decay of 20 ms was assumed for the water signal during signal simulation. PDFF maps were obtained via a complex-based water-fat separation assuming a single T_2^* decay and the same multi-peak fat spectrum that is used for processing the measured phantom data. The water-fat separation problem was solved based on a variable-layer single-min-cut graph-cut algorithm [56]. The GMTF-corrected and uncorrected PDFF maps were compared qualitatively.



GMTF-based trajectory error estimation. (A) The six-echo gradient-echo radial stack-of-stars sequence with pseudo golden angle ordering used for PDFF mapping. (B) GMTF correction of the input gradient waveform for one TR. The gray areas mark the acquisition windows of the single echoes. Zoomed-in views of the second and third echo are shown on the right highlighting the gradient chain induced deviations at the beginning of each echo readout. (C) Corresponding corrected trajectory errors for the x-encoding direction for one TR, based on the GMTF-corrected gradient waveform. The echo times are marked by the vertical black lines. Zoomed in views of echo 2 and 3 are shown on the right in (B) and (C).

2.4 Phantom measurements

Two different phantoms were scanned to evaluate the performance of the GMTF correction. First, a simple water bottle phantom was scanned to replicate the simulation setup, using a 3D six-echo gradientecho radial SoS sequence with pseudo golden angle ordering. The exact sequence parameters were TE = 1.40 ms/2.44 ms/3.47 ms/4.51 ms/5.55 ms/6.59 ms, TR = 10.40 ms, flip angle = 3° , in-plane resolution = $1.5 \times 1.5 \text{ mm}^2$, slice thickness = 3.0 mm, FOV = $450 \times$ $450 \times 159 \text{ mm}^3$, 391 spokes, 300 samples per spoke, 53 slices, bandwidth = 1613 Hz/pixel, scan time of 5 min 13 s. Second, a PDFF phantom (Calimetrix, Madison, USA) containing 15 vials with different fat fractions ranging from 0% to 100% (PDFF reference values of the vials [%]: 0.1 ± 0.6 ; 3.5 ± 0.5 ; 6.2 ± 0.6 ; 9.0 ± 0.5 ; 12.1 ± 0.5 ; 17.3 ± 0.5 ; 23.0 ± 0.5 ; 33.4 ± 0.5 ; 43.2 ± 0.5 ; 53.3 ± 0.6 ; 63.4 ± 0.9 ; 73.1 ± 1.0 ; 82.5 ± 1.1 ; 90.8 ± 1.2 ; 100.2 ± 0.7) was scanned in axial orientation with the following parameters: TE = 1.48 ms/2.55 ms/3.61 ms/4.68 ms/5.75 ms/6.82 ms, TR = 10.31 ms, flip angle = 3°, in-plane resolution = $1.7 \times 1.7 \text{ mm}^2$, slice thickness = 3.0 mm, FOV = $250 \times 250 \times 201 \text{ mm}^3$, 193 spokes, 148 samples per spoke, 67 slices, bandwidth = 2222 Hz/pixel, scan time of 3 min 13 s.

Both phantom scans were reconstructed with the original k-space trajectory and the GMTF-corrected k-space trajectory. The PDFF phantom data was additionally corrected with the spoke alignment method for comparison. Subsequently, PDFF maps were calculated based on a complex-based water-fat separation with a variable-layer single-min-cut graph-cut algorithm. The magnitude discrimination approach was employed to minimize noise bias effects [3]. A peanut oil-based fat model was used to fit the signal of the phantom acquisitions to account for temperature shift and the fat composition of the PDFF phantom [57].

The PDFF reference values of the PDFF phantom were obtained with a Cartesian 6 echo time-interleaved multi-echo gradient echo sequence with two interleaves and monopolar gradient readouts [28] and the following parameters: TE₁ = 1.24 ms, delta TE = 1.0 ms, TR = 7.8 ms, flip angle = 3.0° , resolution = $1.5 \times 1.5 \times 1.5 \text{ mm}^3$, FOV = $190 \times 190 \times 190 \text{ mm}^3$, bandwidth = 1415.2 Hz/pixel, scan duration = 3 min 38 s. The PDFF maps were calculated based on the scanner-reconstructed multi-echo images after phase-correction [28] with a complex-based water-fat separation with a variable-layer single-min-cut graph-cut algorithm.

2.5 In vivo measurements

The effect of the GMTF correction on PDFF quantification was also investigated in two *in vivo* anatomies, the liver and the upper thighs, as a reference for an anatomy without respiratory motion effects. Therefore both anatomies were scanned with the proposed SoS sequence and a low resolution Cartesian reference scan. The radial data was corrected with the proposed GMTF correction and the data-driven spoke alignment method respectively. PDFF maps of the corrected and uncorrected radial data sets were compared with the Cartesian reference PDFF maps qualitatively and via a ROI analysis for the liver data.

In vivo imaging of the liver and the upper thighs was performed on 11 healthy volunteers after approval by the institutional review board (Klinikum rechts der Isar, Technical University of Munich, Munich, Germany) and informed written consent by each subject. Imaging of the liver and the legs was performed axially with a 3D high resolution six-echo radial SoS sequence with pseudo golden angle ordering. The Cartesian z-dimension of the SoS trajectory was aligned with the patient's feet–head axis.

The thighs were scanned with the radial SoS sequence using the following parameters: TE = 1.40 ms/2.43 ms/3.47 ms/4.51 ms/5.55 ms/6.59 ms, TR = 10.40 ms, flip angle = 3° , in-plane resolution = $1.5 \times 1.5 \text{ mm}^2$, slice thickness = 3.0 mm, FOV = $450 \times 450 \times 201 \text{ mm}^3$, 391 spokes, 300 samples per spoke, 42 slices, bandwidth = 1634 Hz/pixel, scan time of 4 min 33 s, and SENSE acceleration factor of 1.6 in the Cartesian-sampled dimension.

Liver radial SoS imaging was performed using the following parameters: TE = $1.38 \text{ ms}/2.41 \text{ ms}/3.44 \text{ ms}/4.48 \text{ ms}/5.51 \text{ ms}/}$ 6.55 ms, TR 10.0 ms, flip angle = 3°, in-plane resolution = $1.5 \times 1.5 \text{ mm}^2$, slice thickness = 3.0 mm, FOV = $450 \times 450 \times$ 201 mm^3 , 391 spokes, 300 samples per spoke, 42 slices, bandwidth = 1603 Hz/pixel, average scan time of 8 min 10 s, and SENSE acceleration factor of 1.6 in the Cartesian-sampled dimension. Breathing motion was compensated with a 1D-pencil beam navigator, placed in the feet-head direction within the liver dome and the diaphragm.

Both anatomies were scanned additionally with a low resolution 3D Cartesian Dixon sequence which is clinically used for PDFF assessment in the liver with the following parameters: TE = 0.92 ms/1.56 ms/2.21 ms/2.85 ms/3.50 ms/4.14 ms, TR = 10.0 ms, flip angle = 3°, in-plane resolution = $3.2 \times 3.2 \text{ mm}^2$, slice thickness = 6.0 mm, FOV = $450 \times 450 \times 198 \text{ mm}^3$, scan time of 17 s (to fit in a single breath-hold for the liver acquisition). The Cartesian breath-hold scans were reconstructed and the water-fat separation steps were performed with the manufacturer's online implementation.

The radial SoS scans were reconstructed offline without any trajectory correction, with the spoke alignment method and with the proposed GMTF correction method. PDFF maps were estimated with a complex-based water-fat separation with a variable-layer single-min-cut graph-cut algorithm. 9-peak fat models were used for the PDFF calculation of the liver [58] and leg [59] data sets.

Mean value and standard deviation of the corrected and uncorrected liver PDFF distributions were determined based on 4 ROIs (area = 4 cm^2) placed at right, left, anterior and posterior positions of the liver [60].

3 Results

3.1 GMTF measurement and trajectory correction

Figure 1A shows the measured first order response to the chirp waveform test gradient for the x-gradient axis. Based on the response, the magnitude (Figure 1B) and phase (Figure 1C) of the GMTF $H(\omega)$ were calculated for each gradient system axis respectively of the 3.0 T system up to a frequency of 10 kHz. The GMTF magnitudes showed similar low pass filter characteristics of the gradient chain for all three gradient dimensions. While the xand y-directions showed a similar GMTF, the GMTF in the z-direction has a higher transfer ratio in the range of 3-10 kHz. This effect can be explained by the different design of the z-gradient coil compared to the x- and y-gradient coils. Single peaks in the low frequency part of the GMTF, for example, at 1.2 kHz, can correspond to mechanical resonances of the gradient system. The phases of the three GMTFs began to deviate above 2 kHz indicating anisotropic delays for the gradient system. The noise level increased in the GMTF measurements with increasing frequency. Since higher frequencies were acquired at the end of the chirp impulse (Figure 1A), the gradient strength was already reduced, resulting in smaller SNR.

Figure 2B shows the result of the convolution of the gradient input waveform with the measured GMTF for the x-gradient direction. Large eddy current induced deviations of the GMTFcorrected waveform from the ideal input waveform were visible during the ramp-up process of each gradient and at the beginning of each echo readout. These errors were caused by short term eddy currents that decayed after the plateau of each gradient was reached. Long term eddy currents were not visible within one TR. Figure 2C shows the resulting trajectory errors for an exemplary radial spoke in the k_x-direction, calculated based on the different sampling positions obtained with the GMTF-corrected gradient waveform and the uncorrected input waveform. The trajectory error varied during the echo readout: The largest errors were observed at the beginning and the end of each readout, where high spatial frequencies were sampled, reaching shifts of up to one k-space sampling step. Varying shifts were observed along the spoke and at the k-space center, while comparing odd and even echoes of the multi-echo acquisition.

3.2 PDFF simulation results

The simulated impact of the trajectory errors propagating into the PDFF quantification is presented in Figure 3. Figure 3A shows the PDFF maps obtained from the simulated six-echo data of a water-only (PDFF = 0) sphere (Figure 3B) reconstructed with the uncorrected trajectory and the GMTF-corrected trajectory. The PDFF map obtained from the uncorrected data showed an inhomogeneous PDFF distribution. PDFF values reached values higher than 2% in the center of the sphere, while the values dropped significantly symmetrically towards the edge of the object to negative PDFF values as low as -8%. In contrast, the corrected PDFF map had a homogeneous distribution around zero as expected, showing only light reconstruction artifacts, stemming



FIGURE 3

Numerical simulations of gradient-chain induced effects. (A) PDFF maps obtained from simulated six-echo data on a water sphere (B) generated with the GMTF-corrected trajectory and reconstructed with the corrected trajectory (upper left) and uncorrected trajectory (upper right). (C) shows PDFF line profiles along the dashed lines in (A). The trajectory errors cause PDFF estimation errors of up to 2% PDFF in the center and up to -8% PDFF at the edge of the numerical phantom.



FIGURE 4

PDFF results of a water bottle phantom scan using a six-echo radial SoS acquisition. The PDFF map obtained from the uncorrected data (left) is inhomogeneous with a severe negative PDFF bias at the bottom right edge of the phantom. The GMTF correction results in a more homogeneous distribution and improves the negative PDFF bias at the bottom edge (right).



GMTF correction results of the PDFF phantom acquisition. (A) uncorrected, GMTF-corrected and spoke alignment-corrected PDFF maps. (B) linear regression comparison of PDFF values obtained from uncorrected data (left), GMTF-corrected data (center) and data corrected with the spoke alignment method (right) to the reference values. (C) corresponding Bland-Altman comparison of PDFF values to the reference values. The dashed line represents the identity in the linear regression plot, the solid black line the linear fit to the data. Bland-Altman plots: the solid black line marks the mean difference (MD), the dashed lines show the limits of agreement (LoA = MD \pm 1.96 standard deviation (SD)) and RPC is the reproducibility coefficient (1.96 SD).

from the NUFFT operations. For example, ringing artifacts were visible at the edge of the phantom or the outer FOV. PDFF line profiles across the phantom are plotted in Figure 3C.

3.3 Phantom results

Figure 4 shows PDFF maps of a water bottle phantom scan that were obtained from data reconstructed with the uncorrected trajectory and with the GMTF-corrected k-space trajectory. Using the uncorrected gradient waveform in the reconstruction process resulted in an inhomogeneous PDFF distribution with a positive PDFF bias towards the upper edge of the phantom and negative PDFF values around –5% PDFF at the bottom edge. In contrast, the reconstruction with the GMTF-corrected trajectory reduced the severe negative PDFF underestimations at the bottom edge, yielding a more homogeneous PDFF map than the uncorrected reconstruction.

The results of the GMTF correction applied to the PDFF phantom scan are presented in Figure 5. Figure 5A displays the PDFF maps estimated with and without the GMTF correction applied during the image reconstruction step as well as PDFF maps from data corrected with the spoke alignment method. The visualization in the 0%–100% PDFF range revealed errors for the high PDFF fraction vials, especially the 82.5% and the 90.8% vial were over- or underestimated, respectively, when no GMTF correction was applied. Linear regression plots over the full 0%–100% PDFF range for uncorrected and corrected data are shown in Figure 5B and the corresponding Bland-Altman plots in Figure 5C. The linear regression test between the uncorrected data and the reference PDFF values yielded a coefficient of determination $r^2 = 0.990$. According to the Bland-Altman



FIGURE 6

Analysis of the GMTF correction results the PDFF phantom acquisition within the 0%-50% PDFF range. (A) uncorrected, GMTF-corrected and spoke alignment-corrected PDFF maps from Figure 5 in a tighter PDFF window highlight negative PDFF quantification errors in the water area of the phantom and the 0% vial. (B) linear regression comparison of PDFF values obtained from uncorrected data (left), GMTF-corrected data (center) and data corrected with the spoke alignment method (right) to the reference values for vials with PDFF in the range 0%-50%. (C) corresponding Bland-Altman comparison of PDFF values to the reference values for vials with PDFF in the range 0%-50%. The dashed line represents the identity in the linear regression plot, the solid black line the linear fit to the data. Bland-Altman plots: the solid black line marks the mean difference (MD), the dashed lines show the linits of agreement [LoA = MD \pm 1.96 standard deviation (SD)] and RPC is the reproducibility coefficient (1.96 SD).

analysis the mean difference (MD) was 1.3% with limits of agreement (LoA) of MD \pm 8.1%.

Applying the GMTF correction improved the linear correlation between the corrected PDFF values and the reference values with a coefficient of determination of $r^2 = 0.999$. The LoA were narrowed down to $1.9\% \pm 5.4\%$. The correction with the spoke alignment method yielded a linear correlation to the reference PDFF values with a coefficient of determination of $r^2 = 0.999$. The LoA of the Bland-Altman analysis after the correction with the spoke alignment method were $2.4\% \pm 4.2\%$. Absolute PDFF values and errors for the PDFF phantom comparison are listed in the Supplementary Table S1.

Since PDFF vials with a PDFF above 50% showed larger errors in the uncorrected and corrected case than the vials below 50%, the same statistical analysis was performed separately for the lower PDFF range of 0%–50% containing 9 vials (see Figure 6). When displaying the corrected and uncorrected PDFF maps in a tighter intensity window (see Figure 6A), PDFF variations in the water-only bath around the fat fraction vials became visible. PDFF values of up to -10% were observed for the uncorrected case in the water bath region of the phantom at the right bottom edge and in locations between the high PDFF vials. Applying the GMTF correction improved the quality of the PDFF map. Linear regression comparisons to the PDFF reference values within the 0%-50% range are shown for the uncorrected, GMTF corrected and spoke alignment-corrected data in Figure 6B and the corresponding Bland-Altman comparison results in Figure 6C. Applying the GMTF correction improved the linear regression test from $r^2 = 0.99$ to $r^2 = 1.00$ in the 0%–50% PDFF range. The Bland-Altman analysis in the same range confirmed an improvement of the PDFF maps by the GMTF correction. MD of the uncorrected data was -0.26% with LoA of MD \pm 2.8% and the GMTF correction achieved MD = 0.12%



FIGURE 7

PDFF map comparison of the upper thigh of a healthy volunteer. PDFF maps obtained from uncorrected images (top left), GMRF-corrected images (top right) and images corrected with the spoke alignment method (bottom left). Bottom right: PDFF map from a low-resolution Cartesian acquisition for comparison. Negative PDFF estimation errors and inhomogeneous PDFF distributions in the muscle tissue of the uncorrected data (indicated by red arrows) are corrected with both the GMTF and the spoke alignment method.

with LoA of MD \pm 1.5%. The spoke alignment method equally improved the linear regression test from $r^2 = 0.99$ to $r^2 = 1.00$ in the 0%–50% PDFF range. The Bland-Altman comparison of the spoke alignment method to the reference values yielded a slightly higher MD = 0.99% with LoA of MD \pm 1.5% than the GMTF correction in the same PDFF range.

3.4 In vivo results

A PDFF map comparison of the upper thigh of a healthy volunteer is displayed in Figure 7. The uncorrected PDFF map showed severe spatial inhomogeneity induced by trajectory errors at the left and right edges of the FOV causing negative PDFF values inside the muscle and quantification errors at the right edge of the right thigh. The PDFF map obtained from the data reconstructed with the proposed GMTF correction showed clear improvement over the uncorrected PDFF map. The quantification errors on the right side of the right thigh and the negative PDFF values in the left thigh could be corrected. The overall appearance of the PDFF map inside the thigh muscle was more homogeneous and in better agreement with the conventional low-resolution Cartesian acquisition. The datadriven spoke alignment approach was also able to correct for trajectory errors of the uncorrected data and achieved a comparable PDFF map quality as the GMTF correction.

A similar PDFF map comparison for an axial liver acquisition of a healthy volunteer is shown in Figure 8. The PDFF map derived from the uncorrected echo images of the radial SoS sequence was heavily affected by trajectory errors, resulting in PDFF inhomogeneities. Quantification errors in subcutaneous adipose tissue were particularly noticeable near the FOV's right edge. Severe negative PDFF underestimations of up to -20% were observed at the lateral back muscles, the spleen and the right liver lobe. The overall quality of the PDFF map estimated from GMTFcorrected images was significantly higher. Errors in the right side's subcutaneous adipose tissue were removed, and the full body outline was restored. The PDFF distribution within the liver was more homogeneous than in the uncorrected case and negative PDFF values in the spleen, right liver lobe and lateral back muscles were corrected as well. The PDFF map agreed in terms of homogeneity with the low-resolution Cartesian breathhold acquisition. The spoke alignment correction was also able to correct most of the trajectory errors and yielded a PDFF map quality comparable to that of the GMTF correction. In the PDFF maps corrected with both the spoke alignment and the GMTFcorrection methods, only minor negative PDFF underestimations were observed at fat-organ tissue and fat-muscle interfaces, as well as in motion-affected areas such as the right liver lobe and the sternum, which are challenging areas for water-fat separation. The liver ROI analysis of the corrected and uncorrected PDFF maps (Figure 9) confirmed that the GMTF correction reduced the standard deviation of the liver PDFF in comparison with the uncorrected data for each subject. After the GMTF correction, the mean liver PDFF agreed with the Cartesian reference mean PDFF within their standard deviation. Comparable correction results were observed for the spoke alignment method. The smallest correction improvements were observed in subjects 3 and 9, which also had the smallest abdominal cavity size.



FIGURE 8

PDFF map comparison from a liver scan of a healthy volunteer. PDFF maps obtained from uncorrected images (top left), GMTF-corrected images (top right) and images corrected with the spoke alignment method (bottom left). Bottom right: PDFF map from a low-resolution Cartesian breath-hold acquisition for comparison. The red arrows highlight PDFF quantification errors for the uncorrected data. GMTF and spoke alignment method remove PDFF errors and achieve a homogeneous PDFF distribution in the liver.



FIGURE 9

Mean liver PDFF over the different liver regions in eleven healthy volunteers. PDFF results obtained from uncorrected data (red), GMTF-corrected data (blue), data corrected with the spoke alignment method (yellow) and data from a low resolution Cartesian breath-hold acquisition (purple). The error bars represent the standard deviation over the different liver regions for a single subject. On the right the mean values over all eleven subjects are shown with the combined standard deviations.

4 Discussion

Eddy currents and gradient system delays are known to cause PDFF quantification errors in both Cartesian [26–28] and radial [16,

17, 19, 20] acquisitions. Radial acquisitions have recently emerged as a viable approach for accomplishing free-breathing PDFF mapping. However, in radial acquisitions a correction of k-space trajectory errors induced by gradient chain imperfections is required to obtain meaningful PDFF maps. The present work shows that a GMTFbased trajectory correction can achieve robust PDFF mapping results when using a radial SoS sequence.

Most previous methods to correct for gradient chain imperfections in radial water-fat imaging relied on the acquisition of calibration lines with opposite polarity [15–21] and recently on estimating the k-space shifts from the radial raw data directly [41]. A GMTF-based trajectory correction method has been applied in cardiac PDFF mapping using a 3D radial sequence lately [61].

The proposed GMTF method provides a flexible correction method for gradient system imperfections which is independent of the applied imaging parameters. A single calibration measurement and a chirp based LTI model [32, 43, 52] are sufficient to obtain the required GMTF to characterize the gradient system chain. As shown previously a GMTF can remain stable over several years [47]. The proposed trajectory correction method could thus be applied in all subsequent image reconstructions from the same scanner without the need of additional calibration scans.

The anisotropic eddy current behavior for the respective gradient axes was measured and corrected based on the extracted GMTF. Only first-order gradient terms responsible for linear eddy currents were considered for the correction step. To include higher-order terms, the phantom method could be extended to 3D by applying phase encoding [45], or a field camera [42, 47] could be used.

The simulation results showed that trajectory errors, predicted by the GMTF measurement, cause magnitude errors in the reconstructed echo images (see Supplementary Figure S1), which translate to PDFF quantification errors. Especially the negative PDFF bias in the numerical phantom could also be observed in the corresponding uncorrected water bottle phantom scan as well as in the water area of the PDFF phantom scan. In both cases the GMTF correction was able to correct the negative PDFF bias in the water area and retrieve a homogeneous PDFF map. In general, the proposed GMTF method improved the PDFF estimation of the PDFF phantom scan, by increasing the linear correlation between the measured and reference PDFF values, as well as reducing the standard deviation of the mean difference between the measured and the reference PDFF over the full range of 0%-100% PDFF. Especially the low fat fraction range of the PDFF phantom scan illustrated the impact of the GMTF-based trajectory correction.

Both scanned *in vivo* anatomies, the upper thigh and the liver, showed the impact of the gradient chain induced trajectory errors on the PDFF maps and proved the need of eddy current correction for precise PDFF mapping in radial Dixon imaging. Uncorrected PDFF maps for both anatomies had significant errors in PDFF values. In both cases the proposed GMTF method was able to correct these errors and to retrieve meaningful PDFF maps. Correcting the liver acquisition data improved the PDFF estimation of the liver, by reducing the standard deviation of the mean liver PDFF across liver regions of each subject and by reducing the bias of the mean liver PDFF.

The GMTF correction achieved comparable results as the datadriven approach based on aligning radial spokes of opposite polarity. Despite comparable results, the two methods operate in different ways. The spoke alignment method can inherently correct only constant shifts along the k-space spoke. While the GMTF method is also able to correct linear phase errors by measuring trajectory shifts in k-space, these shifts are not required to be constant over the whole spoke. On the contrary, the results of the trajectory correction yielded larger shifts at the beginning and the end of each spoke, thus affecting the high spatial frequency components of the reconstructed images. While the presented data-driven approaches and methods relying on the acquisition of calibration scans can only shift k-space data, the GMTF-based correction is able to shift and warp k-space sampling positions.

Furthermore, the proposed GMTF method is faster and less computationally demanding than the spoke alignment method. While the fitting operation of the k-space shift during the spoke alignment has to be performed per imaging slice, per angle and per echo, the GMTF correction requires only a single convolution of the input gradient waveform and its integration to calculate a trajectory for a full TR. The proposed GMTF method added only a few seconds to the reconstruction pipeline. It can be thus easily added to any existing non-Cartesian reconstruction pipeline which is able to use the predicted trajectory in a gridding- or NUFFT-based reconstruction.

The present work has several limitations: First, precise knowledge of the input gradient waveform is necessary for a GMTF-based correction method to predict the real gradient waveform. Depending on the manufacturer and the applied sequence, this knowledge may require access to the pulse sequence programming. Second, the GMTF used in this work was only measured up to 10 kHz. While this range is sufficient to cover the frequencies of typical Dixon readout gradient waveforms, this may not be the case for arbitrary waveforms. In this case, a GMTF with a higher frequency range would be necessary. Third, the effect of B₀ eddy currents was neglected in this work, since the main contributions were expected to stem from linear eddy currents. This correction however could be easily implemented because the B0 transfer function can be extracted from the acquired GMTF data. Fourth, geometric distortions caused by spatial nonlinearities in the gradient fields at the FOV edge are not adapted by the proposed method. These nonlinearities are static and do not account for eddy current-induced dynamic trajectory errors. Fifth, a comparison to a GMTF obtained with NMR field probes could not be investigated unfortunately due to the lack of such a field camera device [32, 34]. Sixth, all SoS acquisitions in this work were performed in the same orientation with the SoS axis being aligned with the z-direction. When using double-oblique orientations, the trajectory can be corrected with the GMTF, but 3D encoding and 3D gridding have to be used instead of 2D gridding. Finally, GMTF correction was only evaluated on a pool of healthy volunteers with normal PDFF values. A further clinical evaluation of patients with increased liver PDFF would be of interest.

5 Conclusion

The present work showed that a simple phantom-based GMTF measurement without any additional scanner hardware can be used to correct k-space trajectories in radial SoS Dixon acquisitions for gradient system imperfections. It was demonstrated that the GMTF-based trajectory correction improves PDFF quantification in radial SoS Dixon acquisitions in phantoms and *in vivo*.

Data availability statement

The datasets presented in this article are not readily available because the *in vivo* data cannot be shared according to the ethics approval. The phantom data can be shared upon request. Basic GMTF measurement results from a Philips system and means to efficiently obtain them are published in (JR, Mazurkewitz P, PB, Nielsen T. Magn Reson Med. 2019 Dec; 82(6):2146-2159). Requests to access the datasets should be directed to christoph.zoellner@tum.de, www.bmrr.de.

Ethics statement

The studies involving humans were approved by the ethics committee of Klinikum rechts der Isar, Technical University of Munich, Munich, Germany. The studies were conducted in accordance with the local legislation and institutional requirements. The participants provided their written informed consent to participate in this study.

Author contributions

CZ, SK, and DK conceived and planned the project. JR and PB developed and provided the fundamentals of the GMTF measurement method. CZ carried out all measurements, performed the numerical simulations, processed and analyzed the data. CB and JS designed the water-fat separation code. KW, JP, SK, DJ, and DK aided in interpreting the results. CZ drafted the manuscript and designed the figures with input from all authors. All authors contributed to the article and approved the submitted version.

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Conflict of interest

The authors KW, JR, PB, and JP are employed by Philips. DK receives research funding from Philips Healthcare outside the scope of the present manuscript.

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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Supplementary material

The Supplementary Material for this article can be found online at: https://www.frontiersin.org/articles/10.3389/fphy.2023.1124980/ full#supplementary-material

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