Whole brain g-ratio mapping using myelin water imaging (MWI) and neurite orientation dispersion and density imaging (NODDI)

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ABSTRACT

MR g-ratio, which measures the ratio of the aggregate volume of axons to that of fibers in a voxel, is a potential biomarker for white matter microstructures. In this study, a new approach for acquiring an in-vivo whole human brain g-ratio map is proposed. To estimate the g-ratio, myelin volume fraction and axonal volume fraction are acquired using multi-echo gradient echo myelin water imaging (GRE-MWI) and neurite orientation dispersion and density imaging (NODDI), respectively. In order to translate myelin water fraction measured in GRE-MWI into myelin volume fraction, a new scaling procedure is proposed and validated. This scaling approach utilizes geometric measures of myelin structure and, therefore, provides robustness over previous methods. The resulting g-ratio map reveals an expected range of g-ratios (0.71 to 0.85 in major fiber bundles) with a small inter-subject coefficient of variance (less than 2%). Additionally, a few fiber bundles (e.g. cortico-spinal tract and optic radiation) show different constituents of myelin volume fraction and axonal volume fraction, indicating potentials to utilize the measures for deciphering fiber tracking.

Keywords:

Myelin, Gradient-echo myelin water imaging (GRE-MWI), g-ratio, MRI, Diffusion, White matter microstructure
INTRODUCTION

White matter in the brain is composed of fibers that have signal conducting axon and signal insulating myelin. The structural dimensions of these microstructures such as axonal diameter and myelin thickness are important physical parameters that determine the conduction velocity of a neuron (Goldman and Albus, 1968; Rushton, 1951; Waxman, 1980) and affect normal functions of the brain (Albert et al., 2007; Dean et al., 2016). A few studies have demonstrated that the ratio of axon diameter (excluding the thickness of myelin sheath) to fiber diameter (including the thickness of myelin sheath) is a key determinant for the conduction velocity of a fiber (Chomiak and Hu, 2009; Goldman and Albus, 1968; Rushton, 1951; Waxman, 1980). This ratio, which is referred to as a g-ratio, may provide important scientific and clinical values (e.g. report brain plasticity from learning and capture de- and remyelination in multiple sclerosis).

However, measuring the g-ratio in-vivo is challenging because of the size of axonal fiber (~µm) compared to MR resolution (~mm). Recently, Stikov et al. have proposed an approach of measuring an area-weighted g-ratio using MRI (Stikov et al., 2011). This approach obtains the axonal and fiber volume fraction in a voxel instead of the axonal and fiber diameters of individual fibers. Then an area- (or volume-) weighted g-ratio is calculated assuming a uniform g-ratio for all fibers in the voxel.

In subsequent MR g-ratio studies, the myelin volume fraction (MVF) and axonal volume fraction (AVF) were measured using myelin-sensitive methods and diffusion-sensitive methods, respectively (Berman et al., 2017; Cercignani et al., 2017; Dean et al., 2016; Duval et al., 2017; Mohammadi et al., 2015; Stikov et al., 2015; Stikov et al., 2011). In particular, magnetization transfer (MT) imaging, such as quantitative magnetization transfer (qMT) (Cercignani et al., 2017; Stikov et al., 2015) and MT saturation (Mohammadi et al., 2015) were applied to estimate MVF, utilizing the sensitivity of the MT contrasts for myelin (Odrobina et al., 2005; Schmierer et al., 2007; Wolff and Balaban, 1989).

This approach of using an MT parameter (e.g. bounded pool fraction in qMT) for MVF requires a scaling factor that translates this parameter to MVF (Cercignani et al., 2017; Mohammadi et al., 2015; Stikov et al., 2015). To date, two approaches have been used (Table 1). The first approach estimates the scaling factor by comparing MR measurements with those of
histology in the same ex-vivo sample (Stikov et al., 2015). The resulting scaling factor is then applied for in-vivo studies. The second approach determines the scaling factor by setting the g-ratio of the splenium of the corpus callosum to be 0.7 and scales the rest of the brain accordingly (Cercignani et al., 2017; Mohammadi et al., 2015). Both of these methods have critical limitations. In the first approach, a range of scaling factors (1.6 to 5.2) has been reported for different studies (Dula et al., 2010; Stikov et al., 2015; Thiessen et al., 2013) and the choice of a particular scaling factor has a significant effect on the resulting g-ratio (Campbell et al., 2017). In the second approach, no histology study exists that reports the g-ratio of the splenium in human and therefore, the choice of 0.7 as the g-ratio of splenium is arbitrary (see Discussion). Moreover, the second approach ignores g-ratio variation in the splenium among subjects. In addition, MT effects originate not only from myelin but also from other macromolecules (West et al., 2016b; Wolff and Balaban, 1989), biasing the scaling results of the MT-based approaches.

**Table 1. Previous studies of the g-ratio in mammalian brains**

<table>
<thead>
<tr>
<th>Article</th>
<th>Target</th>
<th>g-ratio range</th>
<th>Details</th>
</tr>
</thead>
<tbody>
<tr>
<td>-------------------------</td>
<td>---------------------------------------------</td>
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</tr>
<tr>
<td>Blakemore et al., 1974</td>
<td>Mouse brain - superior cerebellar peduncle</td>
<td>0.76 ~ 0.81</td>
<td></td>
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<tr>
<td>Waxman et al., 1976</td>
<td>Rabbit brain - splenium of corpus callosum</td>
<td>0.77 ± 0.05</td>
<td></td>
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<tr>
<td>Graf von Keyserlingk and Schramm, 1984</td>
<td>Human brain - pyramidal tract</td>
<td>Fiber diameter &lt; 5 μm - broadly 0.6</td>
<td>Fiber diameter &gt; 5 μm - over 0.6</td>
</tr>
<tr>
<td>Guy et al., 1989</td>
<td>Guinea pig brain - optic nerve</td>
<td>0.81 ± 0.08</td>
<td></td>
</tr>
<tr>
<td>Chau et al., 2000</td>
<td>Rat brain - optic nerve</td>
<td>0.78 ± 0.003</td>
<td></td>
</tr>
<tr>
<td>Arnett et al., 2001</td>
<td>Mouse brain - corpus callosum</td>
<td>0.758 ± 0.070</td>
<td></td>
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<tr>
<td>Mason et al., 2001</td>
<td>Mouse brain - corpus callosum</td>
<td>0.808 ± 0.011</td>
<td></td>
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<tr>
<td>Benninger et al., 2006</td>
<td>Mouse brain - corpus callosum</td>
<td>0.75 ± 0.005</td>
<td></td>
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<tr>
<td></td>
<td>Mouse brain - optic nerve</td>
<td>0.81 ± 0.014</td>
<td></td>
</tr>
<tr>
<td>Chomiak et al., 2009</td>
<td>Rat brain</td>
<td>0.76 ~ 0.77</td>
<td></td>
</tr>
<tr>
<td>Liewald et al., 2014</td>
<td>Monkey brain - superior longitudinal fasciculus</td>
<td>~ 0.79</td>
<td></td>
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<tr>
<td>Reference</td>
<td>Species</td>
<td>Corpus Callosum</td>
<td>MVF/AVF</td>
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<tr>
<td>Stikov et al., 2015a</td>
<td>Macaque brain</td>
<td></td>
<td>0.6 ~ 0.74</td>
</tr>
<tr>
<td>West et al., 2016a</td>
<td>Mouse brain</td>
<td></td>
<td>0.815 ± 0.003 (mean)</td>
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<tr>
<td></td>
<td>Mouse brain</td>
<td></td>
<td>0.848 ± 0.003 (area-weighted mean)</td>
</tr>
<tr>
<td>Stikov et al., 2015a</td>
<td>Human brain</td>
<td>~ 0.7</td>
<td>MT$^1$ and NODDI$^2$</td>
</tr>
<tr>
<td>Dean et al., 2016</td>
<td>Human brain</td>
<td>0.71 ~ 0.9</td>
<td>mcDESPOT$^3$ and NODDI</td>
</tr>
<tr>
<td>Mohammadi et al., 2016</td>
<td>Human brain</td>
<td>0.55 ~ 0.75</td>
<td>MT and TFD$^4$</td>
</tr>
<tr>
<td>Cercignani et al., 2017</td>
<td>Human brain</td>
<td>0.65 ~ 0.8</td>
<td>MT and NODDI</td>
</tr>
<tr>
<td>Berman et al., 2017</td>
<td>Human brain - corpus callosum</td>
<td>~ 0.69</td>
<td>MTV$^5$ and FA$^6$</td>
</tr>
</tbody>
</table>

\footnotesize{$^1$MT: Magnetic transfer imaging \quad $^2$NODDI: Neurite orientation dispersion and density imaging \quad $^3$mcDESPOT: Multi-component driven equilibrium single pulse observation of T$_1$/T$_2$ \quad $^4$TFD: Tensor fiber density \quad $^5$MTV: Lipid and macromolecular tissue volume \quad $^6$FA: Fractional anisotropy}

An alternative approach for estimating MVF utilizes myelin water imaging (MWI). MWI acquires a T$_2$ (or T$_2^*$) decay curve to estimate the signal fraction of the short T$_2$ (or T$_2^*$) signal from myelin water and the long T$_2$ (or T$_2^*$) signal from axonal and extracellular water (Du et al., 2007; Mackay et al., 1994). Then, the ratio of myelin water signal to total water signal (i.e. the sum of myelin, axonal, and extracellular water signal), which is referred to as myelin water fraction (MWF), is estimated. The method may provide a more direct measurement of myelin signal that is not complicated by other macromolecules (Laule et al., 2008; Laule et al., 2006). Recently, Dean et al., utilized MWF from mcDESPOT (Deoni et al., 2008) for g-ratio mapping without clearly detailing the conversion process from MWF to MVF (Dean et al., 2016). However, MWI signal does not include non-water components (e.g. lipid) of myelin and, therefore, a scaling process is still necessary to translate MWF into MVF.

In this study, we present an approach that scales MWF to MVF by utilizing geometry and mass-density measurements of myelin. Using this approach, an in-vivo human brain g-ratio map is generated by combining MVF from MWI and AVF from neurite orientation dispersion and...
density imaging (NODDI) (Zhang et al., 2012). Additionally, the characteristics of MVF, AVF, and g-ratio in a few major fiber bundles are analyzed. Lastly, the inter-subject coefficient of variance (COV) is estimated to assess the reliability of the MVF, AVF, and g-ratio measurements.

THEORY

Scaling myelin water fraction to myelin volume fraction

In MWI, the signal amplitude of water compartment is determined by MR visible protons at the echo time of milliseconds. Therefore, it cannot report on structures such as myelin lipid bilayer and neurofibril. The signal amplitude of a compartment can be written as the product of the MR visible volume ratio and the volume of the compartment:

\[ A \propto \kappa \cdot V \]  (Eq. 1)

where \( A \) is the signal amplitude, \( \kappa \) is the MR visible volume ratio defined as the ratio of the MR visible volume to the total volume of the compartment, and \( V \) is the total volume of the compartment.

Since MWI estimates the signal amplitude of the three water compartments (i.e. myelin water, axonal water and extracellular water) (Lancaster et al., 2003), MVF can be obtained using the following equation if the MR visible volume ratio of each compartment is known.

\[ MVF = \frac{V_{my}}{V_{my} + V_{ax} + V_{ex}} = \frac{A_{my}/\kappa_{my}}{(A_{my}/\kappa_{my} + A_{ax}/\kappa_{ax} + A_{ex}/\kappa_{ex})} \] (Eq. 2)

where \( V_{my}, V_{ax}, \) and \( V_{ex} \) are the volume of the myelin, axon, and extracellular compartments respectively, \( A_{my}, A_{ax}, \) and \( A_{ex} \) are the signal amplitude of the myelin water, axonal water, and extracellular water, respectively, and \( \kappa_{my}, \kappa_{ax}, \) and \( \kappa_{ex} \) are the MR visible volume ratios of the myelin, axon, and extracellular compartments, respectively.

In order to estimate \( \kappa_{my} \), two different methods are proposed as described as follows. The first method, which is referred to as a geometric approach, calculates \( \kappa_{my} \) using the geometric property of myelin. The second method calculates the value using the mass and density of myelin.
and is referred to as a mass-density approach.

The geometric approach utilizes the lamellar structure of myelin sheath, which is composed of four types of layers (i.e. lipid bilayer, major dense line, lipid bilayer, and intraperiod line) (Peters, 1960). Assuming that only protons in the major dense line and intraperiod line contribute to the myelin water signal and that the contribution from other molecules (e.g. proteins) can be ignored, we can estimate the MR visible volume ratio of the myelin compartment using the thicknesses of the layers, which are reported in X-ray diffraction studies of mammalian CNS nerves (Finean, 1962; Morell and Quarles, 1999). Considering the length (l) of myelinated axon with a particular number of lamellae (n) (Fig. 1), myelin lipid volume (MLV) and myelin water volume (MWV) are calculated as the sum of the alternating volumes of the concentric cylinders (i.e. yellow areas \( \times l \) for MLV and blue areas \( \times l \) for MWV in Fig. 1a):

\[
MLV = \sum_{i=0}^{2n} \pi \left[ \left( w_{lip} + w_{water} \right) \times i + w_{lip} + r_{ax} \right]^2 - \left( w_{lip} + w_{water} \right) \times i + r_{ax}^2 \right] \times l
\]  
\[
(MWV) = \sum_{i=1}^{2n} \pi \left[ \left( w_{lip} + w_{water} \right) \times i + r_{ax} \right]^2 - \left( w_{lip} + w_{water} \right) \times i - w_{water} + r_{ax}^2 \right] \times l
\]

(Eq. 3)

(Eq. 4)

where \( w_{lip} \) and \( w_{water} \) represent the thickness of the lipid layer and water layer, \( r_{ax} \) is the inner radius of the axon, and \( i = 0 \) indicates a myelinated axon with a single lipid bilayer (Fig. 1a). Then, the MR visible volume ratio of the myelin compartment (\( \kappa_{my} \)) can be written as

\[
\kappa_{my} = \frac{MWV}{(MWV + MLV)} = \frac{w_{water}}{\{(1 + 1/2n)w_{lip} + w_{water}\}}.
\]

(Eq. 5)

Thus, once the number of the lamella is known, the MR visible volume ratio of the myelin compartment can be obtained. Note that \( \kappa_{my} \) is not dependent on \( r_{ax} \) because it cancels out when Eq. 5 is expanded. Since this approach is based on the geometric parameters of myelin, it is named as the geometric approach.

Another approach for calculating \( \kappa_{my} \) uses mass-density measurements (Dula et al.,
2010; Laule et al., 2004; West et al., 2016b) and is referred to as the mass-density approach. Since volume is defined as mass divided by (buoyant) density, MWV and MLV can be expressed as follows:

\[
\text{MWV} = \frac{m_{\text{myelin water}}}{\rho_{\text{myelin water}}} \quad (\text{Eq. 6})
\]

\[
\text{MLV} = \frac{m_{\text{myelin lip}}}{\rho_{\text{myelin lip}}} \quad (\text{Eq. 7})
\]

where \(m_{\text{myelin water}}\) and \(m_{\text{myelin lip}}\) represent the mass (fraction) of myelin water layer and myelin lipid layer, respectively and \(\rho_{\text{myelin water}}\) and \(\rho_{\text{myelin lip}}\) are the density of myelin water layer and myelin lipid layer respectively. Then, \(\kappa_{\text{my}}\) can be estimated from literature values of each parameter (Laule et al., 2004).

The same approach can be applied to estimate the MR visible volume ratios of the axon and extracellular compartments. Assuming that the two compartments have the same ratio, we can combine the two compartments and divide the volume into water volume (axonal & extracellular water volume; AWV) and non-water volume (axonal & extracellular non-water volume; ANWV). Then, each volume can be written as

\[
\text{AWV} = \frac{m_{\text{axon & ext water}}}{\rho_{\text{axon & ext water}}} \quad (\text{Eq. 8})
\]

\[
\text{ANWV} = \frac{m_{\text{axon & ext non-water}}}{\rho_{\text{axon & ext non-water}}} \quad (\text{Eq. 9})
\]

where \(m_{\text{axon & ext water}}\) and \(m_{\text{axon & ext non-water}}\) represent the mass (fraction) of axonal-extracellular water and axonal-extracellular non-water, respectively, and \(\rho_{\text{axon & ext water}}\) and \(\rho_{\text{axon & ext non-water}}\) represent the density of axonal-extracellular water and axonal-extracellular non-water, respectively. Then, similar to Eq. 5, \(\kappa_{\text{ax}}\) and \(\kappa_{\text{ex}}\) can be obtained using

\[
\kappa_{\text{ax}} = \kappa_{\text{ex}} = \frac{\text{AWV}}{\text{AWV} + \text{ANWV}}. \quad (\text{Eq. 10})
\]

These MR visible volume ratios of the three compartments are combined with the signal amplitude measurements of MWI to estimate MVF using Eq. 2.

**MATERIALS AND METHODS**

*Parameters of the scaling factors*
The parameters of the scaling factors used literature values of mammalian CNS. In the geometric approach, the thicknesses of lipid bilayer (= $w_{lip}$) and water layer (= $w_{water}$) were set to be 51 Å and 29 Å, respectively (Finean, 1962; Morell and Quarles, 1999). In the mass-density approach, the mass fractions and densities of myelin water and lipid layers were as follows: $m_{myelin\ water} = 0.082$ g per 1 g of white matter, $m_{myelin\ lip} = 0.14$ g per 1 g of white matter, $\rho_{myelin\ water} = 1$ g/ml, and $\rho_{myelin\ lip} = 1.08$ g/ml (Laule et al., 2004). For the calculation of $\kappa_{ax}$ and $\kappa_{ex}$, the following parameters were used for mass fractions and densities of water and non-water components in axonal and extracellular space: $m_{axon\ &\ ext\ water} = 0.638$ g per 1 g of white matter, $m_{axon\ &\ ext\ non-water} = 0.14$ g per 1 g of white matter, $\rho_{axon\ &\ ext\ water} = 1$ g/ml, and $\rho_{axon\ &\ ext\ non-water} = 1.33$ g/ml (Laule et al., 2004).

**Validation of the scaling factors using existing histological data**

To validate the scaling approaches, an existing histo-imaging dataset was utilized (West et al., 2016b). The dataset contained histologically measured MVF and MR measured MWF. Four different ROIs were chosen (genu, mid-body, splenium of corpus callosum, and anterior commissure) from four different types of mouse groups (two hypomyelinated mice (n = 3 each), one hypermyelinated (n = 3), and control mouse groups (n = 6)), yielding a wide range of MVF. To demonstrate the validity of proposed scaling methods, the scaling factors from the geometric approach and the mass-density approach were applied to translate the MRI MWF in the dataset into MRI MVF. Then, a linear regression was applied to MRI MVF and histological MVF, and $R^2$ was calculated to demonstrate the goodness of fit.

**MRI data acquisition**

Five healthy subjects were scanned at a 3 Tesla MRI system (Tim Trio, SIEMENS, Erlangen, Germany) using a 32-channel phased-array head coil. The study was performed under the approval of the internal review board. Multi-echo GRE images and multi-shell diffusion images were acquired for MWI and NODDI, respectively.
The scan parameters for the multi-echo GRE were as follows: multi-slice acquisition, TR = 1400 ms, TE = 2.45 ms to 39.5 ms with 2.2 ms echo spacing, number of echoes = 18, flip angle = 83°, bandwidth = 500 Hz/pixel, FOV = 256 × 208 mm², voxel size = 2 × 2 mm², slice thickness = 2 mm, number of slices = 48, concatenation = 2, bipolar readout gradient, and scan time = 5 min. Both respiratory and cardiac noise were reduced using navigation echoes (17th and 18th echoes) and flow saturation RF pulses (Nam et al., 2015a). To compensate for the eddy current effects from the bipolar readout gradient, the scan was repeated using the opposite readout direction (Lu et al., 2008).

For NODDI, three-shell diffusion weighted imaging data (b = 300 s/mm² with 8 directions; b = 700 s/mm² with 32 directions; b = 2000 s/mm² with 64 directions; b = 0 s/mm² with 13 averages) was acquired using a single shot spin-echo echo-planar-imaging (SE-EPI) sequence with TR = 4000 ms, TE = 95 ms, FOV = 192 × 192 mm², voxel size = 2 × 2 mm², slice thickness = 2 mm, multi-band factor = 2, partial Fourier = 6/8, GRAPPA factor = 2, and scan time = 9.75 min. To compensate for EPI geometric distortion, additional b = 0 s/mm² image was acquired with the opposite phase encoding direction. The geometric distortion was corrected using TOPUP and EDDY (FSL, FMRIB, Oxford, UK) (Andersson et al., 2003).

**MRI data processing**

The multi-echo GRE data were processed as follows: First, respiration induced phase errors were reduced using the navigation echoes (Nam et al., 2015a). Gibbs’ ringing was attenuated using a Tukey window (filter size = 1/3) applied in k-space. Then, multi-channel images were combined by the root sum-of-squared for magnitude and by the complex sum for phase after phase offset correction (Hammond et al., 2008). Eddy current induced artifacts from the bipolar readout were compensated by combining the two GRE images of the opposite readout directions (Buonocore and Gao, 1997).

After generating the combined GRE images, \( A_my \), \( A_ax \), and \( A_ex \) were estimated by fitting a complex three pool model using a nonlinear-least square curve fitting algorithm (lsqnonlin function in MATLAB, TolX = 1e-5, TolFun = 1e-5) with the fitting parameters described in Nam
et al. (Nam et al., 2015b). Finally, the resulting signal amplitudes ($A_{my}$, $A_{ax}$, and $A_{ex}$) were used to calculate MVF using Eq. 2.

For AVF, the diffusion images were processed using AMICO (Daducci et al., 2015). In NODDI, the volume of each voxel was divided into isotropic ($V_{iso}$) and anisotropic volume fraction ($1 - V_{iso}$). Subsequently, the anisotropic volume fraction was divided into anisotropically restricted volume fraction $((1 - V_{iso})V_{ic})$, originating from axonal space, and anisotropically hindered volume fraction $((1 - V_{iso})(1 - V_{ic}))$, originating from extracellular space (Zhang et al., 2012). When calculating these volume fractions, a method that corrects for T2 difference between isotropic compartment (i.e. CSF) and anisotropic compartment has been applied using T2 values of 2000 ms and 90 ms, respectively (Bouyagoub et al., 2016).

Since the myelin signal was not included in NODDI acquisition due to the long TE of diffusion weighted imaging, the estimated volume fraction was scaled by $(1 - MVF)$ (Stikov et al., 2015), which was obtained in MWI. Then, the anisotropically restricted volume fraction was transformed into AVF using the following equation:

$$AVF = (1 - MVF)(1 - V_{iso})V_{ic}. \quad (\text{Eq. 11})$$

This procedure of combining the anisotropically restricted volume fraction and myelin volume fraction requires an accurate registration between the diffusion and GRE images. The registration was performed between the 5th echo GRE image and the $b = 0$ s/mm$^2$ diffusion image using BSpline in Advanced Normalization Tools (ANTS) (Avants et al., 2009) after brain extraction (Jenkinson et al., 2005). The resulting transformation matrix was applied to the MVF map.

Finally, an aggregate g-ratio was calculated from MVF and AVF values using the following equation (Stikov et al., 2015):

$$g = \sqrt{1/(1 + MVF/AVF)}. \quad (\text{Eq. 12})$$

This aggregate g-ratio was calculated in each voxel, generating an in-vivo g-ratio map of the brain.

Additionally, the diffusion weighted images (all b values) were processed to identify major fiber bundles. DTI reconstruction was performed using FSL’s DTIFIT (Jenkinson et al., 2012).
**MRI data analysis**

A white matter mask was created for brain voxels with MVF ranging from 0.01 to 0.50 and AVF higher than 0.2. The resulting mask was smoothed by a Gaussian filter (standard deviation = 2 voxels) and then thresholded at 0.6, generating the final white matter mask. Six regions of interest (ROIs), cingulum bundles (CG), cortico-spinal tract (CST), genu (GN), optic radiation (OR), superior longitudinal fasciculus (SLF), and splenium (SPL) were manually chosen based on the color-coded DTI FA map in each subject (Supplementary Fig. S1). The mean MVF, AVF and g-ratio values of each ROI were computed in each subject.

To demonstrate the reliability of the g-ratio map, the inter-subject COV was calculated as follows:

\[
\text{Inter-subject COV of an ROI} = \frac{(\text{standard deviation of the ROI across subjects})}{(\text{average of the ROI across subjects})}.
\]

(Eq. 13)

Additionally, a scatter plot of MVF and AVF for each subject was generated to demonstrate different constitution of the volume fractions in the fiber tracts.

**RESULTS**

**Scaling myelin water fraction to myelin volume fraction**

Fig. 1b shows the MR visible volume ratio of the myelin compartment (\(\kappa_{my}\)) as a function of the number of lamellae using the geometric approach. The plot reveals that \(\kappa_{my}\) ranges from 0.34 to 0.36 for a myelinated fiber that has more than five lamellae. When the average number of lamellae (=15) in mammalian CNS nerves (Mierzwa et al., 2010) is used, \(\kappa_{my}\) is approximately 0.36. In the mass-density approach, \(\kappa_{my}\) is calculated to be 0.39 (Eqs. 6 and 7), which is close to \(\kappa_{my}\) from the geometric approach. For \(\kappa_{ax}\) and \(\kappa_{ex}\), the result of the mass-density approach (Eqs. 8 to 10) was 0.86.
Figure 1. (a) Schematic diagram of a myelinated axon. One lamella consists of four layers alternating between water and lipid (red arrow). (b) MR visible volume ratio of the myelin compartment ($\kappa_{\text{my}}$) as a function of the number of lamellae (solid black line) in the geometric approach.
Figure 2. Validation of the scaling approaches using existing histo-imaging data (West et al., 2016b). Histologically measured MVF and MRI estimated MVF scaled from MWI using the two proposed scaling methods are plotted (left: scaling using the geometric approach, right: scaling using the mass-density approach). Solid red lines, which represent linear regression lines, are close to the line of unity (dashed black line), suggesting good correspondences between the two measurements.

When these MR visible volume ratios are applied to the existing histo-imaging dataset (West et al., 2016b), the results show good correspondences between histological MVF and MRI MVF as plotted in Fig. 2. Each blue circle represents an ROI measurement with an x value as histological MVF and a y value as MRI MVF translated from MWF. The solid red line is the result of a linear regression whereas the dashed black line is the line of unity. The linear regression lines are close to the line of unity with high $R^2$ (0.69 for the geometric approach in Fig. 2a; 0.68 for the mass-density approach in Fig. 2b), confirming that the proposed scaling approaches are valid. Hence, the MR visible volume ratios of the three compartments, $\kappa_{my} = 0.36$ and $\kappa_{ax} = \kappa_{ex} = 0.86$, are used hereafter. When these volume ratios are applied in Eq. 2, MWF is translated into
MVF as shown in Fig. 3.

**Figure 3.** Plots between myelin water fraction and myelin volume fraction using the geometric approach.

*In-vivo human brain g-ratio mapping*
Figure 4. Color-coded DTI FA (a), myelin volume fraction (MVF) (b), axonal volume fraction (AVF) (c), and g-ratio (d) maps in white matter are displayed for three different slices. The triangle in the lower slice indicates optic radiation (OR). The solid line arrow and dashed line arrow in the middle slice indicate genu (GN) and splenium (SPL), respectively. In the upper slice, the solid line arrow and dashed line arrow indicate cortico-spinal tract (CST) and superior longitudinal fasciculus (SLF), respectively.

The color-coded DTI FA, MVF, AVF, and g-ratio maps of a representative subject are shown in Fig. 4. Three slices that contains the major fiber bundles are displayed. In the lower slice,
the triangle in Fig. 4 indicates OR, which shows high MVF (0.28 ± 0.01; all values hereafter are group-averaged results) and low AVF (0.29 ± 0.01), resulting in a low g-ratio (0.71 ± 0.01). In the middle slice, the solid line arrow shows GN (MVF = 0.26 ± 0.02, AVF = 0.43 ± 0.02, and g-ratio = 0.79 ± 0.01) and the dashed line arrow indicates SPL (MVF = 0.30 ± 0.01, AVF = 0.44 ± 0.02, and g-ratio 0.77 ± 0.01). In the upper slice, the arrow with dashed line indicates SLF (MVF = 0.29 ± 0.01, AVF = 0.38 ± 0.02, and g-ratio = 0.76 ± 0.01) and the arrow with solid line indicates CST (MVF = 0.16 ± 0.01, AVF = 0.43 ± 0.02, and g-ratio = 0.85 ± 0.003). The g-ratio values are markedly higher in CST than in SLF (CST = 0.85 ± 0.003 vs. SLF = 0.75 ± 0.01), which is explained by the lower MVFs in CST than in SLF (CST = 0.16 ± 0.01 vs. SLF = 0.29 ± 0.01) while the AVFs are similar (CST = 0.43 ± 0.02 vs. SLF = 0.38 ± 0.02).

Myelin volume fraction, axonal volume fraction and g-ratio

Fig. 5 shows the average values of MVF, AVF, and g-ratio in each ROI for each subject. Each bar represents the average value of an ROI from each subject with the standard deviation. The subject-averaged MVF was 0.26 ± 0.05 (from 0.16 to 0.30), that of AVF was 0.39 ± 0.06 (from 0.29 to 0.44), and that of g-ratio was 0.78 ± 0.05 (from 0.71 to 0.85). Overall, the inter-subject variation of an ROI was relatively small in all parameters. The inter-subject COV of the MVF, AVF, and g-ratio are reported in Table 2. Most of the COVs except MVF in GN are less than 6%, suggesting the measurements are consistent across the subjects (Table 2).
**Figure 5.** Mean myelin volume fraction (a), axonal volume fraction (b), and aggregate g-ratio (c) with the standard deviation of the ROIs for each subject are plotted (CG: cingulum, CST: corticospinal tract, GN: genu, OR: optic-radiation, SLF: superior longitudinal fasciculus, and SPL: splenium).
Table 2. The inter-subject coefficient of variance (COV) of the MVF, AVF, and g-ratio in each ROI across the subjects (CG: cingulum, CST: cortico-spinal tract, GN: genu, OR: optic-radiation, SLF: superior longitudinal fasciculus, and SPL: splenium)

<table>
<thead>
<tr>
<th></th>
<th>CG</th>
<th>CST</th>
<th>GN</th>
<th>OR</th>
<th>SLF</th>
<th>SPL</th>
</tr>
</thead>
<tbody>
<tr>
<td>MVF</td>
<td>5.1</td>
<td>4.6</td>
<td>9.0</td>
<td>3.2</td>
<td>5.0</td>
<td>4.5</td>
</tr>
<tr>
<td>AVF</td>
<td>4.5</td>
<td>5.0</td>
<td>4.4</td>
<td>4.5</td>
<td>5.7</td>
<td>4.9</td>
</tr>
<tr>
<td>g-ratio</td>
<td>1.5</td>
<td>0.4</td>
<td>1.9</td>
<td>0.8</td>
<td>1.8</td>
<td>1.5</td>
</tr>
</tbody>
</table>

The scatter plot of ROI-averaged MVF and AVF for each subject is shown in Fig. 6. The plot reveals isolated clusters for CST (low MVF and high AVF) and OR (high MVF and low AVF) while the others ROIs are relatively close to each other (high MVF and high AVF). Still, it may be possible to further distinguish SLF (high MVF and mid AVF), CG (mid MVF and mid AVF), and SPL (high MVF and high AVF), if GN, which may be affected by the field inhomogeneity artifacts, is ignored.
**DISCUSSION**

In this study, we developed an approach to generate an *in-vivo* whole human brain g-ratio map by combining MWI and NODDI. The scaling methods that translated MWF into MVF demonstrated a good correspondence with the histological results. The resulting g-ratio map showed consistency across the subjects with small inter-subject COVs. Additionally, we demonstrated that AVF and MVF of a few major fiber bundles have a distinct constituent, which may be helpful in resolving crossing fibers in fiber tracking. The proposed method takes less than 20 min. of scan time and can be applied in a clinical setting.

In our results, the g-ratio varied from $0.71 \pm 0.01$ (OR) to $0.85 \pm 0.01$ (CST). The high g-ratio of CST is because of the low MVF ($0.16 \pm 0.01$) and the high AVF ($0.43 \pm 0.02$), which is partly supported by the histological finding of large axons in CST (Graf von Keyserlingk and Schramm, 1984; Yagishita et al., 1994). The low g-ratio of OR is due to a high MVF ($0.28 \pm 0.01$) and a low AVF ($0.29 \pm 0.01$). This is explained in part by the thick myelin sheath in OR (Kitajima et al., 1996). Unfortunately, there is no histological reference of human g-ratio except Graf von Keyserlingk and Schramm’s study, which only reports rough g-ratios of pyramidal tract (Table 1). Further investigation of fiber tract-wise g-ratio in the human brain is necessary.

One of the interesting results of our study is that the two quantities, MVF and AVF, used for the g-ratio calculation provide important information of the white matter microstructure that can be used to distinguish different fiber bundles as demonstrated in Fig. 6. This information may be helpful in disentangling crossing fibers using an approach similar to the one proposed by De Santis et al., (De Santis et al., 2016). Hence, reporting and utilizing all three parameters will be valuable for future research.

*Robustness of the scaling factors*
In this study, we demonstrated that the scaling parameters of both geometric and mass-density approaches were similar (0.36 and 0.39, respectively) despite differences in the methods. Additionally, we showed that these parameters were consistent with the results of the histomaging data (West et al., 2016b), suggesting robustness of the scaling factors. Still, our method assumed consistent scaling parameters across fibers, regions, and species. In particular, the geometric approach considered the thicknesses of myelin water layer ($w_{\text{water}}$) and lipid layer ($w_{\text{lip}}$) or the ratio of the two thickness ($w_{\text{lip}}/w_{\text{water}}$) to be the same. To our knowledge, two measurements of the thicknesses from two different species (rat optic nerve (Finean, 1962) and rabbit optic nerve (Kirschner and Caspar, 1972); both used fresh tissues) exist in literature and they reported similar results (the thickness ratio was 1.73 in rat and 1.55 in rabbit). When a computer simulation was performed to estimate the effects of this variation in MVF and g-ratio, the results showed very small differences (see Supplementary Information). Additionally, a few studies reported the thickness of lamellar (not the thickness of each layer) that were consistent across fiber bundles and species (11 nm in rat spinal cord (Chen et al., 2017; Rosenbluth et al., 1996), 10.7 nm in rat optic nerve and white matter (Karlsson, 1966), 11.2 nm in dog spinal cord (Raine, 1984)). These results further support that the robustness of the scaling parameter. Recently, a study demonstrated that thickness of lamellar does not change during a demyelinating disease as long as the lamellar structure is intact (Chen et al., 2017), suggesting the potential applicability of the scaling parameter in pathological conditions. However, one has to be cautious in applying the method for pathological conditions (see “accuracy of g-ratio mapping” in Discussion).

The other parameter for the scaling was $\kappa_{\text{ax}}$ ($= \kappa_{\text{ex}}$), which was calculated as 0.86 in the mass-density approach. In other studies, the parameter was suggested to be 0.8 (Finean, 1960), 0.82 (Marjo and van der Knaap, 2005), and 0.9 (LoPachin et al., 1991). When a computer simulation was performed, MVF and g-ratio showed little change for the range of variations (0.8 to 0.9) in $\kappa_{\text{ax}}$, supporting the robustness of our scaling approach (see Supplementary Information). Note that all the parameters reported in this study are from the measurements of mammalian CNS.

In addition to the histological validation results in our study, the application of the same scaling factors can be demonstrated using another study that measured both MWF and MVF (Dula et al., 2010). In the study, MVF of rat spinal cords using light microscopy was reported to range from 51% to 61% (these values are larger than that reported in other studies (Benninger et al., 2006;
Chomiak and Hu, 2009), and MWF using MRI was measured to range from 29% to 36% (except dorsal CST, which has underestimated MWF due to water exchange). When MWF was translated into MVF using our scaling approach, MVF was estimated to range from 49% to 57%. This result demonstrates the consistency of the proposed scaling approach for different species (rat vs. mouse), different regions (spinal cord vs. brain), and different field strengths (7T vs. 15.2T) in the two studies (Dula et al., 2010; West et al., 2016b). Compared to these studies, which scanned rodents using spin echo MWI (SE-MWI), we applied the method for human using GRE-MWI. Because of the aforementioned consistency in the scaling parameters across regions and species, we believe they still hold for the human brain. The two MWI methods (SE-MWI and GRE-MWI) have recently demonstrated to provide similar MWF measurements (Alonso-Ortiz et al., 2017).

**Geometric approach vs. mass-density approach**

When comparing the $\kappa_{my}$ value from mass-density approach with that from the geometric approach, they showed similar values (0.39 in the mass-density approach vs. 0.36 in geometric approach). The difference may originate from the water molecules in the lipid bilayer in the mass-density approach. It was reported that water molecules exist in the myelin lipid bilayer and account for 12% of total water molecules in myelin sheath (Vandenheuvel, 1965). These water molecules may not be visible in MRI due to extremely short T2 and, therefore, need to be excluded in estimating $\kappa_{my}$. In the mass-density approach, however, all water molecules including those in lipid layers are incorrectly classified as the MR visible volume because the water mass fraction of myelin is calculated by comparing the weight of wet and dry white matter (Laule et al., 2004; Marjo and van der Knaap, 2005). Hence, $\kappa_{my}$ is overestimated in the mass-density approach. On the other hand, the geometric approach does not take into account the water molecules in the lipid layer.

In the geometric approach, other molecules such as proteins in the myelin water layer are ignored, assuming the concentration is small. This assumption is reasonable since the concentration has been reported to be less than 5% (Gennis, 1989; West et al., 2016b).

Recently, a study estimated $\kappa_{my}$ using the mass-density approach (West et al., 2016b).
The resulting $\kappa_{my}$ value ($=0.475$) was higher than ours ($=0.39$ in the mass-density approach). The difference may originate from the difference in the water content of myelin (Laule et al., 2004; West et al., 2016b), which is defined as the mass of water molecules in myelin per 1g of myelin. Note that the parameters used in the mass-density approach have been shown to vary with tissue fixation, leading to an error in the measurements (Chen et al., 2017; Finean, 1960). On the other hand, the thicknesses of the lamellar structure used in the geometric approach were measured in fresh tissues (Finean, 1962; Kirschner and Caspar, 1972) and were not affected by fixation. Thus, $\kappa_{my}$ estimated in the geometric approach may provide a more accurate result.

**Estimation of MVF using MRI**

In MRI, a few methods including MWI and MT have been suggested to be sensitive to myelin concentration. Compared to MT-based methods, our methods of scaling MWI to MVF utilize geometric parameters (e.g. thickness of myelin water layer and lipid layer) and have demonstrated consistent scaling parameters for both geometric and mass-density approaches (0.36 and 0.39 respectively). On the other hand, MT-based methods have shown widely varying scaling factors (1.6 to 5.2) (Dula et al., 2010; Stikov et al., 2015; Thiessen et al., 2013). In terms of data acquisition, proposed GRE-based MWI approach is efficient, providing a whole brain myelin map in less than 10 mins. Further reduction of the scan time by optimizing the scan parameters and by using advanced methods (e.g. multi-band imaging) can be achieved, making the method potentially more appealing than MT-based methods.

In our study, the mean g-ratio value of all ROI is 0.78 ± 0.05 (from 0.71 to 0.85) and is close to the theoretically optimum g-ratio of 0.77 (Chomiak and Hu, 2009). This range of g-ratio is in good agreement with most of histological g-ratio studies listed in Table 1 (Arnett et al., 2001; Benninger et al., 2006; Blakemore, 1974; Chau et al., 2000; Chomiak and Hu, 2009; Graf von Keyserlingk and Schramm, 1984; Guy et al., 1989; Liewald et al., 2014; Mason et al., 2001; Waxman and Swadlow, 1976; West et al., 2016a). The result, however, is slightly higher than those from the previous studies using MT approaches for MVF estimation (Table 1) (Cercignani et al., 2017; Mohammadi et al., 2015; Stikov et al., 2015). Several factors may account for the difference. For example, scaling of an MT parameter with a zero intercept has been used in a few studies,
which may suffer from the nonzero intercept demonstrated in more recent studies (Campbell et al., 2017; West et al., 2016b). Additionally, the approach of scaling the g-ratio of the splenium to 0.7, which has been used in a few previous studies, may have resulted in a bias in the g-ratio map. Note that the assumption of the splenium to have a g-ratio of 0.7 is not supported histologically because the reference (Graf von Keyserlingk and Schramm, 1984) does not clearly state a tract-specific g-ratio.

Recently, new approaches such as macromolecular tissue volume (MTV) imaging (Mezer et al., 2013), ultrashort echo time (UTE) myelin proton imaging (Du et al., 2014; Wilhelm et al., 2012), and direct visualization of short transverse relaxation time component (ViSTa) imaging (Oh et al., 2013) have been proposed to acquire myelin sensitive contrasts. These methods also require certain assumptions or scaling factors that translate myelin concentration measurements to myelin volume fraction. For example, the studies that applied MTV for g-ratio mapping suggested to use unscaled MTV values, assuming the volume of non-myelin macromolecules (e.g. glial cells) is equal to myelin water volume (Berman et al., 2017; Duval et al., 2017).

**Accuracy of g-ratio mapping**

Despite our efforts of accurately estimating each parameter, there are a few potential complications that can bias our results. First, MVF can be underestimated by inter-compartmental water exchange between myelin water and axonal and extracellular water (Dula et al., 2010; Kalantari et al., 2011; Sled and Pike, 2001). Secondly, GRE-MWI is sensitive to field inhomogeneity, generating image artifacts at the frontal lobe. The artifacts may be the source for the large inter-subject COV of MVF in GN (9.0%). Recently, methods have been proposed to mitigate the artifacts (Alonso-Ortiz et al., 2016; Lee et al., 2017). In AVF estimated using NODDI, a few parameters (e.g. parallel diffusivity of the axon and extracellular compartment) are fixed in the model, which may lead to estimation errors in AVF (Jelescu et al., 2016; Novikov et al., 2015). Another limitation of our method is that it may not properly estimate g-ratio in gray matter because of the low concentration and low CNR of MVF (Meyers et al., 2009) and the inaccurate estimation of AVF (Guerrero et al., 2016; Lampinen et al., 2017). Lastly, MWF estimation may not distinguish between intact myelin and myelin debris with intact lamellar structure (Chen et al., 2017). This
may result in an unwanted overestimation of MVF in the early stage of demyelination. Furthermore, one has to be extra careful in estimating and interpreting the parametric results in pathological conditions since the models may not correctly explain the conditions (e.g. myelin edema) (Harkins et al., 2013).

CONCLUSION

In this study, we demonstrated an approach of acquiring the whole brain \textit{in-vivo} g-ratio map from the human brain. The approach combined GRE-MWI and NODDI to estimate MVF and AVF, respectively, and provided a reliable estimation of the g-ratio map. In generating the g-ratio map, we developed a new scaling method to translate MWF of GRE-MWI into MVF using the geometric measures of myelin structure. The proposed g-ratio mapping approach may assist us to explore microstructural differences and changes between fiber bundles and different groups.

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