

# $B_1^+$ -correction of magnetization transfer saturation maps optimized for 7T postmortem MRI of the brain

Ilona Lipp<sup>1</sup>   | Evgeniya Kirilina<sup>1</sup>  | Luke J. Edwards<sup>1</sup> | Kerrin J. Pine<sup>1</sup>  |  
Carsten Jäger<sup>1,2</sup> | Tobias Gräßle<sup>3</sup> | EBC Consortium<sup>1</sup> | Nikolaus Weiskopf<sup>1,4</sup> |  
Gunther Helms<sup>1,5</sup> 

<sup>1</sup>Department of Neurophysics, Max Planck Institute for Human Cognitive and Brain Sciences, Leipzig, Germany

<sup>2</sup>Paul Flechsig Institute of Brain Research, Medical Faculty, Leipzig University, Leipzig, Germany

<sup>3</sup>Epidemiology of Highly Pathogenic Microorganisms, Robert Koch-Institute, Berlin, Germany

<sup>4</sup>Felix Bloch Institute for Solid State Physics, Faculty of Physics and Earth Sciences, Leipzig University, Leipzig, Germany

<sup>5</sup>Department of Clinical Sciences and Medical Radiation Physics, Lunds Universitet, Lund, Sweden

## Correspondence

Ilona Lipp, Department of Neurophysics, Max Planck Institute for Human Cognitive and Brain Sciences, Stephanstr. 1a, 04103 Leipzig, Germany.  
Email: [lippi@cbs.mpg.de](mailto:lippi@cbs.mpg.de)

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**Purpose:** Magnetization transfer saturation (MTsat) is a useful marker to probe tissue macromolecular content and myelination in the brain. The increased  $B_1^+$ -inhomogeneity at  $\geq 7$  T and significantly larger saturation pulse flip angles which are often used for postmortem studies exceed the limits where previous MTsat  $B_1^+$  correction methods are applicable. Here, we develop a calibration-based correction model and procedure, and validate and evaluate it in postmortem 7T data of whole chimpanzee brains.

**Theory:** The  $B_1^+$  dependence of MTsat was investigated by varying the off-resonance saturation pulse flip angle. For the range of saturation pulse flip angles applied in typical experiments on postmortem tissue, the dependence was close to linear. A linear model with a single calibration constant  $C$  is proposed to correct bias in MTsat by mapping it to the reference value of the saturation pulse flip angle.

**Methods:**  $C$  was estimated voxel-wise in five postmortem chimpanzee brains. “Individual-based global parameters” were obtained by calculating the mean  $C$  within individual specimen brains and “group-based global parameters” by calculating the means of the individual-based global parameters across the five brains.

**Results:** The linear calibration model described the data well, though  $C$  was not entirely independent of the underlying tissue and  $B_1^+$ . Individual-based correction parameters and a group-based global correction parameter ( $C = 1.2$ ) led to visible, quantifiable reductions of  $B_1^+$ -biases in high-resolution MTsat maps.

**Conclusion:** The presented model and calibration approach effectively corrects for  $B_1^+$  inhomogeneities in postmortem 7T data.

## KEYWORDS

calibration, chimpanzee, magnetization transfer, MRI, postmortem, transmit field, ultra high-field

## 1 | INTRODUCTION

Quantitative MRI (qMRI) is a powerful tool to study brain anatomy.<sup>1</sup> qMRI parameters in the brain provide measures of tissue myelination, and can be compared between in vivo and postmortem, across brain regions, across individuals, and even across species, opening the door for a plethora of neuroscience applications.<sup>1-6</sup> In the post-mortem brain, ultra-high field, ultra-high resolution qMRI facilitates studies of white matter myelination and cortical myeloarchitecture across the whole brain with resolutions down to tens of microns.<sup>7,8</sup>

Magnetization transfer (MT) is a contrast mechanism which is particularly specific to brain myelin. The extended lipid-water interface of myelin facilitates efficient MT between the MRI-visible water protons (the free water pool) and the MRI-invisible protons of macromolecules (the macromolecular pool).<sup>9,10</sup> MT imaging exploits the broad absorption line of macromolecular pool protons.<sup>1,9</sup> The broad line allows selective saturation of the macromolecular pool by off-resonance radiofrequency (RF) pulses applied at a frequency remote from the narrow resonance line of the free water (so-called MT pulses). Subsequent transfer of magnetization between the two pools results in a detectable reduction of the measured free water MRI signal.<sup>11</sup>

Various acquisition schemes for MT quantification have been developed, which differ in complexity, speed, precision, and accuracy.<sup>9,10,12,13</sup> One time-efficient approach is multiparameter mapping (MPM) to acquire maps of the MT saturation (MTsat).<sup>13,14</sup> The MPM approach consists of three RF-spoiled three-dimensional (3D) Fast Low Angle SHot (FLASH) acquisitions and provides a high signal-to-noise ratio (SNR),<sup>13,15,16</sup> it is therefore particularly suitable for ultra-high resolution qMRI on postmortem brains. MTsat values are dependent on the properties of the MT saturation pulse. MTsat cannot, therefore, be strictly considered as a physical parameter of the spin system within the tissue (unlike, e.g., relaxation rates, which are ideally independent of the acquisition method). However, if identical experimental conditions are achieved across the entire brains and between acquisitions (e.g. by using parallel transmit-based approaches<sup>17</sup>), MTsat provides semi-quantitative maps, proportional to the tissue macromolecular pool fraction. Alternatively, MTsat values can be corrected for their dependence on the MT saturation pulse properties (such as shape, length, and amplitude).

While ultra-high field offers advantages with respect to signal-to-noise ratio and is therefore a method of choice for ultra-high resolution imaging, it poses particular challenges for quantitative MT mapping. Increased

inhomogeneity of the RF transmit field ( $B_1^+$ ) at ultra-high field<sup>18,19</sup> results in a spatially varying saturation efficiency of the macromolecular pool which is reflected in a spatial variation in the MT estimates which needs to be corrected for.<sup>13,20-22</sup> The simple correction scheme proposed for in vivo 3T imaging, where the  $B_1^+$  dependency in MTsat is effectively removed by dividing the maps by  $(B_1^+)^2$ ,<sup>13</sup> does not work for the high-power MT pulses which are used in ultra-high resolution postmortem imaging. In vivo work at 3T and 7T demonstrated a correction for  $B_1^+$ -dependency on MTsat based on empirical calibration of the MTsat values.<sup>21,23,24</sup> Here we extend this empirical approach for MT pulses with even higher power for application in post-mortem MTsat imaging.

To achieve an effective correction, below we empirically determine the dependence of MTsat on  $B_1^+$  in a calibration experiment on postmortem chimpanzee brains. We demonstrate that a simple linear model using only one calibration parameter is sufficient to accurately correct the data. Our correction approach also allows for harmonizing across protocols that apply MT pulses with different flip angles. Thus, it can correct not only for spatial inhomogeneities in  $B_1^+$  but also account for different imaging protocols or scanner hardware modifications.

## 2 | THEORY

### 2.1 | Definition of MTsat

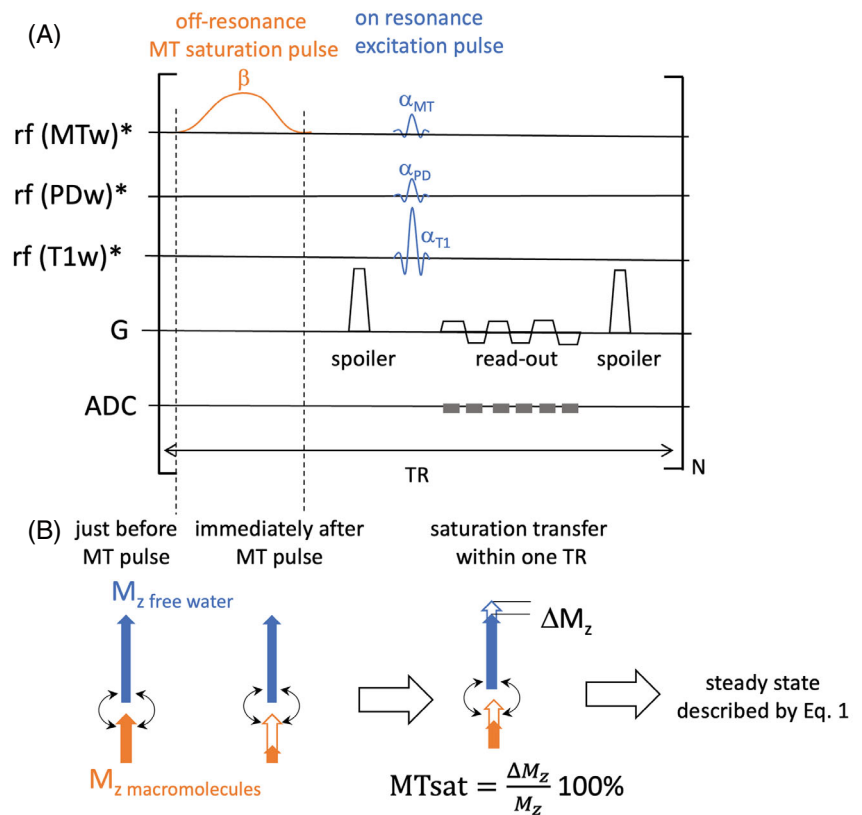
In the MT-weighted acquisition in the MPM acquisition scheme a strong off-resonance MT saturation pulse applied every repetition time (TR) causes selective saturation of the macromolecular pool, which is then transferred to the free water pool (Figure 1).<sup>11</sup> The MTsat is defined as the percentage difference in the free water pool magnetization over one TR period (Figure 1B) relative to the magnetization at the beginning of the TR,<sup>13</sup> and is an indirect measure of tissue macromolecular content,  $m^{25}$ .<sup>1</sup> MTsat is usually reported in percent units (p.u.).

### 2.2 | Estimation of MTsat

MTsat is estimated using three 3D FLASH acquisitions with different weightings (see Figure 1): (i) an MT-weighted acquisition ( $S_{MT}$ ) using a small on-resonance flip angle  $\alpha_{MT}$  and an off-resonance MT pulse with nominal flip angle  $\beta_{nom}$  which selectively

<sup>1</sup>MTsat differs from the commonly used MT ratio (MTR)<sup>12</sup> in that it is inherently corrected for TR and on-resonance excitation flip angle dependence.<sup>13,26</sup>

**FIGURE 1** (A) Schematic representation of the multiparameter mapping pulse sequence for estimation of magnetization transfer saturation (MTsat), consisting of three 3D Fast Low Angle SHot acquisitions, each with a different radiofrequency (RF) excitation scheme (three upper rows). The MT-weighted acquisition uses a low-flip angle on-resonance excitation pulse with flip angle  $\alpha_{MT}$ , and an off-resonance MT pulse with flip angle  $\beta_{nom}$  which selectively saturates the macromolecular pool. The proton density-weighted and T1-weighted acquisitions use low-flip angle ( $\alpha_{PD}$ ) and large-flip angle ( $\alpha_{T1}$ ) on-resonance pulses, respectively. Here we assume repetition time (TR) is the same for all acquisitions. (B) MTsat is defined as the difference in the free water magnetization due to the MT pulse over one TR, expressed in percent units of the free water pool magnetization. MTsat results from the exchange of magnetization between the macromolecular pool, which is selectively saturated by the MT pulse, and the free water pool and is dependent on macromolecular tissue content and the degree of saturation of the macromolecular pool.



saturates the macromolecular pool, (ii) a proton density (PD)-weighted acquisition ( $S_{PD}$ ) using a small on-resonance flip angle  $\alpha_{PD}$ , and (iii) a T1-weighted acquisition ( $S_{T1}$ ) using a large on-resonance flip angle  $\alpha_{T1}$ . Assuming MTsat,  $\alpha_{MT}$  and  $R1 \cdot TR$  are all small, MTsat can be estimated using:<sup>13,14</sup>

$$MTsat = \left( \frac{S_0 \alpha_{MT}}{S_{MT}} - 1 \right) R1 TR - \frac{(\alpha_{MT})^2}{2}, \quad (1)$$

where  $S_0$  is the equilibrium magnetization (proportional to PD) and  $R1$  is the longitudinal relaxation rate. We estimate  $S_0$  and  $R1$  using solutions of an exact algebraization of the Ernst equation<sup>2,27</sup>

$$S_0 = \frac{S_{PD} S_{T1} [\tan(\alpha_{T1}/2) / \tan(\alpha_{PD}/2)] - [\tan(\alpha_{PD}/2) / \tan(\alpha_{T1}/2)]}{2 [S_{T1} \tan(\alpha_{T1}/2) - S_{PD} \tan(\alpha_{PD}/2)]}, \quad (2)$$

and

$$R1 = \frac{2}{TR} \tanh^{-1} \left( \frac{S_{T1} \tan(\alpha_{T1}/2) - S_{PD} \tan(\alpha_{PD}/2)}{[S_{PD} / \tan(\alpha_{PD}/2)] - [S_{T1} / \tan(\alpha_{T1}/2)]} \right). \quad (3)$$

<sup>2</sup>This method of computing  $S_0$  and  $R1$  differs from that in Reference 13, which made use of small angle approximations with respect to  $\alpha_{PD}$  and  $\alpha_{T1}$  to simplify the calculation. We avoid relying on the small angle approximation here because large  $\alpha_{T1}$  are often used in postmortem high resolution imaging to impose sufficient T1-weighting (e.g., in our case  $\alpha_{T1} = 84^\circ$ ).

### 2.3 | $B_1^+$ bias in MTsat

Inhomogeneity of the transmit RF magnetic field  $B_1^+$  results in spatial variation of the flip angles ( $\alpha_{PD}$ ,  $\alpha_{T1}$ ,  $\alpha_{MT}$  and  $\beta_{nom}$ ) across the imaged object such that locally the flip angles are  $f_T \alpha_{PD}$ ,  $f_T \alpha_{T1}$ ,  $f_T \alpha_{MT}$ , and  $f_T \beta_{nom}$ , where  $f_T$  is the dimensionless local relative bias in  $B_1^+$  which is determined in a separate calibration experiment.<sup>28,29</sup> For convenience in the following we define

$$\beta_{loc} = f_T \beta_{nom}, \quad (4)$$

where subscript “loc” denotes the “local” flip angle.

Equations (1), (2), and (3) can be modified to account for the spatial variation of the on-resonance excitation flip angles by substituting  $f_T \alpha_{PD}$ ,  $f_T \alpha_{T1}$ , and  $f_T \alpha_{MT}$  for the respective nominal flip angles. However, Equation (1) contains an implicit dependence on  $\beta_{loc}$  which has not been accounted for: the spatial variation in  $\beta_{loc}$  will result in a spatial variation in the saturation of the macromolecular pool, which will result in a spatial variation of the computed MTsat.

MTsat is thus modulated by two factors: the local macromolecular pool fraction,  $m$ , and the local MT pulse flip angle,  $\beta_{loc}$ .<sup>13</sup> While the dependence of MTsat on  $m$  is an effect of interest used for myelin quantification, its dependence on  $\beta_{loc}$  and therefore on  $B_1^+$  leads to a

spatially dependent bias in MTsat maps which needs to be corrected for to enable accurate whole-brain myelin mapping. Here we present a correction method which aims to render the estimated MT saturation values insensitive to spatial  $B_1^+$  inhomogeneities for postmortem brain imaging at 7T, where formaldehyde-fixed tissue and MT saturation pulses with high power are used.

## 2.4 | Dependence of MTsat on the MT flip angle

### 2.4.1 | Model assumptions and plausibility arguments

We assume<sup>3</sup> that the dependence of MTsat on  $m$  and  $\beta_{loc}$  can be simply factorized into the product:<sup>11</sup>

$$\text{MTsat}(m, \beta_{loc}) = \zeta(m)\varphi(\beta_{loc}). \quad (5)$$

The plausibility of this assumption will be tested experimentally in Section 4. In the following we leave the dependence of MTsat on  $m$  implicit.

To correct for the effect of  $B_1^+$  inhomogeneity on MTsat we need to determine the function  $\varphi(\beta_{loc})$ .

Physical considerations reveal that  $\varphi(\beta_{loc})$  must have a sigmoidal dependence on  $\beta_{loc}$  (Figure 2A). At low  $\beta_{loc}$ , MTsat has a quadratic dependence due to the differential absorption law of the macromolecular pool.<sup>9</sup> On the other hand, it approaches a limiting value at very high  $\beta_{loc}$  as the macromolecular pool becomes fully saturated after every MT pulse. However, the exact functional form of  $\varphi(\beta_{loc})$  is in general unknown and is dependent on experimental conditions (magnetic field strength, MT pulse parameters) and characteristics of the investigated tissue (fixation method, temperature).

To provide a correction of MTsat for a particular experiment we must find a model of the functional form of  $\text{MTsat}(\beta_{loc})$  for the given experimental conditions over the range of  $\beta_{loc}$  observed in the experiment, which we can then use to correct MTsat to a reference MT flip angle,  $\beta_{ref}$ <sup>4</sup>.

Here we develop an empirical approach to obtain a correction for the imaging of postmortem brains at ultra-high fields. This approach is particularly useful for large brains

(such as humans, apes, whales, elephants) where the size of the brain is comparable with the electromagnetic field wavelength and therefore the inhomogeneity of  $B_1^+$  field across the brain is particularly strong. The essence of this approach is to make repeated measurements at a range of  $\beta_{loc}$  by scaling  $\beta_{nom}$  to obtain a set of  $\text{MTsat}(\beta_{loc})$  for each voxel.<sup>21,24</sup> These measured data can then be combined to elucidate the  $\beta_{loc}$  dependence and derive a correction for it. While we examine the specific case of chimpanzee brains at 7T here, the general empirical approach could be applied to derive a correction model for any case.

### 2.5 | Linear model of the dependence of MTsat on the MT flip angle

Figure 2B shows the empirical dependence of MTsat on  $\beta_{loc}$  measured in two distinct regions of a postmortem formalin fixed chimpanzee brain. The shaded area gives the range of  $\beta_{loc}$  values typically used in postmortem experiments. It can be seen that MTsat is empirically linear over the acquired range of  $\beta_{loc}$  for both a gray matter (low myelin) and a white matter (high myelin) region. The linear dependence of  $\beta_{loc}$  suggests that we are in a linear transition region between the quadratic regime and the saturated regime (Figure 2A). We thus propose a linear functional form for the dependence of MTsat on  $\beta_{loc}$ :

$$\varphi(\beta_{loc}) = 1 + (\beta_{loc} - \beta_{ref})A, \quad (6)$$

where  $A$  is a coefficient independent of  $m$  and  $\beta_{loc}$ , and  $\beta_{ref}$  is the reference MT saturation flip angle we want to map MTsat to. Here we have used the fact that the function  $\varphi(\beta_{loc})$  is defined up to a multiplicative factor to define  $\varphi$  such that  $\varphi(\beta_{ref}) = 1$ . Under these assumptions, Equation (5) can be rewritten as:

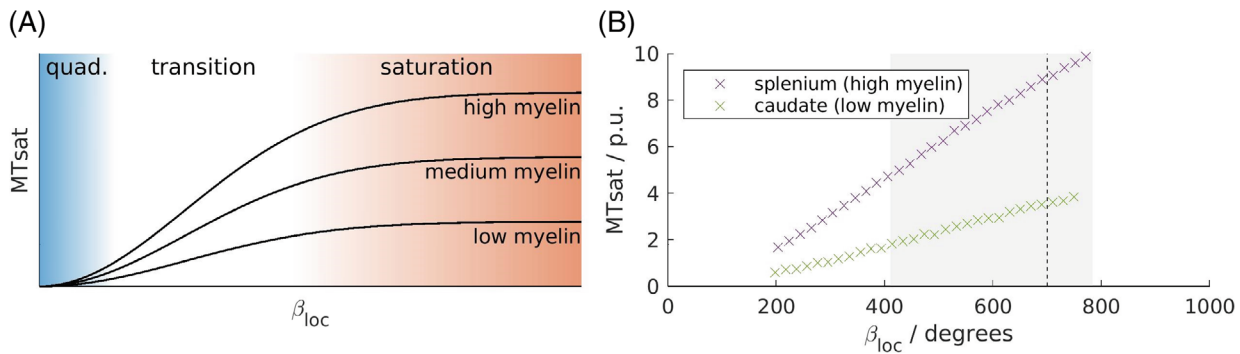
$$\text{MTsat}(\beta_{loc}) = (1 + (\beta_{loc} - \beta_{ref})A)\text{MTsat}(\beta_{ref}), \quad (7)$$

where  $\text{MTsat}(\beta_{ref})$  is the corrected MT saturation at  $\beta_{ref}$ . The assumed multiplicative dependence on  $\zeta(m)$  (see Equation 5) enters this equation through  $\text{MTsat}(\beta_{ref})$ . This model can also be regarded as the first order Taylor series approximation of  $\text{MTsat}(\beta_{loc})$  with respect to  $\beta_{loc}$  around  $\beta_{ref}$ , with  $\text{MTsat}(\beta_{ref})$  the constant term and  $A\text{MTsat}(\beta_{ref})$  the linear term.

The model parameter  $A$  can be estimated from the experimental data by voxel-wise linear regression of Equation (7) using  $\text{MTsat}(\beta_{loc})$  as the dependent variable and  $(\beta_{loc} - \beta_{ref})$  as the independent variable.  $\text{MTsat}(\beta_{ref})$  is then given by the intercept of the linear model, while the parameter  $A$  is given by the slope divided by the intercept of the model.

<sup>3</sup>We implicitly exclude potential contributions from direct saturation of the free pool by the MT pulse,<sup>11</sup> which would scale with  $\beta_{loc}$  but not with  $m$ . Both Bloch simulations of the direct saturation and fitting the experimental data with a model including a MTsat-independent term (data not shown), suggest that the direct saturation contribution is negligible ( $< 0.5$  p.u.) for our off-resonant pulses.

<sup>4</sup>While in a typical MTsat experiment  $\beta_{ref} = \beta_{nom}$ , which would make the definition of  $\beta_{ref}$  redundant, below we experimentally vary  $\beta_{nom}$  and so the distinction between these two angles becomes important.



**FIGURE 2** (A) Schematic representation of the  $\beta_{loc}$  dependence of magnetization transfer saturation (MTsat) for different myelination levels (black lines). Three regimes can be distinguished: a quadratic (quad.) scaling regime at low  $\beta_{loc}$  (blue), a transition region for intermediate  $\beta_{loc}$  (white), and the approach to full saturation of the macromolecular pool at high  $\beta_{loc}$  (red). MTsat values corresponding to anatomical regions with different myelination levels are represented by different multiplicative scalings of a universal function of  $\beta_{loc}$  ( $\varphi(\beta_{loc})$ ; Equation 5). (B) The experimentally observed  $\beta_{loc}$  dependence of MTsat in single voxels in two distinct anatomical regions: the highly myelinated splenium of the corpus callosum (purple) showing higher values of MTsat and the low myelinated caudate nucleus (green) showing lower values of MTsat. The gray shaded area shows the range of  $\beta_{loc}$  values over the sample when  $\beta_{nom} = \beta_{ref}$ , the reference MT flip angle (here  $700^\circ$ ), which is shown by the dashed line. MTsat shows an approximately linear increase with  $\beta_{loc}$  over this shaded area, suggesting that we are in the transition region. The previous  $B_1^+$  correction (used for example in the hMRI toolbox<sup>14</sup>) applies below this range.<sup>21,24</sup> Note that the ability to distinguish these two areas increases with increasing  $\beta_{loc}$ . p.u., percent units

## 2.6 | Correction of MTsat maps using the linear model

Once the value of  $A$  has been determined in a calibration experiment it can be used to correct the bias in MTsat maps. By rearranging Equation (7) for  $MTsat(\beta_{ref})$  we obtain the transformation of MTsat that corrects it from its value at  $\beta_{loc}$  to its value at  $\beta_{ref}$ :

$$MTsat(\beta_{ref}) = \frac{MTsat(\beta_{loc})}{1 + (\beta_{loc} - \beta_{ref})A}. \quad (8)$$

As in Reference 14, Equation (8) can be written in terms of a calibration parameter  $C = \beta_{ref}A$ :

$$MTsat(\beta_{ref}) = \frac{MTsat(\beta_{loc})}{1 + ([\beta_{loc}/\beta_{ref}] - 1)C}. \quad (9)$$

Using Equation (4) we can write  $\beta_{loc}/\beta_{ref} = f_T \beta_{nom}/\beta_{ref}$  and so

$$MTsat(\beta_{ref}) = \frac{MTsat(\beta_{loc})}{1 + (rf_T - 1)C}, \quad (10)$$

where

$$r = \beta_{nom}/\beta_{ref}. \quad (11)$$

The general Equation (10) applies for any  $\beta_{nom}$  within the range of validity of the model. However,

in the usual case where  $\beta_{ref} = \beta_{nom}$  (i.e., we have chosen the reference flip angle for the parameter  $C$  estimation to be the nominal flip angle which we will use for future datasets),  $r = 1$  and so Equation (10) simplifies to

$$MTsat(\beta_{ref} = \beta_{nom}) = \frac{MTsat(\beta_{loc} = f_T \beta_{nom})}{1 + (f_T - 1)C}. \quad (12)$$

For convenience in the following, we can write Equation (12) in terms of a local correction factor  $F(f_T, C)$

$$MTsat(\beta_{ref} = \beta_{nom}) = F(f_T, C)MTsat(\beta_{loc} = f_T \beta_{nom}), \quad (13)$$

where

$$F(f_T, C) = \frac{1}{1 + (f_T - 1)C}. \quad (14)$$

In Equation (12) the utilized  $C$  can be the result of fitting to data from each voxel separately, to data aggregated across all voxels in a single brain or to data aggregated across several brains. Consequently, the correction can be applied with voxel-specific, individual-based global or group-based global parameters to obtain less precise but more robust corrections. Since a generalization of the correction is desirable, that is, a globally fixed parameter would be preferred, a comparison between these approaches is made below.

## 3 | METHODS

### 3.1 | Postmortem tissue specimens

Five whole postmortem chimpanzee brains (two females, aged from 12 to 43 years) were studied. These brains were acquired from wild and captive deceased chimpanzees who died unexpectedly of causes not related to this study (see Table S1 for more information). The general preparation of the tissue is described in References 30–32. postmortem interval before fixation varied from 1 to over 16 hours between brains (Table S1). Different ages and different postmortem intervals before fixation resulted in a broad variation of  $R_1$  (between 1.1 and 2.6 s<sup>-1</sup>) and  $R_2^*$  (between 22 and 37 s<sup>-1</sup>; Table S1). Thus our sample covers the broad variety of animal ages, tissue fixation conditions, and variation of tissue quality typically used in postmortem experiments on hominoid brains.

### 3.2 | Data acquisition setup

For MRI data acquisition, the brains were placed in a spherical acrylic container filled with perfluoropolyether (Fomblin). Constant temperature during the scanning (27.5–33.5°C) was facilitated by a warm air stream and monitored by a sensor (LUXTRON Corporation). See Supplementary Methods in Appendix S1 for more details.

### 3.3 | MRI acquisition

All data were acquired on a 7T whole-body MRI system (Terra 7T, Siemens Healthineers) equipped with a 1-channel transmit/32-channel receive RF head coil (Nova Medical).

#### 3.3.1 | MPM protocol

An MPM protocol<sup>14,15,33</sup> was used (Figure 1) consisting of ultra-high resolution 3D FLASH images recorded with an isotropic resolution of 300  $\mu\text{m}$  (matrix: 432  $\times$  378  $\times$  288; readout bandwidth = 331 Hz/pixel; TR = 70 ms; 12 equidistant echoes (echo times [TE1, ..., TE12] = [3.63, ..., 41.7] ms); excitation flip angles:  $\alpha_{T_1} = 84^\circ$ ,  $\alpha_{PD} = 18^\circ$ ,  $\alpha_{MT} = 18^\circ$  for T1-, PD- and MT-weighted images, respectively). A Hann-filtered Gaussian-shaped MT saturation pulse at 3 kHz off-resonance with

$\beta_{\text{nom}} = 700^\circ$  and length 6 ms gave MT weighting. No partial k-space acceleration was employed.

#### 3.3.2 | Calibration experiment

The calibration experiments were performed using an MPM protocol with similar imaging parameters, but at a lower isotropic resolution of 2.1 mm (matrix: 64  $\times$  56  $\times$  48, bandwidth = 322 Hz/pixel, [TE1, ..., TE12] = [3.6, ..., 41] ms) to accelerate the acquisitions.

T1-weighted and PD-weighted images were acquired once at the beginning of each session, followed by MT-weighted images with  $\beta_{\text{nom}}$  ranging from 220° to 760° (in two brains to 700° due to reaching hardware safety limits), in 20° intervals. The order of the different acquisitions was pseudo-randomized (see Supplementary methods in Appendix S1) to balance out potential drifts related to heating or scanner instabilities.

A brain mask was obtained for each brain through intensity thresholding of the first echo of the T1-weighted images.

#### 3.3.3 | $B_1^+$ mapping

Maps of the RF transmit field  $B_1^+$  were obtained using the method in References 28, 29 using a spin echo–stimulated echo 3D-echo-planar imaging sequence (4 mm isotropic resolution, matrix: 48  $\times$  64  $\times$  48, TR = 500 ms, TE = 40.54 ms; mixing time = 34.91 ms; spin echo flip angles from 120°–330° in 15° increments; GRAPPA acceleration factor = 2  $\times$  2) and B0 mapping using a gradient echo sequence (2 mm isotropic resolution, matrix: 96  $\times$  96  $\times$  64, TR = 1020 ms, TE = 10 and 11.02 ms, excitation flip angle = 20°).

$B_1^+$  maps were computed with the hMRI toolbox.<sup>14</sup> A global reference T1 = 500 ms was used to account for T1 recovery during the mixing time between the spin echo and the stimulated echo.<sup>29,34</sup>  $B_1^+$  maps were smoothed using Gaussian smoothing (8 mm median filter kernel) and then divided by a brain mask smoothed in the same way.  $B_1^+$  maps were then upsampled to the respective resolution of the FLASH images using FSL *flirt* (using image header information only).  $f_T$  was determined by dividing the obtained relative  $B_1^+$  map in p.u. by 100%. Note that by using smoothed, low-resolution maps of  $B_1^+$  we interpolate across variations of  $B_1^+$  on a smaller scale and capture only the large variation of the  $B_1^+$  amplitude on the scale of the size of the brain with our correction.

### 3.4 | Preprocessing and MTsat calculation

MPM maps were computed separately for each nominal MT pulse flip angle ( $\beta_{\text{nom}}$ ).

All weighted FLASH images were corrected for off-resonance-related distortions in the readout direction, which alternated between odd and even echoes due to the bipolar readout scheme. First the geometric mean of the first and third echo was calculated, which is an estimate of a virtual second echo image acquired with the opposite readout polarity.<sup>35</sup> Using the second acquired echo and the virtual echo as input, the HySCO algorithm of the ACID toolbox (<http://diffusio.tools.com/>) was used to estimate the distortions and correct all acquired echoes.<sup>36</sup>

The effective transverse relaxation rate ( $R2^*$ ) was estimated using an ESTATICS<sup>37</sup> weighted log-linear least squares fit, as implemented in the hMRI toolbox.<sup>14,38</sup> No registration was performed between the weighted images. From this fit, the weighted images were extrapolated to  $TE = 0$ . These images were then used to calculate MTsat<sup>13</sup> using Equations (1), (2), and (3). In contrast to Reference 13, we used the local excitation flip angles in the calculation and not the nominal flip angles.

For the analysis with the low-resolution calibration data, we created binary masks excluding areas strongly affected by air bubbles. We did this by fitting a simple ordinary least squares model to describe the signal decay over the echos with an exponential function. Voxels in which the model explained less than 95% of the variance were excluded from the statistical analysis of the calibration data.

### 3.5 | Calibration parameter estimation and correction

For each voxel within the brain mask and each  $\beta_{\text{nom}}$  the local  $\beta_{\text{loc}}$  for the calibration experiment was calculated using Equation (4) and the experimental  $f_T$  map. The experimental dependence of MTsat on  $\beta_{\text{loc}}$  was fit voxel-wise using Equation (7) using nonlinear regression as implemented in MATLAB's `nlinfit(R2021a)`. The fit parameters provide a voxel-wise estimation of the calibration parameter  $A$  in Equation (7). `nlinfit` also gave an estimate of the standard errors of the parameters. The parameter  $C$  in Equation (12) was obtained by multiplying by the nominal target flip angle  $\beta_{\text{ref}}$  ( $700^\circ$  converted to radians). Similarly, the standard error of the fitted parameter  $C$  was obtained by multiplying the standard error of  $A$  by the nominal target flip angle  $\beta_{\text{ref}}$  ( $700^\circ$  converted to radians) and converted to % units relative to the estimated  $C$ .

When the  $\beta_{\text{loc}}$  is too low, MTsat cannot be estimated reliably.<sup>24</sup> We therefore excluded data points with  $\beta_{\text{loc}}$  less than our lowest nominal flip angle ( $220^\circ$ ) from the fit. Low signal-to-noise ratio and artifacts can give rise to negative MTsat estimates; these data points were also excluded.

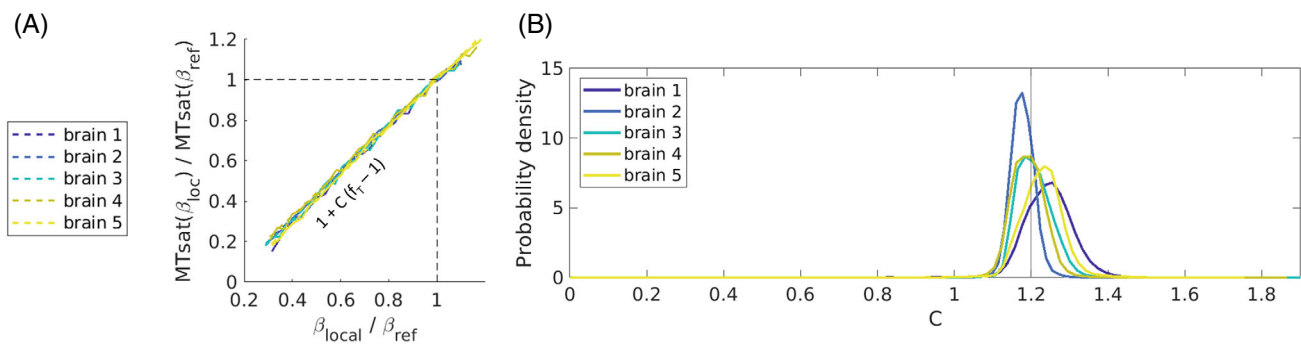
Descriptive statistics of  $C$  were calculated across all voxels in the brain mask. The mean of each calibration parameter across each specimen was used as an individual-based calibration parameter for this specimen. Then, the means of these individual-based parameters across all specimens were used as group-based calibration parameters (a group-based  $C = 1.2$  for  $\beta_{\text{nom}} = 700^\circ$  was determined). Individual-based parameters were rounded to two decimal places before applying bias correction. Finally, two different corrections were applied to high-resolution MTsat maps, using the  $B_1^+$  maps and the two different types of calibration parameters (individual-based vs. group-based).

## 4 | RESULTS

### 4.1 | The calibration coefficient $C$ : estimation uncertainty, within- and between-brain variation

Examples of the MTsat dependence on  $\beta_{\text{loc}}$  obtained in the calibration experiment for individual voxels are shown in Figure 3. Plotted in normalized coordinates the dependencies measured for different voxels and different brains overlapped within experimental error, supporting the plausibility of our model assumption of a universal dependence of MTsat on  $\beta_{\text{loc}}$ . The dependence of MTsat on  $\beta_{\text{loc}}$  was very well described by the proposed linear model for all brains (Table 1, Figures 3 and 4) as reflected in the high goodness of fit, with the average (median)  $R^2$  across the brain exceeding 0.95 in all five specimens (Table 1). The uncertainties of  $C$  estimated for each voxel and averaged across the brain lay between 0.6% and 1.1% of the mean value with the exception of brain 1 with an average uncertainty of 3.2% due to the scanner drift during the calibration experiment for this brain (Table 1, Figure 4). Larger errors in the estimation of  $C$  were observed at the edges of the brain (indicated by the large SD of the model fit in Figure 4) and around air bubbles within the sample, probably due to the low signal and the effect of mechanical and scanner drifts during the calibration experiment.

The means of the calibration coefficient  $C$  for the five brains were all close to 1.2 (varying between 1.18 and 1.24), with a mean of 1.209 and an SD of 0.026 between the brains corresponding to 2% of the mean  $C$  value. The histograms of  $C$  for the five brains showed unimodal overlapping distributions (Figure 3). The between-voxel variation



**FIGURE 3** Distribution of the calibration parameter  $C$ . (A) Voxel-wise dependencies of magnetization transfer saturation (MTsat) on  $\beta_{loc}$  obtained in the calibration experiment for exemplary white matter voxels from five brains (solid lines) together with linear model fit (dashed lines). The dependencies are presented in the normalized coordinates  $MTsat(\beta_{loc})/MTsat(\beta_{ref})$  and  $\beta_{loc}/\beta_{ref}$ . The linear model yielded high goodness of fit, providing voxel-wise estimation of  $C$ . In the normalized coordinates  $C$  corresponds to the slope of the fitted linear dependence. Note that experimental data from all brains nearly overlapped when plotted in normalized coordinates and are therefore described well by similar values of  $C$ . (B) Histograms for  $C$  across all voxels of each specimen. The histograms obtained for five brains overlapped and were centered around  $C = 1.2$  (gray line), with some variation within and between the brains.

**TABLE 1** Estimated values for the calibration parameter  $C$  and related variances

Brain	$C$	Between voxel variance of $C$ (% from mean $C$ )	Within-voxel uncertainty of $C$ (% from fitted $C$ )	Model fit	$\rho_{part}(MTsat) C$ versus $B_1^+$	$\rho_{part}(B_1^+) C$ versus MTsat
Brain 1	$1.242 \pm 0.071$	5.733	$3.233 \pm 23.160$	$0.952 \pm 0.096$	-0.124	-0.635
Brain 2	$1.177 \pm 0.030$	2.567	$0.643 \pm 0.526$	$0.996 \pm 0.013$	-0.503	-0.534
Brain 3	$1.207 \pm 0.046$	3.802	$0.743 \pm 0.978$	$0.993 \pm 0.029$	-0.390	-0.622
Brain 4	$1.191 \pm 0.047$	3.951	$1.149 \pm 10.634$	$0.988 \pm 0.042$	-0.369	-0.574
Brain 5	$1.227 \pm 0.053$	4.332	$0.960 \pm 1.841$	$0.990 \pm 0.039$	-0.312	-0.747

**Notes:** The mean, between-voxel variance and the within-voxel uncertainty of the estimated parameters  $C$  across all voxels for the five brains are provided. Voxels with values of  $C < 0$  (assumption that only positive correlations are physical) or  $C > 1.4$  (assumption that  $MTsat(\beta_{loc})$  is positive for all  $\beta_{loc}$  down to  $220^\circ$  given  $\beta_{ref} = 700^\circ$ ; derived from Equation 10) were excluded from these statistics. Uncertainties and variance are provided in % from  $C$ . The within-voxel uncertainty was estimated from the SD of the fitted parameter and is reported in % from the fitted  $C$ . The between-voxel variance was estimated as the SD across the brain converted to % of mean  $C$ . The quality of the linear model fit (coefficient of determination  $R^2$ ) is provided as mean  $\pm$  SD across the brain. To quantify the dependence of the estimated  $C$  on the underlying  $B_1^+$  and the corrected tissue MTsat, whole-brain voxel-wise partial Spearman correlation coefficients were calculated.

(standard deviation; SD) of  $C$  within each brain ranged from 2.6% to 5.7% of the mean value and exceeded the averaged within-voxel uncertainty of the  $C$  and between brain variation of  $C$  (Table 1).

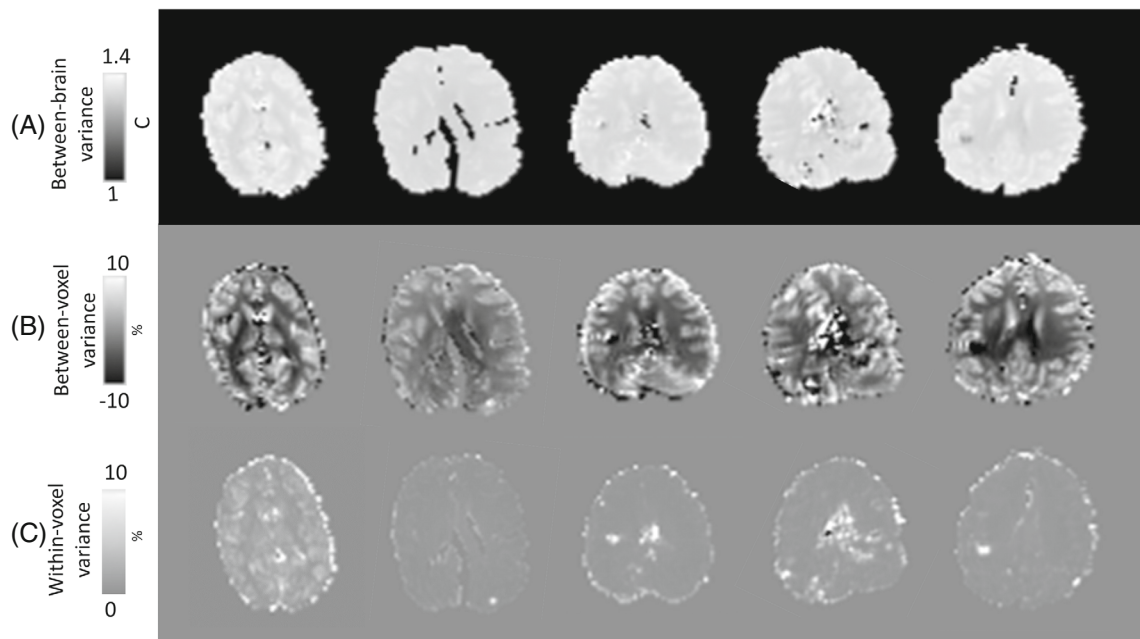
That we obtained similar values of the calibration coefficient for brains with different ages and varying fixation conditions demonstrates the wide generalizability of the proposed calibration approach for postmortem brain imaging.

## 4.2 | Residual tissue-type dependence of $C$

A key assumption behind the applied approach was that  $C$  is independent of the macromolecular content and tissue

type (Equation 5). This assumption was largely supported by the experimentally obtained values of  $C$  (Figure 3), which showed very close agreement across the brains and even between the brains with different fixation conditions and from individuals of different ages. However, the spatial distribution of  $C$  demonstrated residual systematic contrast between gray and white matter regions, with white matter showing on average 2.5% lower values of  $C$  as compared to gray matter (Figure 4). To illustrate the residual dependence of  $C$  on tissue type and underlying  $B_1^+$ , we computed the histograms of  $C$  within cortical and white matter voxels separately in one of the brains (Figure 5). These show different distributions of  $B_1^+$  and  $C$  between cortex and white matter. Also,  $C$  showed small systematic differences between areas of low and high  $B_1^+$ . Figure 5B shows that the effects of  $B_1^+$  and tissue type are partly





**FIGURE 4** Variance of the calibration coefficient  $C$ . (A) Maps of calibration coefficient  $C$  for the five investigated brains obtained by a voxel-wise fit of the experimental dependence of MTsat on  $\beta_{1oc}$ . All five brains demonstrated similar values of the calibration coefficient  $C$  all close to 1.2, with very low variation across the brain. (B) Between-voxel variance in  $C$  is illustrated through % difference from the mean value. (C) Maps of within-voxel uncertainty for  $C$  estimates, expressed in the parameter SD converted to % to  $C$  of that voxel

independent of each other, as the relationship between  $B_1^+$  and  $C$  is clearly visible even if just white matter voxels are considered. Additional analysis of this residual dependence using Spearman correlation is presented in “Supplementary residual  $C$  dependence on tissue type and  $B_1^+$ ” in the Appendix S1.

Next, we tested the practical relevance of this residual dependencies by comparing the bias correction with the voxel-wise versus average  $C$  values.

### 4.3 | Bias correction in MTsat maps: comparing calibration approaches

We tested the performance of the proposed calibration approach for the correction of the bias in MTsat maps resulting from the inhomogeneity of  $B_1^+$ . For each brain we corrected the bias in MTsat maps in three different ways, “voxel-wise,” “individual-based,” and “group-based.” Voxel-wise refers to using each voxel’s estimated  $C$  parameter, and individual-based refers to using the specific brain’s median calibration parameter  $C$  (as reported in Table 1). Group-based refers to correcting with a fixed set of calibration parameters, i.e., the mean across all brains ( $C = 1.2$ ).

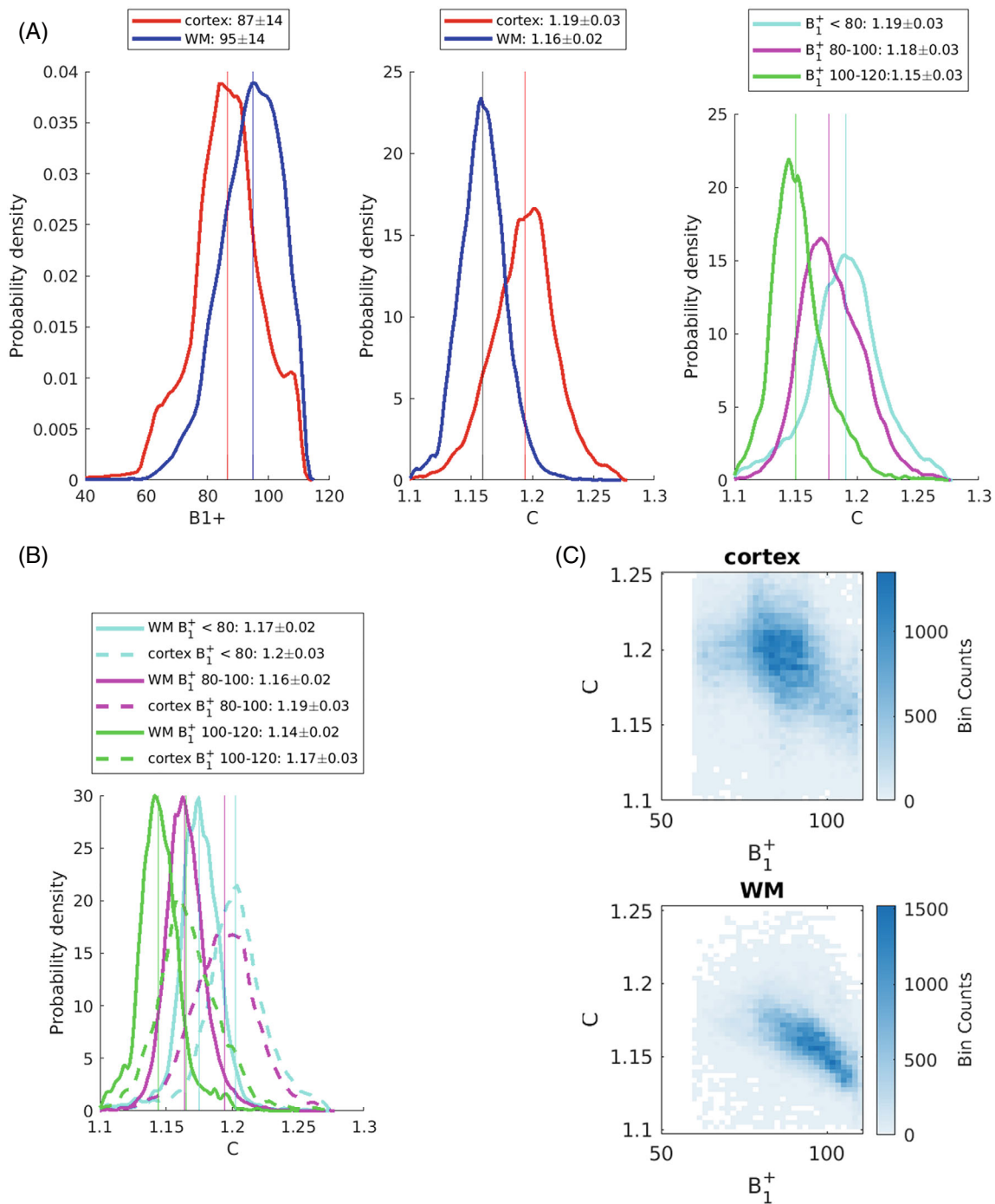
Given the residual spatial dependence of  $C$ , in theory, corrections with voxel-wise  $C$  should provide most accurate results, while requiring a time-consuming calibration

experiment which is only feasible for low-resolution MTsat maps, since repeating the whole calibration experiment for lengthy high resolution scans would require infeasibly long scanning times. Corrections with whole-brain values also requires a calibration experiment on each sample, but would be feasible for ultra-high resolution data by performing a calibration experiment at low resolution. Correction of group-based mean values is less accurate but most time efficient, since it requires a calibration experiment for only a subset of representative samples. In the following we evaluated the difference between these three proposed approaches.

#### 4.3.1 | Voxel-based versus individual-based correction

Figure 6 shows the comparison between voxel-wise and individual-based correction in one exemplary brain. Visible reduction of the bias was achieved with both approaches, with the average difference between corrected maps lying within the  $\pm 0.5$  p.u. interval.

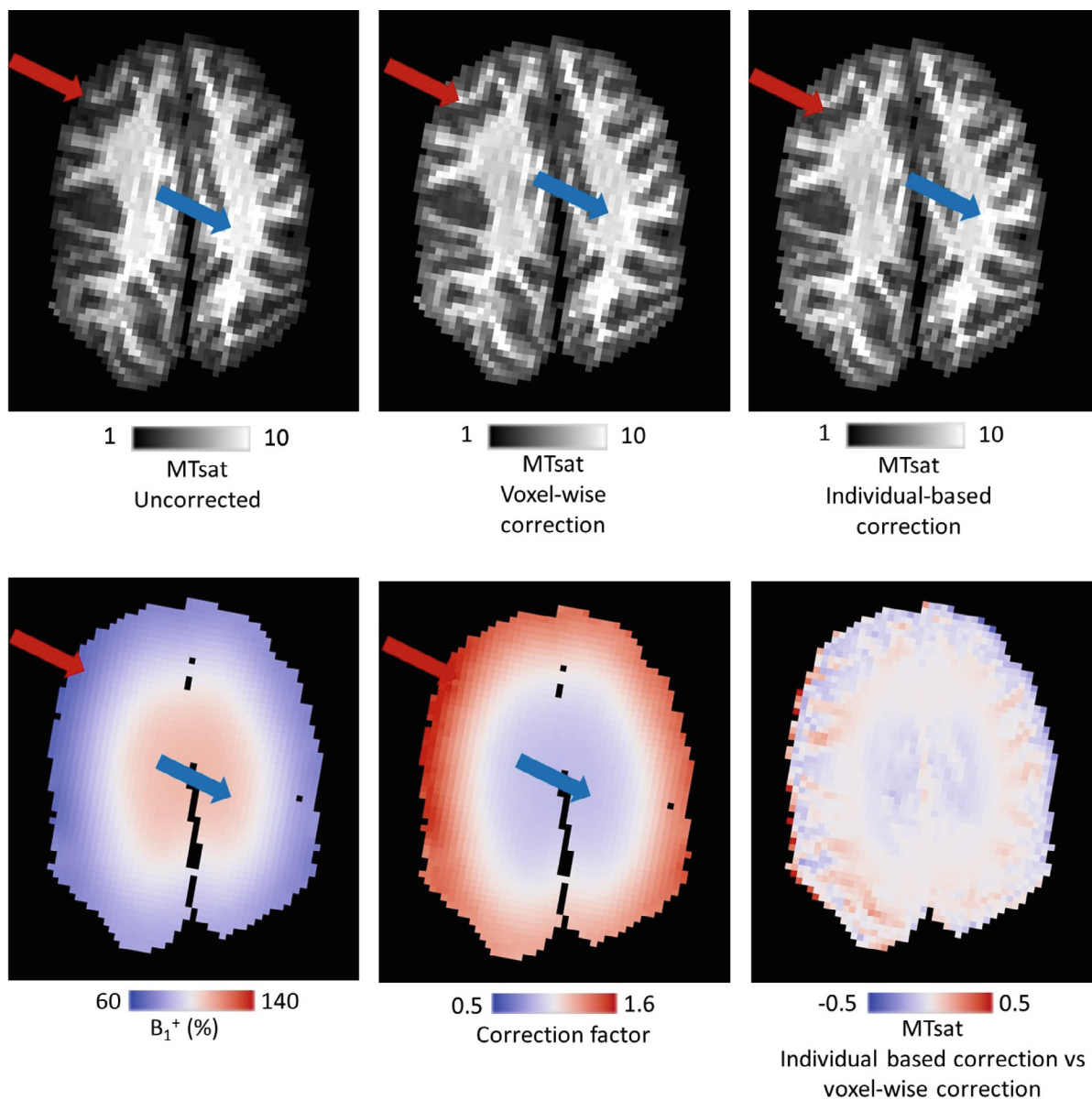
To quantify the  $B_1^+$  bias and the effect of the correction, spatial Spearman correlations were calculated between the apparent MTsat and  $B_1^+$  across the brain. Due to field focussing in head-sized objects in ultra high field MRI,  $B_1^+$  has a maximum at the center of the brain.<sup>18,19</sup> Therefore the  $B_1^+$  distribution is spatially correlated with



**FIGURE 5** Effects of tissue type and  $B_1^+$  on  $C$  in a representative brain (specimen 2). (A) Left: Histograms show lower values of  $B_1^+$  in cortex than in white matter (WM). Middle: Parameter  $C$  is higher in cortical voxels than in WM voxels. Right: Distributions of  $C$  based on  $B_1^+$ , showing the lowest values of  $C$  in regions with the highest  $B_1^+$ . Both tissue type and  $B_1^+$  independently explain variance in  $C$  across the brain, as quantified by partial correlation coefficients (Table 1). Median  $\pm$  interquartile range are provided in the legend. (B) The interaction between tissue type and  $B_1^+$ . The effect of  $B_1^+$  (denoted by different colors) can be seen for both tissue types (WM in solid line, cortex in dashed line). (C) The relationship between  $B_1^+$  and  $C$  is additionally illustrated in density scatter plots for cortex and WM voxels separately. The relationship is particularly visible in the WM. Cortex and white matter masks were obtained using Freesurfer (<https://surfer.nmr.mgh.harvard.edu/>).

anatomy (white matter in the middle of the brain, gray matter in the periphery). We regressed out the effect of anatomy by using the  $B_1^+$ -independent measure  $R2^*$

(Table S2). The correlation coefficients for uncorrected data ranged from  $r = 0.422$  to  $r = 0.633$ . All three correction approaches were able to reduce that bias (reducing



**FIGURE 6** Example maps of MTsat. Top: Correction of the MTsat map. An axial slice of the uncorrected map is shown and compared to a voxel-wise correction and the individual-based correction approach. The red arrow indicates a region of low  $\beta_{loc}$ , whose hypointensity was corrected. The blue arrow indicates an area of high  $\beta_{loc}$ , whose hyperintensity was corrected. Bottom: The  $\beta_{loc}$  and corresponding correction factor (calculated using Equation 14, with either voxel-wise or individual-based mean  $C$ ) are shown, along with the difference map between voxel-wise and individual-based correction. Tissue contrast and features of  $\beta_{loc}$  are visible in this map.

the correlations to between  $r = -0.116$  and  $r = 0.307$ ). The correction had an average effect on MTsat of 11%–16% of the uncorrected value, while the difference between correction approaches lay in the range of less than 0.5% (for all numbers see Table S2). Therefore, the observed variability in  $C$  across the brain is negligible compared to the  $B_1^+$ -bias in MTsat maps. Additionally, comparing the voxel-wise to the individual-based correction revealed some voxels with very large differences (indicated by the max values reported in Table S2) indicating that voxel-wise corrections fail in some regions prone to artifacts induced by

either low signals or scanner instability during the calibration experiment, which can particularly affect brain edges.

#### 4.3.2 | Individual-based versus group-based correction

For each brain we also  $B_1^+$ -bias corrected the ultra-high resolution MTsat maps in two different ways, “individual-based” and “group-based,” as described above.

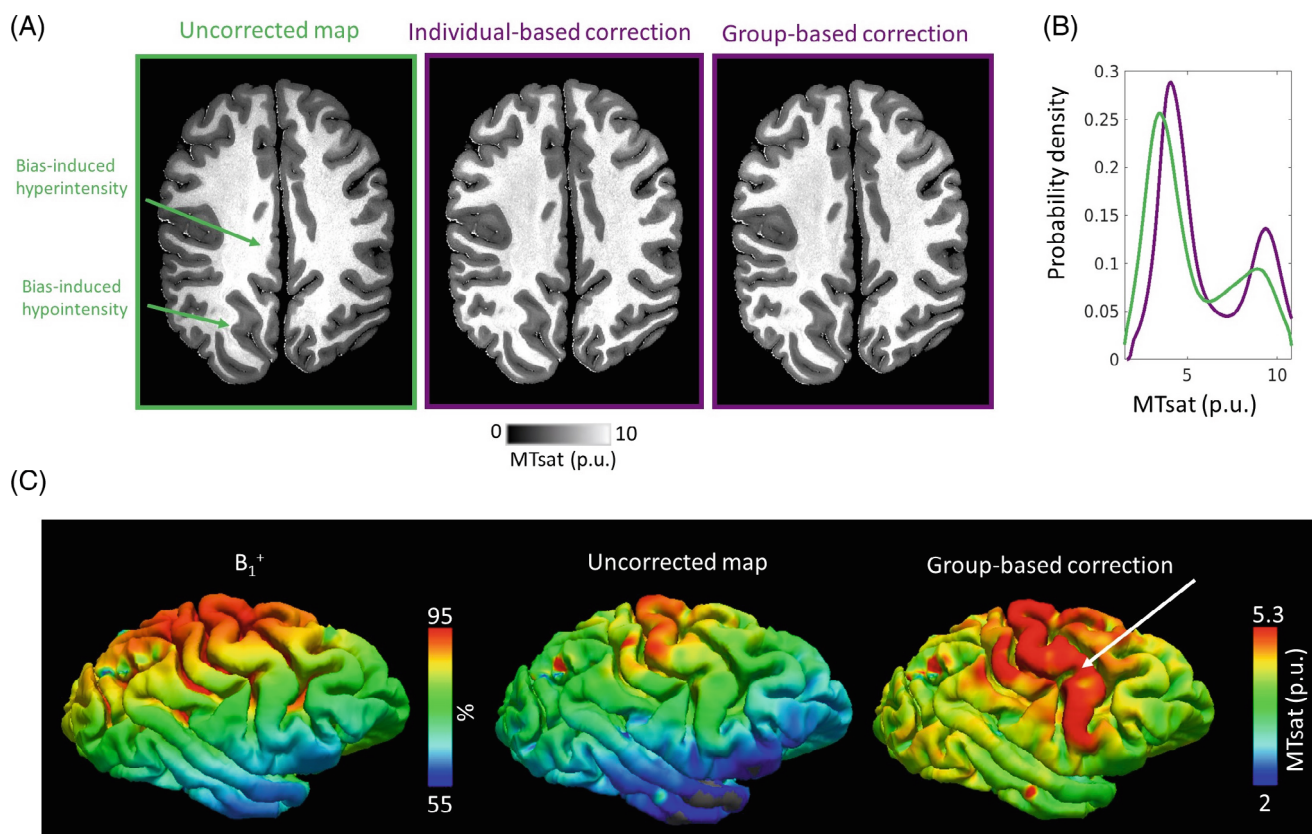
In the  $B_1^+$ -uncorrected data, partial correlation coefficients ranging from  $\rho = 0.440$  to  $\rho = 0.589$  were observed, reflecting the  $B_1^+$ -induced bias in MTsat. The relationship between  $B_1^+$  and MTsat was lower for all corrected data sets and ranged from  $\rho = -0.094$  to  $\rho = 0.196$  with the individual-based correction and from  $\rho = -0.077$  to  $\rho = 0.201$  with the group-based correction. We also quantified the effect of the applied corrections on the MTsat values. Across all voxels, the average effect of applying the correction was an absolute change between 11.75% and 18.90% in MTsat (Table S3). Comparing the two correction approaches to each other, the individually and group-based corrected MTsat maps differed on average by 0.14%–0.42%.

Figure 7A,B illustrates that bias correction of the MTsat values across the brain visibly reduces the bias and yields more distinct histogram peaks of gray and white matter. Figure 7C shows that uncorrected cortical surface MTsat maps strongly resemble the  $B_1^+$  distribution. After

correction the maps show patterns reflecting myeloarchitecture, with the highly myelinated primary cortical areas standing out. For example, the expected high myelination of primary motor and primary somatosensory cortex along the central gyrus only becomes apparent after the correction. This demonstrates that bias correction in MTsat maps is a crucial step when studying cortical myelination across the entire brain and enables quantitative comparison of myelination between cortical areas.

## 5 | DISCUSSION

We have developed a calibration correcting for  $B_1^+$ -induced biases in MTsat maps and demonstrated its efficacy. It extends previous calibration approaches to the higher 7T static magnetic field strength and the stronger MT saturation pulses used for postmortem imaging. A high goodness of fit of the calibration model to experimental data



**FIGURE 7** Correction of high-resolution images. (A) The effect of  $B_1^+$  bias correction on an example high resolution MTsat axial slice from brain 2. (B) A histogram comparing the distribution of the uncorrected map (green) to correction with the individual-based parameters and group-based parameters (purple). The corrected maps show an emphasized bimodal distribution of values, reflecting gray and white matter, while the uncorrected map provides a poorer distinction. The bias visible in the uncorrected map is reduced in both corrected maps. The probability density plot was created with the MATLAB function `ksdensity` using 100 bins and excluding outliers (data points below the second or above the 98th percentile). (C) The distribution of  $B_1^+$  (left) on the Freesurfer (<https://surfer.nmr.mgh.harvard.edu/>) mid-cortical surface of brain 2 is shown, demonstrating that  $B_1^+$  can also systematically vary across cortical regions. If not corrected, this is also reflected in the cortical MTsat maps (middle). Our correction eliminates this bias, revealing the highly myelinated primary cortical areas (right; arrow).

supports the theory-based correction approach. Although the status and quality of the tissue varied significantly, we found a single fixed calibration parameter that significantly reduced the  $B_1^+$ -induced bias in all postmortem data sets, corroborating the generalizability. This simplifies the implementation as a standard tool, since calibration parameters do not need to be estimated for each postmortem specimen individually, which would require the acquisition of additional reference data for each specimen.

## 5.1 | Factors influencing the estimation of $C$

The presented calibration requires the experimental estimation of the calibration parameter  $C$ . A strong overall goodness of fit suggests the validity of the theoretical model. The  $R^2$  of  $>0.95$  we obtained for all brains vastly exceeds the average  $R^2$  of 0.20 that was reported in a similar experiment conducted on humans in vivo, in which the model was fit over large regions which were assumed to be homogeneous.<sup>24</sup> Here, we fit the model and estimate  $C$  voxel-by-voxel, which, unlike region-based analyses, captures spatial inhomogeneities of the calibration curve. However, voxel-wise estimations are more sensitive to variations in statistical noise,  $B_1^+$  errors and the underlying tissue. This limitation is lifted by using whole brain average or group averaged calibration coefficients.

The use of a global  $C$  was based on an assumption that  $C$  does not depend on  $B_1^+$  amplitude, tissue characteristics or tissue type. In our data, the calibration parameters systematically differed between gray and white matter by about 4% (also apparent in a significant spatial correlation between  $C$  with the corrected MTsat values). This indicates that using different values of  $C$  for gray and white matter may provide better correction. Note that in the one brain where we analyzed the calibration coefficient  $C$  separately in gray and white matter the global calibration coefficient  $C$  was closer to the maximum of the distribution of fitted  $C$  values for gray matter voxels than to the white matter maximum. One reason for this may be the fact that gray matter is located on the periphery of the brain, in regions with low  $B_1^+$  and therefore requires stronger correction. Also some residual dependence on  $B_1^+$  was observed. Higher local  $B_1^+$  is expected in regions of higher conductivity but can only be observed using high-resolution  $B_1^+$ -mapping, for example, by MP3RAGE.<sup>39</sup> We calculated global, individual-based calibration parameters for each brain by taking the mean values across all voxels. Corrections using voxel-wise values of  $C$  did not result in a better correction than using brain-averaged values of  $C$

(Table S2), so this residual tissue and  $B_1^+$  dependence has a negligible effect on correction performance.

## 5.2 | Consistency of $C$ across brains

Overall, the variability of the estimated calibration parameters within individual brains (Table 1) was larger than variability between brains. This indicates that factors such as tissue quality, postmortem time etc. play only a minor role and a generalized correction is possible. Correction using the group-based mean calibration parameters of our sample ( $C = 1.2$ ) visibly reduced the bias in all brains. Individually optimising the calibration coefficients may yield more accurate correction results. However, when comparing the individually and group-based corrected maps, average differences of less than 0.5% were obtained. In comparison to the average 12%–19% change that either correction had on the uncorrected maps, this appears minor. Therefore, a correction with the group-based parameters is recommended when individual-based parameters are not available, that is, when additional reference data are not available.

For data obtained postmortem at 7T with the acquisition parameters used in our study, the following equation with group-based  $C = 1.2$  can be applied for correction:

$$\text{MTsat}(\beta_{\text{ref}} = 700^\circ) = \text{MTsat}(\beta_{\text{loc}} = f_T \cdot 700^\circ) \frac{1}{1+1.2(f_T-1)}. \quad (15)$$

## 5.3 | Correction approach reveals true biological variability

As a ground-truth for MTsat is generally not available, we assessed the performance of the  $B_1^+$ -bias correction by directly comparing the different types of corrections with respect to their anatomical validity. All correction approaches had a significant impact on MTsat values, with an average change of up to 20% of the uncorrected value. The two separate gray and white matter peaks in whole sample MTsat histograms became better separated after correction (Figure 7). This indicates that nonbiological sources of variance were being reduced. Additionally, cortical surface projections of uncorrected MTsat maps lead to visible biases that correspond to the  $B_1^+$  distribution across the cortex (Figure 7). Visual assessment of corrected and original uncorrected MTsat maps showed reduced bias and clearer delineation of anatomical structures (e.g. motor and somatosensory cortex; Figure 7).

## 5.4 | Previous approaches

We used an empirical approach to determine a functional form for the correction of  $B_1^+$  bias. The functional form could alternatively be elucidated by using forward models of the MT effect to simulate the dependence of MTsat on  $\beta_{loc}$  and  $m$ .<sup>22</sup> However this method is limited by the need for reliable forward model parameter estimates (exchange times, pool size ranges and relaxation times/lineshapes of the macromolecular and free pools). These are sensitive to tissue preparation methods, difficult to measure, and not generally available.

Comparison of the correction method presented here to the previous in vivo correction method used in References 21,23,24 can be found in the subsection “Comparison with the previous calibration model” in Appendix S1.

## 5.5 | Limitation: potential acquisition protocol dependence

Although a linear dependence on the flip angle is expected in the transition region, this specific calibration approach was only tested on one hardware setup and protocol. We show in Appendix S1 how Equation (15) can be modified (see Equations (S.1) and (S.2)) to map the measured MTsat to different  $\beta_{ref}$  flip angles within the region where the linear model applies.

The value of the global calibration coefficient ( $C = 1.2$ ) reported in this study was obtained for formalin-fixed chimpanzee brains at 7T using a specific MPM protocol and MT saturation pulse. This could potentially limit general applicability of this specific  $C$  coefficient to other MT protocols. However, we note that  $C$  is a normalized estimate of the slope of MTsat with respect to  $\beta_{nom}$ . This normalization can be seen clearly in the case of a two point estimate of  $C$  from MTsat measured at two distinct MT pulse flip angles  $\beta_1$  and  $\beta_2$ ,

$$C = \frac{MTsat(\beta_2) - MTsat(\beta_1)}{\beta_2 - \beta_1} \frac{\beta_{ref}}{MTsat(\beta_{ref})}, \quad (16)$$

where the first product term is the estimate of the slope and the second corresponds to the effective normalization: the numerator is the scaling between  $A$  and  $C$  (i.e.,  $C = \beta_{ref}A$ ) and the denominator corresponds to the factorization of the slope into a product with  $MTsat(\beta_{ref})$  (see Equation 7). Such a normalized parameter will tend to be reasonably robust to small changes in the protocol and flip angles, as changes in the respective numerators and denominators will tend to cancel each other out. Therefore we expect that similar calibration coefficients will be obtained for

different experimental implementations (e.g., for different MR system vendors).

However, if data are obtained with major changes in the acquisition protocol (e.g.,  $\beta_{nom}$  far outside the experimental range used in the calibration), then recalibration may be required using the described calibration experiment. The calibration experiment is generally applicable to the broad range of potential experimental settings and estimation of  $C$  can be easily performed on a small number of postmortem specimens. The resulting group-based calibration parameters can then be used to correct all additional specimens. Future research may investigate which sequence parameters have no or little impact on the calibration parameters.

## 6 | CONCLUSION

We developed a  $B_1^+$  correction of MT saturation (MTsat) maps using a calibration approach. It extends previous calibration approaches and enables quantitative post-mortem MT mapping using high power MT saturation pulses at 7T. We showed that a single correction coefficient can visibly reduce  $B_1^+$ -related biases in low resolution data, high-resolution data and on cortical myelination maps.

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Asimwe (Budongo Conservation Field Station, Uganda), Jenny E. Jaffe (Project Group Epidemiology of Highly Pathogenic Microorganisms, Robert Koch Institute, Berlin, Germany; Tai Chimpanzee Project, Centre Suisse de Recherches Scientifiques, Abidjan, Côte d'Ivoire), Fabian Leendertz (Epidemiology of highly pathogenic microorganisms, Robert Koch-Institute, Berlin, Germany; Helmholtz Institute for One Health, Greifswald, Germany; Tai Chimpanzee Project, CSRS, Abidjan, Ivory Coast), Matyas Liptovszky (Twycross Zoo, UK), Sophie Moitie (Twycross Zoo, UK), and Andrea Pizarro (Tacugama Chimpanzee Sanctuary, Freetown, Sierra Leone). Open Access funding enabled and organized by Projekt DEAL.

### CONFLICT OF INTEREST


The Max Planck Institute for Human Cognitive and Brain Sciences has an institutional research agreement with Siemens Healthcare. NW holds a patent on acquisition of MRI data during spoiler gradients (US 10401453 B2). NW was a speaker at an event organized by Siemens Healthcare and was reimbursed for the travel expenses.

### DATA AVAILABILITY STATEMENT

The analysis code can be found at <https://github.com/IlonaLipp/MTcalibration>. Data associated with this manuscript can be found at <https://doi.org/10.17617/3.803OMM>.

### ORCID

Ilona Lipp  <https://orcid.org/0000-0002-5644-2057>

Evgeniya Kirilina  <https://orcid.org/0000-0003-1304-4363>

Kerrin J. Pine  <https://orcid.org/0000-0002-2486-2970>

Gunther Helms  <https://orcid.org/0000-0002-1371-5123>

### TWITTER

Ilona Lipp  @ilona\_lipp

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## SUPPORTING INFORMATION

Additional supporting information may be found in the online version of the article at the publisher's website.

### Appendix S1. Supporting information

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