Ultra-high field (10.5T) diffusion-weighted MRI of the macaque brain

Mark D. Grier a, Essa Yaacob b,c, Gregor Adriany b,c, Russell L. Lagore b, Noam Harel b,d, Ru-Yuan Zhang f,g,b, Christophe Lenglet b, Kâmil Ügurbil b,c, Jan Zimmermann b,c,e,#, Sarah R. Heilbrunner a,c,#,*

a Department of Neuroscience, University of Minnesota, Minneapolis, MN 55455, United States
b Center for Magnetic Resonance Research, Department of Radiology, University of Minnesota, Minneapolis, MN 55455, United States
c Center for Neuroengineering, University of Minnesota, Minneapolis, MN 55455, United States
d Department of Neurosurgery, University of Minnesota, Minneapolis, MN 55455, United States
e Department of Biomedical Engineering, University of Minnesota, Minneapolis, MN 55455, United States
f Institute of Psychology and Behavioral Science, Shanghai Jiao Tong University, Shanghai 200030, P.R. China
g Shanghai Mental Health Center, Shanghai Jiao Tong University School of Medicine, Shanghai Jiao Tong University, Shanghai 200030, P.R. China

A R T I C L E I N F O

Keywords:
Diffusion MRI
Structural connectivity
Nonhuman primate
Tractography

A B S T R A C T

Diffusion-weighted magnetic resonance imaging (dMRI) is a non-invasive imaging technique that provides information about the barriers to the diffusion of water molecules in tissue. In the brain, this information can be used in several important ways, including to examine tissue abnormalities associated with brain disorders and to infer anatomical connectivity and the organization of white matter bundles through the use of tractography algorithms. However, dMRI also presents certain challenges. For example, historically, the biological validation of tractography models has shown only moderate correlations with anatomical connectivity as determined through invasive tract-tracing studies. Some of the factors contributing to such issues are low spatial resolution, low signal-to-noise ratios, and long scan times required for high-quality data, along with modeling challenges like complex fiber crossing patterns. Leveraging the capabilities provided by an ultra-high field scanner combined with denoising, we have acquired whole-brain, 0.58 mm isotropic resolution dMRI with a 2D single-shot echo planar imaging sequence on a 10.5 Tesla scanner in anesthetized macaques. These data produced high-quality tractograms and maps of scalar diffusion metrics in white matter. This work demonstrates the feasibility and motivation for in-vivo dMRI studies seeking to benefit from ultra-high fields.

1. Introduction

Diffusion-weighted magnetic resonance imaging (dMRI) is a non-invasive neuroimaging modality that is sensitive to the restriction of water molecules within tissue (Le Bihan et al., 1986). In the brain, dMRI has been used to study pathology in multiple disease states, including Alzheimer’s Disease, Parkinson’s Disease and schizophrenia (Granberg et al., 2017; Kamagata et al., 2020; Kubicki et al., 2007). More specifically, diffusion tensor imaging (DTI) uses this information to characterize the microstructure and primary fiber orientation present in each voxel (Le Bihan et al., 2001; Mori and Zhang, 2006). Other, more recent models, including constrained spherical deconvolution (CSD) (Tournier et al., 2008) and neurite orientation dispersion and density imaging (NODDI) (Zhang et al., 2012), also estimate multiple fiber orientations as well as axonal density in brain tissue. These estimates can then be utilized by tractography algorithms to visualize white matter pathways (Behrens et al., 2003; Tournier et al., 2019) and estimate structural brain connectivity.

The majority of dMRI brain studies (in both humans and nonhuman primates) have been conducted at 3 Tesla, whether in individual laboratories or through large-scale efforts like the human connectome project (HCP) (Auria et al., 2020; Sotropoulos et al., 2013; Tounek et al., 2018). Such studies can be more than adequate for some applications of dMRI, such as the assessment of between-population differences in white matter tissue properties. However, structural connectivity estimates derived from such dMRI data through tractography are often inaccurate, producing not only false connections, but also failing to reproduce known ones (Azadakbal et al., 2015; Bullock et al., 2022; Donahue et al., 2016; Grier et al., 2020; Maier-Hein et al., 2019; Reveley et al., 2015b; Schilling et al., 2019; Shen et al., 2019; Thomas et al., 2014). One contributing factor to these inaccuracies is the large dMRI voxel size relative to an axon or axonal fasciculi. An average axon is <1 μm in di-

* Corresponding author: Sarah R. Heilbrunner, 0.2-164 Jackson Hall, 321 Church St SE, Minneapolis, MN, 55455.
E-mail address: heilbr02@umn.edu (S.R. Heilbrunner).
# These authors contributed equally

https://doi.org/10.1016/j.neuroimage.2022.119200.
Received 12 January 2022; Received in revised form 8 March 2022; Accepted 7 April 2022
Available online 13 April 2022.
1053-8119/© 2022 The Author(s). Published by Elsevier Inc. This is an open access article under the CC BY-NC-ND license
(http://creativecommons.org/licenses/by-nc-nd/4.0/)
ameter (Liewald et al., 2014), whereas a typical isotropic dMRI voxel is ≥1.5 mm. This means a single voxel can contain millions of axons with various orientations or exhibit partial volume effects that can confound tractography estimates (Alexander et al., 2001). One potential solution to this dilemma is improved spatial resolution, which requires higher signal-to-noise ratio (SNR). Substantial SNR gains can be achieved by using higher magnetic fields (Pohmann et al., 2016; Tavaf et al., 2021; Ugurbil, 2014, 2014; Vaughan et al., 2001), despite the accompanying decrease in T₂, which is disadvantageous for diffusion encoding with respect to SNR. Contemporary techniques, such as using single echoes (rather than two) enabled by removal of eddy current effects through image processing (Andersson and Sotiropoulos, 2016; Sotiropoulos et al., 2013), acquiring highly undersampled data along the phase encoding direction using parallel imaging and partial Fourier (Vu et al., 2015), and using high performance gradients (Setsomop et al., 2013; Sotiropoulos et al., 2013) make it possible to achieve sufficiently short echo-times (TEs) for diffusion encoding to exploit the intrinsic SNR gains of high magnetic fields (Ugurbil et al., 2013). SNR can be further improved by adding a denoising step when preprocessing dMRI data (Moeller et al., 2021; Veraart et al., 2016b; Veraart et al., 2016c). Additional challenges of increased B₀ and B₁ inhomogeneity and increased tissue energy absorption at the higher magnetic fields are also becoming tractable using new techniques, in particular parallel transmit coils and parallel transmit RF pulse design (Ugurbil, 2018) and optimized RF pulses (Wu et al., 2018). As a result, 7 Tesla is increasingly prevalent in dMRI applications in humans (Guiban et al., 2018; Kida et al., 2016; Lenglet et al., 2012; Moeller et al., 2021; Sotiropoulos et al., 2016; Vu et al., 2015; Wang et al., 2019; Zitella et al., 2015).

Higher field scanners have also successfully been used to conduct dMRI studies, with improved spatial resolution, SNR, and/or contrast-to-noise ratio (CNR) in smaller mammals and ex vivo tissue (Sakai et al., 2020; Schaeffer et al., 2017). Ex vivo dMRI scans of the human brain can be collected on small-bore, very high field scanners typically reserved for animal studies. On such systems, voxel dimensions of a few hundred microns can be achieved, although they suffer from reduced CNR because of the properties of fixed tissue, often requiring scan times of hours to days (Bastiani et al., 2016; McNab et al., 2009; Roebroeck et al., 2019; Seebaus et al., 2015). In vivo dMRI scans of small animals, such as mice, have achieved similarly impressive resolutions (Alomair et al., 2015; Yon et al., 2020). Indeed, higher spatial resolution (0.5 mm vs 1 mm isotropic) achieved by limiting the diffusion encoding to low B values improves estimates of smaller fibers and reduces partial volume effects, which has been shown to improve tractography detection of U-fibers (Tounek et al., 2018). Increased resolution could also improve gray matter applications of dMRI, such as distinguishing cortical regions and defining microstructure (Ganepola et al., 2018; Truong et al., 2014). However, tractography based on extremely high resolution dMRI data can still generate erroneous pathways, especially if the loss of SNR that accompanies higher resolutions pushes the data to a noise dominated, low SNR regime; for example, ex vivo diffusion MRI data acquired at 0.13 mm isotropic resolution produced more erroneous streamlines in multiple fiber pathways than data acquired at lower resolutions (Calabrese et al., 2014). Clearly, the optimal protocol for collecting high resolution data will require a balance between total scan time, angular resolution and B-values. Data obtained with both high and low resolutions can be combined to exploit advantages inherent in the two different data sets (Sotiropoulos et al., 2016).

The challenges and benefits of higher field strengths change according to the species studied, but the macaque is a frequently used neurobiological model for the human brain because of its size and the human-macaque cortical homologies not present in rodents (Hitchson and Everling, 2012; Laubach et al., 2022; Mars et al., 2016; Petrides and Pandya, 2002; Preuss, 1995; Watson and Platt, 2012). Furthermore, because of these similarities, the macaque brain is also used to study a wide array of mechanisms that model brain disorders and treatments (Heilbronner and Chafee, 2019; Johnson et al., 2021; Johnson and Slopesma, 2017; Nelson and Winslow, 2009; Vitek and Johnson, 2019; Worlein, 2014; Xiao et al., 2016). Unlike humans, macaques can undergo invasive procedures, such as tract-tracing, optogenetics, chemogenetics, and single-unit recordings (Cushnie et al., 2020; De et al., 2020; Galvan et al., 2017; Raper et al., 2019; Upright et al., 2018; Watanabe et al., 2020). There is a wealth of anatomical tract-tracing data available for the macaque, including CoCoMac and Core-Nets (Bakker et al., 2012; Markov et al., 2014; Stephan et al., 2001), which have been used to validate and refine diffusion tractography in macaques (Azadabkht et al., 2015; Donahue et al., 2016; van den Heuvel et al., 2015). The combination of strong homology to the human brain and the ability to employ invasive procedures to modulate neural activity and determine anatomical connectivity means that high-resolution in vivo dMRI of the macaque brain would be invaluable.

While very high resolution dMRI has been acquired with ex vivo scans in both humans and macaques (Azadabkht et al., 2015; Calabrese et al., 2015; Fritz et al., 2019; Sébille et al., 2019), there are significant benefits to acquiring high resolution scans in vivo. For example, individualized surgical planning for deep brain stimulation (DBS) in humans and macaques can benefit from in vivo dMRI for targeting (Davidson et al., 2020; Hunsche et al., 2013; Patriat et al., 2018; Planinga et al., 2018; Johnson et al., 2016). Similarly, preoperative dMRI scans can guide intracranial electrode implantation in macaques (Saalmann et al., 2012). Other intraoperative surgeries in humans and/or macaques could benefit from accurate dMRI data on white matter organization and connectivity, including the current push for adeno-associated virus-mediated gene therapies (Haday and Vandenberge, 2019; Weiss et al., 2022). The use of in vivo dMRI also provides the opportunity to perform longitudinal studies that monitor normal development, the progress of various disease states or as a biomarker for neurodegeneration (Andica et al., 2020; Baldarano et al., 2017; de Groot et al., 2016; Fields, 2008; Huber et al., 2018; Krosgrud et al., 2016; Lövdén et al., 2010; Mackey et al., 2012). These approaches are only possible with the use of in vivo dMRI acquisitions, which could be improved with higher resolution scans.

Here, we detail our efforts to acquire high-quality dMRI at 10.5 Tesla in anesthetized rhesus and long-tailed macaques. Macaques are too large for ultrahigh field, small-bore preclinical scanners often used for in vivo studies of smaller model organisms. The 10.5T/88 cm bore magnet utilized in the present study currently represents the highest field, large bore scanner in the world capable of scanning macaques (Yacoub et al., 2022) and humans (He et al., 2020). Our approach produces, to our knowledge, the highest spatial and angular resolution in vivo macaque dMRI data (0.58 mm isotropic voxels) to date and is presented as proof-of-concept for this type of data collection.

2. Materials and methods

2.1. Subjects

We obtained data from 5 female macaque monkeys (4 Macaca fascicularis and 1 Macaca mulatta). Weights ranged from 3.2 kg to 7.8 kgs. Experimental procedures were carried out in accordance with the University of Minnesota Institutional Animal Care and Use Committee and the National Institute of Health standards for the care and use of nonhuman primates. All subjects were pair-housed in a light and temperature-controlled colony room. Animals were fed and watered ad libitum. None of the animals had any prior implants or cranial surgery performed.

2.2. Data acquisition

All data were acquired on a passively shielded, 10.5 Tesla, 88 cm horizontal bore magnet with 60 cm accessible bore within the gradients and bore liner. This magnet was interfaced with a Siemens (Erlangen, Germany) console and electronics (Magnetom 10.5T Plus) and Siemens...
SC72 gradient sub-system operating with a maximum slew rate of 200 mT/m/s, and maximal gradient strength of 70 mT/m. The 10.5T system operates on the E-line (E12U) platform, comparable to clinical platforms (3T Prisma/Skyra, 7T Terra). As such, the user interface and pulse sequences were identical to those running on clinical platforms. A custom designed RF coil (Supplementary Figure 1a) with an 8-channel transmit/receive dipole array consisting of 18 cm long end-loaded dipoles combined with a close-fitting 16-channel loop receive only array head cap, and an 8-channel loop receive only array of 50 × 100 mm size located under the chin (Lagore et al., 2021; Yacoub et al., 2022) was used in all experiments; thus, the total number of receive channels in this coil is equal to 32. The 16 close fitting head cap receive only loops are divided into 14 individual receive loops of 37 mm combined with 2 larger ear loops of 80 mm. All receiver loops were arranged in an overlapping configuration for nearest neighbor decoupling. The combined 32 receive channels support 3 fold acceleration in the phase encode direction. The coil holder was designed to be semi-stereotaxic, capable of holding the head of the animal in a stable, centered position via custom ear bars. The receive elements were modelled to adhere as close to the surface of the animal’s skulls as possible. Transmit phases were fine-tuned for excitation uniformity for one representative animal and the calculated phases were then used for all subsequent acquisitions. Magnetic field homogenization (B0 shimming) was performed using a customized field of view with the Siemens internal 3D mapping routines. Multiple iterations of the shims (using the cardiac adjusted Fov shim parameters) were performed and manual fine adjustments performed on each animal. High-order shim elements (3rd order) were disregarded for these procedures.

On scanning days, anesthesia was induced by intramuscular injection of atropine (0.5 mg/kg), dexmedetomidine (13 μg/kg) and ketamine hydrochloride (7.5 mg/kg). Animals were transported to the anteroom of the scanner and intubated with an endotracheal tube. Anesthesia was maintained using 1.5–2% isoflurane mixed with oxygen (2 L/h flow rate at 100%). Animals were placed in sphinx position onto a custom-built coil bed with integrated ear bars for pseudo-stereotactic head fixation as described previously (Lagore et al., 2021; Yacoub et al., 2022). Rectal temperature, respiration rate, end-tidal carbon dioxide, electrocardiogram and oxygen saturation were monitored using an MRI compatible monitor (IRADMED 3880 MRI Monitor, USA). Temperature was maintained with a circulating water bath, chemical heating pads and blankets for thermal insulation.

2.3. Anatomical scan

T1w images were acquired on all subjects with a 3D-MPRAGE sequence and T2w images were acquired on 3 subjects with a 3D T2 SPACE sequence. T1w: FOV: 131 × 150 mm2; matrix size of 280 × 320 (0.5 mm isotropic resolution); TR/TE of 3300/3.56 ms. T2w: FOV: 128 × 150 mm2; matrix size of 260 × 320 (0.5 mm isotropic resolution); TR/TE of 3000/100 ms; flip angle of 120°. In order to perform structural image processing for subjects without a T2w structural image, the bias field was corrected with the N4biascorrection tool as implemented in ANTS (Tustison et al., 2010). For subjects with a T2w structural image (not always acquired due to early termination of the scan because of animal monitoring issues), a T1w/T2w ratio image was generated by skull stripping both the T1w and T2w images, registering the T2w to the T1w space and then dividing the T1w by the T2w image.

2.4. dMRI acquisition and preprocessing

Pilot studies (not presented here) were performed in order to optimize the custom coil for the 10.5T magnet (Lagore et al., 2021). Testing was performed to achieve the most stable readout direction which resulted in pure z direction. This meant that the smallest field of view for whole brain coverage could be obtained in a sphinx position. A stable and reliable positioning of the animal was of utmost importance for reproducible and reliable data quality. Sequence optimization was performed within the limits set by our gradient and SAR monitoring. In short, we aimed at achieving the highest possible resolution resolvable using a single shot that allowed for coverage of the whole macaque brain and remained manageable with post-hoc distortion correction. Our coil geometry allows roughly 3x in-plane acceleration (GRAPPA in our case) for the small volume needed, limiting the overall matrix size we can achieve. Segmented encoding approaches will likely yield higher resolutions, at the cost of q-space sampling efficiency, but require extreme stability of the animal.

The dMRI sequence utilized was a 2D single shot EPI acquired with reverse phase encoding head/foot (H/F), foot/head (F/H) consistent with the requirements for FSL Topup/Eddy offline EPI distortion correction (Andersson et al., 2003). B1+ field optimization was performed using a vendor provided flip angle mapping sequence and the power was calibrated for each subject. Precalibrated fixed phases were used in all subjects. Each scan included 3 or 4 pairs (H/F and F/H) of 114 diffusion-weighted volumes (TR/TE of 8270/78.4 ms; matrix size, 260 × 144 × 75; 6/8 phase partial Fourier, 0.58 mm isotropic resolution), contingent upon maintenance of acceptable vital signs. An iPAT acceleration of 3 was used without any slice acceleration. Each multi-shell acquisition contained 7 b = 0 s/mm2 vol (one every 16 vol), 54 b = 1000s/mm2 vol and 53 b = 2000s/mm2 images sampled on the whole sphere. Total scan time for the diffusion acquisition was approximately 2 h and 16 min.

Diffusion image preprocessing was performed using a custom pipeline incorporating components from FSL (Jenkinson et al., 2012), MRtrix3 (Tournier et al., 2019), and Freesurfer (Fischl, 2012). Dicom images were converted directly into the MRtrix image format and then back-converted to NIFTI files with the corresponding bvecs/bvals files. The sphinx orientation headers were corrected with Freesurfer’s mri_convert with –sphinx option for further processing with FSL. All acquisitions were visually inspected for gross movement and other artifacts that would prevent further analysis. Data was denoised with MRtrix dwise-noise and Gibbs ringing artifacts were corrected using MRtrix dwise-noise. To increase the SNR for the estimation of the susceptibility induced off-resonance field, an average b0 image for each PE direction was generated and then passed to the TOPUP tool (Andersson et al., 2003) and implemented in FSL (Smith et al., 2004). Eddy current correction was performed using the FSL tool Eddy_Cuda9.1 (Andersen and Sotiropoulos, 2016) using the default settings with the following modifications (nvox=xp=10,000 and –respel (Andersson et al., 2016)). After distortion and eddy current corrections, data was converted back to the MRtrix image format to estimate diffusion tensors and generate fractional anisotropy (FA), apparent diffusion coefficient (ADC) maps and RGB maps of the principal direction of diffusion using the MRtrix dwi2tensor and tensor2metric tools (Tournier et al., 2019). A visual summary of this pipeline is provided in Supplemental Figure 1B.

SNR was quantified by drawing 5 ROIs on the FA map for each animal–Supplemental Figure 2. One gray matter ROI was drawn in the cortex. Two white matter ROIs were drawn (one in the corpus callosum and one in the anterior limb of the internal capsule). One mixed ROI was drawn on the descending posterior corona radiata. A final ROI was drawn outside of the brain to quantify the standard deviation of the noise. The mean signal intensity was measured in each ROI in data that had not been denoised or had Gibbs ringing corrected. SNR was calculated by dividing the mean signal within each ROI by the standard deviation of the noise in each volume. The calculated SNR values from each volume were averaged into a single value for each shell for each subject. These values are reported in Tables 1-2. The same ROIs were also used to calculate fractional anisotropy (FA), mean diffusivity (MD), axial diffusivity (AD) and radial diffusivity (RD). The mean and standard deviation of these metrics are reported in Table 3-4.
Table 1
Signal-to-Noise Ratios 0.58 mm isotropic resolution diffusion weighted images in select regions prior to denoising and Gibb's ringing removal.

| Region          | SNR (b = 0 s/mm\(^2\)) | SNR (b = 1000s/mm\(^2\)) | SNR (b = 2000s/mm\(^2\)) |
|-----------------|-------------------------|----------------------------|----------------------------|
| Corpus Callosum | 17.01±6.95              | 7.36±2.25                  | 5.58±1.57                  |
| ALIC            | 19.66±4.33              | 8.82±2.13                  | 5.66±1.1                   |
| dpcr            | 13.52±1.76              | 7.23±2.13                  | 5.3 ± 1.43                 |
| gray matter     | 14.7 ± 2.23             | 8.28±1.10                  | 5.56±1.63                  |

2.5. Tractography

Tractography was performed with the MRtrix3 software package (Tournier et al., 2019). First, the response functions were estimated from the preprocessed diffusion images with dwi2response using the “dholdealer” algorithm for multishell data. The response functions were then used to estimate fiber orientation directions (FODs) with a constrained spherical deconvolution (CSD) using dwi2fod with the msmt_csd algorithm that generated three separate FODs maps (white matter, gray matter, and CSF) (Jeurissen et al., 2014; Tournier et al., 2004). FODs were normalized for intensity using minormalize (Raffelt et al., 2017). For fiber tracking, tckgen was used with the iFOD2 model (Tournier et al., 2010). Whole brain tractograms were generated from a white matter mask. Individual tracts were assessed with seeds manually drawn on each subject’s FA map. Seeds were defined as the injection sites for previously described pathways as determined by anatomical tract-tracing to assess the quality of the tractography. No waypoints or exclusion masks were used for these tractograms.

3. Results

3.1. Anatomical images

Anatomical imaging of the whole macaque brain was performed using our custom 32 channel receive (32Rxs) coil. Selected images are shown in Fig. 1. Cortical, subcortical, and white matter structures are all clearly visible. A more extensive discussion of the acquisition procedure can be found here (Lagore et al., 2021).

3.2. dMRI

SNR was relatively homogenous throughout the brain. Average SNR maps for each shell (prior to denoising and Gibbs ringing removal) are shown in Fig. 2 and Supplemental Figures 3–6. SNR values for each region of interest before and after denoising/Gibbs ringing removal are reported in Table 1 and Table 2 respectively. On average, SNR was increased by approximately 75% after denoising and removal of Gibbs ringing.

3.3. Denoising

dMRI has an intrinsically low SNR (relative to other MRI modalities). The SNR is proportional to the voxel volume at a fixed field strength. When increasing the resolution of the diffusion scan, there is a corresponding reduction in SNR. To maintain SNR at high resolution (especially at higher b-values which provide higher diffusion weighting, but at the penalty of increased noise), a denoising algorithm was applied to these data. Prior to any interpolation, the dwidenoise tool in MRtrix based on the MPPCA algorithm (Veraart et al., 2016b, 2016c) was applied to the data. Fig. 3 shows the dMRI data after eddy-current distortions correction (top half) and eddy-current distortion corrected plus dwidenoise denoising (bottom half). Note that the more superficial slices (thus closer to the coil with higher SNR) show less improvement with denoising, though the boundaries of cortical sulci become more defined. Deeper structures show much cleaner boundaries. This is readily apparent in the optic radiation. Some sulci that are not well-defined in the eddy-current corrected only data can be visualized more readily in the denoised dataset.

3.4. Gibbs ringing artifact correction

Gibbs ringing is an imaging artifact observed at boundaries between tissue types. This artifact manifests as parallel lines adjacent to high contrast tissue boundaries in the brain, often observed near the ventricle and cortex/CSF interface. It is important to correct these artifacts as they can significantly alter modeling of the diffusion tensor and kurtosis parameters (Kellner et al., 2016; Perrone et al., 2015; Veraart et al., 2016a). Gibbs ringing is a truncation artifact caused by partial k-space reconstructions and can be corrected with the MRtrix tool mrdegbibs (Kellner et al., 2016). One way to minimize the appearance of Gibbs ringing is to increase the matrix size of the acquisition. However, while this would minimize the ringing artifact, it is not ideal given the already long readout time, which would increase further along with geometric distortions, signal dropouts, and echo time. Fig. 4 shows the removal of Gibbs ringing. This artifact is clearly visible in the b0 images and where the enlarged area demonstrating Gibbs ringing is highlighted with red boxes. The ringing artifact is clearly visible in the upper panel near the ventricles and cortical surface. The lower panel shows the data with the ringing removed by the MRtrix tool mrdegbibs.

3.5. Fractional anisotropy/MD/DTI

Fractional anisotropy (Fig. 5), mean diffusivity (Fig. 6), and color-coded fractional anisotropy (Fig. 7) maps were generated after denoising and Gibbs ringing removal. Scalar metrics calculated before and after denoising/Gibbs ringing removal are summarized in Tables 3 and 4, respectively. Exemplar fiber orientation distribution functions are displayed in the colored boxes in Fig. 7. The high resolution scans allow for differentiation of smaller fiber bundles and structures that are often affected by partial volume effects. These include U-fibers and gray matter/white matter boundaries in the cortex.

Table 3
Scalar Diffusion Metrics prior to denoising and Gibbs’s ringing removal. Data reported as mean ± standard deviation.

| Region          | FA (x10^-3/μm^2/s) | MD (x10^-3/μm^2/s) | RD (x10^-3/μm^2/s) | AD (x10^-3/μm^2/s) |
|-----------------|---------------------|--------------------|--------------------|-------------------|
| Corpus Callosum | 0.691±0.041         | 0.743±0.095        | 0.420±0.058        | 1.379±0.194      |
| ALIC            | 0.637±0.065         | 0.537±0.087        | 0.288±0.025        | 1.034±0.225      |
| dpcr            | 0.548±0.041         | 0.600±0.024        | 0.405±0.040        | 0.992±0.061      |
| gray matter     | 0.128±0.007         | 0.701±0.038        | 0.655±0.065        | 0.791±0.045      |
3.6. Tractography

Fig. 8 shows whole brain streamlines generated from the white matter FODs. To demonstrate the feasibility of tractography at 10.5T with 0.58 mm isotropic resolution, Fig. 9 shows tractography derived from well-described pathways with tract-tracing validation from the literature. The tractography seeds are shown in supplemental figure 7. In Fig. 9A, a seed was drawn in the motor cortex (Case 24 from (Schmahmann and Pandya, 2006)). Fibers can be seen clearly descending in the internal capsule, as well as crossing the corpus callosum, travelling in the superior longitudinal fasciculus and traversing toward the external capsule. Although all of these belong to the correct bundles, we also see some streamlines descending in the contralateral internal capsule, which represents a false positive. In Fig. 9B, a seed was placed in the white matter of the caudal corpus callosum (connectivity previously defined in (Kennedy et al., 1986; Kennedy and Dehay, 1988; Sunderland, 1940)), resulting in streamlines reaching large portions of the parietal and occipital lobes, while avoiding more frontal and ventral cerebral cortical regions. In Fig. 9C, a seed was placed in the precuneus (Case 17 from (Schmahmann and Pandya, 2006)). Here, streamlines can be seen traversing the dorsal cingulum bundle (but not temporal or subgenual components, (Heilbronner and Haber, 2014) and the superior longitudinal fasciculus I. Finally, in Fig. 9D, a seed was placed in the ventrolateral prefrontal cortex (Case 33 from (Schmahmann and Pandya, 2006)). Streamlines can be seed traveling dorsally and medially, around the striatum, to reach the corpus callosum and cross to the contralateral hemisphere.

3.7. U fibers

Sulci present specific issues for tractography; namely, at the fundus, it is difficult to track both U-fibers (local association fibers coursing around the sulcus) and the deep, penetrating fibers that represent longer-range connections (Reveley et al., 2015a). Here, we seeded the fundus of the arcuate sulcus and the intraparietal sulcus. Fig. 10 shows both U-fibers and deep penetrating fibers. Thus, with this acquisition protocol, we are able to visualize the U-fibers described in ex vivo scans (Oishi et al., 2011) and with in vivo 3D-mseEPI acquisitions (Toumaki et al., 2018). Supplemental figures 8 and 9 demonstrate consistency across multiple subjects for the same U-fibers in the arcuate and intraparietal sulci, respectively.

|                  | FA  | MD (s10^−3)mm²/s | RD (s10^−3)mm²/s | AD (s10^−3)mm²/s |
|------------------|-----|-----------------|-----------------|-----------------|
| Corpus Callosum  | 0.691±0.041 | 0.721±0.095 | 0.420±0.062 | 1.325±0.181 |
| ALNC             | 0.630±0.070 | 0.568±0.078 | 0.277±0.018 | 0.997±0.020 |
| dpcc             | 0.521±0.041 | 0.583±0.023 | 0.410±0.023 | 0.928±0.043 |
| gray matter      | 0.118±0.009 | 0.685±0.033 | 0.644±0.031 | 0.766±0.038 |

Fig. 1. Structural images of the macaque brain acquired at 10.5T. Top panel shows T1w images. Middle panel shows T2w images. The bottom panel shows a T1w/T2w ratio image.
Fig. 2. Signal-to-noise ratio maps for 0.58 mm isotropic resolution dMRI data acquired at 10.5T from a single subject, I. The average SNR for each shell is shown. Selected axial slices show a relatively homogenous SNR through the entire brain.

Fig. 3. Effect of MPPCA denoising on diffusion weighted images. The top panel shows 5 axial slices of data processed through Eddy and Topup. The bottom panel shows Eddy corrected with additional denoising.

4. Discussion

Diffusion neuroimaging is a noninvasive imaging modality that provides information about the diffusion of water molecules in tissue in both clinical and nonclinical populations. Of particular relevance to neuroscience is the ability to model white matter in the brain from these properties and to generate estimates of anatomical connectivity from tractography. Such connectivity measures are expected to improve with higher spatial resolutions. However, going beyond the typically supramillimeter resolutions commonly employed today to submillimeter resolution dMRI is challenging due to the reduction in SNR caused by the correspondingly smaller voxel volume. Utilizing an ultra-high field magnet is one method to improve the SNR when compared to the same voxel size acquired at lower fields, assuming the echo time for diffusion...
encoding, TE, is not undesirably long. Macaques, like humans, are too large for other high field, small-bore MRIs that have been used extensively with smaller mammals (Motovylyak et al., 2019; Schaeffer et al., 2017; Schilling et al., 2017; Wu and Zhang, 2016). Here, we present, for the first time, in vivo dMRI of macaque brains imaged on a large bore 10.5 Tesla scanner capable of human imaging. We have obtained multi-shell, high angular resolution, 0.58 mm isotropic resolution data. Utilizing a combination of freely available software packages, we have generated tractography maps that recapitulate known white matter pathways when compared to established anatomical data.

Using a single shot 2D EPI protocol in the anesthetized macaque, we were able to collect up to 4 repeated reverse phase encoded acquisitions (34 min each) of high angular resolution, multi-shell data to increase the SNR. We did not note any motion artifacts or corruption of data resulting from physiological motion of the brain, which can be a problem with 3D-multiecho EPI acquisitions (Liu et al., 2016). Single shot EPI acquisitions are prone to have artifacts (image distortion and signal loss) due to magnetic field inhomogeneities (Andersson et al., 2003); these become more pronounced with increasing static magnetic field used for the study and long echo train lengths required for high resolution scans. Nevertheless, we found the distortions to be well corrected by the FSL Topup/Eddy tools (Supplemental Figure 10).

Macaque imaging comes with unique challenges and oftentimes requires specific tools and significant adaptations for optimal results. One major hurdle is that the macaque head is significantly larger than those of rodents, marmosets, or squirrel monkeys (preventing the use of high field, small bore MRIs with significantly stronger gradients), but are approximately 10 times smaller than the human brain (Donahue et al., 2018). This means that higher spatial resolution is required to resolve similar structures in the macaque brain compared to humans. Along with the aforementioned SNR limitation, another problem is the lack of a commercially available coils specific to macaque imaging, particularly at ultra-high fields. A fully custom 32 channel head coil was developed for the 10.5T scanner, which required extensive optimization (Lagore et al., 2021). This coil has also been used to study resting-state functional MRI in macaques, which provides the opportunity to perform multimodal imaging studies within-subject at 10.5T (Yacoub et al., 2022).

The majority of whole brain submillimeter resolution macaque dMRI studies have been performed ex vivo. The long scan times and lack of breathing-related motion artifacts means that ex vivo scans can achieve stunning results. Ex vivo scans are capable of attaining resolutions as high as 0.13 mm (Calabrese et al., 2014). (Calabrese et al., 2015) used a 7T scanner with a single shell (b = 1500s/mm²), 12 diffusion directions, and a total scan time of 46 h per subject. Ex vivo whole brain coverage at 0.2 mm isotropic resolution has also been achieved with an 11.4T magnet and a diffusion weighted 3D segmented EPI acquisition with 3 shells over 73 h (Sébille et al., 2019). Another ex vivo study conducted at 4.7T collected dMRI data on 2 brains, one imaged at
0.8 mm isotropic resolution with 4 averages of 61 directions each over 64 h and a second imaged at 0.43 mm isotropic resolution with 3D-EPI with 120 directions over 27 h (Azadbakht et al., 2015). Resolutions as high as 0.4 mm isotropic have been reported for ex vivo human whole brain coverage with a 53 hour acquisition (Fritz et al., 2019). Calabrese et al. also conducted an ex vivo study holding total scan time constant at 60 h, while varying the number of b-values, spatial and angular resolution (Calabrese et al., 2014). This study generated datasets acquired between 0.13 and 0.6 mm isotropic resolution. While these studies have provided valuable information regarding the limits of dMRI, complementary protocols that take advantage of the non-invasive nature of the technique are needed. Additionally, fixation alters the fundamental diffusion properties of brain tissues (Shatil et al., 2018), further limiting the translation of these techniques.  

In vivo macaque dMRI with 1–1.5 mm isotropic resolution has been widely reported and easily attainable with most 3T scanners and commercially available coils (Feng et al., 2017; Liu et al., 2016; Meng et al., 2018; Rane and Duong, 2011; Shen et al., 2019; Shimony et al., 2018; Weiss et al., 2022). 1.1 mm isotropic dMRI data has been acquired with a knee coil (Shimony et al., 2018). Utilizing a standard head coil, 1.5 mm isotropic resolution diffusion MRI was collected on ten macaques in vivo; this dataset was further refined by registering the high resolution ex vivo dMRI to the in vivo diffusion space to generate a population average of white matter tracks (Feng et al., 2017). Liu et al. also acquired
1 mm isotropic dMRI with a custom built 3 channel coil (Liu et al., 2016). Another study performed with a pediatric head coil acquired 1 mm isotropic dMRI data as well as 0.5 mm isotropic T1 and T2 images to generate a population average white matter and gray matter maps (Weiss et al., 2022). However, there has been only a small number of in vivo dMRI studies with submillimeter isotropic resolution in anesthetized macaques, mostly at 3T. Autio et al. (2020), presented 0.9 mm dMRI data acquired on a 3T scanner with 2D-single shot EPI with simultaneous multi-slice accelerations (Autio et al., 2020). This study used the same b-value scheme ($b = 1000, 2000$ and $3000$ s/mm$^2$) as in the HCP but utilized a much higher number of diffusion directions (500 vs 270 in the HCP) to improve the data quality in a 30 min acquisition. Another study acquired dMRI at 0.7 mm isotropic resolution on a 3T scanner, with a custom made, subject specific implanted 8-channel coil array using a single shell ($b = 1000$s/mm$^2$) 256 direction acquisition (Janssens et al., 2012). The coil array was attached to the animal’s skull with ceramic screws and encased in dental acrylic during a standard headpost surgery. While the implanted coil produces very high-quality data with 3 h of acquisition, the invasive nature of the coil is a significant limitation for a non-invasive imaging modality. Finally, another study reported 0.5 mm isotropic resolution (Tounekti et al., 2018) utilizing 3D multi-shot echo planar imaging on a 3T clinical scanner with 30 diffusion directions applied with a $b = 1000$s/mm$^2$ and a total acquisition time of 130 min. While 3D-msEPI is capable of increasing resolution, it is also highly susceptible to physiological motion artifacts. In Tounekti et al. (2018), approximately 8% of diffusion weighted images were discarded due to artifacts. At 7T, Wang et al. present the optimization of readout segmented echo planar imaging (rsEPI) for whole macaque brain coverage. A single 30 direction, $b = 1000$s/mm$^2$ diffusion scan was collected at 0.8 mm isotropic resolution in a single phase encode direction over 45 min (Wang et al., 2019). As noted above, with the exception of Autio et al. (2020), all of these studies used single $b$ value of 1000$s$/mm$^2$ for diffusion weighting. Higher $b$ values lead to SNR loss not only because of the higher diffusion weighting but also the longer TE that is required to acquire the data. Since the TE for the entire experiment is set by the TE for the highest $b$ value employed, in a multi-shell experiment, the lower $b$ value shell also suffers SNR losses. Avoiding higher $b$ values in a multi-shell experiment, on the other hand, negatively impacts the quality of tractography estimates significantly (Sotiropoulos et al., 2013)

The brief summary of submillimeter resolution macaque dMRI given above highlights the challenges of high resolution in vivo dMRI protocols

Fig. 8. Representative whole brain tractograms from a single subject. Selected axial slices showing whole brain streamlines.

Fig. 9. Tractography in dMRI data collected at 10.5T. Tractography from a single subject with seeds in the motor cortex (A), caudal corpus callosum (B), precuneus (C), and ventrolateral prefrontal cortex (D). Streamlines are overlaid on a T1 image registered to the diffusion space.
that often require tradeoffs among acquisition time, acceleration factors, angular resolution, maximum b values employed, and potentially invasive procedures. The work presented in our study is non-invasive, acquired with both high spatial and angular resolution using two shells ($b = 1000$ and $2000$) and a $2D$ spin echo sequence in approximately two hours. The preprocessed data produces high quality scalar metric maps and tractography streamlines.

Tractography allows for the visualization of white matter pathways and estimation of connectivity strength between brain regions in vivo and non-invasively (Lenglet et al., 2012; Soares et al., 2013). dMRI is currently the only technique available that can estimate structural connectivity in human brains in vivo (dissection and polarized imaging techniques allow for ex vivo human structural connectivity calculations, but this limits our ability to assess population differences and brain-behavior relationships). Thus, optimization and validation of the technique is of the utmost importance. Future studies will compare the results of our high resolution tractography to anatomical databases, to within-subject anatomical data generated from these animals, and to lower field dMRI results. While tractography has improved dramatically with the advent of newer methodology, e.g. CSD (Tournier et al., 2007), NODDI (Zhang et al., 2012), etc., it has historically had the significant limitation that it is only moderately correlated with true anatomical connectivity (as estimated with tract-tracing) (Donahue et al., 2016; Grier et al., 2020; Shen et al., 2019; Thomas et al., 2014; van den Heuvel et al., 2015). As discussed in the Introduction section, spatial resolution is one of the primary factors in resolving crossing fibers (Donahue et al., 2016). The inherently low SNR of diffusion MRI images, especially when high b values are employed, makes increasing the resolution challenging because as spatial resolution improves, the signal reduces even further (Jones, 2010). In this study, we utilized both SNR gains available at $10.5T$ magnetic field and denoising using the MPPCA approach as implemented in MRtrix (Veraart et al., 2016b, 2016c). Future diffusion data acquired on the $10.5T$ scanner will be acquired consistent with the requirements of the recently published Noise reduction with Distribution Corrected (NORDIC) PCA method, and we expect to see further improvements in data quality (Moeller et al., 2021). The collection of high-quality dMRI data, as with all MRI modalities, involves a complex tradeoff of many imaging parameters. Especially important are the field strength of the magnet and the gradient slew rate and amplitude. Within the possible parameter space available with current instrumentation, we opted for targets that minimized distortions as much as possible and provided sufficient SNR for reliable tract reconstruction over the entire brain. Additional significant gains will be achieved using a gradient set that can operate with substantially higher maximal gradient strength and high slew rates at $10.5T$, enabling more extensive q-space sampling and additional optimization of scan parameters such as resolution, scan time, and q-space coverage. Such an optimization is expected to lead to better data quality. Ultimately, however, the validation of our data will depend on comparison with ground truth anatomical data.

Much can be learned from diffusion scans on anesthetized macaques that will improve or inform the methods when applied to the study of human brains, especially when combined with neural tract tracing experiments within the same subjects that will allow direct comparison of diffusion tractography data to known anatomical connectivity. This
highlights one important advantage of using nonhuman animal models: the ability to compare to ground-truth neural tract-tracing data to assess the accuracy of dMRI (Azadabkht et al., 2015; Girard et al., 2020; Shen et al., 2019; van den Heuvel et al., 2015).

5. Conclusions

Using a 10.5T large bore MRI, a custom 32 channel coil, and a denoising approach, this study presents high spatial resolution (0.58 mm isotropic voxels), high angular resolution (115 directions), multi-shell diffusion MRI data in anesthetized macaques achieved with 2D-single shot EPI protocols in approximately 2 h. Our study demonstrates the feasibility of collecting these data at ultra-high field. Despite the challenges of the intrinsically low SNR of diffusion MRI, this acquisition protocol at 10.5T combined with denoising produces tractography results in agreement with known pathways and is capable of differentiating short association fibers in the cortex that are often lost to partial volume effects. Further optimization of acquisition protocols with the macaque model as well as significantly improved hardware, such as higher channel count receivers (48+) and a high performance head gradient insert employed at 10.5T, and improved denoising techniques will further push the dMRI methods to truly explore the ultimate limits of in vivo 10.5T diffusion MRI.

Ethics statement

Animal procedures were described by the University of Minnesota Institutional Animal Care and Use Committee and adhered to the National Institute of Health standards for the care and use of nonhuman primates. Code will be deposited on github upon manuscript acceptance. Data will be made available through the PRIMeATe Data Exchange (PRIME-DE).

Funding statement

This work was supported by NIH grants R01MH118257 (SRH), P50MH119569 (SRH), P41EB027061 (CL, KU), U01EB025144 (KU), T32DA007234 to P. Mermelstein (MDG), P30DA048742 (JZ, SRH), R56EB031765 (JZ), P50NS098753 (NH), a UMN AIRP award (JZ, SRH), MnDrive Brain Conditions (SRH).

Credit authorship contribution statement

Mark D. Grier: Conceptualization, Methodology, Writing – original draft, Data curation, Software, Investigation. Essa Yacoub: Conceptualization, Methodology, Writing – review & editing, Investigation. Gregor Adriany: Methodology, Writing – review & editing, Investigation. Russell L. Lagore: Methodology, Writing – review & editing, Investigation. Noam Harel: Conceptualization, Methodology, Writing – review & editing, Investigation. Ru-Yuan Zhang: Methodology, Writing – review & editing, Software. Christophe Lenglet: Conceptualization, Methodology, Writing – review & editing. Kamil Uğurbil: Conceptualization, Methodology, Writing – review & editing, Funding acquisition. Jan Zimmermann: Conceptualization, Methodology, Writing – original draft, Funding acquisition, Data curation, Software, Investigation. Sarah R. Heilbronn: Conceptualization, Methodology, Writing – original draft, Funding acquisition, Investigation.

Acknowledgements

We thank Steve Jungst for continuing support with our coils and hardware setup. We thank Hannah Lee, Jen Holmberg, Adriana Cusnie, Tanya Casta, and Megan Monko for support with animal care and data acquisition. We thank Research Animal Resources at UMN, especially Whitney McGee and Anne Merley, for helping us implement new and improved anesthesia protocols.

Supplementary materials

Supplementary material associated with this article can be found, in the online version, at doi:10.1016/j.neuroimage.2022.119200.

References

Alexander, A.L., Hasan, K.M., Lazar, M., Tsuruta, J.S., Parker, D.L., 2001. Analysis of partial volume effects in diffusion-tensor MRI. Magn. Reson. Med. 45, 1105–1110.
Alomair, O.I., Breiten, L.M., Smith, M.T., Galloway, G.J., Kurniawan, N.D., 2015. In vivo high angular resolution diffusion-weighted imaging of mouse brain at 16.4 Tesla. PLOS ONE. doi:10.1371/journal.pone.0143133.
Anderson, J.L.R., Graham, M.S., Zeodis, E., Sotiropoulos, S.N., 2016. Incorporating outlier detection and replacement into a non-parametric framework for movement and distortion correction of diffusion MR images. Neuroimage 141, 556–572.
Anderson, J.L.R., Skare, S., Ashburner, J., 2003. How to correct susceptibility distortions in spin-echo echo-planar images: application to diffusion tensor imaging. Neuroimage 20, 870–888.
Anderson, J.L.R., Sotiropoulos, S.N., 2016. An integrated approach to correction for off-resonance effects and subject movement in diffusion MR imaging. Neuroimage 125, 1063–1079.
Andica, C., Kamagata, K., Hatanou, T., Saio, Y., Ogaki, K., Hattori, N., Aoki, S., 2020. MR biomarkers of degenerative brain disorders derived from diffusion imaging. J. Magn. Reson. Imaging 52, 1620–1636.
Austin, J.A., Glanser, M.F., One, T., Donahue, C.J., Bastiani, M., Ohno, M., Kawabata, Y., Urushihara, Y., Murata, K., Nishigori, K., Yamaguchi, M., Hori, Y., Yoshida, A., Go, Y., Coakon, T.S., Jhabsi, S., Sotiropoulos, S.N., Kennedy, H., Smith, S., Van Essen, D.C., Hayashi, T., 2020. Towards HCP-Style macaque connectomes: 24 Channel 3T multi-array coil, MR1 sequences can be used. Neuroimage 215, 116800.
Azadabkht, H., Parkes, L.M., Haroon, H.A., Augath, M., Logothetis, N.K., de Crespigny, A., D’Arceuil, H.E., Parker, G.J.M., 2015. Validation of High-Resolution Tractography Against In Vivo Tracing in the Macaque Visual Cortex. Cereb. Cortex 25, 4289–4309.
Bakker, R., Wachter, T., Diemann, M., 2012. CoCoMac 2.0 and the future of tract-tracing databases. Front. Neuroinform. 6, 30.
Baldarova, D., Khenomen, A., Kobor, I., Bogdahn, U., Gorges, M., Kansubek, J., Müller, H.-P., 2017. Longitudinal diffusion tensor imaging-based assessment of tract alterations: an application to amyotrophic lateral sclerosis. front. Hum. Neurosci. 11, 567.
Bastiani, M., Oros-Peusquers, A.-M., Seehaus, A., Brenner, D., Müllerhoff, K., Ce-Sik, A., Felder, J., Bratske, H., Shah, N.J., Galunke, R., Goebel, R., Roebroek, A., 2016. Automatic segmentation of human cortical layer-plexuses and architectural areas using Ex vivo diffusion MRI and its validation. Frontiers Neurosci. doi:10.3389/fnins.2016.00487.
Behrens, T.E.J., Woolrich, M.W., Jenkinson, M., Johansson-Berg, H., Nunes, R.G., Clare, S., Matthews, P.M., Brady, J.M., Smith, S.M., 2003. Characterization and propagation of uncertainty in diffusion-weighted MR imaging. Magn. Reson. Med. 50, 1077–1088.
Bihan, D.L., Le Bihan, D., Mangin, J.-F., Poupon, C., Clark, C.A., Pappata, S., Molko, N., Chabrier, H., 2001. Diffusion tensor imaging: concepts and applications. J. Magn. Reson. Imaging 13, 1002–1016.
Bullock, D.N., Hayden, E.A., Grier, M.D., Tang, W., Pestilli, F., Heilbronn, S.R., 2022. A taxonomy of the brain’s white matter: Twenty-one major tracts for the 21st century. Cerebral Cortex doi:10.1093/cercor/bhab500, bhab500.
Calabrese, E., Badea, A., Coe, C.L., Lubach, G.R., Stynor, M.A., Johnson, G.A., 2014. Investigating the tradesoffs between spatial resolution and diffusion sampling for brain mapping with diffusion tractography: Time well spent? Hum. Brain Mapp 35. doi:10.1002/hbm.22579.
Calabrese, E., Badea, A., Coe, C.L., Lubach, G.R., Shi, Y., Stynor, M.A., Johnson, G.A., 2015. A diffusion tensor MRI atlas of the postmortem rhesus macaque brain. Neuroimage 117, 408–416.
Cusnie, A.K., El Nahal, H.G., Bohlen, M.O., May, P.J., Basso, M.A., Grimaldi, P., Wang, M.Z., de Velasco Ezequiel, M.F., Sommer, M.A., Heilbronn, S.R., 2020. Using aAV2V2-etro in rhesus macaque: promise and caveats for circuit manipulation. J. Neurosci. Methods 345, 108859.
Davidson, B., Lipsman, N., Meng, Y., Robin, J.S., Giacobbe, P., Hamani, C., 2020. The use of tractography-based targeting in deep brain stimulation for psychiatric indications. Front. Hum. Neurosci. 14, 584425.
De, A., El-Shamayleh, Y., Horvitz, G.D., 2020. Fast and reversible neaural inactivation in macaque cortex by optogenetic stimulation of GABAergic neurons. Elife 9. doi:10.7554/eLife.25658.
De Groot, M., Cremers, L.G.M., Ikrar, M.A., Hofman, A., Krestin, G.P., van der Lugt, A., Niessen, W.J., Vansoe, M.W., 2016. White matter degeneration with aging: longitudinal diffusion MR imaging analysis. Radiology 279, 532–541.
Donahue, C.J., Glasser, M.F., Preus, T.M., Rilling, J.K., Van Essen, D.C., 2018. Quantitative assessment of prefrontal cortex in humans relative to nonhuman primates. Proc. Natl. Acad. Sci. U. S. A. 115, E1813–E1819.
Donahue, C.J., Sotiropoulos, S.N., Jhabsi, S., Hernandez-Fernandez, M., Behrens, T.E., Dryby, T.B., Coakon, T., Kennedy, H., Knoblauch, K., Van Essen, D.C., Glasser, M.F., 2016. Using diffusion tractography to predict cortical connection strength and distance: a quantitative comparison with tracers in the monkey. J. Neurosci. 36, 6758–6770.
Fields, R.D., 2008. White matter in learning, cognition and psychiatric disorders. Trends Neurosci 31, 361–370. doi:10.1016/j.tins.2008.04.001.
loop receive array for non-human primate head imaging at 10.5 T. NMR Biomed. doi:10.1002/nbm.4472.
Le Bihan, D., Breton, E., Lallemand, D., Grenier, P., Cabanas, E., Laval-Jeantet, M., 1986. MR imaging of intravoxel incoherent motions: application to diffusion and perfusion in neurologic disorders. Radiology 161, 401–407.
Lengel, R., Talosio, A., Yacoub, M., Ghi, S., Harel, N., 2012. Comprehensive in vivo mapping of the human basal ganglia and thalamic connectives in individuals using 77 MRI. PLoS One 7. doi:10.1371/journal.pone.0029153.
Liewald, D., Miller, R., Logothetis, N., Wagner, H.-J., Schüz, A., 2014. Distribution of axon diameters in cortical white matter: a neuroanatomical micro-CT study on three human brains and a macaque. Biol. Cybern. 108, 541–557.
Li, W., Zhao, X., Ma, Y., Tang, X., Gao, J.-H., 2016. DWI using navigated interleaved multishot diffusion with realigned GRAPPA reconstruction. Magn. Reson. Med. 75, 280–286.
Liu, H., Zhang, L., Lin, X., Xia, Y., Gaufaumar, N., Lin, Z., Wu, Y., He, H., Tempelmann, C., Heine, H.-J., Düzel, E., Schmiedel, F., Lindenberger, U., 2010. Experience-depend- dent plasticity of white matter microstructure extends into old age. Neurology 48, 3870–3883.
Mackey, A.P., Whitsel, K.J., Bunge, S.A., 2012. Experience-dependent plasticity in white matter microstructure: reasoning training alters structural connectivity. Front. Hum. Neurosci. 6, 32.
Maier-Hein, K.H., Neuber, P.P., Houde, J.-C., Gise, M.-A., Garyfallidis, E., Zhong, J., Chamberland, M., Yeh, F.C., Lin, Y.C., J.I., Q., Reddick, W.E., Glass, J.O., Chen, D.Q., Feng, Y., Gao, C., Wu, Y., Ma, J., He, R., Li, Q., Weston, C.-F., Deslauriers-Gaut- thier, S., González, J.O., Paquette, M.-S., Jean, S., Giraud, E., Raveaud, F., Sidhu, J., Tax, C.M.W., Guo, F., Mesri, H.Y., David, S., Froeling, M., Heemskerk, A.M., Lee-, man, J., Zarate, A., Bodi, C., Demers, M., Brabant, S., Doyon, J., Sarica, A., Vasta, R., Cerasa, A., Quattrone, A., Yeatman, J., Khan, A.R., Hodges, W., Alexander, S., Romascano, D., Barakovic, M., Auria, A., Esteban, O., Lemkad- dem, A., Thiran, J.-P., Cetingul, H.E., Odry, B.L., Maille, B., Nadar, N.S., Maza, -galli, R., Dehay, C., Clarivogner, S., Giraud, P., Sappey-Marinier, D., Dufay, C., Corcos, D., Knobloch, K., Kan Esem, D.C., C., Lebed-Hendy, C., 2014. A white matter di- rected interconnectivity matrix model for human brain cortex. Cereb. Cortex 24, 36–39.
Mars, R.B., Verhaagen, L., Gladwin, T.E., Neubert, F.X., Ballester, J.R., Rushworth, M.F.S., 2016. Comparing brains by matching connectivity profiles. Neurosci. Biobehav. Rev. 60, 90. doi:10.1016/j.neubiorev.2015.01.008.
McNab, N., Jamb, A., Jhabal, Sai, Deoni, Sean C., Dowsett, Gwenaëlle, Behrens, Timo- thy E.J., Miller, Karla L., 2009. High resolution diffusion weighted imaging in fixed human brain using diffusion-weighted steady state free precision. Neuroimage 46 (3), 775–785.
Meng, Y., Hu, X., Zhang, B., Bachevalier, J., 2012. Diffusion tensor imaging reveals microstructural alterations in brain white matter in adults with neonatal hippocampal lesions. Hippocampus 22, 838–845.
Moeller, Steen, Pishaard, Pramod Kumar, Ramanna, Sudhir, Lengeth, Christophe, Wu, Xi- coping, Doowle, Logan, Yacoub, Esa, Ugurbil, Kamil, Austry, S., 2021. Cortical Tractography using diffusion weighted MRI in monkey brain at 11.7 Tesla. Neuroimage 86, 116085.
Mothury, A., Skinner, N.P., Schmit, B.D., Wilkins, B., Kurpad, S.N., Buddle, M., 2019. Longitudinal diffusion in vivo magnetic resonance imaging remote from the lesion site in rat spinal cord injury. J. Neurotrauma doi:10.1089/neu.2018.6253.
Nicolaou, S.K., Donnelly, A.M., Oxford University Press.
Kamadaya, K., Andica, C., Hatanan, O., Tawada, G., Takahashi-Amano, H., Ogaki, K., Akashi, T., Hagiwara, A., Fujita, S., Aoki, S., 2020. Advanced diffusion magnetic reso- nance imaging in patients with Alzheimer's and Parkinson's disease. Neuro. Regener. 15, 1590–1600.
Kellner, E., Dhital, B., Kistelev, V.G., Reisert, M., 2016. Gibbs-ringing artifact removal based on local subvoxel-shifts. Magn. Reson. Med. 76, 1574–1581.
Kennedy, H., Dehay, C., 1988. Functional implications of the anatomical organization of the cellular projection to visual areas V1 and V2 in the macaque monkey. Behav. Brain. Res. 29, 225–236.
Kennedy, H., Dehay, C., 1986. Organization of the cellular connec- tions of visual areas V1 and V2 in the macaque monkey. J. Comp. Neurol. 402, 470–409.
Kida, I., Ieguchi, T., Matsuka, Y., Zhou, K., Steemer, A., Porter, D., 2016. Compani- son of diffusion-weighted imaging in the human brain using readout-sensed EPI and PROPELLER turbo spin echo with single-shot EPI at 7 T MRI. Invest. Radiol. doi:10.1097/rid.0000000000001284.
Krogsgaard, S.K., Fjell, A.M., Tamnes, C.K., Grydeland, H., Mork, L., Due-Tonnnessen, P., Bjørnerud, A., Sampo-Baptista, C., Anderson, J., Johansen-Berg, H., Wallhov, K.B., 2016. Diffusion MRI in non-human primate brain-a longitudinal diffusion tensor imaging study of children from 4 to 11 years of age. Neuroimage 124, 473–486.
Kubicki, M., McCarley, R., Westin, C.-F., Park, H.-J., Maier, S., Sikinis, R., Rodes, F.A., Shenton, M.E., 2007. Prevalence of diffusion tensor imaging studies in schizophrenia. J. Psychiatr. Res. 41, 51–50.
Lagore, R.L., Moeller, S., Zimmermann, J., DelaBarre, L., Radder, J., Grant, A., Ugur- bil, K., Yacoub, E., Harel, N., Adriany, G., 2021. An 8-bit transpose and 24-
Sakai T., Hata J., Shinatku Y., Ohita H., Sogabe K., Mori S., Okano H.J., Hamada Y., Hirabayashi T., Minaminoto T., Sadato N., Okano H., Otishi K. The monkey monkey makes brain imaging primary repository image of the postmortem postmortem magnet resonance imaging: the second phase of the archive of digital records. bioRxiv 2020.08.23.265317; doi: 10.1101/2020.08.23.265317.

Schilling, K., Gao, Y., Stepnevskis, I., Choe, A.S., Landman, B.A., Anderson, A.W., 2017. Reproductibility and variation of diffusion measures in the squirrel monkey brain, in vivo and ex vivo. Magn. Reson. Imaging 35, 29–36. doi: 10.1016/j.mri.2016.07.011.

Schilling, K., Nath, N., Pfeuffer, C., Parvathaneni, P., Blaber, J., Gao, Y., Neher, P., Aydogan, D.B., Shi, Y., Ocampo-Pineda, M., Sciavi, S., Daducci, A., Girard, G., Barakovic, M., Rafael-Palomino, J., Komosniewski, D., Pizzolato, M., Batey, A., Fischl, B., Thirion, J.P., Canales-Rodriguez, E.J., Hua, C., Zhu, H., Zhong, L., Ceballos, B.A., Rohnert, F., Theieu, G., Houde, J.-C., Sidhu, J., Chamberland, M., Westin, C.-F., Dyrbj, T.B., Verma, R., Rathi, Y., Ifragnoulou, M.O., Thomas, C., Pierpaoli, C., Descoteaux, M., Anderson, A.W., Landman, B.A., 2019. Limit to anatomical accuracy of diffusion tractography using modern approaches. Neuroimage 208, 1–13. doi: 10.1016/j.neuroimage.2019.01.064.

Schumann, J.D., Pantazis, N.S., 2006. Fiber Pathways of the Brain doi: 10.1093/9780195104223.001.0001.

Sibon, P., Rolland, M., Benali, K., Benali, K., Santin, M.D., 2019. Post mortem high resolution diffusion MRI for large specimens imaging at 11.7 T with 3D segmented echo-planar imaging. J. Neurosci. Methods. doi: 10.1016/j.jneumeth.2018.10.010.

Seehaus, A., Roebrock, A., Bastiani, M., Fonseca, L., Bratze, H., Lori, N., Villanova, A., Goebel, R., Galuske, R., 2015. Histological validation of high-resolution DTI in human post-mortem tissue. Front. Neurol. 6, 98.

Shatil, A.S., Uddin, M.N., Matsuda, K.M., Figrle, C.R., 2018. Quantitative MRI changes due to progressive familial fixation in whole human brain specimens: longitudinal characterization of diffusion, relaxometry, and myelin water fraction measurements at 3T. Front. Med. 5, 31.

Shen, K., Goulas, A., Grayson, D.S., Eusebio, J., Gati, J.S., Menon, R.S., McIntosh, A.R., Everling, S., 2019. Exploring the limits of network topology estimation using diffusion tractography and tracer studies in the macaque cortex. Neuroimage 191, 81–92.

Smith, S.M., Jenkinson, M., Woolrich, M.W., Beckmann, C.F., Behrens, T.E.J., Janssen-Berg, H., Bamister, P.R., De Luca, M., Drobnjak, I., Fitzne, D.E., Nizio, R.K., Sobesky, J., Vickson, K., van der Stafika, V., van der Staefika, N., Brady, J.M., Matthews, P.M., 2004. Advances in functional and structural MRI model image analysis and implementation as FSL Neuroimage 23 (Suppl 1), S208-S219.

Soares, J.M., Marques, P., Alves, V., Sousa, N., 2013. A hitchhiker’s guide to diffusion tensor imaging progress. Front. Neuroanat. 7, 3589/finwa.2013.00031.

Sotirovopoulos, S.N., Jhabhdi, S., Xu, J., Anderson, J.L., Moeller, S., Auerbach, E.J., Glasser, M.F., Hernandez, M., Sapico, G., Jenkinson, M., Feinberg, D.A., Yacoub, E., Lenglet, C., Eisen, D.C., Van Ugurbil, K., Behrens, T.E.J., 2013. Advances in diffusion MRI acquisition and processing in the Human Connectome Project. Neuroimage 80, 125-143. doi: 10.1016/j.neuroimage.2013.05.057.

Sotirovopoulos, S.N., Hernández-Fernández, M., Vu, A.T., Anderson, J.L., Moeller, S., Yacoub, E., 2017. High-resolution diffusion imaging at 7T: an atlas of the human connectome project. Neuroimage 134, 396-409.

Stephan, K.E., Kemper, L., Kokott, A., Burns, G.A., Young, M.P., Köttler, R., 2001. Advanced database methodology for the collection of connectivity data on the macaque brain (CoMmac). Philos. Trans. R. Soc. Lond. B Biol. Sci. 356, 1159-1186.

Sunderland, S., 1940. The distribution of commissural fibres in the corpus callosum in the monkey. J. Comp. Neurol. 72, 241-270.

Tavaf, N., Lagore, R.L., Jung, S., Gunamony, S., Radder, J., Grant, A., Moeller, S., Auerbach, E.J., Yacoub, E., Ugurbil, K., Adiyan, G., Van de Moortele, P.F., 2021. A self-decomposed 32-channel receive array for human-brain MRI at 10.5 T. Magn. Reson. Med. 86, 664-672. doi: 10.1002/mrm.28065.

Thomas, C. Ye, F.Q., Irfagno, M.O., Modii, P., Saleem, K.S., Leopold, D.A., Pierpaoli, C., 2014. Anatomical accuracy of brain connections derived from diffusion MRI is inherently limited. Proc. Natl. Acad. Sci. U.S.A. 111, 16574-16579.
Yacoub, E., Grier, M.D., Auerbach, E.J., Lagore, R.L., Harel, N., Ugurbil, K., Adriany, G., Zilverstand, A., Hayden, B.Y., Heilbronner, S.R., Zimmermann, J., 2022 n.d. Ultra-high field (10.5 T) resting state fMRI in the macaque. doi:10.1101/2020.05.21.109595.

Yon, M., Bao, Q., Chitrit, O.J., Henriques, R.N., Shemesh, N., Frydman, L., 2020. High-resolution 3D in vivo brain diffusion tensor imaging at ultrahigh fields: following maturation on juvenile and adult mice. Front Neurosci doi:10.3389/fnins.2020.590900.

Zhang, H., Schneider, T., Wheeler-Kingshott, C.A., Alexander, D.C., 2012. NODDE: practical in vivo neurite orientation dispersion and density imaging of the human brain. Neuroimage 61, 1000–1016.

Zitella, L.M., Xiao, Y., Teplitzky, B.A., Kastl, D.J., Duchin, Y., Baker, K.B., Vitek, J.L., Adriany, G., Yacoub, E., Harel, N., Johnson, M.D., 2015. In vivo 7T MRI of the non-human primate brainstem. PloS ONE 10, e0127049.