Fibre-specific white matter changes in multiple sclerosis patients with optic neuritis

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A R T I C L E  I N F O

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A B S T R A C T

Long term irreversible disability in multiple sclerosis (MS) is thought to be primarily driven by axonal degeneration. Axonal degeneration leads to degenerative atrophy, therefore early markers of axonal degeneration are required to predict clinical disability and treatment efficacy. Given that additional pathologies such as inflammation, demyelination and oedema are also present in MS, it is essential to develop axonal markers that are not confounded by these processes. The present study investigated a novel method for measuring axonal degeneration in MS based on high angular resolution diffusion magnetic resonance imaging. Unlike standard methods, this novel method involved advanced acquisition and modelling for improved axonal sensitivity and specificity. Recent work has developed analytical methods, two novel axonal markers, fibre density and cross-section, that can be estimated for each fibre direction in each voxel (termed a “pixel”). This technique, termed a pixel-based analysis, thus simultaneously estimates axonal density and white matter atrophy from specific white matter tracts. Diffusion-weighted imaging datasets were acquired for 17 patients with a history of acute unilateral optic neuritis (35.3 ± 10.2 years, 11 females) and 14 healthy controls (32.7 ± 4.8 years, 8 females) on a 3 T scanner. Fibre density values were compared to standard diffusion tensor imaging parameters (fractional anisotropy and mean diffusivity) in lesions and normal appearing white matter. Group comparisons were performed for each pixel to assess putative differences in fibre density and fibre cross-section. Fibre density was observed to have a comparable sensitivity to fractional anisotropy for detecting white matter pathology in MS, but was not affected by crossing axonal fibres. Whole brain pixel-based analysis revealed significant reductions in fibre density and fibre cross-section in the inferior fronto-occipital fasciculus (including the optic radiations) of patients compared to controls. We interpret this result to indicate that this pixel-based approach is able to detect early loss of fibre density and cross-section in the optic radiations in MS patients with a history of optic neuritis. Fibre-specific markers of axonal degeneration should be investigated further for use in early stage therapeutic trials, or to monitor axonal injury in early stage MS.

1. Introduction

Multiple sclerosis (MS) is the most common neurological disorder affecting young adults. Pathologically, MS is characterised by the presence of inflammatory demyelinating lesions, associated with axonal loss within the central nervous system. Axonal degeneration leads to neurodegeneration visible as atrophy that is known to be associated with clinical disability (De Stefano et al., 2014). Therefore, early markers that are indicative of atrophy are required to predict clinical disability.

In acute MS lesions, inflammatory cells can directly injure axons leading to axonal transection and subsequently Wallerian and retrograde degeneration (Trapp and Stys, 2009). Additionally, chronic demyelination can lead to axonal degeneration via excitotoxic...
mechanisms due to the reliance on excess sodium channels required to transmit action potentials along demyelinated axonal segments (Trapp and Stys, 2009). Trans-synaptic degeneration can also occur due to loss of synaptic inputs from degenerating afferent neurons (Getti et al., 1972; Cook and Wisniewski, 1987; Ginsberg and Martin, 2002) or loss of efferent targets (Reich et al., 2009). With therapeutic interventions and clinical trials aiming to reduce or reverse axonal degeneration, there is a need for quantitative in vivo markers that can quantify axonal degeneration specifically and sensitively in early stage MS.

Diffusion weighted MRI, a technique that is sensitive to diffusion of water molecules, can provide information about the underlying microstructure of the brain. Diffusion tensor imaging, a method introduced over 20 years ago (Basser et al., 1994), remains the standard approach for quantifying diffusion in clinical neuroimaging studies. Voxel based metrics derived from diffusion tensor imaging include fractional anisotropy, reflecting the degree of diffusion anisotropy, and mean diffusivity, reflecting the overall magnitude of diffusion. Cross-sectional diffusion tensor imaging studies in MS patients have repeatedly reported reduced fractional anisotropy and increased mean diffusivity within lesions and normal appearing white matter (Ciccarelli et al., 2001; Gratasias et al., 2015). However, diffusion tensor imaging only models diffusion assuming a model of Gaussian diffusion for a single population of parallel axons, and is thus confounded by partial volume effects in voxels containing multiple fibre populations (up to 90% of white matter voxels), CSF and other non-axonal cell types (Alexander et al., 2001; Metzler-Baddeley et al., 2012; Jeurissen et al., 2013).

More recently, methods combining high angular resolution and higher diffusion weighting with higher-order diffusion models have been developed to overcome limitations of diffusion tensor imaging (Tuch et al., 2003; Tournier et al., 2004; Assaf and Basser, 2005; Wedeen et al., 2005; Wang et al., 2011; Raffelt et al., 2012b; Zhang et al., 2012; Dell’Acqua et al., 2013). However, some of these diffusion models suffer limitations including: (1) the estimation of diffusion distributions (Tuch et al., 2003; Wedeen et al., 2005) rather than fibre distributions that are better able to delineate crossing fibres of < 90° separation (Tournier et al., 2008), or (2) estimation of diffusion parameters averaged over whole voxels without considering each fibre population separately (Wang et al., 2011; Zhang et al., 2012).

Here we utilize a diffusion analysis method that seeks to overcome these limitations, broadly termed “Fixel Based Analysis”. Using constrained spherical deconvolution, it is possible to estimate the continuous fibre orientation distribution for each voxel (Tournier et al., 2007). By segmenting the fibre orientation distributions within each voxel, individual underlying fibre populations (termed “fixels”) can be identified (Raffelt et al., 2017). Previous studies have shown that the fibre orientation distribution amplitude is proportional to the apparent fibre density of axons aligned in that direction (Raffelt et al., 2012b). Also, nonlinear registration of fibre orientation distributions provides information regarding local macroscopic volumetric changes in the white matter, termed fibre cross-section. Fixels can thus be used to represent fibre density and cross-section for each fibre population at a sub-voxel level for group-level statistical comparison (Raffelt et al., 2015). Fixel based analysis therefore allows for simultaneous group analysis of microscopic fibre density and macroscopic white matter tract atrophy across the entire white matter.

Ultimately fixel-based analyses could provide early markers of fibre specific axonal loss in MS for use in disease monitoring and prognosis. To that end, we first undertook a study aiming to examine pathological changes in FD in the white matter of people with MS, and the qualitative and quantitative relationships between FD and standard diffusion tensor imaging metrics. We then used whole brain fixel-based analysis to compare FD and FC between people with MS who presented with optic neuritis and healthy control subjects. We hypothesised that FD there would be reduced fibre density and cross-section in the primary visual pathway consistent with previous observations of microstructural changes in this pathway in patients with optic neuritis (Kolbe et al., 2012a).

2. Methods

2.1. Subjects

Seventeen patients (age 21–50 years; 11 females, 6 males) with a history of optic neuritis (disease duration 4.48 ± 0.61 years) were recruited from the Royal Victorian Eye and Ear Hospital, Australia. These patients were a subset of a cohort of patients (n = 38) who were recruited to study optic pathway changes after acute optic neuritis (van der Walt et al., 2013). This subset agreed to return for a research MRI scanning session three and five years after initial recruitment to study longer term outcomes after optic neuritis. At the time of testing, all patients had been diagnosed with clinically definite MS according to the 2010 McDonald criteria and twelve out of seventeen patients were receiving treatment. All patients were experiencing acute optic neuritis at the time of recruitment and some patients also experienced subsequent relapses prior to scanning for this study. Fourteen healthy controls (age 27–43 years; 8 females, 6 males) with no reported history of neurological disease were recruited for comparisons. Disease severity was assessed in all patients with the Expanded Disability Status Scale (EDSS) (median: 1; range: 0–2).

This study was conducted in accordance with the Declaration of Helsinki and was approved by the Human Research Ethics Committees of the Royal Victorian Eye and Ear Hospital and the Royal Melbourne Hospital. All study participants provided voluntary, written consent.

2.2. Ocular testing

Optic nerve and retinal injury were measured using two techniques: (1) optical coherence tomography of the retina to measure the thickness of the retinal nerve fibre layer, and (2) multifocal visual evoked potential amplitude to measure loss of visual pathway conductance associated with axonal loss. Optical coherence tomography was performed using an OCT-3 scanner (Stratus™, software version 3.0, Carl Zeiss Meditec Inc.) using the Fast RNFL protocol. The acquisition consists of three circular 3.4 mm diameter scans centred on the optic disc. Scans with signal strength of < 7 were discarded. Multifocal visual evoked potentials were recorded using the Accumap™ (ObjectiVision: Opera, Sydney, Australia) by a trained technician. The procedure entailed recording from 58 sectors of the visual field with four electrodes placed over the inion on the rear of the skull. Eyes were individually stimulated for 10 to 12 runs until a sufficient signal to noise ratio was reached.

2.3. MRI acquisition

Image acquisition was performed using a Siemens Trio 3 T MRI system (Siemens, Erlangen, Germany) with a 32 channel head coil. Diffusion weighted imaging (DWI) data were acquired with the following parameters: repetition time = 7800 ms, echo time = 112 ms, voxel size = 2.5 × 2.5 × 2.5 mm³. Sixty diffusion-weighted images (b = 3000 s/mm²), and seven non-diffusion weighted images (b = 0) were acquired with a spin-echo echo planar imaging sequence. A pair of non-diffusion weighted images with opposite phase encoding was also acquired to correct for susceptibility induced distortion. In addition, T1-weighted MPRAGE images (repetition time = 1900, echo time = 2.63 ms; flip angle = 9°; voxel size = 0.8 × 0.8 × 0.8 mm³) were acquired. Lastly, T2-weighted double inversion recovery SPACE images (repetition time = 7400 ms, effective echo time = 105 ms; inversion times = 3000 and 450 ms, echo train length = 194, flip angle = 120°; voxel size = 1.1 × 1.1 × 1.1 mm³ sagittal acquisition) were obtained for patients to identify T2 hyperintense brain lesions. Lesions were segmented and verified by a neuroradiologist of nine years
experience (EL), who was blinded to each patient’s clinical status using MRicro (www.cabi.gatech.edu/micro/mricro/).

2.4. Diffusion imaging processing

2.4.1. Diffusion-weighted MRI pre-processing

Diffusion-weighted MRI pre-processing involved field-map correction of EPI distortions (TOPUP, FSL, v.5.0.4), motion correction with correction for current-induced distortions (refer to Appendix A of Raffelt et al., 2012b), bias field correction (N4, http://stnava.github.io/ANTs/). Data was also intensity normalised across all subjects using the median $b = 0\text{s/mm}^2$ intensity within a white matter mask and upsampling by a factor of two to improve fibre orientation distribution registration and statistics (Raffelt et al., 2012b).

Diffusion tensor imaging analysis was performed with MRtrix3. Fractional anisotropy and mean diffusivity maps were derived for each subject. Note that voxels were masked out if the mean diffusivity values were deemed physically implausible (negative or 100 times greater than the median).

2.4.2. Fibre orientation distribution calculation

To generate a more accurate, specific and sensitive representation of the white matter fibre orientation distribution, single-shell 3-tissue constrained spherical deconvolution (SS3T-CSD) was applied using tissue-specific response functions for white matter, grey matter and cerebrospinal fluid (Dhollander and Connelly, 2016; Dhollander et al., 2016). These response functions are tissue-specific signal models, derived from the diffusion data itself. While subsequent analysis steps focus exclusively on fibre density measured from the white matter fibre orientation distributions, the inclusion of grey matter and cerebrospinal fluid components in the model accounts for the remainder of the complete diffusion-weighted MRI signal due to healthy tissues as well as contributions from certain inflammatory cells (Dhollander et al., 2017). Both response function estimation and SS3T-CSD are fully unsupervised and directly driven by the diffusion-weighted MRI data itself.

2.4.3. Fibre orientation distribution template construction

A population-averaged fibre orientation distribution template was generated from a subset of subjects (10 patients and 10 controls). In brief, the fibre orientation distribution of each subject was repeatedly non-linearly registered to optimise the template (Raffelt et al., 2011, 2012a). For each subject, a deformation field that encodes nonlinear transformations to the template space was generated. The fibre orientation distributions in the template are first segmented to identify pixels that are mapped reasonably well across all subjects in both populations—we will refer to this outcome as the template voxel orientation mask.

2.4.4. Fibre density and cross-section calculation

Firstly, each subject’s fibre orientation distribution map was transformed to the fibre orientation distribution template and segmented to identify the subject fixes in each voxel. Secondly, the fixes were re-oriented and correspondences between the subject fixes and the template fixes were computed.

Fixed-specific measures are calculated for each fixed in the voxel template analysis mask. The integral of fibre orientation distribution lobes is proportional to the intra-axonal volume of axons oriented along the fixed direction, and can be interpreted as the apparent fibre density (Raffelt et al., 2012b, 2017). Unlike standard diffusion tensor imaging metrics, each individual fixed is associated with a specific fibre density (FD) metric, even in voxels that contain multiple fibre populations. Furthermore, the deformation field created from the registration of each fibre orientation distribution map to the template contains morphological information about cross-sectional volume changes of white matter fibre bundles. The Jacobian determinant of the deformation field is able to quantify such local volumetric changes. Local expansion and contraction perpendicular to the fixed orientation provides a quantitative assessment of fibre cross-section (FC) differences (Raffelt et al., 2017). By multiplying fibre density and fibre cross-section at each fixed, a combined measure of fibre density and cross-section (FDC) is generated as well. The FDC accounts for both the microscopic and macroscopic changes in a fibre bundle.

2.4.4.1. Comparisons between fibre density and diffusion tensor imaging in pathological white matter

To evaluate fibre density against standard diffusion tensor imaging metrics, we performed qualitative regional and quantitative whole brain comparisons. These comparisons required us to convert fibre density from a fixed-wise measure to a simplified voxel-wise measure, and thus reduced the amount of potential information encoded in the fixed-based analysis. The angular integral of the fibre orientation distribution in each voxel was considered as the total fibre density.

For the qualitative analysis, we identified regions of interest in a single MS patient to evaluate fibre density in two lesions of differing severity and corresponding regions in a healthy control were also investigated. First, the fibre orientation distribution maps were transformed into template space to allow for comparisons between subjects. Voxel-wise measures, total fibre density and fractional anisotropy, were examined within the regions of interest (31 × 31 voxels), which included the lesion and surrounding white matter. To examine the non-linear difference between fractional anisotropy from total fibre density, a spatial representation of the difference between the two measures was calculated by subtracting the two measures after normalisation by their maximum and minimum values. Given that the scales for total fibre density and fractional anisotropy are different, normalisation provides values that are within the same scale. For all images, the maximum FD and FA voxels were not outliers (all data points are clearly presented in Fig. 1B scatterplots).

For the quantitative analysis, each subject’s T1-weighted image was segmented to generate a binary white matter mask using MRtrix. A rigid affine registration (ANTs, version 1.5) was used to transform tissue masks to diffusion MRI space using nearest neighbour interpolation. In patients, binary lesion masks were registered from double inversion recovery image space to diffusion space with a rigid affine registration with a nearest neighbour interpolation. For patients, the lesion mask was subtracted from the white matter mask to create a normal appearing white matter mask.

2.4.4.2. Whole-brain statistical analysis of fixed measures

FD, FA and FDC were compared between groups using fixed-based analyses. Template-based probabilistic tractography (angle = 22.5, maxlength = 250 mm, minlength = 10 mm, power = 1.0, tracks = 20 million) was performed to identify structurally connected fixeds that are more likely to share underlying anatomy and pathology. Tractograms were filtered to minimise tractography seeding biases (Smith et al., 2013) using MRtrix. This process resulted in a final track count of 2 million that was used for connectivity-based smoothing and statistical inference. In short, the template-based tractography map was used to identify fixes that are anatomically connected given that each fibre tract can be pathologically affected differently.

2.4.4.3. Correlations between fixed-wise measures and brain and ocular measures

Post-hoc correlation analyses were performed between the fixed-wise measures and both brain volume and ocular measures in patients in order to better interpret the sources of pathological variance in fixed measures. Average FD and FA were obtained from regions of interest including all significant fixeds identified using fixed-based analysis.

Measures of brain and optic nerve injury were calculated to correlate with fixed measures. From the optical coherence tomography the average retinal nerve fibre layer thickness was calculated for each eye. From
multifocal visual evoked potential, the mean amplitude was calculated for each visual field sector and then for each eye as a whole. All measures from the affected eye were expressed as a normalised asymmetric percentage to account for inter-subject variation, while all measures from the unaffected eye were expressed as raw values. Brain volume measures were calculated with FreeSurfer (version 5.1.0) using the standard pipeline (Fischl et al., 2002) and expressed as fractions of intracranial volume to account for differences in head size. Specifically, lateral ventricle, cortical grey matter, subcortical grey matter (encompassing the thalamus, caudate, accumbens and ventral diencephalon) were examined.

2.5. Statistical analyses

2.5.1. Comparison between fibre density and diffusion tensor imaging

The average total fibre density, fractional anisotropy and mean diffusivity were extracted using the binary masks for each subject. A paired t-test was used to compare lesions to normal appearing white matter in patients, and independent sample t-tests were used to compare lesions and normal appearing patient white matter to healthy white matter in control subjects. Correction for multiple comparisons was performed with FDR.

2.5.2. Whole-brain voxel-based analysis

To compare quantitative measures at each pixel, general linear models were used to identify significantly reduced FD, FC and FDC for each pixel in patients compared to controls. Unlike voxel-based analysis, voxel-based analysis takes into consideration that a white matter voxel can contain multiple fibre populations, and each population can be independently affected by distal or proximal damage. Family-wise error corrected p-values were assigned to each pixel with nonparametric permutation testing using 5000 permutations (Raffelt et al., 2015). Significance was denoted by family-wise error corrected p < 0.05. To visualise the findings, the tractography map was cropped to include the streamlines that correspond with significant fibres.

2.5.3. Correlation between voxel-wise measures and, brain and ocular injury

Volumetric measures were compared between groups with independent samples t-tests (for equal and unequal variances) or Mann-Whitney U tests when violations of the assumption of normality occurred. Spearman correlation analyses were performed between regional fibre measures and structural and functional measures of visual and brain injury, after correcting for age as a covariate using linear regression. Correlations were considered significant at p < 0.05. Corrections for multiple comparisons were performed with FDR.

3. Results

Table 1 summarises demographic and disease related data for patients and control subjects. Patients exhibited extant loss of retinal nerve fibre layer thickness and visual evoked potential amplitude for the affected eye. Compared to controls, patients displayed significant atrophy of cortical grey matter but no ventricular enlargement, and cerebral white matter and subcortical grey matter atrophy.

3.1. Qualitative comparisons between total fibre density and diffusion tensor imaging metrics in pathological tissue

The fibre orientation distributions generated with SS3T-CSD were qualitatively assessed in two lesions of differing severity (Fig. 1A). Close inspection of lesion 1 revealed reduced fibre orientation distribution amplitudes associated with reduced fixed FD compared to healthy white matter. Within lesion 2, the fibre orientation distribution amplitudes were reduced, yet the fibre orientations were preserved relative to the healthy control data.

Voxel-wise total fibre density and fractional anisotropy metrics were qualitatively compared within the regions of interest in the MS patient and healthy control (Fig. 1B). The difference between the two measures provides a spatial representation of the nonlinearities between the measures in pathological tissue and crossing-fibres. Overall, total fibre density appears more homogeneous than fractional anisotropy in lesions and surrounding white matter. Importantly, reduction in fractional anisotropy is not specific to regions of pathology, but can also be seen in healthy areas where fibre tracts cross, including within the periventricular regions that were lesion in the patient. Scatterplots showing the relationships between normalised total fibre density and fractional anisotropy provide further qualitative insights into the nonlinearities between the two measures. As expected, voxels that deviate the most from the line of unity lie within regions of crossing fibres.

3.2. Quantitative comparisons between total fibre density and diffusion tensor imaging metrics in pathological tissue

Averaged measures for total fibre density, fractional anisotropy and mean diffusivity were calculated for each subject to characterize the degree of difference in these measures in lesions and normal appearing white matter compared to normal healthy white matter (Fig. 2). Both total fibre density and fractional anisotropy were significantly different for the following comparisons: (1) lesions vs healthy white matter (total fibre density: d = 5.70, p = 0.86 × 10\(^{-12}\), fractional anisotropy: d = 3.56, p = 5.57 × 10\(^{-9}\), (2) lesions vs normal appearing white matter (total fibre density: d = 4.32, p = 0.61 × 10\(^{-9}\), fractional anisotropy: d = 3.08, p = 0.29 × 10\(^{-6}\), and (3) normal appearing vs healthy white matter (total fibre density: d = 0.88, p = 0.026; fractional anisotropy: d = 1.06, p = 0.007). For mean diffusivity, a difference was detected between lesions and healthy white matter (d = 6.05, p = 3.86 × 10\(^{-12}\)) and normal appearing white matter (d = 5.68, p = 13.78 × 10\(^{-12}\)), yet no difference was detected between healthy and normal appearing white matter (d = 0.73, p = 0.061).

3.3. Changes in voxel-specific measures after optic neuritis

Whole brain voxel-based analyses reported significant bilateral reduction of FD, FC, and FDC in the inferior longitudinal fasciculi (ILF), and reduction of FDC in the cingulum bundles in patients compared to controls (Fig. 3). Refer to Supplementary Fig. 1 for voxel-wise Cohen’s d values. Average fibre density within the significant regions of interest did not correlate with visual (retinal nerve fibre layer thickness or visual evoked potential amplitude) or brain volumetric measures (lateral ventricle, subcortical or cortical grey matter) (Supplementary Table 1). Loss of fibre density was moderately associated with enlargement of the lateral ventricles (p = −0.52, p = 0.035).

4. Discussion

This study aimed to investigate the use of voxel-specific measures...
Table 1
Demographics and clinical data. Measures describing the visual and MRI volumetric changes in patients and controls. Measures from the affected eye reported as an asymmetric percentage, normalised to the unaffected eye. Measures from the unaffected eye reported as a raw value. A Mann-Whitney U test was performed. RNFL = retinal nerve fibre layer, mVEP = multifocal visual evoked potential.

<table>
<thead>
<tr>
<th></th>
<th>MS patients (n = 17) 11 females</th>
<th>Healthy controls (n = 14) 8 females</th>
<th>p-Value</th>
</tr>
</thead>
<tbody>
<tr>
<td>Age (years)</td>
<td>35.3 ± 10.2</td>
<td>32.7 ± 4.8</td>
<td>0.390</td>
</tr>
<tr>
<td>Lesion volume (mm³)</td>
<td>2811.13</td>
<td>–</td>
<td>–</td>
</tr>
<tr>
<td>EDSS score (median; range)</td>
<td>1; 0.4</td>
<td>–</td>
<td>–</td>
</tr>
<tr>
<td>Number of relapses (median; range)</td>
<td>2; 1–4</td>
<td>–</td>
<td>–</td>
</tr>
<tr>
<td>Disease duration (years)</td>
<td>4.48 ± 0.61</td>
<td>–</td>
<td>–</td>
</tr>
<tr>
<td>Visual measures</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>RNFL</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Affected eye (%)</td>
<td>–14.06 ± 12.94</td>
<td>–</td>
<td>–</td>
</tr>
<tr>
<td>Unaffected eye (µm)</td>
<td>97.87 ± 13.23</td>
<td>–</td>
<td>–</td>
</tr>
<tr>
<td>mVEP amplitude</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Affected eye (%)</td>
<td>–9.22 ± 16.86</td>
<td>–</td>
<td>–</td>
</tr>
<tr>
<td>Unaffected eye (nV)</td>
<td>170.74 ± 35.98</td>
<td>–</td>
<td>–</td>
</tr>
<tr>
<td>Volumetric measures (× 10⁻³)</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Lateral ventricle fraction¹</td>
<td>1.11 ± 0.67</td>
<td>0.80 ± 0.26</td>
<td>0.173</td>
</tr>
<tr>
<td>Subcortical grey matter fraction¹</td>
<td>2.08 ± 0.19</td>
<td>2.12 ± 0.14</td>
<td>0.512</td>
</tr>
<tr>
<td>Cortical grey matter fraction</td>
<td>35.12 ± 4.21</td>
<td>39.74 ± 2.75</td>
<td>0.001</td>
</tr>
<tr>
<td>White matter fraction</td>
<td>31.91 ± 2.73</td>
<td>31.55 ± 2.53</td>
<td>0.712</td>
</tr>
</tbody>
</table>

(fibre density (FD), fibre cross-section (FC), and fibre density & cross-section (FDC)) as markers for axonal loss in the context of MS. Fixel-based analysis has previously been used to investigate axonal changes in motor neurone disease, temporal lobe epilepsy and pubertal development (Raffelt et al., 2012b; Genc et al., 2017; Raffelt et al., 2017; Vaughan et al., 2017). In MS, white matter pathologies additional to axonal loss can include inflammation, demyelination, oedema and gliosis. However, by focusing on the anisotropic white matter component of the diffusion MRI signal, fibre density measures are highly specific for axonal density changes. Additionally, fixel-based analysis quantifies macroscopic volume changes. Given that axonal degeneration leads to progressive atrophy, which has been associated with irreversible disability, an early precursor of atrophy may aid to predict clinical disability. Axonal degeneration is a potential early marker. Here we have applied a fixel-based analysis in a cohort of early MS patients with a history of acute optic neuritis and shown that indeed fixel-based loss was detected in the visual pathway as we hypothesised.

4.1. Fibre density and fibre cross-section of the ILF is reduced in optic neuritis patients

Using fixel-based analysis we compared FD, FC and FDC in healthy participants to those in white matter of patients with historical optic neuritis. Consistent with our expectations, we observed a relatively specific loss of fibre density, and to a lesser extent fibre cross-section in patients within the ILF, a white matter tract that connects the occipital and temporal lobes. Anatomically it sits closely with the optic radiations posterior to the lateral geniculate nucleus. In addition, reduced FD was reported in the cingulum, an important fibre bundle involved in communication between the frontal, temporal and parietal lobes. Previous studies have shown that anterograde trans-synaptic degeneration after optic neuritis is associated with atrophy to the lateral geniculate nucleus, axonal degeneration to the optic radiation and cortical thinning to the visual cortex (Evangelou et al., 2001; Rocca et al., 2013; von Glehn et al., 2014; Tur et al., 2016). Indeed, previous diffusion tensor imaging studies by our group have reported loss of axial diffusivity in the optic radiations of patients with optic neuritis (Kolbe et al., 2012a, 2016). Loss of axial diffusivity is difficult to interpret pathologically but has been observed in the context of acute axonal injury (Song et al., 2003; Wu et al., 2007; Zhang et al., 2009). In contrast to previous studies, we employed a diffusion-weighted MRI sequence with high diffusion weighting (b-value = 3000 s/mm²), which results in near complete attenuation of the signal from fast diffusing extracellular free-water and longitudinal axonal diffusion. Also, the relatively long echo time employed for diffusion imaging (112 ms) was such that the T2 from myelin water was fully decayed. In addition, S3ST-CSD can improve the specificity to the white matter intra-axonal space by removing signal contributions from certain isotropic diffusion components that could be associated with inflammatory cell infiltration. We therefore postulate that the observed findings in this cohort with a relatively short disease duration reflect a loss of fibre density in visual pathways as the initial manifestation of damage, which is then followed by a reduction in fibre bundle cross-section. Considering that fibre density and cross-section are derived from two separate sources of information (diffusion signal and nonlinear transformation field respectively), the multiplication of their effects as FDC provides a more comprehensive measure that characterises both microscopic and macroscopic fibre changes simultaneously (Raffelt et al., 2017).

Fig. 2. Pathological effect size comparisons between fibre density and diffusion tensor imaging metrics. Subject-wise comparisons of average total fibre density (FD(total)), fractional anisotropy (FA) and mean diffusivity (MD) extracted from T2 lesions, normal appearing white matter (NAWM) and normal white matter (NWM) (**p = 0.001, *p = 0.01, p = 0.05).
Contrary to our expectations, loss of optic radiation FDC was not correlated with the degree of optic nerve axonal loss or brain volume loss (apart from a trend with ventricular enlargement). We have previously shown that loss of optic radiation axial diffusivity correlated with affected optic nerve axonal loss measured using multifocal visual evoked potentials and optical coherence tomography (Kolbe et al., 2012b, 2016). Despite this apparent conflict, we are more confident in the current results given the greater pathological specificity of fibre density over diffusion tensor imaging markers such as axial diffusivity whose pathological substrate remains elusive. Also, the lack of correlation between pixel measures and grey matter atrophy further supports the notion that grey and white matter damage should be examined independently to provide a comprehensive assessment of neurodegeneration in people with MS.

4.2. Diffusion metrics within lesions and normal appearing white matter

To better contextualise fibre density in terms of standard DTI parameters, we performed qualitatively and quantitative comparisons of lesions and NAWM. We observed a loss of fibre density in lesions and NAWM compared to healthy white matter. Focused assessments of lesions demonstrated clear demarcation of lesion boundaries using FD. Despite the ability to identify lesions with fractional anisotropy, the metric exaggerates the loss of fibres due to crossing fibres and partial volume inclusion from other non-white matter tissue types, resulting in low fractional anisotropy. However, from the fibre orientation distribution and pixel overlays it is apparent that the lesions contain crossing fibres. Since some lesions have greater fibre loss than others, fibre density measures provide the opportunity to differentiate these lesions and their likely clinical consequences by using pixel-based analysis.

4.3. Advantages of pixel-based analysis: diffusion model

Diffusion MRI sequences commonly used for diffusion tensor imaging employ low diffusion weighting ($b$-value ≤ 1000 s/mm$^2$) which provides limited angular contrast and sensitivity to more rapidly moving molecular pools outside the axon. Furthermore, the diffusion tensor model does not account for the partial volume inclusion of other tissue types. Therefore, despite significant correlations reported between diffusion tensor imaging metrics and axonal density in post-mortem histology/imaging studies, diffusion tensor imaging parameters

![Fig. 3. Significant loss of fibre density, cross-section and fibre density and cross-section in the visual pathways of patients with a history of optic neuritis. To enable visualisation, streamlines from the template-derived whole-brain tractography map were cropped to include streamlines that correspond with significant pixels (family-wise error corrected $p$-value < 0.05), and coloured by $p$-value. Significant tracts are overlayed on a 2D glass-brain in the axial, coronal and sagittal plane. A close-up of the significant pixels are displayed in the lowest panel. Results are displayed on the population-specific template. FD = fibre density, FC = fibre cross-section, FDC = fibre density and cross-section.](image-url)
are not solely affected by axonal density (Schmierer et al., 2007). Secondly, the diffusion tensor model assumes a single fibre orientation within each voxel; however, it is now known that ~90% of white matter voxels in the brain contain more than one fibre population (Jeurissen et al., 2013). Here, we overcome these limitations by estimating the complete fibre orientation distribution. In the context of MS, several other higher-order diffusion models have been proposed that can account for inflammation associated cellularity, gliosis and oedema (Assaf and Basser, 2005; Wang et al., 2011; Zhang et al., 2012). Similar to these methods, SSST-3SD incorporates additional response functions for grey matter and CSF, which allow the partial volume fraction estimation of these signals in the white matter to improve the specificity to axonal contributions and damage (Dhollander and Connelly, 2016). It has been shown before that these “grey matter-like” and “CSF-like” signals can also account for signal contributions due to inflammatory processes, gliosis and oedema (Dhollander et al., 2017).

4.4. Advantages of fixed-based analysis: statistical analysis

MS has increasingly been viewed as a disconnection syndrome with disruptions to white matter connectivity so a method for examining specific white matter tracts across multiple subjects is highly desirable (Rocca et al., 2015). The tract-based spatial statistics method has gained popularity in this context for the population analysis of diffusion tensor imaging data (Dineen et al., 2009; Roosendaal et al., 2009). Fixed-based analysis provides an alternative approach to statistically compare fixed-specific measures across subjects (Raffelt et al., 2015) by registering all image data to a common template (Raffelt et al., 2011, 2012a, 2017). Unlike voxel-based approaches, fixed-based analysis considers the connectivity between fibres in a directionally meaningful way such that signals from orthogonally intersecting tracts are not combined during statistical clustering. This is a critical innovation that recognises the fact that each fibre tract can be pathologically affected differently (Raffelt et al., 2015). Fixed connectivity is estimated by performing a template-based tractography to smooth and enhance fibres that are structurally connected. By doing so, fixed-based analysis can identify specific white matter tracts that exhibit a disruption to fibre connectivity allowing for more meaningful interpretation of white matter damage in terms of patient functional outcomes.

4.5. Limitations and future work

In the current study we examined a relatively small clinically homogenous sample of patients. However, the inclusion of patients with a consistent pathological insult (optic neuritis) provided a useful opportunity to examine the anatomical specificity of fixed-based analysis given the very specific hypothesis we were able to form regarding the location of axonal damage in the visual pathways. It should be noted that the current study included a cohort of early MS patients with minimum lesional damage, therefore there would not be a significant problem during the image registration steps. However, pathologically severe lesions that show complete loss of fibre density in the lesion core can potentially affect image registration locally. In future studies, we intend to examine more severe MS cases where the presence of lesions can be a problem during image registration. In addition, we aim to examine more heterogeneous and larger samples of patients over time to understand the temporal and spatial pattern of fixed-specific changes. Specifically, longitudinal studies will provide an opportunity to investigate the sensitivity of fixed measures for detecting changes in individual patients to provide effect size estimates for potential trials using fixed measures as endpoints. Additionally, the growing use of more realistic diffusion models to assess the underlying microstructure will require future research to compare these models in specific disease contexts. Finally, while we have inferred loss of fibre density to be associated with axonal loss based on solid theory, validation of the measure against gold standard histological measures of axonal density would be advantageous for correctly interpreting fibre density changes in patients.

5. Conclusion

In summary, we have used fixed-based analysis to detect tract-specific loss of fibre density and cross-section in the posterior visual pathway following optic neuritis. We contend that the advantages provided by fixed-based analysis compared to conventional diffusion tensor methods (i.e. greater specificity for axonal density, sensitivity to damage in multiple fibre populations within a voxel, simultaneous measurement of microstructural and macroscopic neurodegenerative changes) have enabled improved understanding of the characteristics of white matter lesions and of more widespread white matter injury in MS. Importantly, this study has demonstrated that the fixed-based analysis as a potential early marker of clinical disability in MS. With advanced markers of axonal degeneration it can potentially aid in treatment efficacy and predict clinical outcome.

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Appendix A. Supplementary data

Supplementary data to this article can be found online at https://doi.org/10.1016/j.jicll.2017.09.027.

References


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