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Age related diffusion and tractography changes in typically developing pediatric cervical and thoracic spinal cord

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Age related diffusion and tractography changes in typically developing pediatric cervical and thoracic spinal cord

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A R T I C L E   I N F O
Keywords:
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A B S T R A C T

Background and objective: Diffusion tensor imaging (DTI) and diffusion tensor tractography (DTT) are two techniques that can measure white matter integrity of the spinal cord. Recently, DTI indices have been shown to change with age. The purpose of this study is (a) to evaluate the maturational states of the entire pediatric spinal cord using DTI and DTT indices including fractional anisotropy (FA), mean diffusivity (MD), mean length of white matter fiber tracts and tract density and (b) to analyze the DTI and DTT parameters along the entire spinal cord as a function of spinal cord levels and age.

Method: A total of 23 typically developing (TD) pediatric subjects ranging in age from 6 to 16 years old (11.94 ± 3.26 mean ± standard deviation, 13 females and 10 males) were recruited, and scanned using 3.0 T MR scanner. Reduced FOV diffusion tensor images were acquired axially in the same anatomical location prescribed for the T2-weighted images to cover the entire spinal cord (C1-mid L1 levels). To mitigate motion induced artifacts, diffusion directional images were aligned with the reference image (b0) using a rigid body registration algorithm performed by in-house software developed in Matlab (MathWorks, Natick, Massachusetts). Diffusion tensor maps (FA and MD) and streamline deterministic tractography were then generated from the motion corrected DTI dataset. DTI and DTT parameters were calculated by using ROIs drawn to encapsulate the whole cord along the entire spinal cord by an independent board certified neuroradiologist. These indices then were compared between two age groups (age group A = 6–11 years (n = 11) and age group B = 12–16 years (n = 12)) based on similar standards and age definitions used for reporting spinal cord injury in the pediatric population. Standard least squared linear regression based on a restricted maximum likelihood (REML) method was used to evaluate the relationship between age and DTI and DTT parameters.

Results: An increase in FA (group A = 0.42 ± 0.097, group B = 0.49 ± 0.116), white matter tract density (group A = 368.01 ± 236.88, group B = 440.13 ± 245.24) and mean length of fiber tracts (group A = 48.16 ± 20.48 mm, group B = 60.28 ± 23.87 mm) and a decrease in MD (group A = 1.06 ± 0.23 × 10^-3 mm^2/s, group B = 0.82 ± 0.24 × 10^-3 mm^2/s) were observed with age along the entire spinal cord. Statistically significant increases have been shown in FA (p = 0.004, R^2 = 0.57), tract density (p = 0.0004, R^2 = 0.58), mean length of fiber tracts (p < 0.001, R^2 = 0.51) and a significant decrease has been shown in MD (p = 0.002, R^2 = 0.59) between group A and group B. Also, it has been shown DTI and DTT parameters vary along the spinal cord as a function of intervertebral disk and mid-vertebral body level.

Conclusion: This study provides an initial understanding of age related changes of DTI values as well as DTT metrics of the spinal cord. The results show significant differences in DTI and DTT parameters which may result from decreasing water content, myelination of fiber tracts, and the thickening diameter of fiber tracts during the...
1. Introduction

Conventional diagnostic techniques such as magnetic resonance imaging (MRI) have been used to demonstrate significant microstructural changes of neural tissue during the development and maturation of children through their adolescence. Reports have shown that through childhood and adolescence there is a global increase in brain white matter (WM) density as well as a decline in gray matter density (Giedd et al., 1999; Giorgio et al., 2010; Giorgio et al., 2008). In recent years, techniques like diffusion tensor imaging (DTI), a powerful quantitative tool used to measure the diffusivity of water molecules in tissue (Mulcahey et al., 2013; Beaulieu, 2002; Maier and Mamata, 2005; Stiles and Jernigan, 2016; Hüppi and Dubois, 2006) as well as diffusion tensor tractography (DTT) (Kaneko et al., 2013), a visual and quantitative whiter matter fiber tracking technique, have been used to investigate these changes. These increases/decreases in tissue density have been found to strongly correlate to age-related increases/decreases in DTI and DTT metrics (Hüppi and Dubois, 2006; Kaneko et al., 2013; Izbudak et al., 2015; Singhi et al., 2012); fractional anisotropy (FA), apparent diffusion coefficient (MD), tract density and length of fiber tracts (average Euclidean length in mm for all streamlines within a given segment) in the brain. These findings are highly supportive of the hypothesis that age-related maturation and development continue through to adulthood.

DTI and DTT have been used in past studies to investigate the WM and its axonal projections in the brain. Reports have shown the reliability of DTI and DTT in the analysis of pediatric epilepsy (Helen et al., 2014), severity of WM damage in the brain following stroke (Iannetti et al., 2011), and the maturation of brain white matter tracts through typical development (Cancelliere et al., 2013). However, to the best of our knowledge, little has been reported on the white matter changes that occur in the spinal cord, especially with pediatric subjects. This is a result of the smaller size of the spinal cord compared to the brain and the challenges associate with diffusion imaging of the cord. Presence of anatomical structures such as muscle and bone that are in close proximity to the spinal cord increase image artifacts due to susceptibility changes (Barakat et al., 2012). Furthermore, image degradation can occur as a result of physiological and patient motion. Typically, gating strategies are used to mitigate motion effects from patient movement, CSF pulsation, cardiac, and respiratory cycles. These gating techniques increase acquisition time, which poses a problem when scanning children (Iannetti et al., 2011; Cancelliere et al., 2013; Barakat et al., 2012). In pediatric cases, these problems are more pronounced due to the smaller overall size of the cord compared to adults. In the last several years, development of new pulse sequences like reduced due to the smaller overall size of the cord compared to adults. In the last several years, development of new pulse sequences like reduced due to the smaller overall size of the cord compared to adults. In the last several years, development of new pulse sequences like reduced due to the smaller overall size of the cord compared to adults. In the last several years, development of new pulse sequences like reduced due to the smaller overall size of the cord compared to adults. In the last several years, development of new pulse sequences like reduced due to the smaller overall size of the cord compared to adults. In the last several years, development of new pulse sequences like reduced due to the smaller overall size of the cord compared to adults. In the last several years, development of new pulse sequences like reduced due to the smaller overall size of the cord compared to adults.
In order to keep the acquisition time to a minimum neither gating nor anesthesia were used during scanning.

3. Pre-processing of data

3.1. Motion correction

While the implementation of the 2DRF diffusion sequence helped to mitigate image contamination due to patient movement, and some cardiac and respiratory motion within the cervical cord, motion correction was performed to further enhance tensor estimation accuracy (Middleton et al., 2014). The diffusion data sets were corrected for motion induced artifacts based on 3D registration technique using an in-house software developed in Matlab (MathWorks, Natick, Massachusetts) (Saksena et al., 2016; Middleton et al., 2014). The technique uses rigid body transformation with 6 degrees of freedom (3 translational, 3 axis rotations) paired with normalized mutual information as cost function to align target images (20 diffusion directional images) with the reference image (b0).

3.2. Diffusion tensor estimation

Using the preprocessed data, DTI maps such as FA and MD were produced using robust estimation of tensor by outlier rejection (RESTORE). This RESTORE technique takes the reweighted least-squares regression to determine and exclude any existing signal outliers caused by physiologically generated artifacts (Chang et al., 2005).

4. Generation of fiber tracts

Following DTI image acquisition and pre-processing of the data, DTT images were generated using the deterministic fiber tracking approach (Fig. 1 and Fig. 2). Neural fiber seed and termination points were defined using the DTI estimated FA parameter map as a threshold value to incorporate regions of high diffusion directionality (WM). The FA threshold was set to 40–50% below the averaged FA value (0.18–0.23) across all subjects. Fiber assignments of deterministic tractography was performed at each of the seed points (voxel) belonging to the WM (with FA greater than the pre-defined threshold) of every slice to generate whole spinal cord tracts followed by propagation along the dominant diffusion orientation by setting the length and step size of the fiber trajectories. A tract is terminated when it reaches a voxel with FA values lower than the predefined threshold (0.18–0.23) as well as an angular constraint for turning fiber orientation (i.e., 70°) (Alizadeh et al., 2017; Knoche et al., 2015). These tracts were then restricted to the user defined ROIs placed at each intervertebral disk level and at the mid-vertebral body level of the cervical and thoracic spinal cord (approximately 40 ROIs).

5. ROI definition

Region of interests (ROIs) were drawn manually on each FA map at each intervertebral disk level and at the mid-vertebral body level of the cervical and thoracic spinal cord to compute DTI and DTT measures. These ROIs were anatomically localized by an independent board certified neuroradiologist, per subjects using sagittal turbo spin echo (TSE)-T1-weighted scan localized with axial T2-weighted GRE images. Fig. 3 provides an example of ROIs for several positions along the spinal cord.

Manual ROIs were drawn around the gray and white matter of the cord in the FA map. To avoid any partial volume artifacts which occur at the cord/CSF interface, the ROI borders were drawn approximately 1–2 voxels away from the perimeter of the spinal cord (Conklin et al., 2016). The following levels were included: C1, mid-dens, base dens, mid-C2, C2–C3, mid-C3 (upper cervical cord); C3–C4, mid-C4, C4–C5, mid-C5 (middle cervical cord); C5–C6, mid-C6, C6–C7, mid-C7, C7–T1 (lower cervical cord); mid-T1, T1–T2, mid-T2, T2–T3, mid-T3, T3–T4, mid-T4, T4–T5 (upper thoracic cord); mid-T5, T5–T6, mid-T6, T6–T7, mid-T7, T7–T8, mid-T8, T8–T9 (middle thoracic cord); mid-T9, T9–T10, mid-T10, T10–T11, mid-T11, T11–T12, mid-T12, T12–L1, mid-L1 (lower thoracic cord).

Fig. 1. Sagittal reconstruction of FA color maps and MD maps of 2 overlapping slabs (A and B). The cervical and middle thoracic regions (A): FA color map (left), MD (middle) and tractography (right). The lower cervical through lower thoracic (B): FA color map (left), MD (middle) and tractography (left). (For interpretation of the references to color in this figure legend, the reader is referred to the web version of this article.)
5.1. Statistics

Intra-age group and inter-age group statistics were performed at the levels defined in Section VI. The mean and standard deviation of each DTI and DTT measure, including FA, MD, tract density, and length of tracts for every subject along the entire spinal cord at each intervertebral disk level and at the mid-vertebral body level of the cervical and thoracic spinal cord were calculated (Table 1). These indices then were compared between two age groups (age group A = 6–11 years (n = 11) and age group B = 12–16 years (n = 12)) based on standard least squared linear regression model and the restricted maximum likelihood (REML) method.

This model was constructed looking at group differences by assuming ROI level and groups (controls/patients) composition as the fixed effects and subjects as the random effects.

The REML methodology performs maximum likelihood estimation of a restricted likelihood function that does not depend on the fixed-effect parameters. This yields estimates of the variance components that are then used to obtain the estimates of the fixed effects. Estimates of precision are based on estimates of the covariance matrix which modeled as unstructured and repeated based on level. It provides useful estimates, tests, and confidence intervals. To determine statistical significance, a P value of 0.05 was used throughout (Chang et al., 2010).

DTI and DTT parameters were compared between age groups as a function of cord levels (intervertebral disk levels and mid-vertebral body levels) (Fig. 4) and age (Fig. 5). Additionally, eight specific spinal cord regions were defined: cervical, thoracic, upper-cervical (CUP), mid-cervical (CMID), lower-cervical (CLOW), upper-thoracic (TUP), mid-thoracic (TMID), and lower-thoracic (TLOW). The mean values among the upper, middle, and lower cervical and thoracic spinal cord were calculated and corresponding regions were compared between age groups (Table 2).

6. Results

The mean FA values for subjects in group A were 0.42 ± 0.097 while group B produced a mean FA of 0.49 ± 0.116. Statistics revealed that there was a significant increase in FA in the more mature group B.

Fig. 2. Axial color FA, FA, MD and T2W-GRE images for a single subject at different spinal cord levels.
(p = p = 0.004, R² = 0.57) compared to the younger group A. The same was true for white matter tract density and mean length of fiber tracts. Tract density significantly increased (p < 0.001, R² = 0.5) in group B (440.13 ± 245.24) compared to group A (368.01 ± 236.88). Mean length of fiber tracts significantly increased (p < 0.0001, R² = 0.58) in group B (61.56 ± 35.65 mm) compared to group A (48.16 ± 20.48 mm). MD was the only parameter to show a significant decrease (p = 0.002, R² = 0.59). The older group B had an MD = 1.06 ± 0.23 × 10⁻³ mm²/s while group A’s MD = 0.82 ± 0.24 × 10⁻³ mm²/s.

Additionally, average DTI and DTT parameters of each age group were plotted as a function of individual cord levels. Fig. 4 shows a significant increase in mean FA, tract density, and mean length of tract of group B compared to group A. Also, there was a significant decrease in mean MD of group B compared to group A.

Next, DTI and DTT parameters were plotted against age using a simple linear regression analysis. As seen in Fig. 5, there was an increase in FA, tract density, and mean length of tract while MD decreased as age increased. Also, the size of ROIs as a function of cord levels were plotted in Fig. 6. The Pearson correlation coefficient were used to correlate DTI and DTT measures with size of cord. The results reveal strong correlation between spinal cord size and DTI and DTT indices (FA [r = 0.99], MD [r = −0.7], tract density [r = 0.98] and length of tract [r = 0.55]).

To study the changes in DTI and DTT parameters among different levels of the cervical and thoracic spinal cord, we averaged DTI and DTT values across the following regions per subject: CUP, CMID, CLOW, TUP, TMID and TLOW. The mean values were used for comparison among the upper, middle, and lower cervical and thoracic spinal cord across all subjects. Table 2 shows the p-value for each subregion. The significant values are shown in bold.

7. Discussion

Previous studies in adult’s cervical spinal cord DTI have been shown decrease in FA and increase in MD as a function of age (Chan et al., 2015; Vedantam et al., 2013; Wang et al., 2014) while in pediatric, FA increases and MD decrease. The observed increase in mean FA may be a result of increased diameter of the fiber tracts following the increase in axonal membranes and myelin density during the process of maturation. The decrease in MD is likely tied to the decrease in extracellular space and water content (Saksena et al., 2016). To the best of our knowledge, this is first study looking for age related DTI and DTT changes along cervical and thoracic spinal cord. The older group B, ages 12–16, revealed a significant increase in mean FA and decrease in mean MD compared to the younger group A, ages 6–11. As an extension of the brain, the spinal cord is likely to undergo similar maturational processes. The observed increase in mean FA may be a result of increased diameter of the fiber tracts following the increase in axonal membranes and myelin density. The decrease in mean MD is likely tied to the decrease in extracellular space and water content (Saksena et al., 2016). Previous studies in brain shows myelination is a major contributor and is likely accompanied by increased axonal packing and decreasing water content (Lebel and Deoni, 2018).

To gain further quantitative and empirical insight, DTT has become a supplementary tool to DTI. Along with DTI parameters FA and MD, DTT produces informing measures such as tract density and length of fiber tracts. Few reports on the use of fiber tracking in the typically developing pediatric spinal cord have been published. Most studies that have been done have used healthy pediatric controls as comparisons to abnormal spinal cord conditions. For example, fiber tracking has been used to investigate the inflammatory cord in patients with myelitis (Renoux et al., 2006), as a predictor of the resectability of spinal cord

![Fig. 3. An example of ROIs for several positions along the spinal cord.](image-url)
tumors (Setzer et al., 2010), and the severity of spinal cord injuries (Chang et al., 2010; Alizadeh et al., 2017; Facon et al., 2005). In all these cases, healthy pediatrics were used as a comparison to the abnormal patients. A major shortcoming of these studies is the failure to consider natural age related variation in control groups. What is considered typically developing in an eight year old child may not be considered healthy for an adolescent. To test for this natural variation, length of fiber tracts and tract density were calculated by use of deterministic fiber tracking. Analysis of the data revealed significant increases in both tract density and length of tracts.

Significant increases and decreases in both DTI and DTT metrics is a clear indicator of biological development from childhood through adolescence. Giedd et al. conducted a study on 145 subjects ranging from ages 4–20 years. They found a steady 12% significant increase in brain white matter volume as the subjects age (Giedd et al., 1999). Paus et al found that in the cortico-spinal tract and left arcuate fasciculus, white matter volume significantly increases between ages 4 and 17 (Paus, 2010). Another relevant study looked at the age-related changes in axonal diameter from childhood to adulthood. They observed that the largest axons of the brain, located in the internal capsule, increase
from 1 μm at birth, to 12 μm at age 7 and finally up to 24 μm in adulthood. This increase in axonal diameter corresponds to the increase in axonal caliber and myelin sheath. As part of the CNS, we can hypothesize that these biological changes occur in the spinal cord as well (Lassek, 1942; Verhaart, 1950). Therefore, in the older aged group B of our study, the significant increases in FA, tract density, length of tracts and decrease in MD is likely a cause of an increase in white matter volume and axon diameter.

Table 2
the p-value calculated for regional DTI and DTT parameters.

<table>
<thead>
<tr>
<th>Value</th>
<th>FA</th>
<th>MD</th>
<th>Tract Density</th>
<th>Length of Tract</th>
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<tr>
<td></td>
<td>CervicalA vs CervicalB:</td>
<td>ThoracicA vs ThoracicB:</td>
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<td>P &lt; 0.001</td>
</tr>
<tr>
<td></td>
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<td>CMIDA vs CMIDB:</td>
<td>P = 0.97</td>
<td>P = 0.29</td>
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<tr>
<td></td>
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<td>P = 0.77</td>
<td>TUPA vs TUPB:</td>
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<tr>
<td></td>
<td>TMDA vs TMIDB:</td>
<td>P = 0.46</td>
<td>TLOWA vs TLOWB:</td>
<td>P = 0.19</td>
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<tr>
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<td>P = 0.034</td>
<td>TLOWA vs TLOWB:</td>
<td>P = 0.03</td>
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<tr>
<td></td>
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<td>P = 0.17</td>
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<td></td>
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<tr>
<td></td>
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<td>TLOWA vs TLOWB:</td>
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<td>TLOWA vs TLOWB:</td>
</tr>
<tr>
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<td>P = 0.008</td>
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<td>P &lt; 0.001</td>
<td>TLOWA vs TLOWB:</td>
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<tr>
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<tr>
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<td>TLOWA vs TLOWB:</td>
<td>P = 0.03</td>
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<tr>
<td></td>
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<td>CMIDA vs CMIDB:</td>
<td>P = 0.89</td>
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<td>P = 0.01</td>
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<td>TLOWA vs TLOWB:</td>
<td>P = 0.03</td>
</tr>
</tbody>
</table>

(105x52) Size of ROIs manually drown at each intervertebral disk levels and mid-vertebral body levels. The error bars represent the standard deviation.
sectional size and shape, and active research is being conducted by several groups around the world in this field (Prados et al., 2017). Currently, no automatic methods exist for delineating WM and GM from DTI spinal cord images along cervical and thoracic spinal cord in pediatric. Therefore, in future, an automated or semi-automated segmentation method is required for accurately delineating the GM and WM in the spinal cord. There is also the need to acquire high resolution DTI at a higher field strength, with improved radiofrequency coils and multiband DTI techniques, which will allow imaging small voxels while still maintaining a relatively short imaging time.

The fiber tracts of the cord were generated using a deterministic streamline approach. This method has proven to be an effective algorithm capable of estimating the trajectories of white matter tracts in the brain and the cord (Chang et al., 2010; Alizadeh et al., 2017; Mori and Van Zijl, 2002). However, deterministic fiber tracking has limitations, particularly in voxels where fibers are crossing, bending or kissing. The spinal cord has various ascending and descending tracts in close proximity to each other that turn and cross at different levels. With a voxel size of 0.8 × 0.8 × 6.0 mm, it is possible that a voxel will contain fibers of different orientations. The origin of this crossing fibers problem lies in the fact that DTI requires a relatively low number of diffusion weighted directions, causing the tract to terminate or be generated inaccurately. One way to resolve this problem is to use other diffusion imaging techniques that use a higher number of gradient directions such as High Angular Resolution Diffusion Imaging (HARDI) or Neurite Orientation Dispersion and Density Imaging (NODDI) (Zhang et al., 2012; Berman et al., 2013) in conjunction with probabilistic fiber tracking. Probabilistic fiber tracking evaluates all possible propagation directions to generate the neural tracts. With more gradient directions and therefore a higher angular resolution, more accurate production of fibers is possible. However, diffusion weighted images with such high b-value have low SNR but due to the increased contrast-to-noise ratio in the angular domain they provide better delineation of crossing fibers (Tournier et al., 2012). Previous studies in the brain suggest that the optimal b-value to use for estimation of fiber orientations in crossing fiber voxel is ~2500–3000 s/mm², although good results can still be obtained using lower b-value (~1000 s/mm²) (Tournier et al., 2012; Tournier et al., 2007). The optimal number of diffusion weighted directions is still unresolved and depends of different factors such as the robustness of the post processing algorithms, voxel size, available acquisition time and scan parameters (e.g., TE) (Tournier et al., 2012; Tournier et al., 2007).

DTT and DTI have been under explored for the evaluation of the pediatric spinal cord because of the relatively small size of the spinal cord and the motion artifacts induced by CSF pulsation, cardiac and respiration (Barakat et al., 2012; Saksema et al., 2016). However, development of newer pulse sequence methods such as reduced FOV has enabled reliable DTI collection and enabled further exploration of the spinal cord. The bulk physiologic motion about the cord, due to CSF pulsation, cardiac and respiration or due to swallowing hamper quality of diffusion weighted images. Some mitigation can be achievable with cardiac gating (Andre and Bammer, 2010). Cardiac gating was not used in this study. The lowest cervical levels, thoracic and thoracolumbar are the most sensitive to cardiac motion. Therefore, some cardiac related artifacts may have biased DTT and DTI measures. While this may be considered a limitation in this study, it is important to note that gating increases acquisition time and does not completely prevent subject motion. However, the absence of triggering might be linked to the relatively poor WM/GM contrast on the DTI data. Keeping the acquisition time relatively short was the priority for imaging in the pediatric population. This limitation could be overcome in the future by imaging the spinal cord using parallel or multiband DTI techniques, which will allow faster imaging.

8. Conclusion

DTI and DTT have proven to be effective techniques in the investigation and quantification of neurodevelopmental changes in pediatric subjects. However, few imaging studies have been conducted on the maturation of the pediatric spinal cord. In this study, these diffusion and fiber tracking techniques were successfully utilized to detect significant age-related changes of the spinal cord. This normative data and analysis could potentially help facilitate a more thorough understanding of the maturation process in pediatrics and serve as a basis for the detection of abnormal spinal conditions.

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