

Reducing Distortion in DTI Using SENSE: An Experimental Study with 13 Datasets

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Synopsis

The statistics of the normalized diffusion tensor field is useful for quantitatively characterizing pathological changes and white-matter-structure variations. In previous studies, co-registered T_1 -weighted images were used to normalize DT images, in which disagreement between the DTI and the T_1 -weighted images caused by EPI-related image distortion may affect the validity of the normalization. In this presentation, we minimized this shortcoming by using SENSE imaging and averaged DTI maps obtained from normal volunteers. T_1 -DTI discrepancy map confirmed a small amount of distortion. Various white matter structures could be clearly appreciated in the average map, suggesting a possibility of quantitative study.

Introduction

In order to perform group analysis, spatial normalization is a powerful quantitative technique. However, with local biological microstructure information encoded, diffusion tensor (DT) fields contain much higher dimensional signals than intensity-based images. Therefore, warping DT fields is more complicated. Several DTI warping algorithms were postulated [1, 2], which can properly reorient tensors through a statistical analysis of local displacement in a small neighborhood. In this study, HAMMER [3], a very high dimensional elastic transformation procedure in 3D volume space was employed for this purpose, which can effectively determine correspondence between a pair of intensity images co-registered with the DT images. In previous studies, we performed spatial normalization of diffusion tensor fields using co-registered T_1 -weighted images. However, validity of the results was affected by the EPI distortion and errors from the mismatches of co-registered T_1 -weighted image and DT image. It would be highly preferable if inter-subject normalization is based solely on DT images. Instead of using the transformation information determined from co-registered T_1 -weighted images, we adopted that directly from the averaged DTI map. Then it becomes crucial to minimize EPI-related image distortion and to know the extent of residual space distortion. In this study, SENSitivity Encoding (SENSE) parallel imaging was used to reduce diffusion weighted image (DWI) distortion. We measured residual discrepancy between T_1 -weighted image and DTI and assessed the quality of the averaged DTI map using data from 13 healthy volunteers.

Methods

Data acquisition: *In vivo* human data of 13 subjects were acquired using a 1.5 T Philips Gyroscan NT system. A single-shot EPI sequence with SENSE parallel imaging scheme (SENSitivity Encoding, reduction factor $R = 2.5$) was used for DTI data acquisition, with an imaging matrix of 96×96 and a field of view of 240×240 mm (nominal resolution 2.5 mm), which was then zero-filled to 256×256 . Axial slices thickness was 2.5 mm parallel to the anterior–posterior commissure line. A total of 50 to 55 slices covered the entire brain and brainstem leaving no gap. The diffusion weighting was encoded along 30 independent directions and the b-value was $700 \text{ mm}^2/\text{sec}$. Five additional images with minimal diffusion weighting ($b = 33 \text{ mm}^2/\text{sec}$) were also acquired. Co-registered magnetization-prepared rapid gradient echo (MPRAGE) images of the same resolution were also recorded for anatomical guidance. **Image warping:** We randomly selected one of the 13 subjects as a template, and subsequently normalized the other 12, using the algorithm presented in [1]. Instead of taking the co-registered T_1 -weighted images for HAMMER [3] to derive a displacement field between a pair of DT images, we used their averaged DWIs, which guaranteed perfect co-registration inherently. By averaging all 13 normalized DT images, we obtained an average DT image, displaying the common aspects of this small population of 13 subjects. In order to demonstrate the difference of using co-registered T_1 -weighted images and using average DWIs, we calculated and normalized their difference images, with the same batch of displacement fields. In a similar averaging process, we obtained the average difference.

Results

Figure 1 is one slice of the color map of one individual's principal direction (PD), and Figure 2 is the same slice of the average result. We showed the colormap by assigning PD's x-component (horizontal) to red, y (vertical) to green and z (perpendicular to screen) to blue. Figure 3 is the average difference between T_1 -weighted images and DWIs with gray matter and white matter assigned to value 250 and all others to 0. Although the brain shapes before and after normalization are very different, the tensors are properly reoriented and thus fiber pathways are preserved.

Discussion

By using SENSE imaging, mismatches between T_1 -weighted images and DTI images have been largely ameliorated. Small disagreement is shown in the front and temporal regions, while most of the brain volumes match well. This demonstrates that the anatomical information revealed by our DT imaging with SENSE is, in general, compatible with that contained in T_1 -weighted images and have less volume distortion. From a comparison between the individual subject in Fig 1 and the average map in Fig. 2, it can be seen that the high order elastic warping could preserve many details of white matter anatomy after normalization. This technique will allow us to quantitatively compare white matter anatomy between subject groups for these structures visible in the averaged maps.

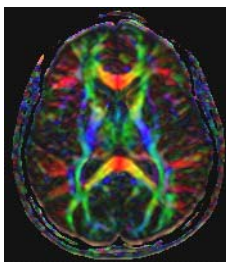


Fig. 1 A slice of one individual's DT colormap

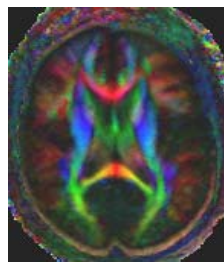


Fig. 2 A slice of the average DT colormap of 13 normalized healthy subjects.

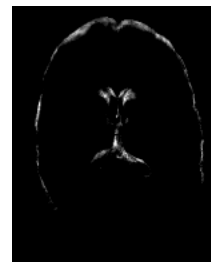


Fig. 3 . Average difference between 13 subjects' DWIs and T_1 -weighted images

Acknowledgement This study was supported by NIH R01 AG20012 and NIH P41 RR15241.

References: [1] D. Xu. et al. *MRM*, 50:175-182, 2003. [2] DK. Jones. et al. *NeuroImage*, 17:592-617, 2002. [3] D. Shen. et al. *IEEE TMI*, 21(11):1421-1439, 2002.