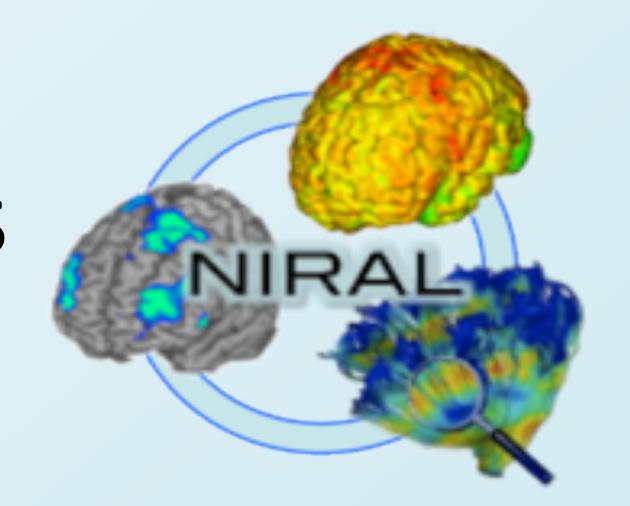


# FADTTSter: Accelerating Hypothesis Testing With Functional Analysis of Diffusion Tensor Tract Statistics

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### INTRODUCTION

Functional Analysis of Diffusion Tensor Tract Statistics (FADTTS) is a powerful statistical tool box developed to outline the evolution of diffusion properties such as axial diffusivity (AD), radial diffusivity (RD), mean diffusivity (MD) and fractional anisotropy (FA) along white matter fiber tracts and their correlation with a set of covariates of interest, such as age or gender.

## METHODS

**Quality Control** 3D Slicer -Convert DICOM to NRRD DTIPrep -Automated Quality Control -Manual Quality Control 3D Slicer -Create DTI from quality controlled DWI -Check FA -Check Color FA -Check Glyphs -Fiducial Tractography **Atlas Creation** bet2 or AutoSeg -Create Brain Mask 3D Slicer -Edit Brain Mask -Create DTI files with Brain Mask Applied DTIAtlasBuilder -Creates an atlas from your sample using affine, & diffeomorphic registrations Tractography 3D Slicer -Label Map Tractography -ROI Fiber Bundle Cleaning FiberViewerLight -Cleaning & Clustering Fiber Bundles **Statistical Analysis** DTIAtlasFiberAnalyzer -Fiber Parameter Profiles **FADTTSter** -Statistical Analysis MergeStatWithFiber Merge statistics along fiber bundles

Figure 1: UNC-Utah NA-MIC DTI frame- work

3D Slicer

Visualize statistics

merged with fiber bundles

The main contribution of this work is enabling researches to drill down in the statistical analysis produced by FADTTS using interactive charts.

Additionally, this tool facilitates setting up FADTTS execution and guides the user through a series of steps including quality control

#### MATERIALS

Images from healthy full-term infants (75 males and 53 females) were taken from a larger study designed to investigate early brain development. All 128 infants were less than one year old at the time of the first imaging session. Using the DTIs from this population of subjects, an atlas is generated with a set of fiber bundles of interest for each subject. These bundles are used in the statistical analysis.

#### RESULTS

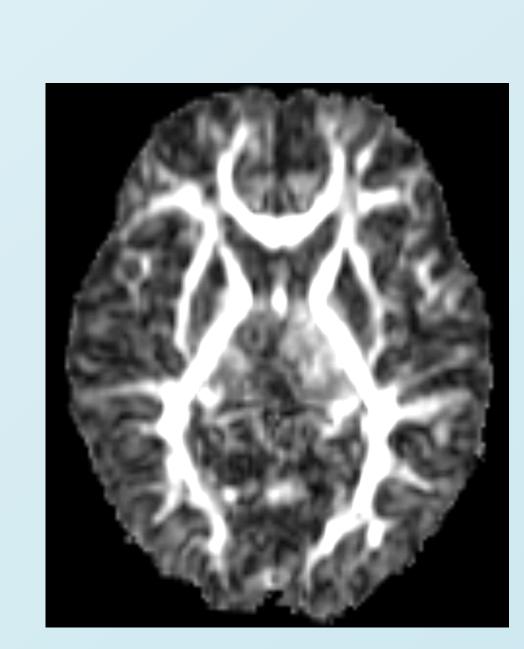


Figure 2: FA image

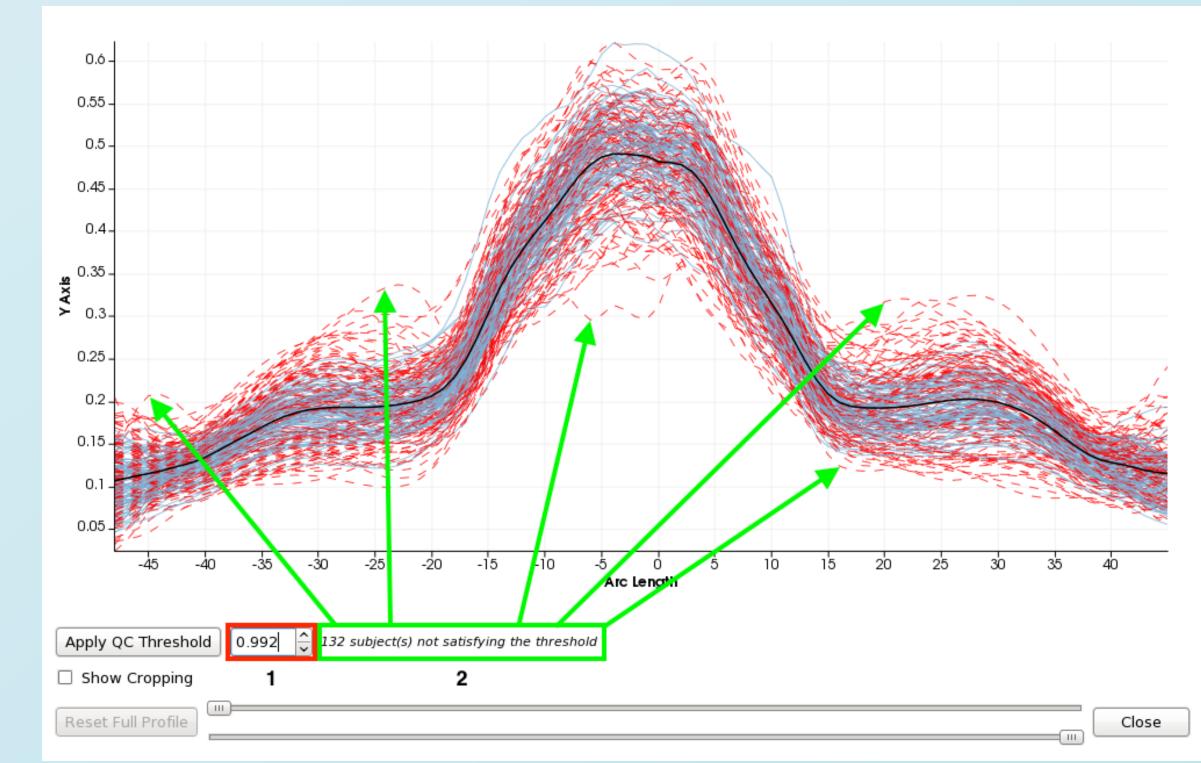


Figure 3: Subject filtering

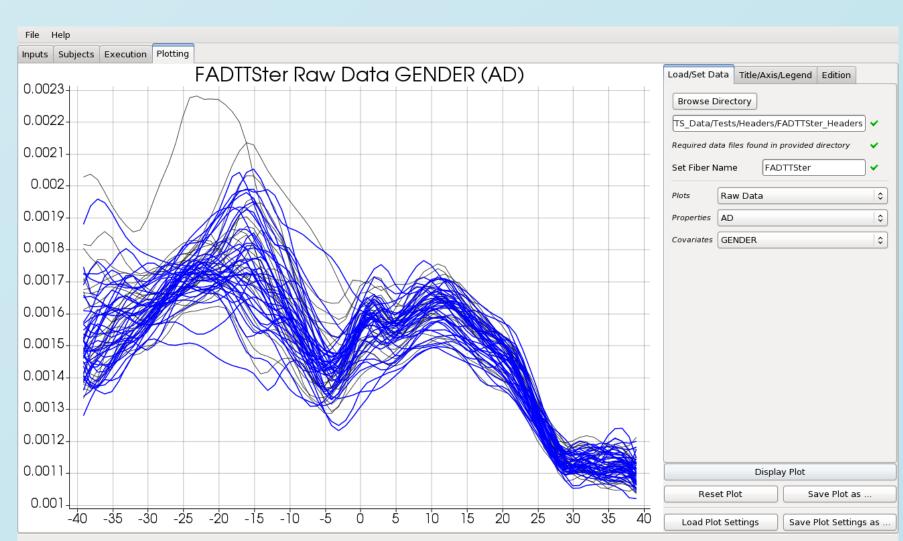


Figure 4: Raw data plot

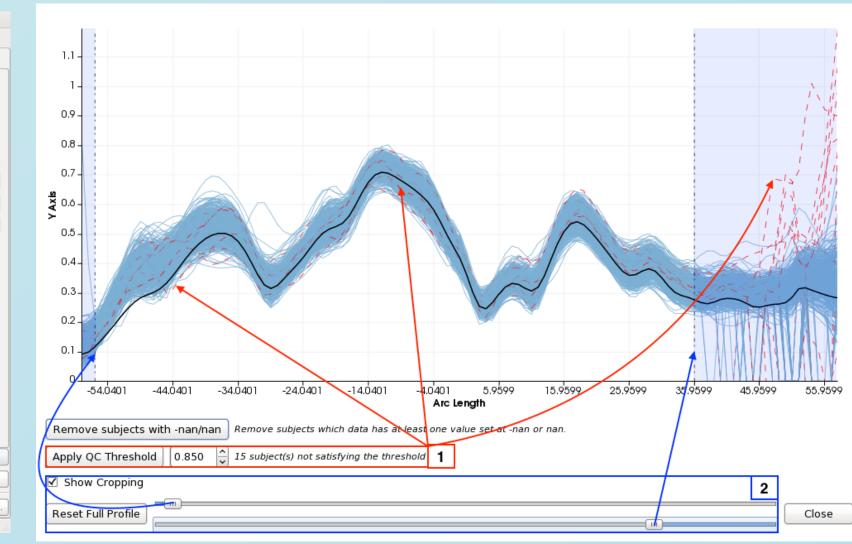


Figure 5: Fiber profile cropping

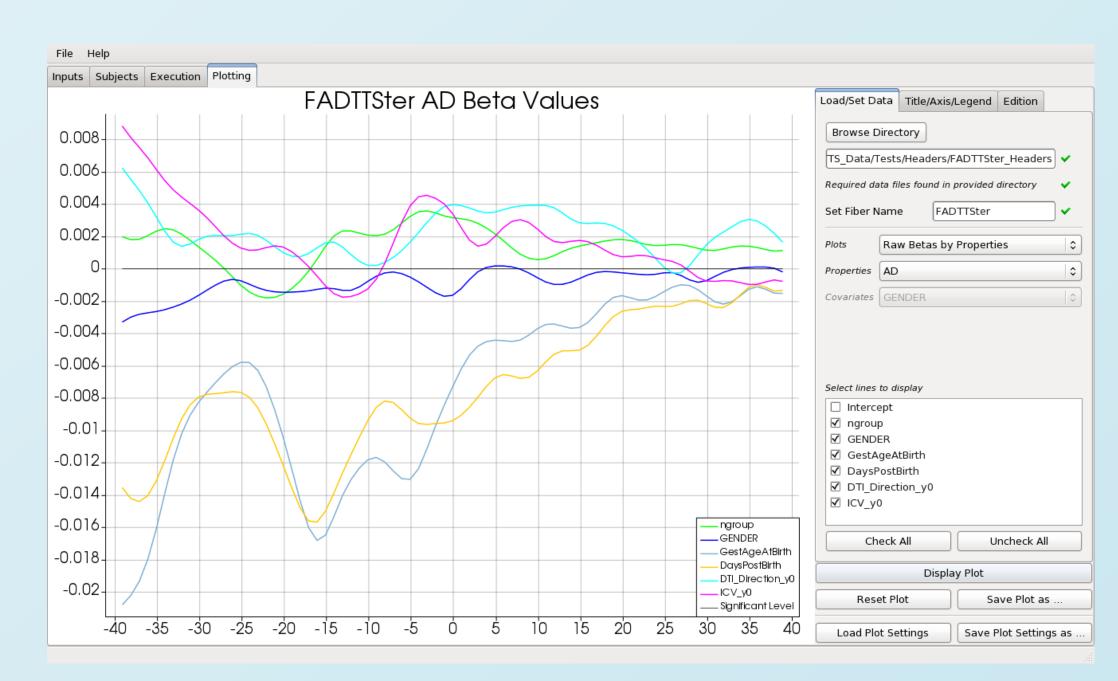


Figure 6: Omnibus FDR significant betas by properties

#### CONCLUSIONS

The first contribution presented in this work is to enable FADTTS to non-technical users. The tool guides users through a series of steps which simplifies setting up FADTTS. FADTTSter is actively being used by researchers at the University of North Carolina.

The second contribution presented here is the new set of interactive plots. FADTTS outputs may be difficult to understand with out having the possibility to drill down on the raw data and statistics.

We have shown several plots that will enhance the researcher's capability to understand the results.

By enabling FADTTS to a broader audience, we seek to accelerate hypothesis testing in neuroimaging studies involving heterogeneous clinical data and diffusion tensor imaging. We expect that this novel tool will lead to new findings in our clinical applications.

#### REFERENCES

[1] Feldman, H. M., Yeatman, J. D., Lee, E. S., Barde, L. H., and Gaman-Bean, S., "Diffusion tensor imaging: a review for pediatric researchers and clinicians," Journal of developmental and behavioral pediatrics: JDBP 31(4), 346 (2010).

[2] Basser, P., Mattiello, J., and Lebihan, D., "MR Diffusion Tensor Spectroscopy and Imaging," Biophysical Journal 66, 247–254 (January 1994).
[3] Basser, P., Mattiello, J., and Lebihan, D., "Estimation of the Effective Self-Diffusion Tensor from the NMR

Spin Echo," Journal of Magnetic Resonance Series B 103, 247–254 (March 1994).

[4] Zhu, H., Kong, L., Li, R., Styner, M., Gerig, G., Lin, W., and Gilmore, J. H., "FADTTS: Functional analysis of diffusion tensor tract statistics," NeuroImage 56, 1412–1425 (June 2011).

[5] Verde, A. R., Budin, F., Berger, J.-B., Gupta, A., Farzinfar, M., Kaiser, A., Ahn, M., Johnson, H., Matsui, J., Hazlett, H. C., Sharma, A., Goodlett, C., Shi, Y., Gouttard, S., Vachet, C., Piven, J., Zhu, H., Gerig, G., and Styner, M., "UNC-Utah NA-NAMIC framework for dti fiber tract analysis," Frontiers in Neuroinformatics 7 (January 2014).

[6] Goodlett, C. B., Fletcher, P. T., Gilmore, J. H., and Gerig, G., "Group analysis of dti fiber tract statistics with application to neurodevelopment," Neuroimage 45(1), S133–S142 (2009).

[7] Joshi, S., Davis, B., Jomier, M., and Gerig, G., "Unbiased diffeomorphic atlas construction for

computational anatomy," NeuroImage 23, S151–S160 (2004). [8] Fletcher, P. T. and Joshi, S., "Riemannian geometry for the statistical analysis of diffusion tensor data," Signal Processing 87(2), 250–262 (2007).

[9] Prieto, J. C., Yang, J. Y., Budin, F., and Styner, M., "Autotract: automatic cleaning and tracking of fibers," in [SPIE Medical Imaging], 978408–978408, International Society for Optics and Photonics (2016).

