

Contextual Enhancements on DW-MRI

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Where innovation starts

Categories and Approach

DTI Category

b-value = 1200, 32 gradients

HARDI Category

b-value = 3000, 64 gradients

Categories and Approach

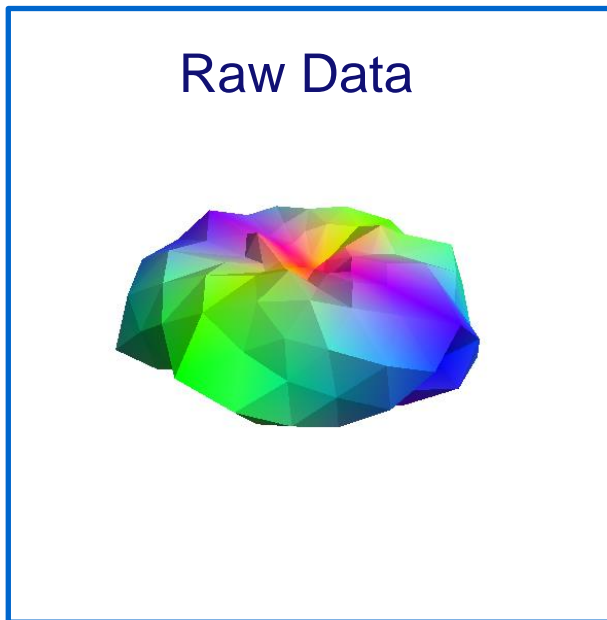
DTI Category

B-value = 1200, 32 gradients

HARDI Category

B-value = 3000, 64 gradients

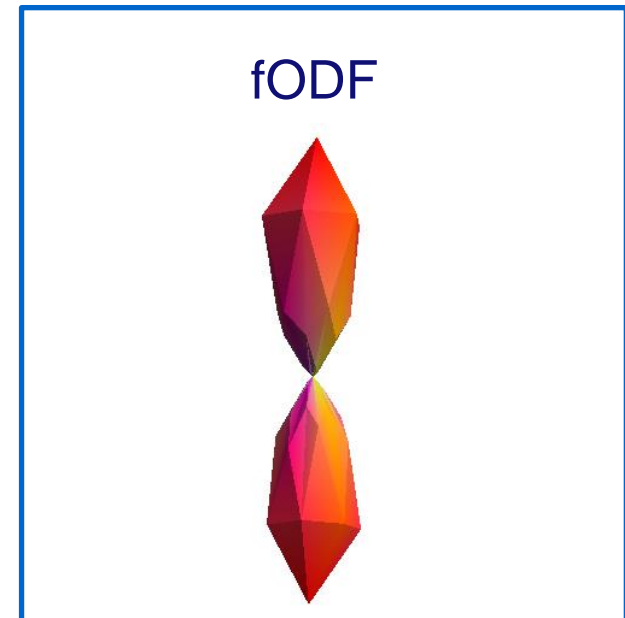
Two-Step Approach



Constrained
Spherical
Deconvolution

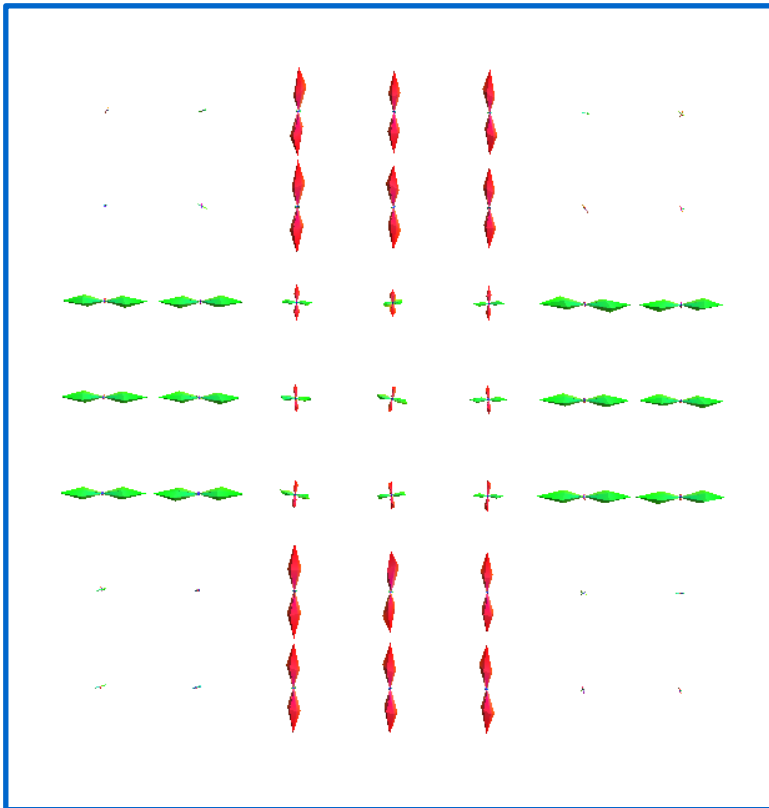


1



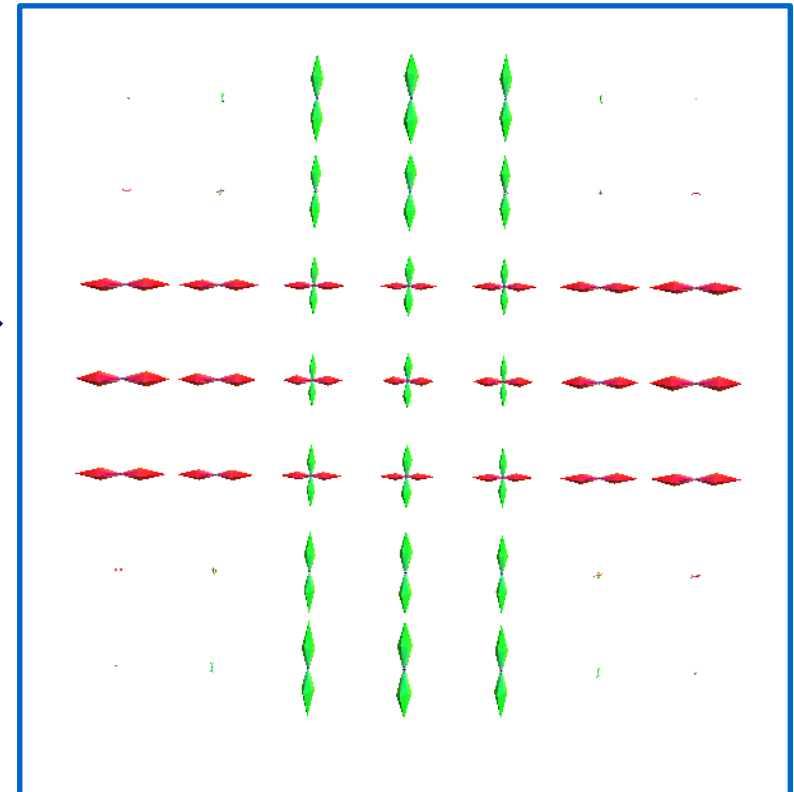
Enhancement of fODF Field

Artificial Fiber Crossing
B-value=1200, 32 gradients



2

Glyphs are aligned with surrounding!



Enhancement of fODF Field

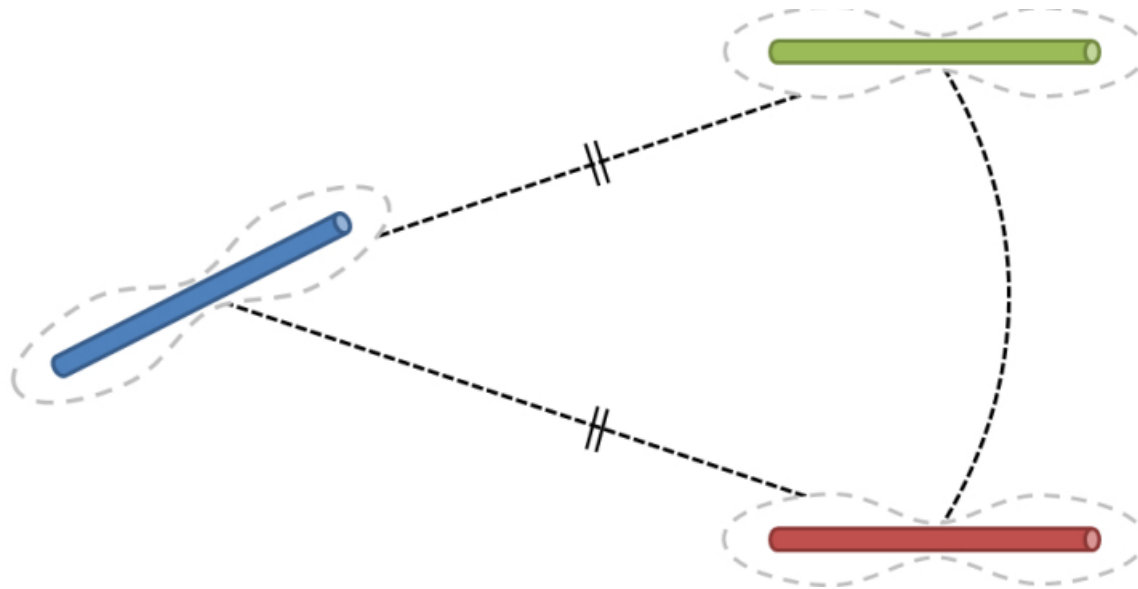
Detection of alignment requires coupling of position and orientation!

$$U : \mathbb{R}^3 \times S^2 \rightarrow \mathbb{R}^+ : (\underline{\mathbf{y}}, \underline{\mathbf{n}}) \rightarrow U(\underline{\mathbf{y}}, \underline{\mathbf{n}})$$

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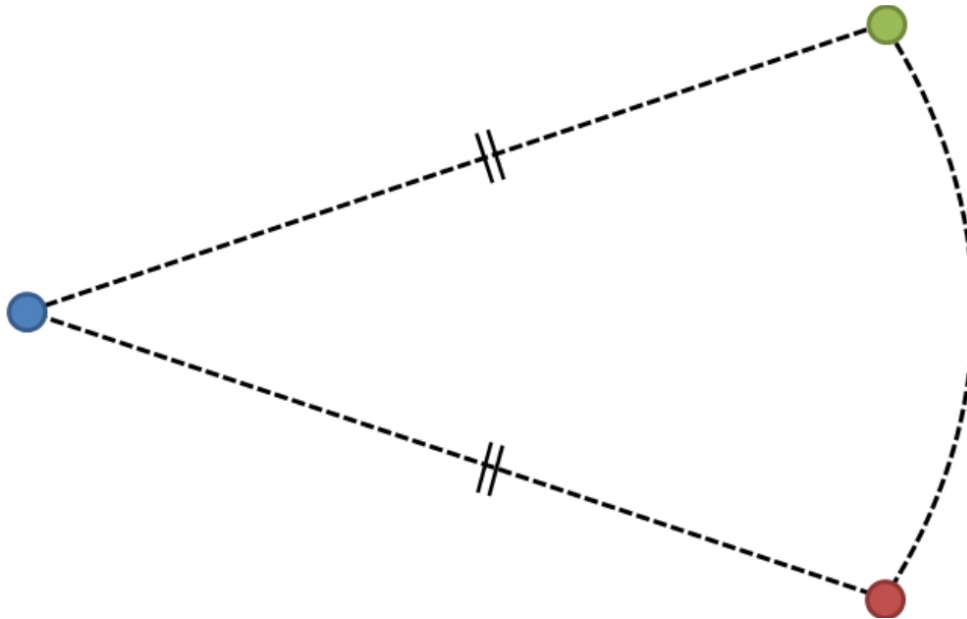
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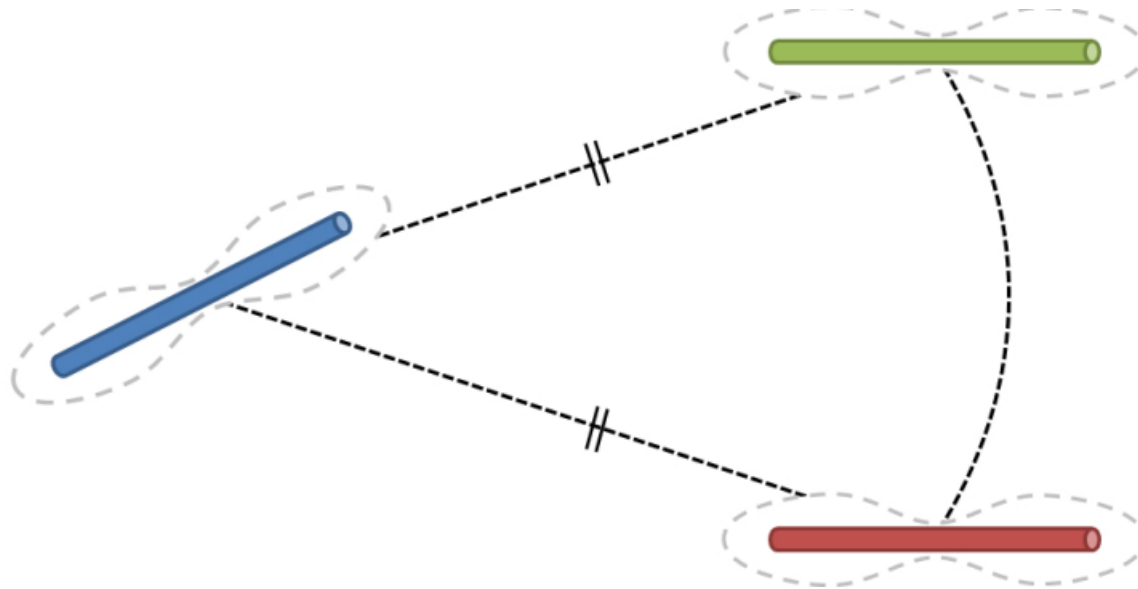
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Enhancement of fODF Field

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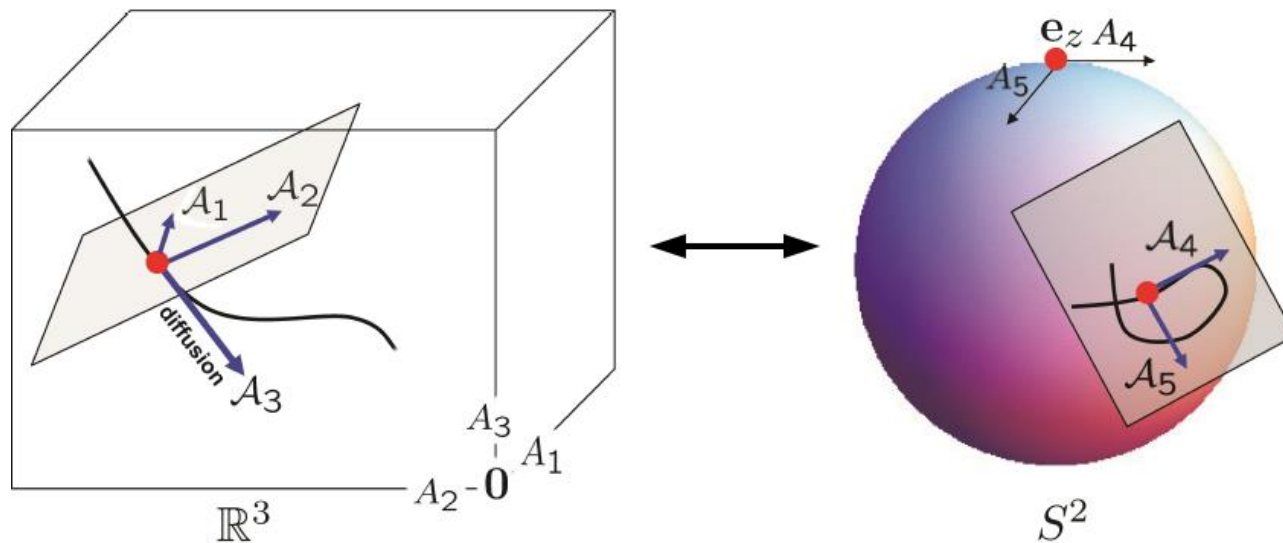
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Enhancement of fODF Field

We define directional derivatives in this 5-dimensional space

$$U : \mathbb{R}^3 \times S^2 \rightarrow \mathbb{R}^+ : (\underline{\mathbf{y}}, \underline{\mathbf{n}}) \rightarrow U(\underline{\mathbf{y}}, \underline{\mathbf{n}})$$

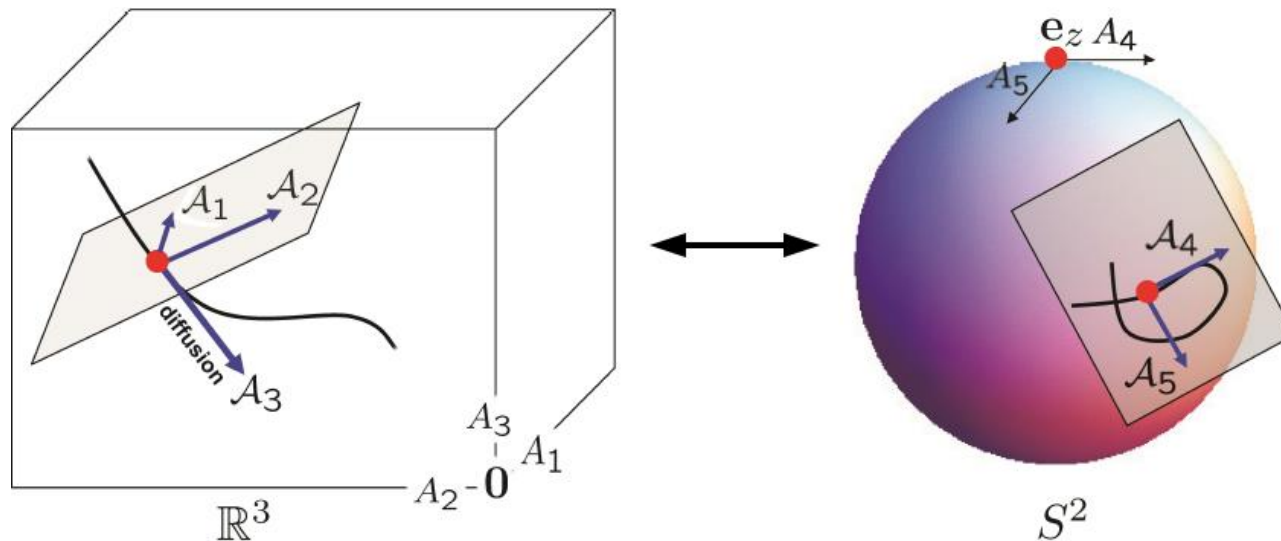


Diffusion Equation

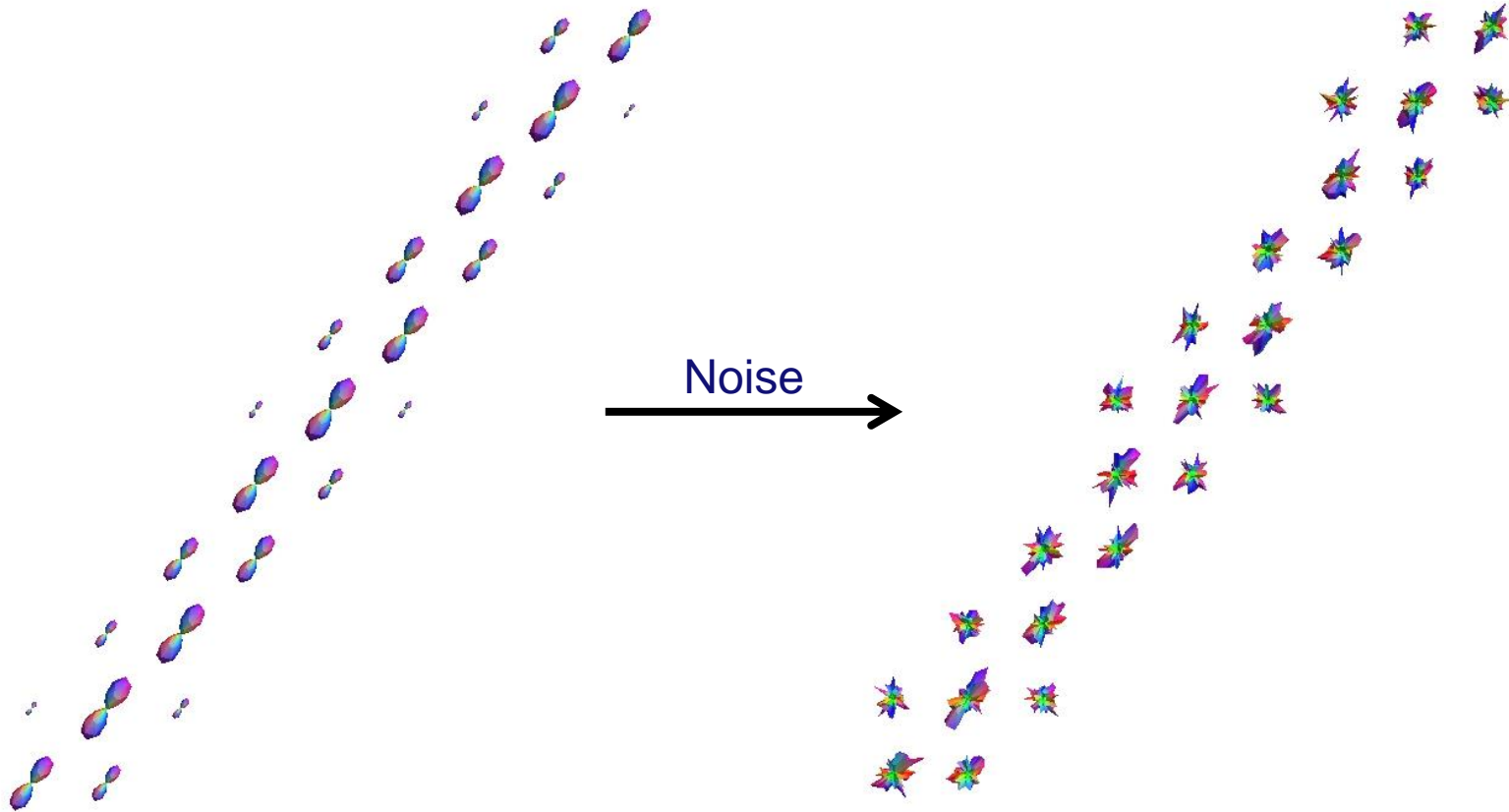
Diffusion equation on $\mathbb{R}^3 \times S^2$

$$\begin{cases} \partial_t W(\mathbf{y}, \mathbf{n}; t) = \left(\alpha (\mathcal{A}_3)^2 + \beta ((\mathcal{A}_4)^2 + (\mathcal{A}_5)^2) \right) W(\mathbf{y}, \mathbf{n}; t) \\ W(\mathbf{y}, \mathbf{n}; 0) = U(\mathbf{y}, \mathbf{n}) \end{cases}$$

α and β set balance between spatial and angular diffusion

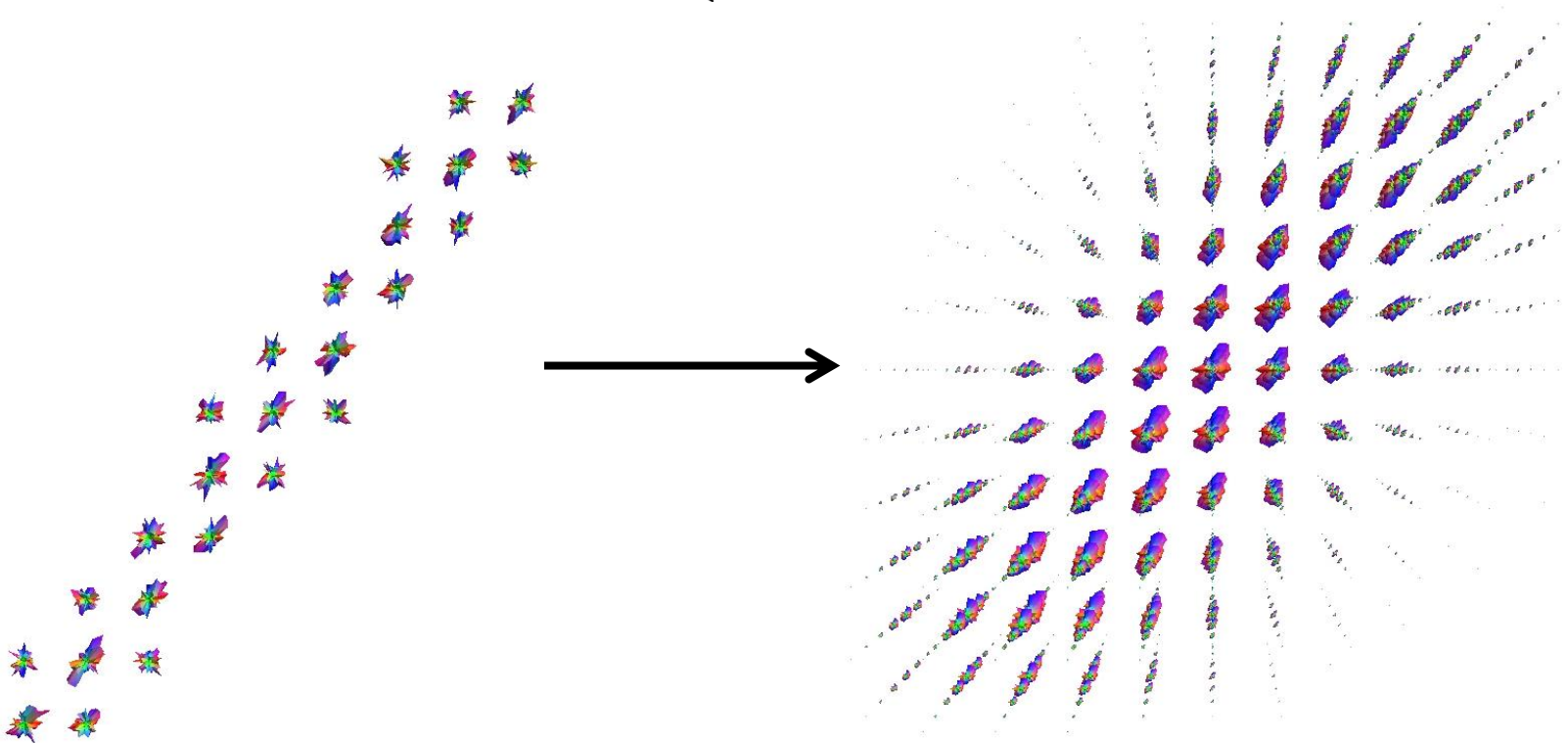


A Simple Example



Isotropic Spatial Diffusion

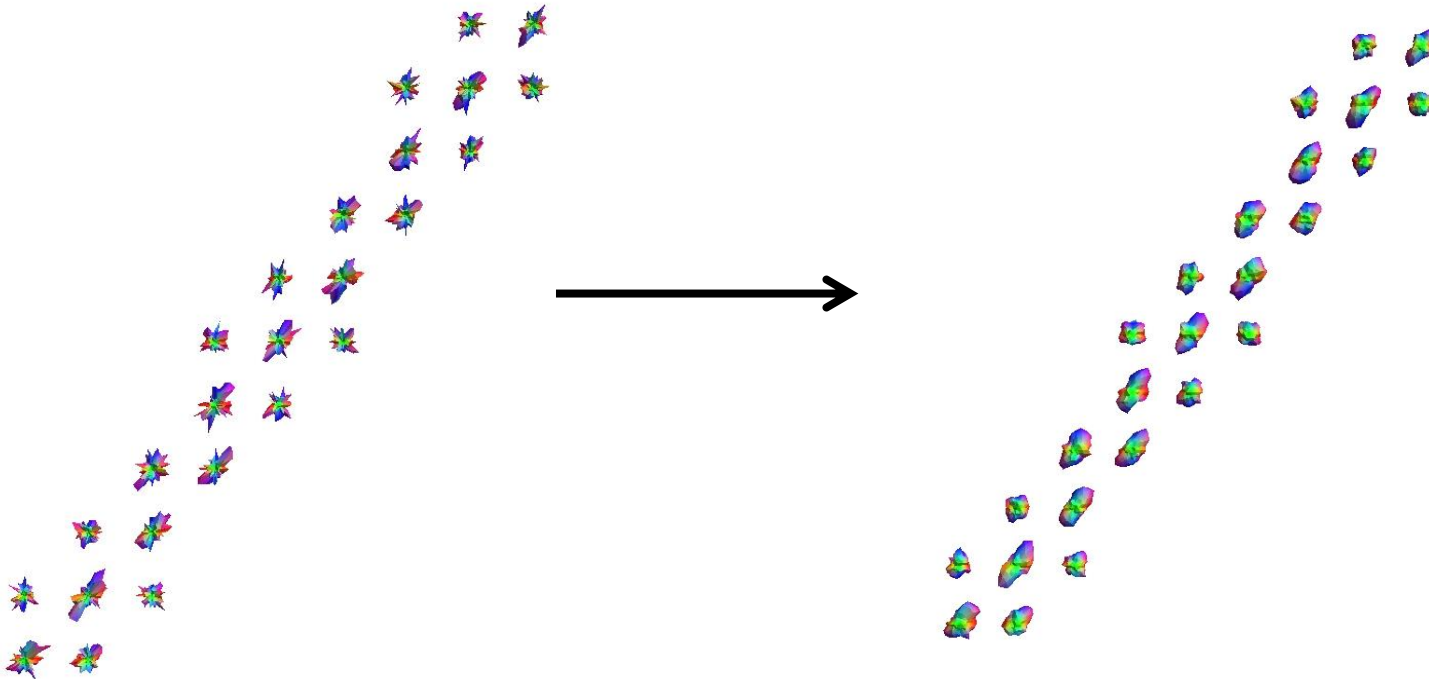
Isotropic Spatial Diffusion:
$$\begin{cases} \partial_t W(\mathbf{y}, \mathbf{n}; t) = ((\mathcal{A}_1)^2 + (\mathcal{A}_2)^2 + (\mathcal{A}_3)^2) W(\mathbf{y}, \mathbf{n}; t) \\ W(\mathbf{y}, \mathbf{n}; 0) = U(\mathbf{y}, \mathbf{n}) \end{cases}$$



Angular Diffusion

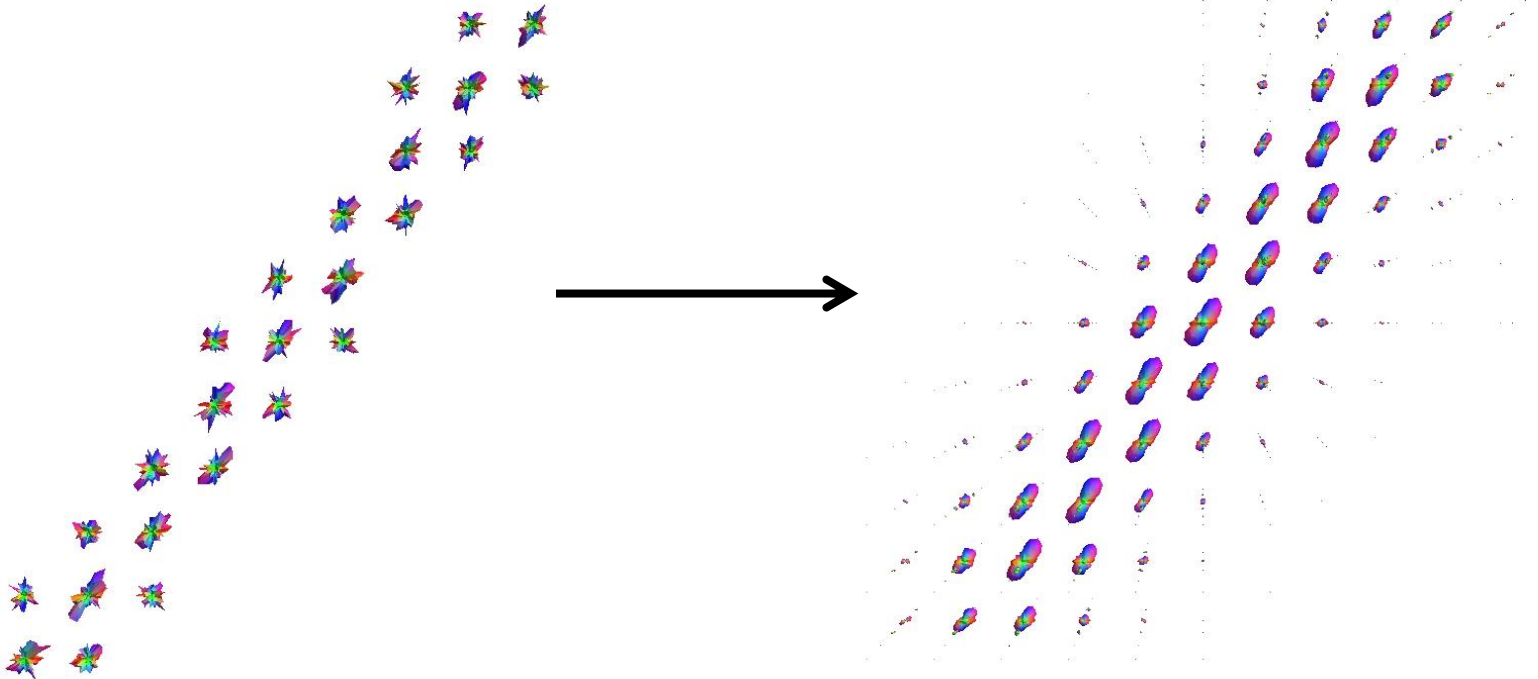
Angular diffusion:

$$\begin{cases} \partial_t W(\mathbf{y}, \mathbf{n}; t) = ((\mathcal{A}_4)^2 + (\mathcal{A}_5)^2) W(\mathbf{y}, \mathbf{n}; t) \\ W(\mathbf{y}, \mathbf{n}; 0) = U(\mathbf{y}, \mathbf{n}) \end{cases}$$



Contour Enhancement

Coupled diffusion:
$$\begin{cases} \partial_t W(\mathbf{y}, \mathbf{n}; t) = \left(\alpha (\mathcal{A}_3)^2 + \beta ((\mathcal{A}_4)^2 + (\mathcal{A}_5)^2) \right) W(\mathbf{y}, \mathbf{n}; t) \\ W(\mathbf{y}, \mathbf{n}; 0) = U(\mathbf{y}, \mathbf{n}) \end{cases}$$



Edge Preserving Diffusion

- Uses a K parameter which lowers α when the changes along the orientation are large
- Effect: Less shooting out of the curve

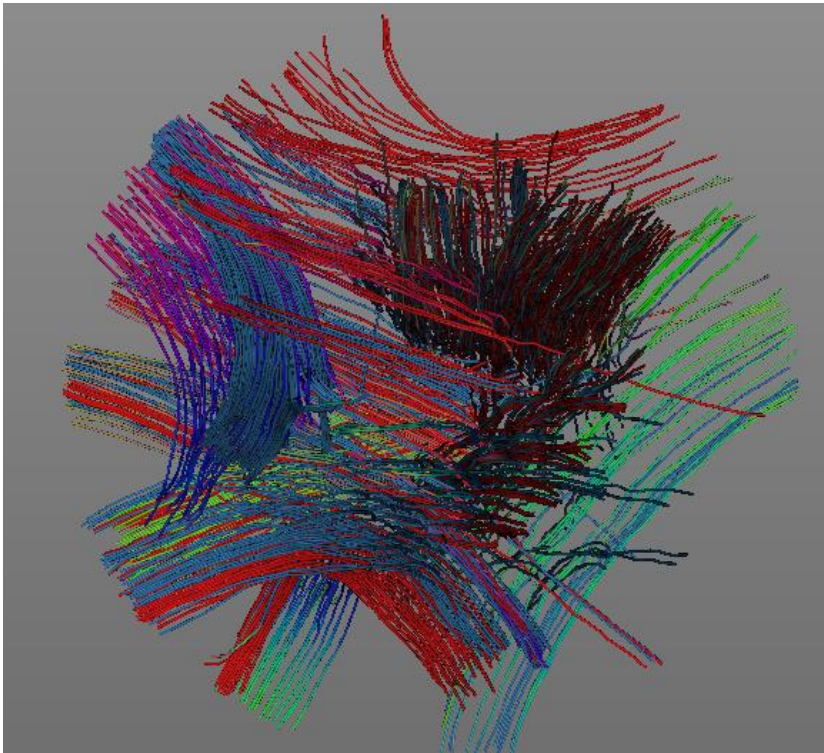
$$\begin{cases} \partial_t W(\mathbf{y}, \mathbf{n}; t) = (\mathcal{A}_3 \tilde{\alpha} \mathcal{A}_3 + \beta ((\mathcal{A}_4)^2 + (\mathcal{A}_5)^2)) W(\mathbf{y}, \mathbf{n}; t) \\ W(\mathbf{y}, \mathbf{n}; 0) = U(\mathbf{y}, \mathbf{n}) \end{cases}$$

$$\tilde{\alpha}(\mathbf{y}, \mathbf{n}) = \alpha e^{-\frac{(\mathcal{A}_3 W(\mathbf{y}, \mathbf{n}; t))^2}{K^2}}$$

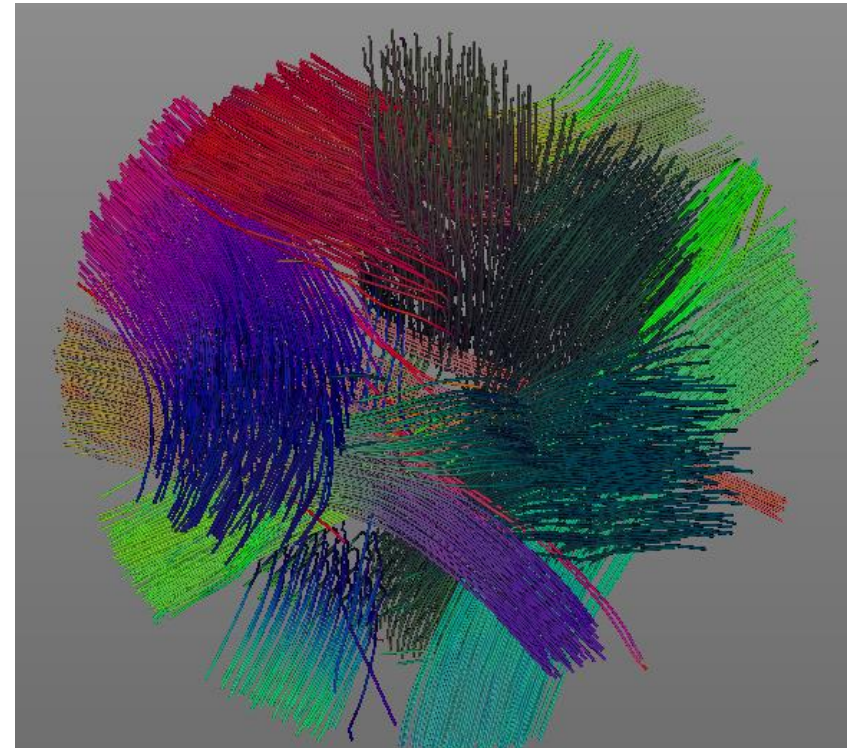
Preliminary Fiber Tracking Results

32 measured orientations, SNR=10

Only CSD



CSD + Enhancements



Summary

We introduce a new contextual enhancement method using a coupled domain of $\mathbb{R}^3 \times S^2$

Aligns fiber fragments according to their surrounding

Allows tracking through crossings when using low number of gradients.

Thank you for your attention

Are there any questions?

- E.J. Creusen, R. Duits, A. Vilanova, L.M.J. Florack, "Numerical Schemes for Linear and Non-linear Enhancement of DW-MRI", *Numer. Math. Theor. Meth. Appl.*, 6, 2013
 - <http://bmia.bmt.tue.nl/people/RDuits/CreusenDuits-NM-TMA2011.pdf>
- R. Duits and E. Franken. "Left-invariant diffusions on the space of positions and orientations and their application to crossing-preserving smoothing of HARDI images." *International Journal of Computer Vision*, 40, 2010.
 - <http://bmia.bmt.tue.nl/people/RDuits/IJCV2010fulltext.pdf>
- R. Duits, T. Dela Haije, E.J. Creusen, A. Ghosh, "Morphological and Linear Scale Spaces on $\mathbb{R}^3 \times \mathbb{S}^2$ for Enhancement of Crossing Fibers in DW-MRI", *Journal of Mathematical Imaging and Vision*
 - <http://bmia.bmt.tue.nl/people/RDuits/JMIVDuits2011final.pdf>
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