

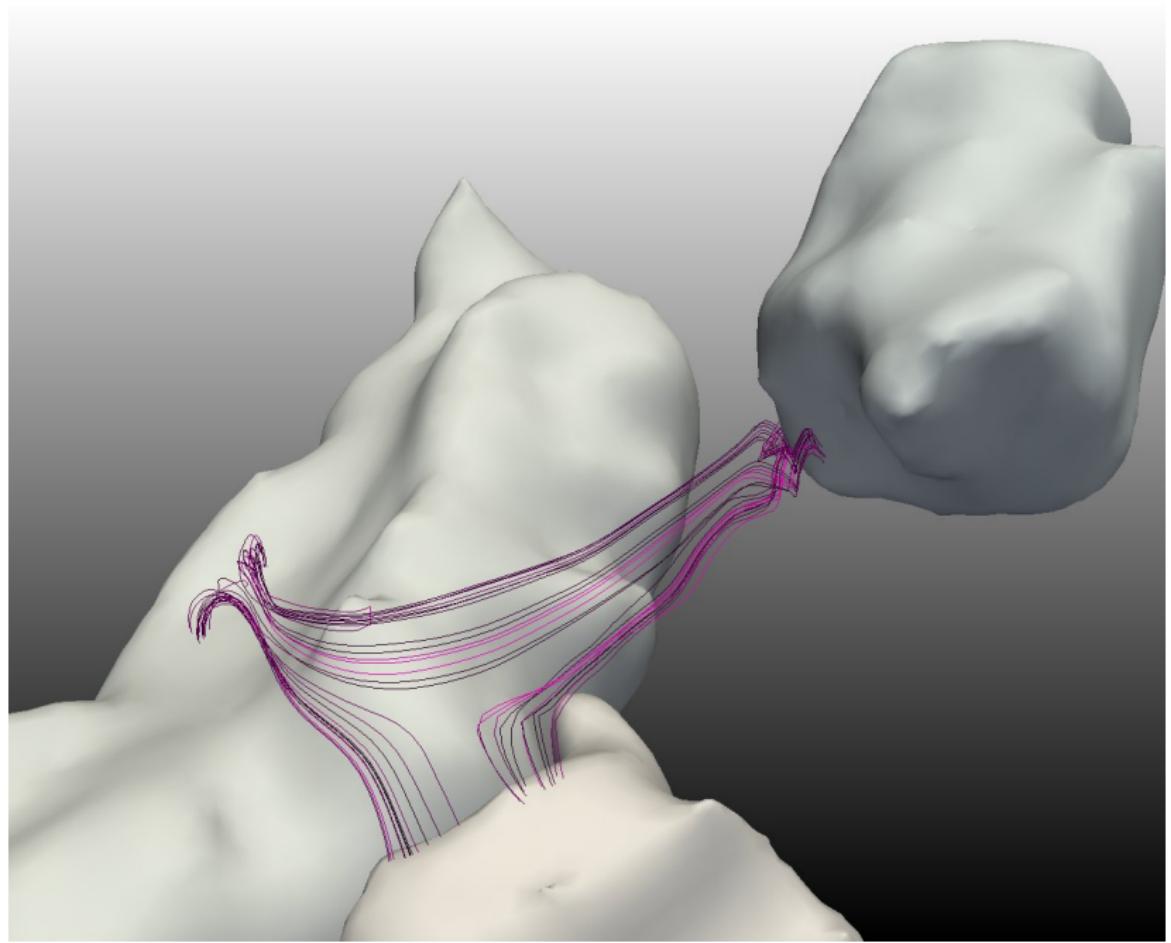
# A Comprehensive Riemannian Framework for the Analysis of White Matter Fiber Tracts

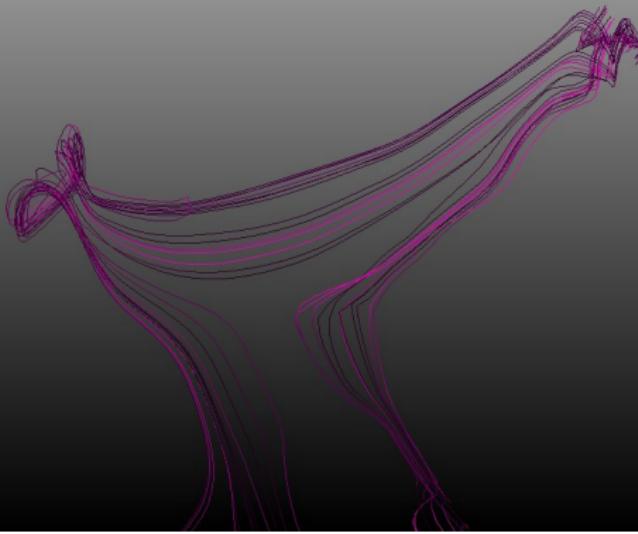
Meena Mani<sup>1</sup> Sebastian Kurtek<sup>2</sup>  
Christian Barillot<sup>1</sup> Anuj Srivastava<sup>2</sup>

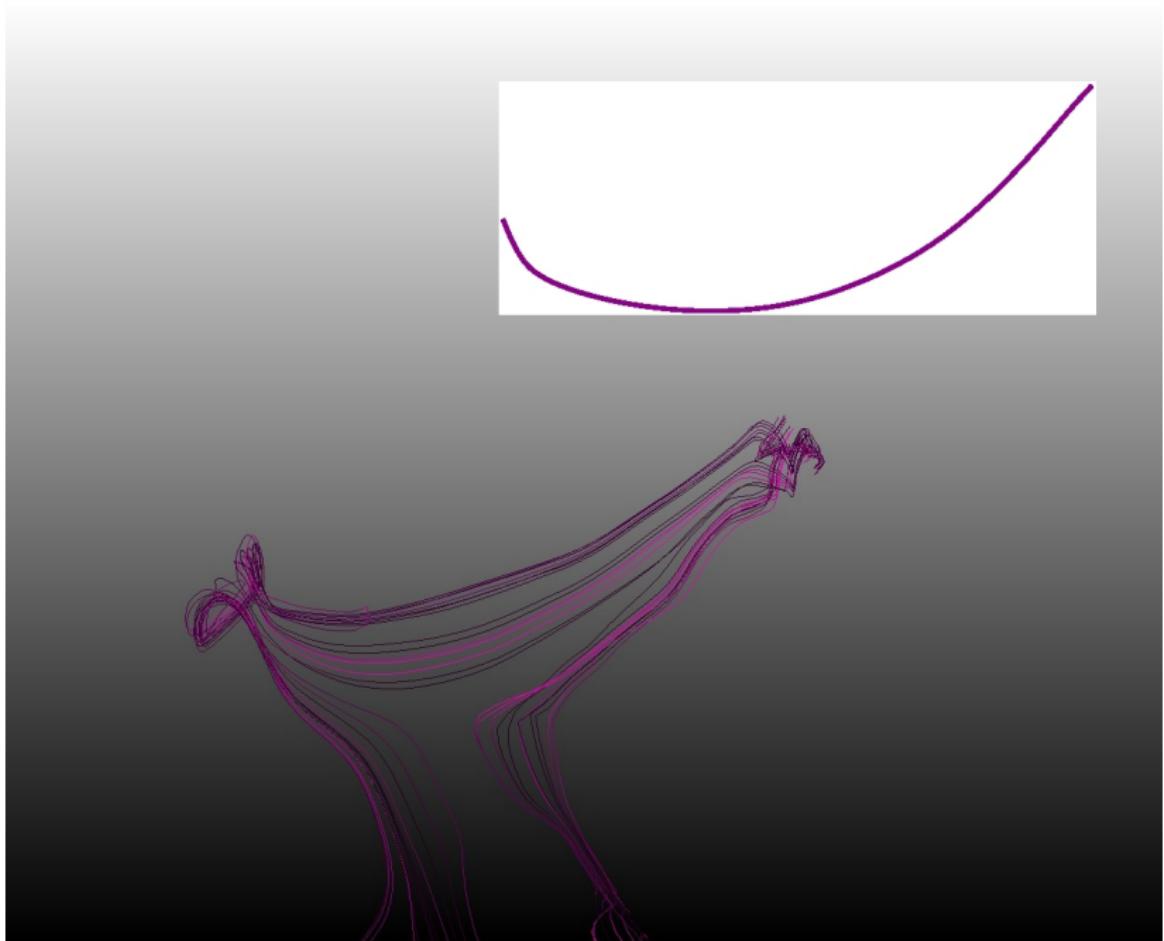
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<sup>2</sup>Department of Statistics, Florida State University, Tallahassee, USA

ISBI 2010  
April 16, 2010







# DTI Fiber: Physical Features

position

orientation

scale

shape

# DTI Fiber: Physical Features

position



orientation

scale

shape



# DTI Fiber: Physical Features

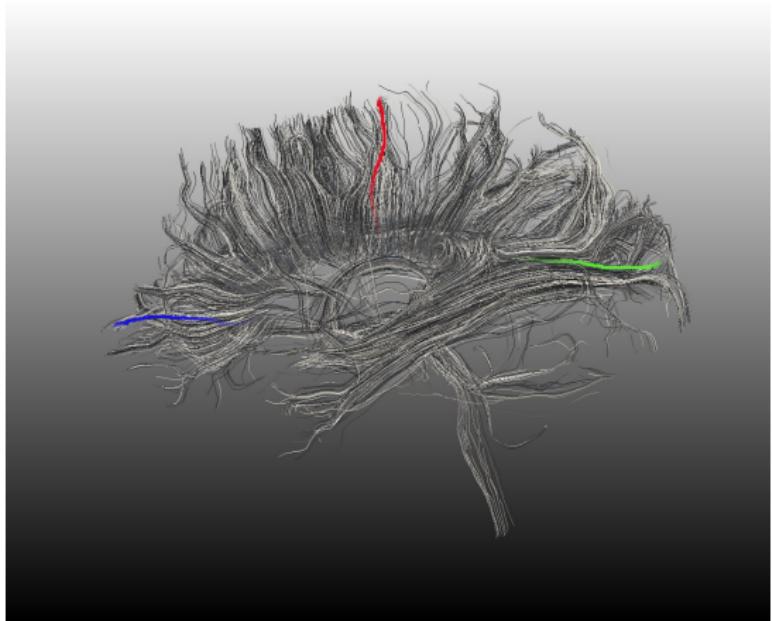
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orientation

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shape



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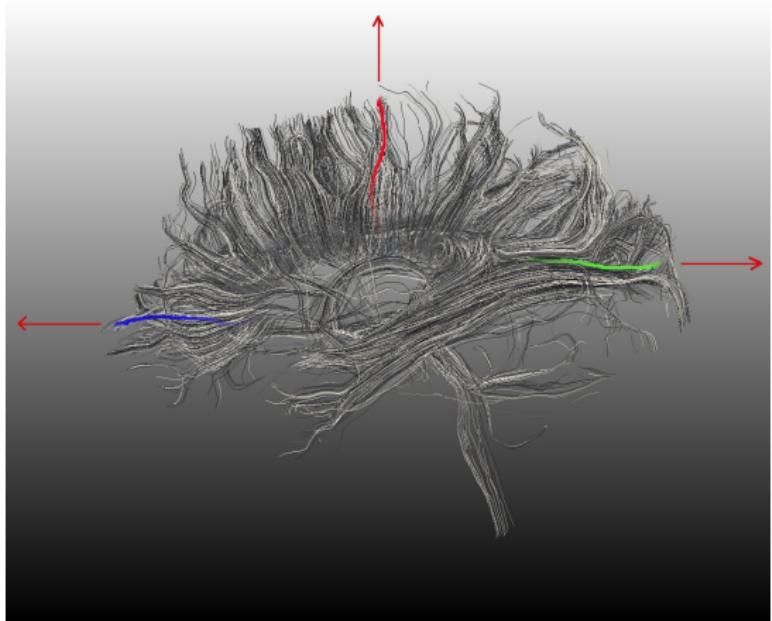
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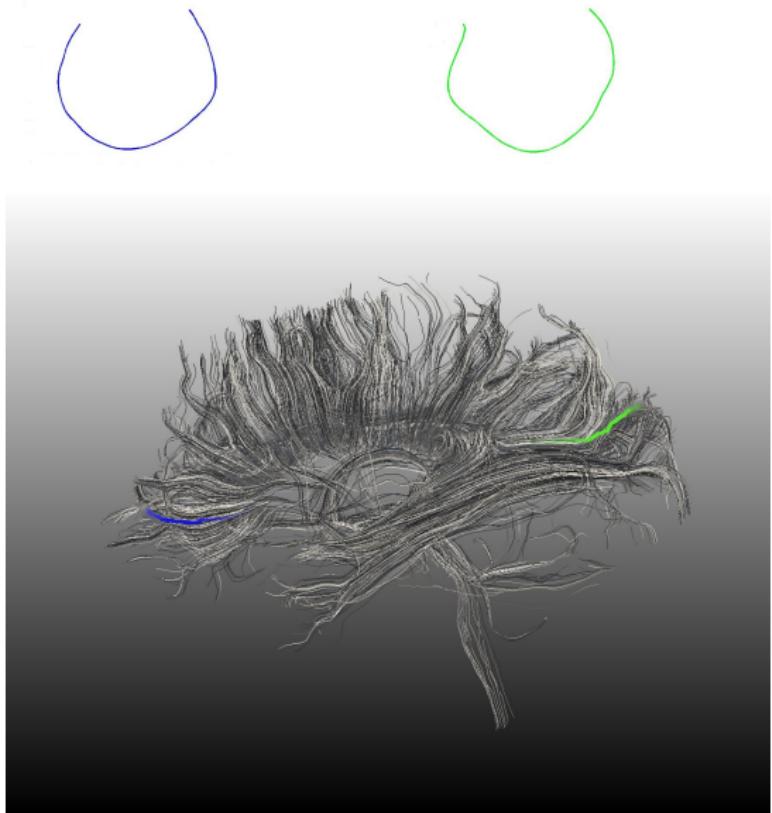
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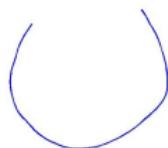
scale

shape



# DTI Fiber: Physical Features

position



orientation



scale

shape

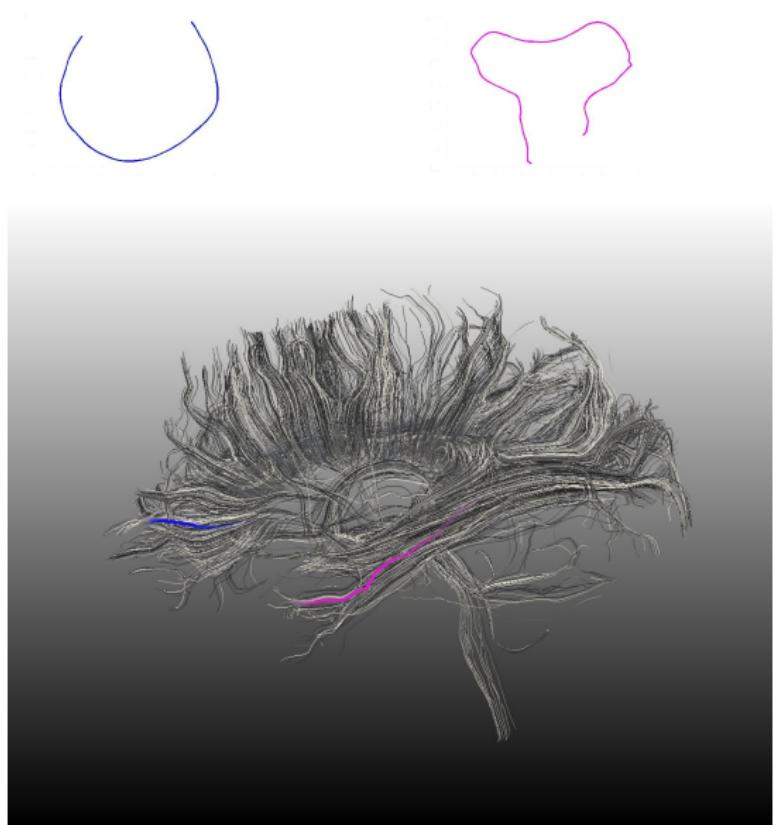
# DTI Fiber: Physical Features

position

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# DTI Fiber: Physical Features

**position**

**orientation**

**scale**

**shape**

These features either individually or in combination can be used to design metrics and feature spaces

## White Matter Fiber Analysis

- varied applications
- different goals and perspectives

# MOTIVATION

## White Matter Fiber Analysis

- varied applications
- different goals and perspectives



## White Matter Fiber Analysis

- varied applications
  - different goals and perspectives
- ↓
- need to design appropriate metrics

# Outline

## 1 Mathematical Framework

- Fiber Tract Representation
- $\mathcal{S}_1$ : Shape + orientation + scale + position
- $\mathcal{S}_2$ : Shape + orientation + scale
- $\mathcal{S}_3$ : Shape + scale
- $\mathcal{S}_4$ : Shape + orientation
- $\mathcal{S}_5$ : Shape

# Outline

## 1 Mathematical Framework

- Fiber Tract Representation
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## 2 Example: Clustering the Corpus Callosum

# Outline

## 1 Mathematical Framework

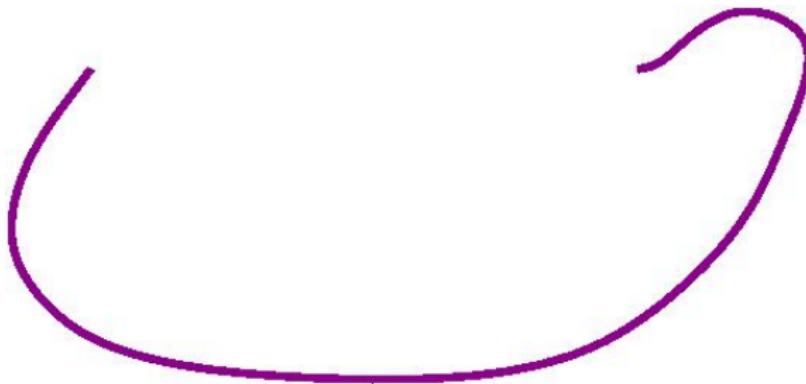
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## 2 Example: Clustering the Corpus Callosum

## 3 Statistics

## Fiber Tract Representation

Let  $\beta : [0, 1] \rightarrow \mathbb{R}^3$  be a parameterized curve



# Comparing Curves

## Common Technique:

$\mathbb{L}^2$  distance:  $\|\beta_1 - \beta_2\| = \sqrt{\int_0^1 \|\beta_1(t) - \beta_2(t)\|^2 dt}$

For example, methods based on Fourier representation use this metric.

**Problem:** This metric is not invariant to reparametrization.

Let  $\gamma : [0, 1] \rightarrow [0, 1]$  be a diffeomorphism ( $\gamma$  acts as a re-parameterization function). A curve  $\beta$  is re-parameterized by  $\gamma$  as  $(\beta \circ \gamma)(t) = \beta(\gamma(t))$ .

It can be shown that  $\|\beta_1 \circ \gamma - \beta_2 \circ \gamma\| \neq \|\beta_1 - \beta_2\|$  in general.

**Solution:** Seek a new representation that preserves  $\mathbb{L}^2$  distance under re-parameterization. The choice of representation depends on the features we want to include in the analysis.

Shape + orientation + scale + position ( $\mathcal{S}_1$ )

**Curve Representation:** “ $h$ ”-function

$$h(t) = \sqrt{\|\dot{\beta}(t)\|} \beta(t), \quad h : [0, 1] \rightarrow \mathbb{R}^3$$

The  $h$ -function of a re-parameterized curve is:  $(h, \gamma) \equiv h(\gamma(t))\sqrt{\dot{\gamma}(t)}$

## Shape + orientation + scale + position ( $S_1$ )

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**Motivation:**  $\|(h_1, \gamma) - (h_2, \gamma)\| = \|h_1 - h_2\|$  for all  $\gamma$

**Curve-Comparison/Distance function:**

$$\gamma^* = \operatorname{argmin}_{\gamma \in \Gamma} (\|h_1 - (h_2, \gamma)\|), \quad d_a(\beta_1, \beta_2) = \|h_1 - (h_2, \gamma^*)\|$$

$d_a$  is invariant to re-parameterizations of  $\beta_1$  and  $\beta_2$

**Geodesic path or Deformation:** Straight line

$$\psi(\tau) = (1 - \tau)h_1 + \tau(h_2, \gamma^*)$$

## Shape + orientation + scale ( $\mathcal{S}_2$ )

**Curve Representation:** “ $q$ ”-function or square-root velocity function

$$q(t) = \frac{\dot{\beta}(t)}{\sqrt{\|\dot{\beta}(t)\|}}$$

Since  $q$  is defined using  $\dot{\beta}$ , it is invariant to the translation of  $\beta$

The  $q$ -function of a re-parameterized curve is:  $(q, \gamma) \equiv q(\gamma(t))\sqrt{\dot{\gamma}(t)}$

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**Curve-Comparison/Distance function:**

$$\gamma^* = \operatorname{argmin}_{\gamma \in \Gamma} (\|q_1 - (q_2, \gamma)\|), \quad d_b(\beta_1, \beta_2) = \|q_1 - (q_2, \gamma^*)\|$$

**Geodesic path or Deformation:** Straight line

$$\psi(\tau) = (1 - \tau)q_1 + \tau(q_2, \gamma^*)$$

## Shape + scale ( $\mathcal{S}_3$ )

We **remove rotation** from the representation as follows:

**Curve Representation:**  $q$ -function

$$q(t) = \frac{\dot{\beta}(t)}{\sqrt{\|\dot{\beta}(t)\|}}$$

**Curve Alignment:**

$$(O^*, \gamma^*) = \operatorname{argmin}_{O \in SO(3), \gamma \in \Gamma} (\|q_1 - O(q_2, \gamma)\|)$$

**Distance function:**

$$d_d(\beta_1, \beta_2) = \|q_1 - O^*(q_2, \gamma^*)\|$$

**Geodesic path:** Straight line

$$\psi(\tau) = (1 - \tau)q_1 + \tau(O^* q_2, \gamma^*)$$

## Shape + orientation ( $\mathcal{S}_4$ )

Keep orientation but remove scale by rescaling all curves

**Curve Representation:**  $q(t) = \frac{\dot{\beta}(t)}{\sqrt{\|\dot{\beta}(t)\|}}$

**Distance function:** Arc-Length on a sphere

$$d_c(\beta_1, \beta_2) = \min_{\gamma \in \Gamma} \left( \cos^{-1} \left( \int_0^1 \langle (q_1, \gamma)(t), (q_2, \gamma)(t) \rangle dt \right) \right)$$

**Geodesic path:** Great circle!

$$\psi(\tau) = \frac{1}{\sin(\theta)} [\sin(\theta - \tau\theta) q_1 + \sin(\tau\theta) (q_2, \gamma^*)]$$

where  $\theta = d_c(\beta_1, \beta_2)$

## Shape ( $\mathcal{S}_5$ )

Now we **remove** all variables: **rotation, translation and scale**

**Curve Representation:**  $q(t) = \frac{\dot{\beta}(t)}{\sqrt{\|\dot{\beta}(t)\|}}$

**Curve Alignment:**

$$(O^*, \gamma^*) = \operatorname{argmin}_{O \in SO(3), \gamma \in \Gamma} \left( \cos^{-1} \left( \int_0^1 \langle (q_1, \gamma)(t), O(q_2, \gamma)(t) \rangle dt \right) \right)$$

**Distance function:**

$$d_e(\beta_1, \beta_2) = \cos^{-1} \left( \int_0^1 \langle (q_1, \gamma)(t), O^*(q_2, \gamma)(t) \rangle dt \right)$$

**Geodesic path:** Great circle!

$$\psi(\tau) = \frac{1}{\sin(\theta)} [\sin(\theta - \tau\theta) q_1 + \sin(\tau\theta) (O^* q_2, \gamma^*)]$$

for  $\theta = d_e(\beta_1, \beta_2)$

# In Summary

Manifold	function representation	pre-shape space	shape space
shape + orientation + scale + position	$h(t) = \sqrt{\ \dot{\beta}(t)\ } \beta(t)$	$\mathbb{L}^2$ space $\mathcal{C}_1$	$S_1 = \mathcal{C}_1 / (\Gamma)$
shape + orientation + scale	$q(t) = \frac{\dot{\beta}(t)}{\sqrt{\ \dot{\beta}(t)\ }}$	$\mathbb{L}^2$ space $\mathcal{C}_2$	$S_2 = \mathcal{C}_2 / (\Gamma)$
shape + scale	$q(t) = \frac{\dot{\beta}(t)}{\sqrt{\ \dot{\beta}(t)\ }}$	$\mathbb{L}^2$ space $\mathcal{C}_2$	$S_3 = \mathcal{C}_2 / (\Gamma \times SO(3))$
shape + orientation	$q(t) = \frac{\dot{\beta}(t)}{\sqrt{\ \dot{\beta}(t)\ }}$	$\mathcal{C}_3$ hypersphere	$S_4 = \mathcal{C}_3 / (\Gamma)$
shape	$q(t) = \frac{\dot{\beta}(t)}{\sqrt{\ \dot{\beta}(t)\ }}$	$\mathcal{C}_3$ hypersphere	$S_5 = \mathcal{C}_3 / (\Gamma \times SO(3))$

# Geodesic Paths

Evolution of one curve into another



$\mathcal{S}_2$ : shape+orientation+scale



$\mathcal{S}_3$ : shape+scale



$\mathcal{S}_4$ : shape+orientation



$\mathcal{S}_5$ : shape

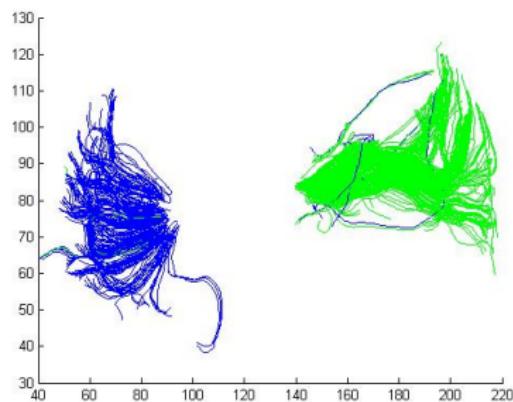
## Application: Clustering Fibers in the Corpus Callosum



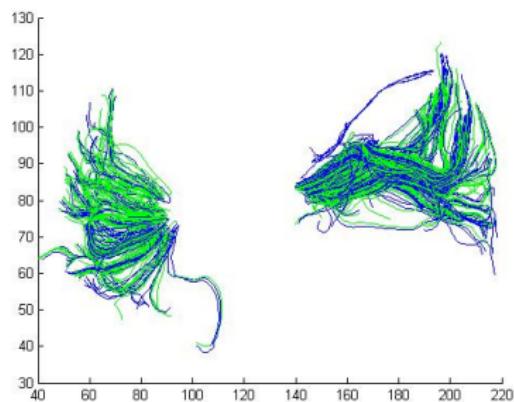
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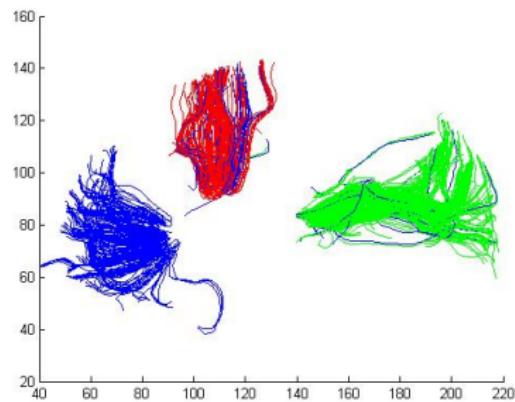
(a) shape+orientation+scale ( $d_b$ )



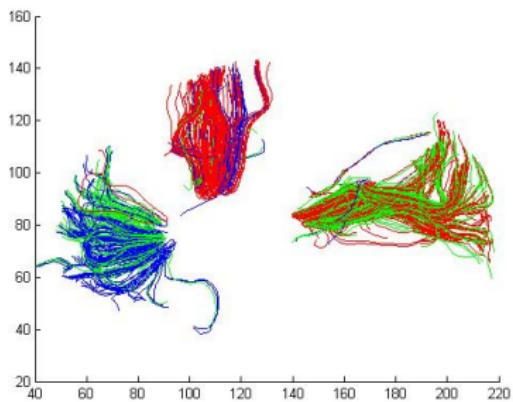
(b) shape ( $d_e$ )

Clustering the genu and splenium, the anterior and posterior sections of the CC. Here, shape information alone (b) is not adequate for clustering.

## Application: Clustering Fibers in the Corpus Callosum



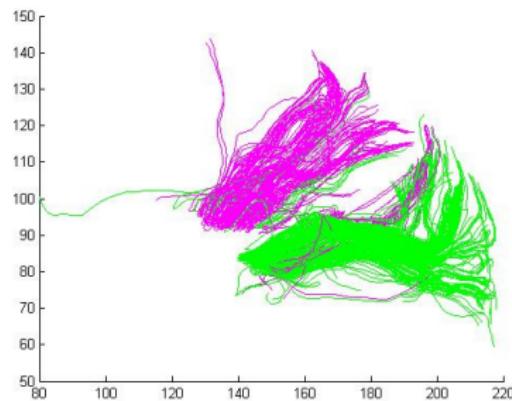
(a) shape+orientation ( $d_c$ )



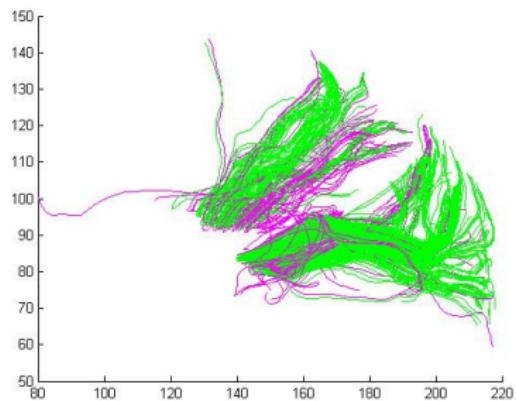
(b) shape+orientation+scale ( $d_b$ )

Clustering of the genu, corpus and splenium, the anterior, middle and posterior sections of the CC. Including the scale information results in poorer clustering (b).

## Application: Clustering Fibers in the Corpus Callosum



(a) shape+orientation ( $d_c$ )



(b) shape+scale ( $d_d$ )

Clustering the isthmus and splenium, the posterior CC

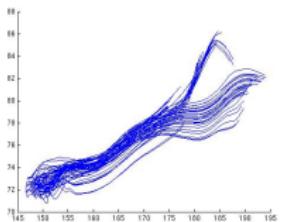
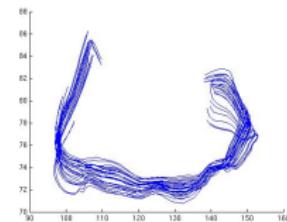
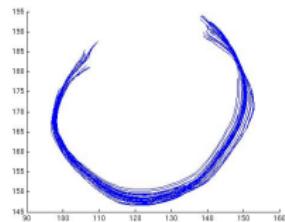
# Statistics

- Summary statistics
- Statistical inference
  - ▶ Stochastic models
  - ▶ Hypothesis testing

# Statistics

## Mean Curves of a DTI Fiber Bundle: Splenium

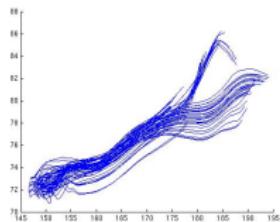
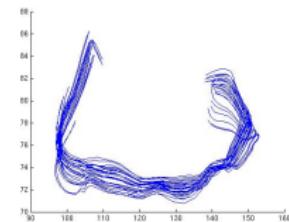
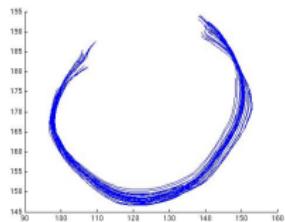
fiber  
bundle



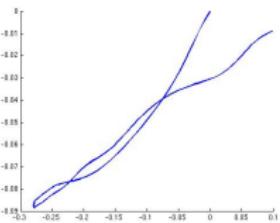
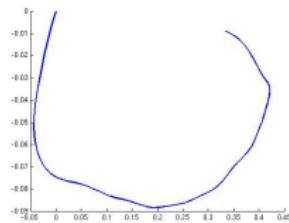
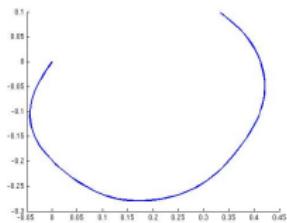
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fiber  
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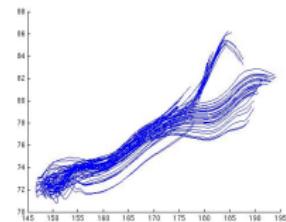
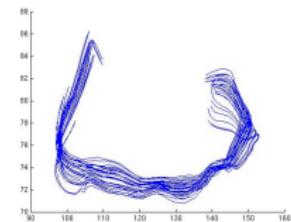
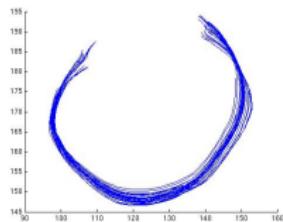
shape



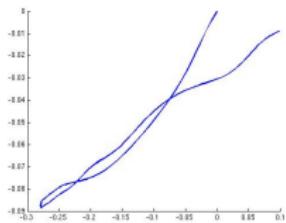
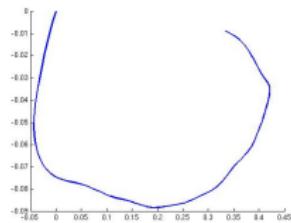
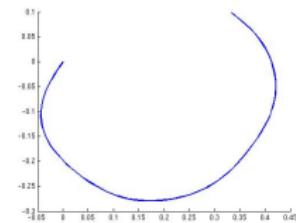
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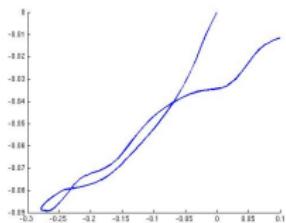
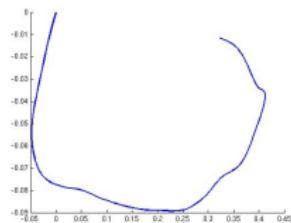
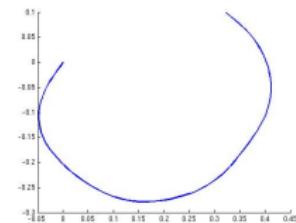
fiber  
bundle



shape



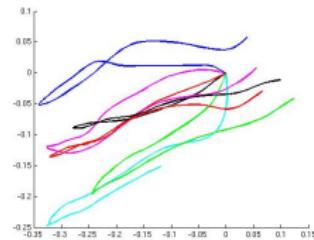
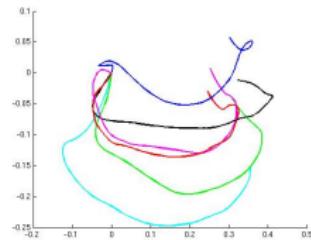
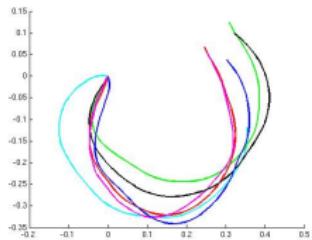
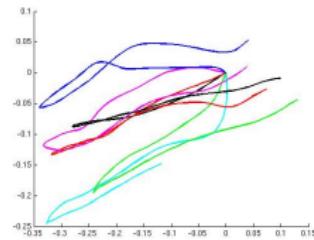
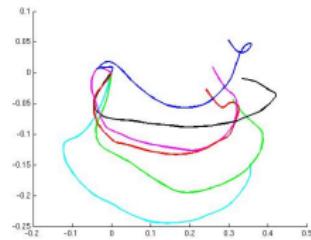
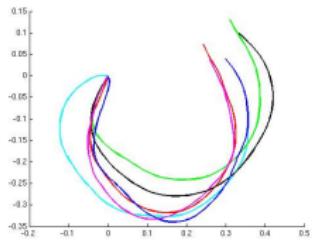
shape+  
orientation



# Statistics

## Mean Curves of a Population: Spleniums of 6 Subjects

shape  
shape+  
orientation



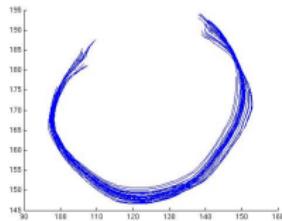
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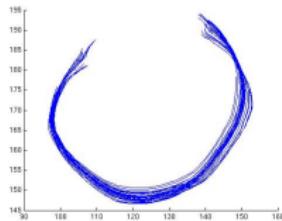
Search terms: Anuj Srivastava, FSU, Publications

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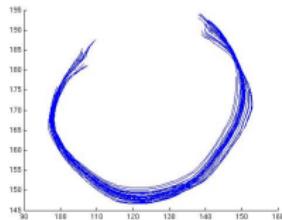
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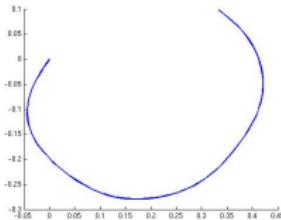
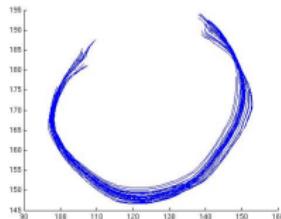
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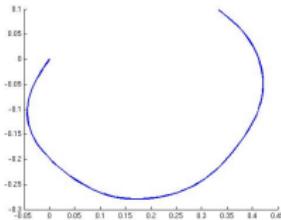
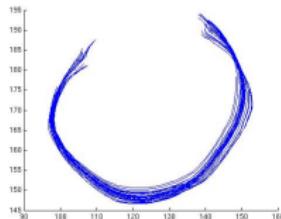
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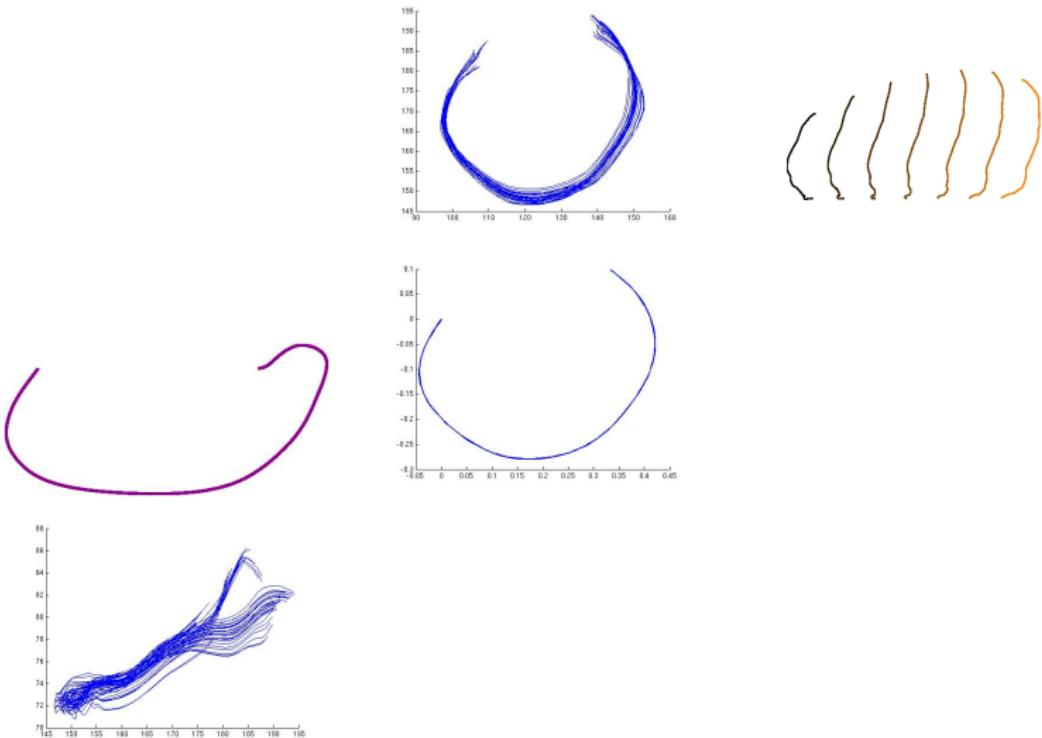
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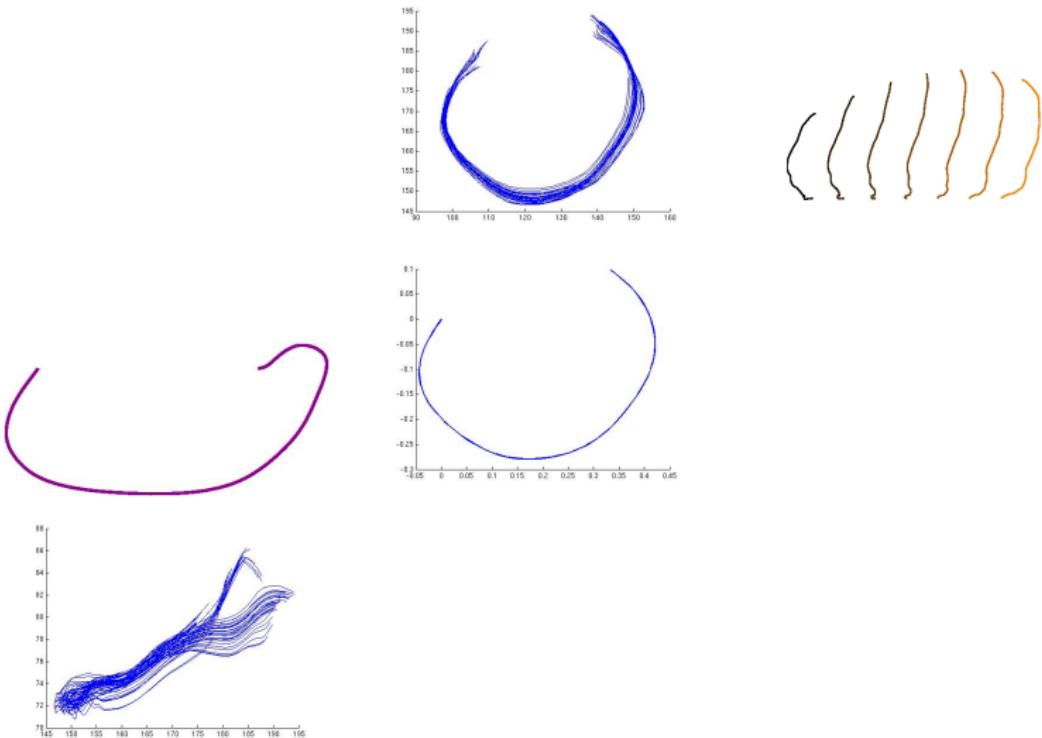
# Thank You



# Thank You



# Thank You



# Thank You

