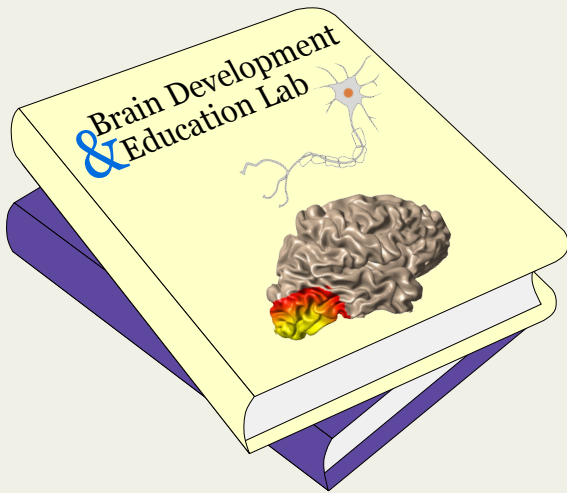


Models: From voxels, to fascicles, to brain development and cognition



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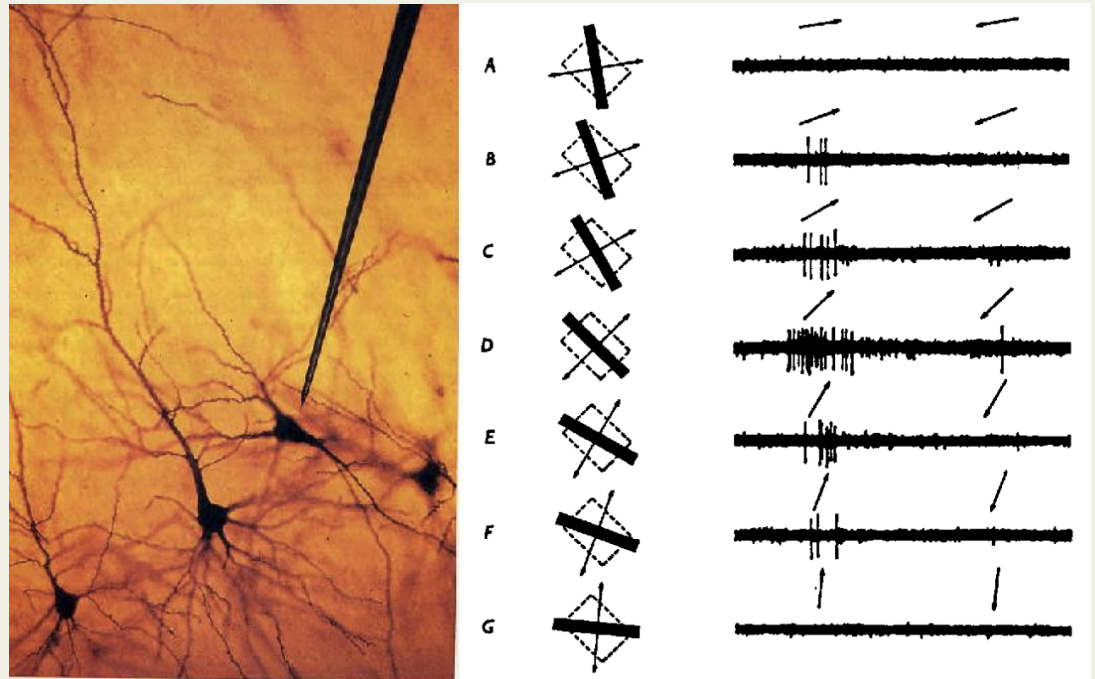
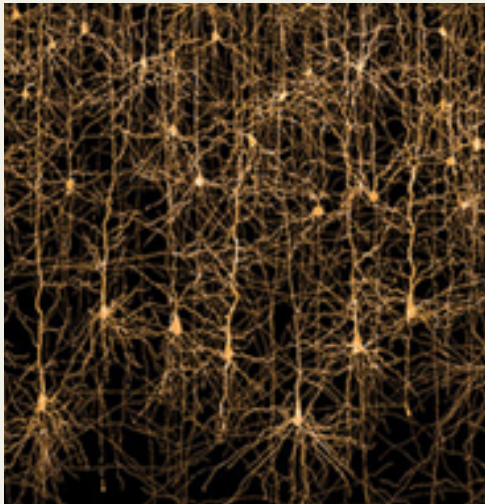


INSTITUTE FOR LEARNING & BRAIN SCIENCES

UNIVERSITY of WASHINGTON

Cognition and the neuron doctrine

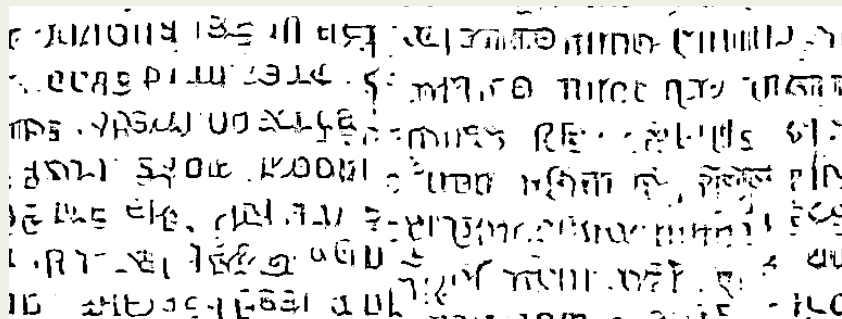
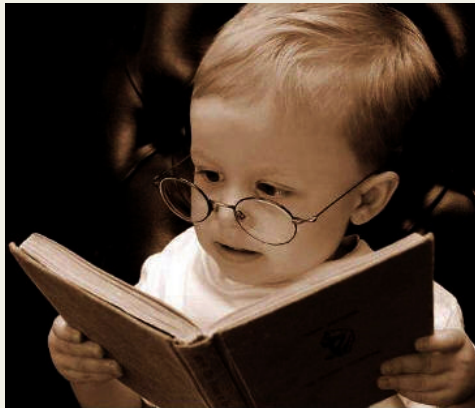
Cognitive function can be attributed to localized neural activity



Hubel and Wiesel (1959 onward)

Developmental changes in behavior occur over much longer time scales

- For example ~10 years to become a skilled reader.
- Learning to read requires brain circuits to modify their structure in response to years of training (Wandell & Yeatman, 2013).



Portilla & Simoncelli, 2000

to have your cream opens paper
robin seeds shoot intent struck fa
s wildly tower shut theirs faster b
broad nato native lists polite cafe
black found shed market mass swing
baked powers minds round hide laws
al sound growth occupy bath alter h
blue remain court others makes a

Cognitive development depends on tissue changes that occur over correspondingly long time-scales

- Understanding development requires measurements that are sensitive to changes in glia, axons, myelin and vasculature.

MRI can be used to quantify brain tissue properties and model the interplay between the development of brain circuits and cognitive functions.

But how do we go from MR signals to models of development?

(LaMantia & Rakic, 1990)

Ture (2000)

puzo (1998)

Neuron

Astrocyte

Allen & Barres (2009)

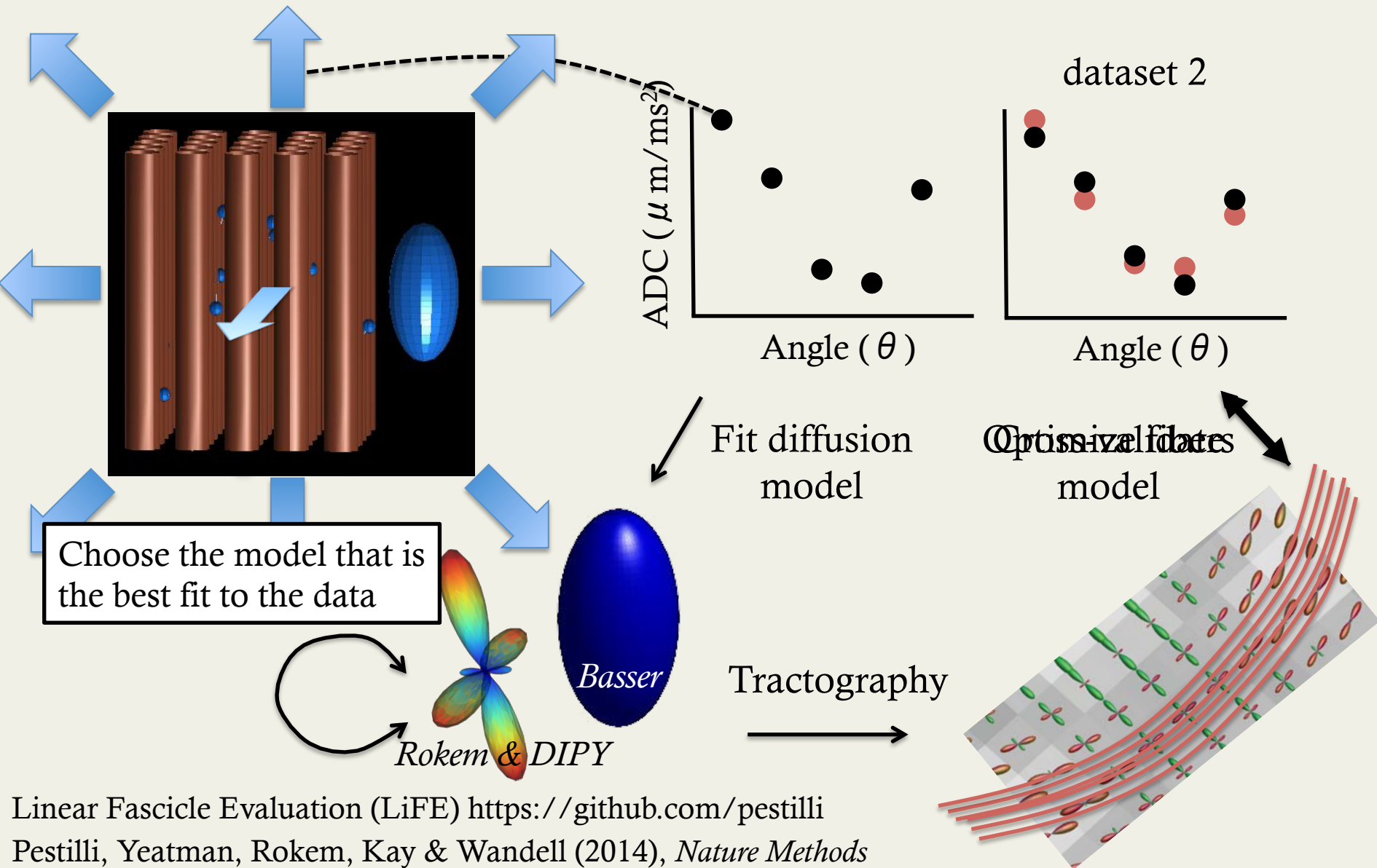
Outline

1. From diffusion to fascicles: Models of an individual's white matter.
2. Quantitative MRI measurements of tissue volume and composition.
3. Combining multiple measurements to dissociate developmental processes.
 - Testing models of development.

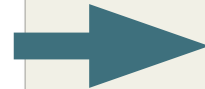
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Building a wiring diagram with dMRI



Fitting a tensor model to diffusion measurements



$$\begin{pmatrix} \sigma_{xx} & \sigma_{xy} & \sigma_{xz} \\ \sigma_{yx} & \sigma_{yy} & \sigma_{yz} \\ \sigma_{zx} & \sigma_{zy} & \sigma_{zz} \end{pmatrix}$$



Diffusion signal
measured in
different directions

Diffusion tensor model:
Predicts Gaussian diffusion
in three dimensions

Visualization of
tensor model:
Ellipsoid

Fitting a tensor model to diffusion measurements

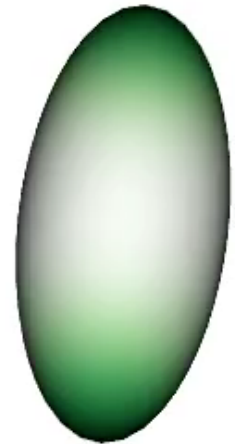


Diffusion signal
measured in
different directions

Predict diffusion measures
from tensor model

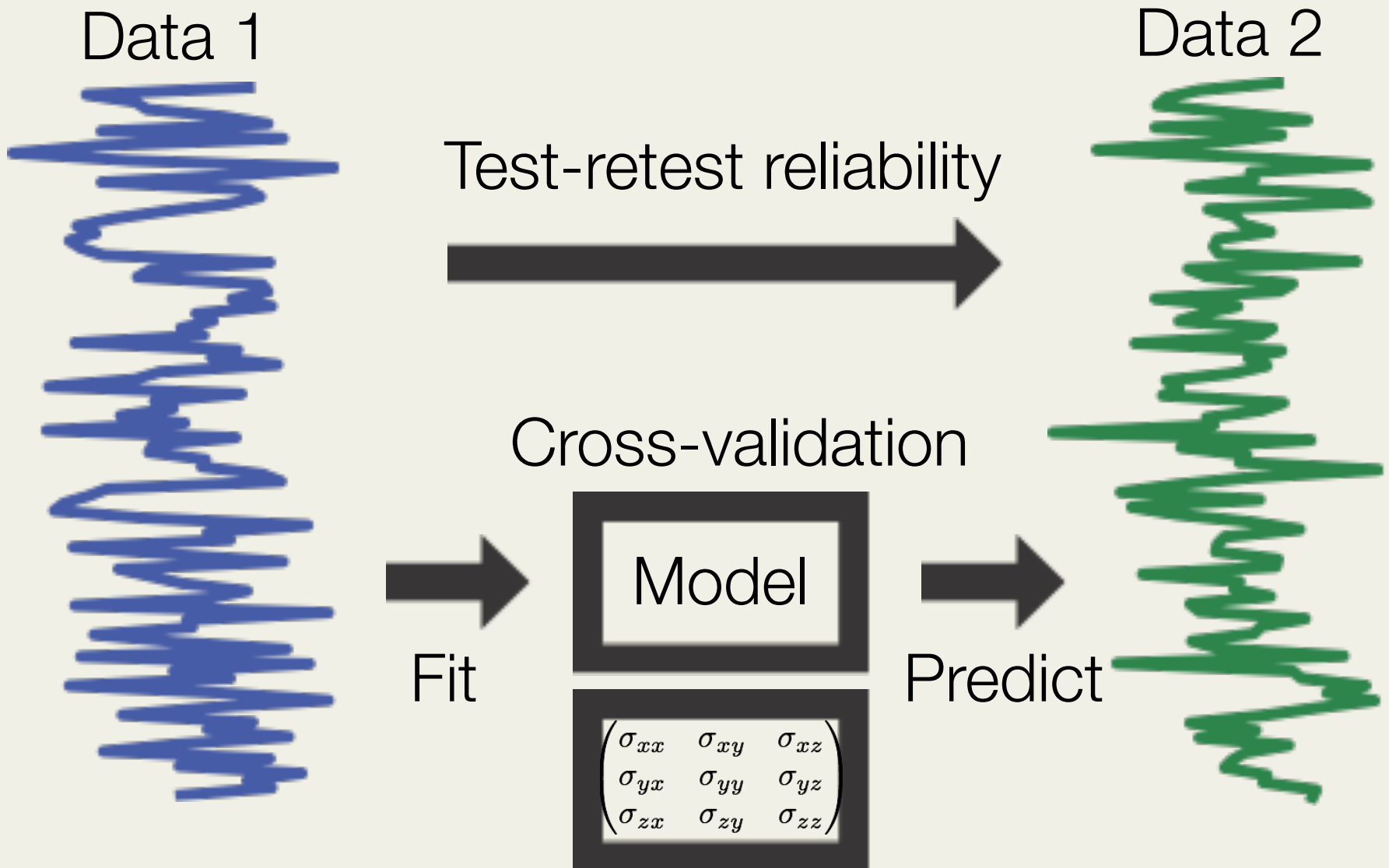
$$\begin{pmatrix} \sigma_{xx} & \sigma_{xy} & \sigma_{xz} \\ \sigma_{yx} & \sigma_{yy} & \sigma_{yz} \\ \sigma_{zx} & \sigma_{zy} & \sigma_{zz} \end{pmatrix}$$

Diffusion tensor model:
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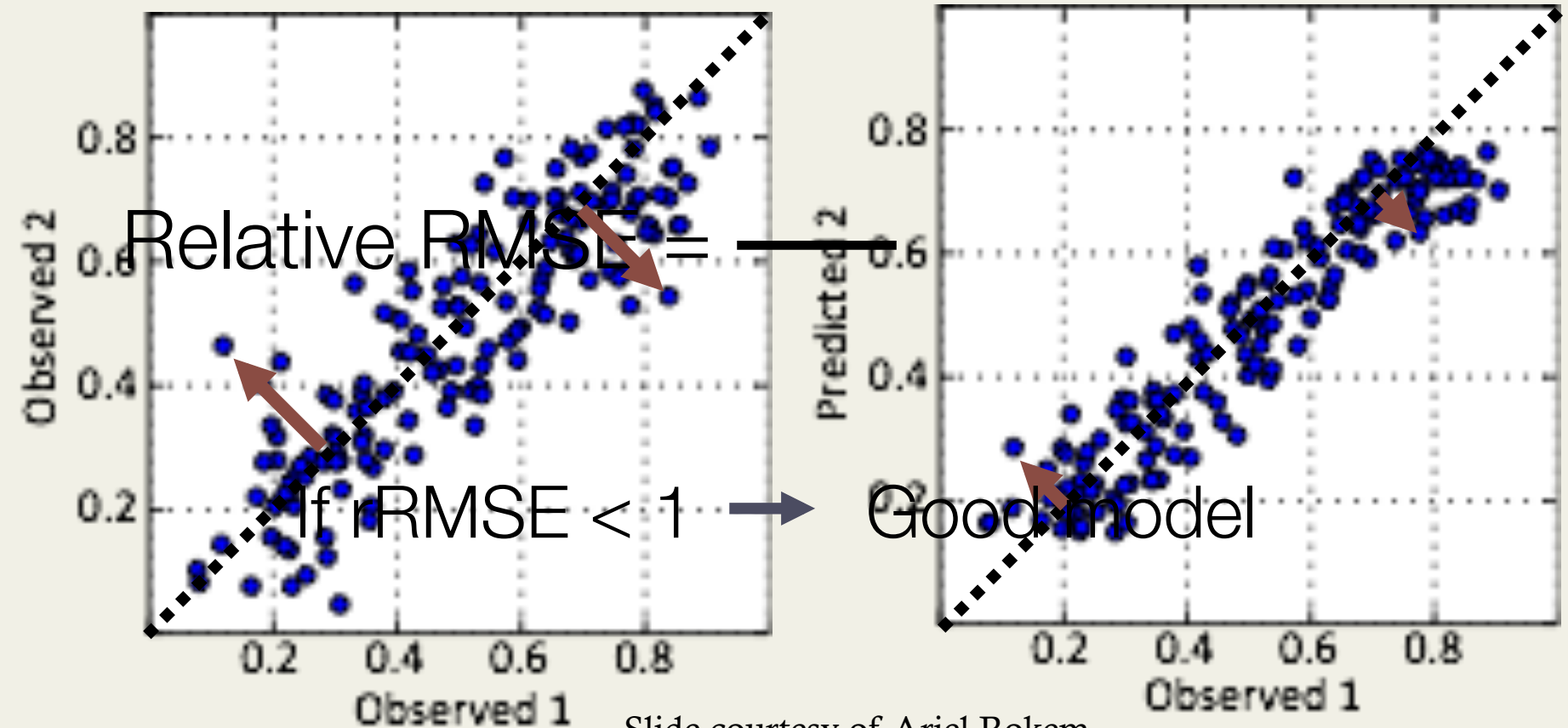
Evaluating the tensor model



Evaluating tensor model with cross validation

Data => data

Model => data

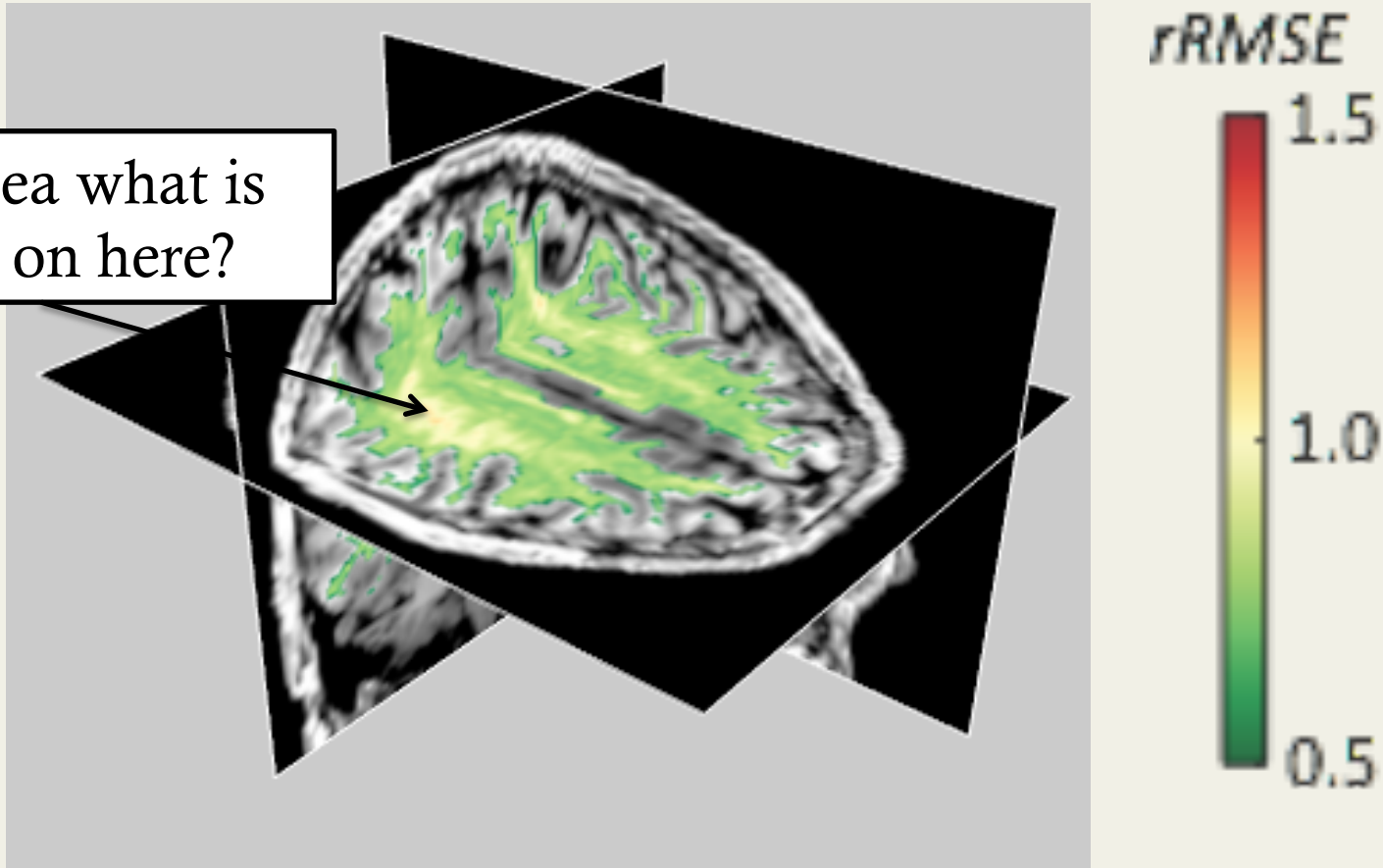


Slide courtesy of Ariel Rokem

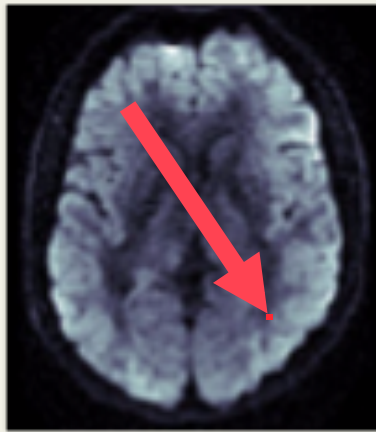
Rokem, Yeatman, Pestilli, Kay, Mezer, Van der Walt and Wandell, (2014), *PLoS ONE*

The tensor model is a good fit through much of the brain

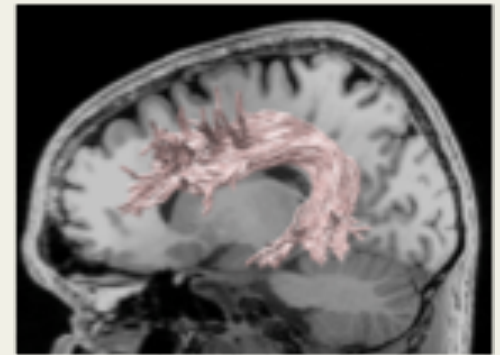
Any idea what is going on here?



Linear Fascicle Evaluation (LiFE): Validating tractography

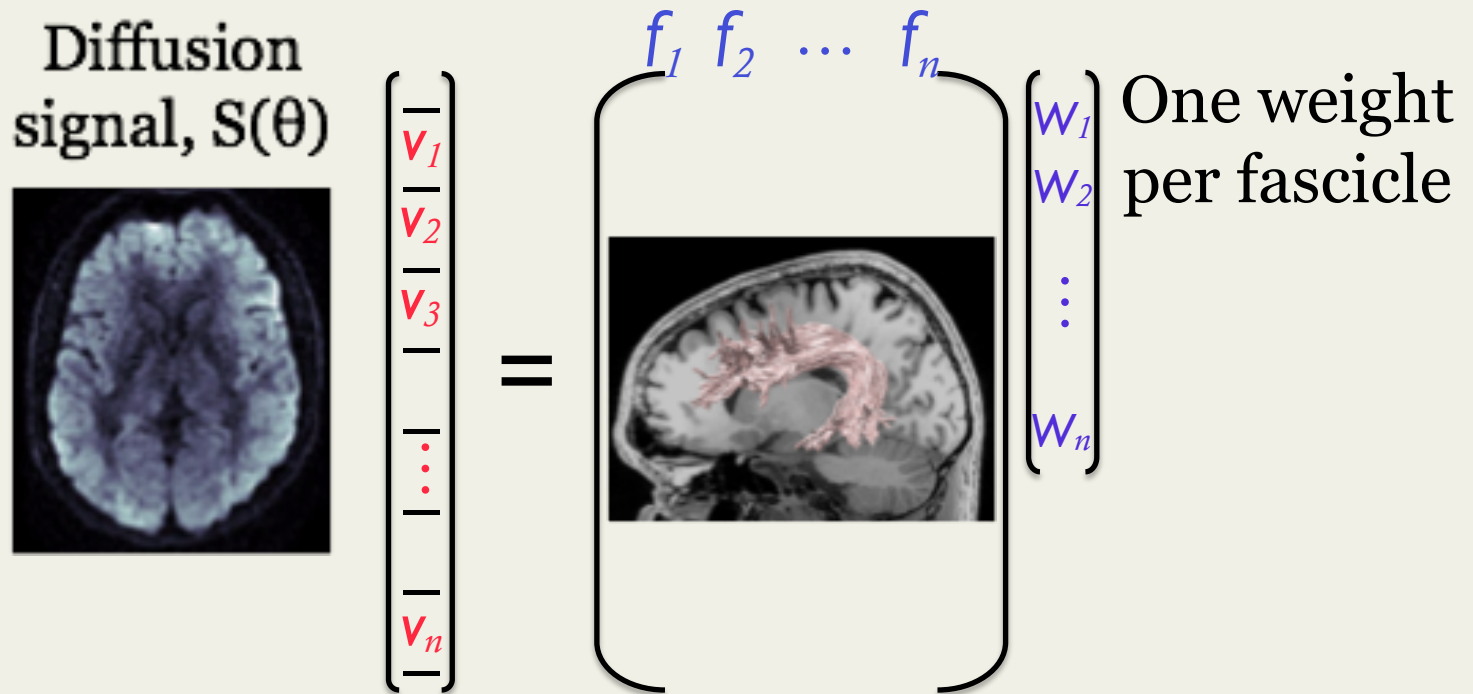


$$\begin{array}{c} \text{Voxel signal} \\ \left[\begin{array}{c} \phantom{\text{Fascicle 1}} \\ \phantom{\text{Fascicle 2}} \\ \phantom{\text{Fascicle 3}} \\ \phantom{\text{Fascicle 4}} \end{array} \right] \end{array} = \begin{array}{c} \text{Fascicle 1} \\ \left[\begin{array}{c} \phantom{\text{Fascicle 1}} \\ \phantom{\text{Fascicle 2}} \\ \phantom{\text{Fascicle 3}} \\ \phantom{\text{Fascicle 4}} \end{array} \right] \end{array} + \begin{array}{c} \text{Fascicle 2} \\ \left[\begin{array}{c} \phantom{\text{Fascicle 1}} \\ \phantom{\text{Fascicle 2}} \\ \phantom{\text{Fascicle 3}} \\ \phantom{\text{Fascicle 4}} \end{array} \right] \end{array} + \begin{array}{c} \text{Fascicle 3} \\ \left[\begin{array}{c} \phantom{\text{Fascicle 1}} \\ \phantom{\text{Fascicle 2}} \\ \phantom{\text{Fascicle 3}} \\ \phantom{\text{Fascicle 4}} \end{array} \right] \end{array} + \begin{array}{c} \text{Fascicle 4} \\ \left[\begin{array}{c} \phantom{\text{Fascicle 1}} \\ \phantom{\text{Fascicle 2}} \\ \phantom{\text{Fascicle 3}} \\ \phantom{\text{Fascicle 4}} \end{array} \right] \end{array}$$



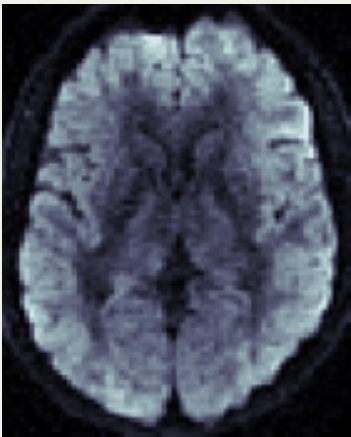
Linear Fascicle Evaluation (LiFE): Validating tractography

Validation: Eliminate fascicles with zero weight



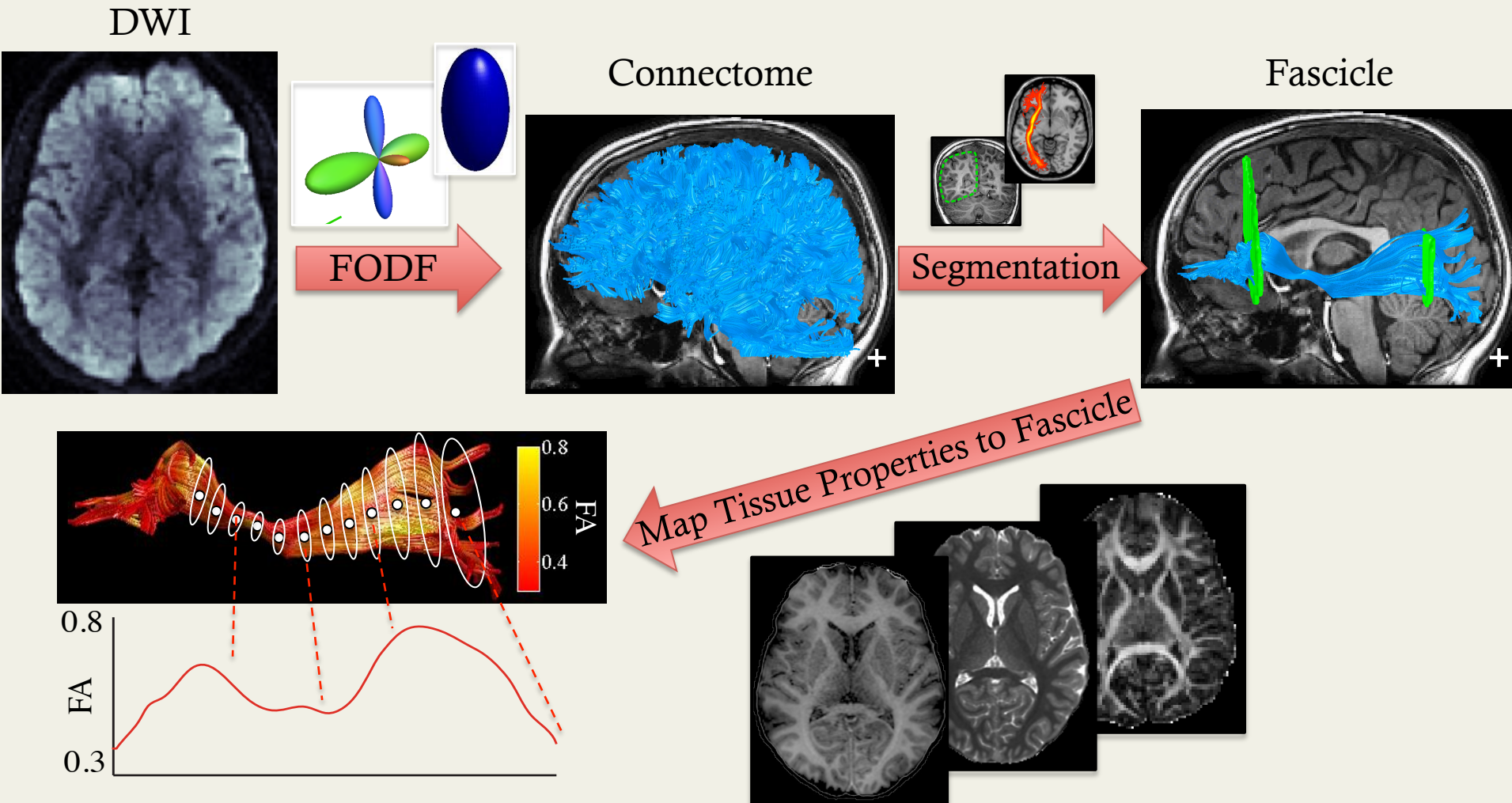
Automated fiber tract quantification: Segmenting an individual's white matter

DWI



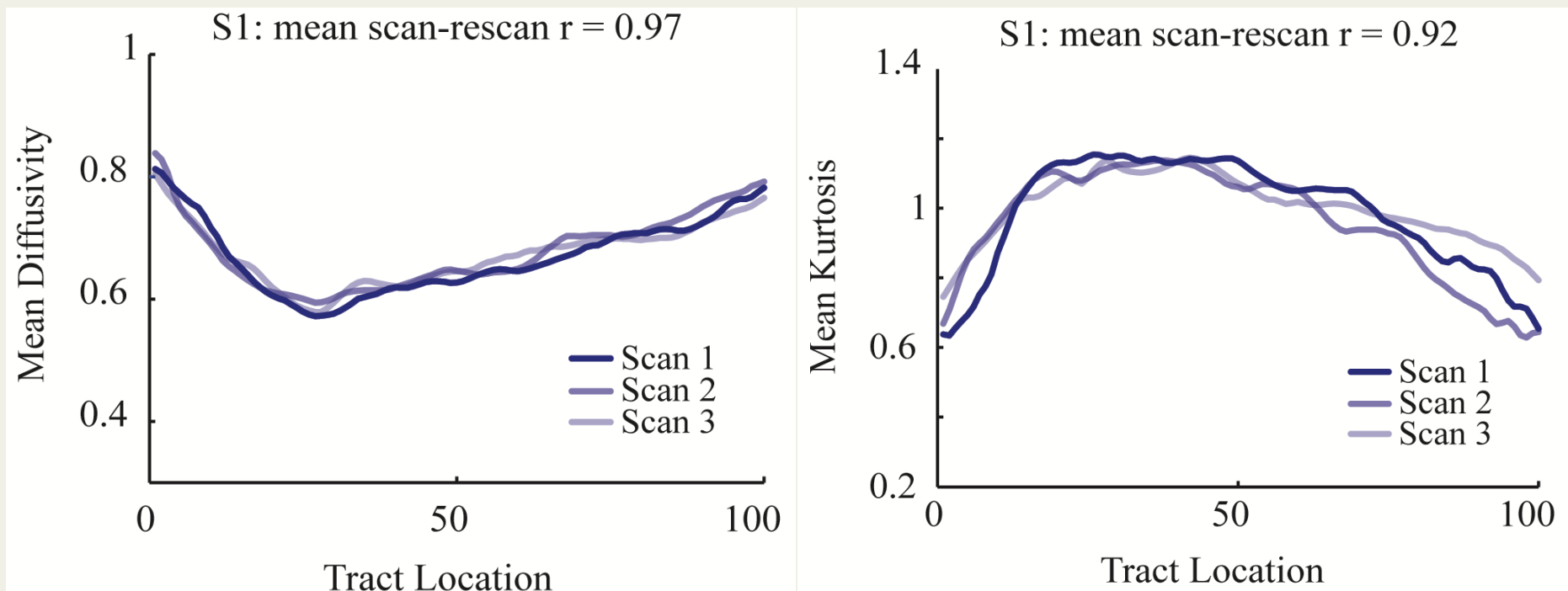
The goal of all this modeling is to generate accurate estimates of an individual's white matter connections

Automated fiber tract quantification: Segmenting an individual's white matter



So with all this work, how well do we do?

- Measures of white matter microstructure are **highly** reliable.



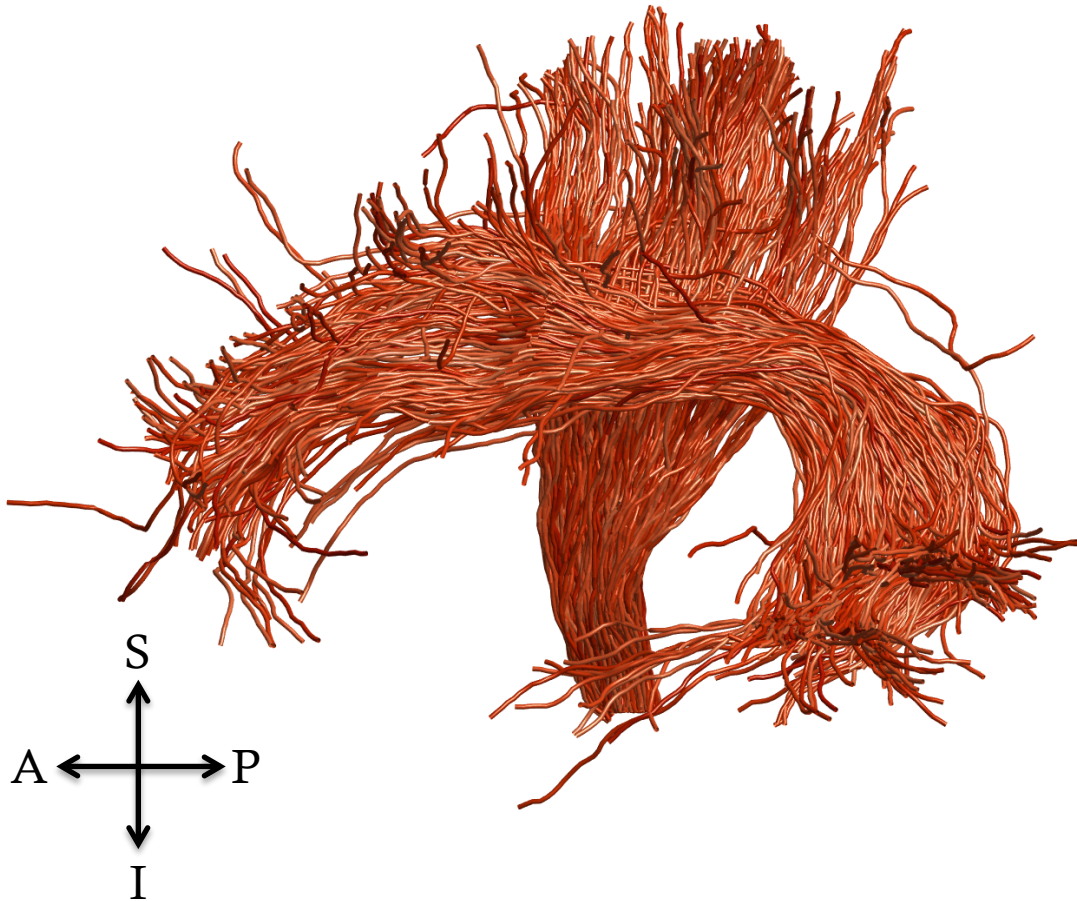
Arcuate Fasciculus – dMRI acquisition: 32 directions $b=800$; 64 directions $b=2000$

How might we select the optimal tractography algorithm to use with AFQ?

- Consider two use cases:
 - Clinical data collected on children with traumatic brain injury versus Human Connectome Project data.
- What might be the pros and cons to using a tensor model with deterministic tractography versus spherical deconvolution with probabilistic tractography?

The choice of algorithm has a substantial impact on the results

Spherical Deconvolution
Probabilistic tractography



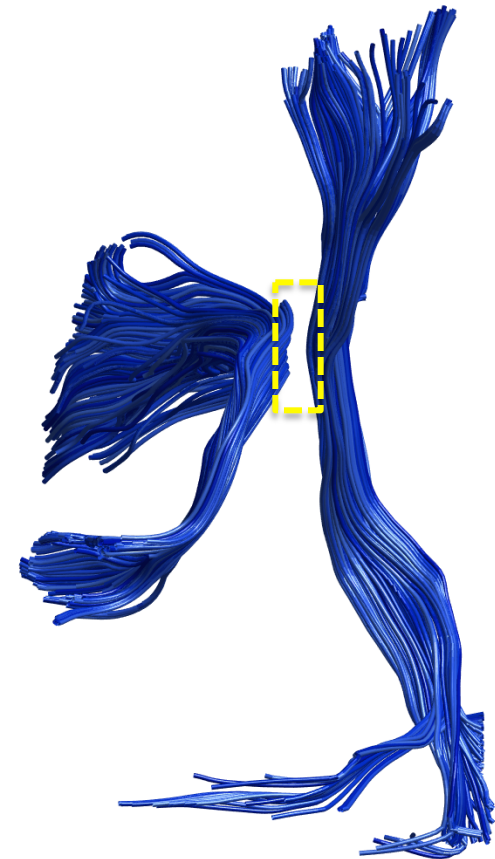
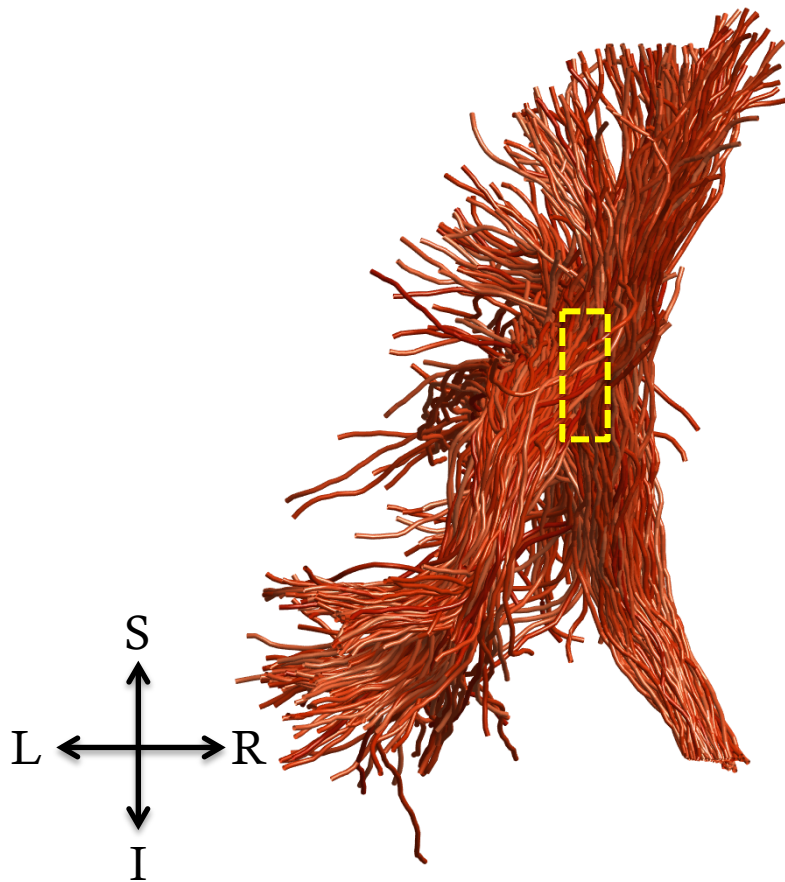
Tensor Model
Deterministic Tractography



The choice of algorithm has a substantial impact on the results

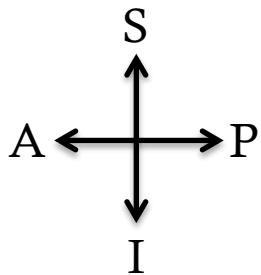
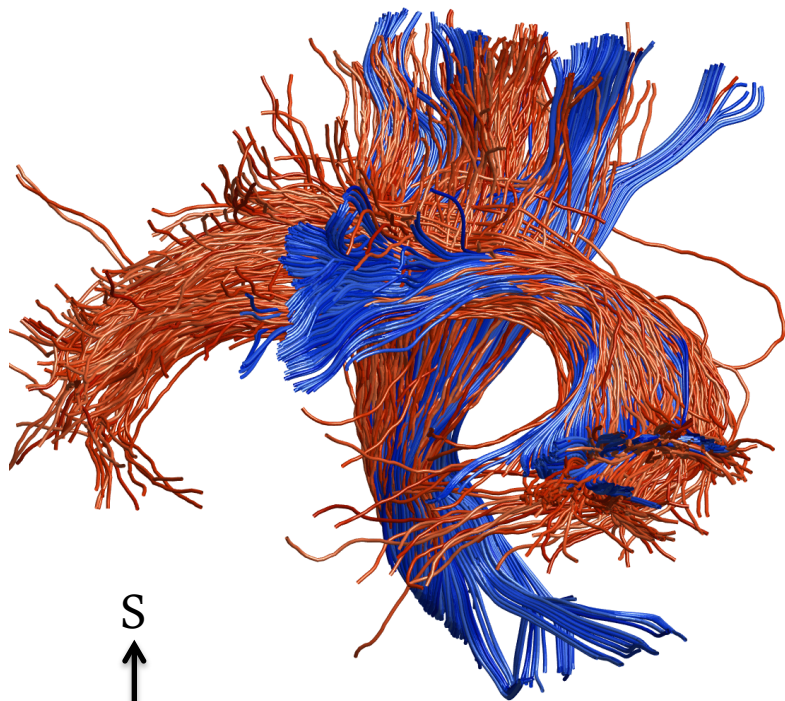
Spherical Deconvolution
Probabilistic tractography

Tensor Model
Deterministic Tractography

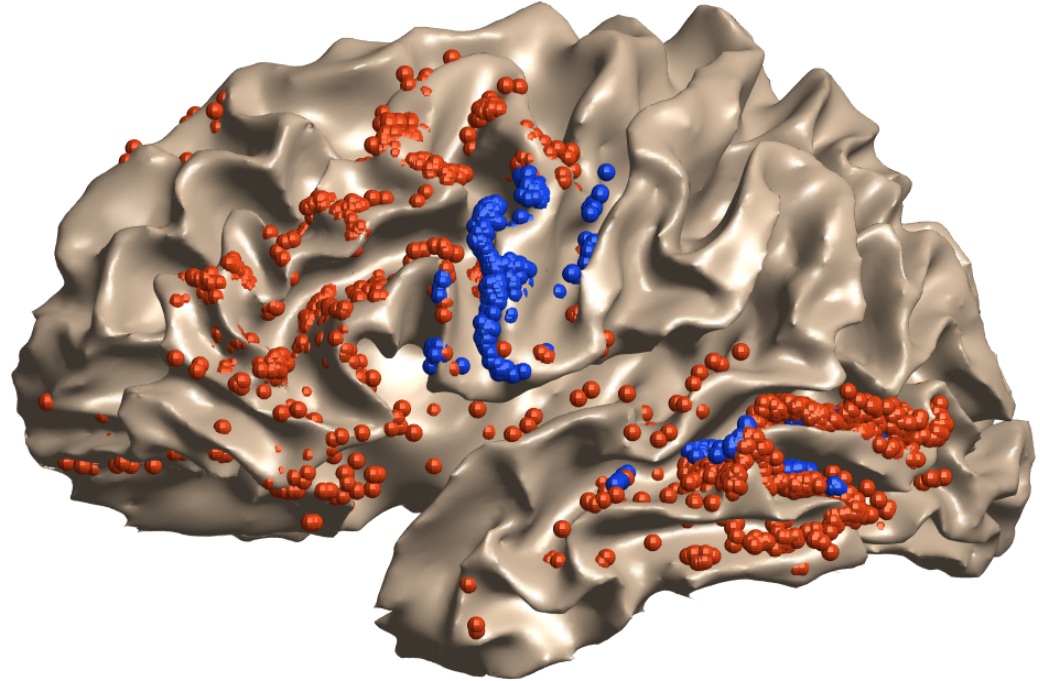


The core of the fascicle is consistent but the cortical endpoints differ

- Select the appropriate algorithm based on the goals of the study.
- Test the fit of the fascicles to the diffusion measurements.
 - Remember that tractography is a model



The same “core”
fascicle



Substantially different
endpoints

Summary: From diffusion to fascicles

- Based on diffusion measurements we can fit a model of the fascicles that pass through a voxel and quantify the fit of the model with cross-validation.
 - http://nipy.org/dipy/examples_built/kfold_xval.html#example-kfold-xval
- Inferences about brain connectivity depend on selecting the appropriate diffusion model and tractography algorithm for your research question.
 - Fascicles are themselves a model and we should use cross-validation to test how well they predict the data.
 - <https://github.com/pestilli>

Outline

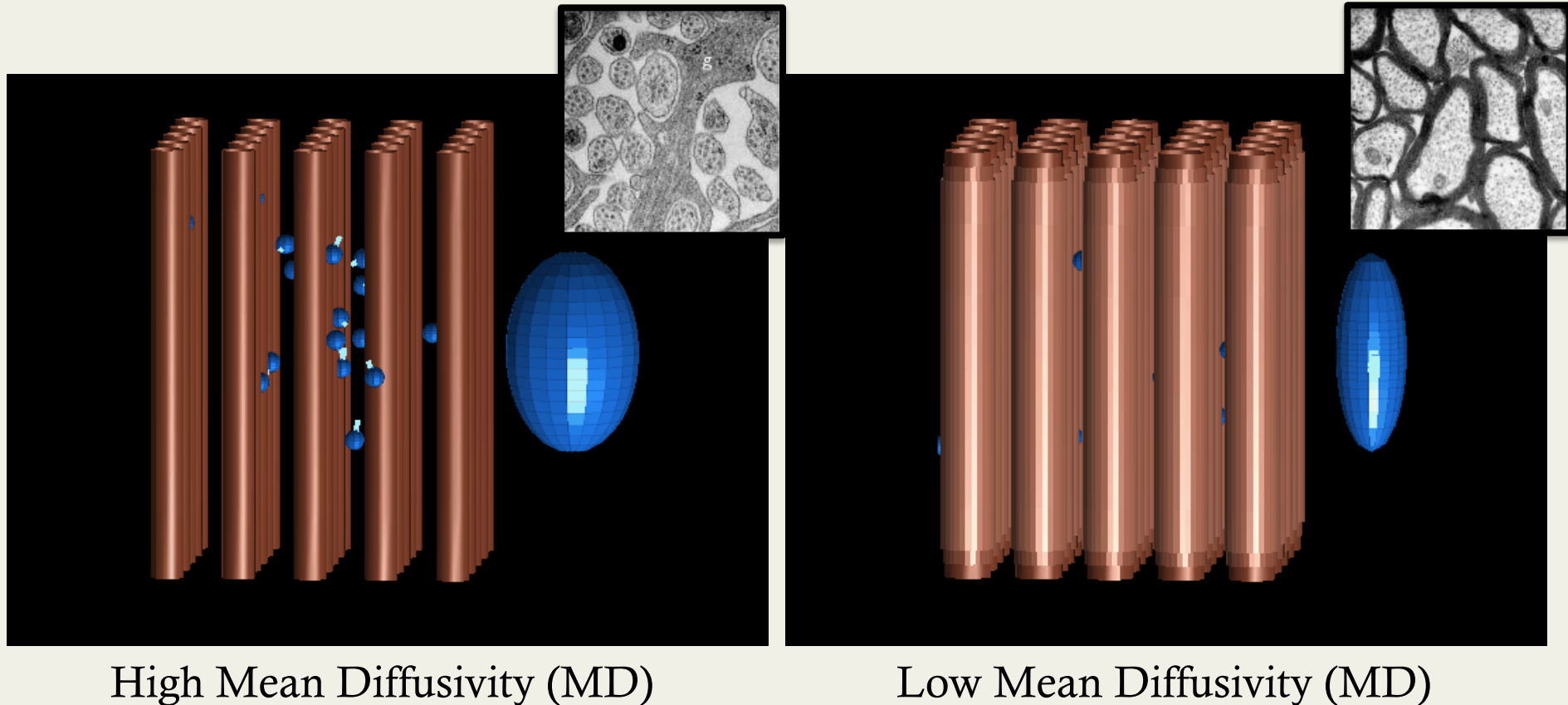
- 1. From diffusion to fascicles: Segmenting an individual's white matter.**
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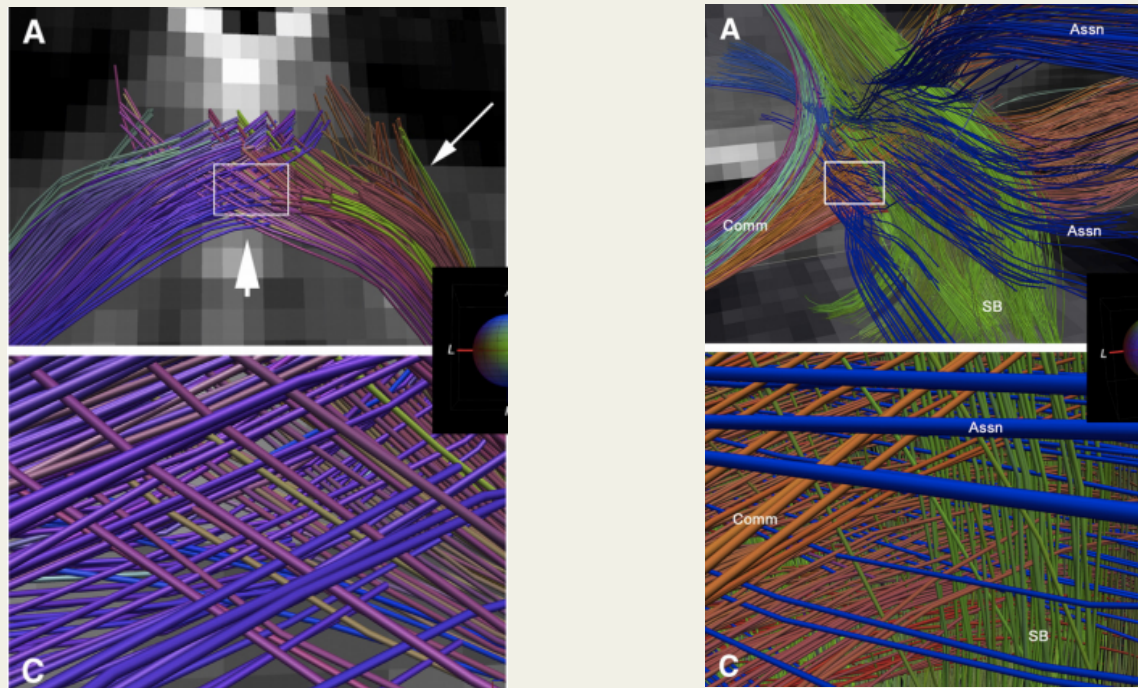
Inferring tissue biology from diffusion

- Diffusion is very sensitive to tissue changes and can help generate hypotheses about potential biological processes.



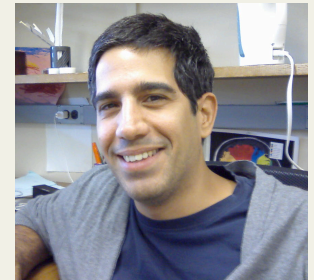
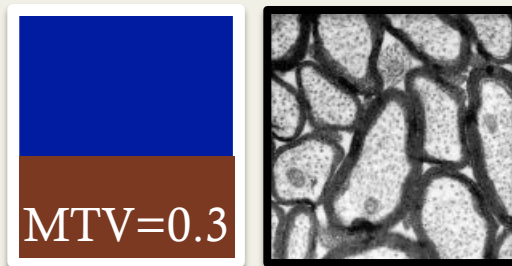
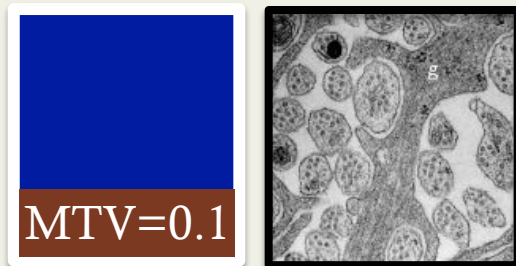
Diffusion is affected by many tissue properties

- It's amazing that water diffusion correlates with behavior (e.g., Klingberg et al., 2000).
- The relationship between water diffusion and tissue biology is not straightforward (Beaulieu, 2002; Jones, Knosche & Turner 2013).
- Additional measurements modalities will help constrain our models.

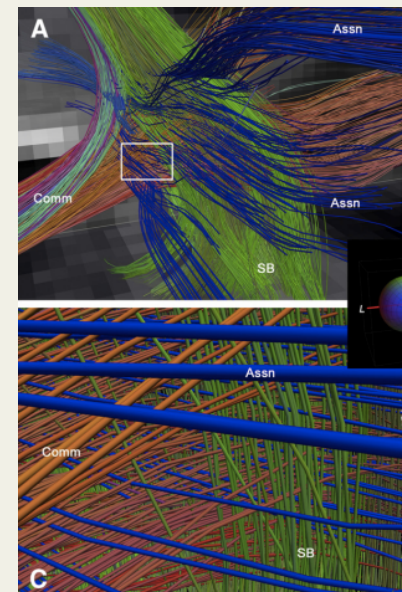
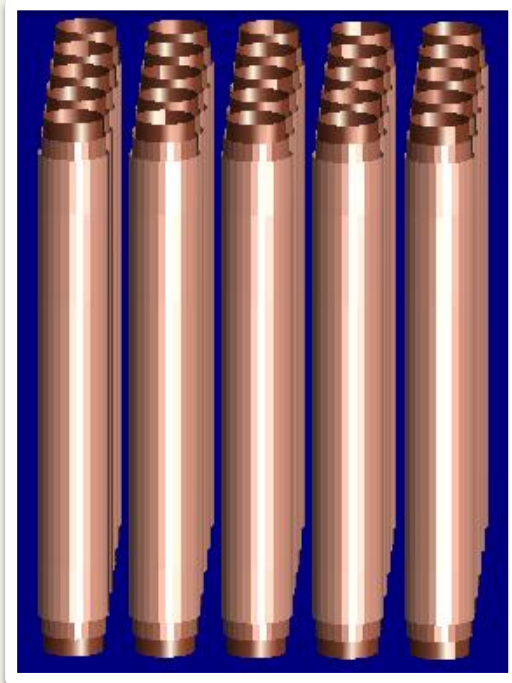
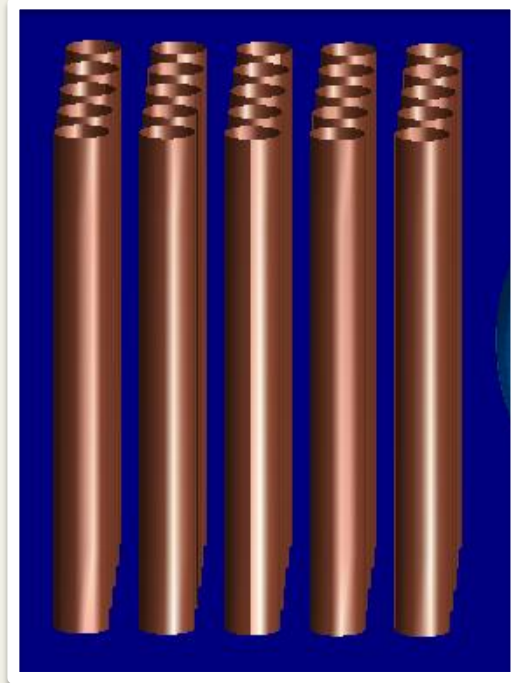


Wedeen et al., (2008, 2012)

Quantitative MRI measurements of tissue volume and composition

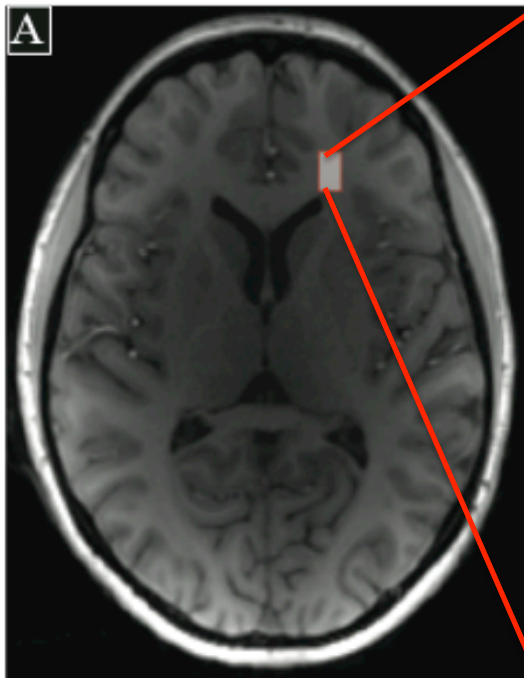


Aviv Mezer

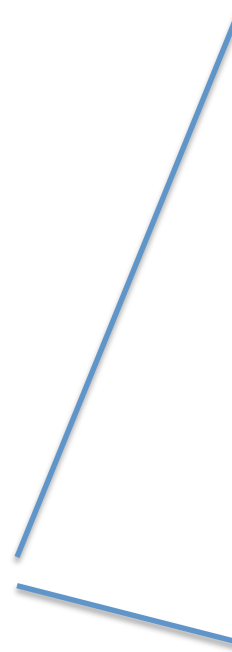


In vivo histology with quantitative MRI

Aviv Mezer
Hebrew University

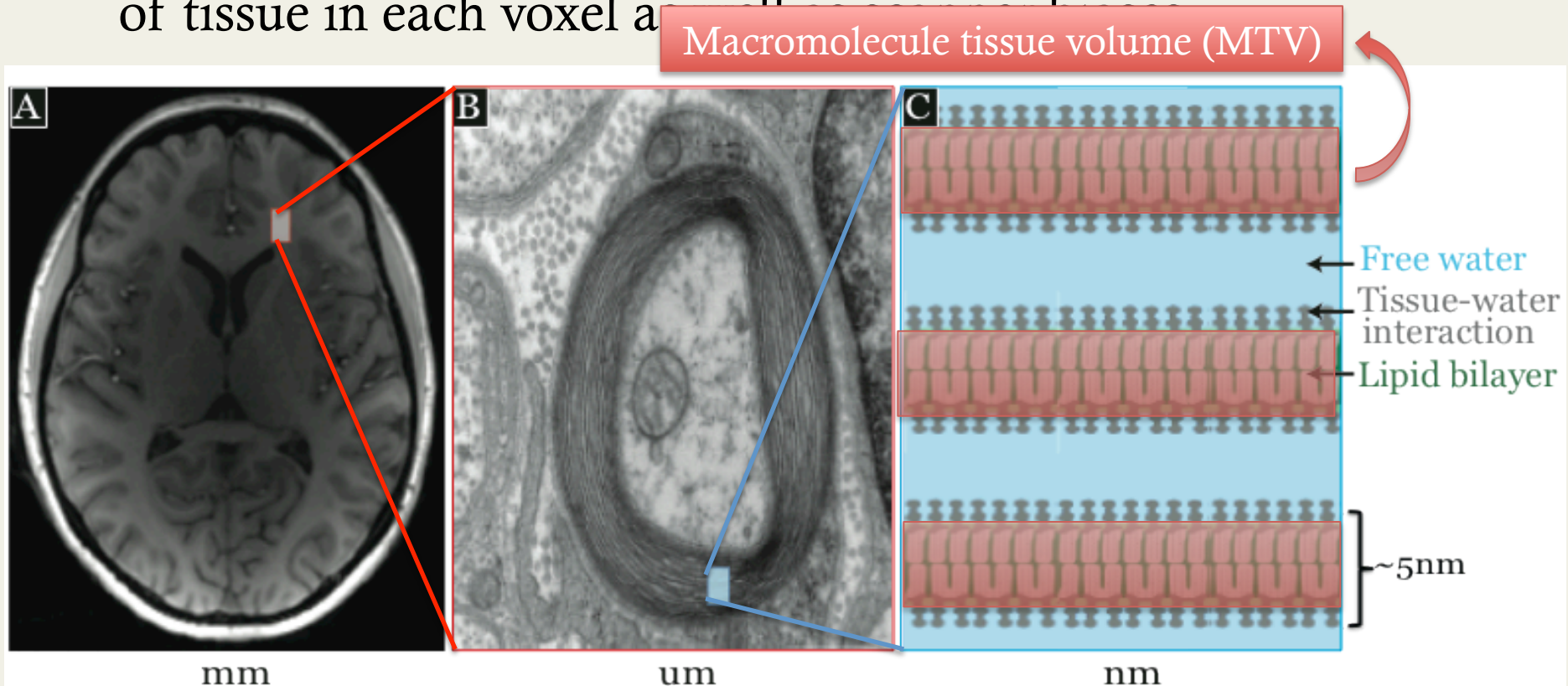


mm



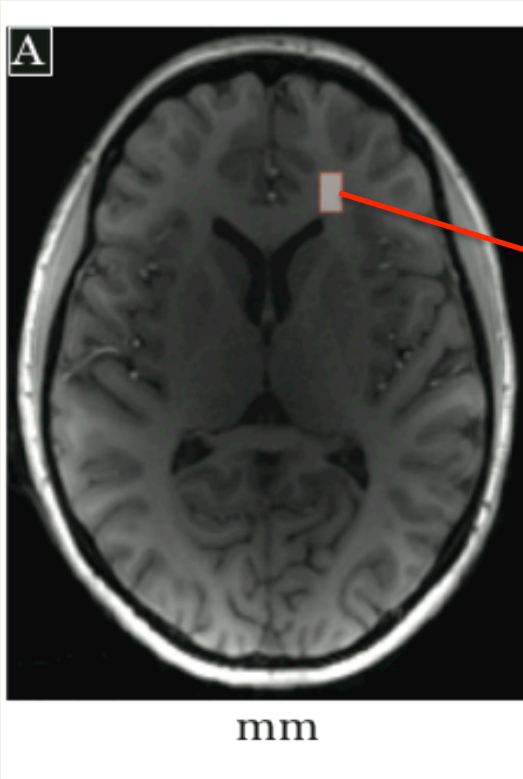
In vivo histology with quantitative MRI

- MR signals (T1) from water protons change when the protons interact with membranes.
- T1 image intensity depends on the **amount** and **composition** of tissue in each voxel and the **macromolecule tissue volume (MTV)**.



In vivo histology with quantitative MRI

- MR signals (T1) from water protons change when the protons interact with membranes.
- T1 image intensity depends on the amount and composition of tissue in each voxel as well as scanner biases.



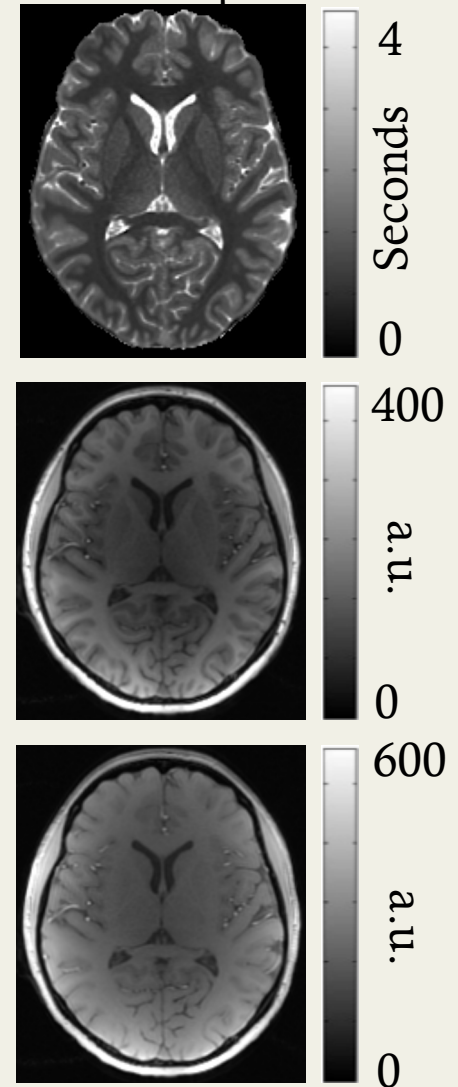
$$\text{Image intensity} = f(\overset{\text{coil gain}}{\cancel{S_0}}, \overset{\text{scan parameters}}{\cancel{TR}}, \overset{\text{tissue properties}}{T1}, \overset{\text{tissue properties}}{MTV})$$

In the white matter T1 relaxation rate is principally driven by myelin content (Stuber et al., 2014).

What does “quantitative MRI” mean?

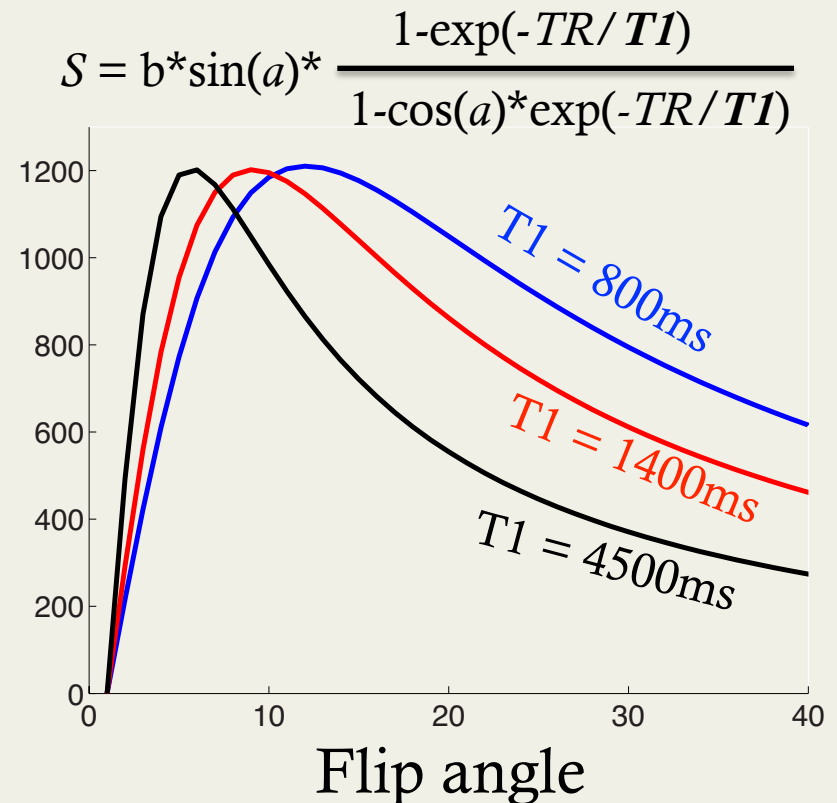
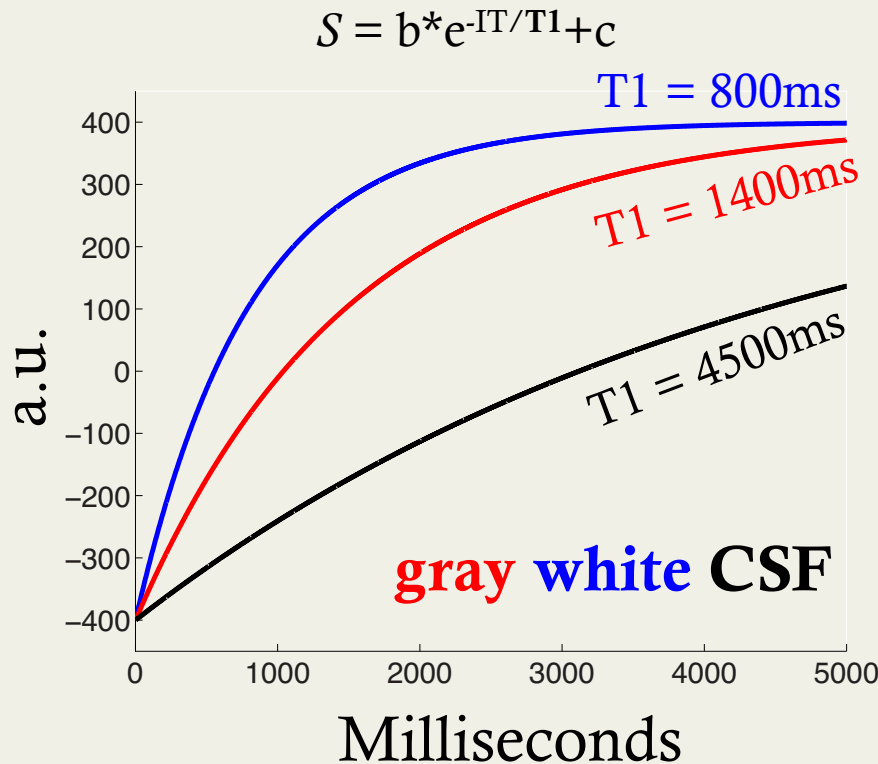
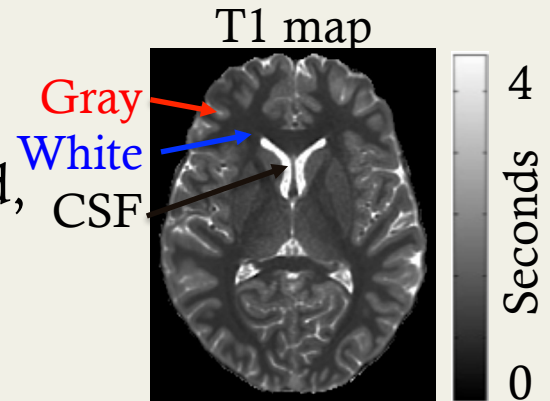
- **T1 (s)** - The T1 relaxation rate is a physical property of water protons in a magnetic field, has units, and does not depend on scanner hardware/pulse sequence.

T1 map

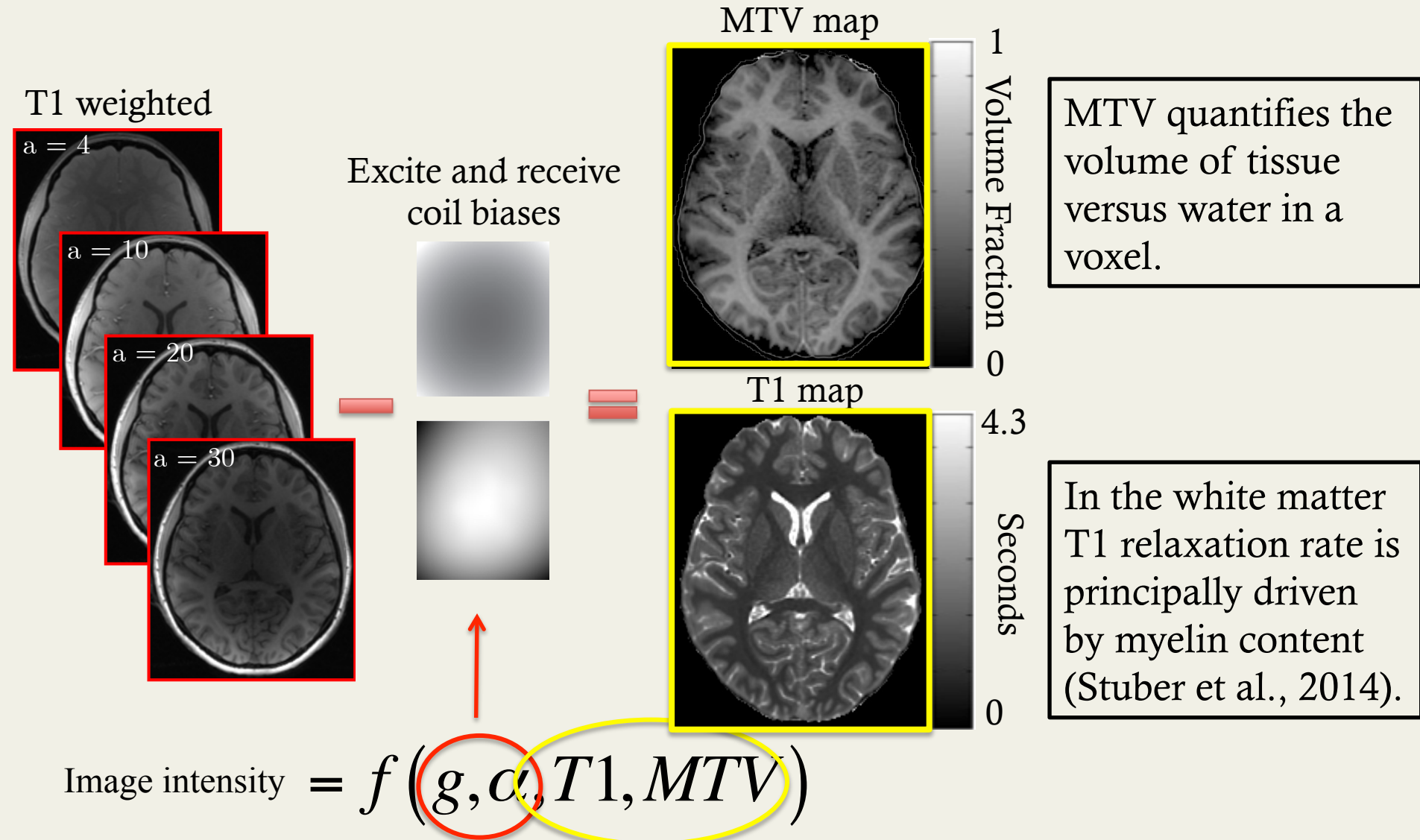


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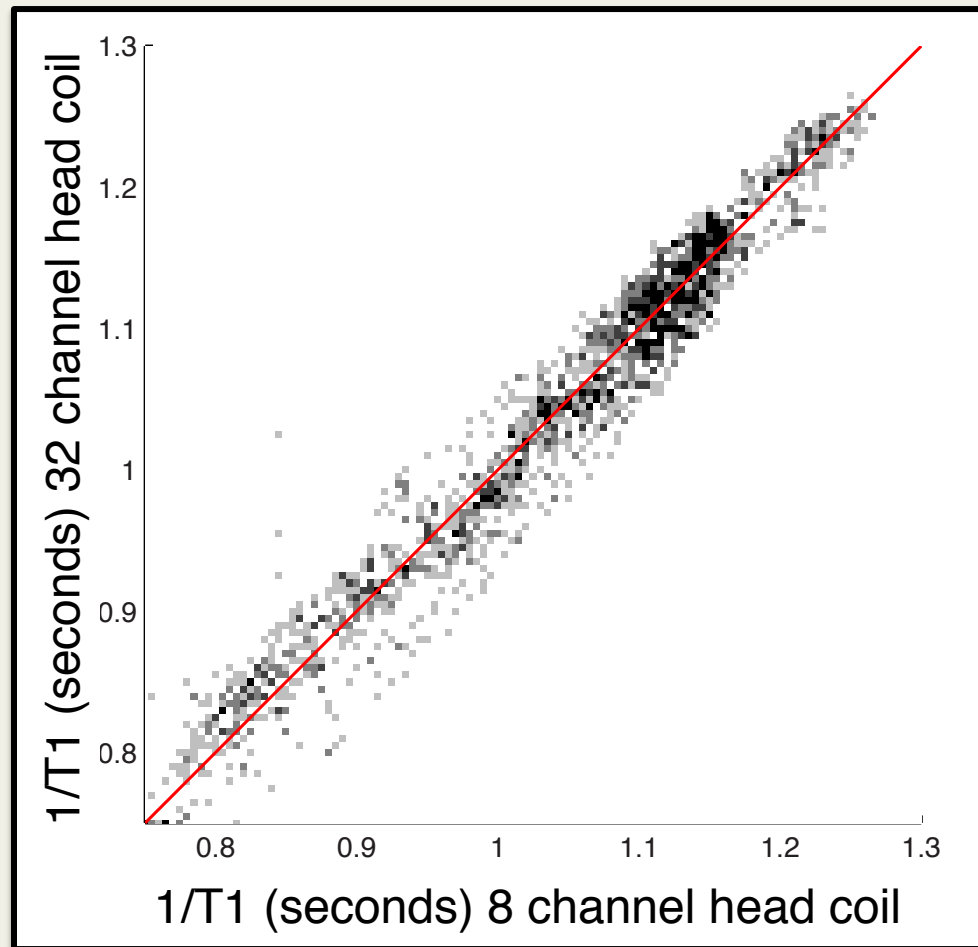
From images to quantitative tissue maps



MTV quantifies the volume of tissue versus water in a voxel.

In the white matter T1 relaxation rate is principally driven by myelin content (Stuber et al., 2014).

Quantitative MRI measures are independent of scanner hardware



Summary: Quantitative MRI

- MRI can be used to quantify many important properties of the tissue.
 - Volume of tissue macromolecules (**MTV**).
 - **T1** relaxation rate is sensitive to myelin (Stuber et al., 2014).
- Quantitative MRI measurements are independent of the specific scanner hardware and pulse sequence.
 - Opens up new diagnostic applications.

Outline

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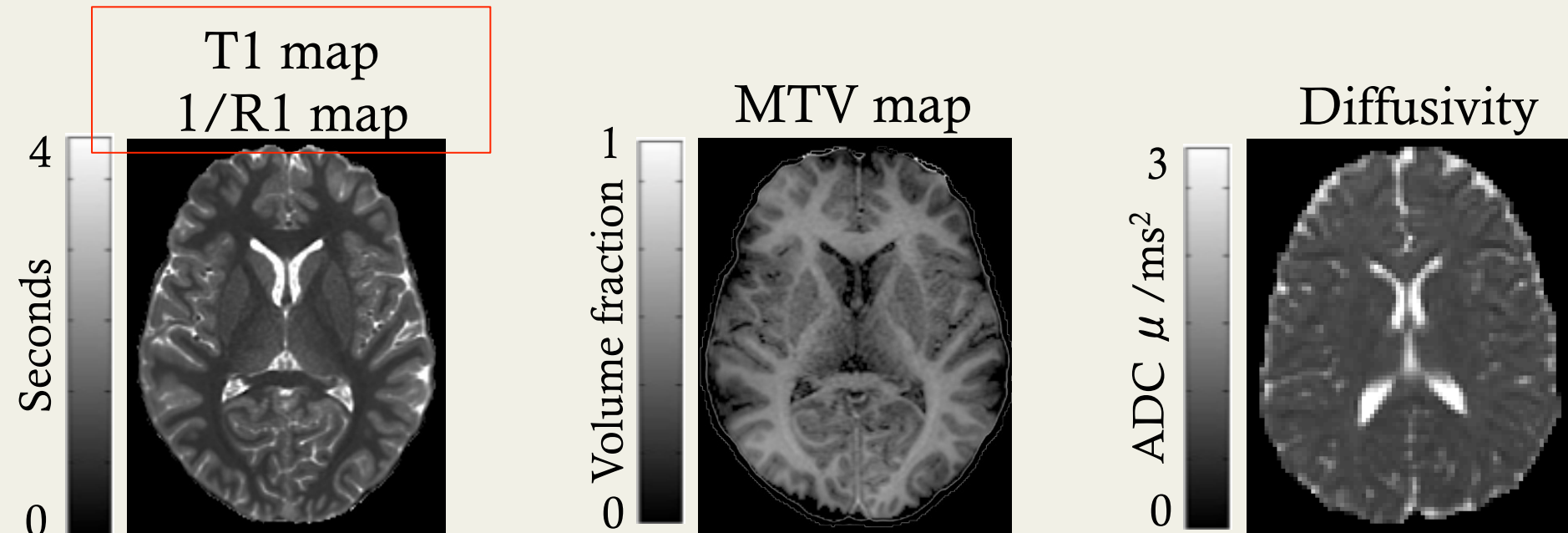
In vivo histology:

Combining measures to model brain tissue

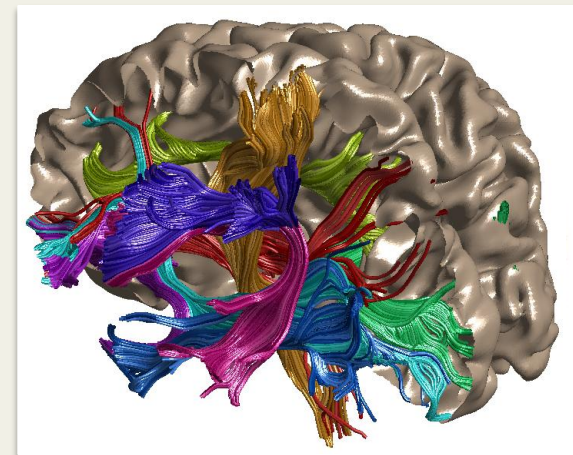
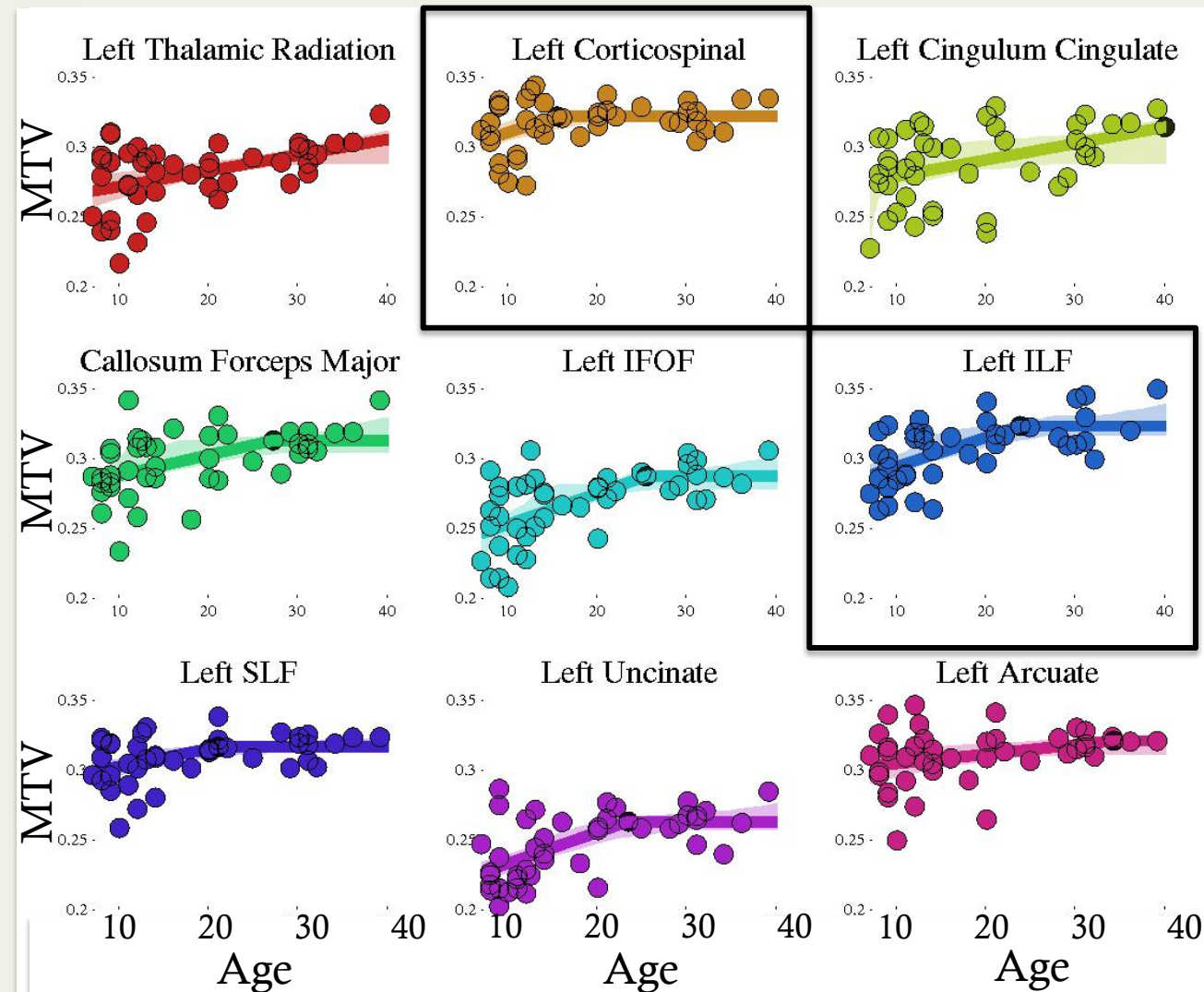
- What can we learn about development with qMRI:
 - Do different types of tissue have distinct maturational time-courses (e.g., myelin vs. astrocytes)?
 - Which properties of the white matter are related to behavior?
 - Can we model how properties of the white matter affect cortical computation (i.e., why do white matter measures predict behavior)?

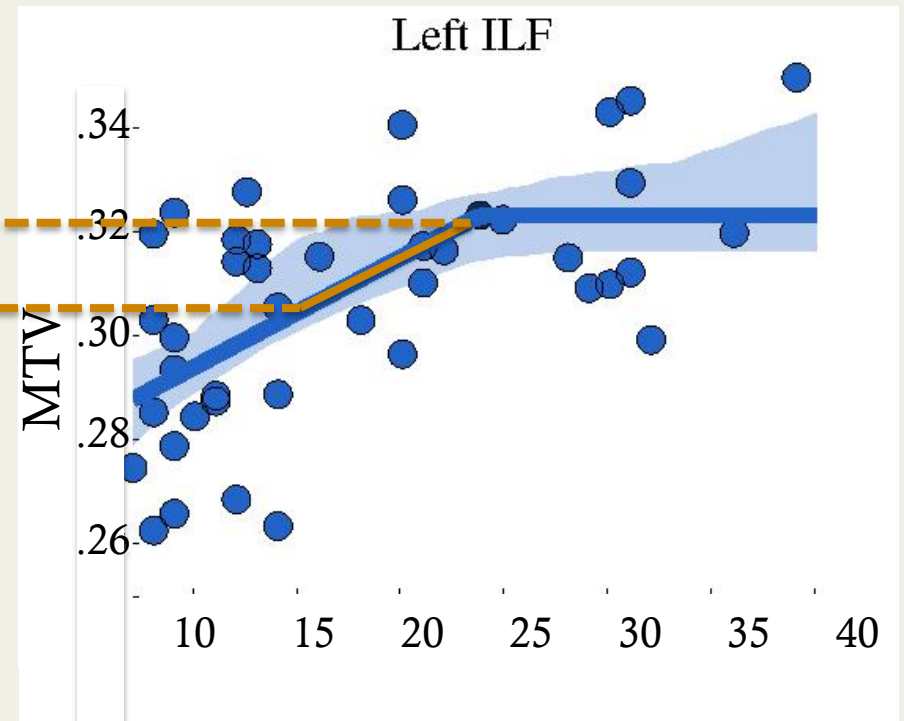
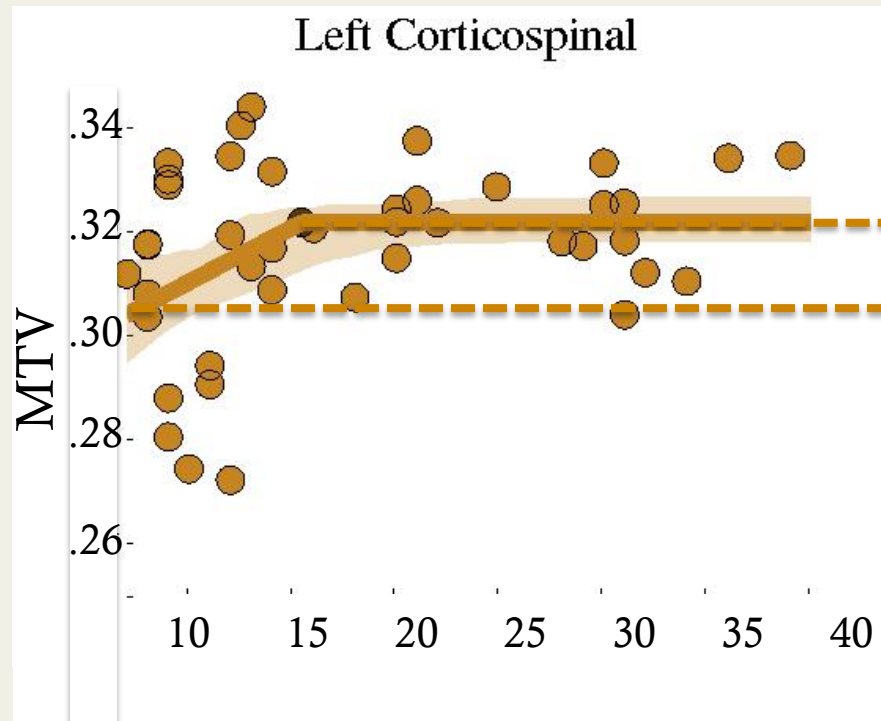
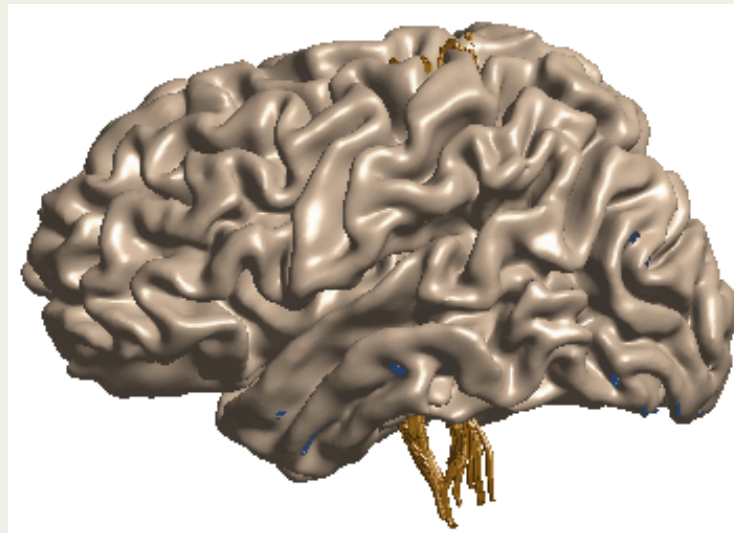
In vivo histology:

Combining measures to model brain tissue



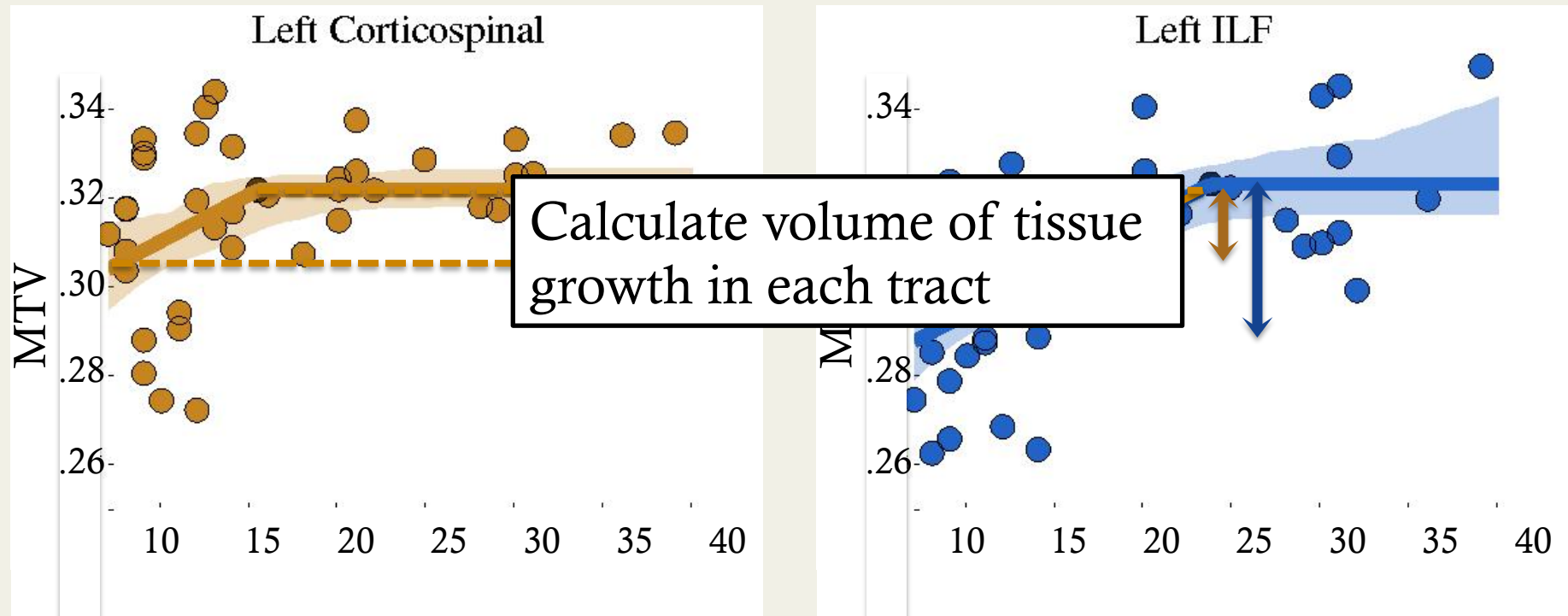
Measuring the creation of new tissue in the developing brain





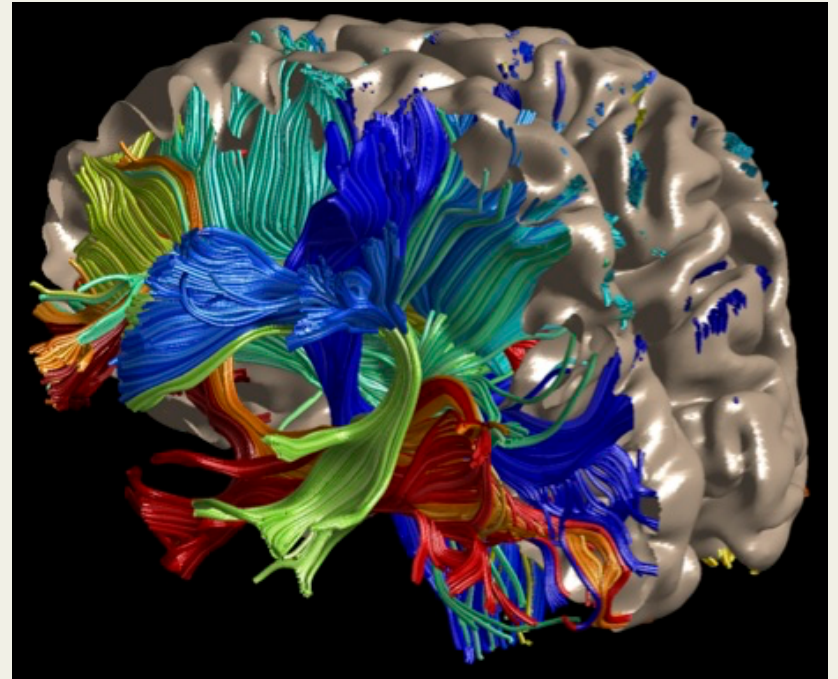
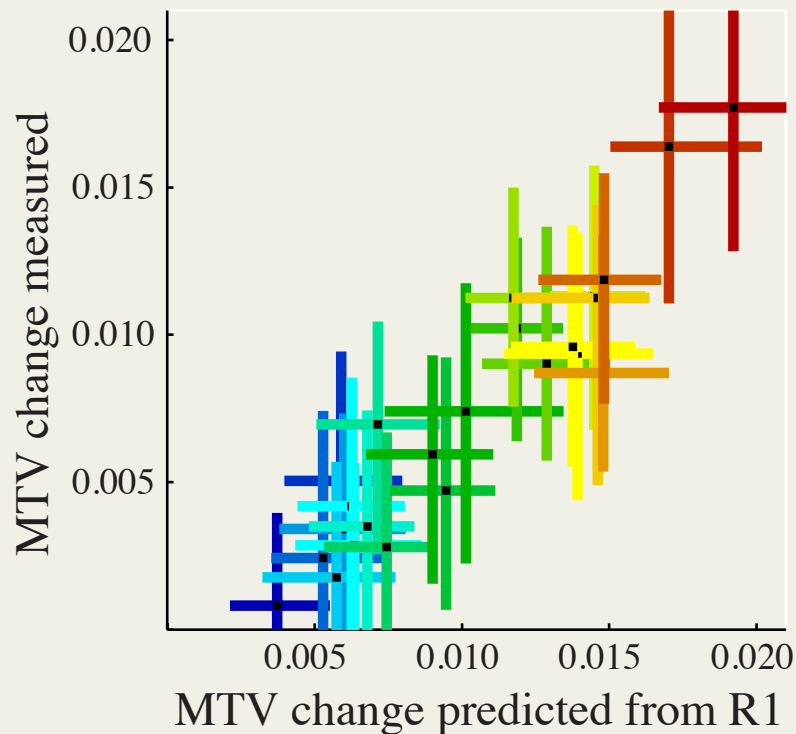
Does each qMRI parameter measure the same underlying biological properties?

Can we detect multiple developmental processes in the white matter?



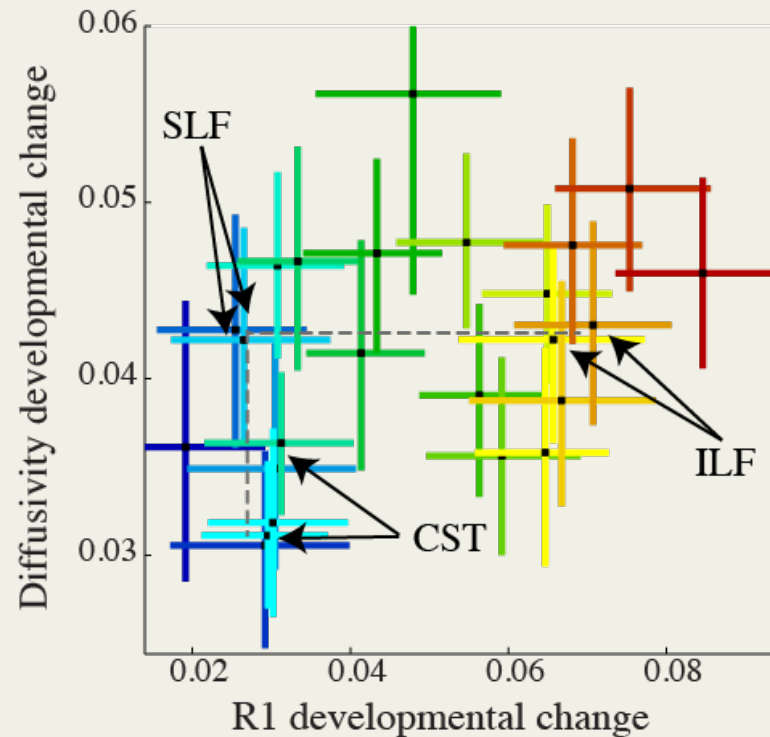
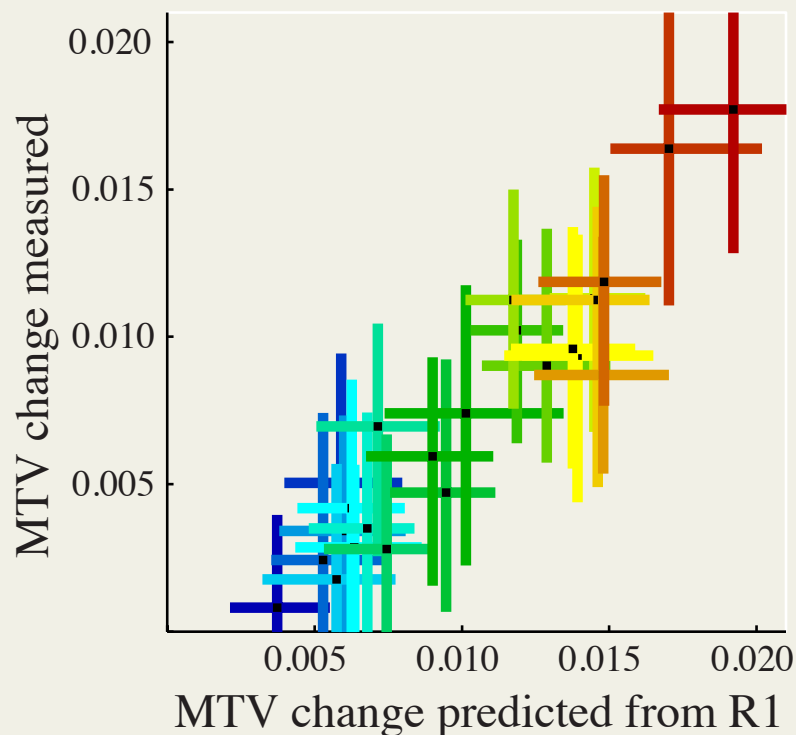
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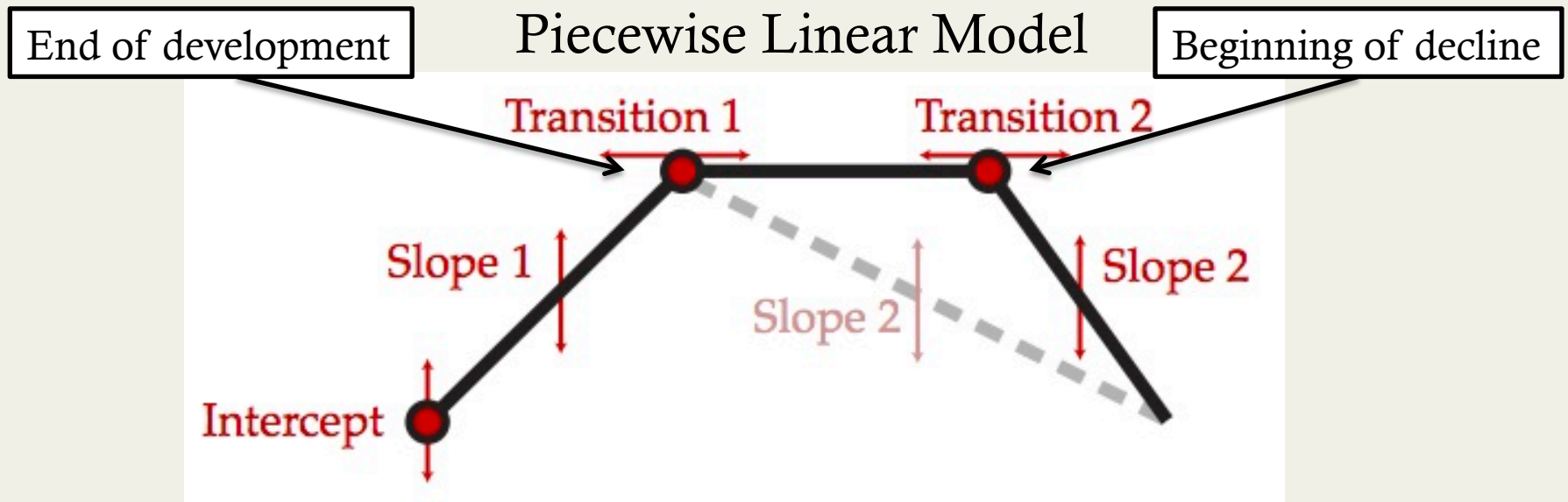
Can we detect multiple developmental processes in the white matter?



- R1 and MTV are sensitive to the same developmental processes.
- Diffusion is sensitive to independent processes.

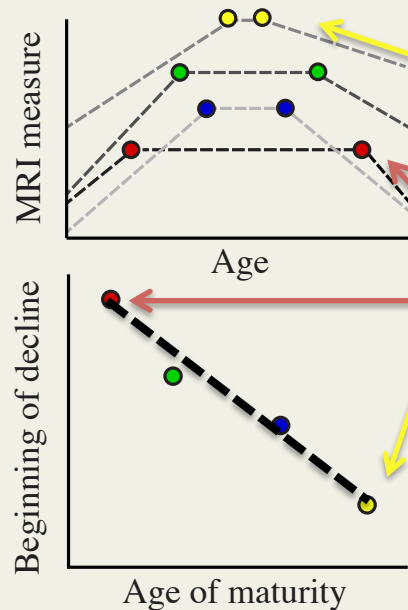
Last-in-first-out hypothesis: Late developing brain regions are more susceptible to degenerative processes.

- Are the last regions to develop, the first to degenerate?
 - Important implications for models of aging and disease.
- How might we test this hypothesis?
 - Implement a model and test the fit to the data



Last-in-first-out hypothesis: Late developing brain regions are more susceptible to degenerative processes.

a



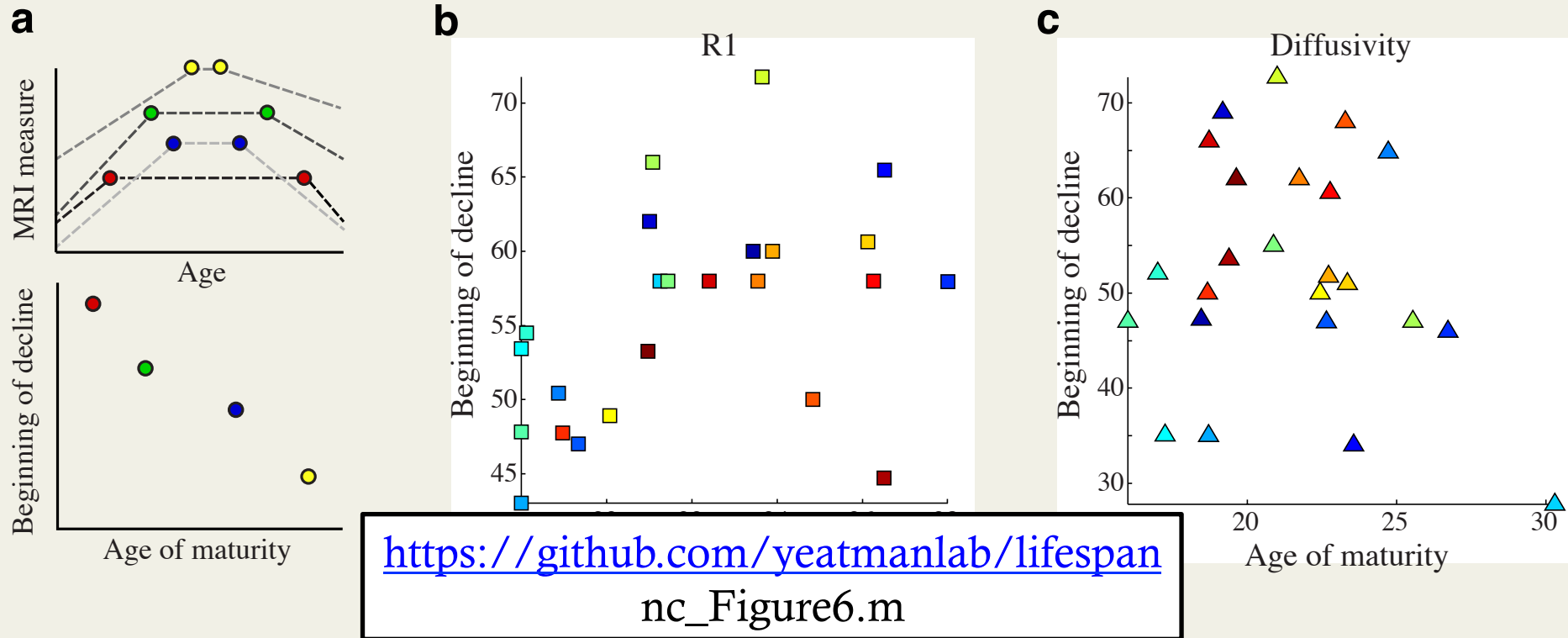
Prediction 1: Tracts that develop late decline early

Prediction 2: Tracts that develop early decline late

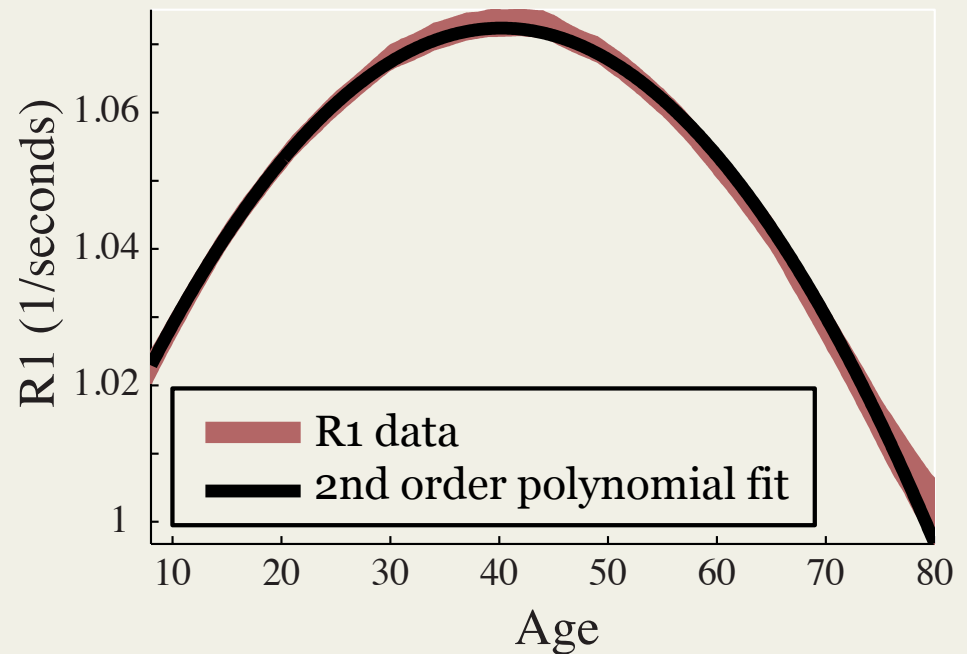
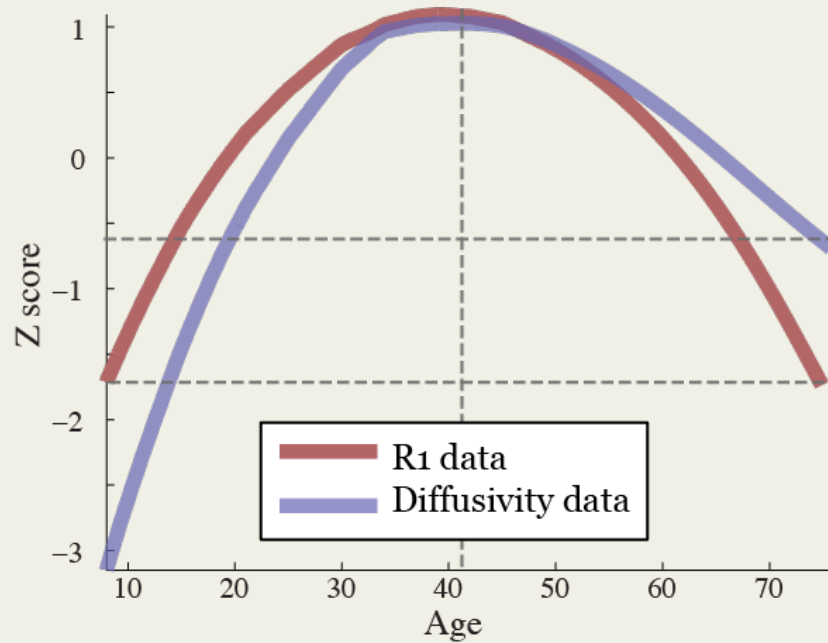
Prediction 3: Negative correlation between the two transition points

Last-in-first-out hypothesis: Late developing brain regions are more susceptible to degenerative processes.

- There is no relationship between the age of maturation and the age of degeneration for R1 or diffusivity.
- We can reject the Last-in-first-out hypothesis.

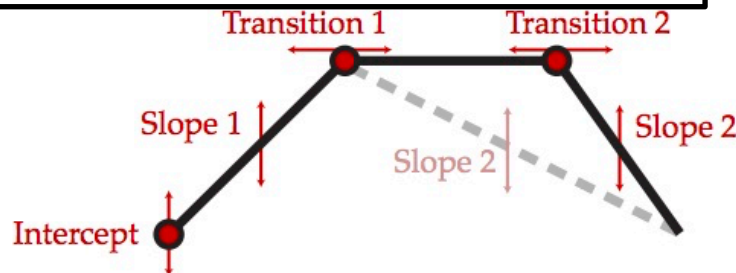
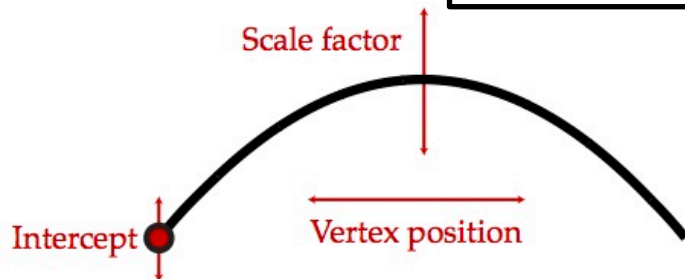


Symmetry hypothesis: The amount of growth predicts the amount of decline

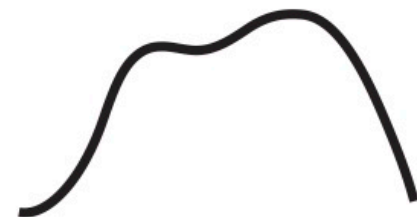


<https://github.com/yeatmanlab/lifespan>
nc_Figure4.m nc_Figure5.m

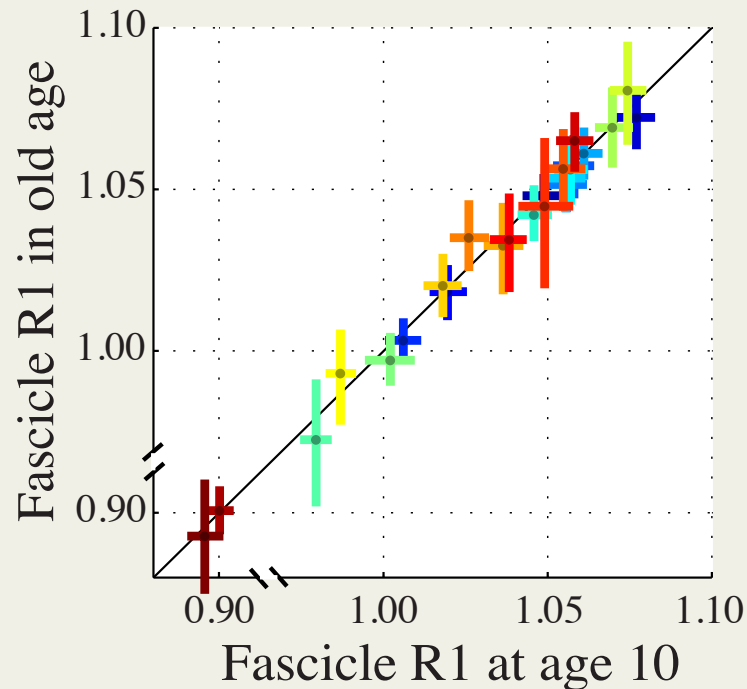
a. Second order polynomial



c. Local regression



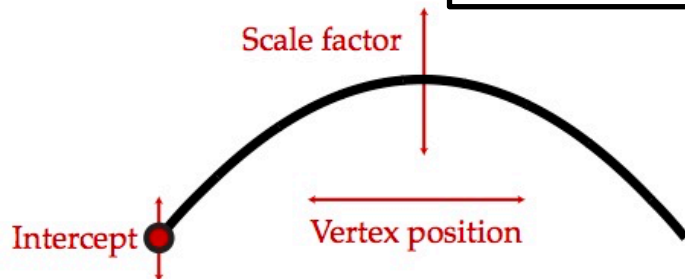
Symmetry hypothesis: The amount of growth predicts the amount of decline



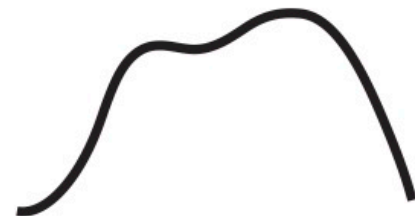
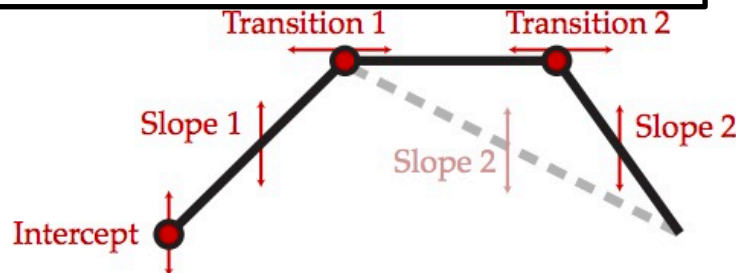
1. Testing the prediction of the parabolic model: Lifespan changes should be symmetric.
2. Growth of myelin during development predicts degeneration during aging.

<https://github.com/yeatmanlab/lifespan>
nc_Figure5.m

a. Second order polynomial



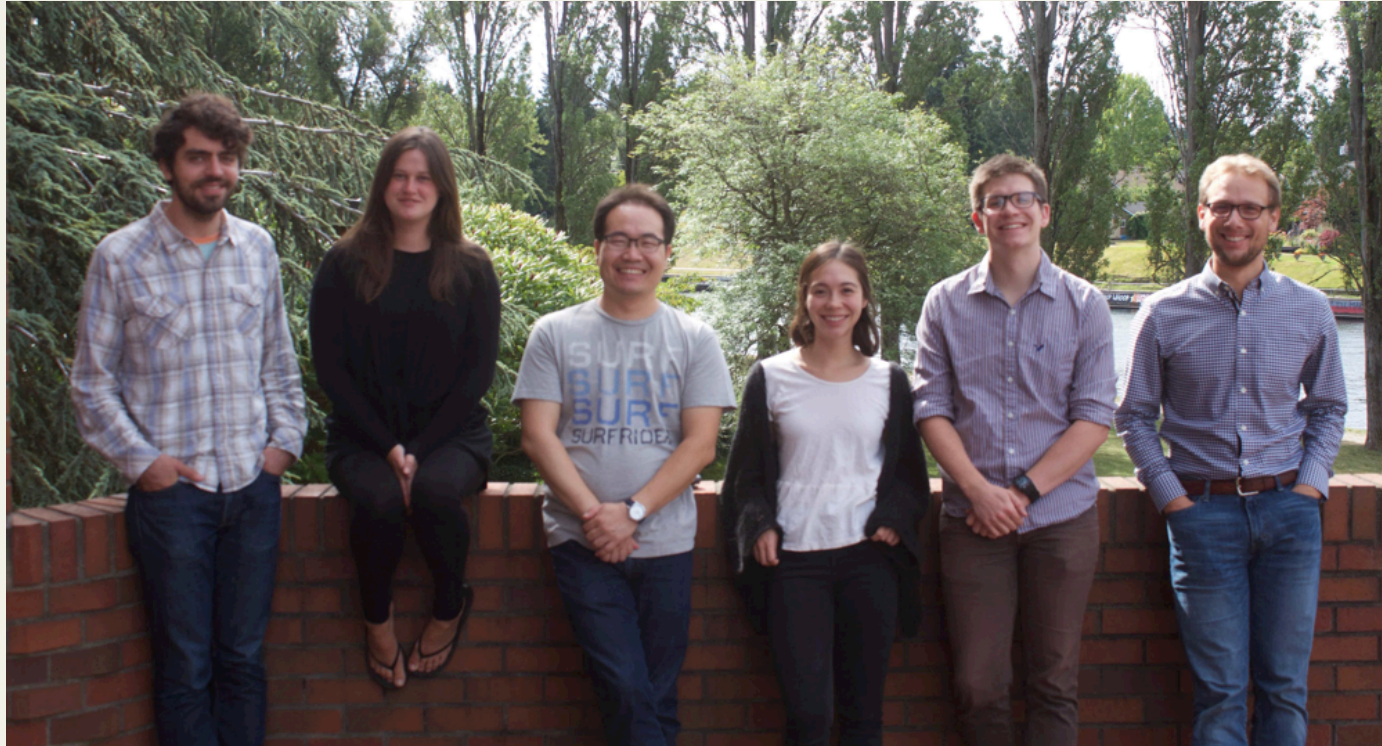
c. Local regression



Summary

- The time-courses of R1 and diffusion changes demonstrate that multiple biological processes drive changes in the white-matter over the lifespan.
 - qMRI can dissociate different tissue changes.
- There is not a systematic relationship between the age of maturation and degeneration
 - Last-in-first out does not fit the data
- A symmetric model predicts R1 changes over the lifespan.
 - Models provide insight into mechanisms and generate testable predictions.
- We can define, test and interpret models at the level of voxels, tracts and lifespan maturation.

Thank you!



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Removing bias and computing MTV

Each coil sees the same underlying MTV value but has its own gain function

Solve for the each coil's gain function to uncover the true MTV value

$$g_i * (1 - MTV) \quad \min_{g_i} \left\{ \sum (MTV_i - MTV)^2 \right\}$$

