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350

Diffusion MRI of Brain Tissue

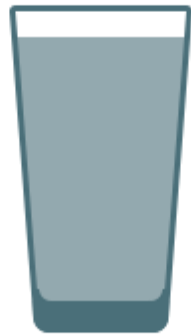
Importance of Axonal Trajectory

JAN BRABEC, JAN PALLON, MARKUS NILSSON



Introduction: Magnetic Resonance Imaging

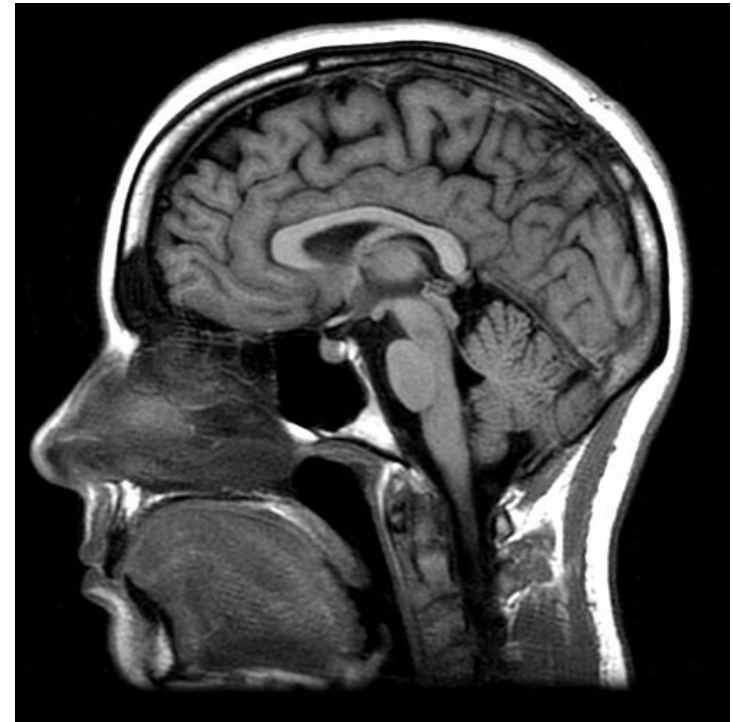
- Imaging modality.
- Probes magnetic properties of tissue.
- Probes properties of water.
- Most MR methods study macrostructure.
- dMRI aims at probing microstructure.



[1]

Conventional MR image of head

[2]



Introduction: Clinical application of dMRI

Stroke Diagnostics

Neurosurgery Planning

CT

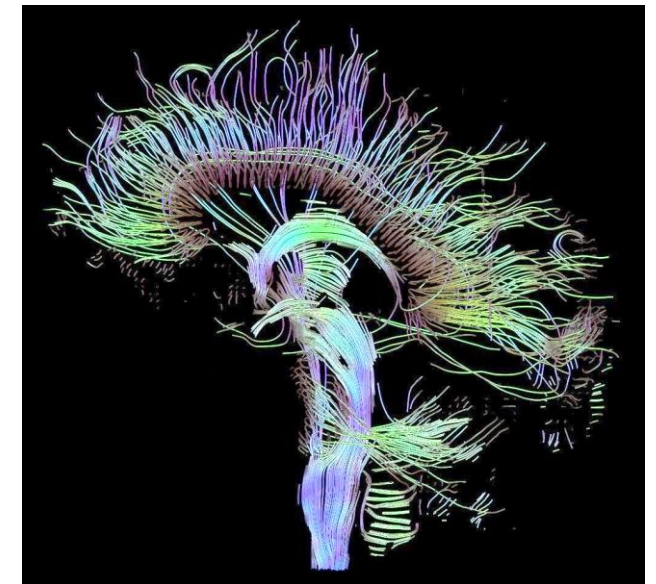
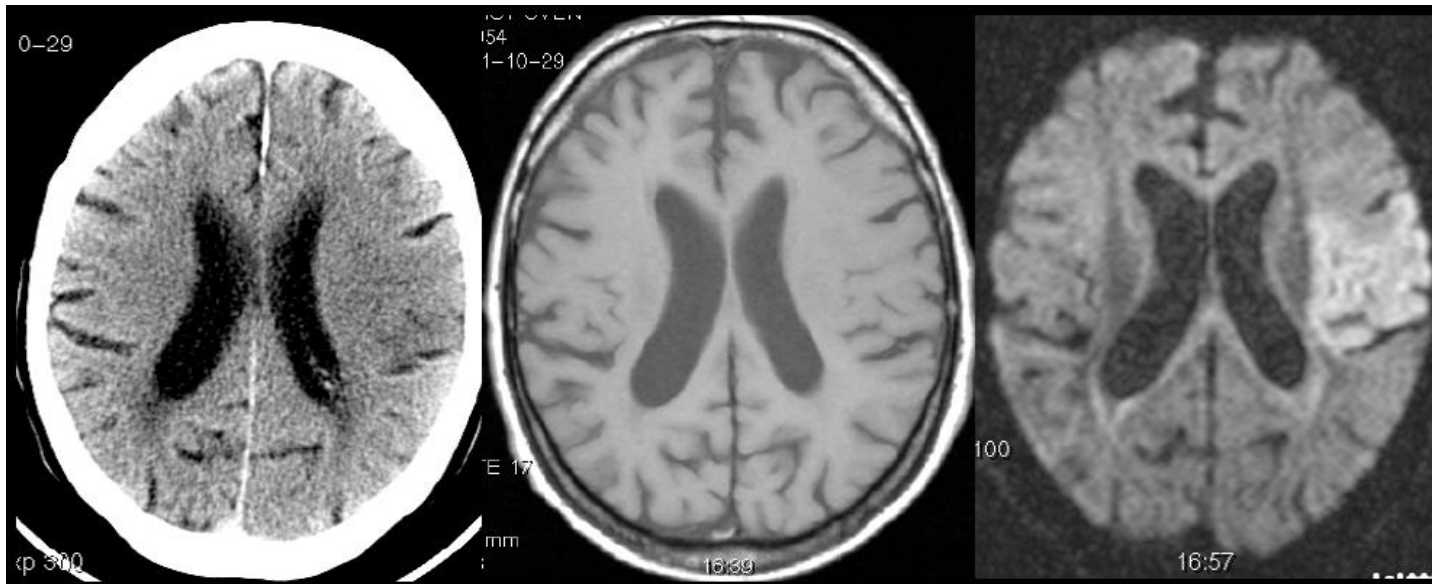
T1W

DWI

[1]

Tractography

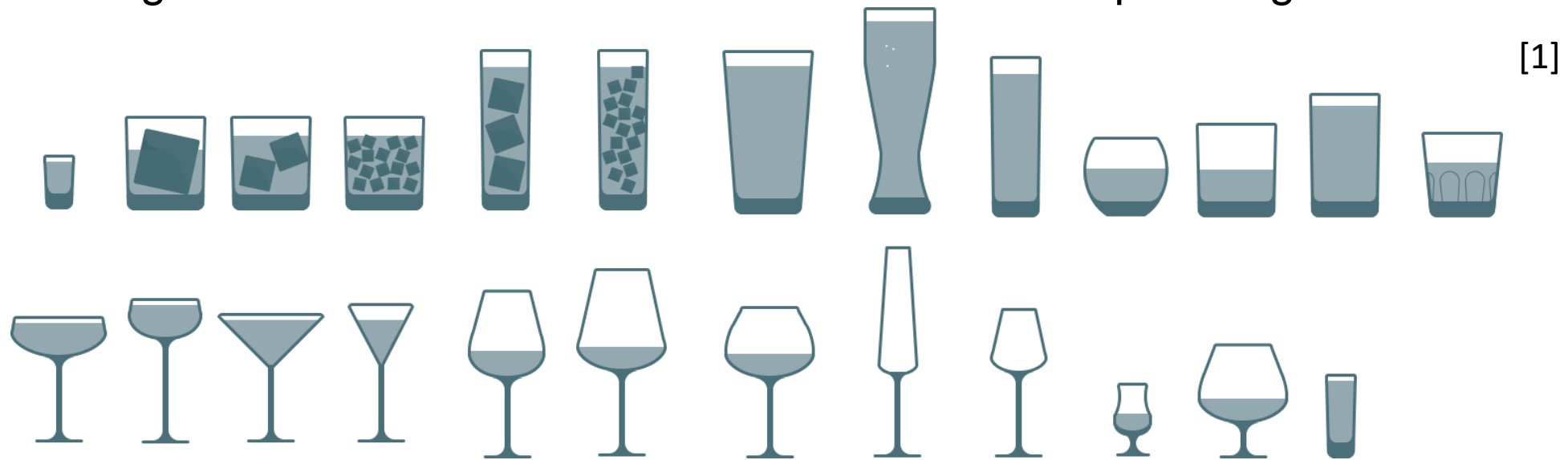
[2]



[1] In-house Lund data; [2] Schultz (2006), *Wikimedia*

Introduction: Principles of dMRI

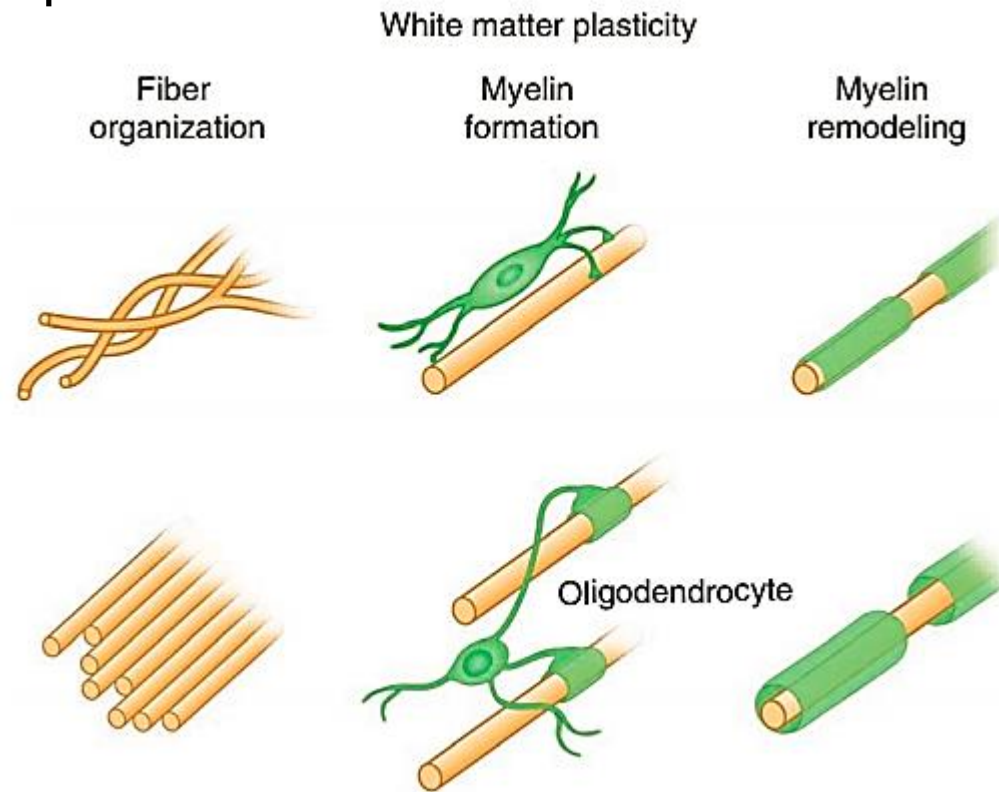
- Diffusion MRI studies diffusion of water.
- Diffusion is modulated by surrounding environment.
- Diffusion MRI contains indirectly information on the geometry of the environment.
- Different geometries can be linked to different functions/pathological conditions.



[1] Koziel (2016), *bhcamouflage.us*

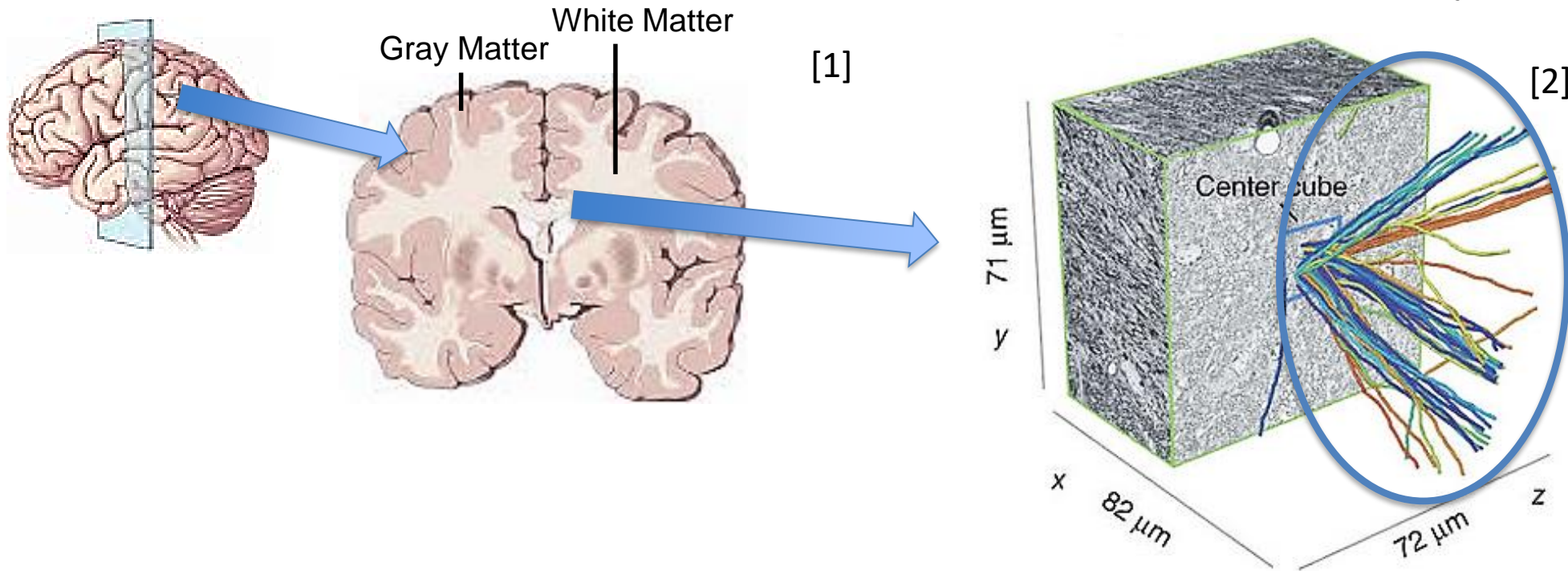
Introduction: Axon structure ↔ function

- Convey information as electrical impulses.
- Projections of nerve cells.
- Part of brain white matter.



Introduction: Axons \leftrightarrow wiring of the brain

- Architecture is called *axonal trajectory*.

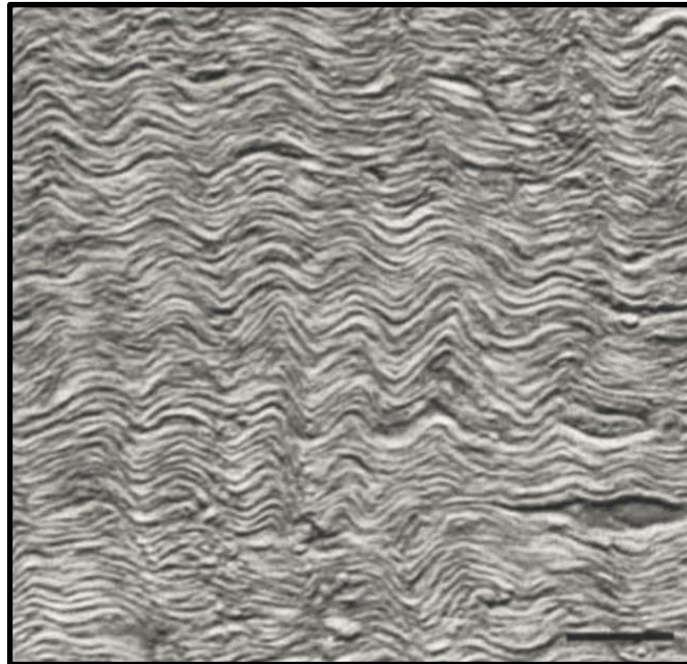


Axons:
their architecture:
axonal trajectory

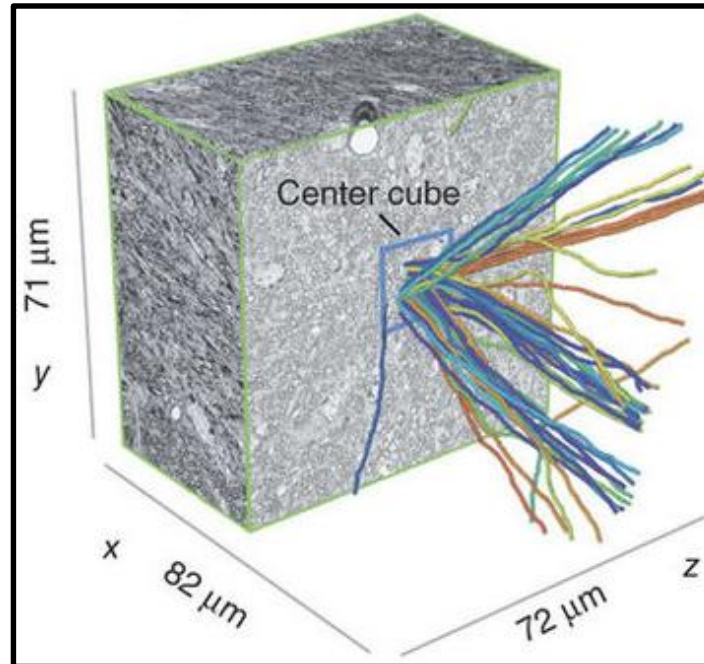
[1] AliceD (2015), *Stackexchange*; [2] Mikula et al. (2012), *Nat. Methods*

Introduction: Undulations are ubiquitous

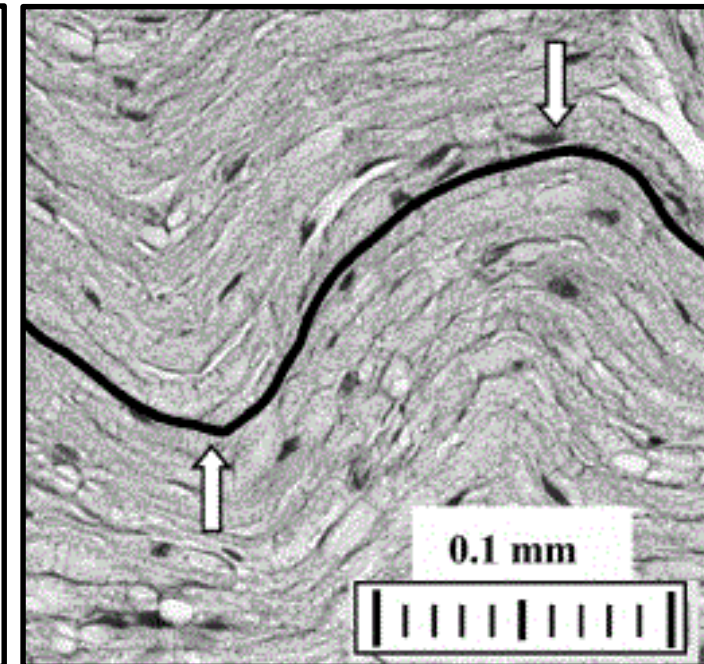
Optic Nerve [1]



Corpus Callosum [2]

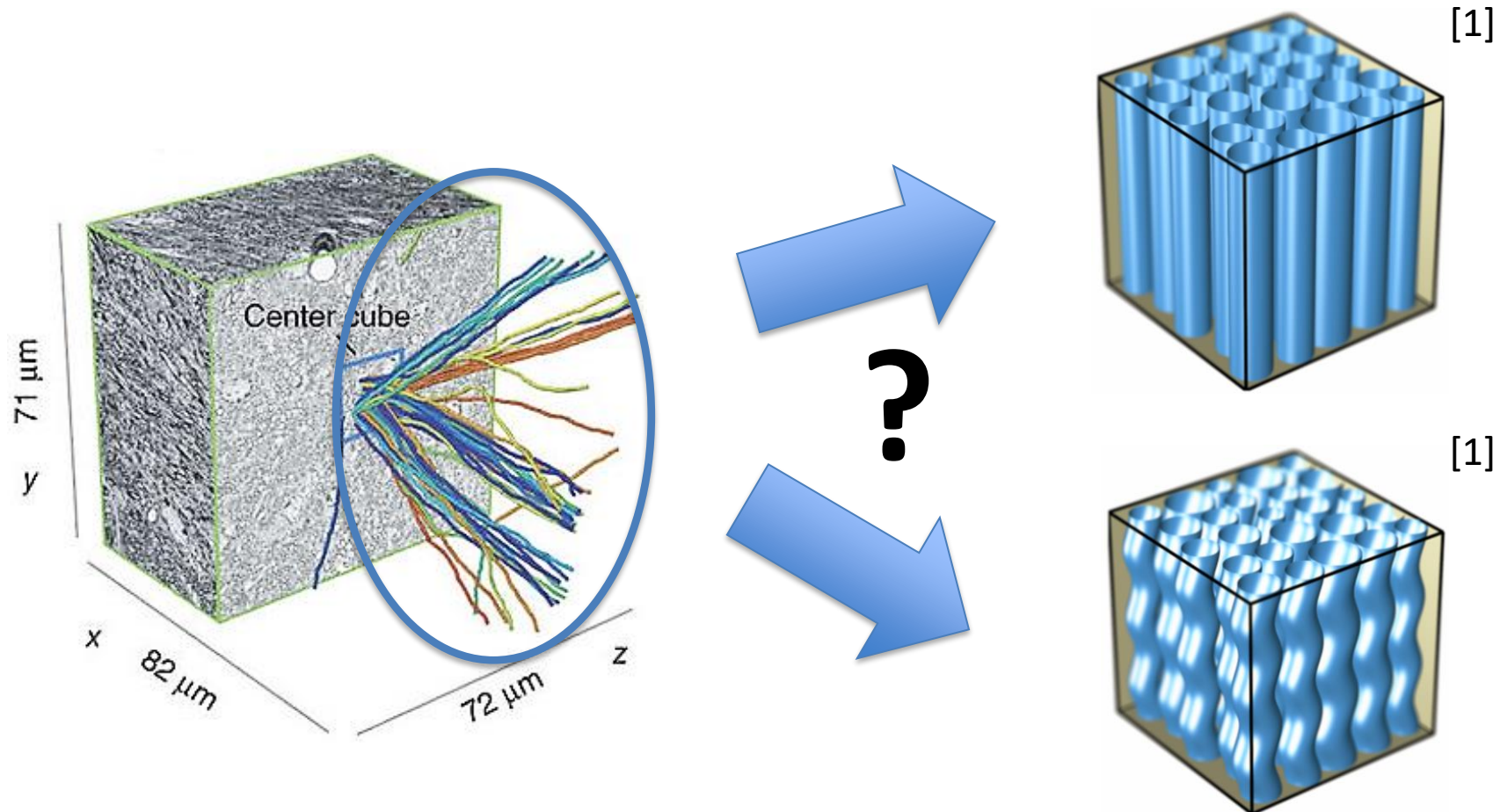


Phrenic Nerve [3]



[1] Jeffery (1996), *J. Comp. Neurol*; [2] Mikula et al. (2012), *Nat. Methods*; [3] Lontis (2009), *IEEE Trans. Biomed. Eng.*

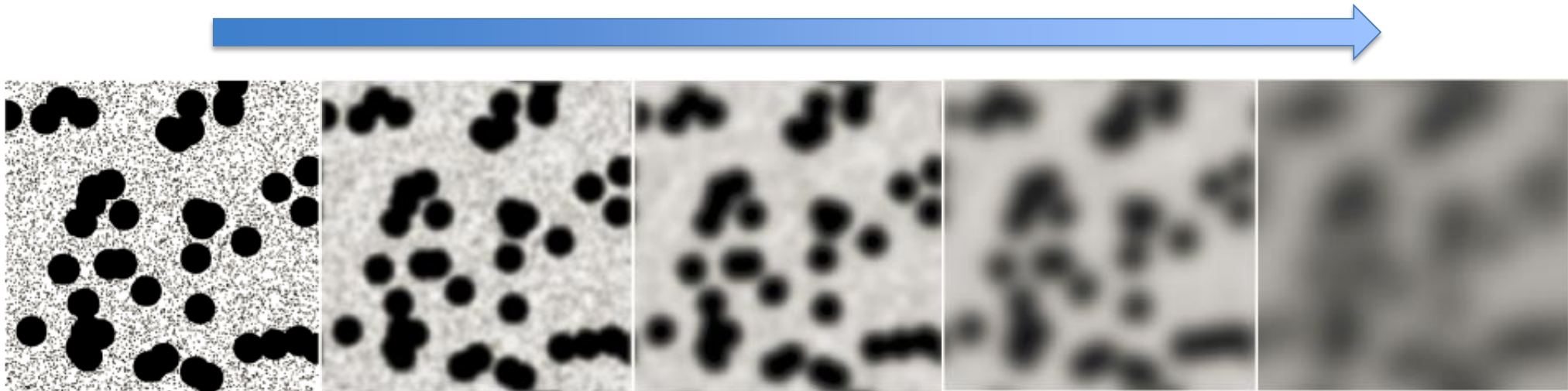
The Problem: How to model axons?



[1] Nilsson (2016), *Proc. ISMRM 24*

Methods: Physics at different diffusion times

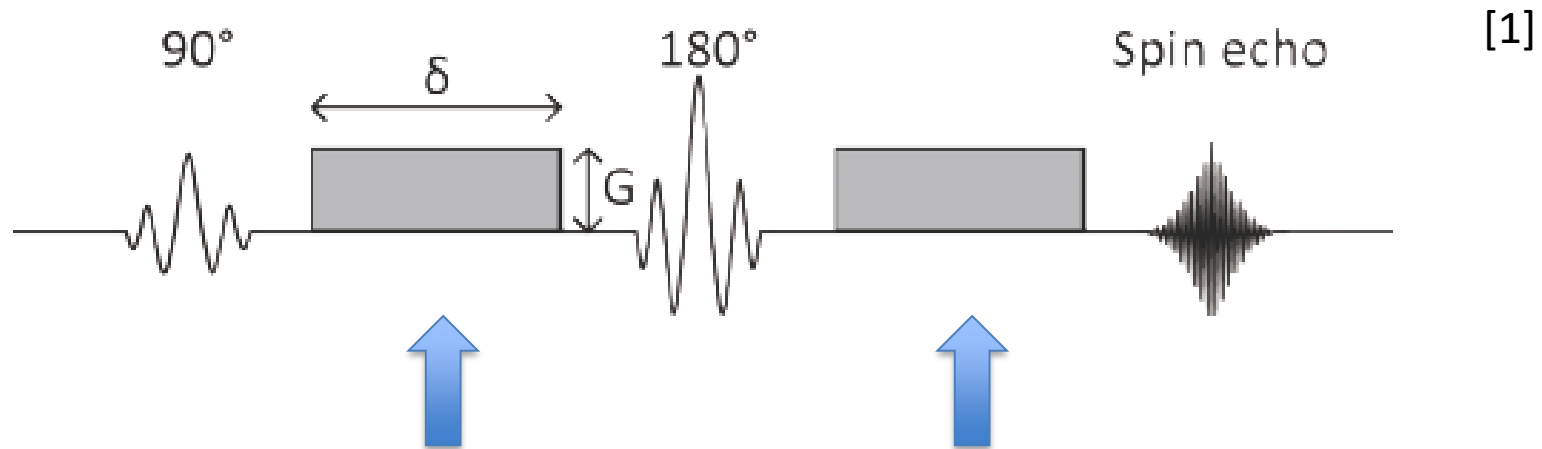
Diffusion Time



[1] Novikov (2016), *arXiv preprint*

Encoding information into signal

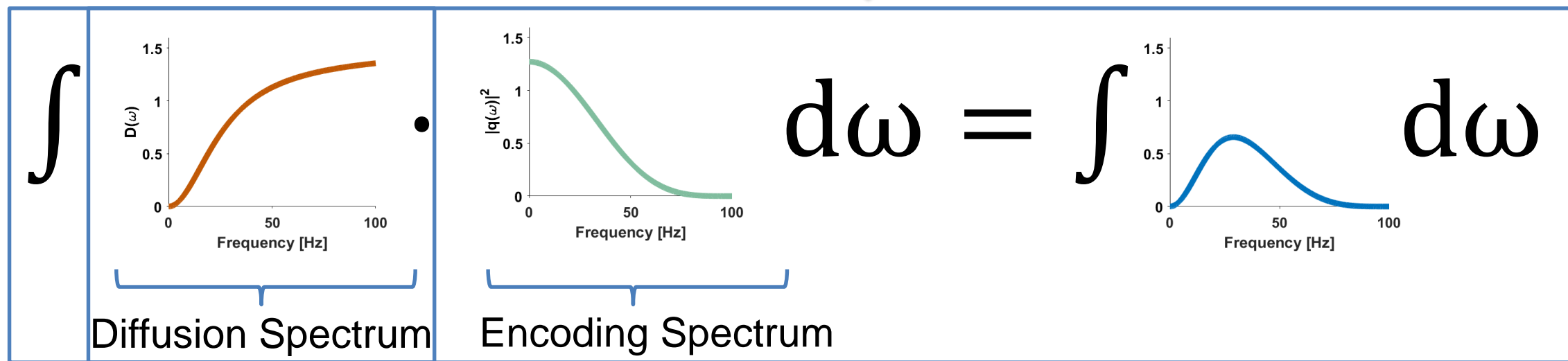
- Pulsed Gradient Spin Echo Sequence



[1] Winston (2012), *Quant Imaging Med Surg*.

Signal = $f(\text{Diffusion Spectrum, Gradient Waveform})^{[1]}$

$$S = e^{-\int D(\omega) \cdot |q(\omega)|^2 d\omega}$$



[1] Stepišnik (1993), *Physica B: Condensed Matter*

Aims

Model:

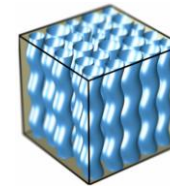


or

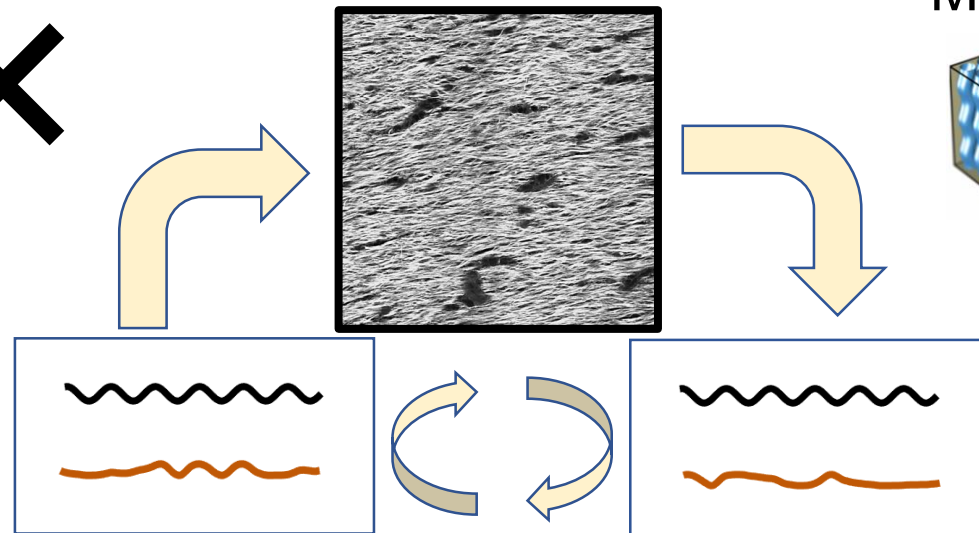
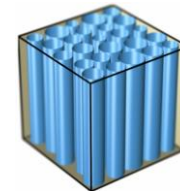


[1]

Model:



or

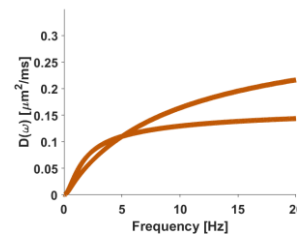
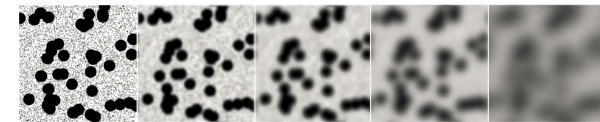


In-vivo application:

[2]



Method:

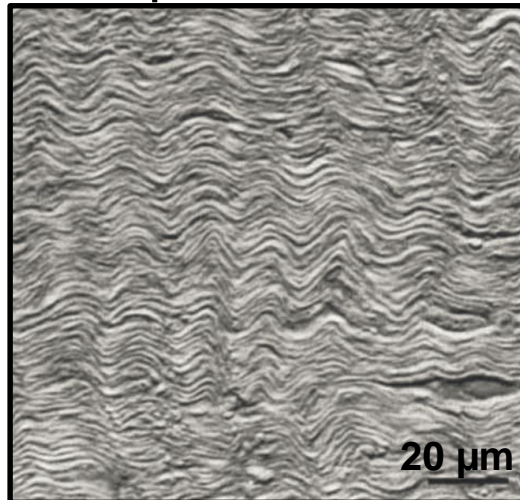


[1] Schilling (2018), *NeuroImage*; [2] _DJ_ (2005), *flickr.com*

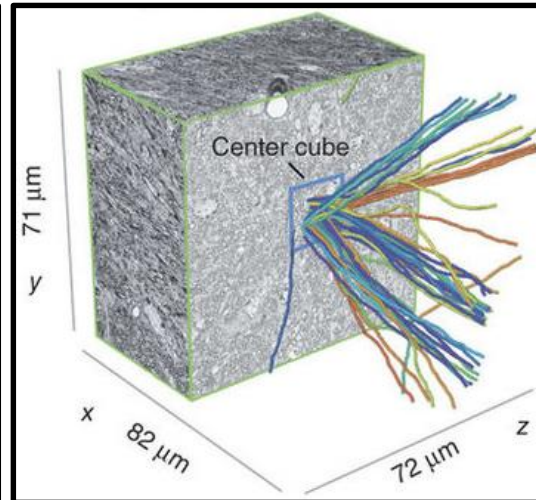
Model of axonal trajectory \leftrightarrow thin undulating wire

$$y(x) = a \cdot \sin(\Phi) = a \cdot \sin\left(2\pi \cdot \frac{x}{\lambda} + r(x)\right)$$

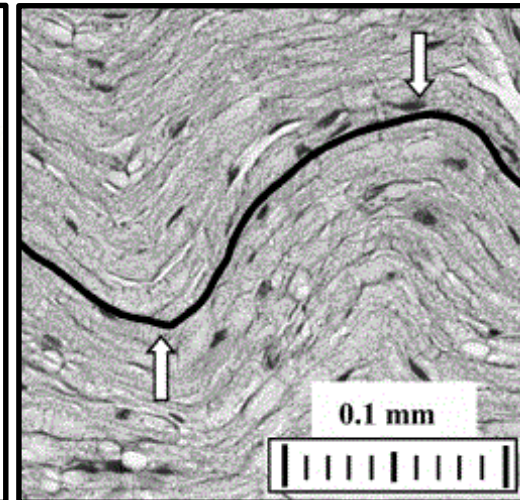
Optic Nerve



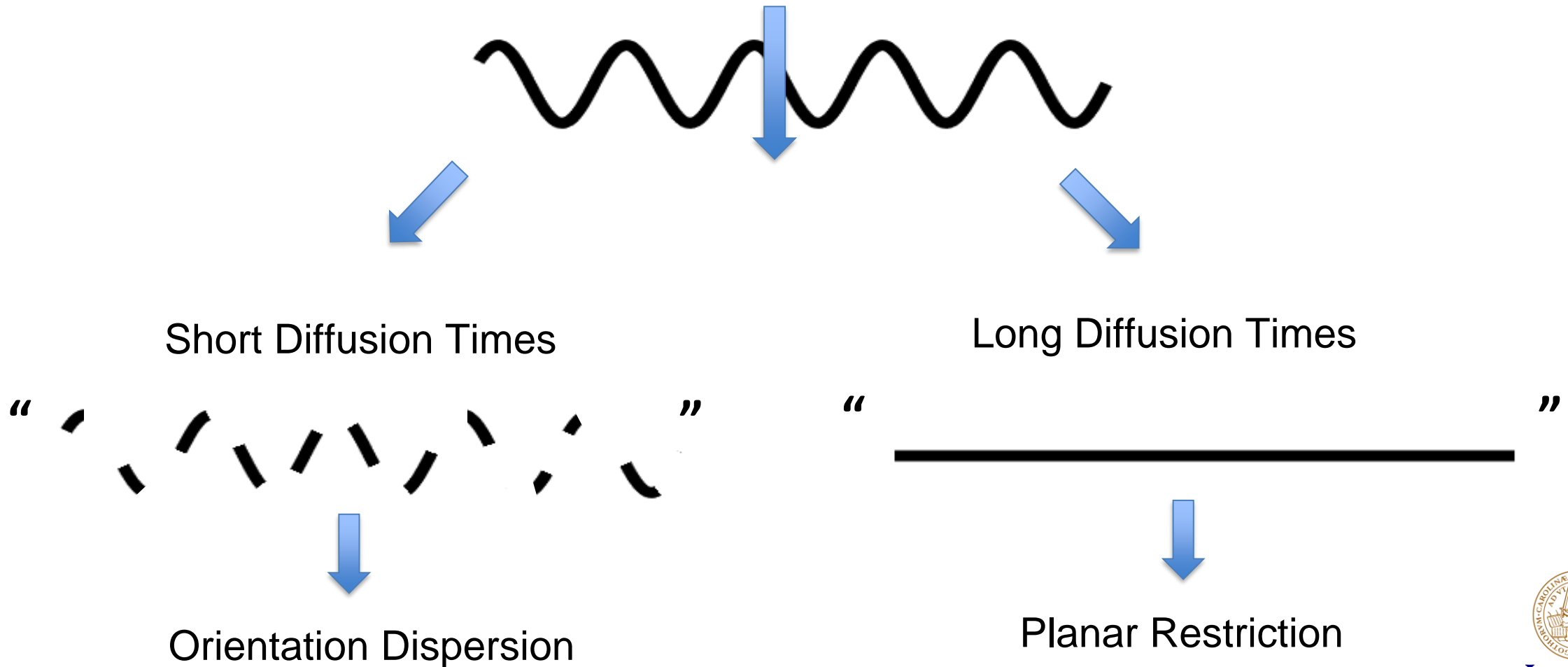
Corpus Callosum



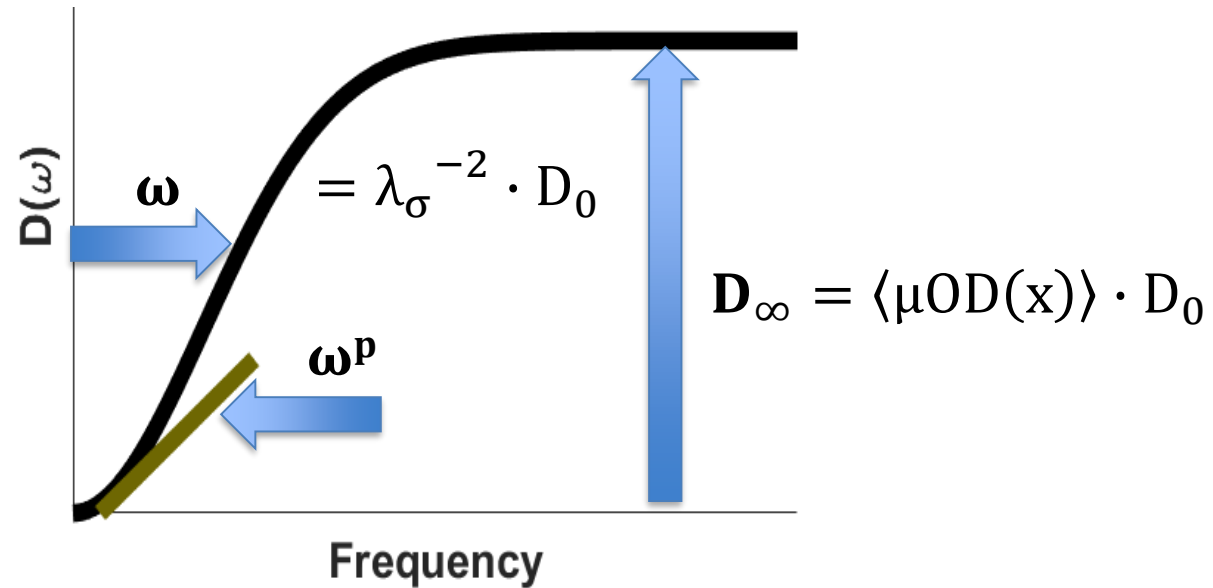
Phrenic Nerve



Diffusion in harmonic waves

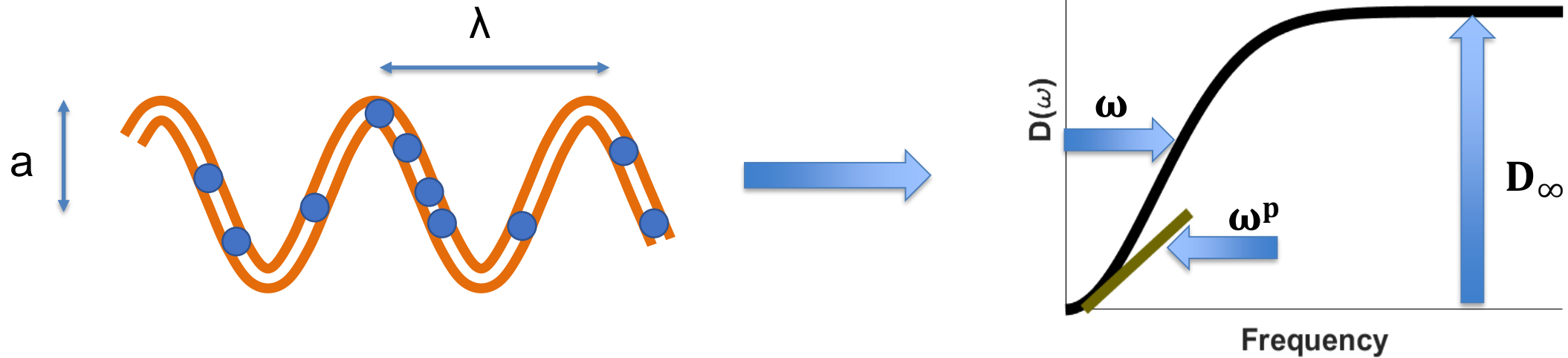


How to characterize diffusion spectrum?



$$D(\omega|D_\infty, \sigma, p) = D_\infty \left(1 - e^{-\frac{\omega^p}{\sigma^2}} \right)$$

Methods: Numerical Simulations



Descriptors of axonal Trajectories:

- 1) Amplitude a , wavelength λ
- 2) Microscopic orientation dispersion μOD , dispersion-weighted wavelength λ_σ

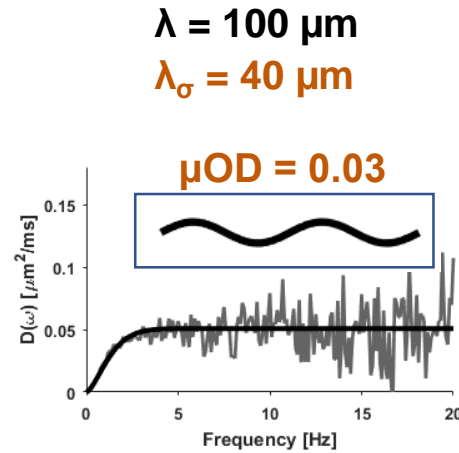
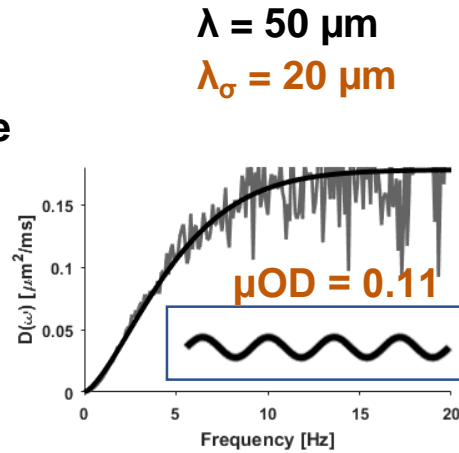
Results: Diffusion Spectra of Axonal Trajectories

Harmonic Axonal Trajectories

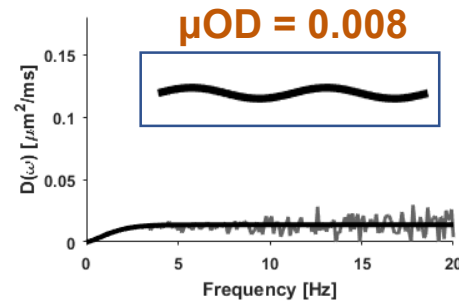
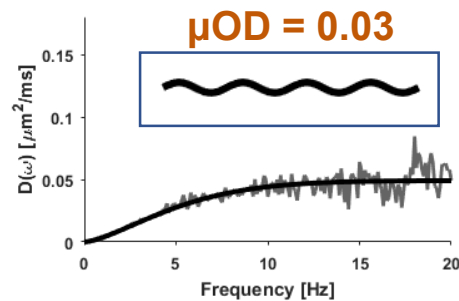
Stochastic Axonal Trajectories

Amplitude

4 μm



2 μm

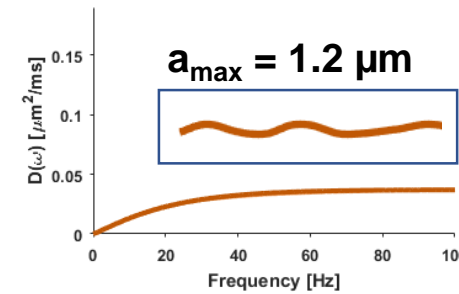


μOD

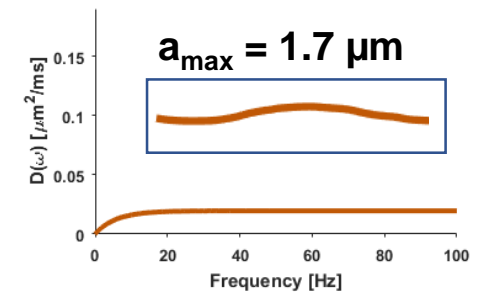
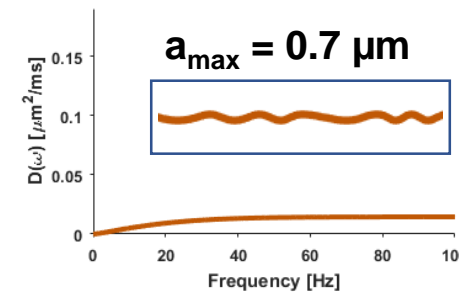
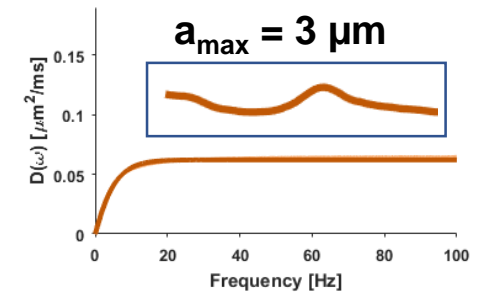
0.03

0.01

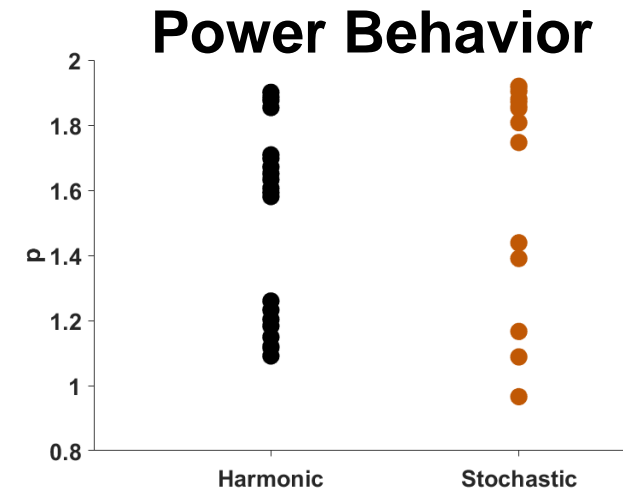
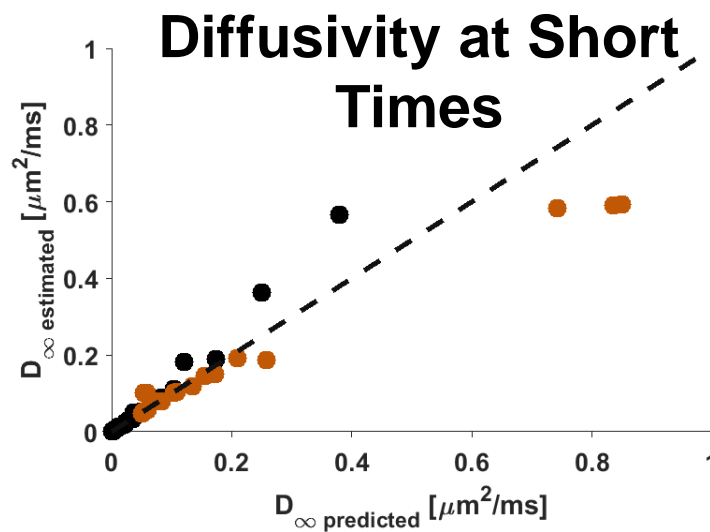
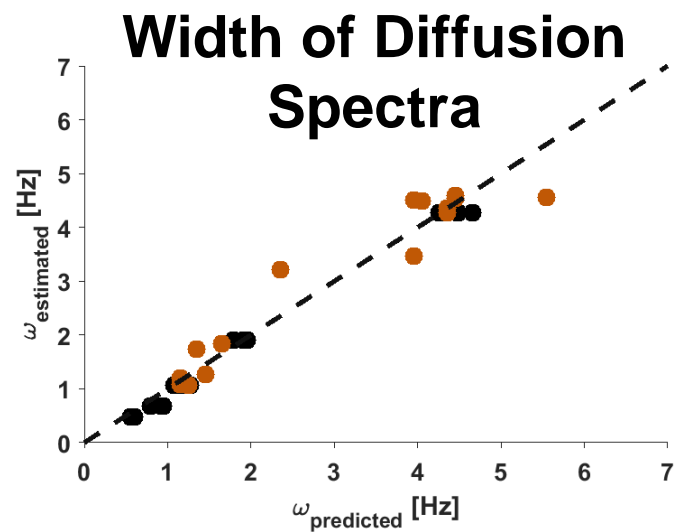
$\lambda_\sigma = 10 \mu\text{m}$
 $\lambda = \text{undefined}$



$\lambda_\sigma = 20 \mu\text{m}$
 $\lambda = \text{undefined}$



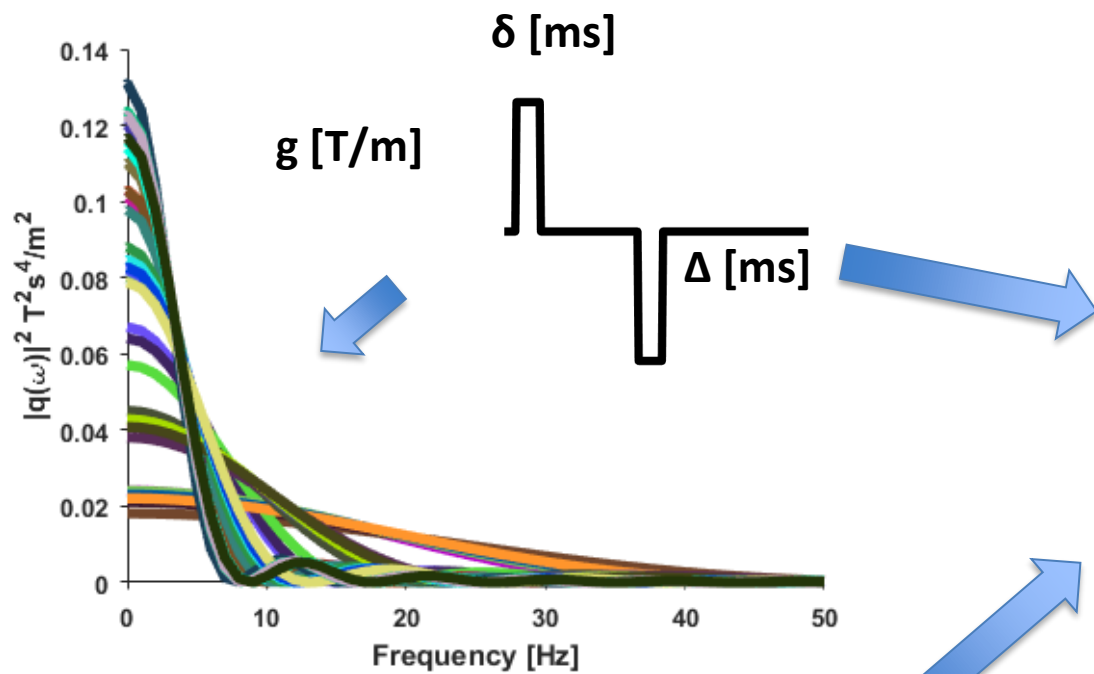
Results: Predicted vs. Estimated parameters



- Harmonic
- Stochastic

Results: Solving the inverse problem

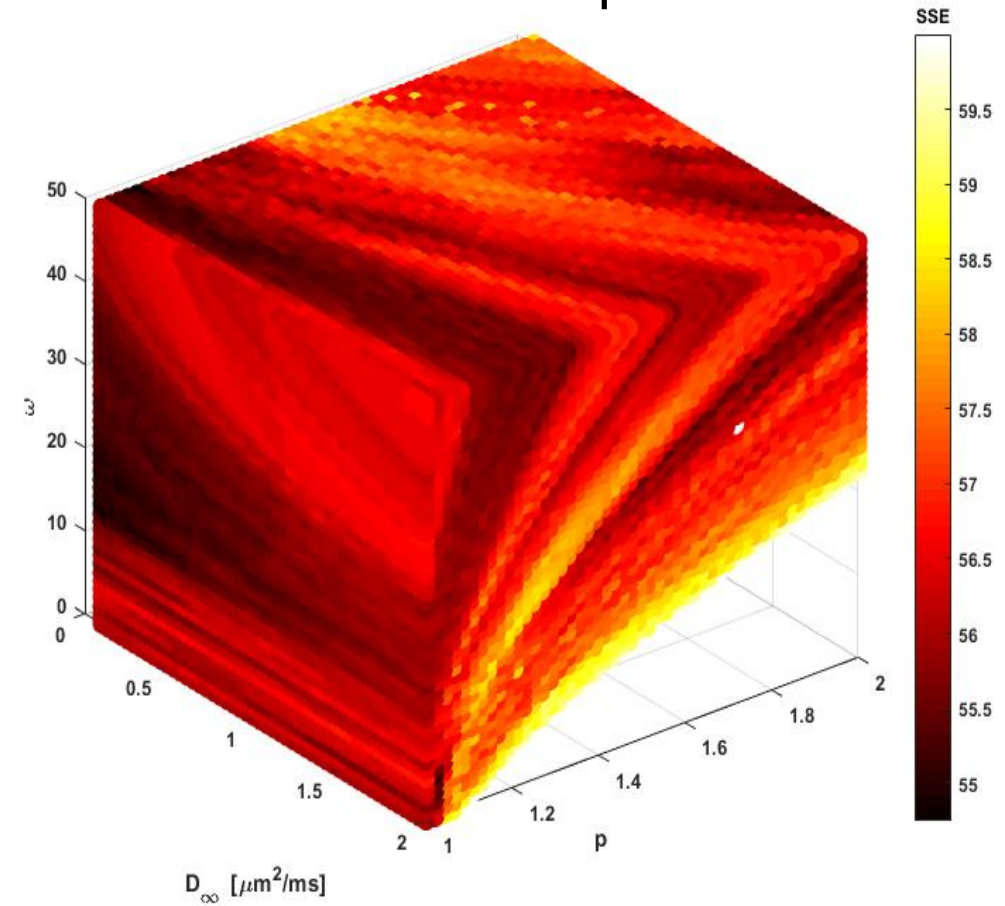
Gradient Waveforms



dMRI Signal



Parameter Space



Results: Estimating Axonal Trajectories

$$\omega = 10 \text{ Hz}$$



$$D_\infty = 0.1 \text{ } \mu\text{m}^2/\text{ms}$$

$$p = 2$$

$$\lambda_\sigma = 13 \text{ } \mu\text{m}, \mu\text{OD} = 0.06$$



$$a = 1.8 \text{ } \mu\text{m}$$

$$\lambda = 32 \text{ } \mu\text{m}$$



$$a_{\text{max}} = 2.5 \text{ } \mu\text{m}$$

$$\lambda = \text{undefined}$$

$$\omega = 33 \text{ Hz}$$



$$D_\infty = 0.4 \text{ } \mu\text{m}^2/\text{ms}$$

$$p = 2$$

$$\lambda_\sigma = 7.2 \text{ } \mu\text{m}, \mu\text{OD} = 0.24$$



$$a = 2.3 \text{ } \mu\text{m}$$

$$\lambda = 18 \text{ } \mu\text{m}$$



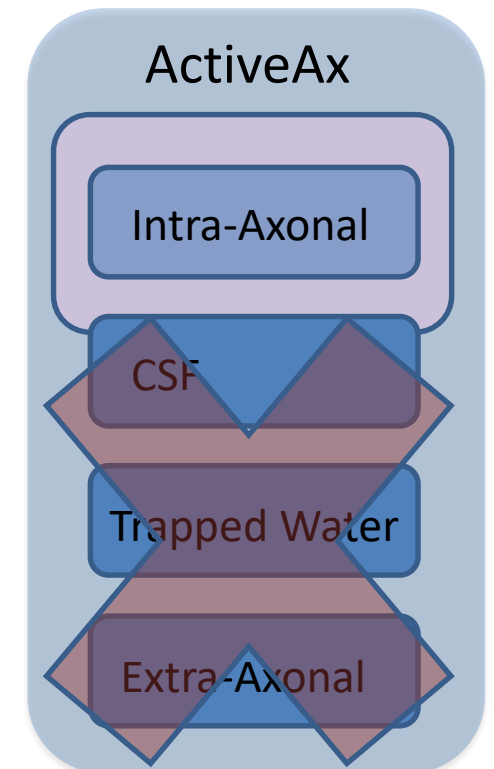
$$a_{\text{max}} = 3 \text{ } \mu\text{m}$$

$$\lambda = \text{undefined}$$

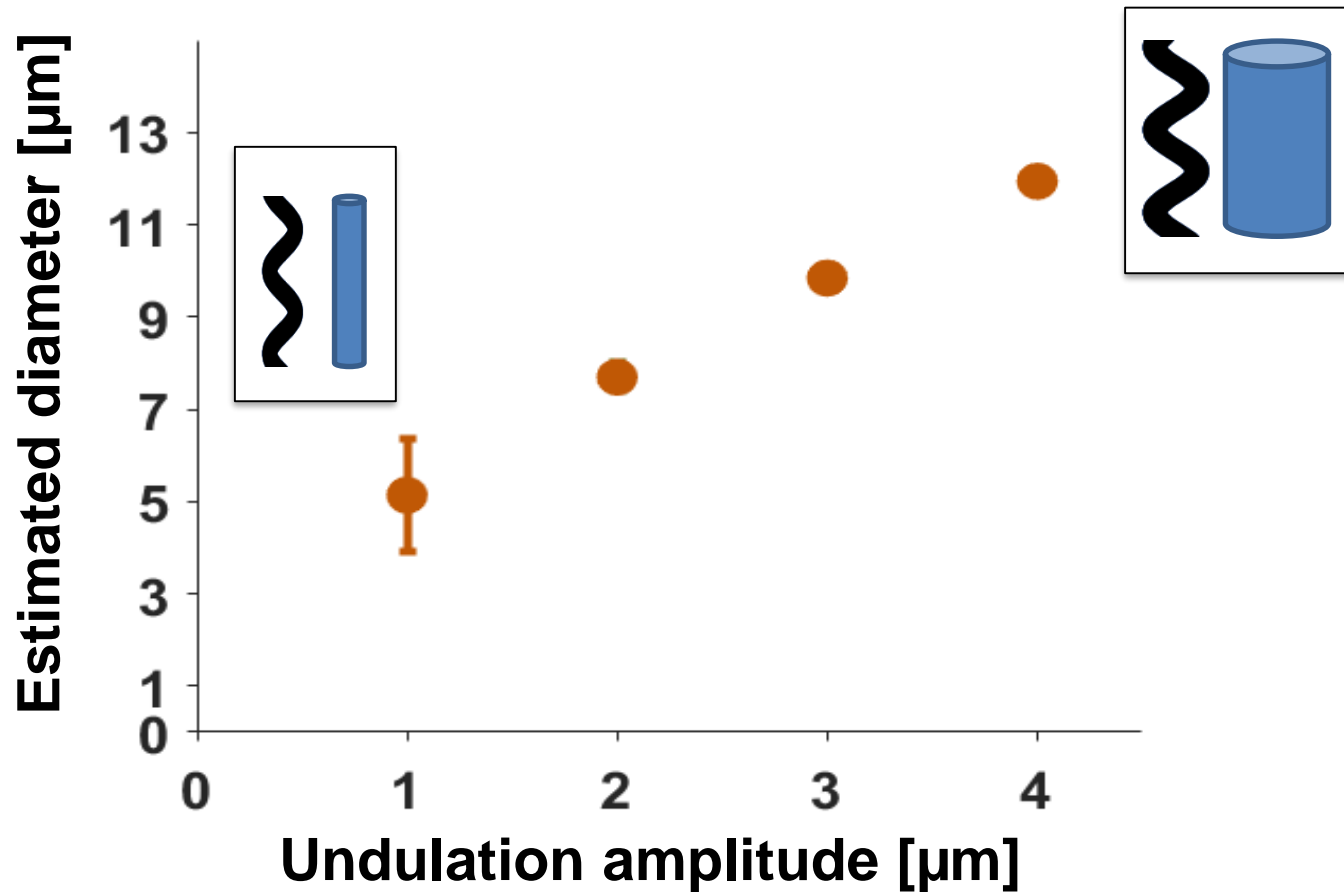


Are non-straight axonal trajectories necessary?

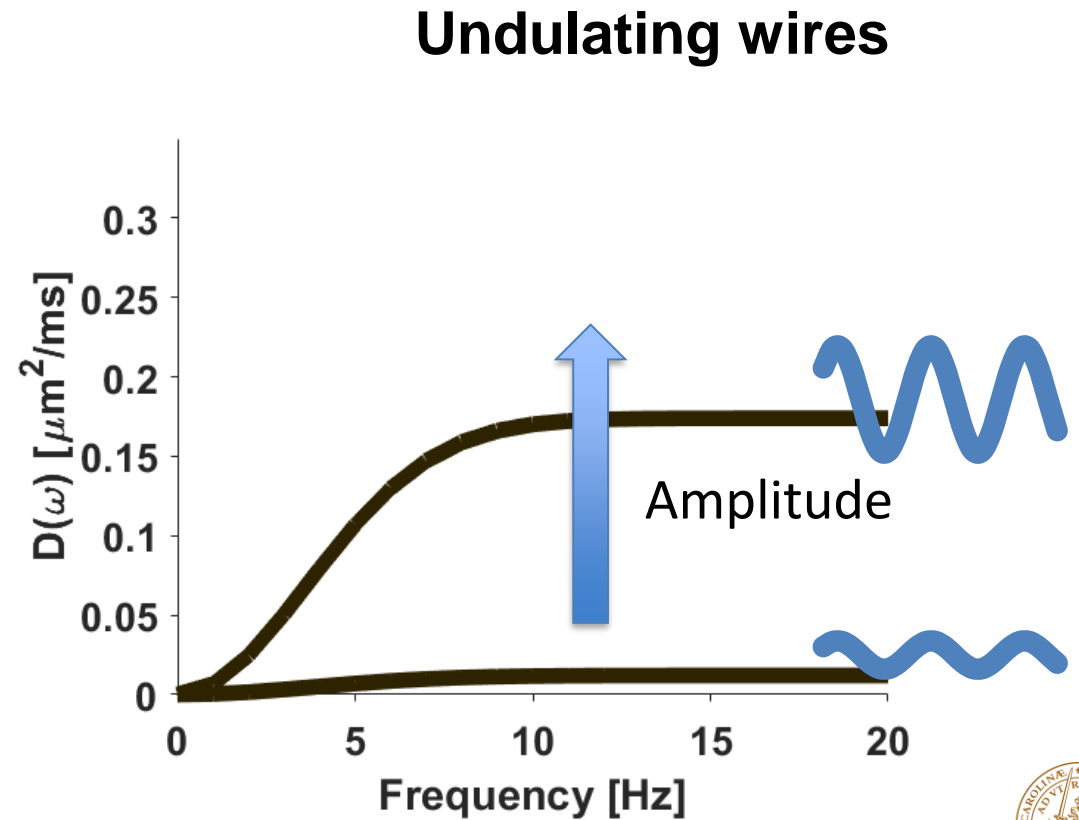
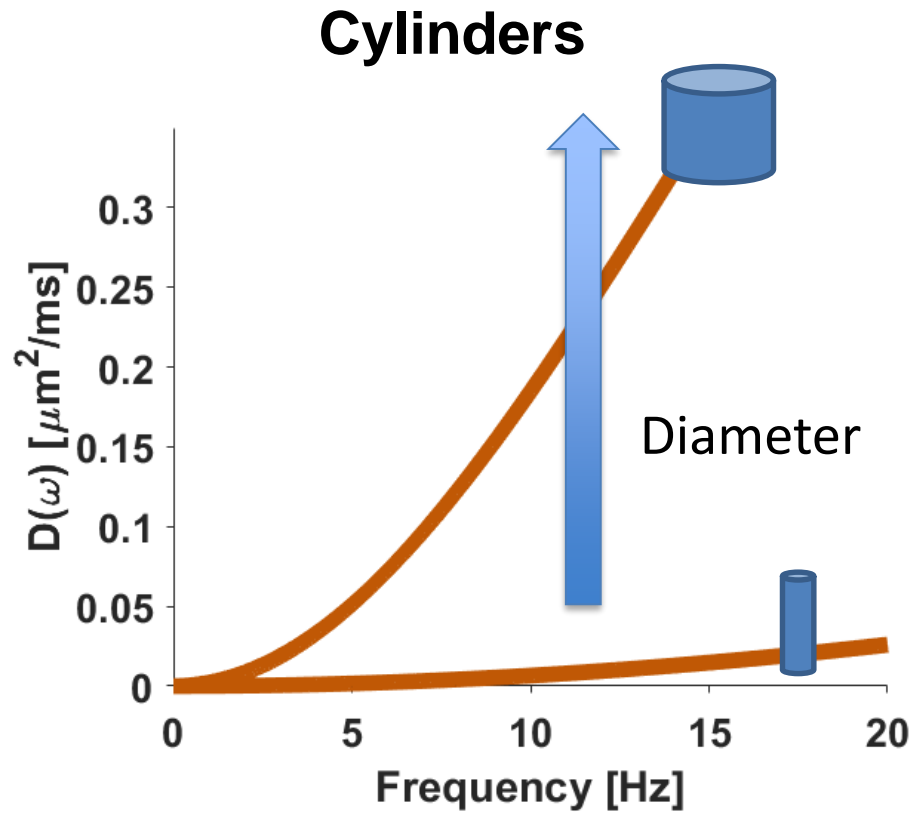
- Questions:
 - Do undulations bias axon diameter estimation in models that assume straight cylinders? [1]
 - Can the bias be explained?
- Signal and diffusion spectra from Monte Carlo.
- Restricted diffusion compartment model constrained to intra-axonal part only [2].
- Gradient waveforms with short & long diffusion times [2].
- Explanation through diffusion spectra.



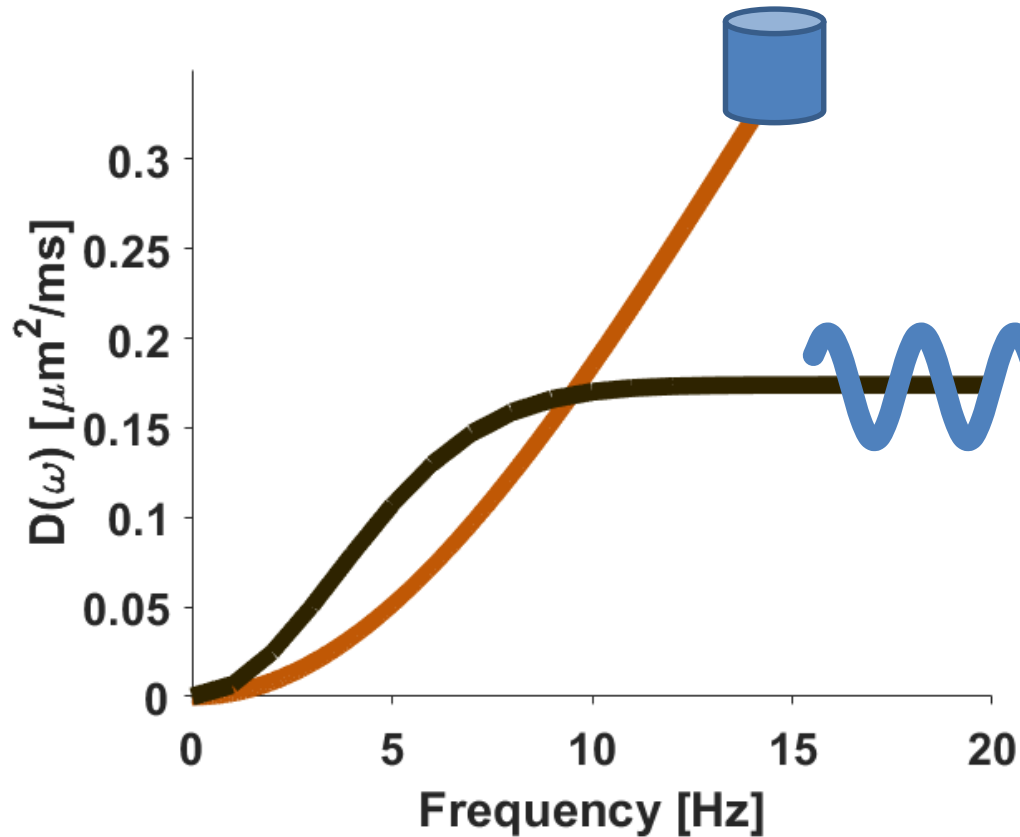
Results: Undulations interpreted as cylinders



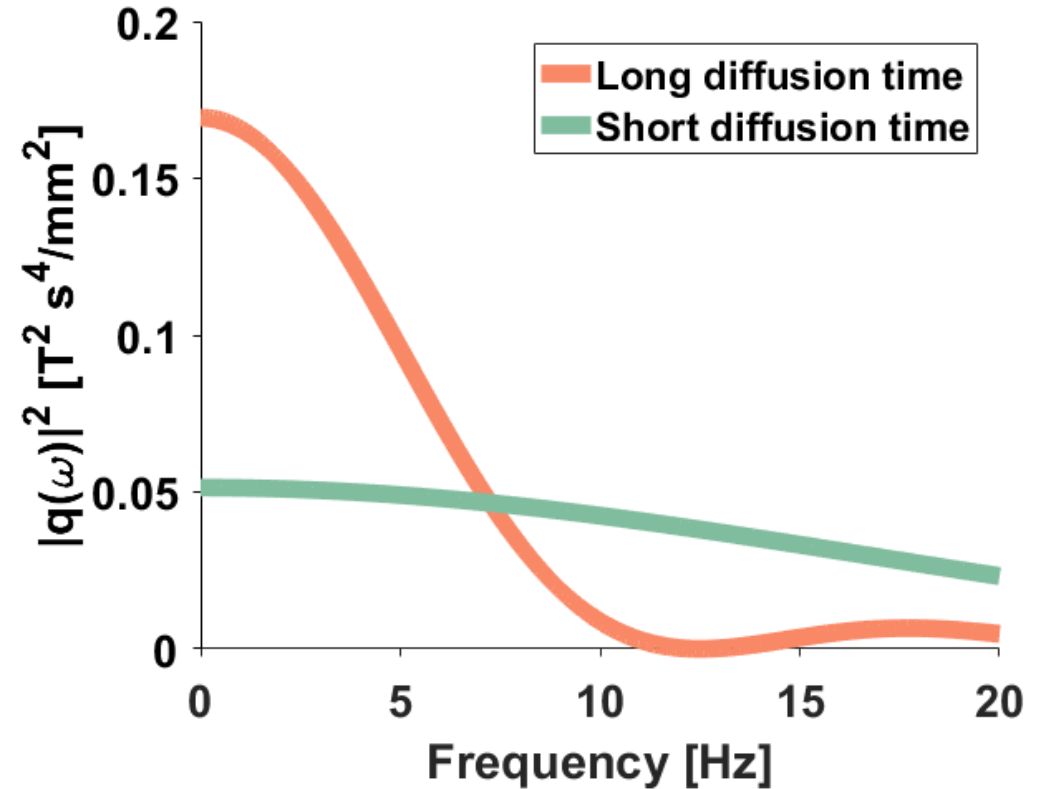
Tissue Parameters \leftrightarrow Diffusion Spectrum



What & Where is Fitted?



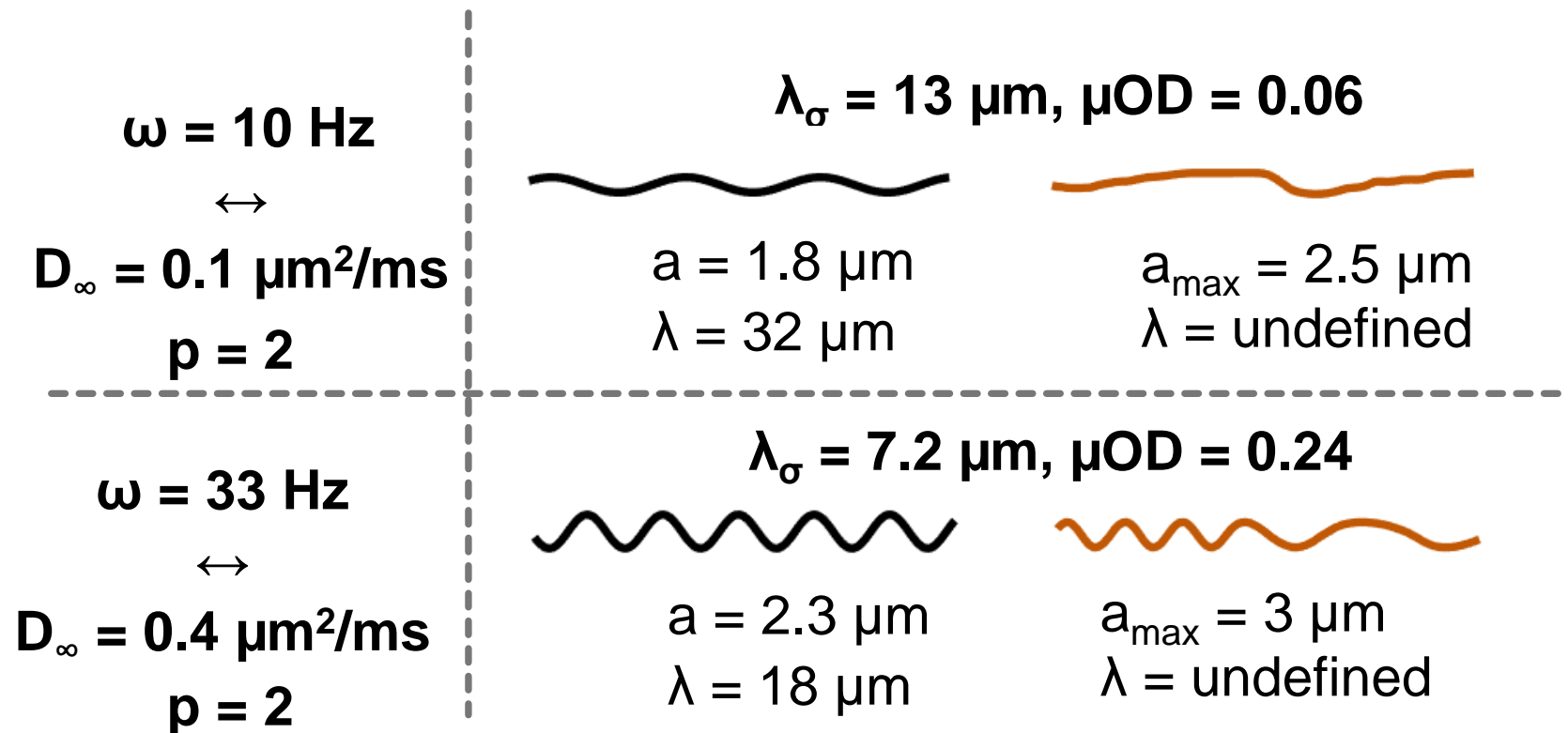
Diffusion Spectra



Encoding Spectra

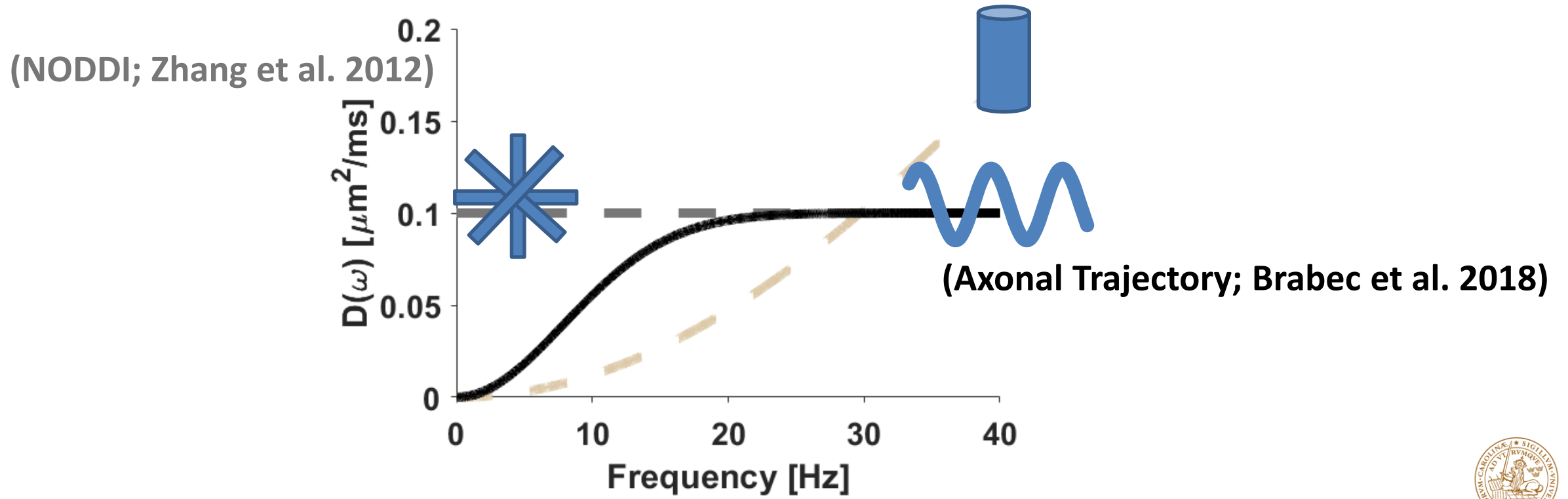
Conclusions

- Axonal trajectory model is congruent with histology, unlike models that assume parallel straight cylinders.



Conclusions

(ActiveAx; Alexander et al. 2010)

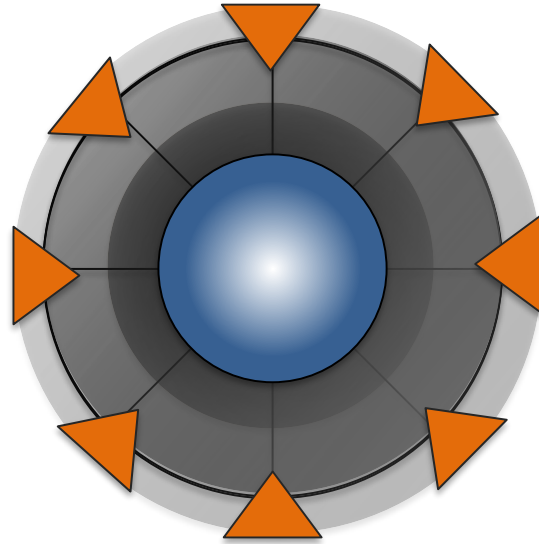


Importance of Physics

- A simple 1D-toy model with far-reaching consequences.
- Scientific fields are no longer separate.



Thank you for your attention



Acknowledgements:

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