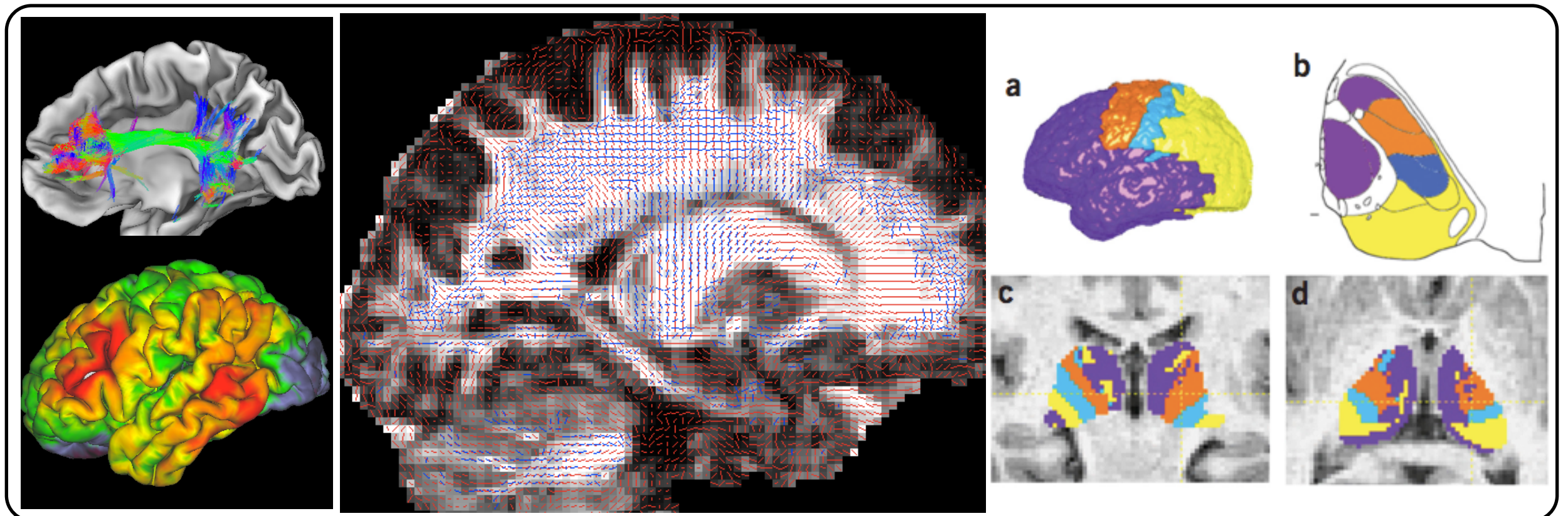


Diffusion Tractography

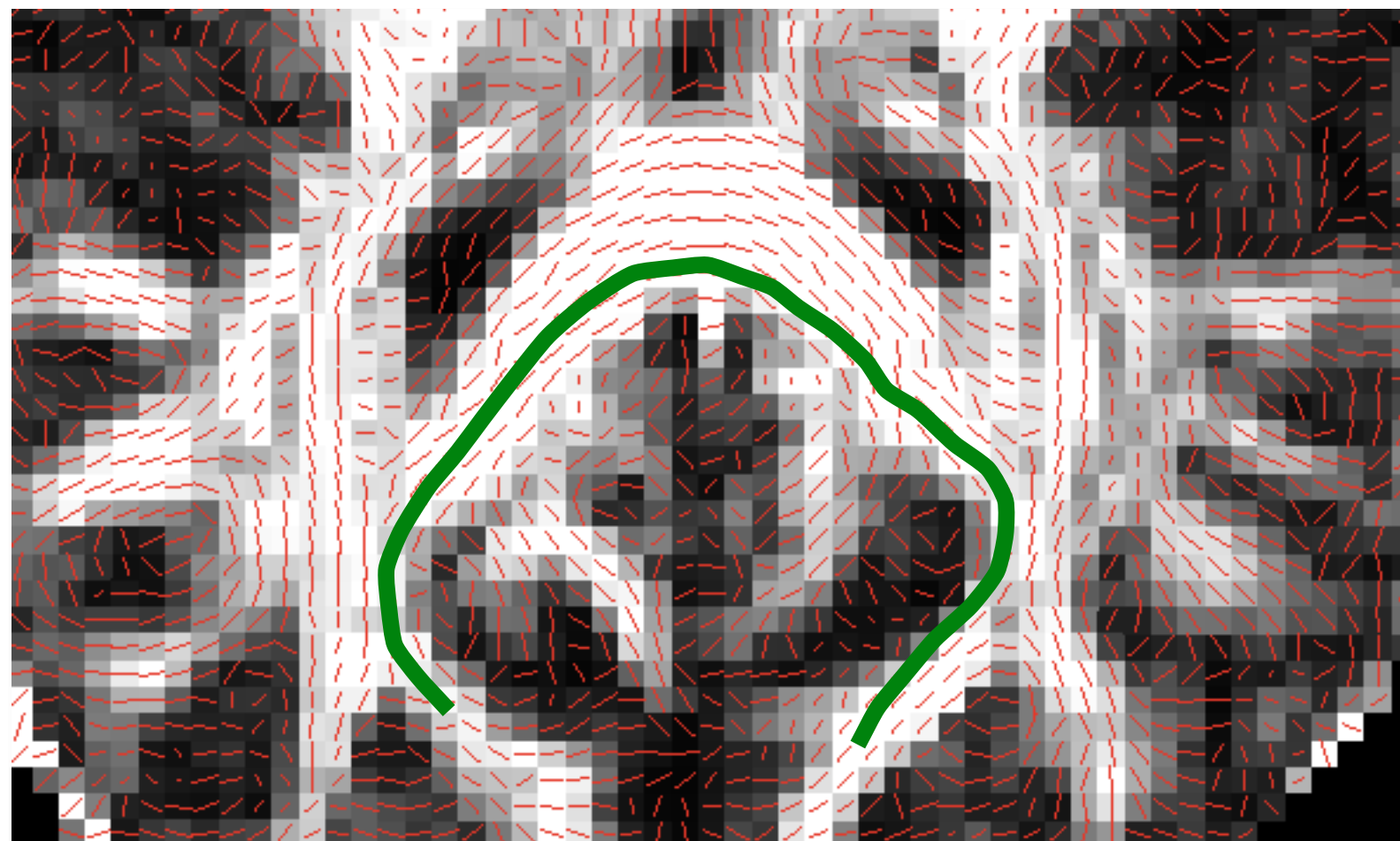
- Goal of tractography
- Estimating Fibre Orientations - BEDPOSTX
- Probabilistic Tractography - PROBTRACKX
- ProbtrackX outputs
- Tractography limitations



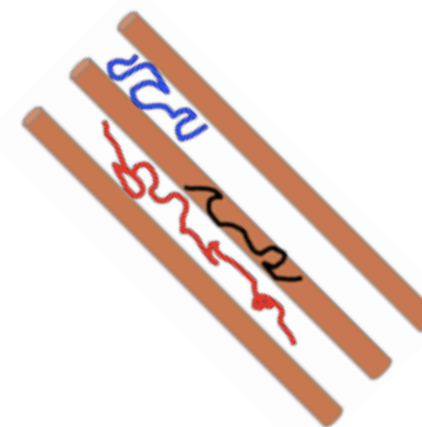
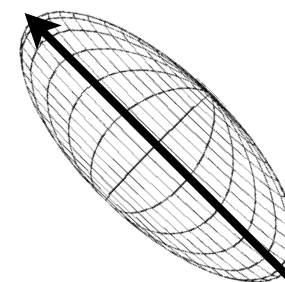


DTI tractography

v_1 map
Principal Diffusion Direction



Principal Diffusion
Direction

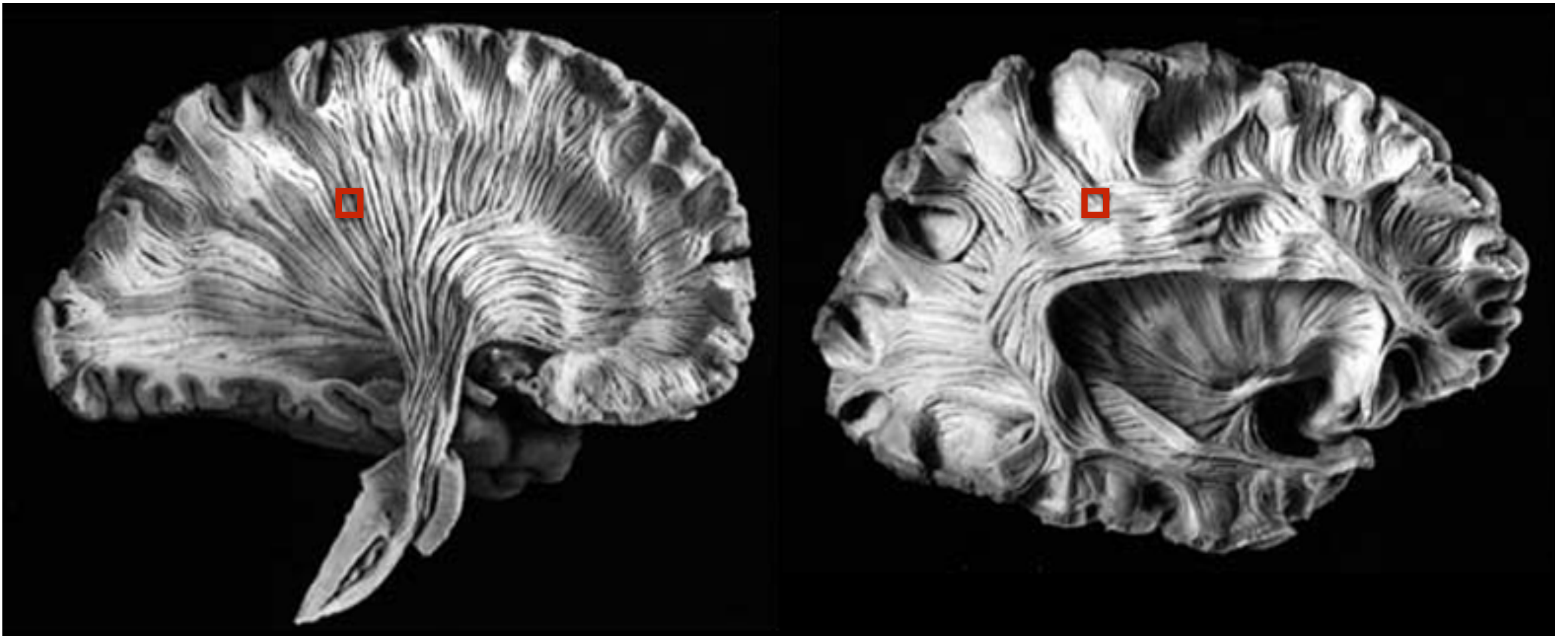


Assumption:

Direction of maximum diffusivity
(in anisotropic voxels)
is an estimate of the major fibre
orientation.



But is WM always coherently organised within a voxel?



Unfortunately not, complex fibre patterns (e.g. crossings) are very common at the voxel scale.

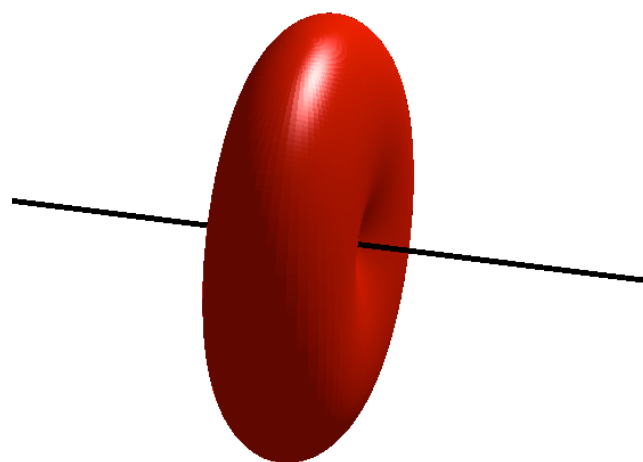
Williams, Gluhbegovic, and Jew, "The Human Brain: Dissections of the Real Brain", Virtual Hospital, University of Iowa, 1997



Predictions from the tensor model no crossing fibres

One orientation

Measured
Signal
Shape



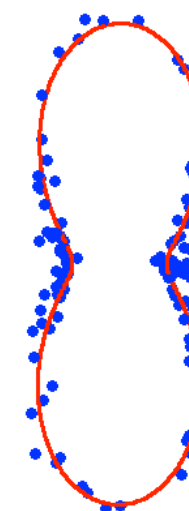
Predicted
Signal
Shape



DTI
Ellipsoid



Prediction &
Measurements
in 2D

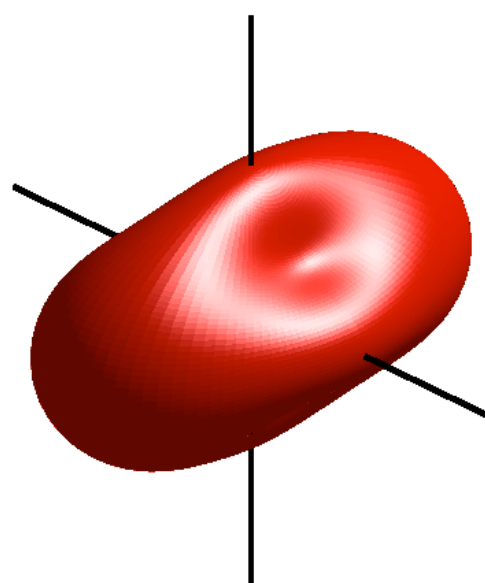




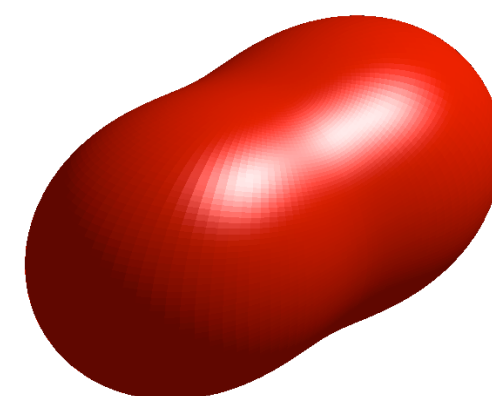
Predictions from the tensor model crossing fibres

Two orientations

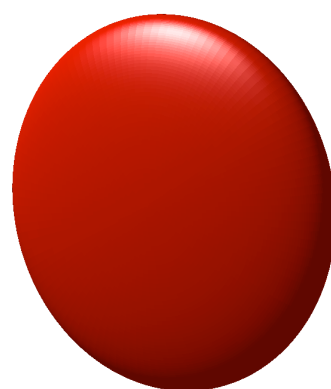
Measured
Signal
Shape



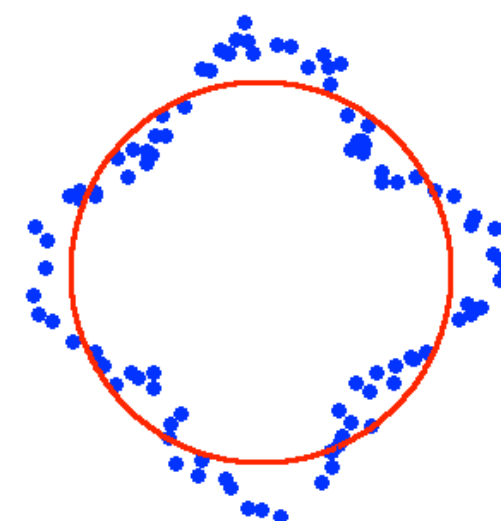
Predicted
Signal
Shape



DTI
Ellipsoid



Prediction &
Measurements
in 2D

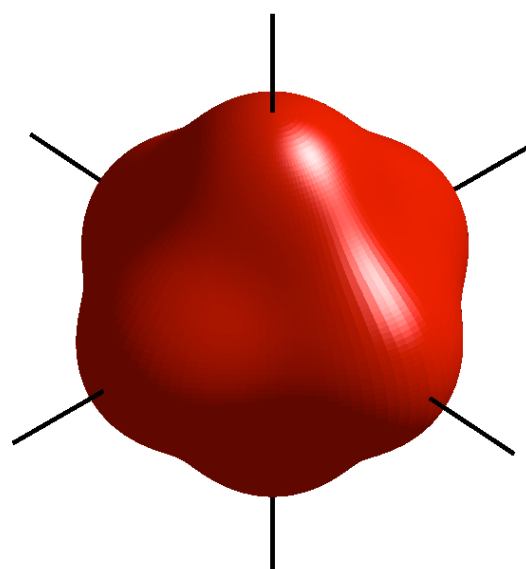




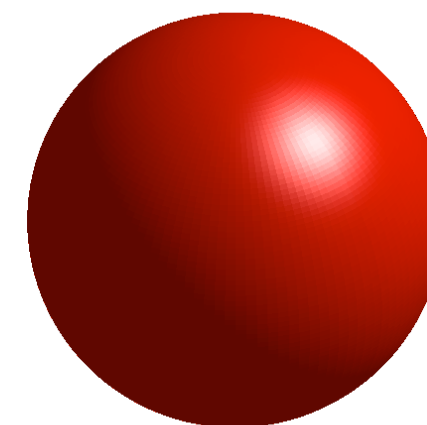
Predictions from the tensor model crossing fibres

Three Orientations

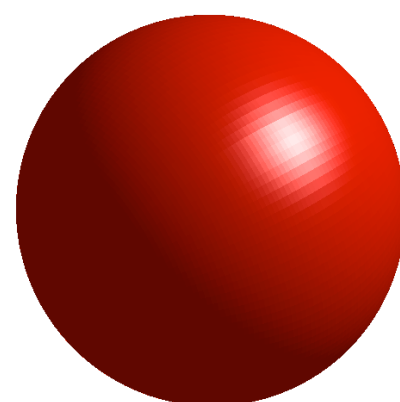
Measured
Signal
Shape



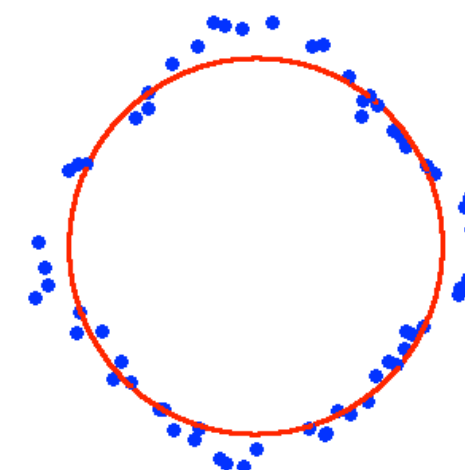
Predicted
Signal
Shape



DTI
Ellipsoid



Prediction &
Measurements
in 2D



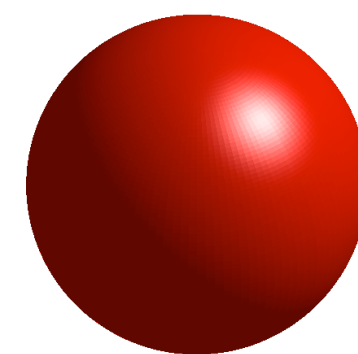
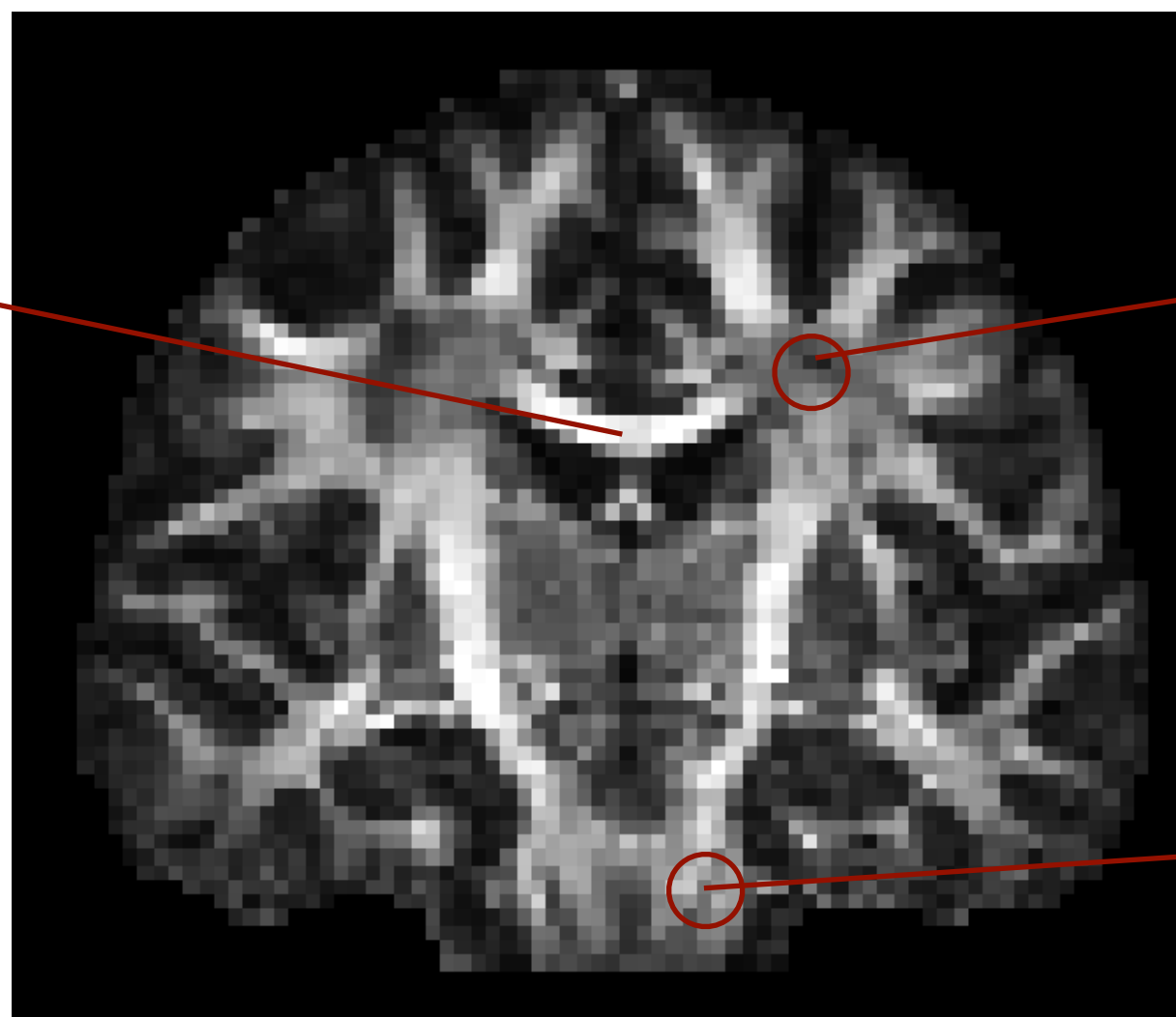


How good is the DTI Model in regions with crossing fibres?

- In voxels containing two crossing bundles, the tensor ellipsoid is pancake-shaped (oblate, planar tensor).
- In voxels containing three crossing bundles, the tensor ellipsoid is spherical.
- In these areas, DTI $\mathbf{v}_1$ is meaningless.



Prolate Tensor
 $\lambda_1 \gg \lambda_2, \lambda_3$



Spherical Tensor
 $\lambda_1 = \lambda_2 = \lambda_3$



Oblate Tensor
 $\lambda_1 = \lambda_2 \gg \lambda_3$

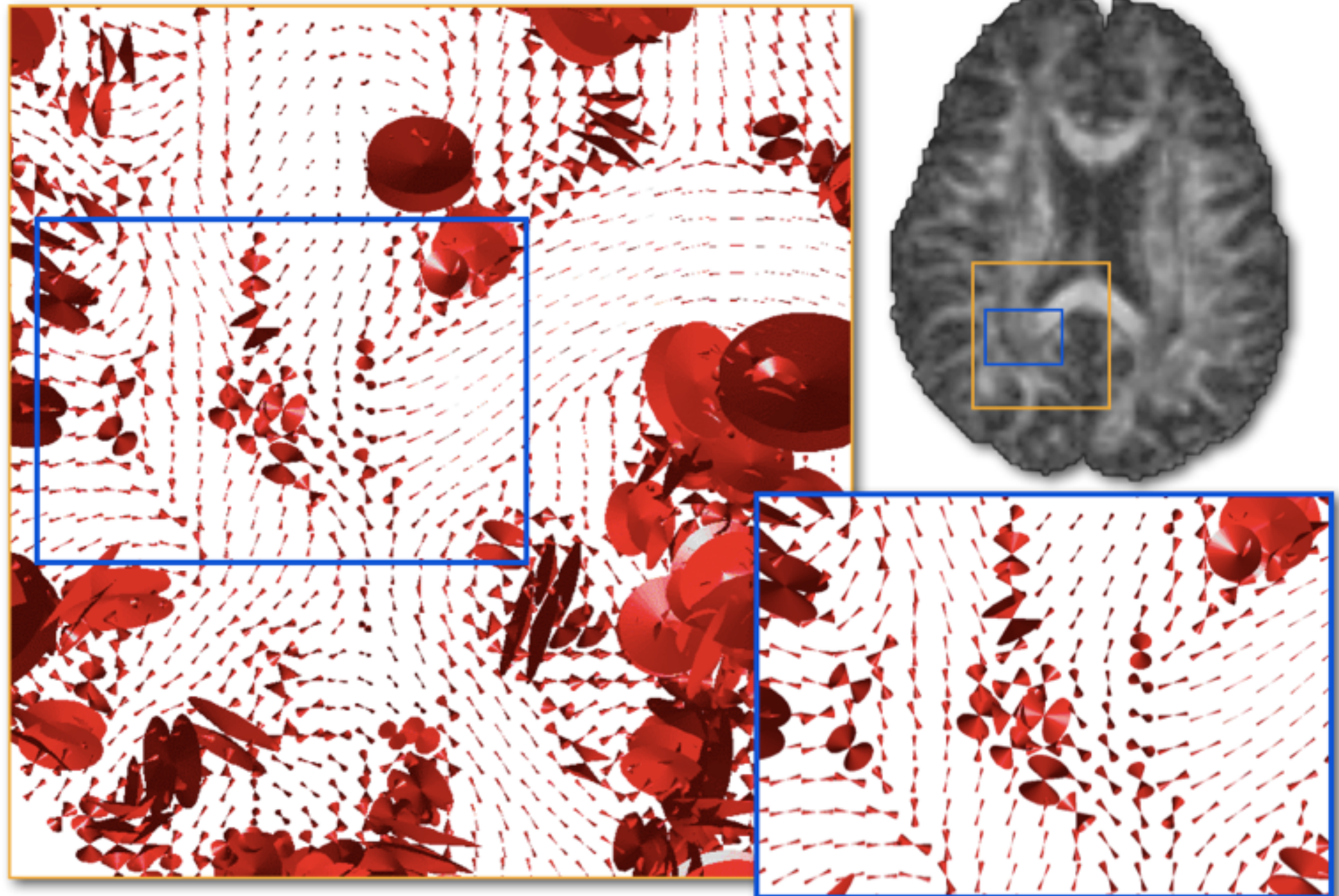


Uncertainty on DTI Fibre Orientation Estimates

Repeat an acquisition many times and obtain the variability in $\mathbf{v}_1$ from the different datasets.

Uncertainty Sources

- Modelling errors
- Noise



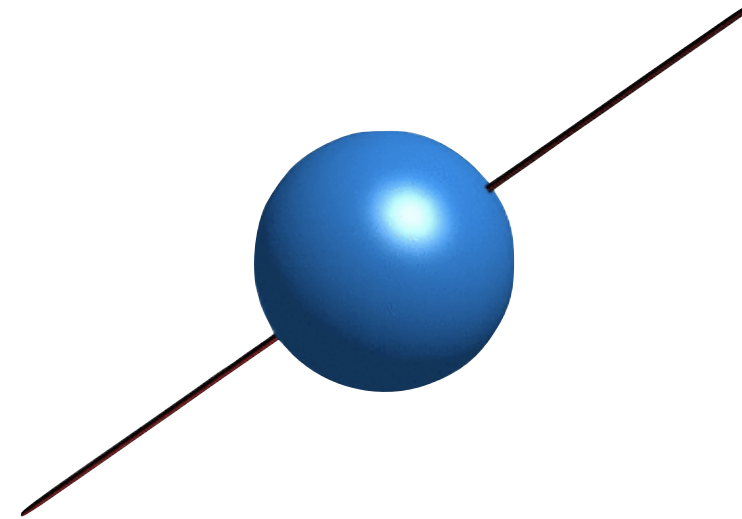
Cones of uncertainty on DTI $\mathbf{v}_1$

Do we have to use the DTI model to estimate orientations? Not really, many models exist

DTI model (dtifit)



Ball & sticks model (bedpostx)



Anisotropic Volume Fraction (unknown)

Diffusivity (unknown)

Fibre Orientation (unknown)

$$s_j = s_0 [(1-f)\exp(-b_j d) + f \exp(-b_j d (\mathbf{x}_j^T \mathbf{v})^2)]$$

Measured Signal for Gradient j

b-value for gradient j (known)

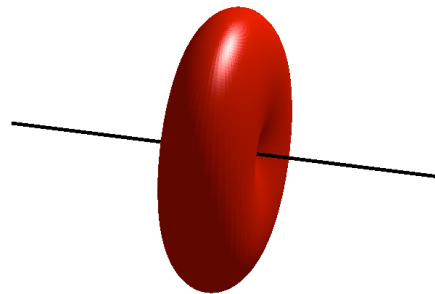
Unit vector representing the direction of gradient j (known)



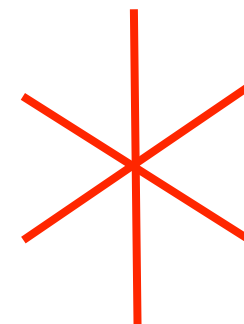
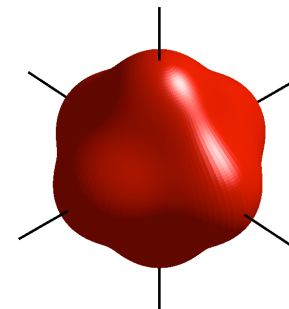
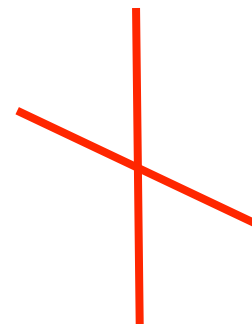
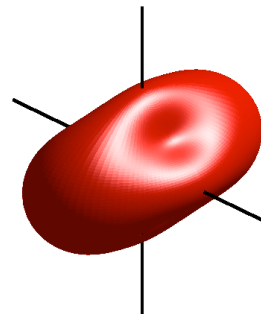
Ball & Sticks Model

Unlike the DT model, it can represent many orientations

Measured
Signal



Estimated
Stick
Orientations



- Anisotropic tensors (sticks) with isotropic background (ball)
- Fibre Orientations modelled explicitly and separated from isotropic partial volumes

Anisotropic Volume
Fractions (unknown)

Diffusivity
(unknown)

Max number
of sticks
(user-defined)

Fibre Orientation
(unknown)

$$s_j = s_0 [(1 - \sum f_n) \exp(-b_j d) + \sum f_n \exp(-b_j d (\mathbf{x}_j^T \mathbf{v}_n)^2)]$$

Measured Signal
for Gradient j

b-value for gradient j
(known)

Unit vector representing the direction of
gradient j (known)

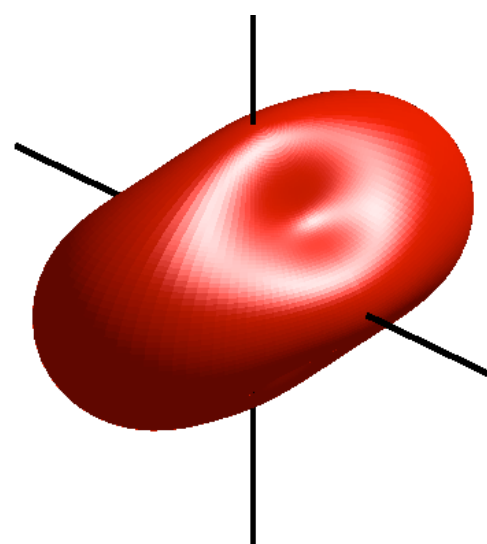


Predictions from the ball and sticks model crossing fibres

Measured
Signal Shape

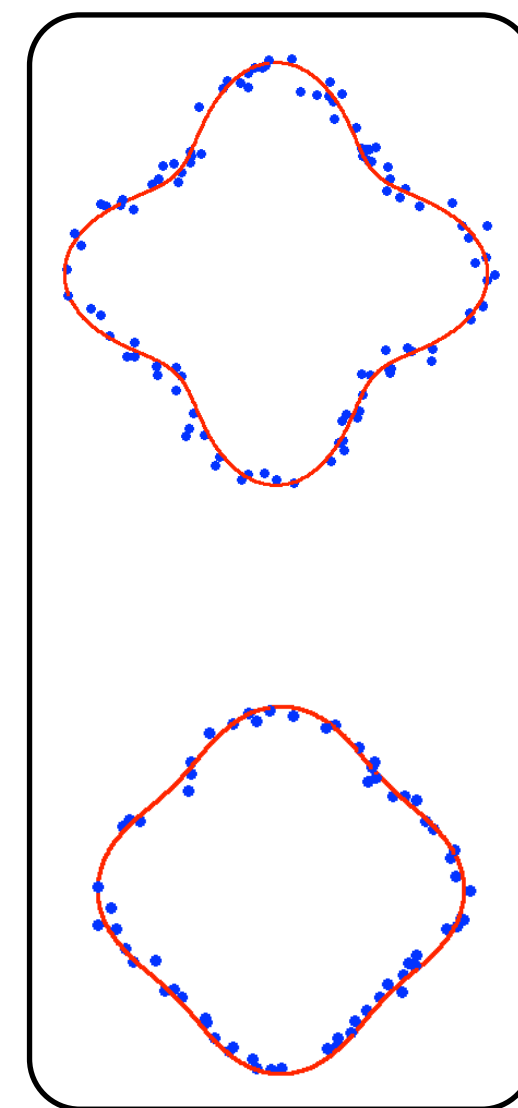
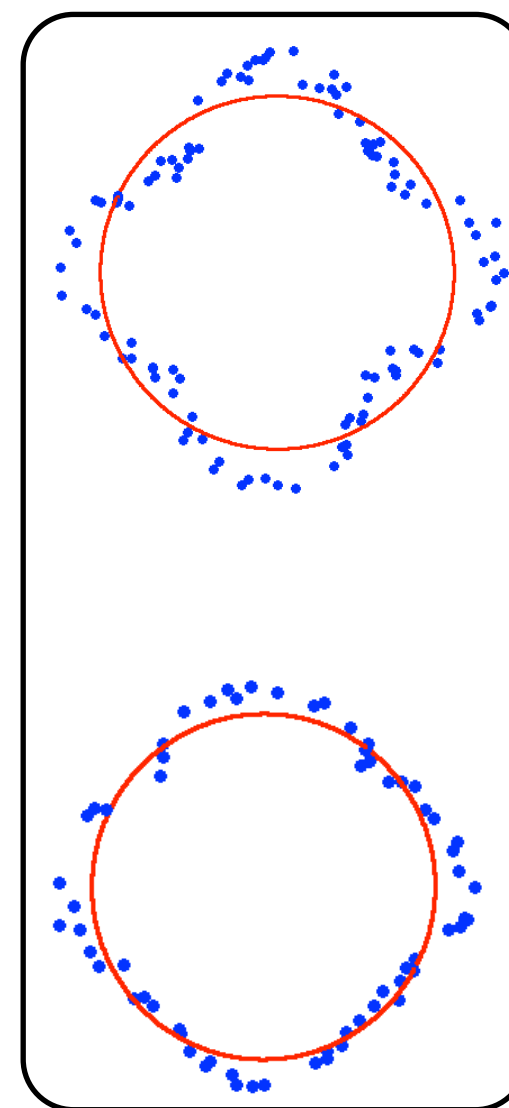
Prediction &
Measurement in 2D

Two orientations

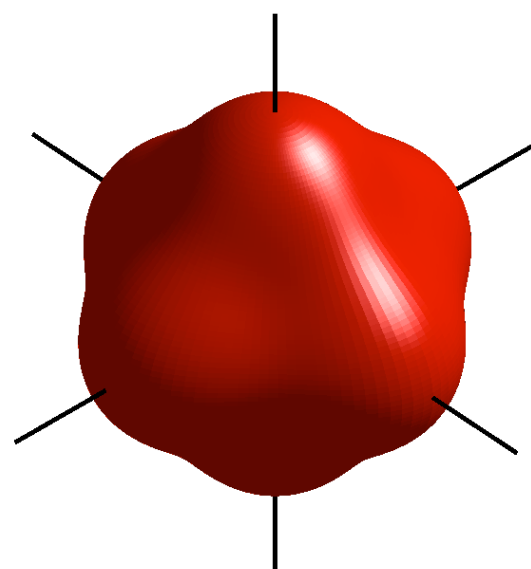


DTI

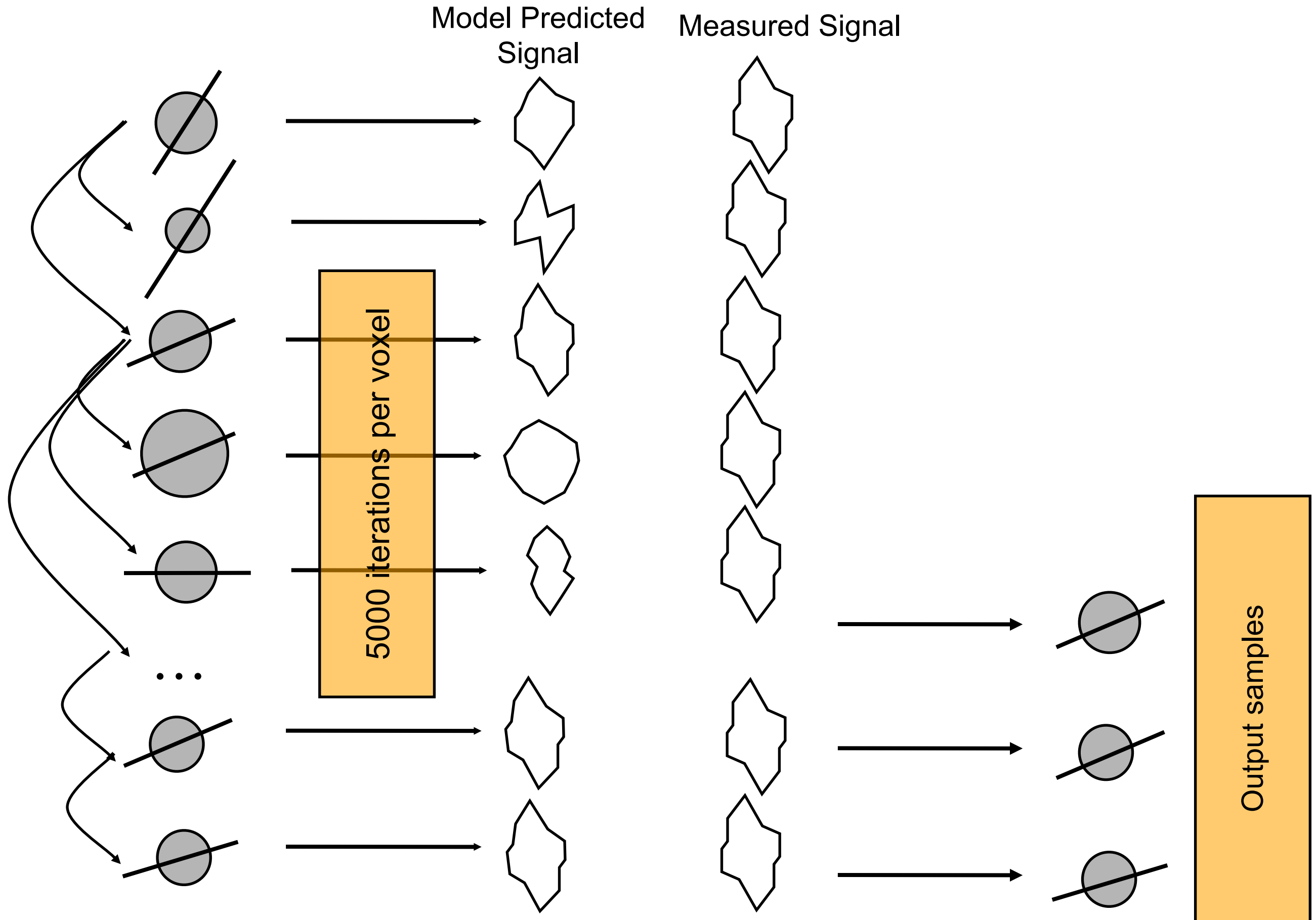
Ball & sticks



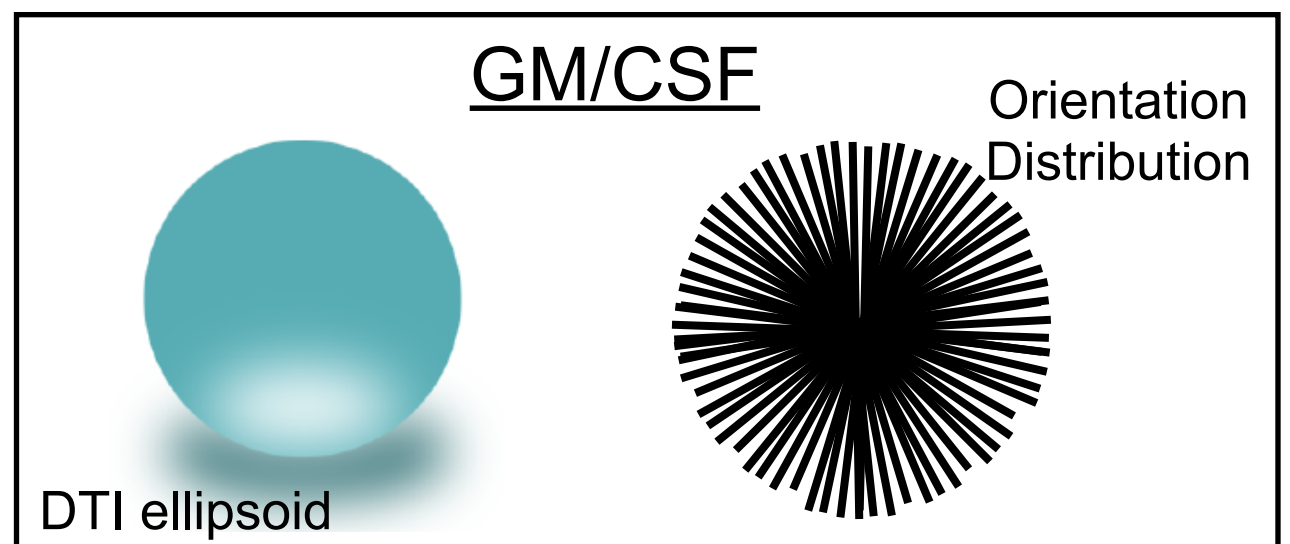
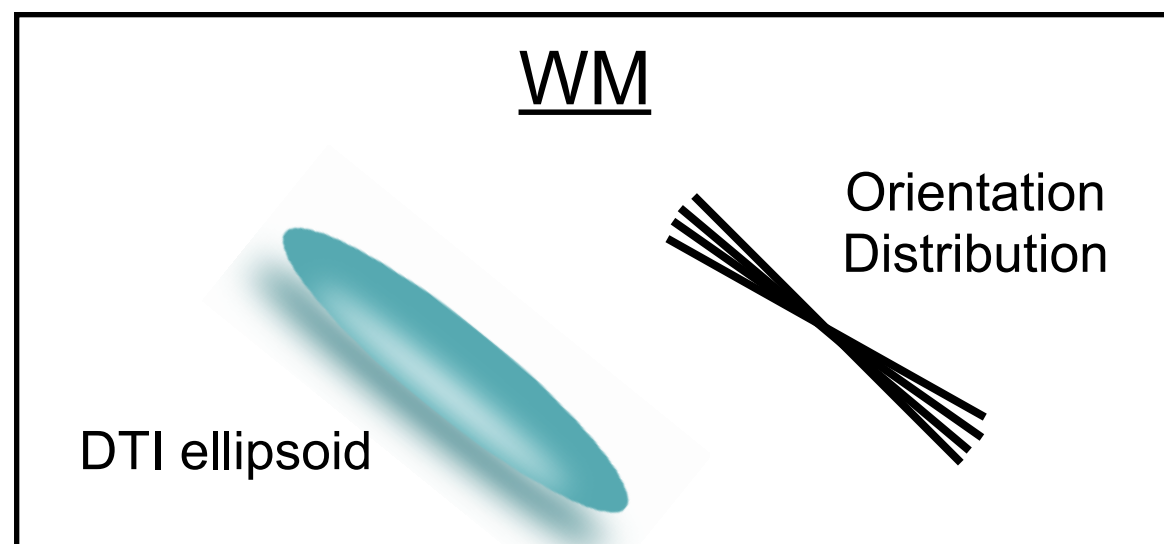
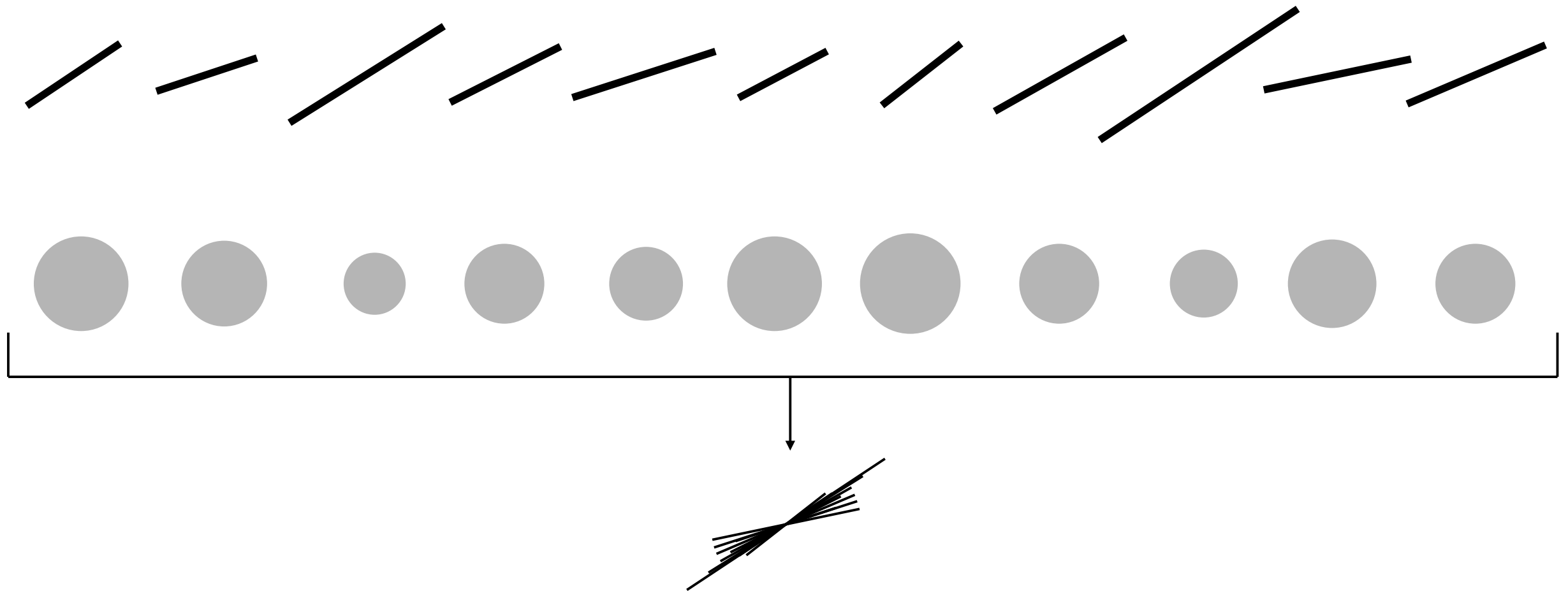
Three orientations



Markov Chain - Monte Carlo (MCMC) Sampling



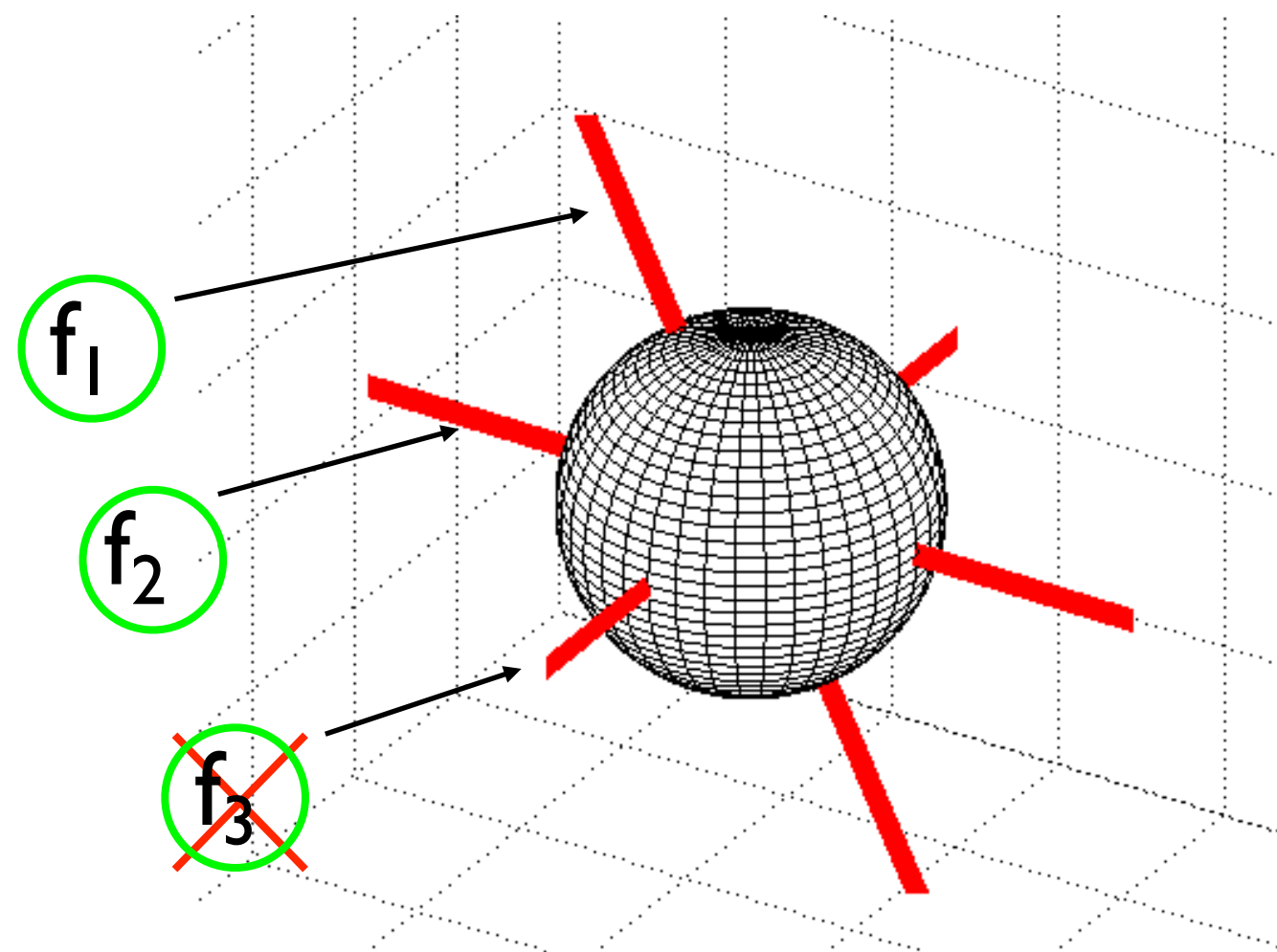
Output in Each voxel = Distributions of Parameters





Ball & Sticks Model Selection

- Model selection problem: One, two or more fibres within a voxel?
- Automatic Relevance Determination: Only estimate complexity that is supported by the data

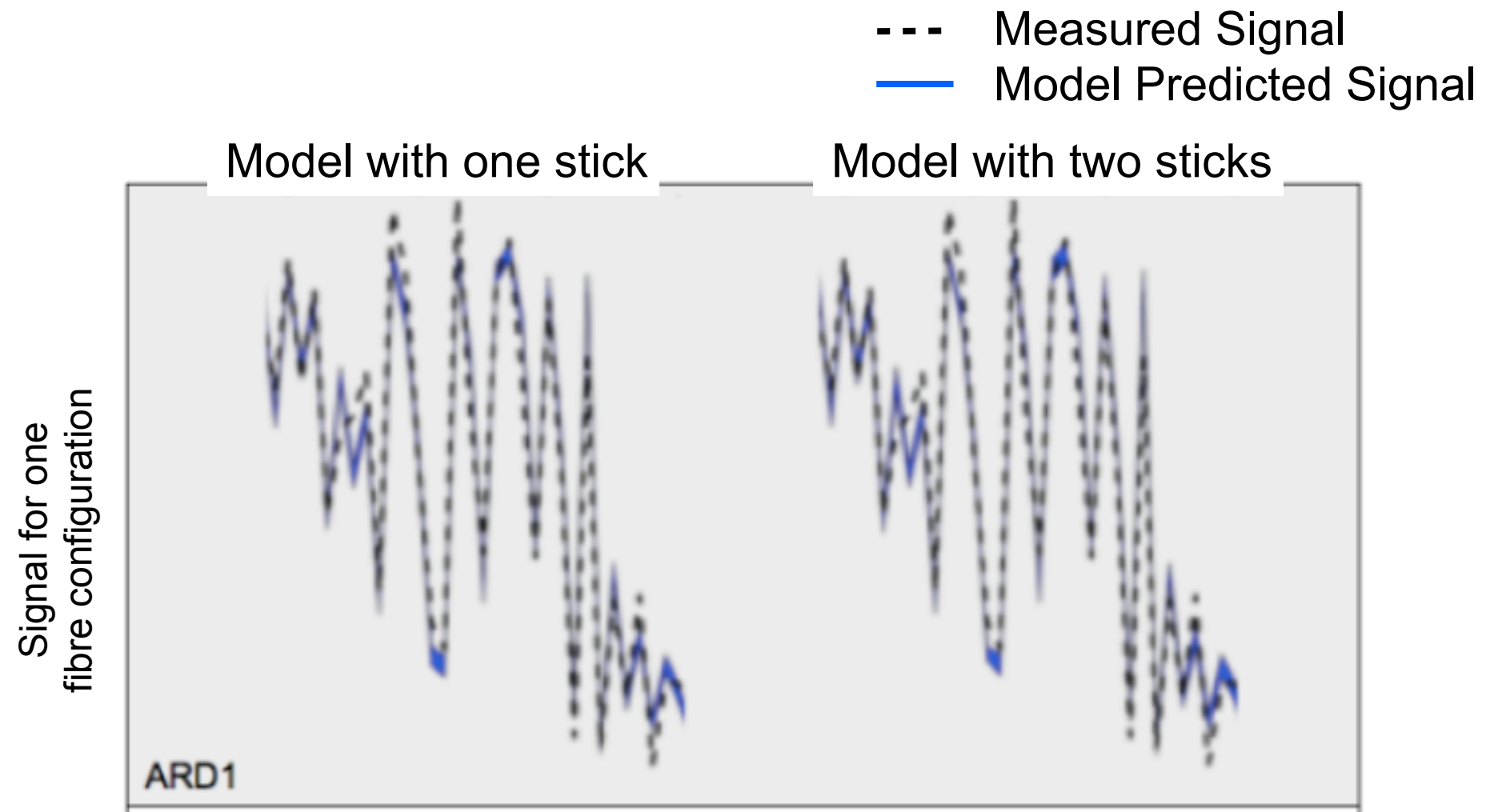


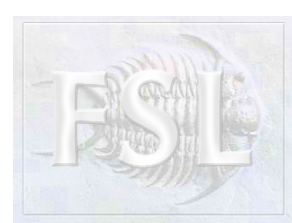


Modelling Complex Fibre Architectures

Automatic Relevance Determination (A.R.D.)

- No benefit from including a 2nd fibre
=> 2nd volume fraction goes to zero



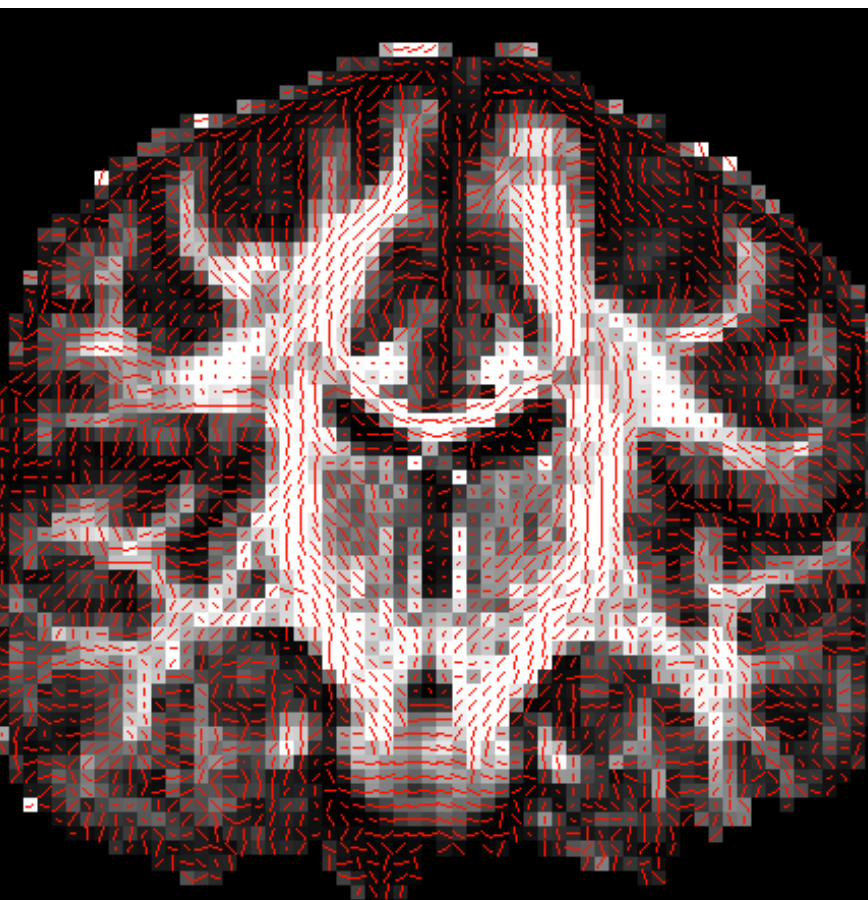


Modelling Complex Fibre Architectures

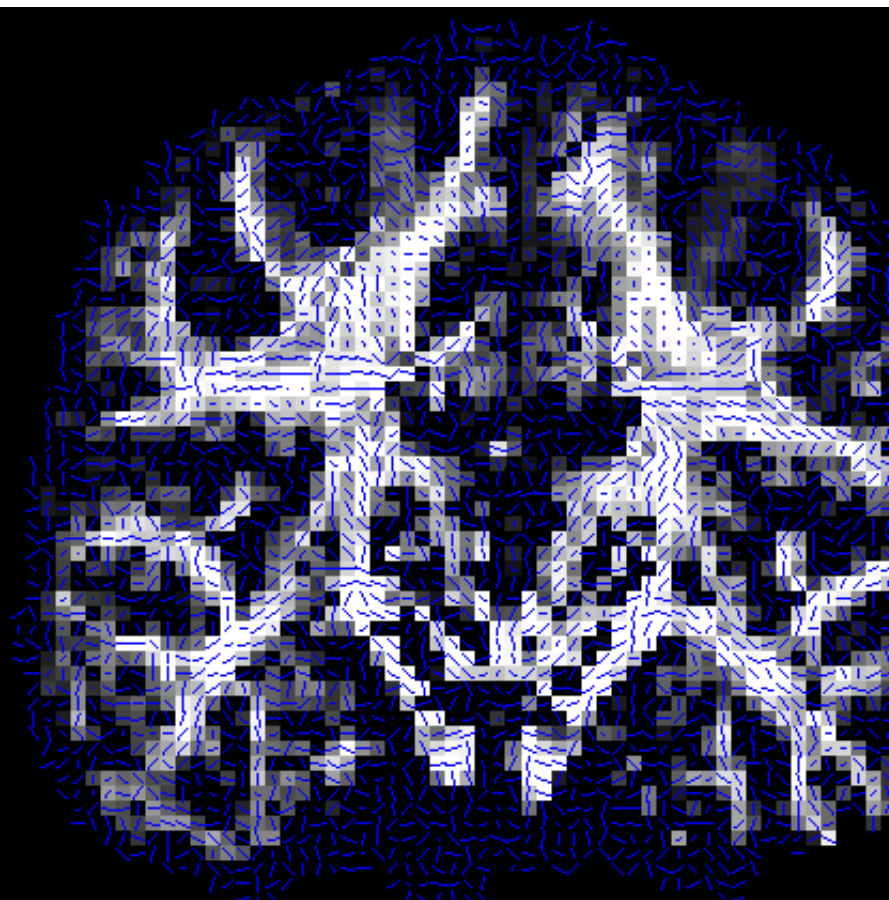
Automatic Relevance Determination (A.R.D.)

- After running BedpostX all voxels will have estimated parameters for the maximum number of sticks requested.
- But due to ARD, the sticks that are not supported in a voxel will have an almost zero volume fraction.
- We use a threshold (e.g. $>5\%$) to **exclude sticks with tiny volume fraction**.

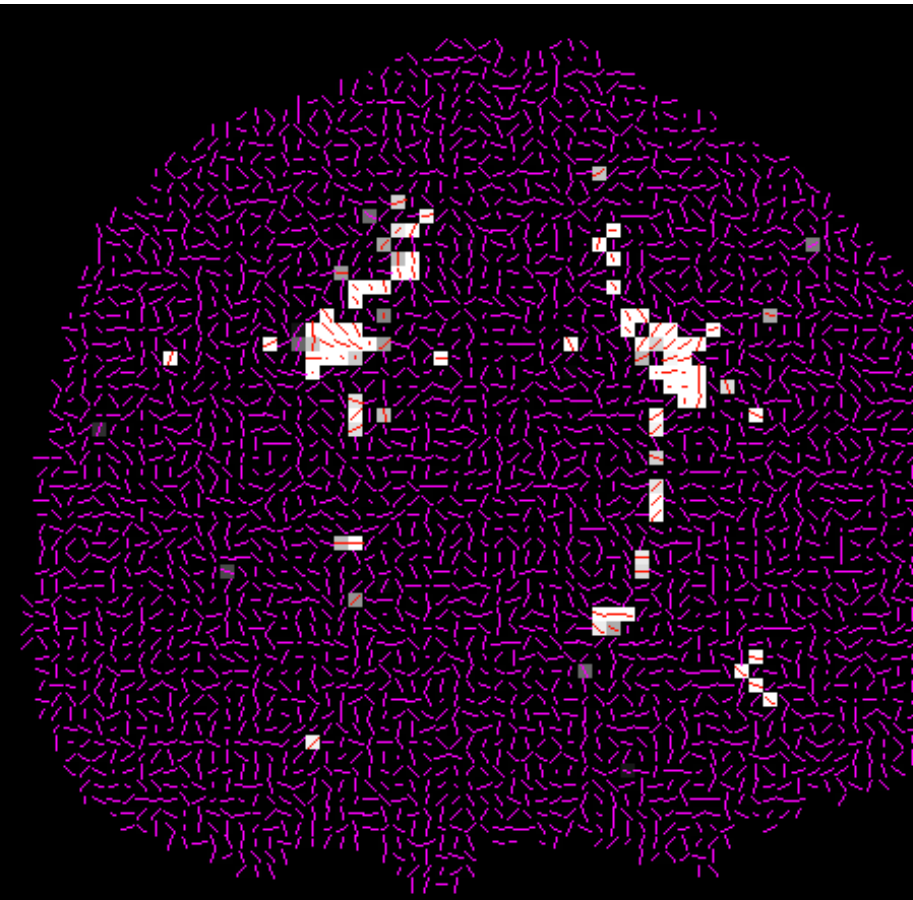
Stick1



Stick2



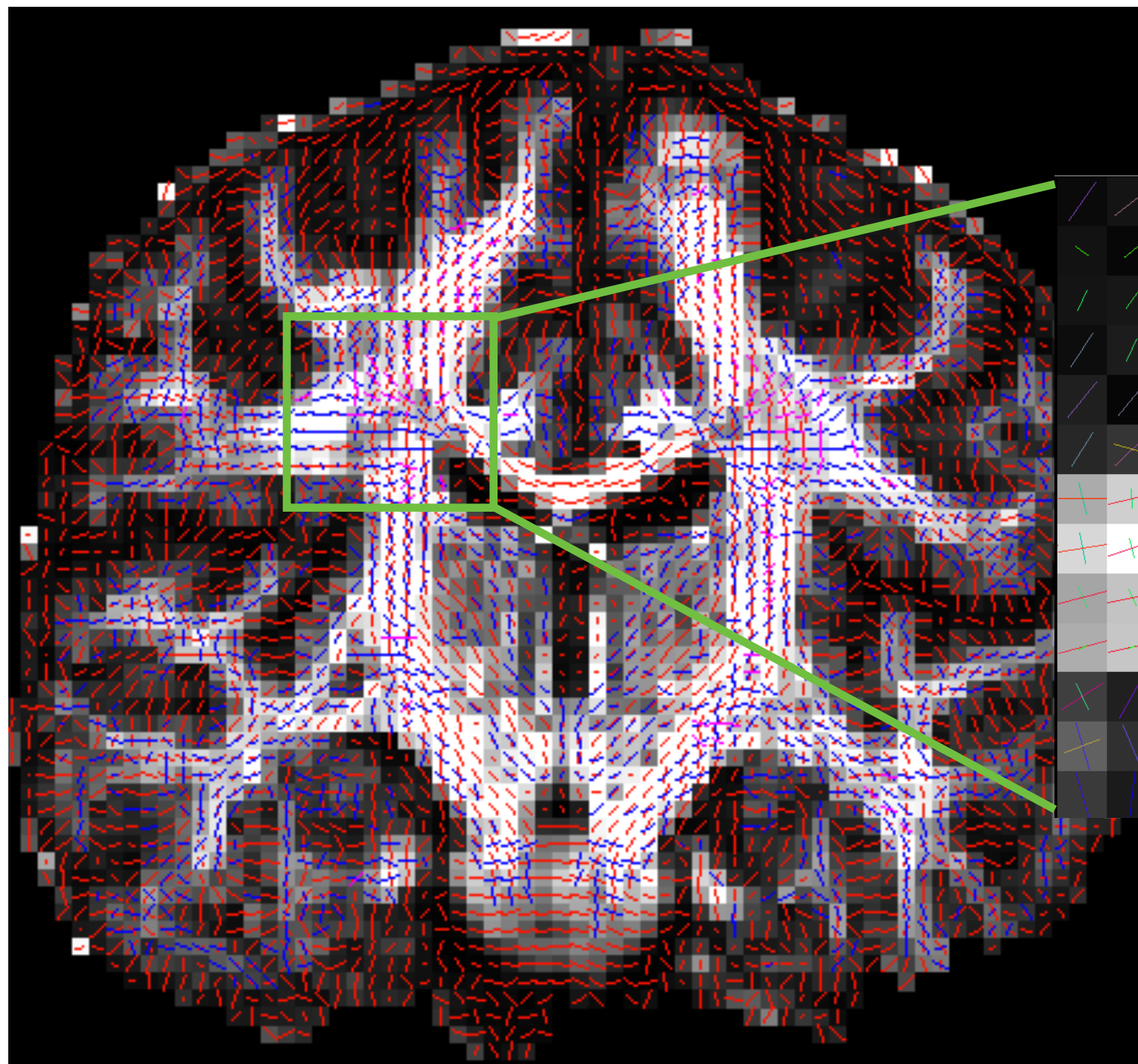
Stick3



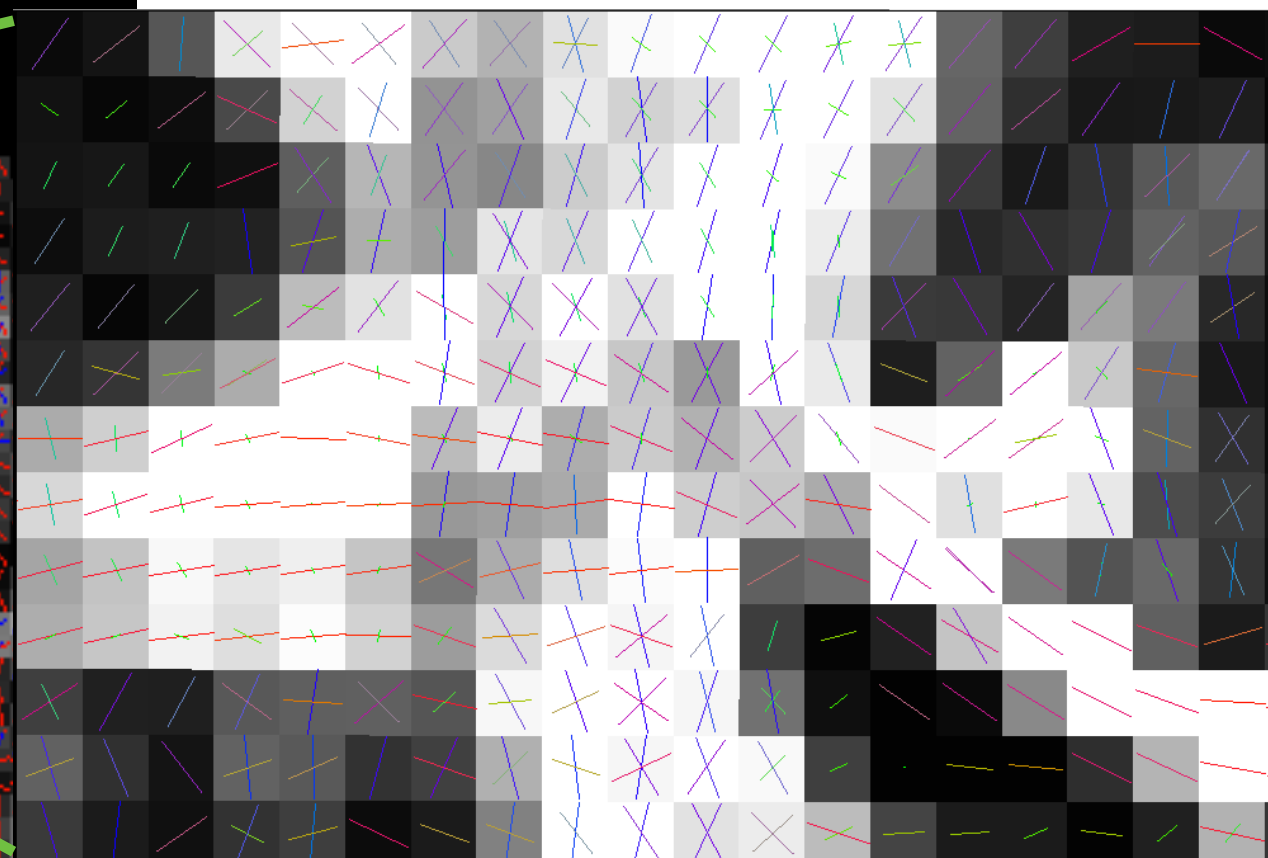


Ball & Sticks Orientations

All sticks, with secondary ones
thresholded ($f_n > 5\%$)



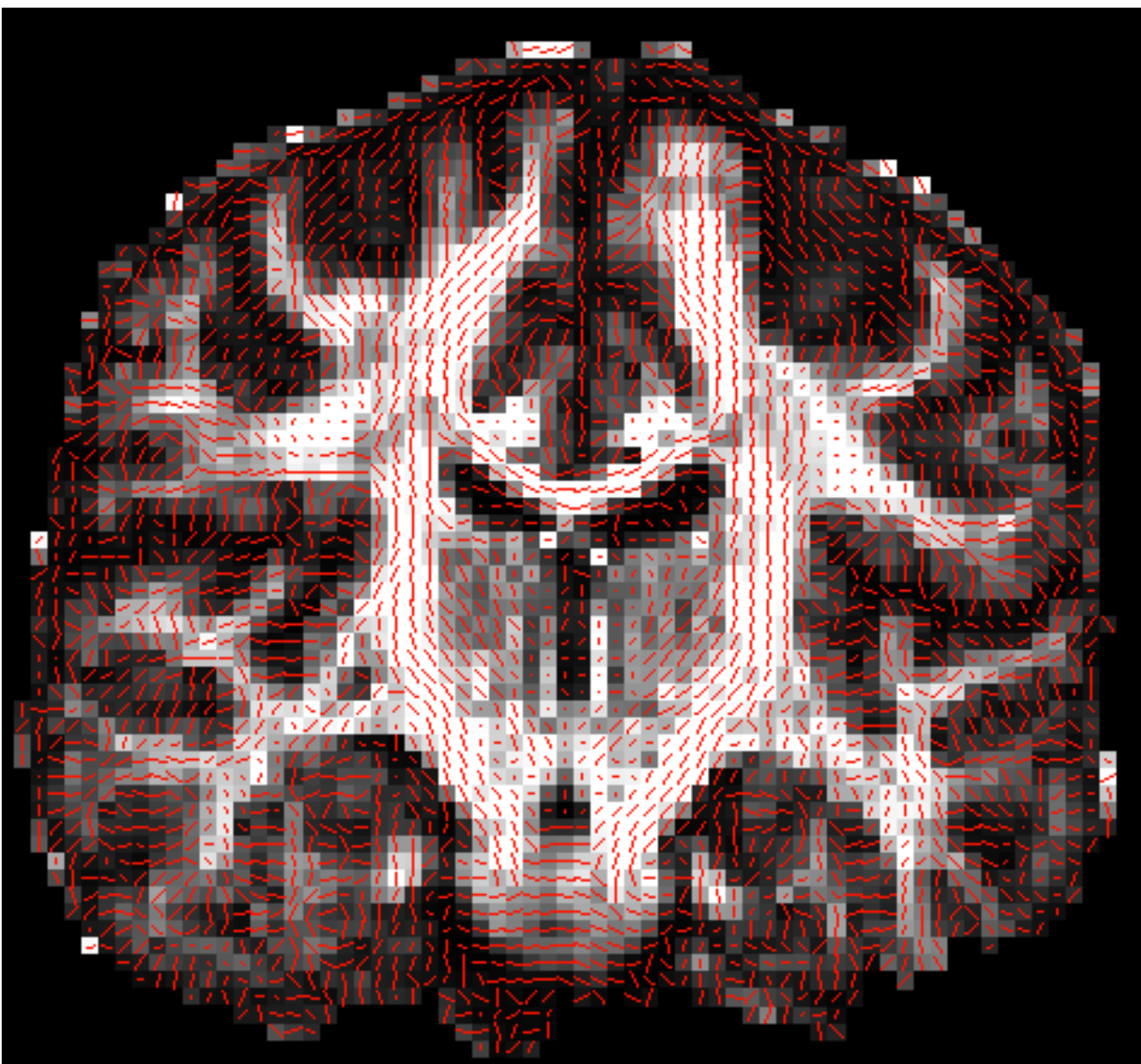
Orientations RGB-colour coded



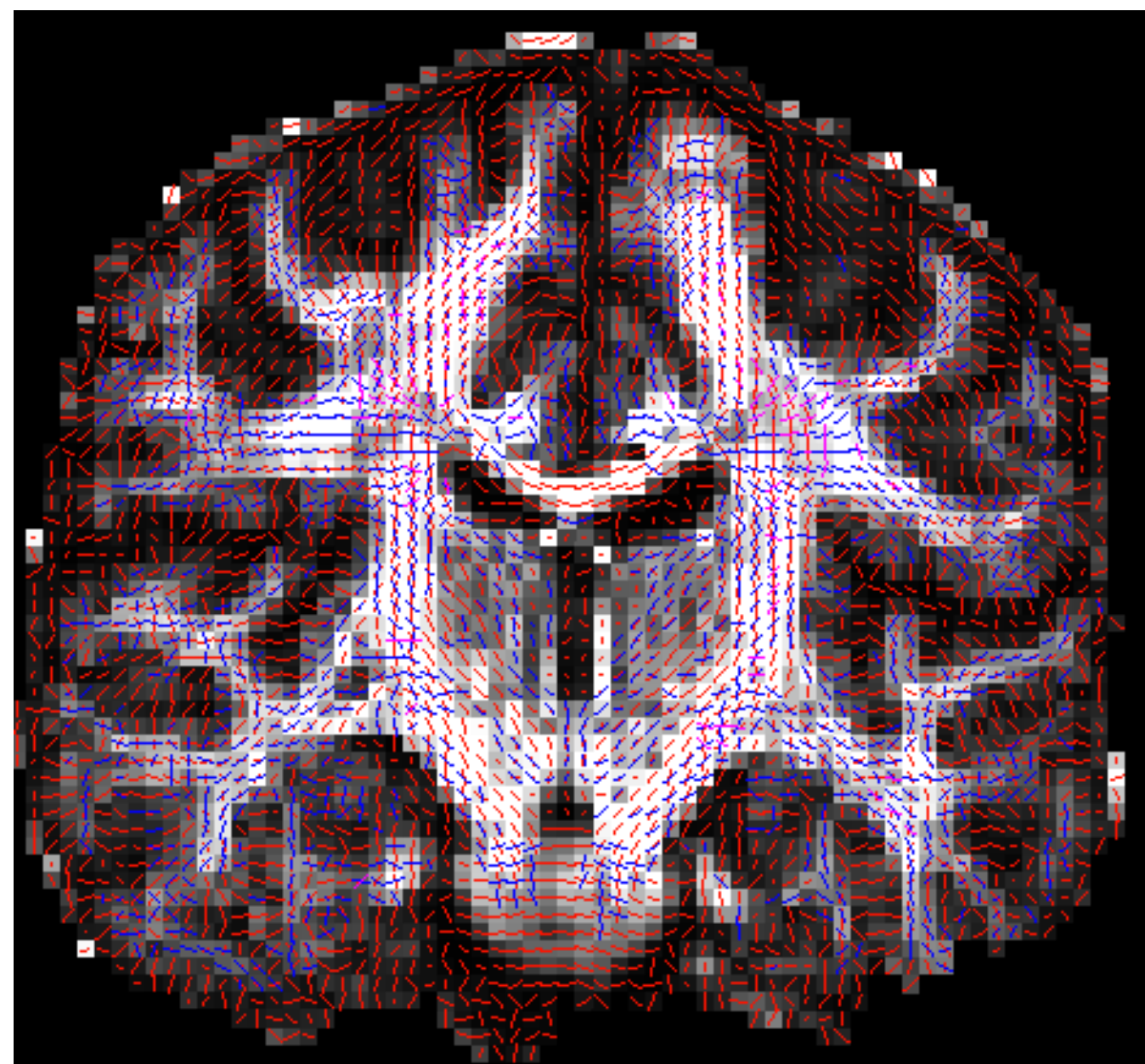


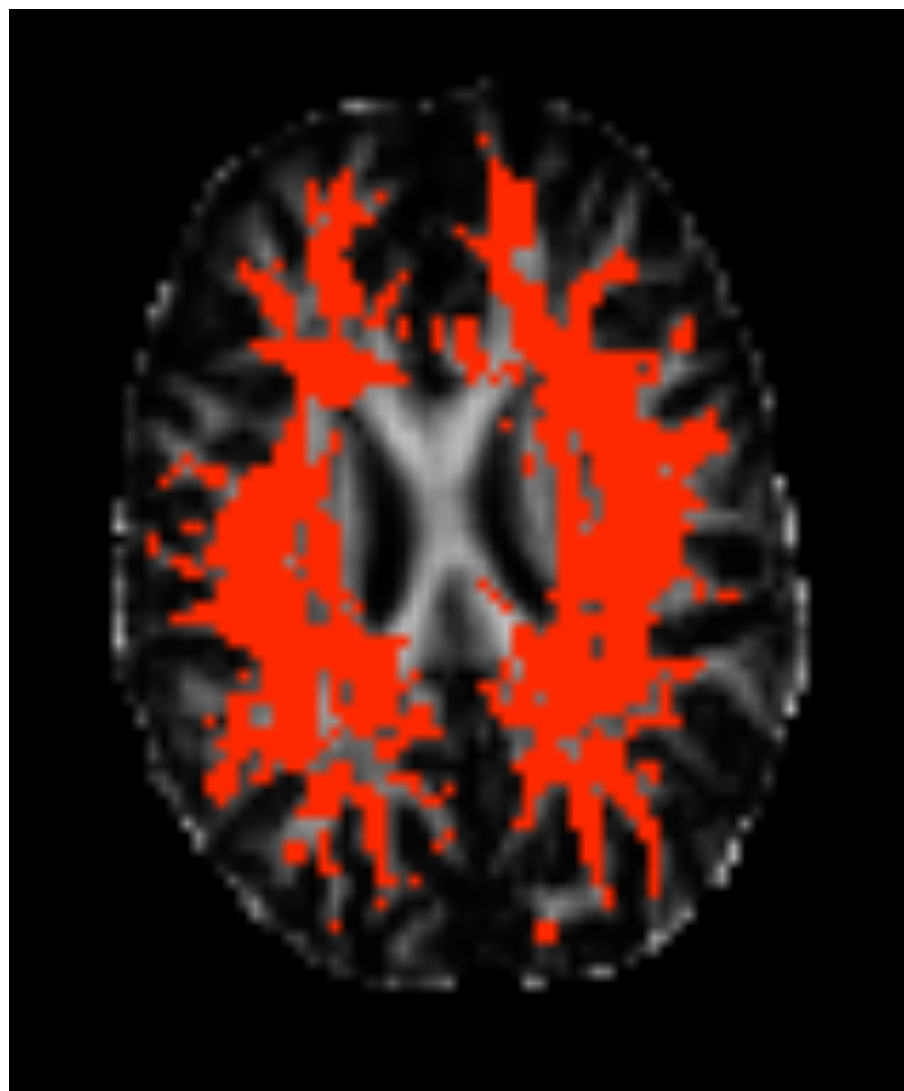
DTI vs Ball & Sticks Orientations

DTI

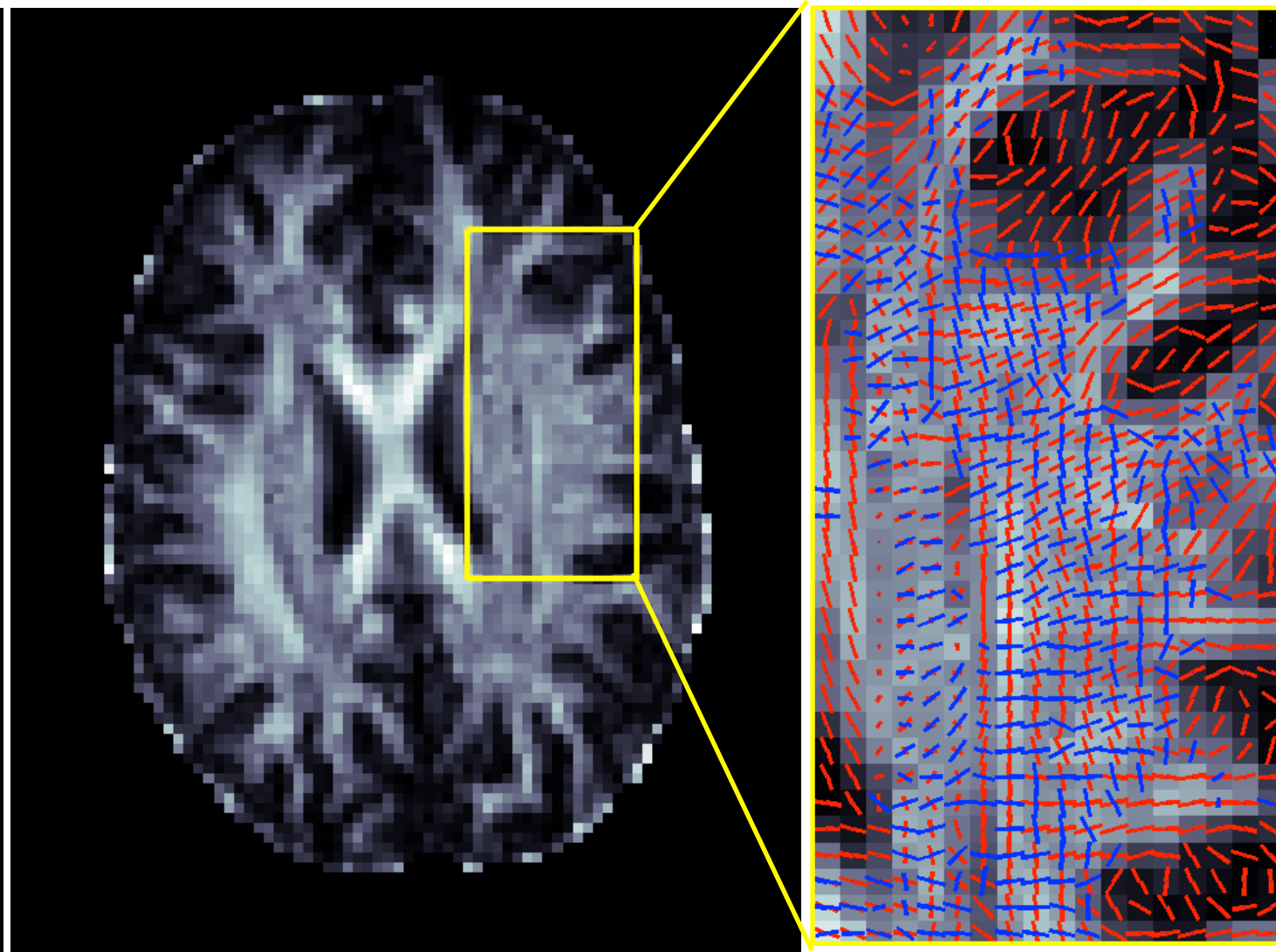


Ball & Sticks





A large portion of the WM supports crossing fibres



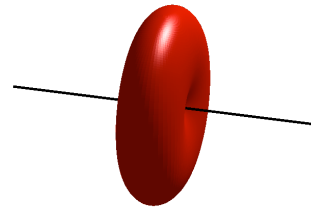
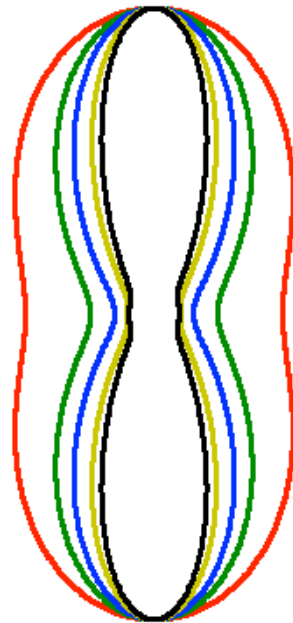
Coherence in orientations shows that we are not over-fitting (the ARD works)



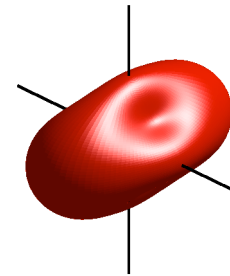
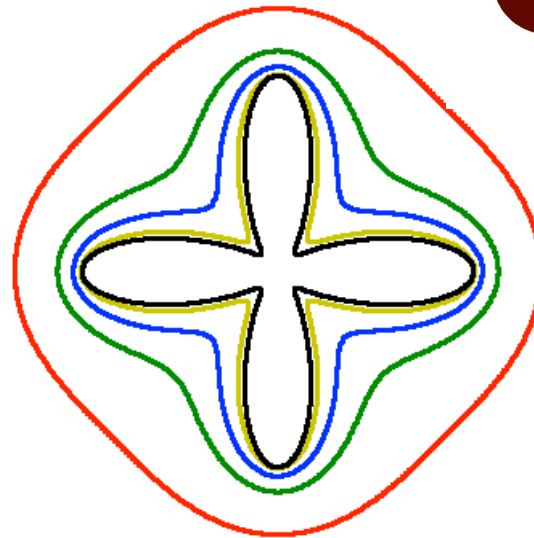
Multi-Shell Diffusion Acquisitions

Why bother?

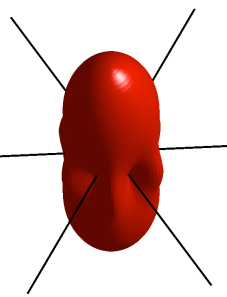
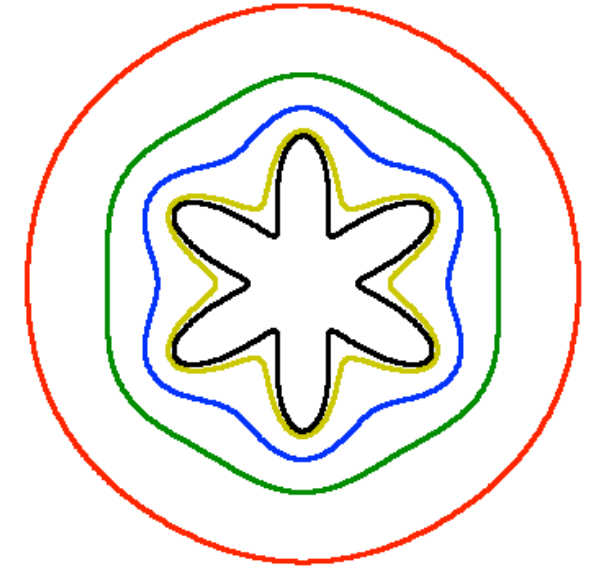
One Orientation



Two Orientations



Three Orientations



Signal at
different
b values
(s/mm²)
b=1000
b=2000
b=3000
b=4000
b=5000

Higher b value gives us more angular contrast!!!

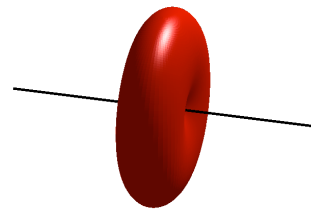
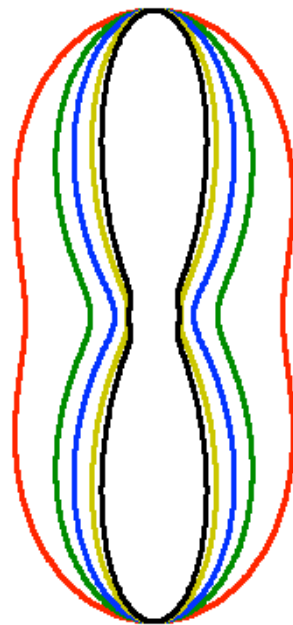




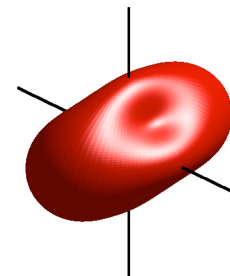
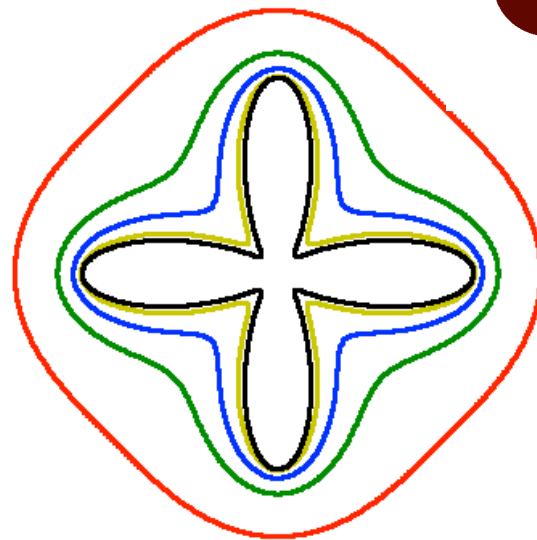
Multi-Shell Diffusion Acquisitions

Why bother?

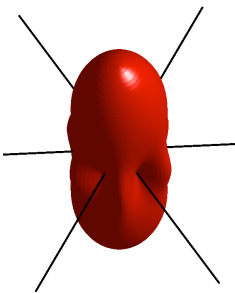
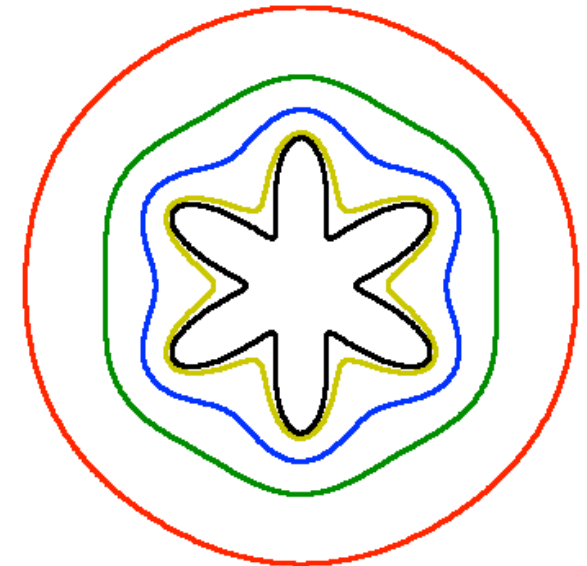
One Orientation



Two Orientations



Three Orientations



b=300

b=1000

b=2000

b=3000



But SNR goes down very quickly with b...



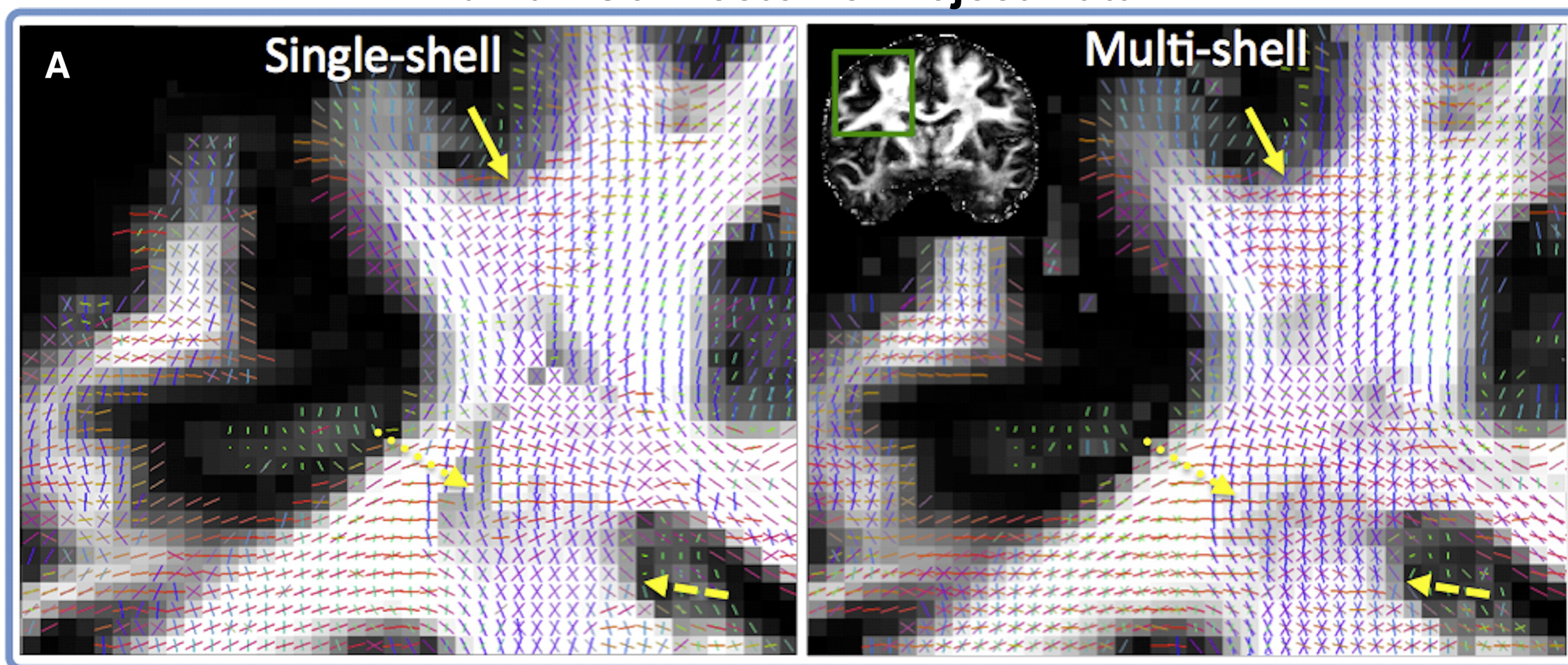


Generalised Ball & sticks Model

Gets best of both worlds

- Multi-shell model (or model=2) in Bedpostx options.
- Allows representation of multiple diffusivities within a voxel (rather than just one).
- More accurate model for multi-shell data & partial volume effects.

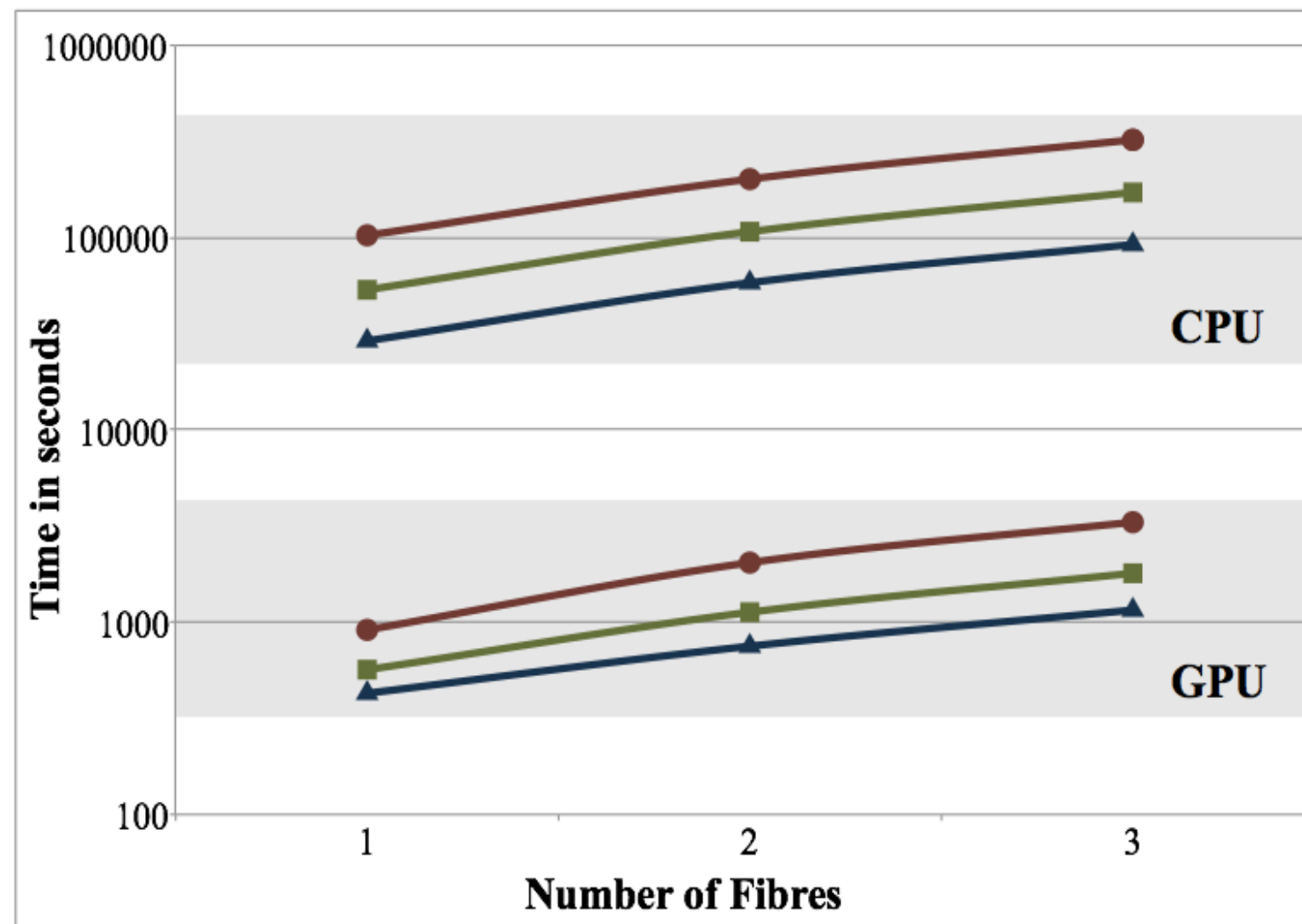
Human Connectome Project Data



*Jbabdi, Sotiropoulos et al, MRM 2012

* Sotiropoulos, Jbabdi et al, NeuroImage 2013

Faster bedpostx on GPUs



Directions number: 64 128 256

50x-150x Speedup using GPUs