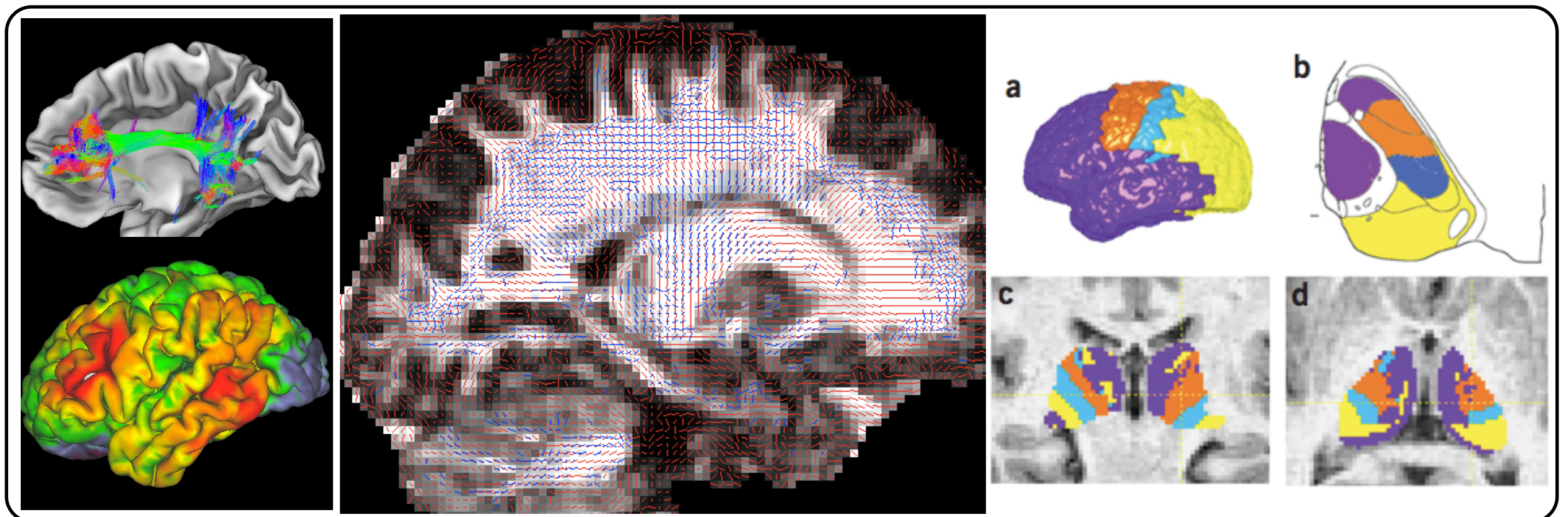




# Diffusion Tractography

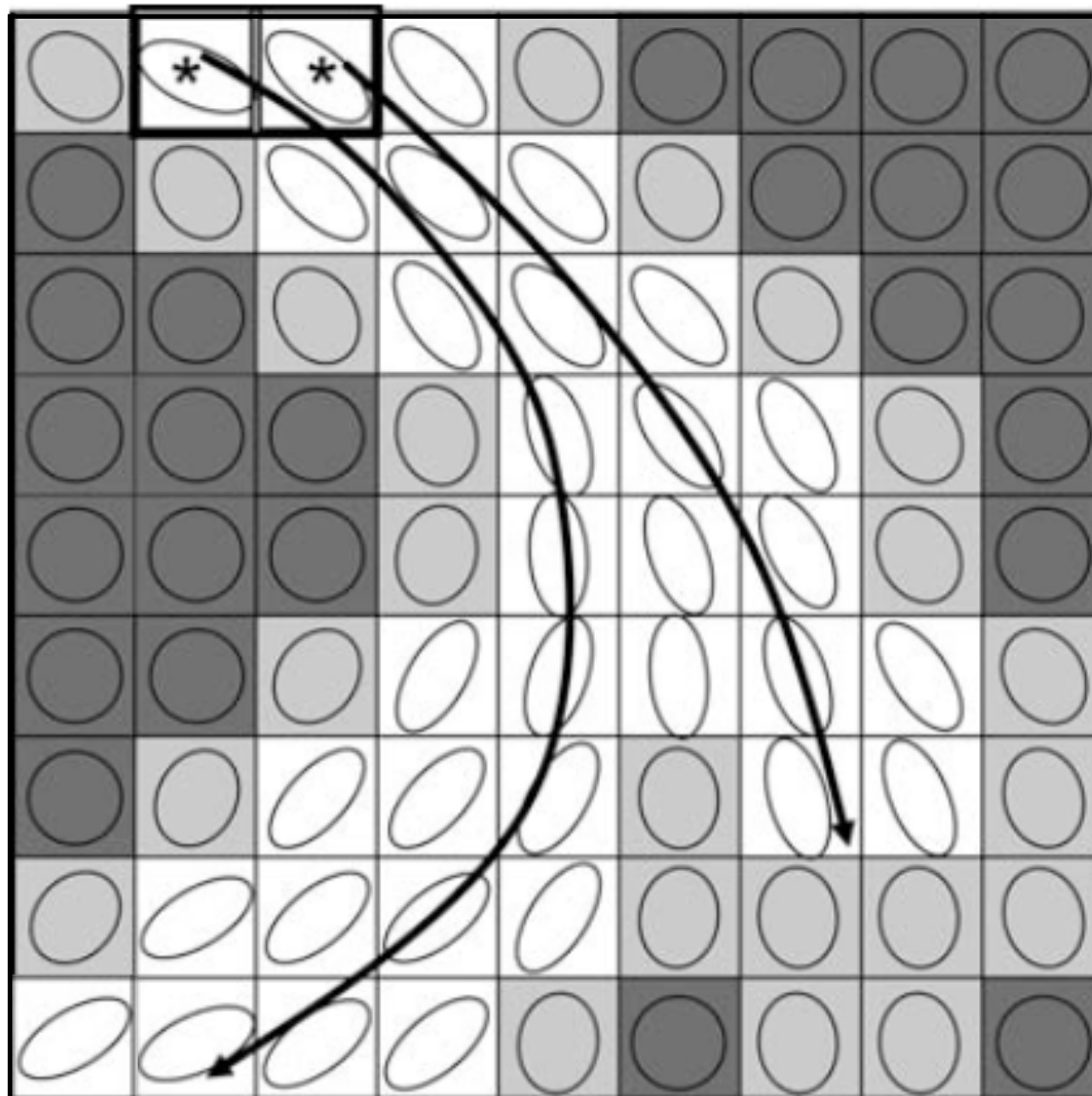
- Goal of tractography
- Estimating Fibre Orientations - BEDPOSTX
- Probabilistic Tractography - PROBTRACKX
- ProbtrackX outputs
- Tractography limitations





# DTI Streamline Tractography

Seed  
region



Formally, we solve numerically the differential equation:

$$\frac{d\mathbf{r}(s)}{ds} = \mathbf{v}_1(\mathbf{r}(s)), \quad \mathbf{r}(0) = \mathbf{r}_0$$

Position  
along a curve

Principal eigenvector  
 $\mathbf{v}_1$  at position  $\mathbf{r}(s)$

Starting  
Position



# DTI Streamline Tractography

But When to Stop?

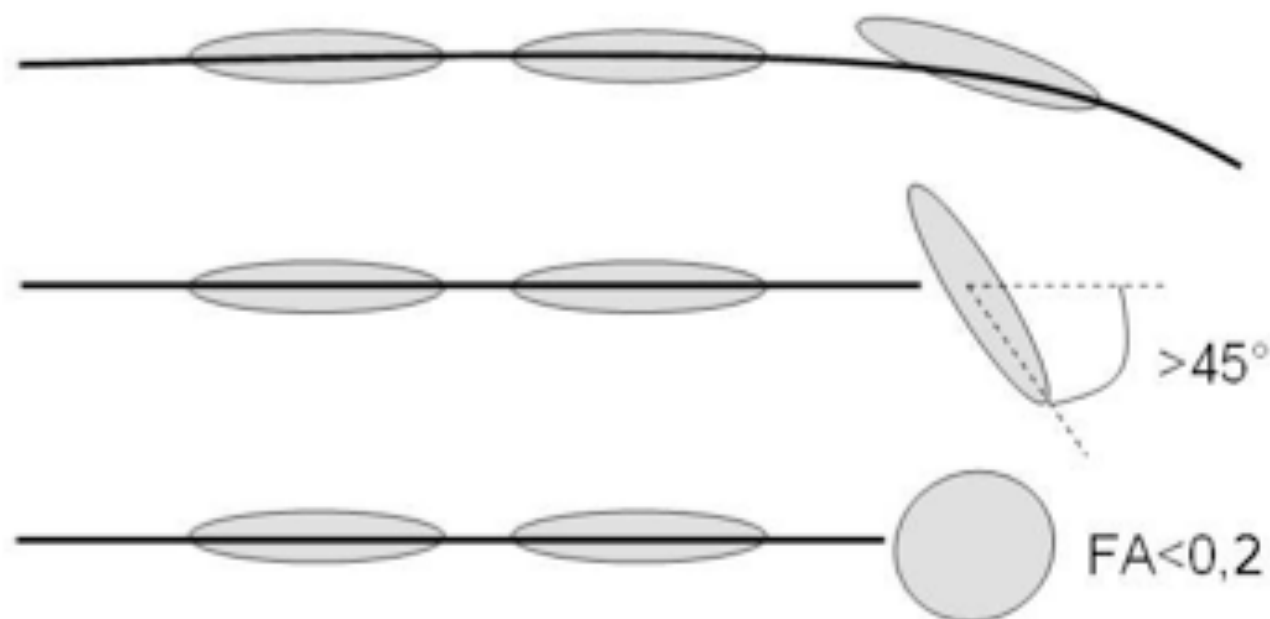
Heuristics to avoid error propagation.

+ Knowledge of the anatomy

**Curvature Change Threshold:** To avoid crossings of boundaries and very bended trajectories, impose a smoothness criterion.

**Anisotropy Threshold:** To avoid propagating in regions where  $\mathbf{v}_1$  is meaningless.

**Anatomical criteria** (e.g. reach grey matter)







# Streamline tractography can dissect major bundles



arcuate fasciculus



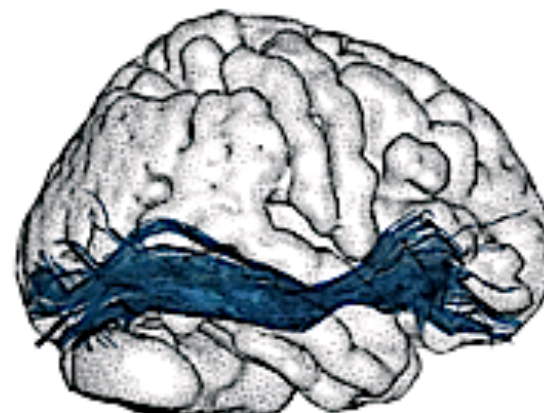
cingulum bundle



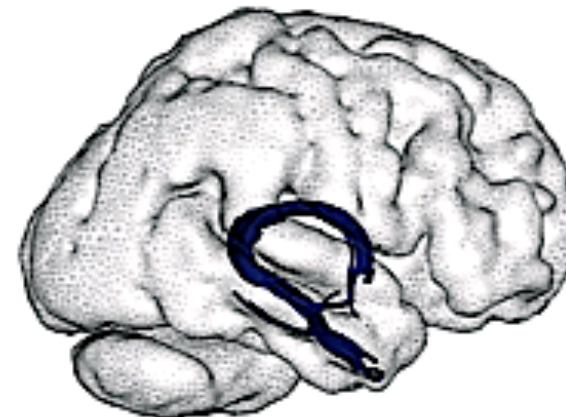
inferior longitudinal fasciculus



corpus callosum



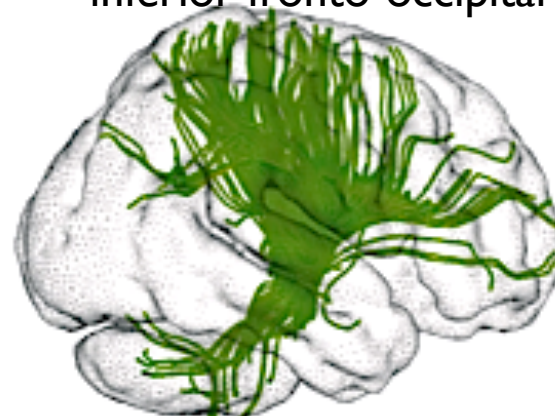
inferior fronto-occipital



fornix



uncinate fasciculus



corona radiata



cerebellar tracts



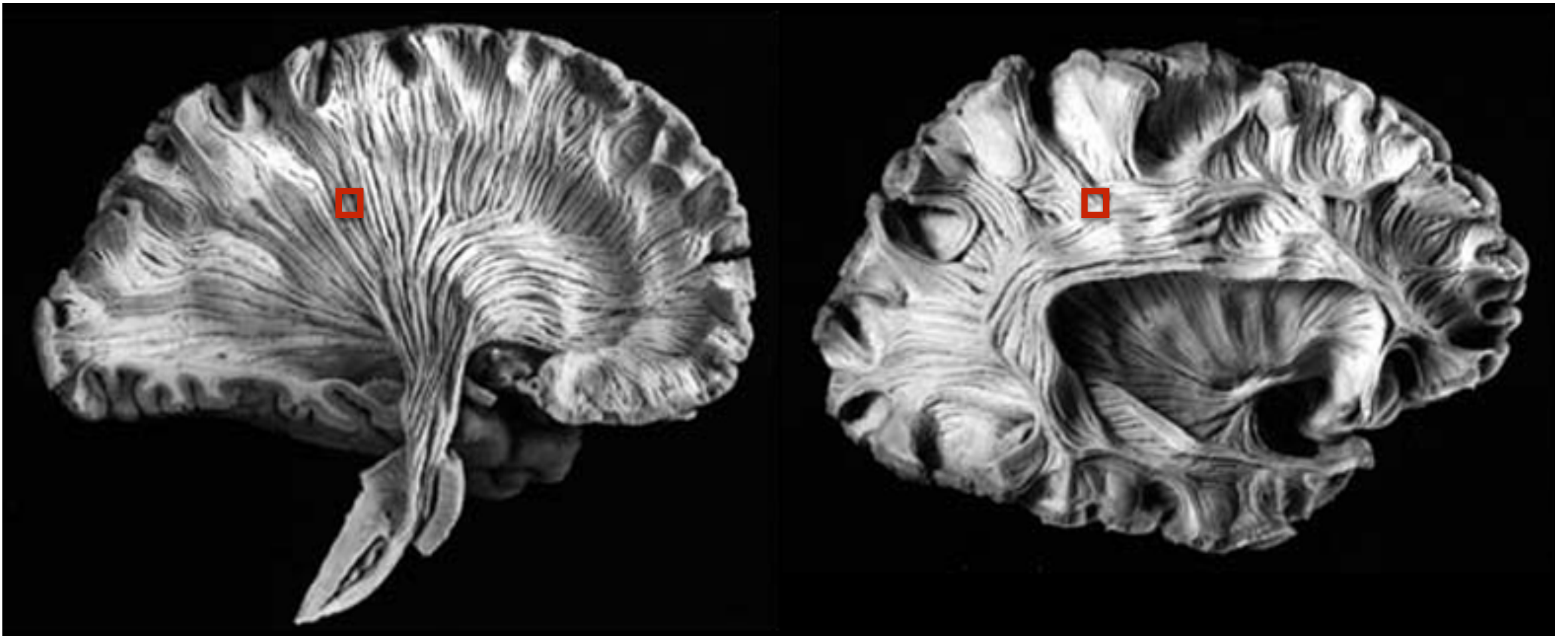
## DTI Streamline Tractography Summary

- Use the major axis of the DTI ellipsoid as a fibre orientation estimate.
- Propagate curves within this vector field until empirical thresholds are exceeded.
- Major fibre bundles can be reconstructed.





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



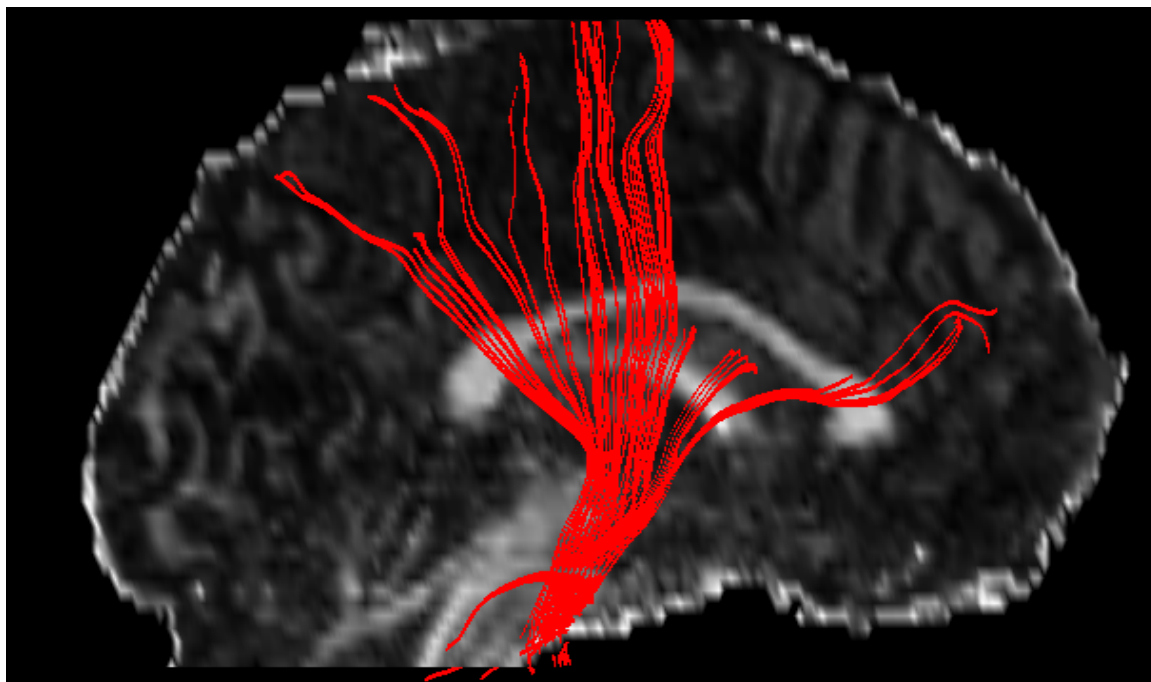
## Streamlining reproducibility

Repeat an acquisition many times and repeat streamline tracking.

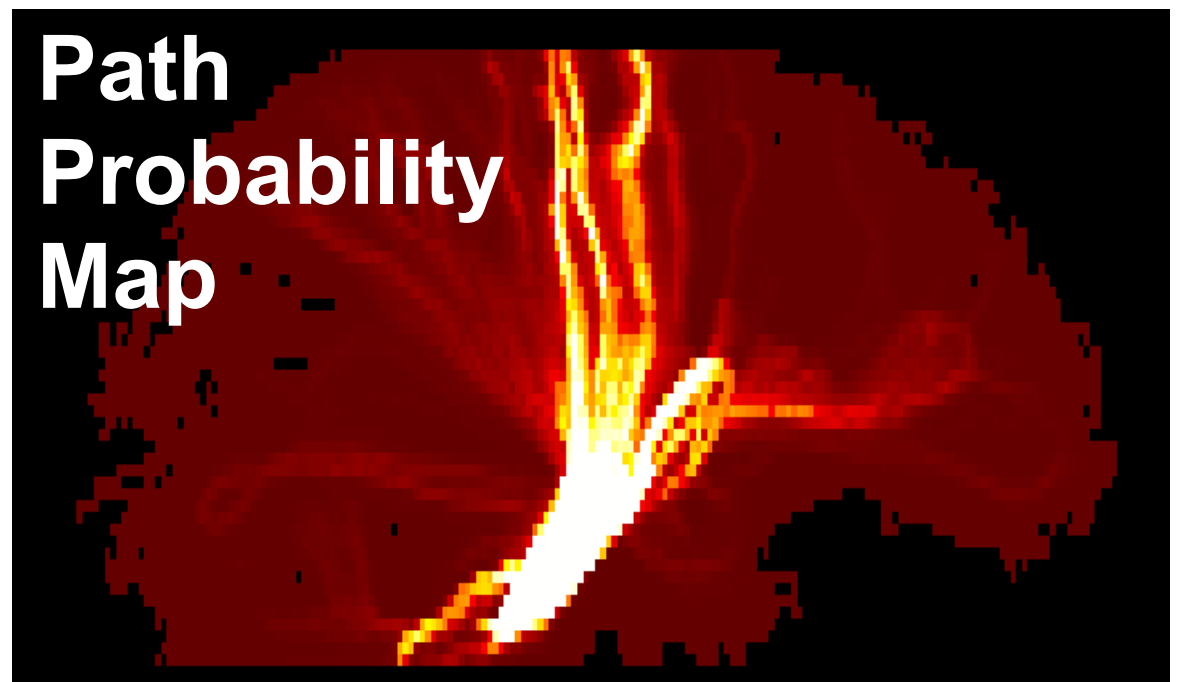
Due to uncertainty in  $\mathbf{v}_1$ , curves will not perfectly overlap

Create a map that shows the degree of overlap across the trials.

Streamlines from a single dataset



Map that shows where results  
across datasets overlap



Low Reproducibility

High Reproducibility





# Probabilistic Tractography

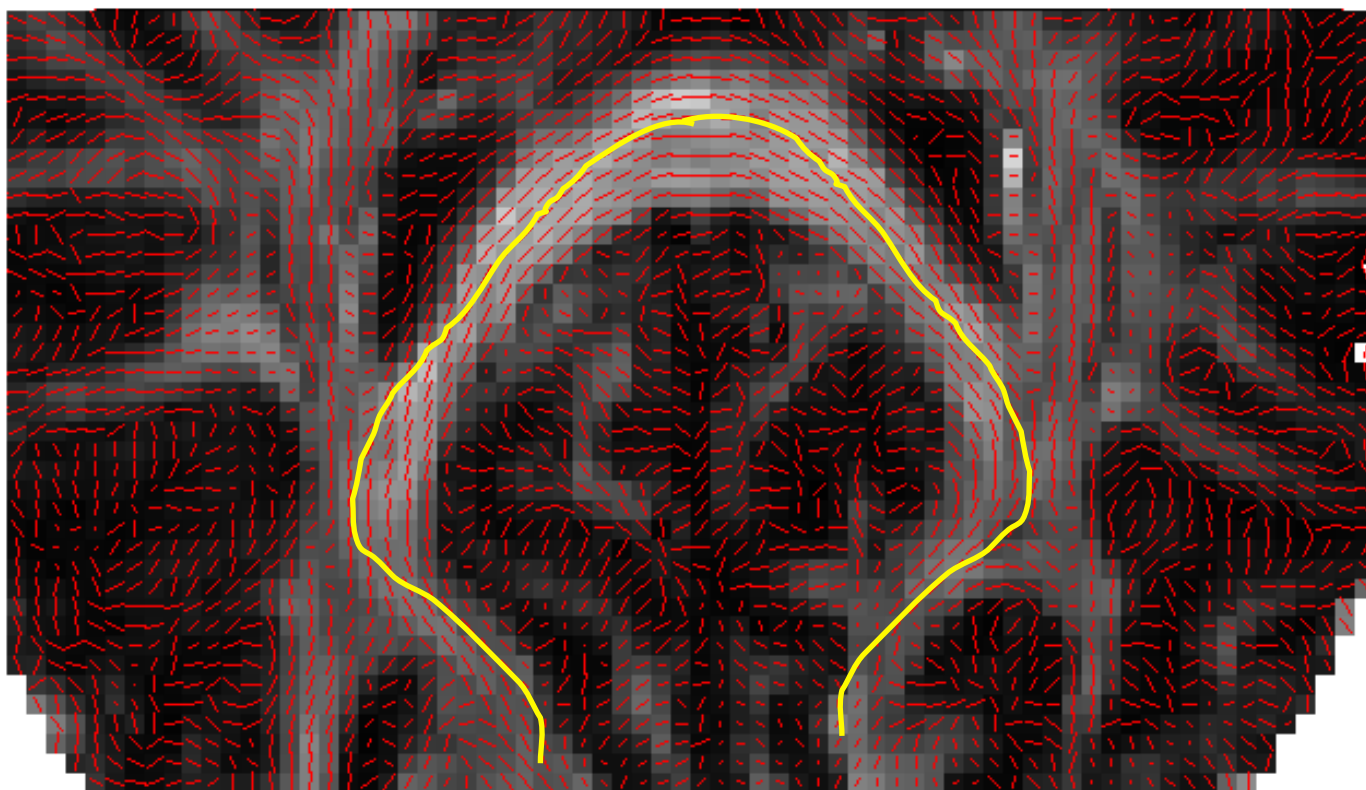
- We normally have one dataset per subject, not many.
- Probabilistic Tractography as a two-step process:
  - a) Use DWI data and a model to infer a fibre orientation **and its uncertainty** in each voxel.
  - b) Use the estimates **and the uncertainty to build a path probability map** to a seed.



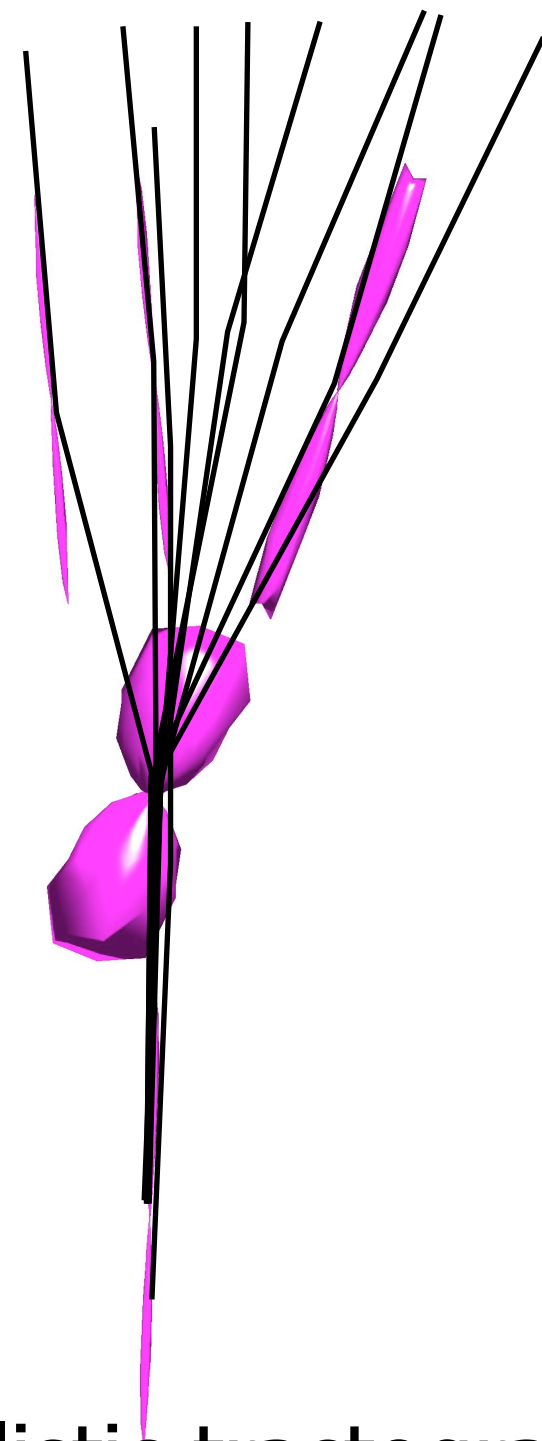


# Probabilistic tractography

- But now, we no longer have a single direction at each voxel. How can we do tractography?



'Streamlining'

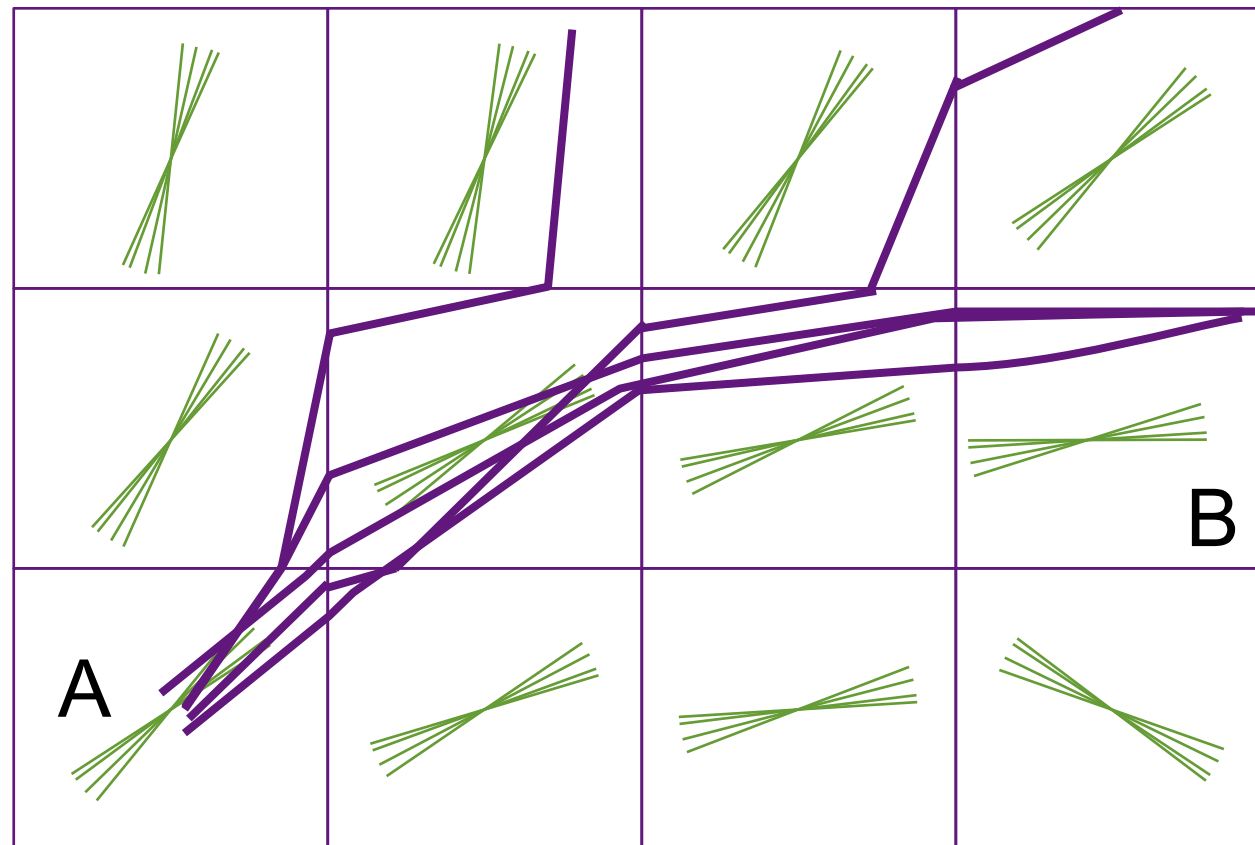


Probabilistic tractography

Behrens et al, 2003, Parker et al. 2003,  
Hagmann et al 2003, Jones et al. 2004



# Probabilistic Tractography - Propagating the Uncertainty

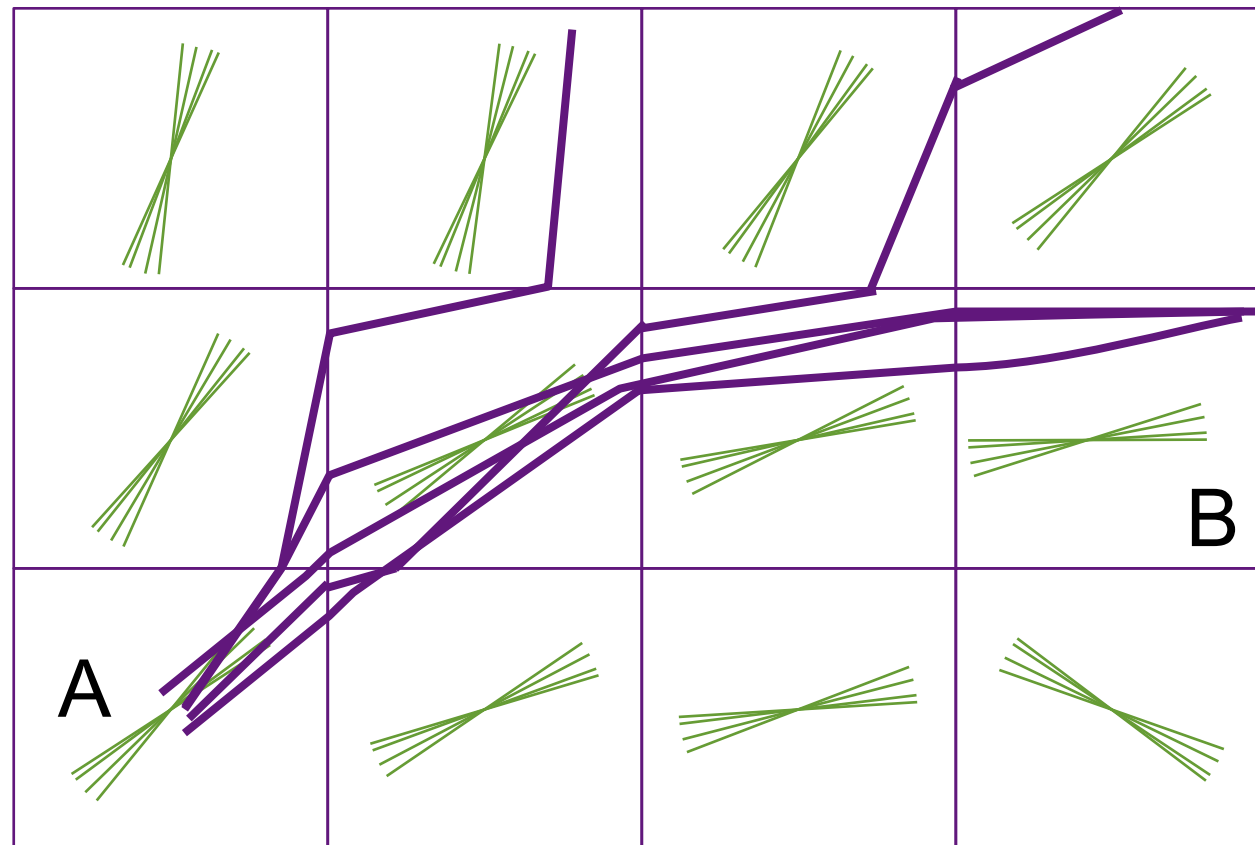


Behrens et al, 2003  
Parker et al, 2003

- Propagate N streamlines from a seed, but for each propagation step choose randomly an orientation from the underlying distribution.
- Build a spatial distribution of curves that mimics the overlapped results from multiple deterministic tracking on multiple scans



# Probabilistic Tractography - Propagating the Uncertainty



Behrens et al, 2003  
Parker et al, 2003

Define the degree of overlap at each location B, as:

$$P_{AB} = M/N$$

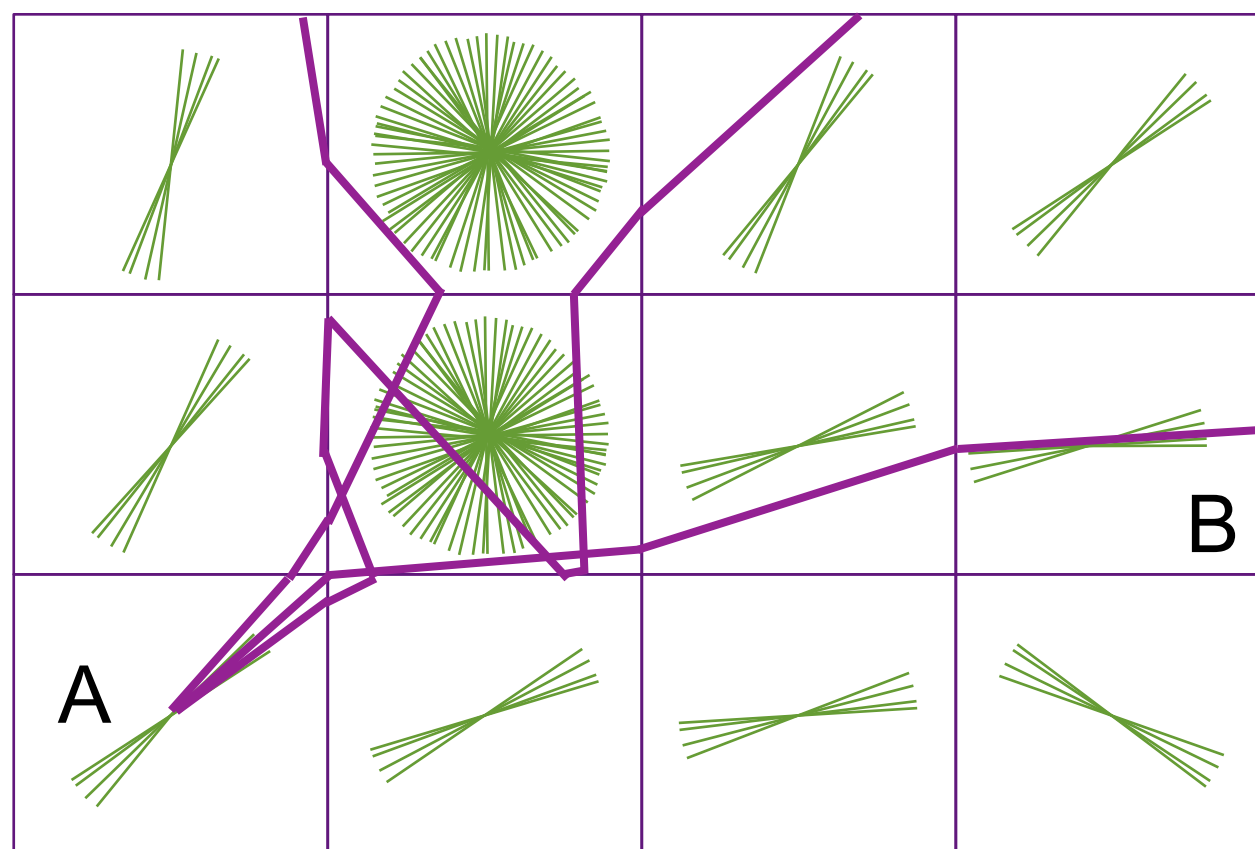
M: number of streamlines that go through B  
N: total streamlines generated from A

This is the probability of a curve starting at A and going through B.





# Probabilistic Tractography - Propagating the Uncertainty

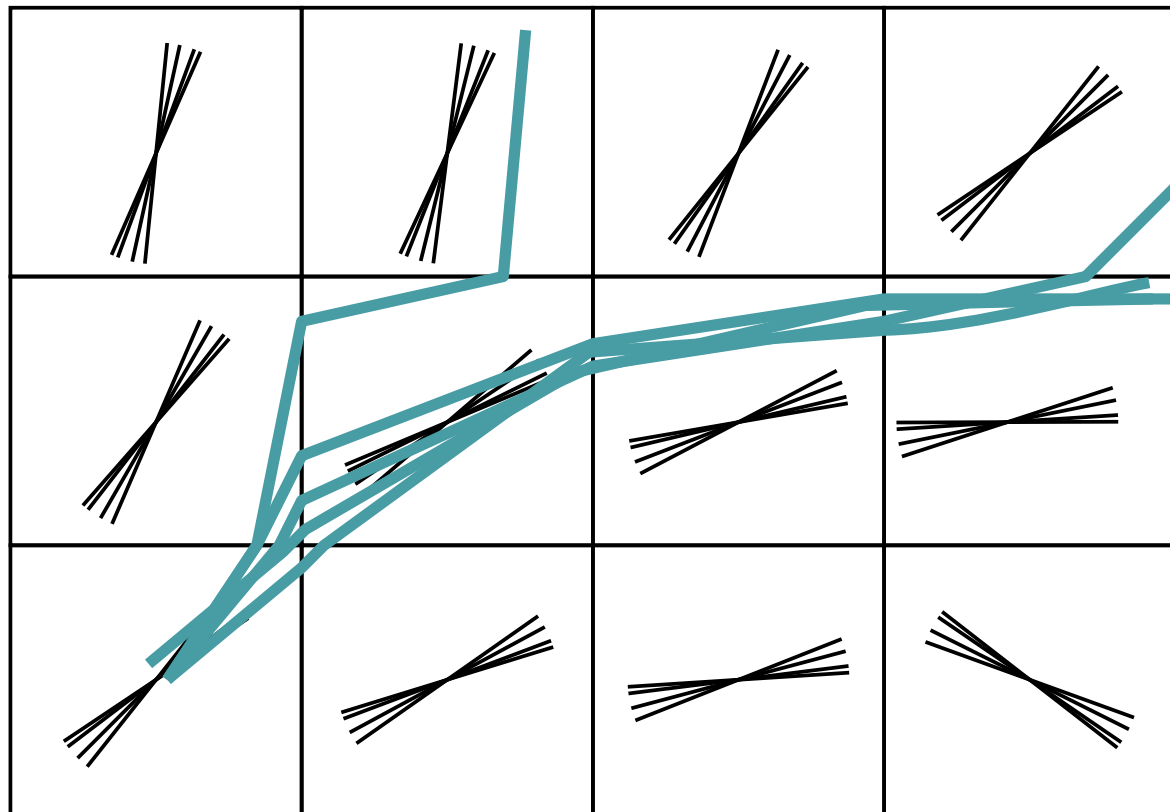


Behrens et al, 2003  
Parker et al, 2003

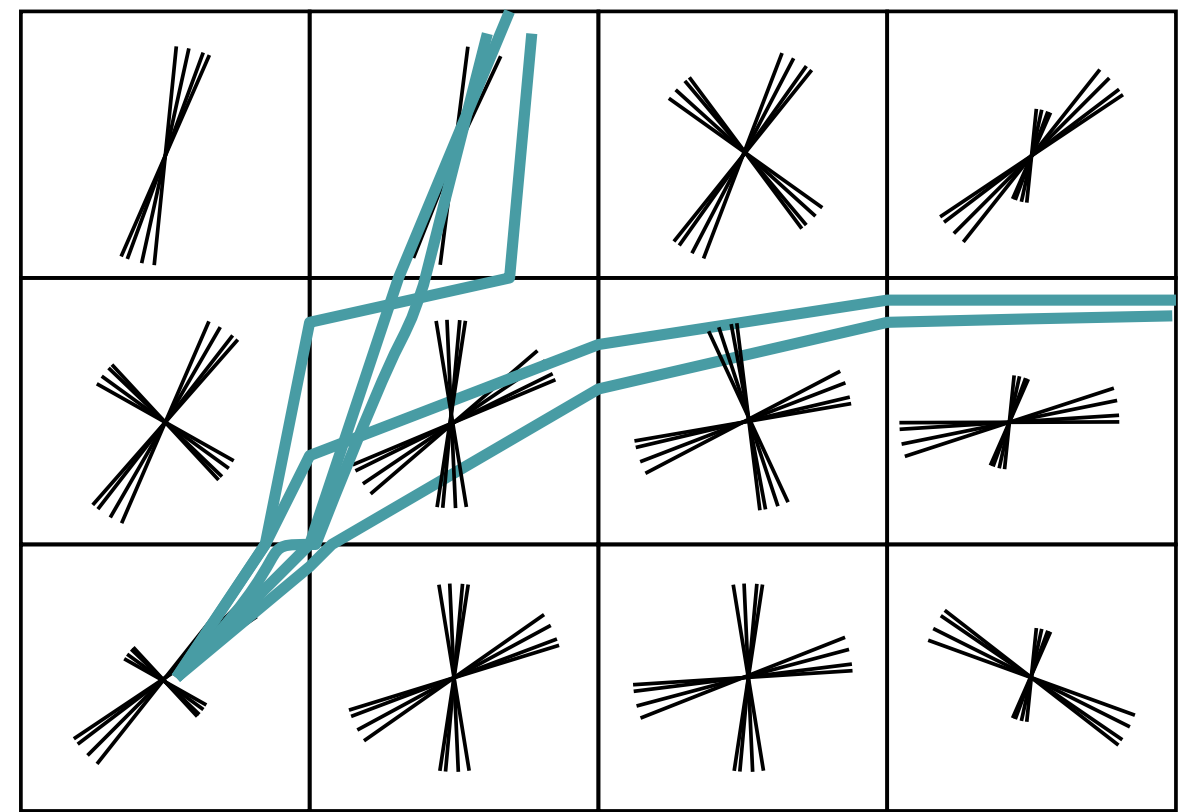
- Can now propagate through isotropic regions (e.g. GM).
- Do not need to stop when anisotropy is low, as in deterministic tracking.
  - The high uncertainty will be reflected in the probability map.
- Still impose a curvature threshold to avoid swirled trajectories.



# Probabilistic Tractography in Multi-Fibre Fields



Behrens et al, 2003, Parker et al. 2003,  
Hagmann et al 2003, Jones et al. 2004



Parker & Alexander 2003,  
Behrens et al, 2007

When multiple fibre orientations exist in a voxel, choose the one that is most compatible with the incoming trajectory.

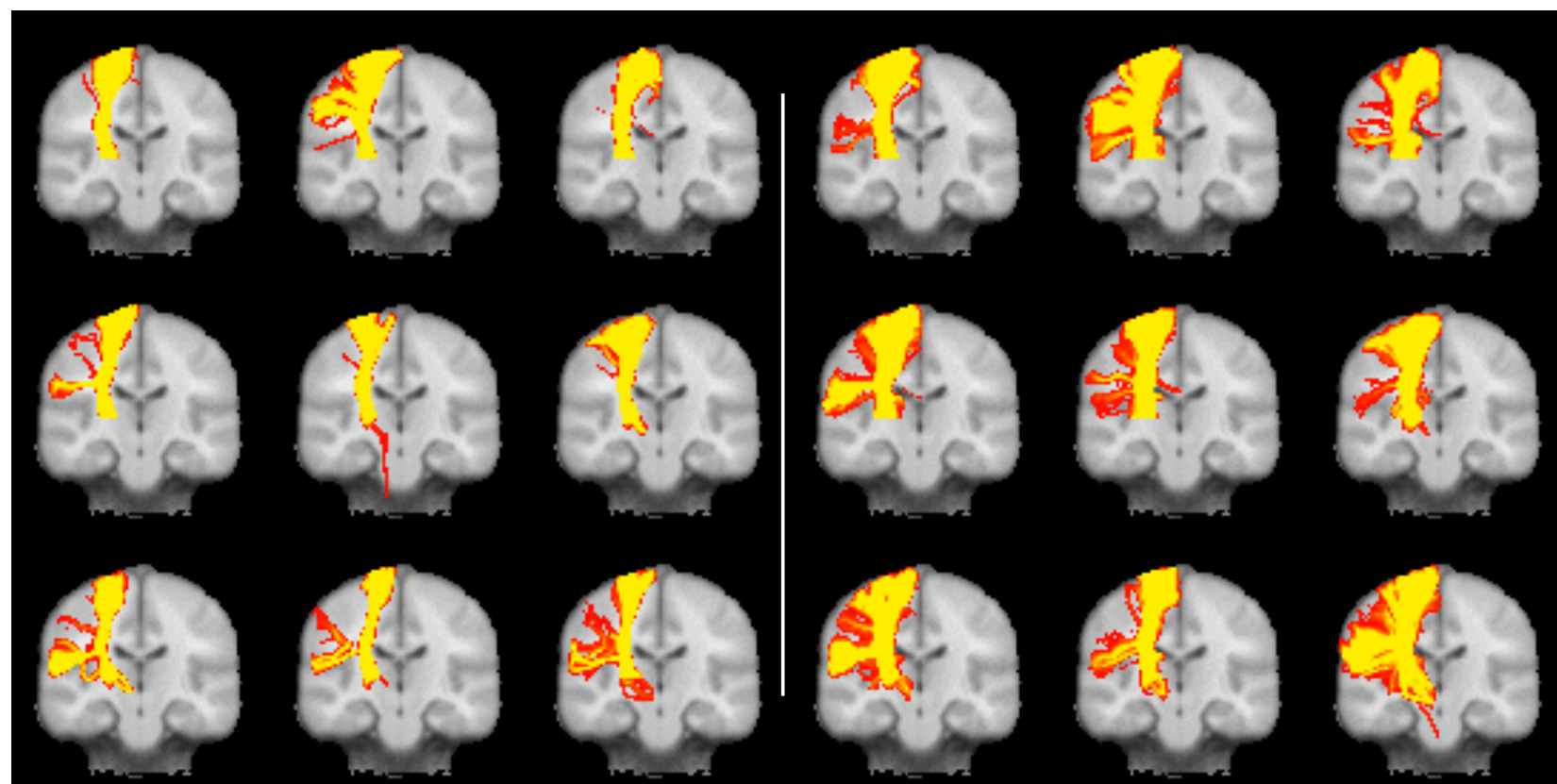
# Probabilistic Tractography in Multi-Fibre Fields

## Examples

Cortico-spinal tracts.  
9 subjects

Behrens et al, 2007

SLF  
Internal capsule --- Primary motor cortex



one fibre

two fibres

\* If one fibre is modelled and we track through a crossing, a) we may not make it through the crossing, b) if we make it, the connectivity index will be relatively low.



# Probabilistic Tractography in Multi-Fibre Fields

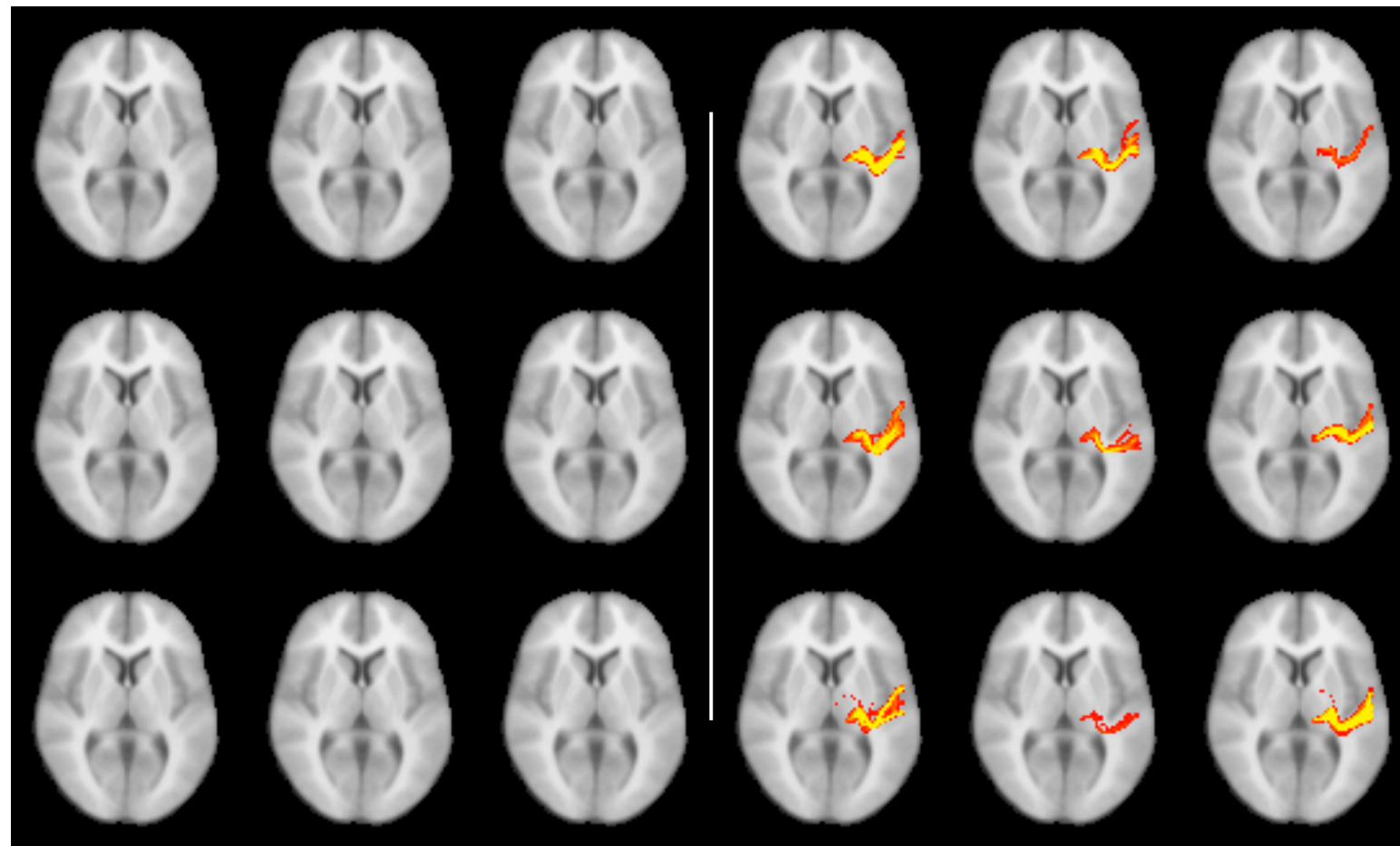
## Examples

Acoustic radiations.

9 subjects

Behrens et al, 2007

Opt.R  
MGN --- Primary Auditory cortex



one fibre

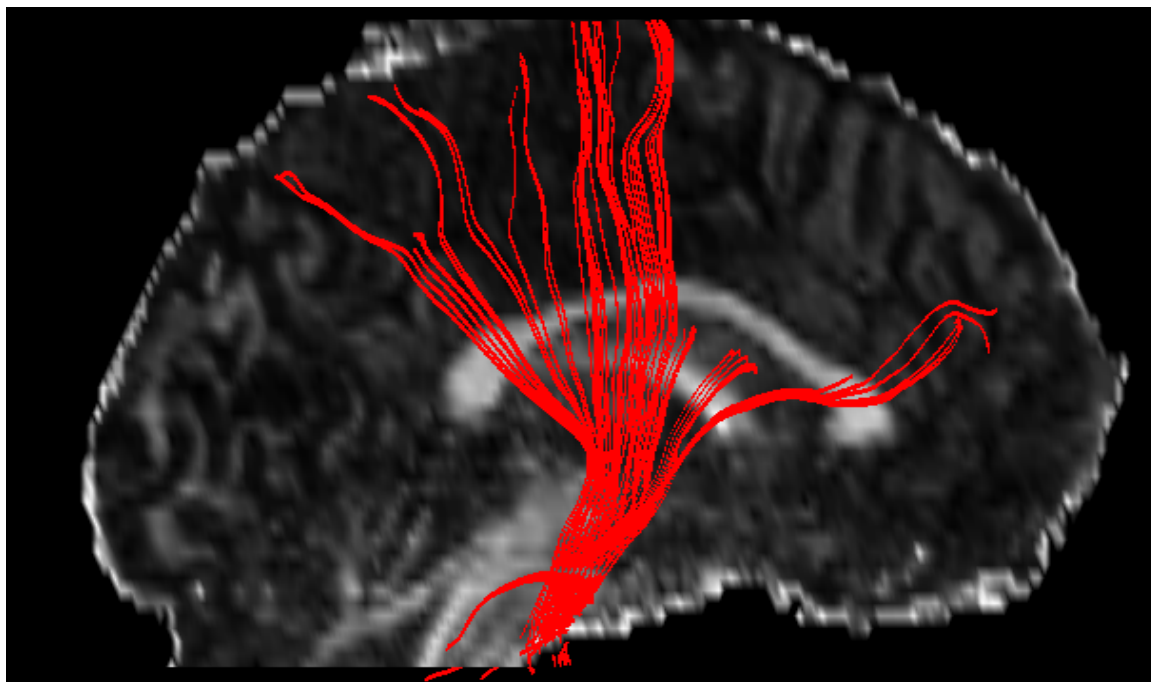
two fibres



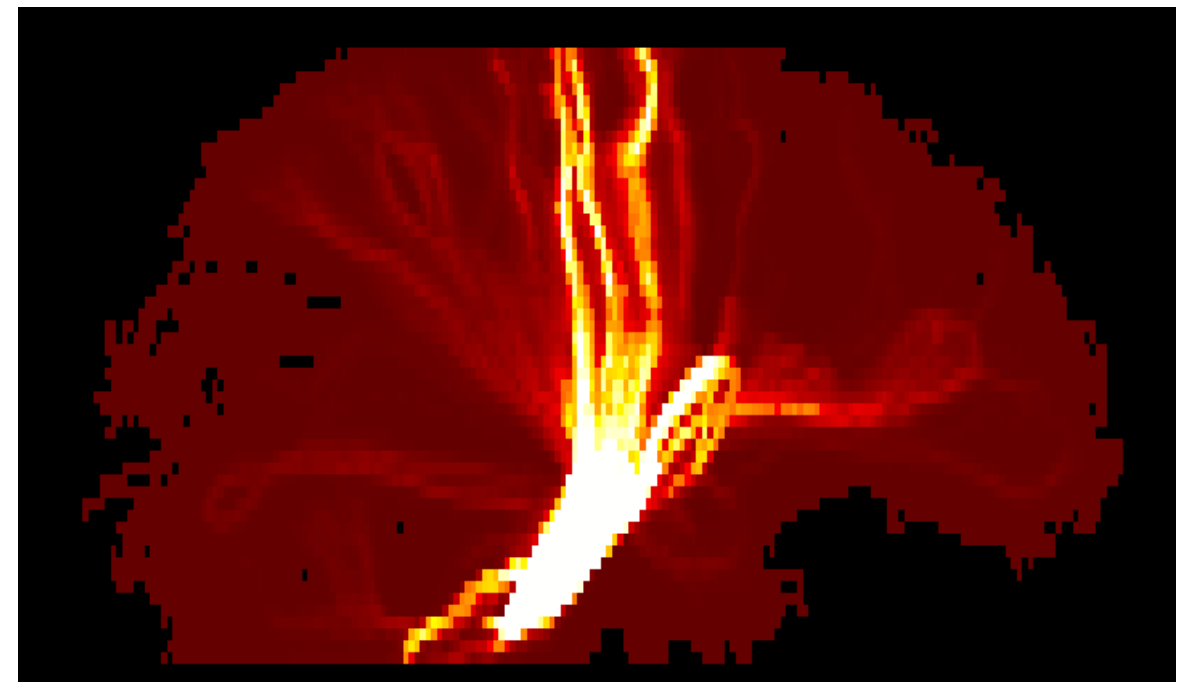
## Path Probability Map

- Recall that it assesses how reproducible results are
- Often called “connection probability”, “connectivity index”, “connectivity strength”. But it does not quantify how strong a connection is...
- Rather, how robust it is against noise/uncertainty

Deterministic Tractography



Probabilistic Tractography



Low Probability

High Probability





# Probabilistic Streamline Tractography Summary

- Needs apart from orientation estimates, an estimate of their uncertainty. Does not need to be the ball and stick model, the DTI model can be used instead!
- Propagate streamlines repeatedly from a seed, but the orientation field is no longer deterministic. In each propagation step choose randomly an orientation from the underlying distribution.
- A connection probability value  $\geq 0$  can be obtained from a seed A to any voxel in the brain B. This assesses **the reproducibility of the path from A to B, along which water molecules preferably diffuse.**

