

# Contrôle qualité et analyse des données IRM de diffusion (DWI)

## exemples d'outils performants et standardisés

*Réunion Mensuelle de Neuro-Imagerie*

23 / 11 / 2023

Julien Sein (INT)



# Introduction / Motivation for this talk

- Diffusion weighted MRI is one of the oldest MRI technique  
Stejskal EO, Tanner JE. 1965  
Le Bihan et al. 1986  
Moseley et al. 1990
- Many choices in acquisition
- Complicated preprocessing/ analysis steps and many ways and tools to do so
- Functional MRI much more used in the MRI Center – INT @ CERIMED:  
tools in place: SPM, FSL, fmriprep, nilearn
- Acquisition shorter for DWI experiments than fMRI in general
- 3T Siemens Prisma scanner (and soon Cima.X) is particularly performant for DWI with strong gradients and limited Tx bias and susceptibility distortion.
- Can we get more studies using DWI at the MRI Center / Can we analyse now studies that already acquired DWI ?

# QSIPREP: Standardized Preprocessing and reconstruction tool for DWI

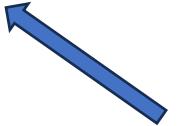
- **Documentation:** <https://qsiprep.readthedocs.io/en/latest/>
- **BIDSApp:** understands your dataset
- Integrates **cutting edge tools** from different software: MRTrix, FSL, Dipy, pyAFQ, DSI Studio, AMICO, ...
- Very **versatile**: able to deal automatically with different sampling schemes: DTI, HARDI, DSI, single shell, multi-shells, multi-runs...
- Very **modular**: possibility to remove some steps, or use very advanced step (`eddy_cuda` available with almost all its options)
- **Well maintained**, good support (github and neurostars), frequent releases
- **Easy installation**: Docker and Singularity images

# QSIPREP: Standardized Preprocessing and reconstruction tool for DWI

- Usage on the mesocentre (SLURM):

```
singularity run --cleanenv -B /scratch/jsein/BIDS:/work \
--nv /scratch/jsein/my_images/qsiprep-0.18.1.sif /work/$study \
/work/$study/derivatives participant --participant_label $sub \
-w /work/temp_data_${study}_test --output-resolution 1.8 --fs-license-file /work/freesurfer/license.txt \
--eddy-config /work/$study/code/qsiprep/eddy_params.json \
--b0-threshold 50 --unringing-method mrdegibbs --denoise-method dwidenoise \
--anat-modality T1w --distortion-group-merge none --recon-spec
/work/qsirecon_custom/mrtrix_multishell_msmt_pyafq_tractometry_JS.json \
--freesurfer-input /work/$study/derivatives/fmriprep/sourcedata/freesurfer
```

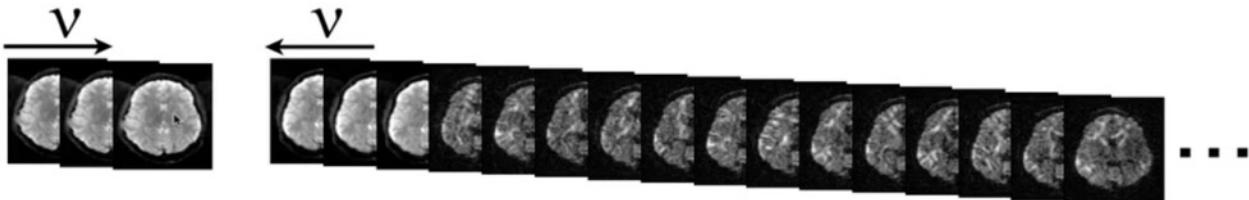
- 8h execution



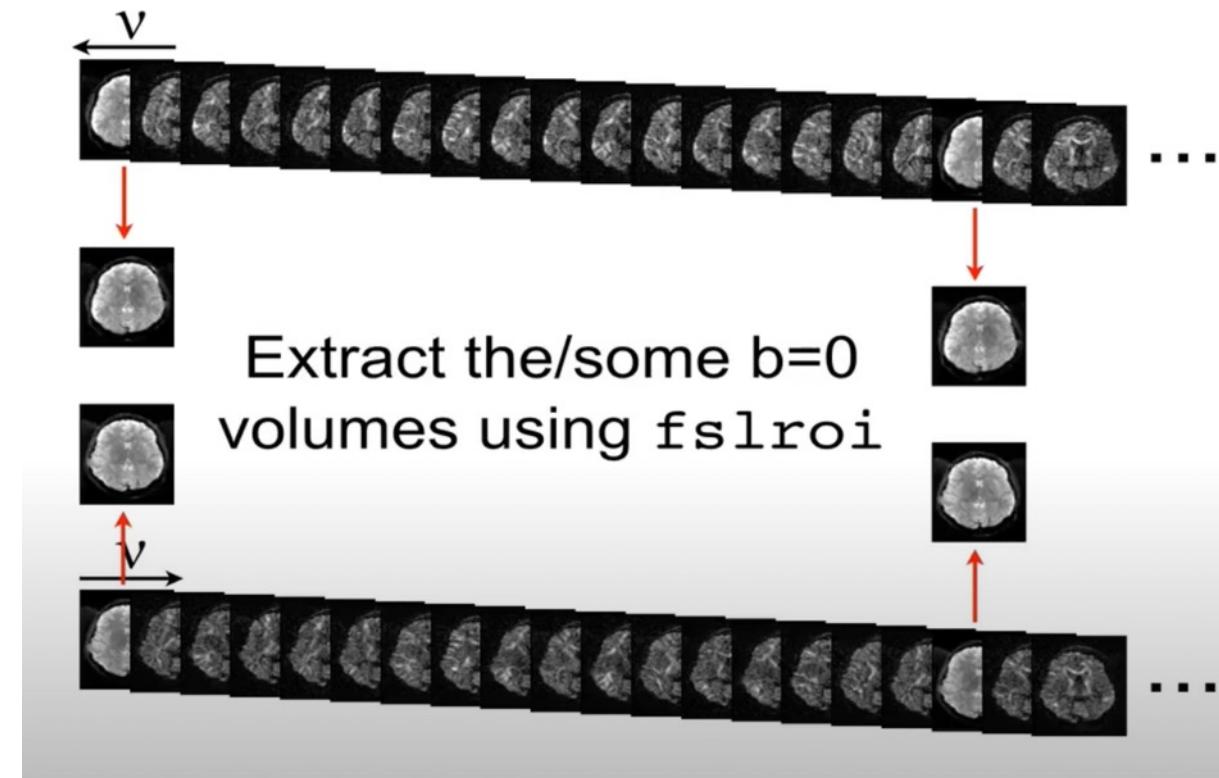
FreeSurfer run by FMRIprep

# Organization of BIDS dataset for DWI (BIDS valid)

- Several acquisition cases:



**Few b0 volumes in one direction and all diffusion volumes in the opposite direction (UK BioBank- style)**



**All the diffusion volumes (along with extra b0s) in both phase encoding direction (HCP-style)**

# Organization of BIDS dataset for DWI (BIDS validation necessary)

- Several acquisition schemes, all automatically understood by QSIPREP

DWI in one direction + external fieldmap

```
sub-test
  anat
    sub-test_T1w.json
    sub-test_T1w.nii.gz
  dwi
    sub-test_dir-AP_dwi.bval
    sub-test_dir-AP_dwi.bvec
    sub-test_dir-AP_dwi.json
    sub-test_dir-AP_dwi.nii.gz
    sub-test_dir-PA_dwi.bval
    sub-test_dir-PA_dwi.bvec
    sub-test_dir-PA_dwi.json
    sub-test_dir-PA_dwi.nii.gz
```

or

```
sub-test
  anat
    sub-test_T1w.json
    sub-test_T1w.nii.gz
  dwi
    sub-test_dir-AP_dwi.bval
    sub-test_dir-AP_dwi.bvec
    sub-test_dir-AP_dwi.json
    sub-test_dir-AP_dwi.nii.gz
  fmap
    sub-test_dir-PA_epi.json
    sub-test_dir-PA_epi.nii.gz
```



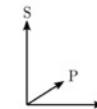
Full PEPOLAR: DWI repeated in AP and PA

+ other more “exotic schemes: DSI, CS-DSI, ...

## Preprocessing

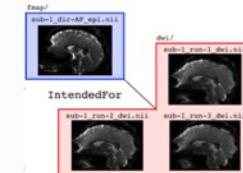
# QSIPREP: Preprocessing

- Each preprocessing step is modular and scalable
- BIDS organization is essential
- Possibility to use and configure `eddy_cuda` (with GPU) with specific parameters defined in a JSON configuration file.
- Output in T1w-ACPC space
- QC measures provided at the end of the preprocessing



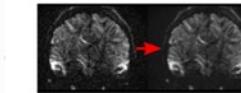
### Conform Image and Gradient Orientation

Ensure that gradient directions and image axes are consistent and will be handled correctly.



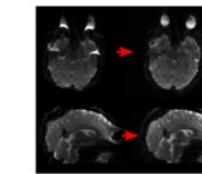
### Group by Distortion

Find images with similar distortion, based on their fieldmap and phase encoding direction. Group them for motion correction and susceptibility distortion correction.



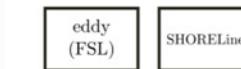
### Denoising

Use MP-PCA or patch2self to non-aggressively denoise diffusion-weighted images.



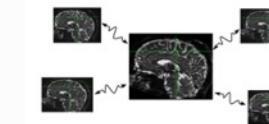
### Distortion Correction

Most types of fieldmaps are supported, including Gradient Recall Echo, TOPUP-style references, real fieldmaps, and fieldmapless (SyN).



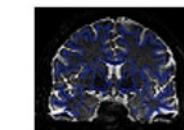
### Head Motion Correction

FSL's `eddy` and our novel `SHORELine` method can motion correct any sampling scheme (DSI, DTI, Multi-SHELL, etc.).



### Build Subject B0 Template

Multiple B0 reference images can be used to build a single unbiased reference image that all DWI scans will be registered to: great for longitudinal voxelwise analysis.

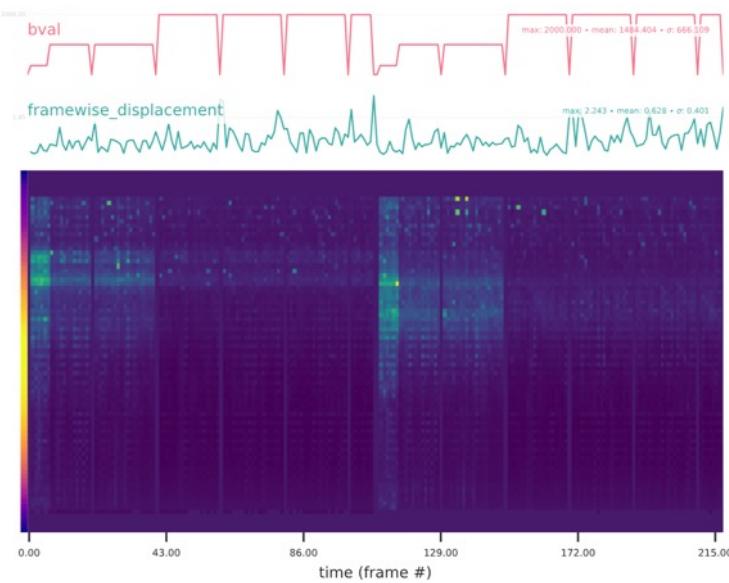


### Registration and Normalization

Register the B0 reference image to the skull-stripped T1w image. ANTs registration to MNI used for derivatives.

# QSIPREP: QC

- HTML report (similar to FMRIprep report)
- dwiqc.json file to be read by dmiprep-viewer  
<http://www.nipreps.org/dmriprep-viewer/#/>
- Possibility to run EddyQC by saving temporary nipype files



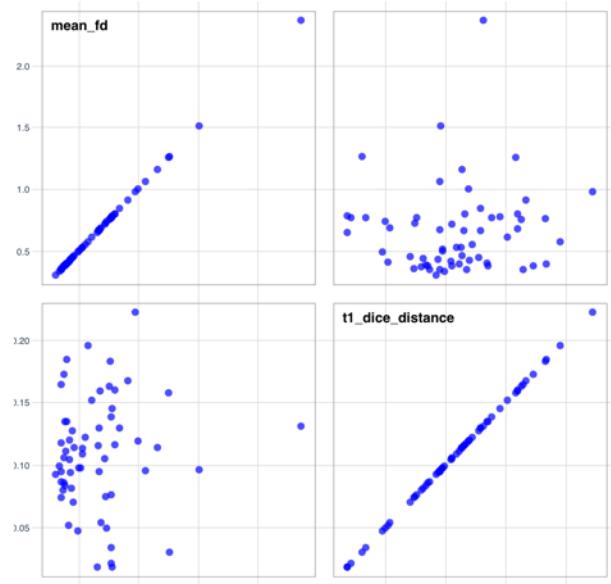
Subject eddy_corrected.nii.gz	
Volume-to-volume motion	
Average abs. motion (mm)	0.98
Average rel. motion (mm)	0.49
Average x translation (mm)	0.11
Average y translation (mm)	-0.71
Average z translation (mm)	-0.33
Average x rotation (deg)	0.22
Average y rotation (deg)	-0.12
Average z rotation (deg)	0.15
Within-volume motion	
Avg std x translation (mm)	0.03
Avg std y translation (mm)	0.28
Avg std z translation (mm)	0.05
Avg std x rotation (deg)	0.08
Avg std y rotation (deg)	0.04
Avg std z rotation (deg)	0.13
Outliers	
Total outliers (%)	0.10
Outliers ( $b=300 \text{ s/mm}^2$ )	0.72
Outliers ( $b=1000 \text{ s/mm}^2$ )	0.12
Outliers ( $b=2000 \text{ s/mm}^2$ )	0.04
Outliers ( $\text{PE dir}=[0, 1, 0]$ )	0.09
Outliers ( $\text{PE dir}=[0, -1, 0]$ )	0.10
SNR/CNR	
Average SNR ( $b=0 \text{ s/mm}^2$ )	26.55
Average CNR ( $b=300 \text{ s/mm}^2$ )	0.92
Average CNR ( $b=1000 \text{ s/mm}^2$ )	1.23
Average CNR ( $b=2000 \text{ s/mm}^2$ )	1.65
Eddy currents	
Std Dev EC linear term (x)	0.05
Std Dev EC linear term (y)	0.08
Std Dev EC linear term (z)	0.11
Susceptibility distortions	
Std Dev voxel displacement	1.25

# dmriprep-viewer : group QC

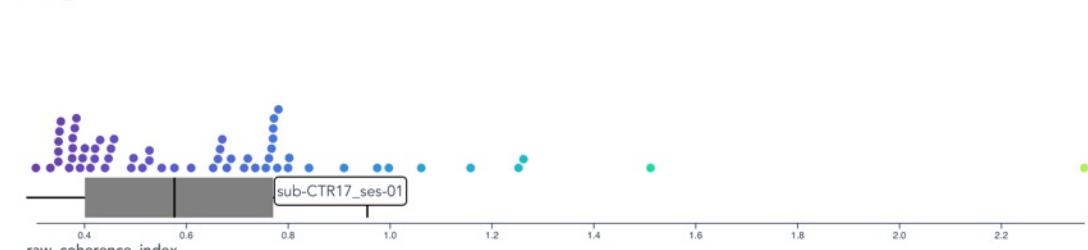
- dmiprep-viewer <http://www.nipreps.org/dmriprep-viewer/#/>

Select scatterplot metrics

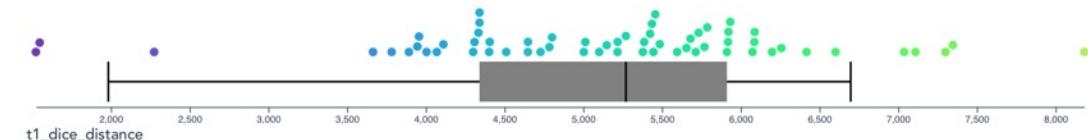
acq\_id  dir\_id  max\_fd  max\_rel\_rotation  max\_rel\_translation  max\_rotation  max\_translation  mean\_fd  
 raw\_coherence\_index  raw\_dimension\_x  raw\_dimension\_y  raw\_dimension\_z  raw\_incoherence\_index  raw\_max\_b  
 raw\_neighbor\_corr  raw\_num\_bad\_slices  raw\_num\_directions  raw voxel\_size\_x  raw voxel\_size\_y  raw voxel\_size\_z  rec\_id  
 run\_id  space\_id  t1 coherence\_index  t1\_dice\_distance  t1 dimension\_x  t1 dimension\_y  t1 dimension\_z  
 t1 incoherence\_index  t1 max\_b  t1 neighbor\_corr  t1 num\_bad\_slices  t1 num\_directions  t1 voxel\_size\_x  
 t1 voxel\_size\_y  t1 voxel\_size\_z  task\_id



mean\_fd



t1\_dice\_distance

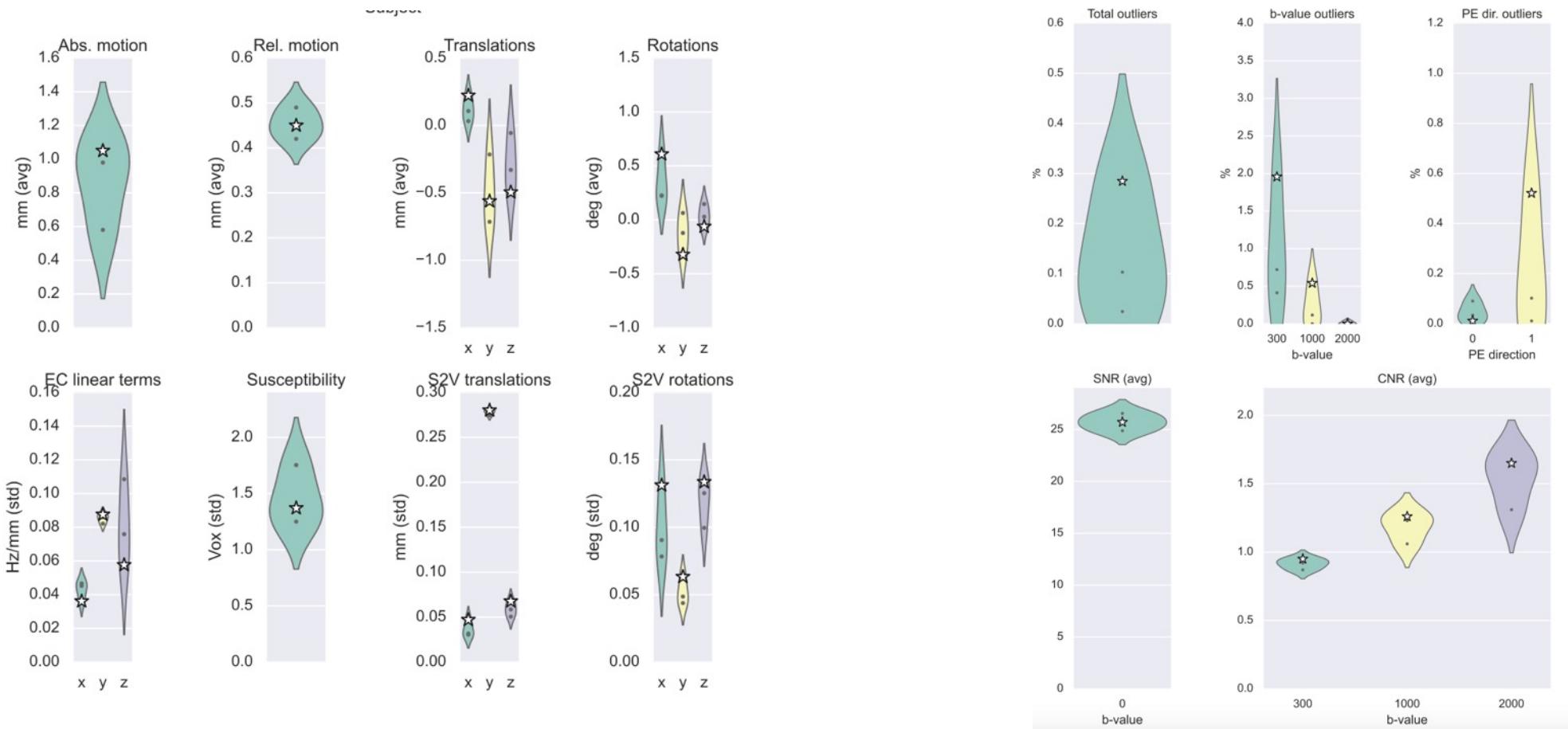


0.02 0.04 0.06 0.08 0.10 0.12 0.14 0.16 0.18 0.20 0.22



# EddyQC : group QC

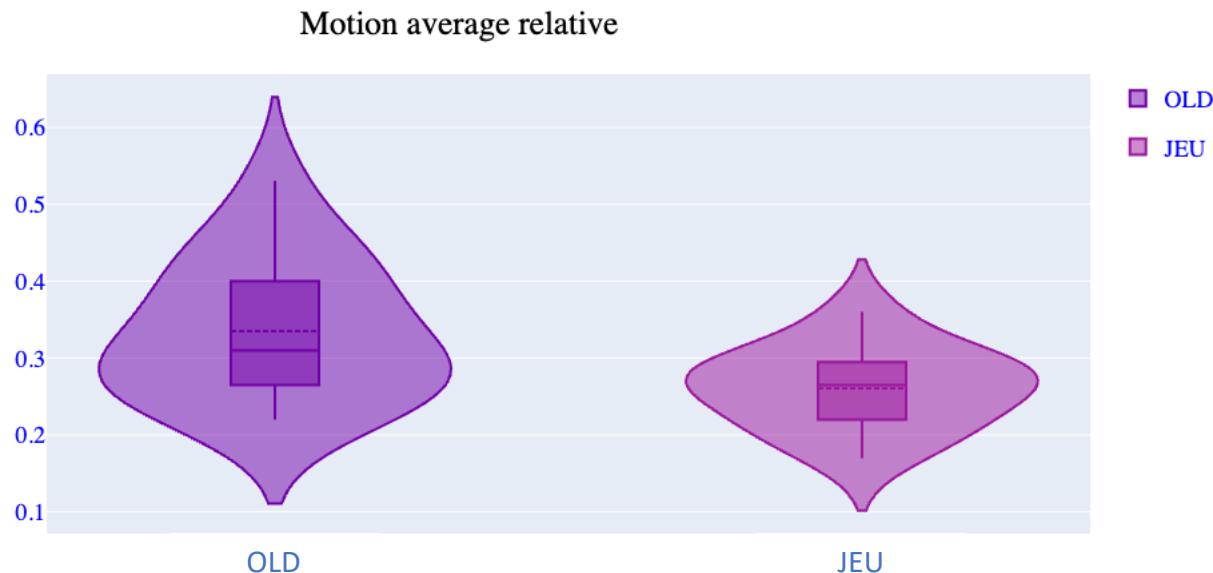
- EddyQC: command suad



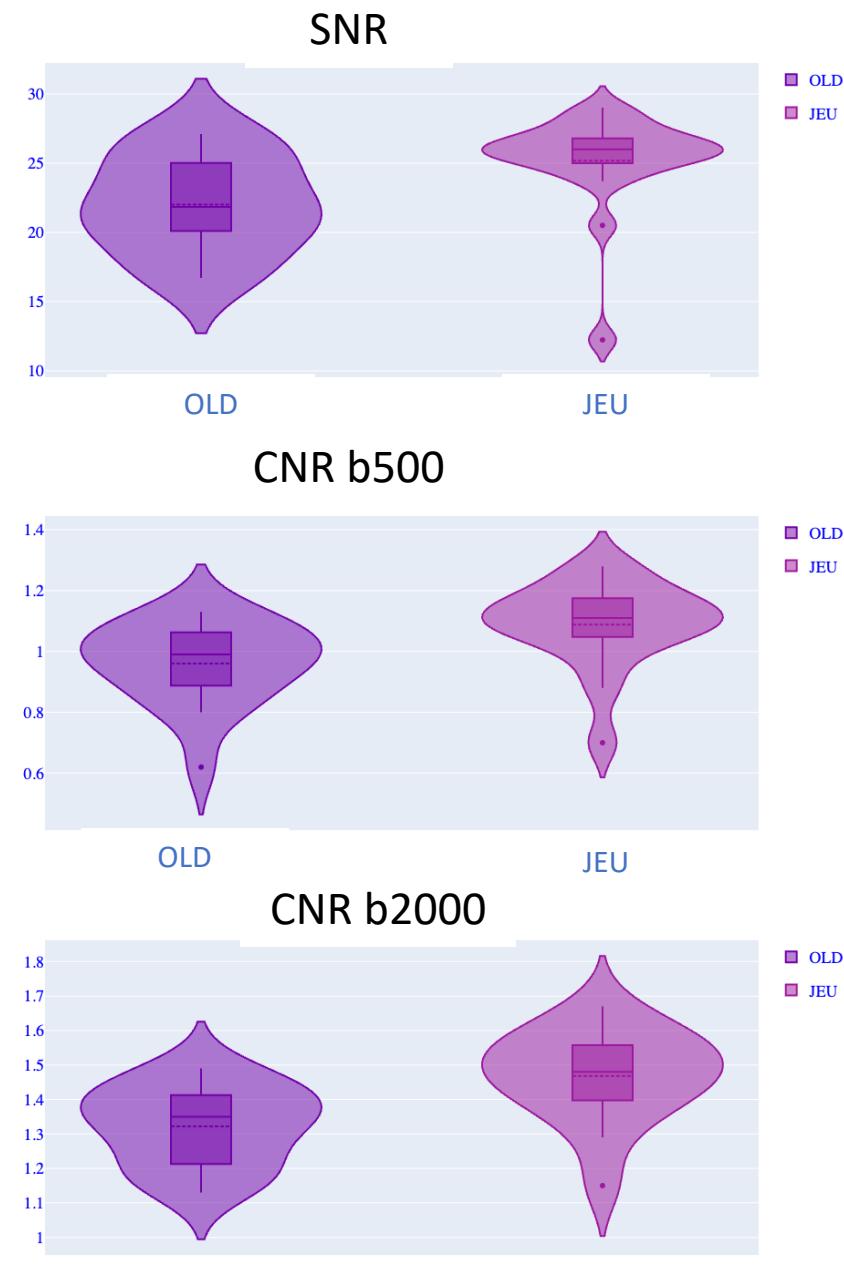
## EddyQC : group QC

Example with Aging study (Kavounoudias et al.) :

- two populations: Young (JEU) and Elderly (OLD)
- 2 b-values: 500 (6 dir) and 2000 (64 dir) s/mm<sup>2</sup>



**Open question:** impact on Tractography and Tractometry?



# DWI Reconstruction with qsiprep

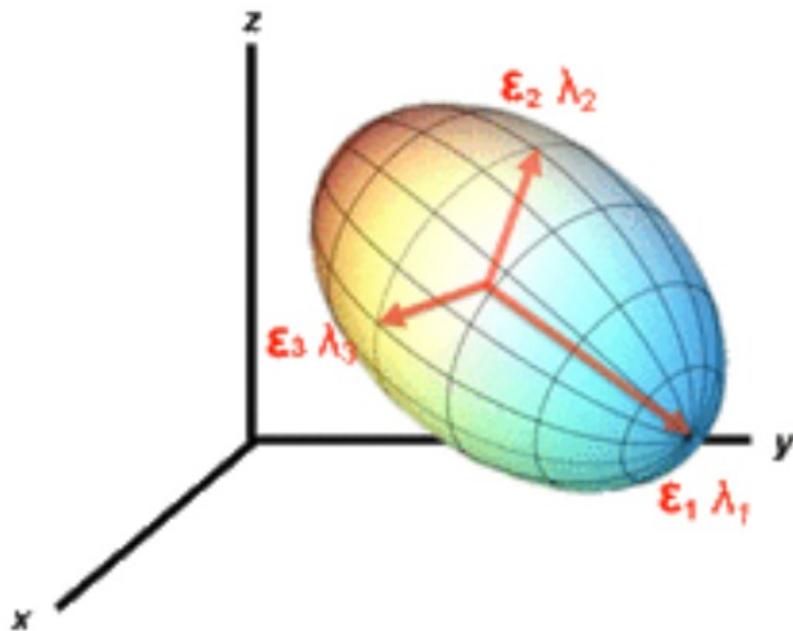
Option	MultiShell	DSI	DTI	Tractography
mrtrix_multishell_msmt_ACT-fast*	Yes	No	No	Probabilistic
mrtrix_multishell_msmt_ACT-hsvs	Yes	No	No	Probabilistic
mrtrix_multishell_msmt_noACT	Yes	No	No	Probabilistic
mrtrix_singleshell_ss3t_noACT	No	No	Yes	Probabilistic
mrtrix_singleshell_ss3t_ACT-hsvs	No	No	Yes	Probabilistic
mrtrix_multishell_msmt_ACT-fast*	No	No	Yes	Probabilistic
pyafq_tractometry	Yes	No	Yes	Both
mrtrix_multishell_msmt_pyafq_tractometry	Yes	No	Yes	Both
amico_noddi	Yes	No	No	None
dsi_studio_gqi	Yes	Yes	Yes*	Deterministic
dipy_mapmri	Yes	Yes	No	Both
dipy_3dshore	Yes	Yes	No	Both
csdsi_3dshore	Yes	Yes	No	Both
reorient_fslstd	Yes	Yes	Yes	None

+ custom  
reconstruction  
through  
configuration files ! 

# DTI metrics

DTI scalar metrics:

- FA =  $\sqrt{\frac{(\lambda_1 - \lambda_2)^2 + (\lambda_2 - \lambda_3)^2 + (\lambda_1 - \lambda_3)^2}{2(\lambda_1^2 + \lambda_2^2 + \lambda_3^2)}}$
- MD =  $(\lambda_1 + \lambda_2 + \lambda_3)/3$
- RD =  $(\lambda_2 + \lambda_3)/2$
- AxD =  $\lambda_1$



Representation of the diffusion as an ellipsoid with three unit eigenvectors, ( $\epsilon_1$ ,  $\epsilon_2$ , and  $\epsilon_3$ ), with corresponding lengths ( $\lambda_1$ ,  $\lambda_2$ , and  $\lambda_3$ ), the eigenvalues.

# DTI metrics and usage

## DTI scalar metrics interpretation

- Increased FA: higher fiber density, lower membrane permeability, greater myelination, ...
- MD : more robust than FA, RD and AxD to crossing fibers



**Figley et al. Potential Pitfalls of Using Fractional Anisotropy, Axial Diffusivity, and Radial Diffusivity as Biomarkers of Cerebral White Matter Microstructure. (2022)**

Recommends complimented measures:

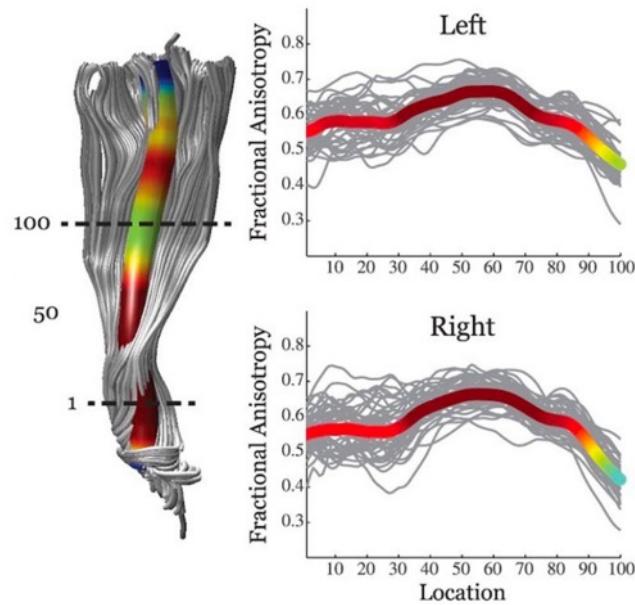
- **DKI, CSD, NODDI, Fixel Based Analysis**
- other than DWI: **T1w/T2w, ihMT, MWF**

# DWI Reconstruction : Tractometry

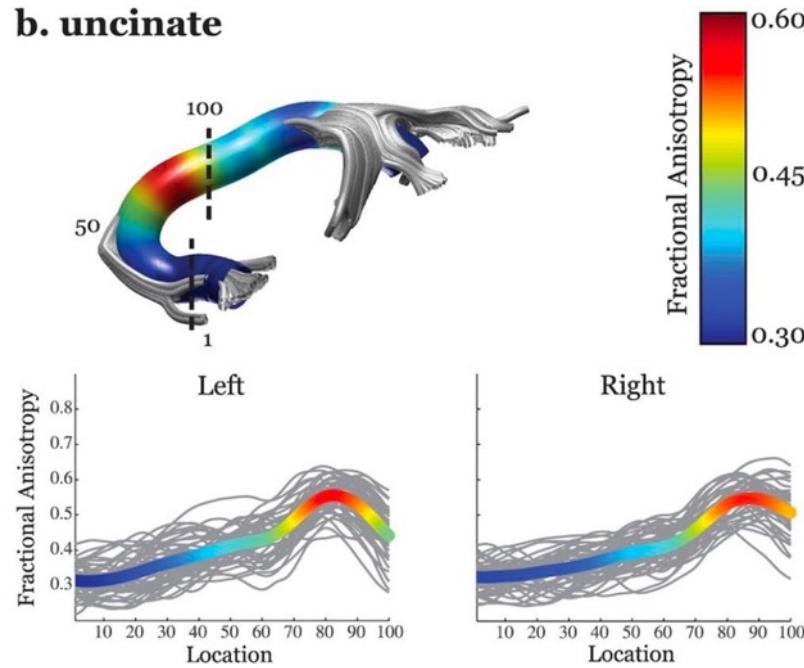
“Taken together, **computational tractography**, **bundle recognition**, and **diffusion modeling** provide so-called **tract profiles**: estimates of microstructural properties of tissue along the length of major pathways. This is the basis of **tractometry**: statistical analysis that compares different groups or assesses individual variability in brain connection structure”

Kruper et al. *Aperture Neuro* 2021

a. corticospinal

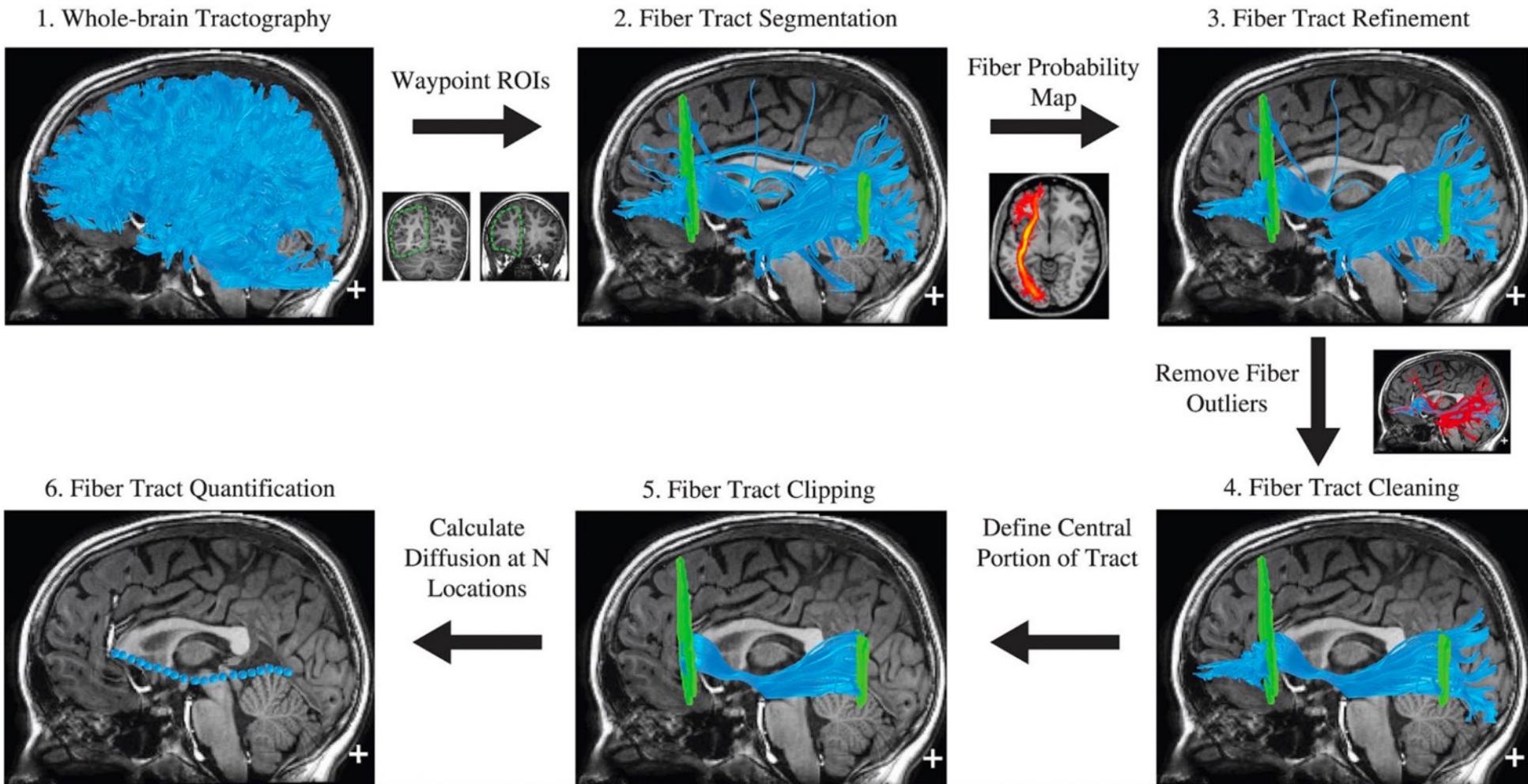


b. uncinate



Andica et al.  
Automated Three-Dimensional Major White Matter Bundle Segmentation Using Diffusion Magnetic Resonance Imaging.  
*Anat Sci Int* 2023

# pyAFQ Reconstruction within qsiprep



Andica et al. 2021, Yeatman et al. 2012

# pyAFQ Reconstruction within qsiprep

```
"description": "Use pyAFQ to perform the  
Tractometry pipeline, with tractography from  
qsiprep with DKI in addition",  
    "space": "T1w",  
    "name":  
"mrtrix_multishell_msmt_pyafq_tractometry_JS,  
    "atlases": [],  
    "nodes": [  
        {  
            "name": "msmt_csd",  
            "software": "MRTrrix3",  
            "action": "csd",  
            "output_suffix": "msmtcsd",  
            "input": "qsiprep",  
            "parameters": {  
                "mtnormalize": true,  
                "response": {  
                    "algorithm": "dhollander"  
                },  
                "fod": {  
                    "algorithm": "msmt_csd",  
                    "max_sh": [8, 8, 8]  
                }  
            }  
        }  
    ]  
}
```

```
"name": "track_ifod2",  
"software": "MRTrrix3",  
"action": "tractography",  
"output_suffix": "ifod2",  
"input": "msmt_csd",  
"parameters": {  
    "use_5tt": false,  
    "use_sift2": true,  
    "tckgen": {  
        "algorithm": "iFOD2",  
        "select": 1e6,  
        "max_length": 250,  
        "min_length": 30,  
        "power": 0.33  
    },  
    "sift2": {}  
}
```

```
"name": "pyafq_tractometry",  
"software": "pyAFQ",  
"action": "pyafq_tractometry",  
"input": "track_ifod2",  
"output_suffix": "PYAFQ_TRACTOMETRY_ET",  
"parameters": {  
    "use_external_tracking": true,  
    "export": "all",  
    "directions": "prob",  
    "max_angle": 30.0,  
    "sphere": "",  
    "seed_mask": "",  
    "seed_threshold": 0,  
    "n_seeds": 1,  
    "random_seeds": false,  
    "rng_seed": "",  
    "stop_mask": "",  
    "stop_threshold": 0,  
    "step_size": 0.5,  
    "min_length": 50,  
    "max_length": 250,  
    "odf_model": "CSD",  
    "tracker": "local",  
    "nb_points": false,  
    "nb_streamlines": false,  
    "seg_algo": "AFQ",  
    " "
```

# pyAFQ Reconstruction within qsiprep

Specific option in pyAFQ (through QSIPREP):

"scalars":

"['dti\_fa', 'dti\_md', 'dti\_rd', 'dti\_ad', 'dki\_fa', 'dki\_md']",

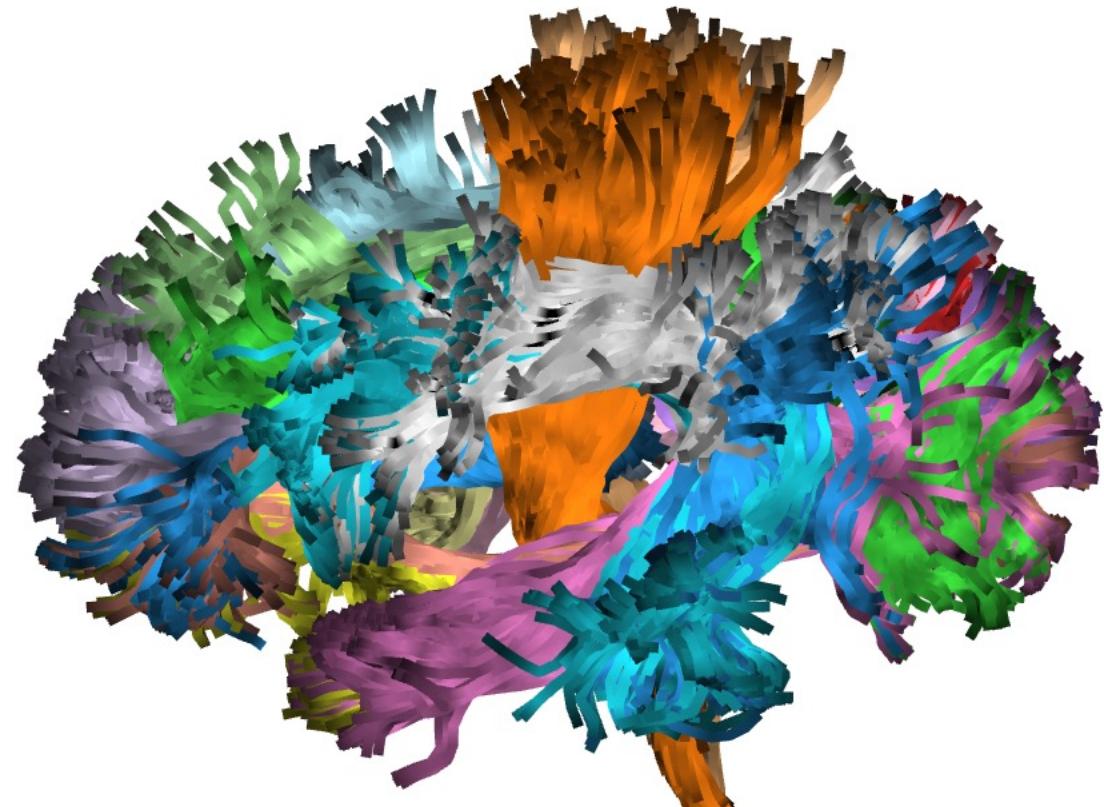


**sub-SUB\_space-T1w\_desc-preproc\_dwi\_space-RASMM\_model-probCSD\_algo-AFQ\_desc-profiles\_dwi.csv**

	tractID	nodeID	dti_fa	dti_md	dti_rd	dti_ad	dki_fa	dki_md
0	ATR_L	0	0.20570933854551643	0.000686916065784494	0.0006151699570349627	0.0008304082832835536	0.23724199868892912	0.0008574057502554258
1	ATR_L	1	0.235723937038021	0.0006726244134407995	0.0005910810428346782	0.000835711154653046	0.2701782451199084	0.0008293609318190809
2	ATR_L	2	0.2692432407500829	0.000666069032143725	0.0005716939706152982	0.000854819155200578	0.3007053672774038	0.0008210591738175878
3	ATR_L	3	0.30230813333267226	0.0006621957152211032	0.0005541436993604519	0.0008782997469424069	0.3280628871756977	0.000819582586150776
4	ATR_L	4	0.33778103072740817	0.0006571018774746051	0.0005342394656032334	0.0009028267012173492	0.35847578610972647	0.000817585508623988
5	ATR_L	5	0.3720022284451804	0.0006507187349331651	0.0005137055316980438	0.0009247451414034052	0.38877468371008156	0.0008151938069753821
6	ATR_L	6	0.4053563495129063	0.0006440213266363467	0.0004931299407467528	0.0009458040984155365	0.41807084685300433	0.0008144467516062641
7	ATR_L	7	0.43423053017046875	0.0006383732725740044	0.0004752454445847734	0.0009646289285524639	0.44181102793176036	0.0008156335089100516
8	ATR_L	8	0.45619211389941133	0.0006341570153181982	0.00046208220224638847	0.000978306641461819	0.458830337790342	0.0008187808832119736
9	ATR_L	9	0.47142632821755165	0.000631969647024736	0.0004538612617678625	0.0009881864175384824	0.47025880763211164	0.0008232013445819059

**Table S1.** Abbreviations of the major white matter pathways recognized by pyAFQ.

ARC L	Left arcuate
ARC R	Right arcuate
ATR L	Left thalamic radiation
ATR R	Right thalamic radiation
CGC L	Left cingulum cingulate
CGC R	Right cingulum cingulate
CST L	Left corticospinal
CST R	Right corticospinal
FA	Callosum forceps minor
FP	Callosum forceps major
IFO L	Left inferior fronto-occipital fasciculus
IFO R	Right inferior fronto-occipital fasciculus
ILF L	Left inferior longitudinal fasciculus
ILF R	Right inferior longitudinal fasciculus
SLF L	Left superior longitudinal fasciculus
SLF R	Right superior longitudinal fasciculus
UNC L	Left uncinate
UNC R	Right uncinate



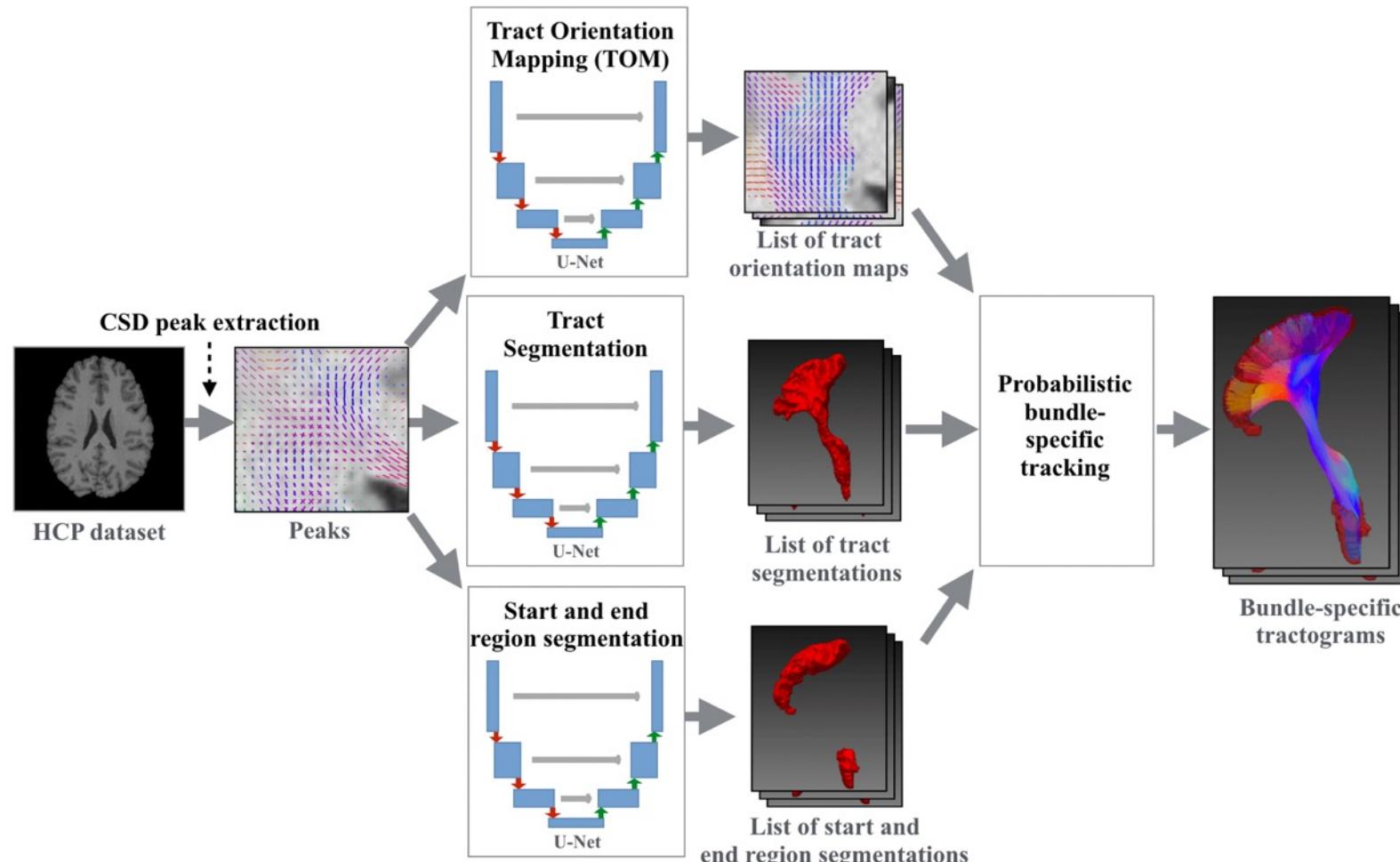
Andica et al. 2021, Yeatman et al. 2012

# pyAFQ Reconstruction within qsiprep

Demo on Web browser and image viewers

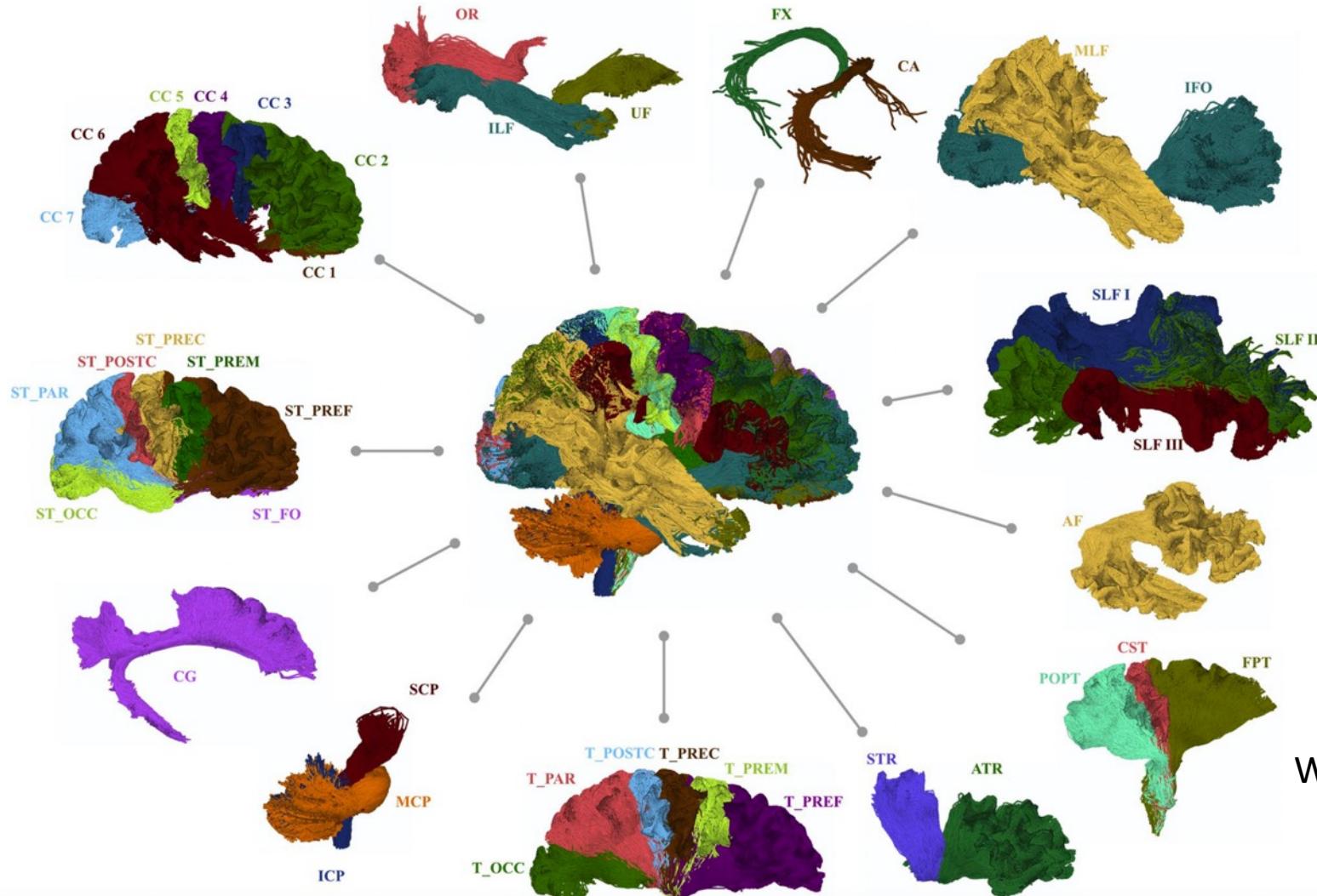
# Another Tractometry tool: TractSeg

- <https://github.com/MIC-DKFZ/TractSeg>
- 72 white matter tracts automatically segmented



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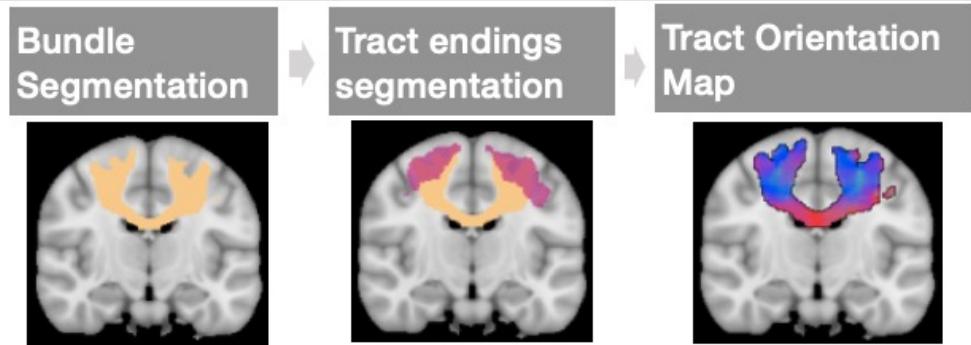
Wasserthal et al. 2018

0: AF_left	(Arcuate fascicle)	31: POPT_left	(Parieto-occipital pontine)	63: ST_PREM_right
1: AF_right		32: POPT_right		64: ST_PREC_left (Striato-precentral)
2: ATR_left	(Anterior Thalamic Radiation)	33: SCP_left	(Superior cerebellar peduncle)	65: ST_PREC_right
3: ATR_right		34: SCP_right		66: ST_POSTC_left (Striato-postcentral)
4: CA	(Commissure Anterior)	35: SLF_I_left	(Superior longitudinal fascicle I)	67: ST_POSTC_right
5: CC_1	(Rostrum)	36: SLF_I_right		68: ST_PAR_left (Striato-parietal)
6: CC_2	(Genu)	37: SLF_II_left	(Superior longitudinal fascicle II)	69: ST_PAR_right
7: CC_3	(Rostral body (Premotor))	38: SLF_II_right		70: ST_OCC_left (Striato-occipital)
8: CC_4	(Anterior midbody (Primary Motor))	39: SLF_III_left	(Superior longitudinal fascicle III)	71: ST_OCC_right
9: CC_5	(Posterior midbody (Primary Somatosensory))	40: SLF_III_right		
10: CC_6	(Isthmus)	41: STR_left	(Superior Thalamic Radiation)	
11: CC_7	(Splenium)	42: STR_right		
12: CG_left	(Cingulum left)	43: UF_left	(Uncinate fascicle)	
13: CG_right		44: UF_right		
14: CST_left	(Corticospinal tract)	45: CC	(Corpus Callosum - all)	
15: CST_right		46: T_PREF_left	(Thalamo-prefrontal)	
16: MLF_left	(Middle longitudinal fascicle)	47: T_PREF_right		
17: MLF_right		48: T_PREM_left	(Thalamo-premotor)	
18: FPT_left	(Fronto-pontine tract)	49: T_PREM_right		
19: FPT_right		50: T_PREC_left	(Thalamo-precentral)	
20: FX_left	(Fornix)	51: T_PREC_right		
21: FX_right		52: T_POSTC_left	(Thalamo-postcentral)	
22: ICP_left	(Inferior cerebellar peduncle)	53: T_POSTC_right		
23: ICP_right		54: T_PAR_left	(Thalamo-parietal)	
24: IFO_left	(Inferior occipito-frontal fascicle)	55: T_PAR_right		
25: IFO_right		56: T_OCC_left	(Thalamo-occipital)	
26: ILF_left	(Inferior longitudinal fascicle)	57: T_OCC_right		
27: ILF_right		58: ST_FO_left	(Striato-fronto-orbital)	
28: MCP	(Middle cerebellar peduncle)	59: ST_FO_right		
29: OR_left	(Optic radiation)	60: ST_PREF_left	(Striato-prefrontal)	
30: OR_right		61: ST_PREF_right		
		62: ST_PREM_left	(Striato-premotor)	

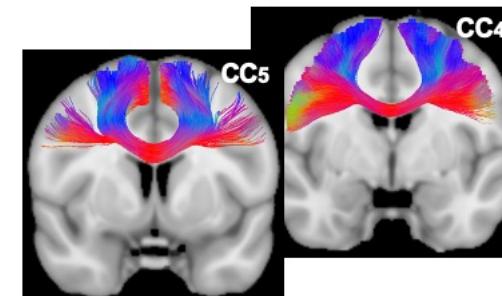
**TractSeg bundles**  
Wasserthal et al. 2018



# Exemple with TractSeg: project Aging (A. Kavounoudias)

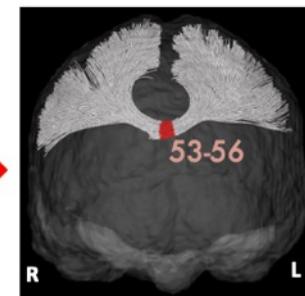
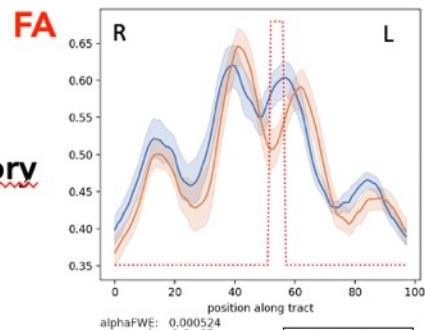


Tractometry

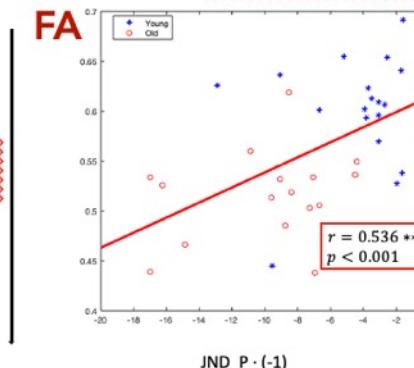


**Tractography:** age-related changes within the interhemispheric motor and somatosensory tract.

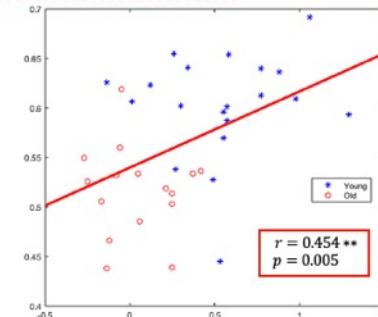
Somatosensory Tract



High  
↔  
Low

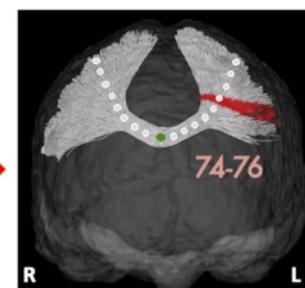
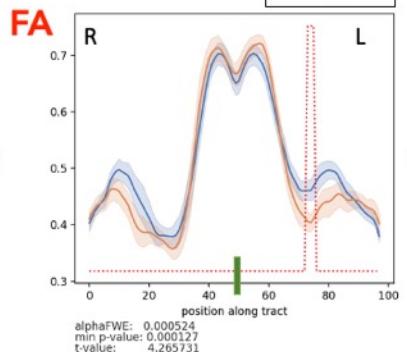


Correlations with functional indices



Pinzon et al.  
OHBM 2022

Motor Tract



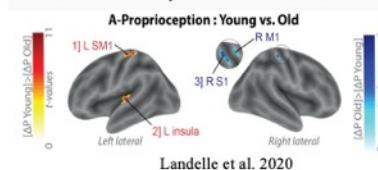
not significant  
↔

Proprioceptive performance (JND)



Landelle et al. 2018

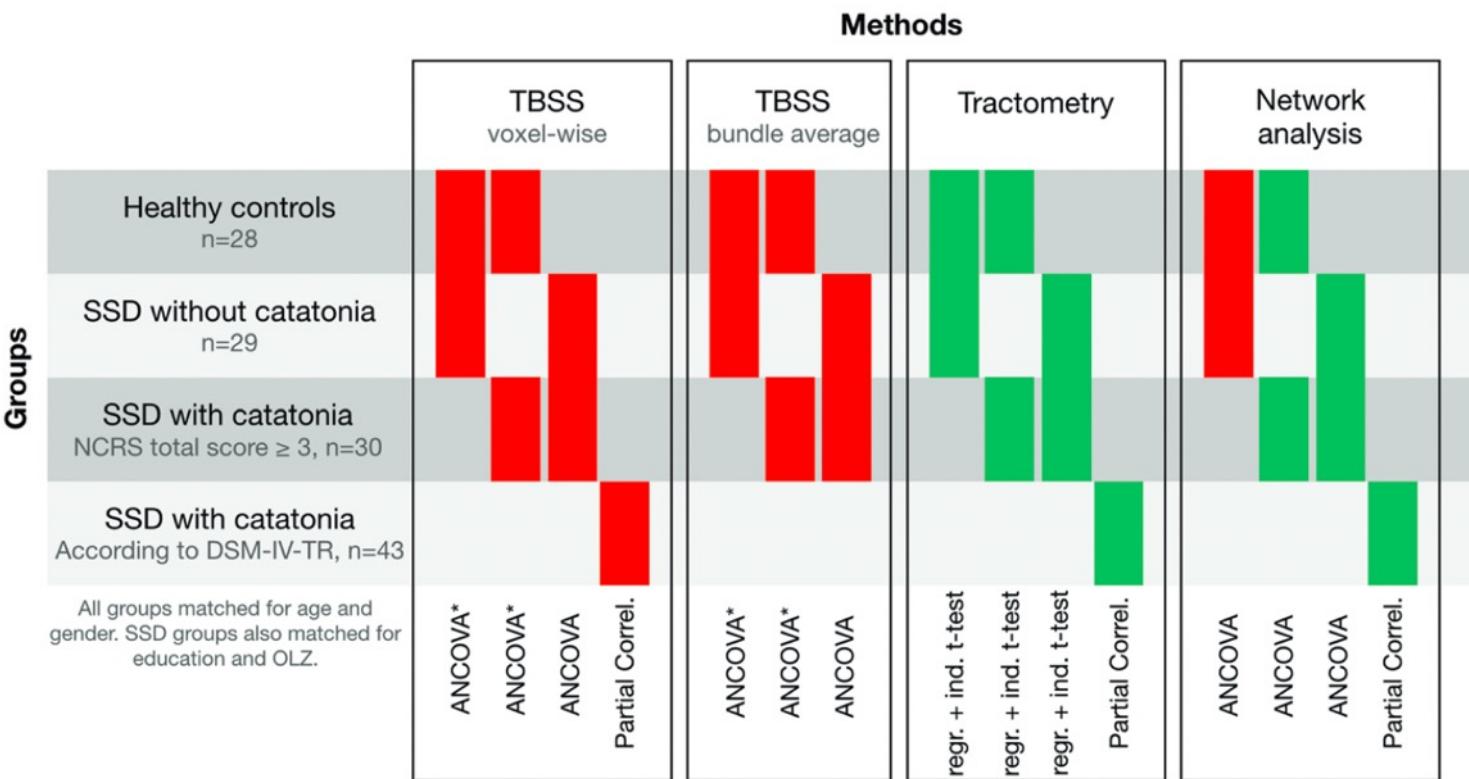
Inter-hemispheric Difference



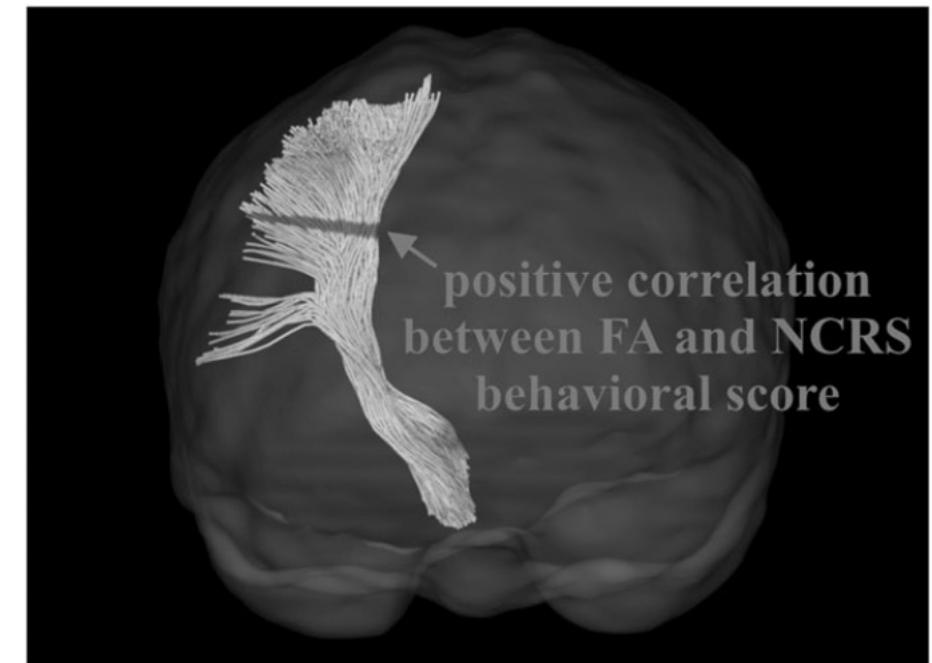
RD in the sensory tract significantly correlated with the functional indices

# Examples of Tractometry studies (TractSeg)

Wasserthal et al. Multiparametric Mapping of White Matter Microstructure in Catatonia. *Neuropsychopharmacol.* 2020



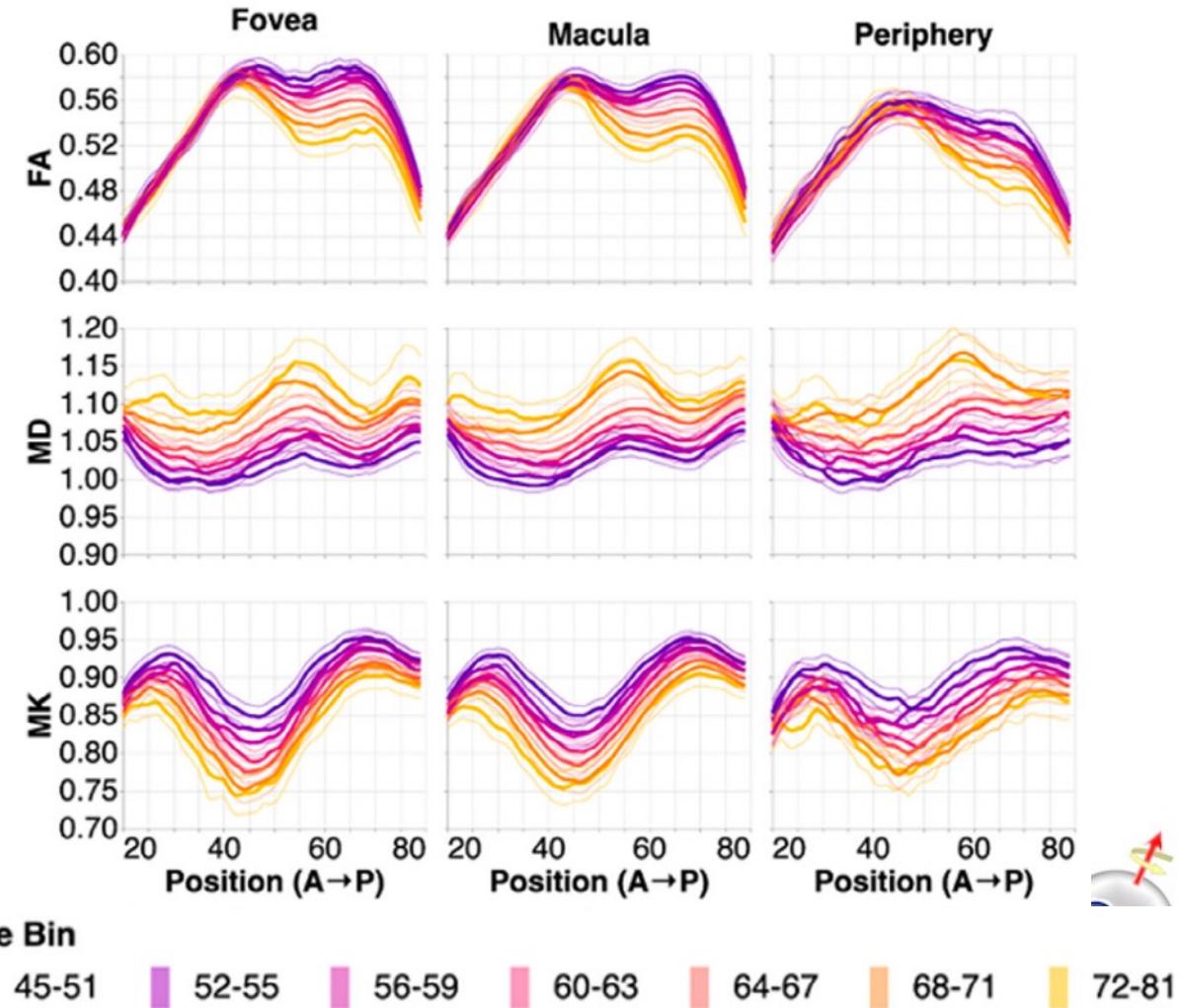
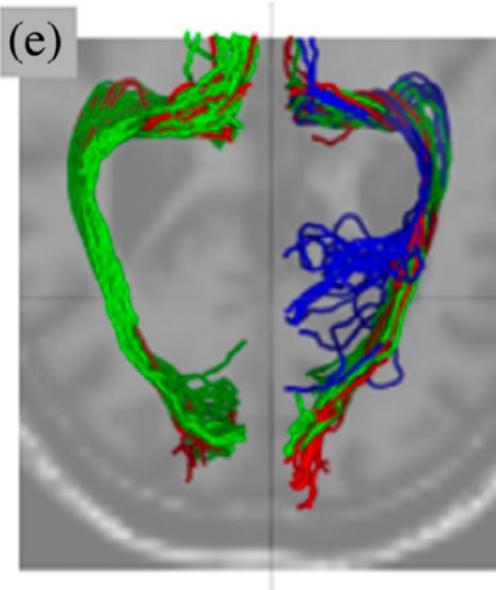
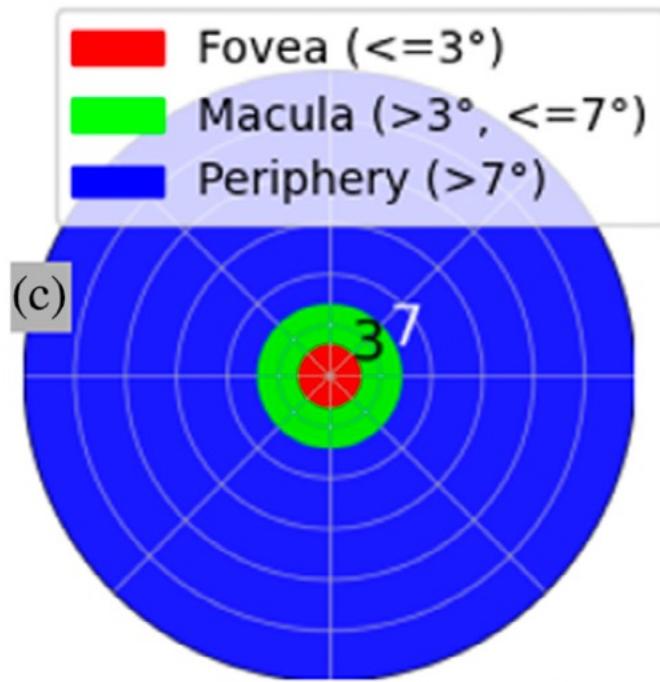
**left CST (Corticospinal tract)**



# Examples of Tractometry studies ( pyAFQ)

Kruper et al. Optic Radiations Representing Different Eccentricities Age Differently. *Human Brain Mapping* 2023

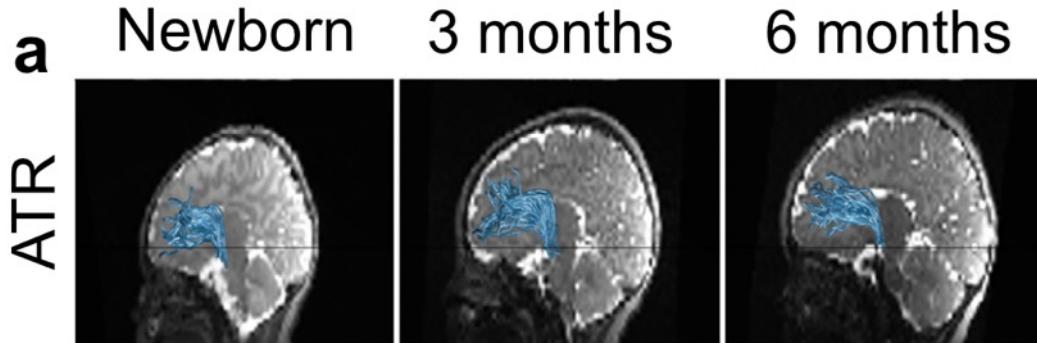
dMRI data from 5382 subjects of UK BioBank , age 45-81



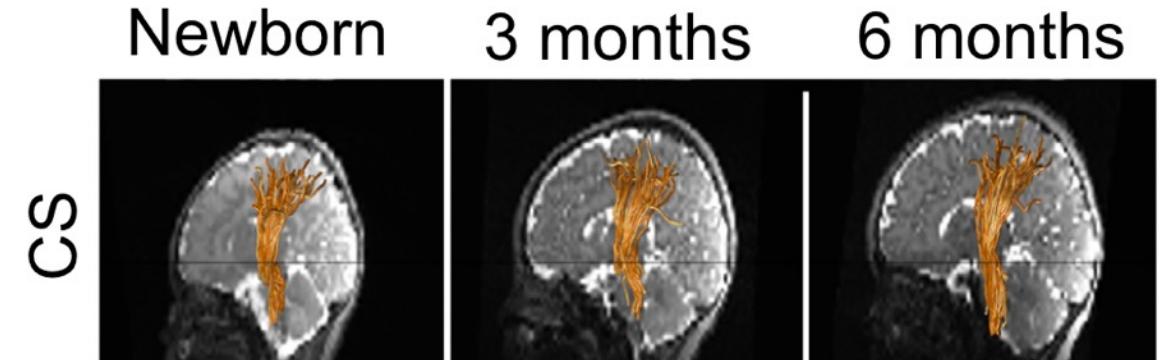
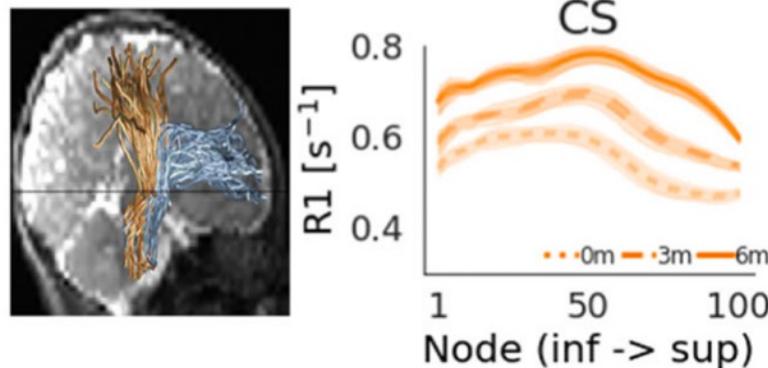
# Examples of Tractometry studies ( Baby AFQ)

Grotheer et al. Matter Myelination during Early Infancy Is Linked to Spatial Gradients and Myelin Content at Birth.

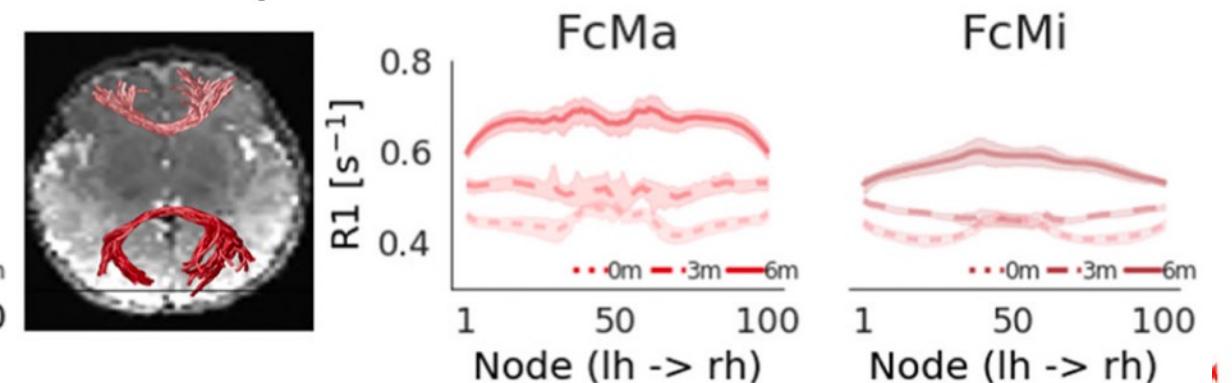
*Nat Commun* 2022



Projection bundles



Inter-hemispheric bundles



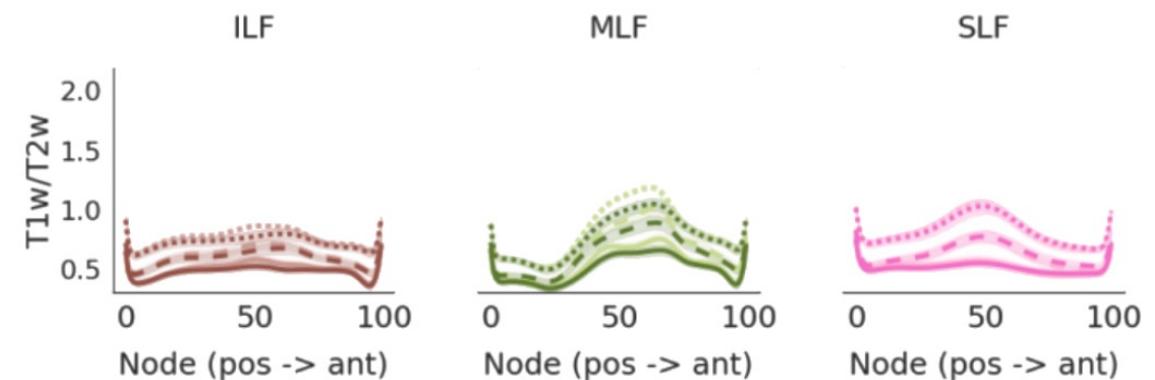
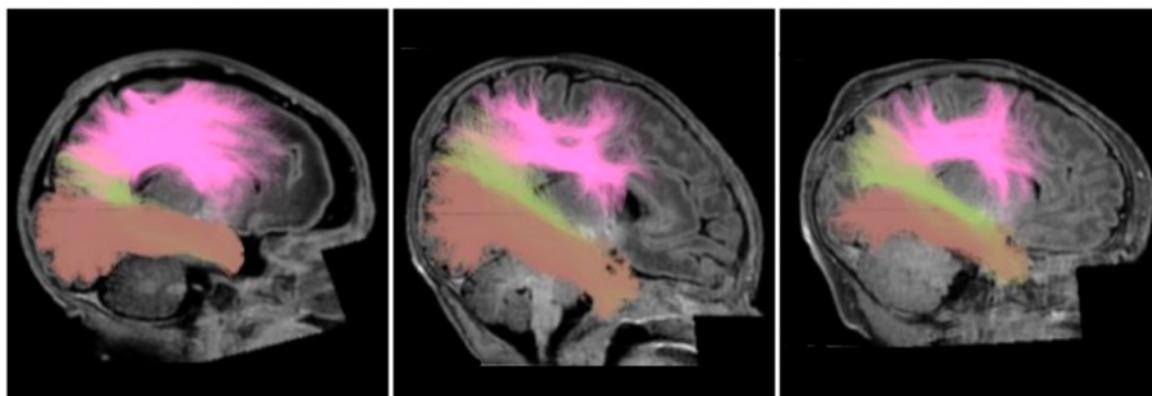
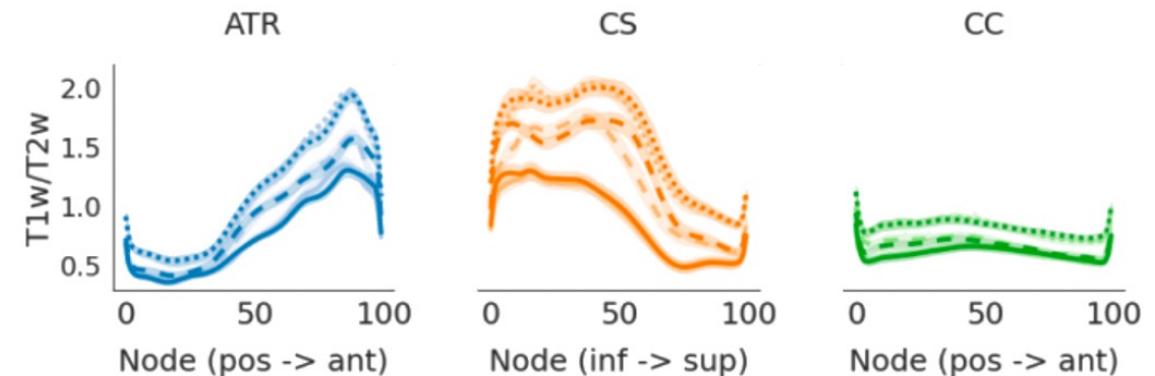
## Examples of Tractometry studies ( Baby AFQ)

- Grotheer et al. . Human White Matter Myelination Rate Slows down at Birth; preprint; Neuroscience, 2023.

34 weeks

38 weeks

40 weeks



— Scan Age: 33-<37 weeks    — Scan Age: 37-<41 weeks    ······ Scan Age: 41-<45 weeks

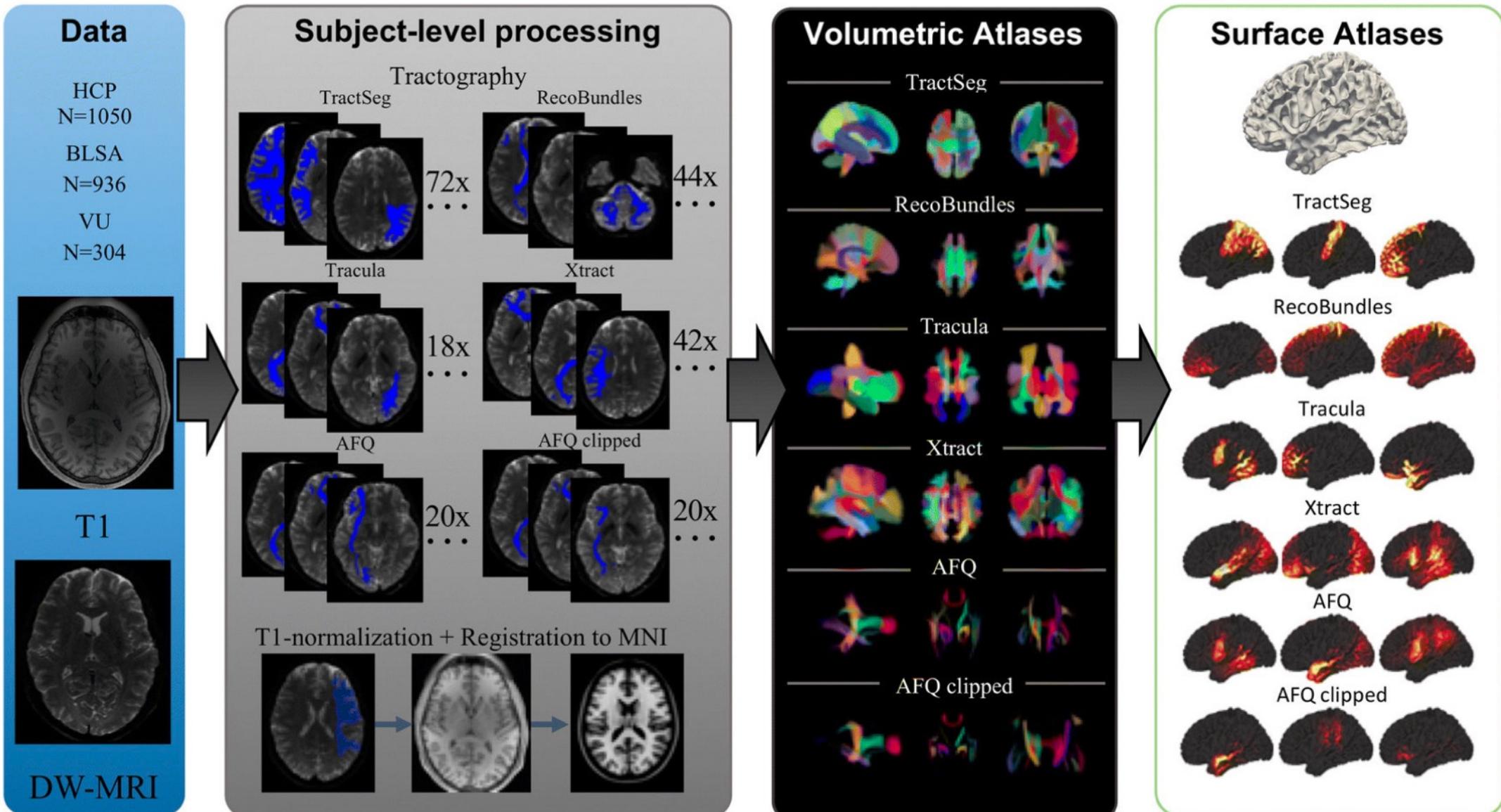


# Tractometry tools

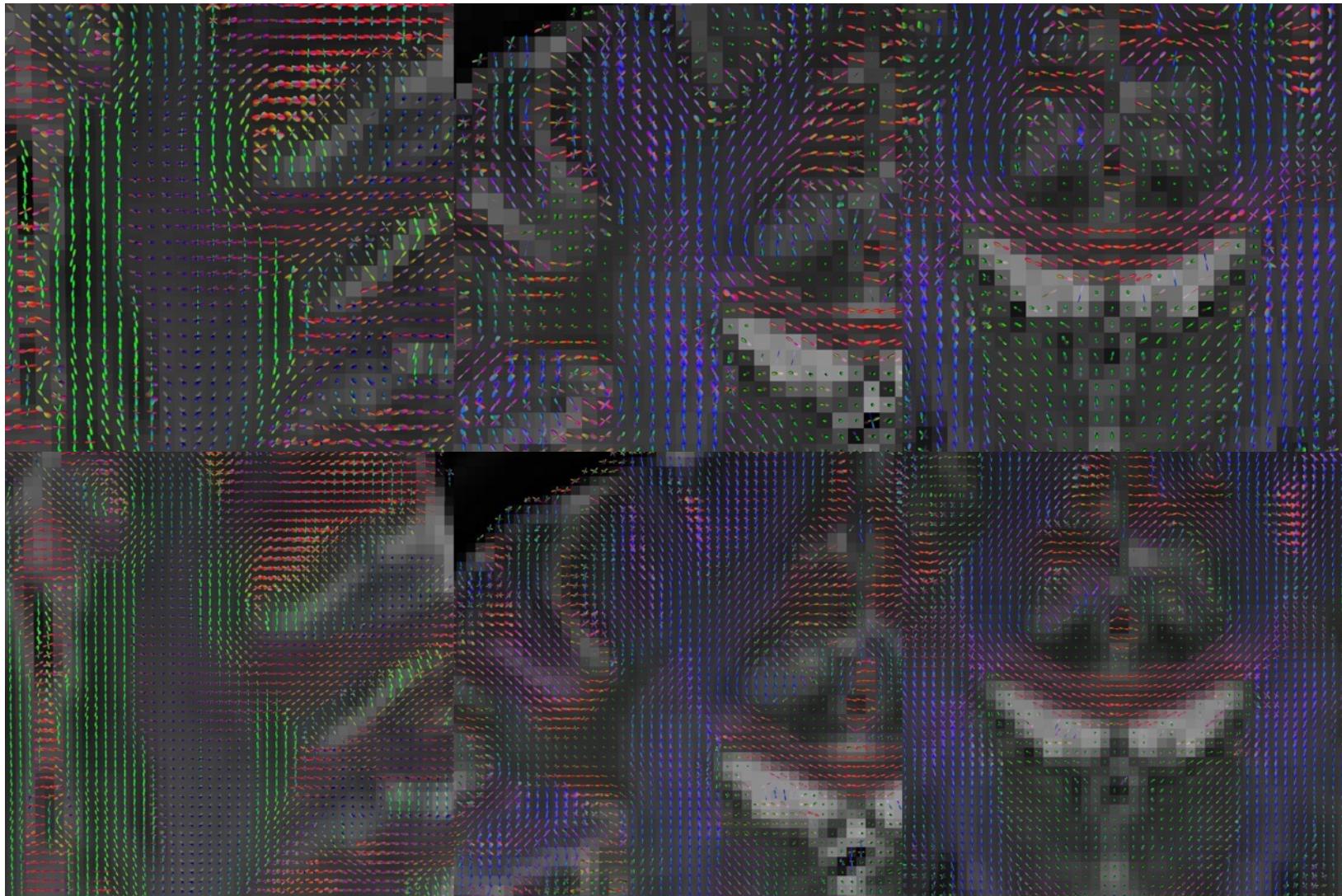
- **AFQ** (matlab) and its successor pyAFQ (python) ( implemented within QSIPREP )  
Yeatman et al. **Tract Profiles of White Matter Properties: Automating Fiber-Tract Quantification.** *PLOS ONE* 2012.  
Kruperet al. **Evaluating the Reliability of Human Brain White Matter Tractometry.** *Aperture Neuro* 2021

- Comparison between **TRACULA**, **AFQ** and **TractSeg** :  
Andica et al. **Automated Three-Dimensional Major White Matter Bundle Segmentation Using Diffusion Magnetic Resonance Imaging.** *Anat Sci Int* 2023, 98 (3), 318–336. <https://doi.org/10.1007/s12565-023-00715-9>
- Atlas based on different Bundle Segmentation Algorithms: **TRACULA**, **AFQ**, **XTRACT**, **RecoBundles**, **TractSeg**  
Hansen et al. **Pandora: 4-D White Matter Bundle Population-Based Atlases Derived from Diffusion MRI Fiber Tractography.** *Neuroinformatics* 2021

# Pandora atlas



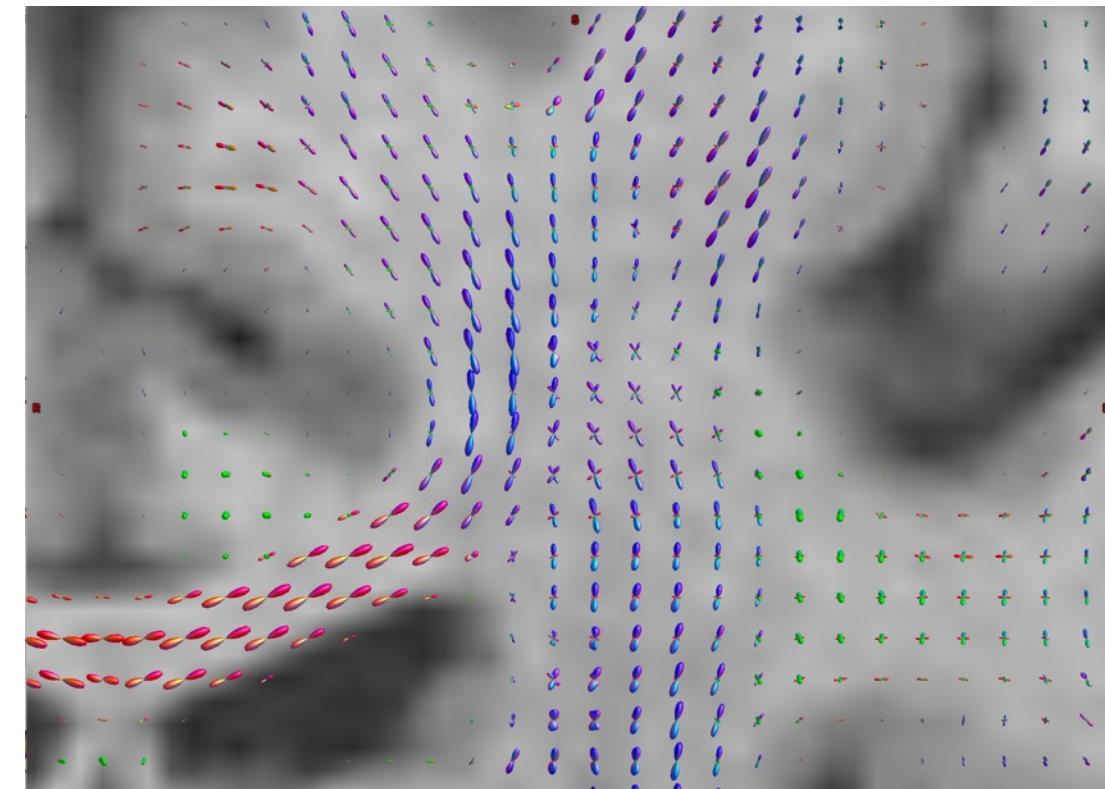
## Methodological use case: Upsampling during preprocessing?



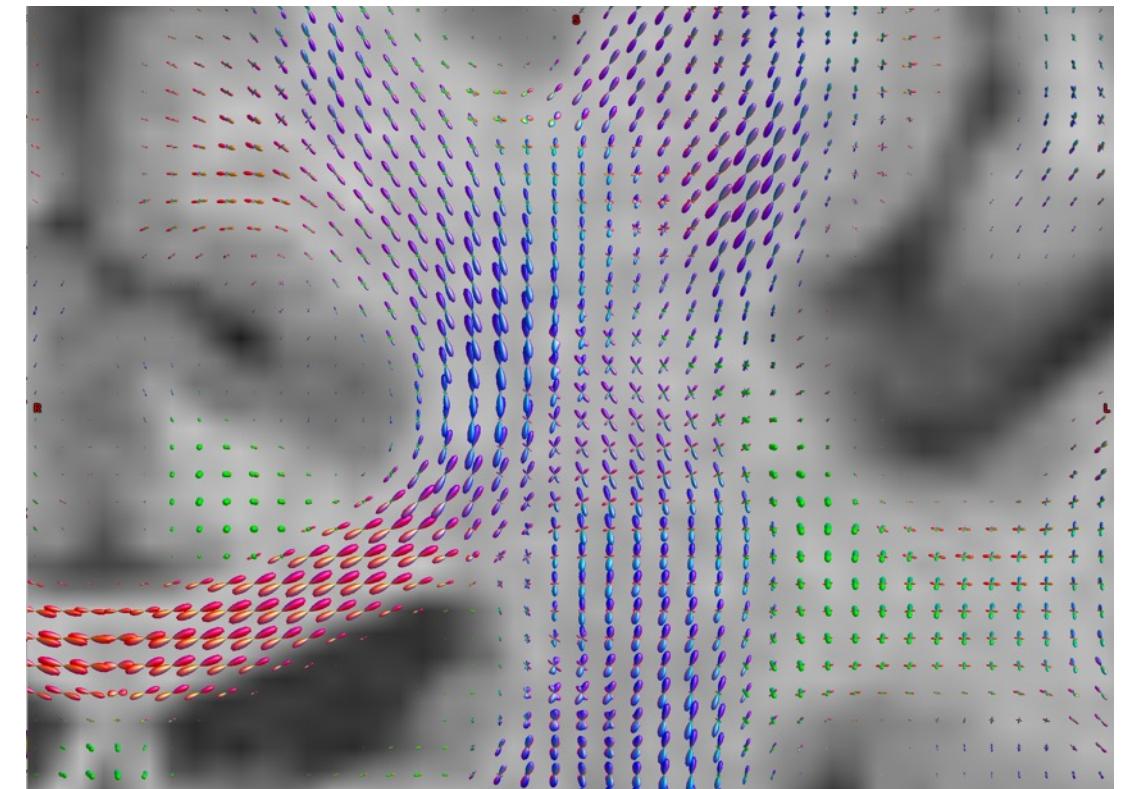
ODFs after  
preprocessing  
at original  
resolution:  
**1.8mm iso**

ODFs after  
preprocessing at  
upsampled  
resolution:  
**1.2mm iso**

## Upsampling during preprocessing?

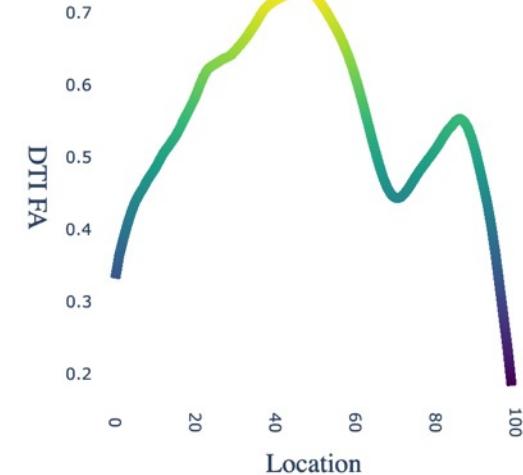
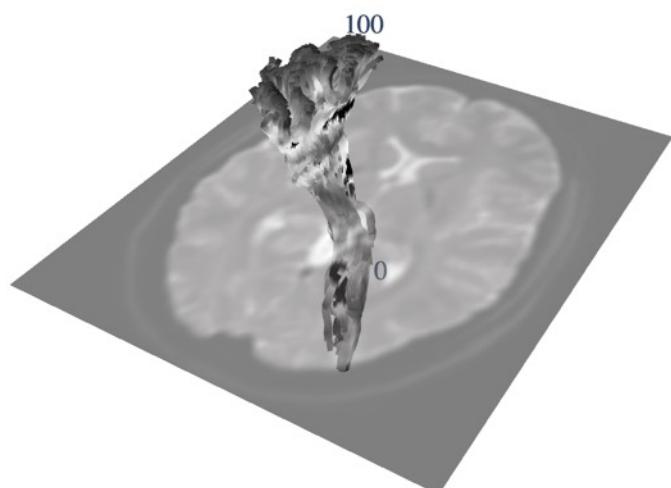
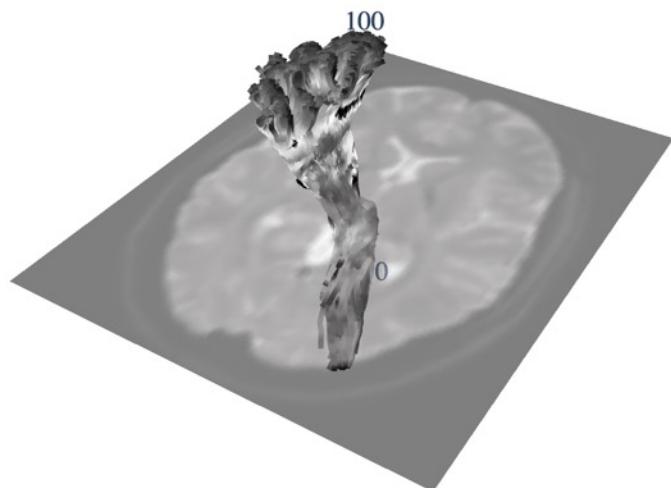


ODFs after preprocessing at  
original resolution: 1.8mm iso

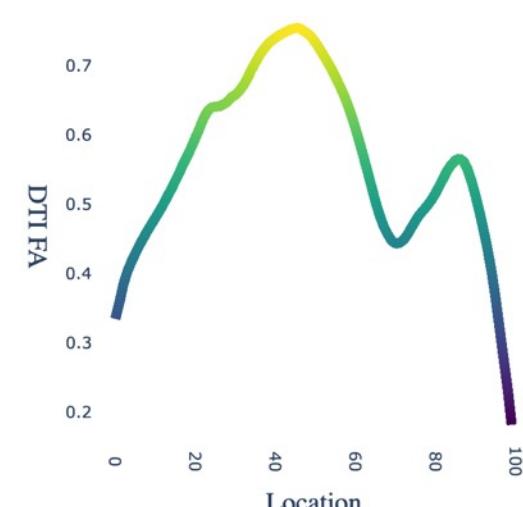


ODFs after preprocessing at  
upsampled resolution: 1.2mm iso

## Upsampling during preprocessing?

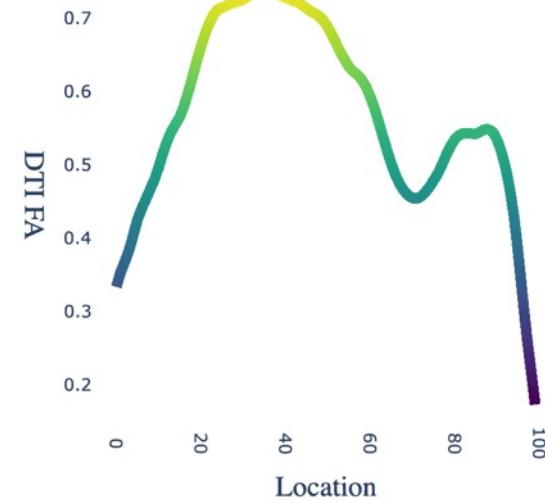
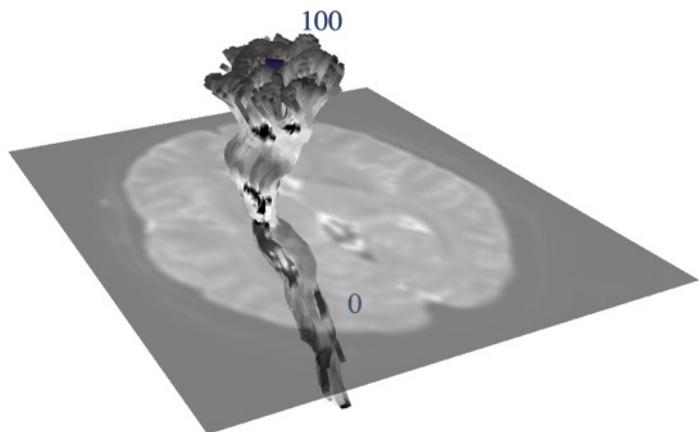


Original  
resolution:  
1.8mm iso

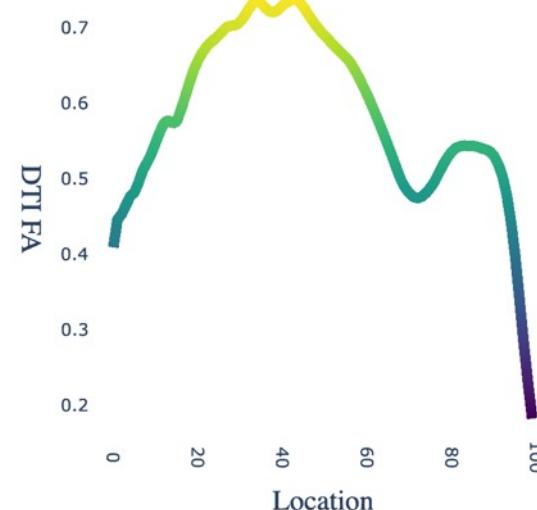
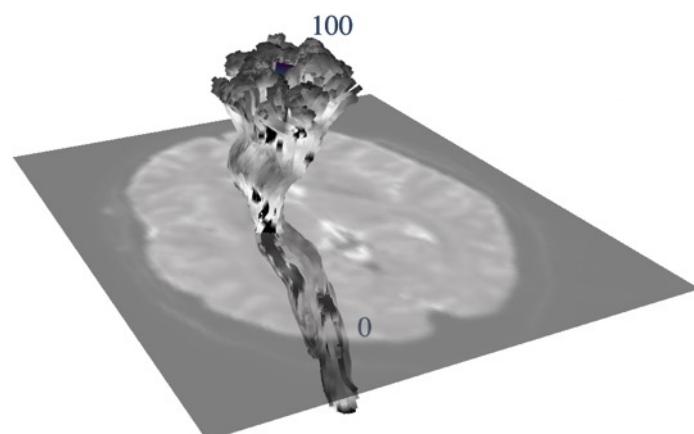


Preprocessed  
data  
upsampled at  
1.2mm iso

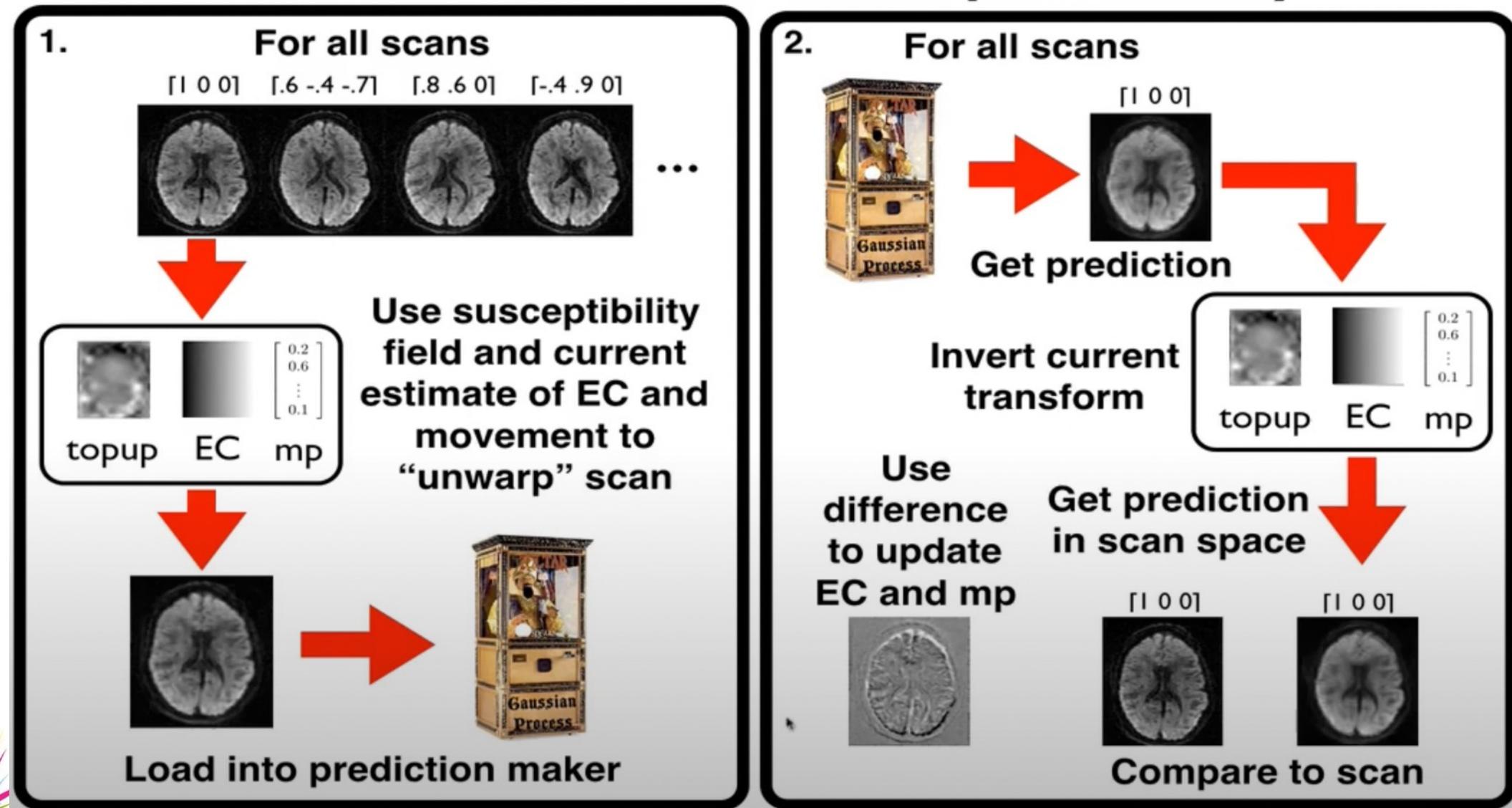
## DWI acquisition with only bvec in AP?

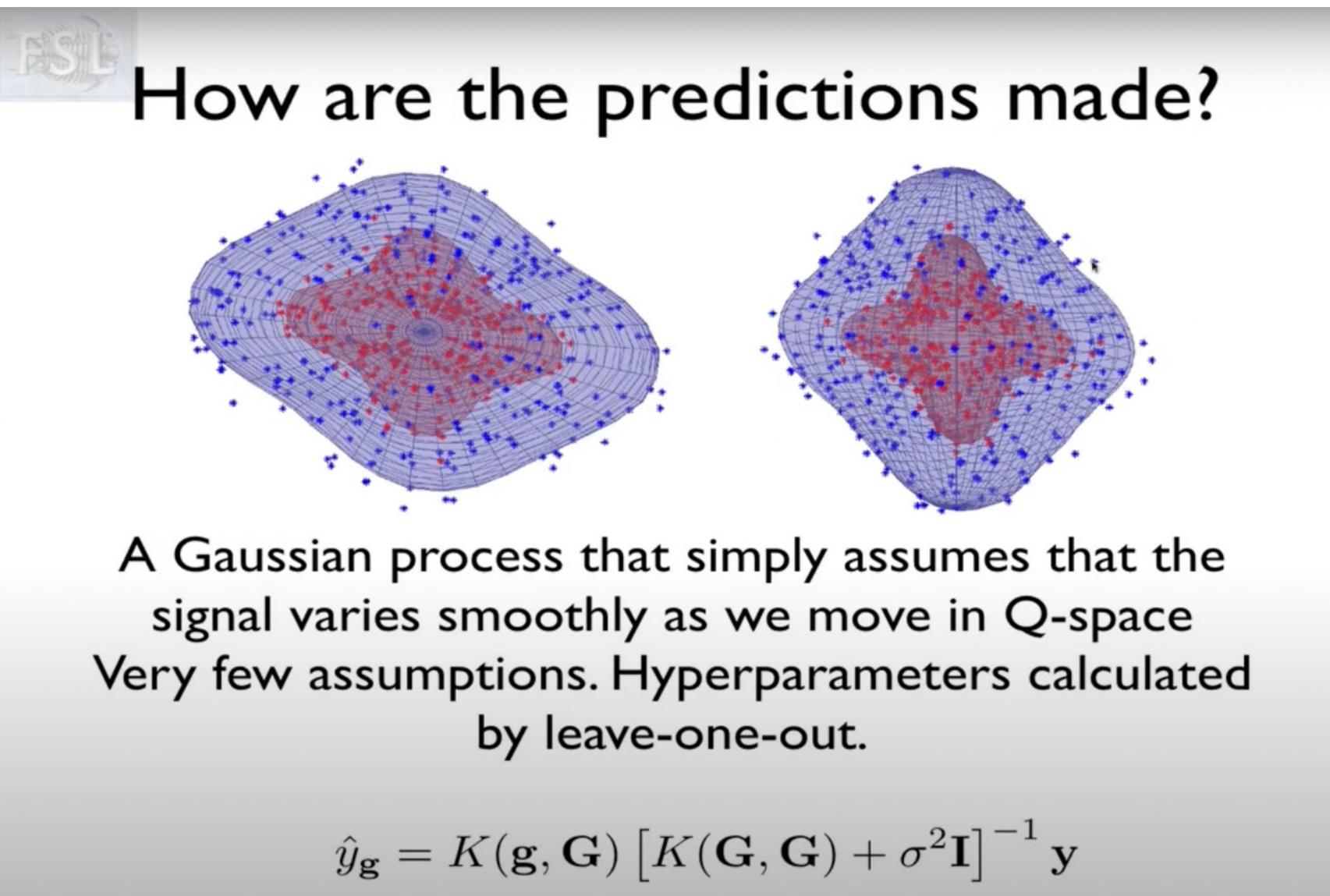


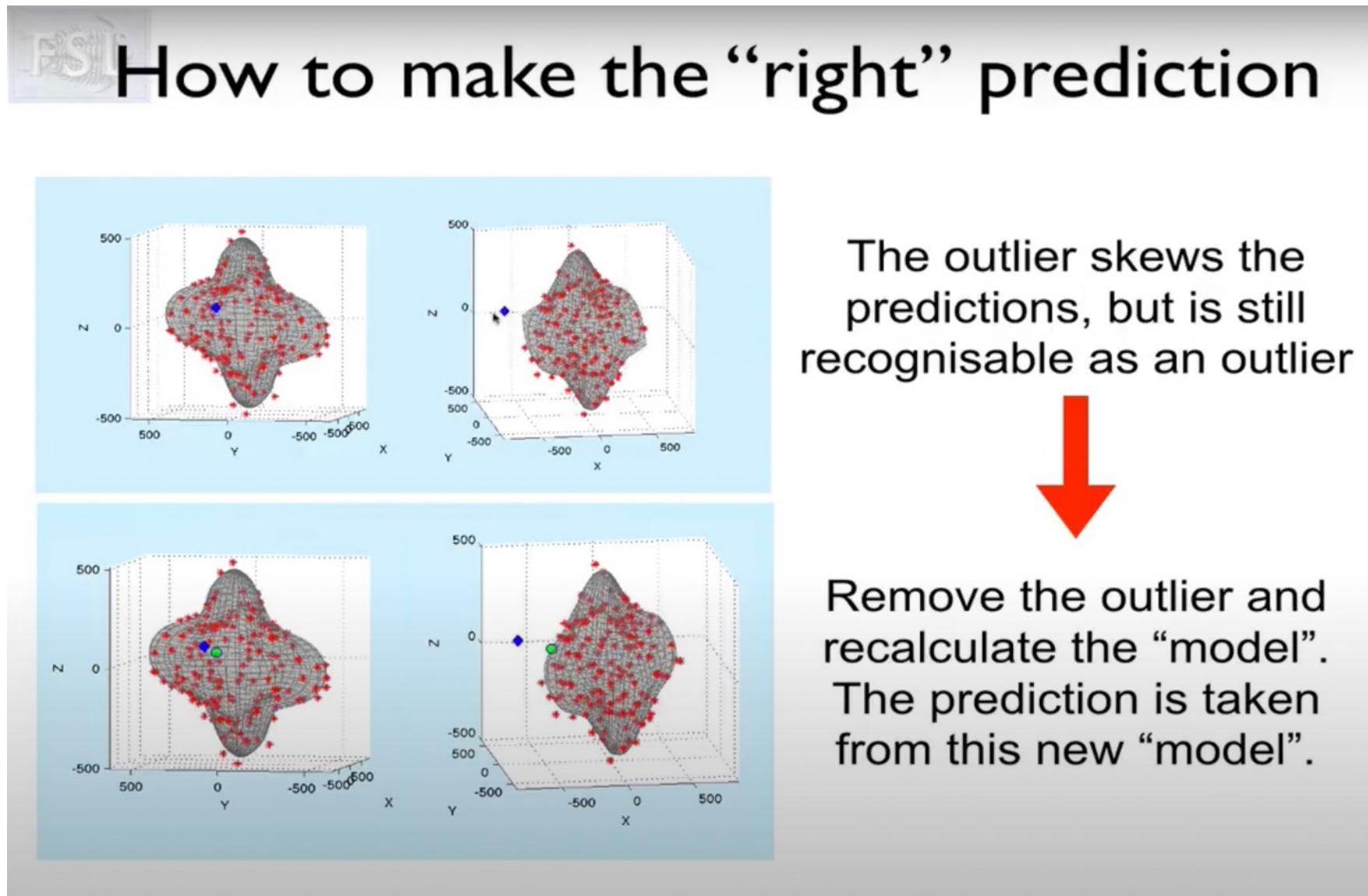
Full  
PEPOLAR  
acquisition



Only Diffusion  
direction in AP

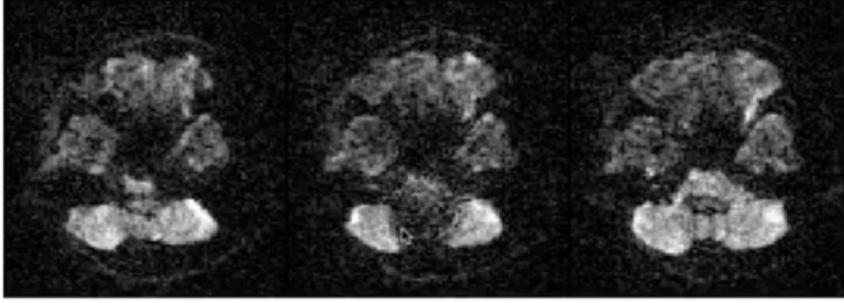






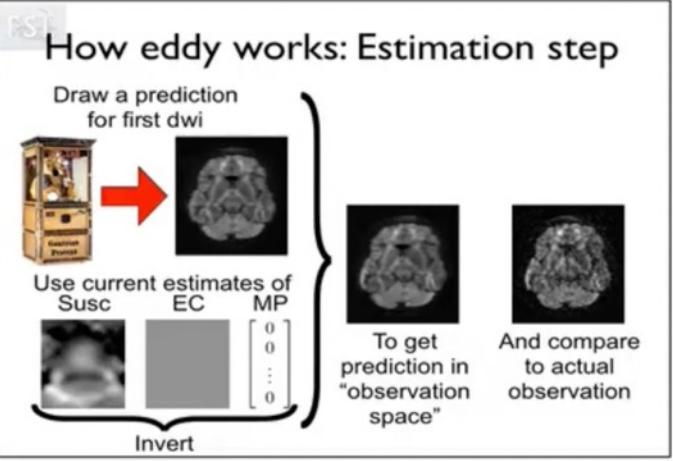
# Outlier detection

Observed data



How eddy works: Estimation step

Draw a prediction for first dwi



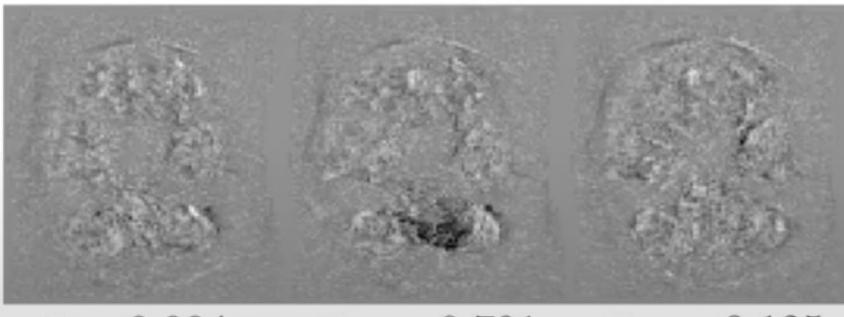
To get prediction in "observation space"

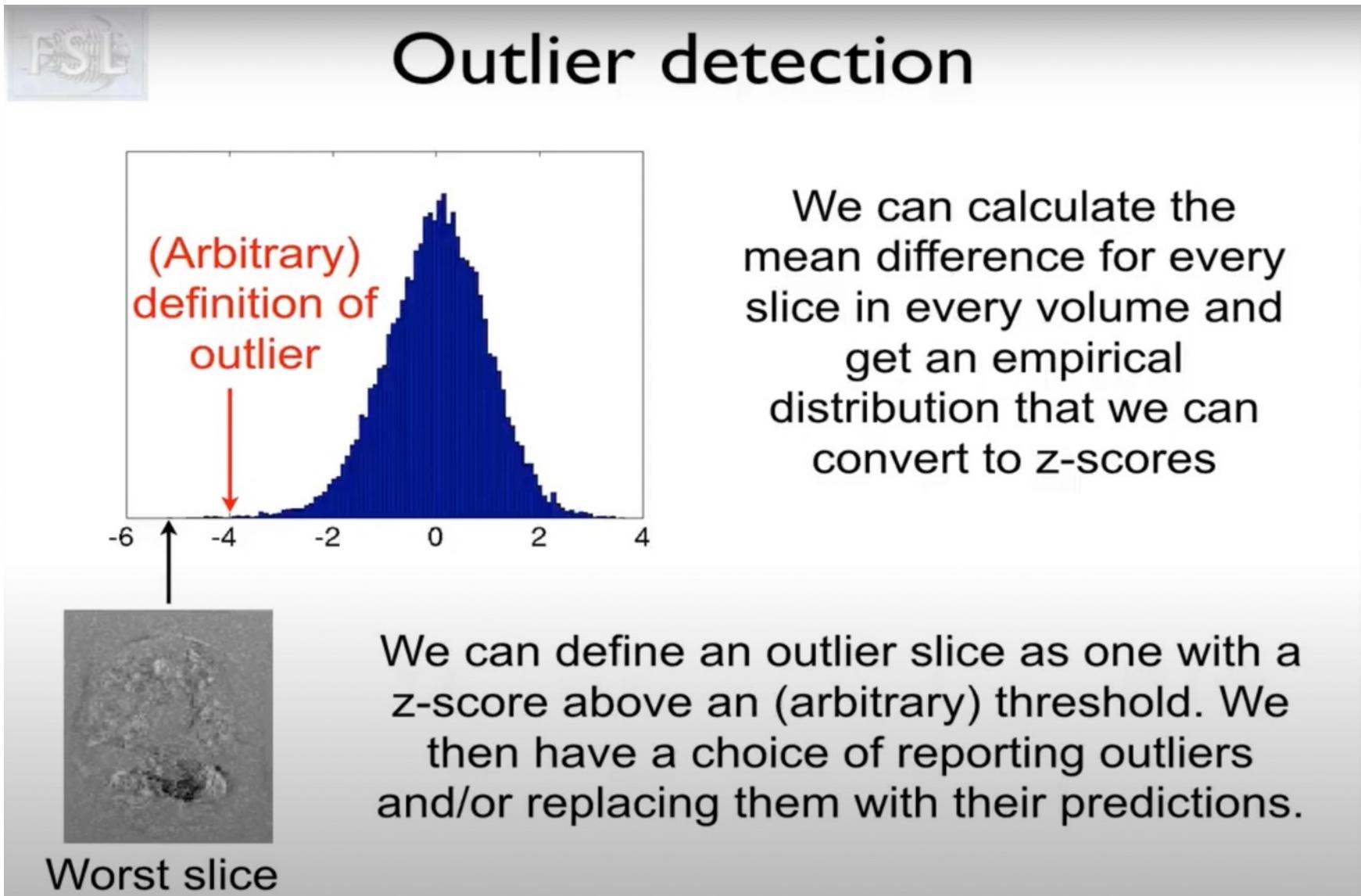
And compare to actual observation

Remember that we do all comparisons in observation space.

This allows us to calculate the per-slice mean difference between observation and prediction

Observed - predicted


$$\bar{x} = 0.084 \quad \bar{x} = -0.791 \quad \bar{x} = -0.125$$



# Recent Tractometry litterature

- (1) Zhang, S.; Zhao, F.; Yang, X.; Tan, Q.; Li, S.; Shao, H.; Su, X.; Gong, Q.; Yue, Q. Multiparametric Mapping of White Matter Reorganizations in Patients with Frontal Glioma-Related Epilepsy. *CNS Neuroscience & Therapeutics* **2023**, *29* (8), 2366–2376. <https://doi.org/10.1111/cns.14322>.
- (2) Morita, T.; Takemura, H.; Naito, E. Functional and Structural Properties of Interhemispheric Interaction between Bilateral Precentral Hand Motor Regions in a Top Wheelchair Racing Paralympian. *Brain Sci* **2023**, *13* (5), 715. <https://doi.org/10.3390/brainsci13050715>.
- (3) Lerma-Usabiaga, G.; Liu, M.; Paz-Alonso, P. M.; Wandell, B. A. Reproducible Tract Profiles 2 (RTP2) Suite, from Diffusion MRI Acquisition to Clinical Practice and Research. *Sci Rep* **2023**, *13* (1), 6010. <https://doi.org/10.1038/s41598-023-32924-7>.
- (4) Kruper, J.; Benson, N. C.; Caffarra, S.; Owen, J.; Wu, Y.; Lee, A. Y.; Lee, C. S.; Yeatman, J. D.; Rokem, A.; UK Biobank Eye and Vision Consortium. Optic Radiations Representing Different Eccentricities Age Differently. *Human Brain Mapping* **2023**, *44* (8), 3123–3135. <https://doi.org/10.1002/hbm.26267>.
- (5) Grotheer, M.; Bloom, D.; Kruper, J.; Richie-Halford, A.; Zika, S.; Aguilera González, V. A.; Yeatman, J. D.; Grill-Spector, K.; Rokem, A. *Human White Matter Myelination Rate Slows down at Birth*; preprint; Neuroscience, 2023. <https://doi.org/10.1101/2023.03.02.530800>.
- (6) Fuelscher, I.; Hyde, C.; Thomson, P.; Vijayakumar, N.; Sciberras, E.; Efron, D.; Anderson, V.; Hazell, P.; Silk, T. J. Longitudinal Trajectories of White Matter Development in Attention-Deficit/Hyperactivity Disorder. *Biological Psychiatry: Cognitive Neuroscience and Neuroimaging* **2023**, *8* (11), 1103–1112. <https://doi.org/10.1016/j.bpsc.2023.03.008>.
- (7) Clemente, A.; Attyé, A.; Renard, F.; Calamante, F.; Burmester, A.; Imms, P.; Deutscher, E.; Akhlaghi, H.; Beech, P.; Wilson, P. H.; Poudel, G.; Domínguez D, J. F.; Caeyenberghs, K. Individualised Profiling of White Matter Organisation in Moderate-to-Severe Traumatic Brain Injury Patients. *Brain Research* **2023**, *1806*, 148289. <https://doi.org/10.1016/j.brainres.2023.148289>.
- (8) Chen, Q.; Wang, M.; Wu, G.; Li, W.; Ren, X.; Wang, Y.; Wei, X.; Wang, J.; Yang, Z.; Li, X.; Li, Z.; Tang, L.; Zhang, P.; Wang, Z. Characteristics of White Matter Alterations along Fibres in Patients with Bulimia Nervosa: A Combined Voxelwise and Tractography Study. *European Journal of Neuroscience* **2023**, *58* (3), 2874–2887. <https://doi.org/10.1111/ejn.16077>.
- (9) Bianco, K. M.; Fuelscher, I.; Lum, J. A. G.; Singh, M.; Enticott, P. G.; Caeyenberghs, K.; Hyde, C. Individual Differences in Procedural Learning Are Associated with Fiber Specific White Matter Microstructure of the Superior Cerebellar Peduncles in Healthy Adults. *Cortex* **2023**, *161*, 1–12. <https://doi.org/10.1016/j.cortex.2023.01.006>.
- (10) Andica, C.; Kamagata, K.; Aoki, S. Automated Three-Dimensional Major White Matter Bundle Segmentation Using Diffusion Magnetic Resonance Imaging. *Anat Sci Int* **2023**, *98* (3), 318–336. <https://doi.org/10.1007/s12565-023-00715-9>.
- (11) Muncy, N. M.; Kimbler, A.; Hedges-Muncy, A. M.; McMakin, D. L.; Mattfeld, A. T. General Additive Models Address Statistical Issues in Diffusion MRI: An Example with Clinically Anxious Adolescents. *NeuroImage: Clinical* **2022**, *33*, 102937. <https://doi.org/10.1016/j.nicl.2022.102937>.
- (12) Grotheer, M.; Rosenke, M.; Wu, H.; Kular, H.; Querdasi, F. R.; Natu, V. S.; Yeatman, J. D.; Grill-Spector, K. White Matter Myelination during Early Infancy Is Linked to Spatial Gradients and Myelin Content at Birth. *Nat Commun* **2022**, *13* (1), 997. <https://doi.org/10.1038/s41467-022-28326-4>.
- (13) Kruper, J.; Yeatman, J. D.; Richie-Halford, A.; Bloom, D.; Grotheer, M.; Caffarra, S.; Kiar, G.; Karipidis, I. I.; Roy, E.; Chandio, B. Q.; Garyfallidis, E.; Rokem, A. Evaluating the Reliability of Human Brain White Matter Tractometry. *Aperture Neuro* **2021**, *2021* (1), 25. <https://doi.org/10.52294/e6198273-b8e3-4b63-babb-6e6b0da10669>.
- (14) Grotheer, M.; Rosenke, M.; Wu, H.; Kular, H.; Querdasi, F. R.; Natu, V.; Yeatman, J. D.; Grill-Spector, K. *White Matter Myelination during Early Infancy Is Explained by Spatial Gradients and Myelin Content at Birth*; preprint; Neuroscience, 2021. <https://doi.org/10.1101/2021.03.29.437583>.
- (15) Chamberland, M.; Genc, S.; Tax, C. M. W.; Shastin, D.; Koller, K.; Raven, E. P.; Cunningham, A.; Doherty, J.; van den Bree, M. B. M.; Parker, G. D.; Hamandi, K.; Gray, W. P.; Jones, D. K. Detecting Microstructural Deviations in Individuals with Deep Diffusion MRI Tractometry. *Nat Comput Sci* **2021**, *1*, 598–606. <https://doi.org/10.1038/s43588-021-00126-8>.