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ALZHEIMER'S ASSOCIATION INTERNATIONAL CONFERENCE®

JULY 16-20 > AMSTERDAM, NETHERLANDS, AND ONLINE

ISTAART Neuroimaging PIA THE BASICS OF NEUROIMAGING SEMINAR SERIES



ISTAART Neuroimaging PIA

The Basics of Neuroimaging Series



@NeuroimagingPIA

@alexa_pichetb



BASICS OF NEUROIMAGING

DIFFUSION MRI

DR ALEXA PICHET BINETTE

LUND UNIVERSITY, LUND, SWEDEN

Available on demand very soon!

The Basics of Neuroimaging

Data Structure and Formats

Moderator:

Alexis Moscoso Rial, PhD

Speaker:

Ludovica Griffanti, PhD

Wednesday, April 5, 9 a.m. CT

The Basics of Neuroimaging

Structural Magnetic Resonance Imaging (MRI)

Moderator:

Tavia Evans, PhD:
Erasmus MC, Netherlands

Panelists:

David Cash, PhD:
University College London, United Kingdom

Friday, April 14, 9 a.m. CT

The Basics of Neuroimaging

Positron Emission Tomography (PET)

Moderator:

Lyduine Collij, Ph.D.

Panelists:

Tobey Betthausen, Ph.D.

Wednesday, April 19, 12 p.m. CT

The Basics of Neuroimaging

Diffusion-Weighted Imaging (DWI)

Moderator:

Tom Veale, Ph.D.

Panelists:

Alexa Pichet Binette, Ph.D.

Friday, April 21, 9 a.m. CT

The Basics of Neuroimaging

Functional Magnetic Resonance Imaging (fMRI)

Moderator:

Betty Tijms, Ph.D.

Speaker:

Luigi Lorenzini, Ph.D.

Wednesday, April 26, 10 a.m. CT

By the end of this session, you should be able to:

- Understand the acquisition and data structure of diffusion MRI
- Outline the basic preprocessing steps needed for diffusion MRI data and how to look for issues in the data and processing
- Describe the main outputs from diffusion tensor models

NEUROIMAGING DATA ANALYSIS: A BLUEPRINT FOR DIFFUSION MRI

1. Data acquisition

Aim: obtain good quality and consistent data

2. Data preprocessing

Aim: Reduce noise and prepare data for further analyses

3. Single-subject analysis

Aim: Obtain measure of interest for each subject (often an image)

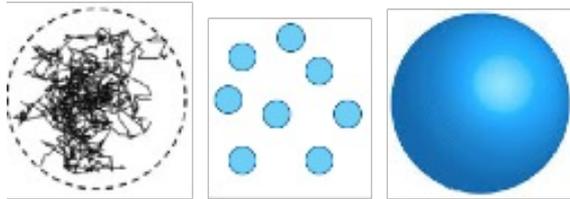
4. Group-level analysis

Aim: Compare single-subject results across groups

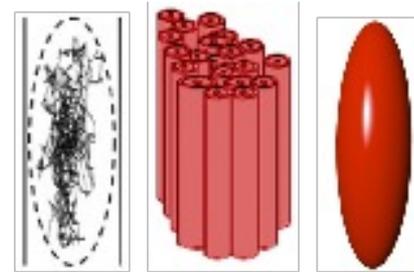
5. Statistical inference

Aim: test reliability of results and generalizability to the general population

1. Data acquisition

Basic principles of diffusion:

Free diffusion – Isotropy (in CSF)



Restricted diffusion – Anisotropy (in WM)

We are measuring the movement of water molecules along certain directions in the brain.

1. Data acquisition

Orientation is key in DWI. We measure the diffusion of water molecules along different directions

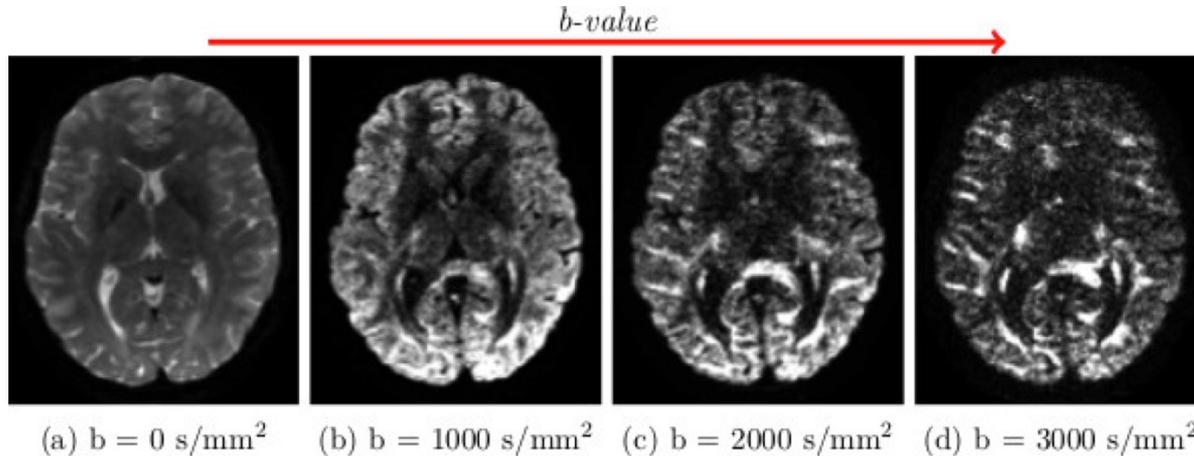
The diffusion-weighting signal varies according to the gradient direction in which it measures the image. If the gradient direction is aligned with the underlying white matter structure, the signal in the image is attenuated. It is weaker.



On each image
a different
direction (x,y,z)
is applied

1. Data acquisition

Another key aspect is the **b-value**, which is related to the strength and duration of the gradients used to generate diffusion-weighted images.

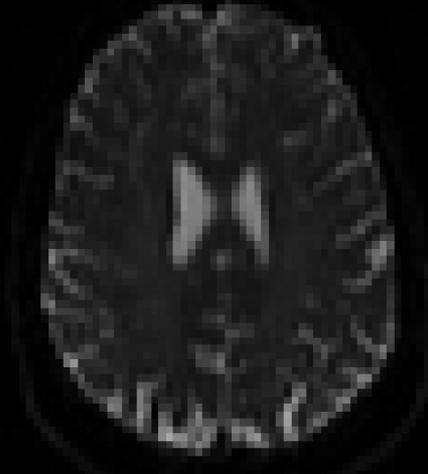
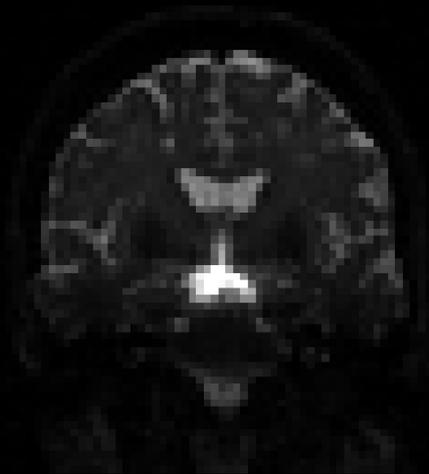
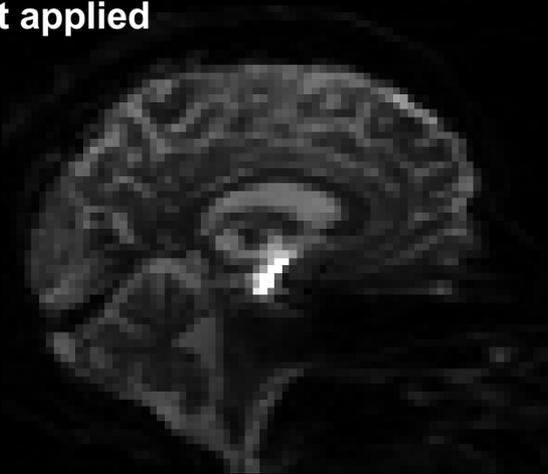


Higher b-values =
“longer time to let
molecules
diffuse”

b=1000 is
standard is most
basic research
MRI sequence

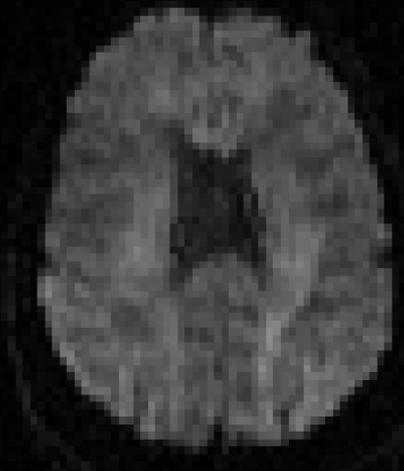
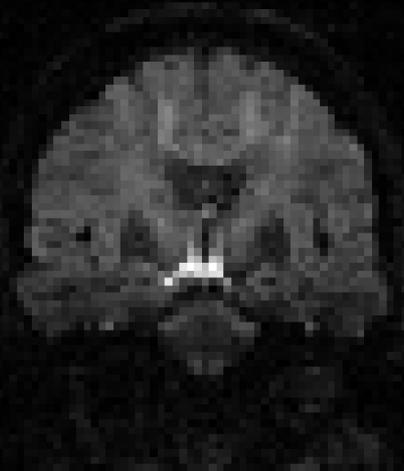
b-value = 0

No gradient applied

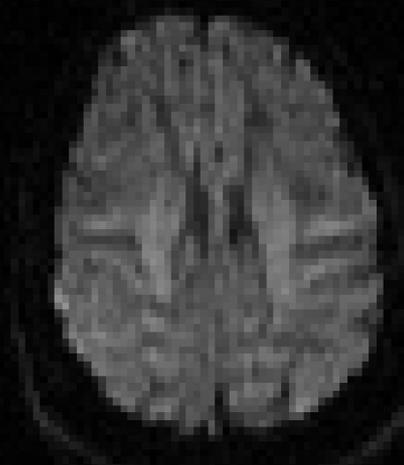
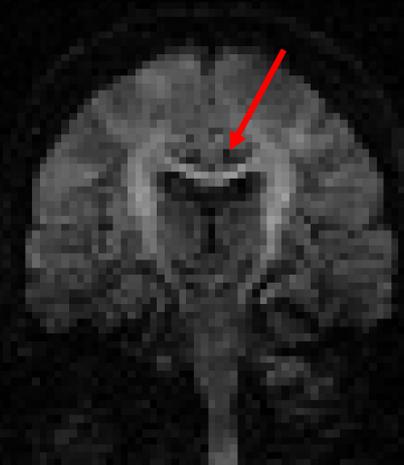


b-value = 1000

**Gradient in the x-direction
Left-Right**



**Gradient in the y-direction
Anterior-Posterior**



DIFFUSION IMAGING: DATA ORGANIZATION

1. Data acquisition

1.b Data organization



<https://bids.neuroimaging.io>

```
my_dataset/  
├── participants.tsv  
├── sub-01/  
│   ├── anat/  
│   │   └── sub-01_T1w.nii.gz  
│   ├── func/  
│   │   ├── sub-01_task-rest_bold.nii.gz  
│   │   └── sub-01_task-rest_bold.json  
│   └── dwi/  
│       ├── sub-01_dwi.nii.gz  
│       ├── sub-01_dwi.json  
│       ├── sub-01_dwi.bval  
│       └── sub-01_dwi.bvec
```

T1-Weighted



April 14
webinar

DIFFUSION IMAGING:
DATA ORGANIZATION

1. Data acquisition

1.b Data organization

<https://bids.neuroimaging.io>4-D file where each image has a specific diffusion
gradient and orientation

```
my_dataset/  
├── participants.tsv  
├── sub-01/  
│   ├── anat/  
│   │   └── sub-01_T1w.nii.gz  
│   ├── func/  
│   │   ├── sub-01_task-rest_bold.nii.gz  
│   │   └── sub-01_task-rest_bold.json  
│   └── dwi/  
│       ├── sub-01_dwi.nii.gz  
│       ├── sub-01_dwi.json  
│       ├── sub-01_dwi.bval  
│       └── sub-01_dwi.bvec
```

DWI scan



DIFFUSION IMAGING: DATA ORGANIZATION

1. Data acquisition

1.b Data organization



<https://bids.neuroimaging.io>

Vectors of length equal to the total number of directions of the diffusion scan

```

my_dataset/
├── participants.tsv
├── sub-01/
│   ├── anat/
│   │   └── sub-01_T1w.nii.gz
│   ├── func/
│   │   ├── sub-01_task-rest_bold.nii.gz
│   │   └── sub-01_task-rest_bold.json
│   └── dwi/
│       ├── sub-01_dwi.nii.gz
│       ├── sub-01_dwi.json
│       ├── sub-01_dwi.bval
│       └── sub-01_dwi.bvec
  
```



b-values

Orientation (x,y,z)

bval

bvec

0

0, 0, 0

1000

0.999, -0.003, -0.003

1000

0.001, 0.999, -0.003

...

...

...

...

1000

0.265, 0.960, 0.082

DIFFUSION IMAGING: PREPROCESSING

1. Data acquisition



2. Data preprocessing

Key steps:

- Correcting for susceptibility-induced distortions
- Correcting for eddy currents and movement

DIFFUSION IMAGING: SUSCEPTIBILITY-INDUCED ARTEFACTS

1. Data acquisition

2. Data preprocessing

Key steps:

-Correcting for susceptibility-induced distortions

- Correcting for eddy currents and movement

Susceptibility induced distortions



Some parts of the brain can appear distorted depending on their magnetic properties.

One common way to correct the distortions with DWI data is by acquiring a b_0 image acquired with a different phase-encoding, and merging the two types of images running [TOPUP](#).

Source: FSL Diffusion Toolbox

DIFFUSION IMAGING: SUSCEPTIBILITY-INDUCED ARTEFACTS

1. Data acquisition



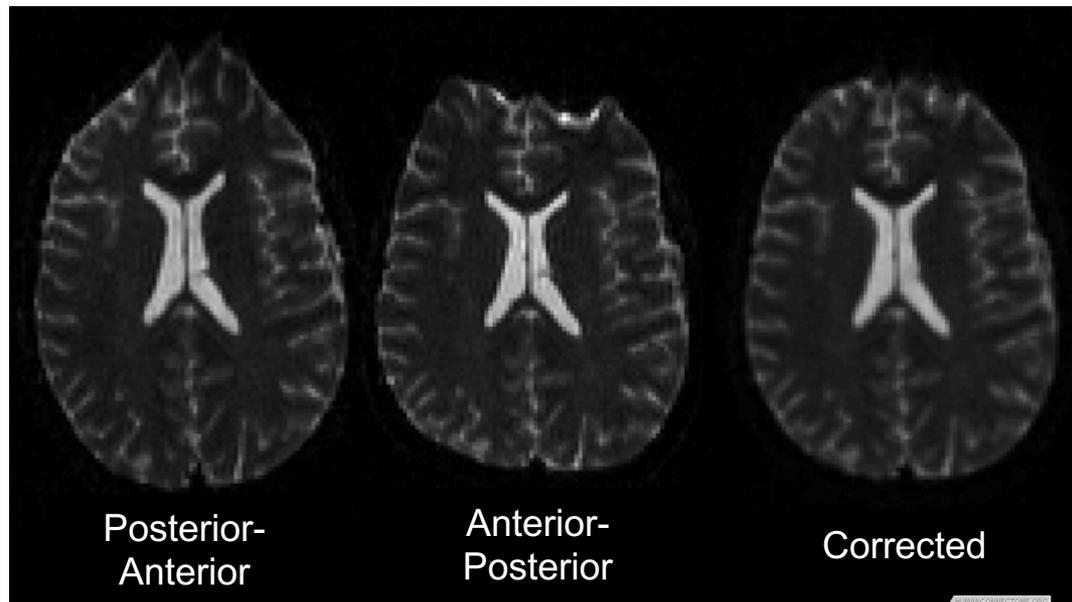
2. Data preprocessing

Key steps:

-Correcting for susceptibility-induced distortions

- Correcting for eddy currents and movement

Example from HCP:



DIFFUSION IMAGING: SUSCEPTIBILITY-INDUCED ARTEFACTS

1. Data acquisition



2. Data preprocessing

Key steps:

-Correcting for susceptibility-induced distortions

- Correcting for eddy currents and movement

Eddy currents arise from electric current due to strong and fast changing gradients.

FSL's eddy is a tool to correct for eddy current-induced distortions and movement on the image. It also does outlier detection and will replace signal loss by non-parametric predictions.

DIFFUSION IMAGING: EDDY CURRENTS

1. Data acquisition



2. Data preprocessing

Key steps:

-Correcting for susceptibility distortions

- **Correcting for eddy currents and movement**



Example of diffusion scan with artefact

DIFFUSION IMAGING: EDDY CURRENTS

1. Data acquisition

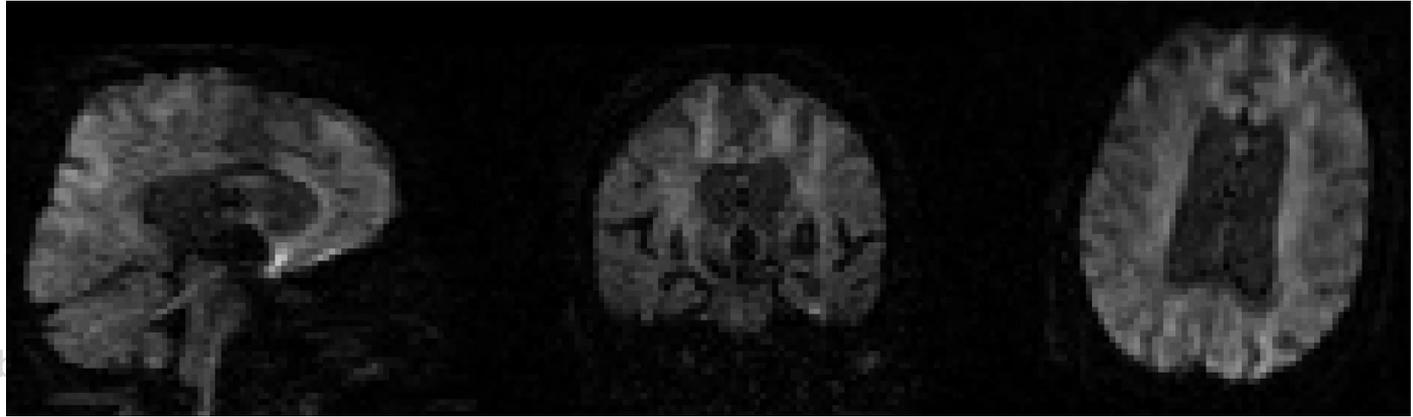


2. Data preprocessing

Key steps:

-Correcting for susceptibility distortions

- **Correcting for eddy currents and movement**



After eddy correction

DIFFUSION IMAGING: EDDY CURRENTS

1. Data acquisition



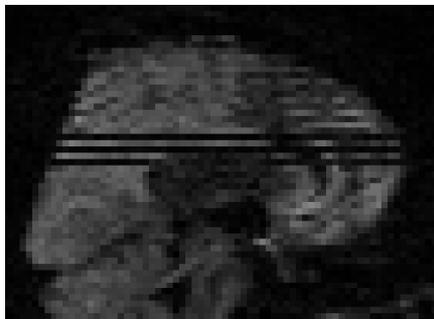
2. Data preprocessing

Key steps:

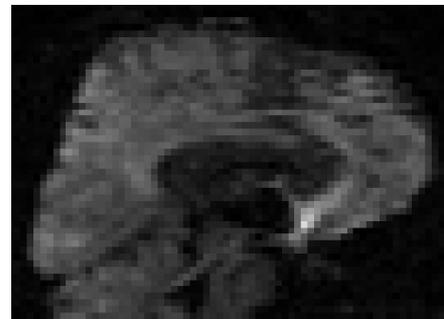
-Correcting for susceptibility-induced distortions

- **Correcting for eddy currents and movement**

Raw data



After eddy correction



IMPORTANT TO INSPECT YOUR DATA!

FROM THE DIFFUSION SCAN TO SUBJECT-LEVEL MEASURES

1. Data acquisition

2. Data preprocessing

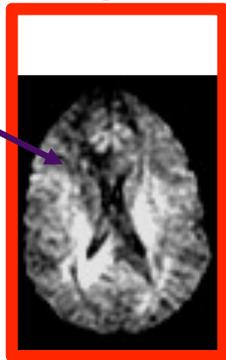
3. Single-subject analysis

How to go from the diffusion scan to representing the underlying white matter microstructure organization?

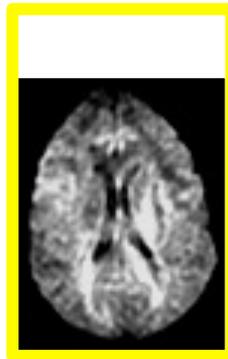
TENSOR TO REPRESENT DIFFUSION SIGNAL



xy



xz



yz



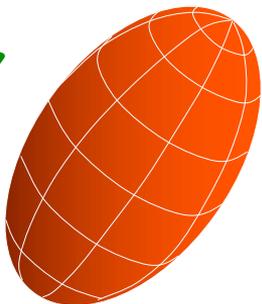
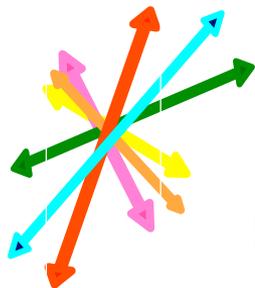
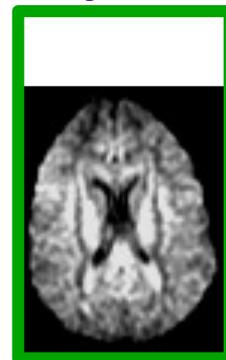
-xy



-xz



y-z



Basser et al, 1994

1. Data acquisition

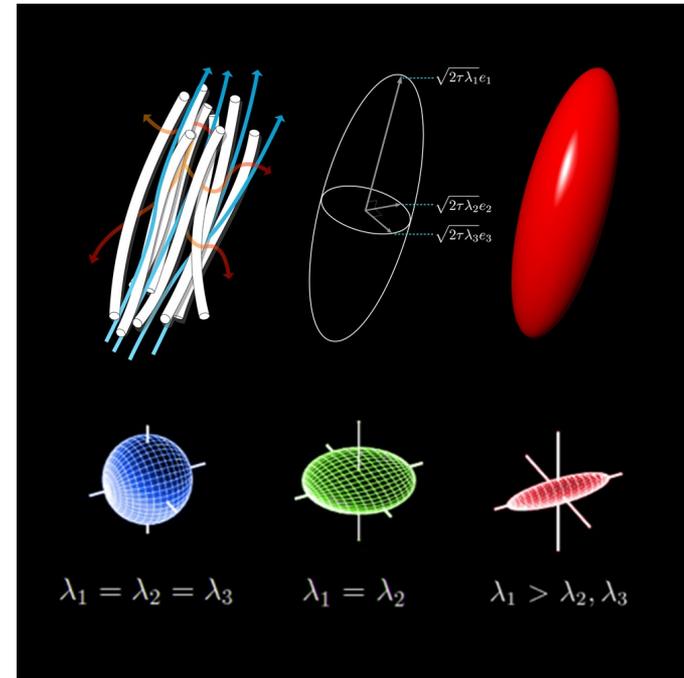
2. Data preprocessing

3. Single-subject analysis

We can represent the diffusion direction in each voxel using **tensors**.

Each tensor is described by 3 eigen vectors which represent diffusivity along 3 axes, with the first eigenvalue being the main diffusion axis.

The **diffusion tensor model** is the most common way to fit diffusion data. It is often a prerequisite for any diffusion imaging analysis pipeline.



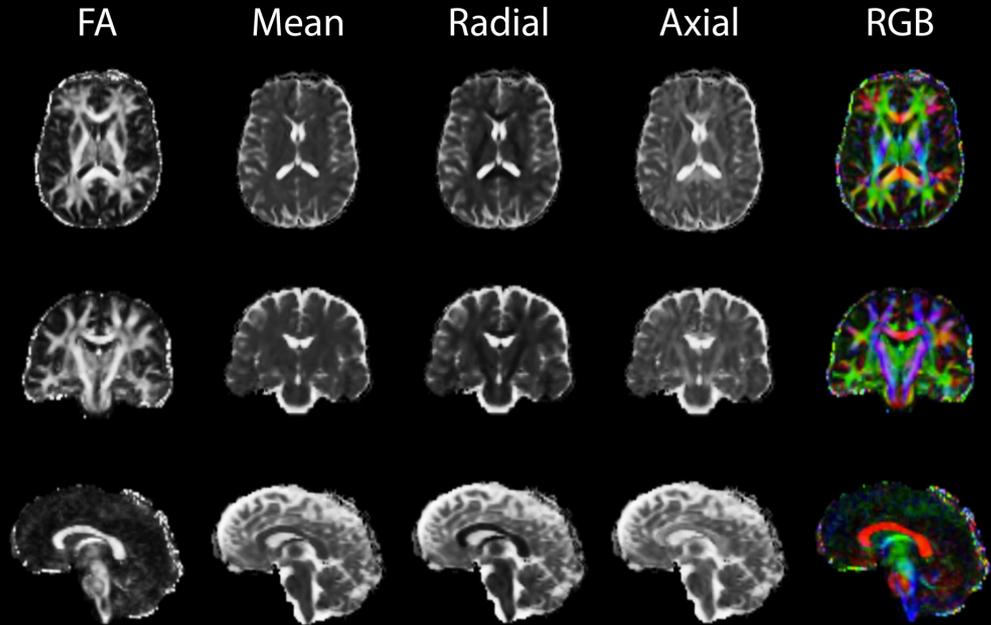
1. Data acquisition

2. Data preprocessing

3. Single-subject analysis

There are key measures that can be derived from the diffusion tensor model, namely:

- **Fractional anisotropy (FA)**
- **Mean, Radial and Axial diffusivities**

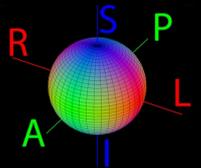


$$\text{FA} : \frac{3}{2} \cdot \sqrt{\frac{(\lambda_1 - \lambda_2)^2 + (\lambda_1 - \lambda_3)^2 + (\lambda_2 - \lambda_3)^2}{\lambda_1^2 + \lambda_2^2 + \lambda_3^2}}$$

$$\text{Mean} : \frac{\lambda_1 + \lambda_2 + \lambda_3}{3}$$

$$\text{Radial} : \frac{\lambda_2 + \lambda_3}{2}$$

$$\text{Axial} : \lambda_1$$



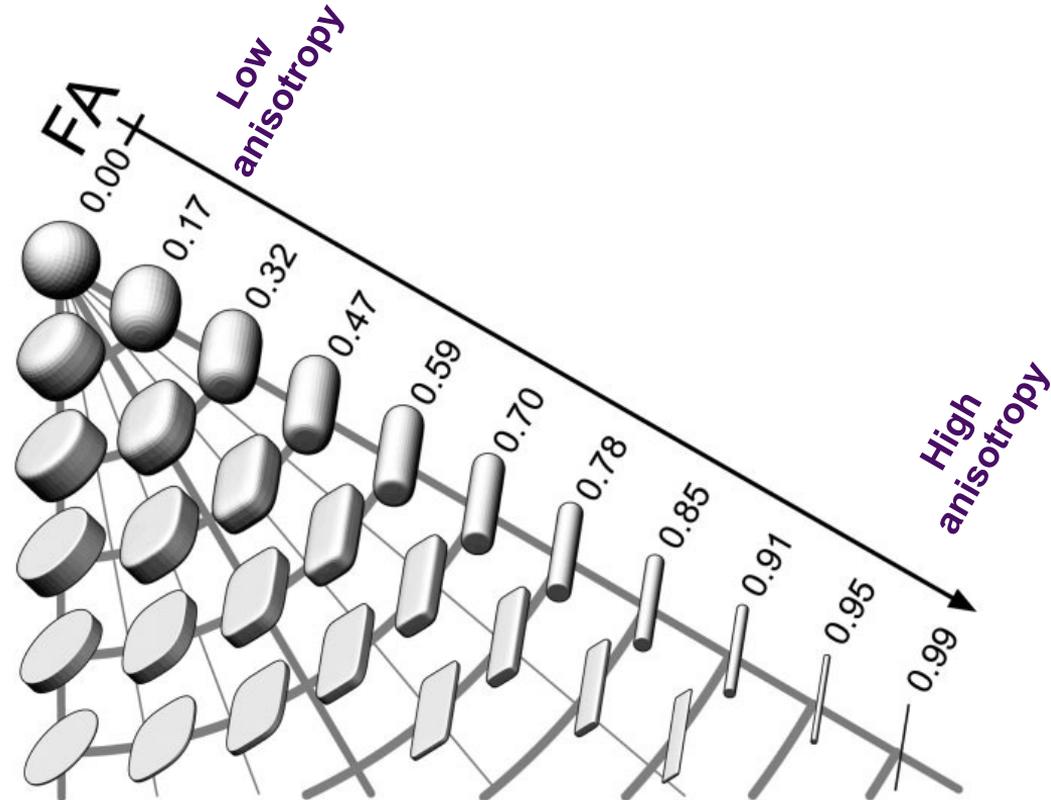
1. Data acquisition

2. Data preprocessing

3. Single-subject analysis

There are key measures that can be derived from the diffusion tensor model, namely:

- **Fractional anisotropy (FA)**
- Mean, Radial and Axial diffusivities



GROUP-LEVEL ANALYSIS OF DIFFUSION DATA

1. Data acquisition

2. Data preprocessing

3. Single-subject analysis

4. Group-level analysis

5. Statistical inference

- **Numbers** fed into 'classic' stats software
- **Images** require specific stats (usually within imaging software tools)
- Input = Single subject image
- Output = Statistical maps in pseudocolours shows **significant voxels**, overlaid on template.
- **Atlases** can help interpreting results

GROUP-LEVEL ANALYSIS OF DIFFUSION DATA – NUMERICAL VALUES

1. Data acquisition

2. Data preprocessing

3. Single-subject analysis

4. Group-level analysis

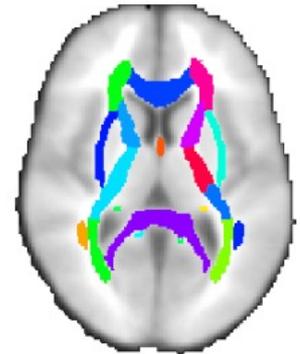
5. Statistical inference

- **Numbers** fed into 'classic' stats software

Example: We can extract diffusion measures (FA, MD, ...) for each participant in known white matter tract
Each label corresponds to a white matter tract.

- **Atlases** can help interpreting results

JHU DTI Atlas
John Hopkins University



GROUP-LEVEL ANALYSIS OF DIFFUSION DATA – IMAGES AS INPUT

1. Data acquisition

2. Data preprocessing

3. Single-subject analysis

4. Group-level analysis

5. Statistical inference

- **Numbers** fed into 'classic' stats software
- **Images** require specific stats (usually within imaging software tools)
- Input = Single subject image
- Output = Statistical maps in pseudocolours shows **significant voxels**, overlaid on template.
- **Atlases** can help interpreting results

GROUP-LEVEL ANALYSIS OF DIFFUSION DATA – IMAGES AS INPUT

1. Data acquisition

2. Data preprocessing

3. Single-subject analysis

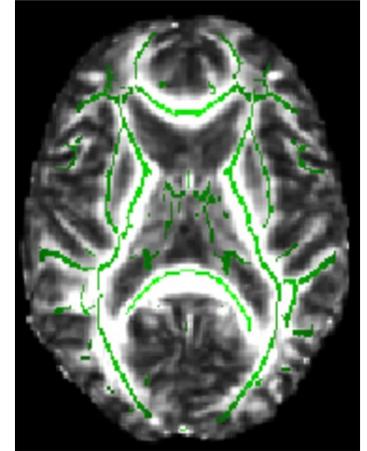
4. Group-level analysis

5. Statistical inference

Example: Tract-Based Spatial Statistics

All participants' FA maps are aligned into a common space

A mean FA skeleton which represents all tracts common to the group is created.



GROUP-LEVEL ANALYSIS OF DIFFUSION DATA – IMAGES AS INPUT

1. Data acquisition

2. Data preprocessing

3. Single-subject analysis

4. Group-level analysis

5. Statistical inference

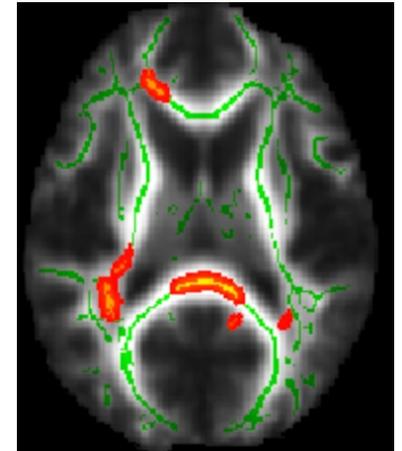
Example: Tract-Based Spatial Statistics

All participants' FA maps are aligned into a common space

A mean FA skeleton which represents all tracts common to the group is created.

Each participant's aligned FA data is then projected onto this skeleton.

Voxelwise analyses can then be performed across participants.

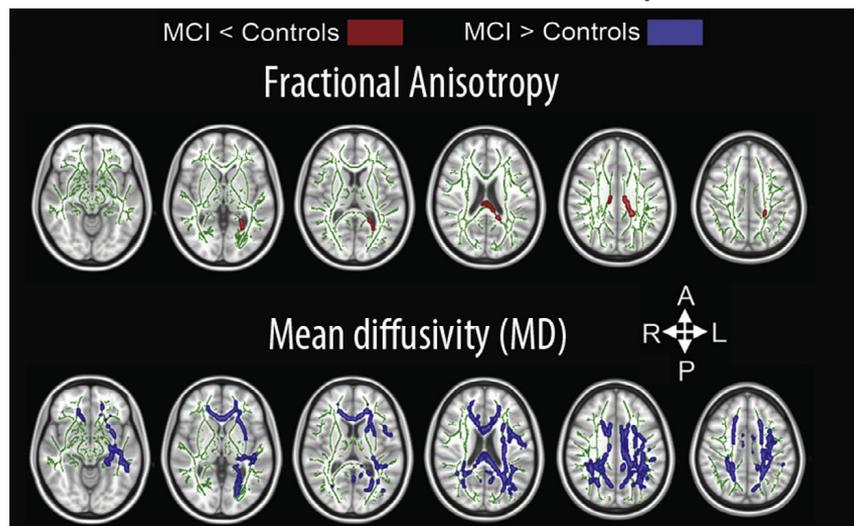


Significant voxels that are different between two groups, related to a variable of interest, etc

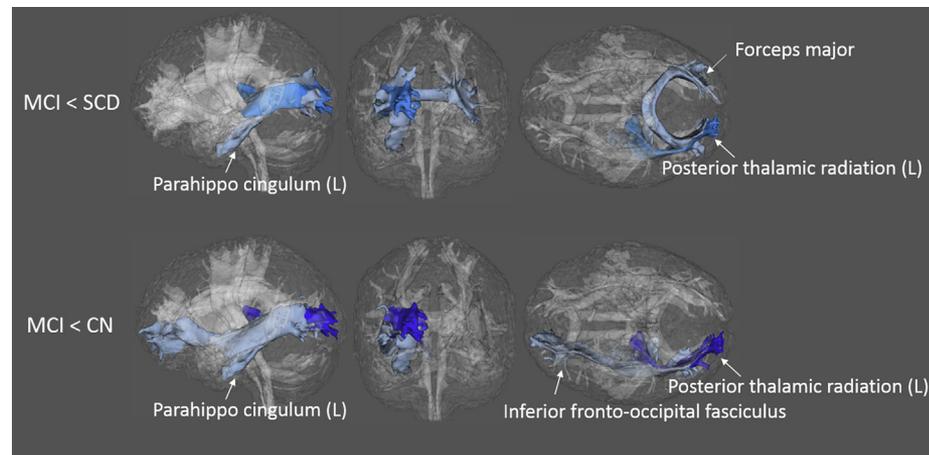
DIFFERENCES IN DIFFUSION MEASURES ACROSS THE AD CONTINUUM

Differences in white matter microstructure across diagnostic groups

↓ FA and ↑ MD across the AD spectrum



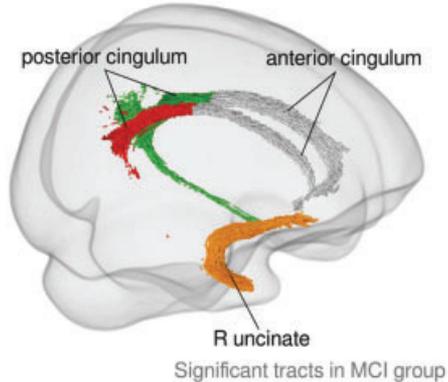
Amlien et al., *Radiology*, 2013



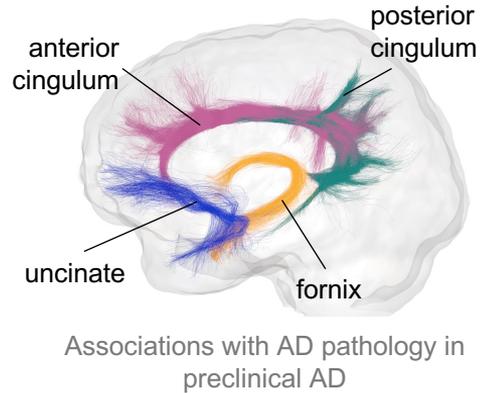
Wen et al., *Alz Dem: DADM*, 2019

DIFFUSION MEASURES IN RELATION WITH AD PATHOLOGY IN VIVO

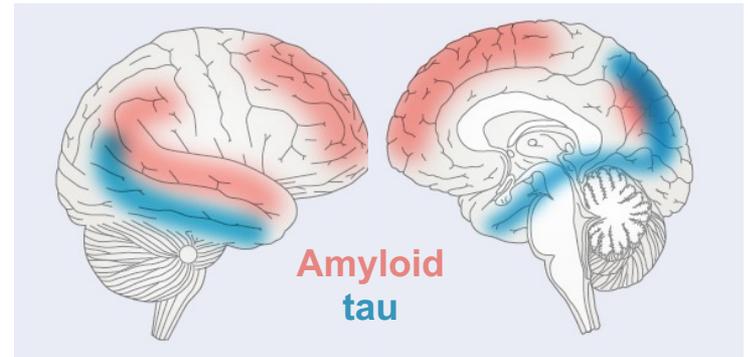
There are commonly affected white matter bundles in AD ... that connect regions where AD pathological proteins accumulate



Mito et al., *Brain*, 2018
Jacobs et al., *Nature Neuro*, 2018



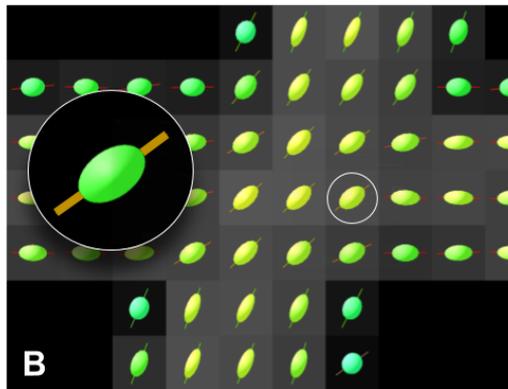
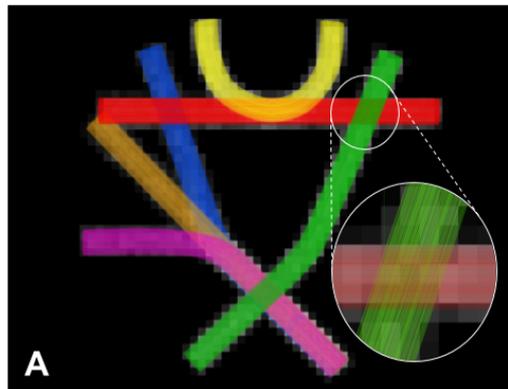
Pichet Binette et al., *eLife*, 2021



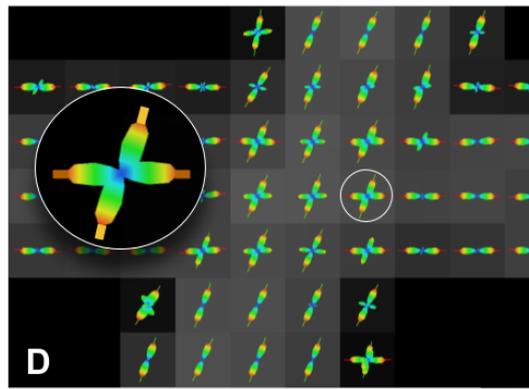
Jagust, *Nature Reviews*, 2018

- Advanced diffusion model:

There are limitations to the diffusion tensor models, and another commonly used model that can better recapitulate crossing fibers are the **fiber orientation distribution functions (fODF)**



Tensor

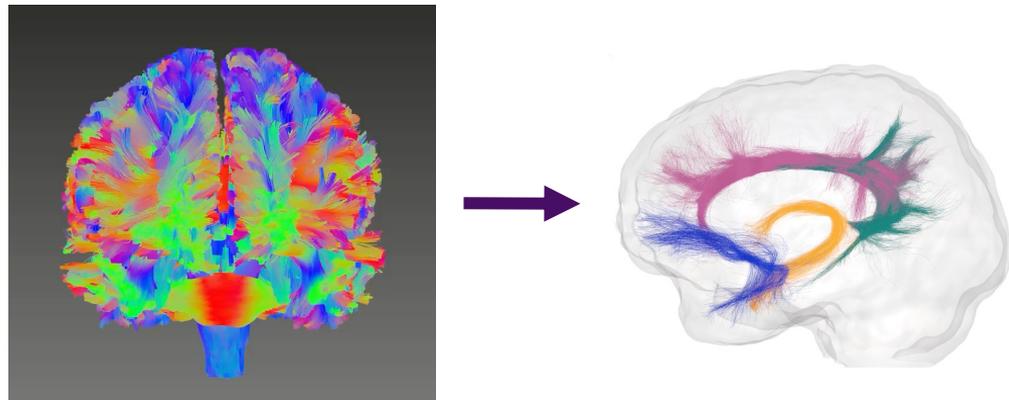


Fiber orientation distribution function (fODf)

- Tractography

After fitting the data to a model, another possibility is to do tractography.

Tractography is a way to "reconstruct" the white matter fibers to generate a whole-brain tractogram.

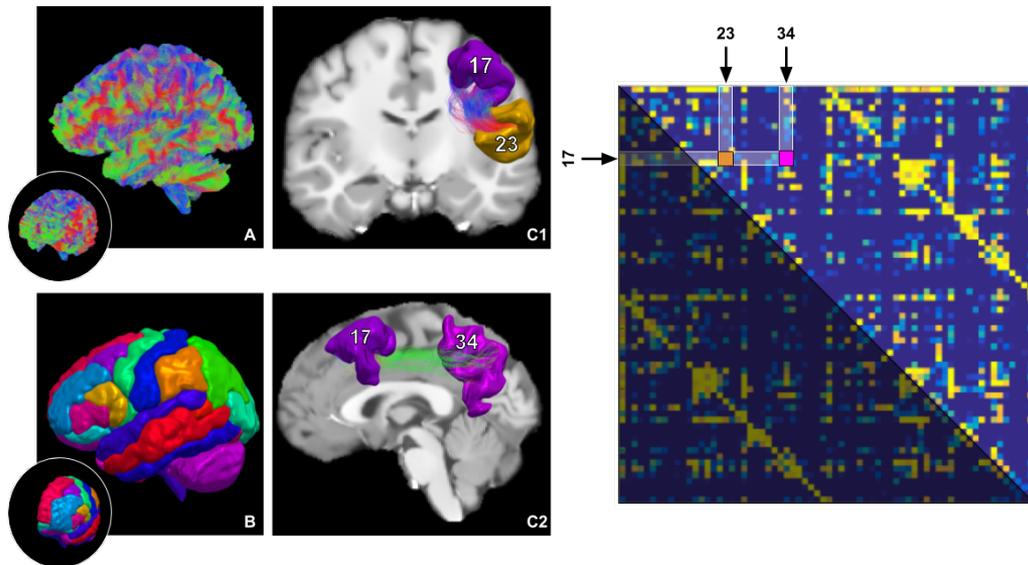


We can extract known anatomical white matter bundles from the tractogram

- Structural connectivity

After fitting the data to a model, another possibility is to do tractography.

Tractography is a way to "reconstruct" the white matter fibers to generate a whole-brain tractogram.



We can generate a structural connectivity matrix from the tractogram

NEUROIMAGING DATA ANALYSIS: A BLUEPRINT FOR DIFFUSION MRI

1. Data acquisition

Aim: obtain good quality and consistent data

Diffusion sequence has different b-values and each image has a specific gradient direction. Data includes diffusion scan, bvec and bval vectors. Optimize protocol for research aim.

2. Data preprocessing

Aim: Reduce noise and prepare data for further analyses

Susceptibility-induced and eddy currents distortions: correct for magnetic field inhomogeneities, eddy-induced currents, motion and signal loss. Requires careful checking.

3. Single-subject analysis

Aim: Obtain measure of interest for each subject (often an image)

Image of different diffusion measures: Fractional anisotropy, Mean diffusivity, Axial diffusivity, Radial diffusivity, ... Examples limited to DTI, other measure can be derived from other models.

4. Group-level analysis

Aim: Compare single-subject results across groups

Voxelwise analyses with images
Individual numerical values from regions of interest can be extracted

5. Statistical inference

Aim: test reliability of results and generalizability to the general population

Comparisons between groups
Associations of diffusion measures with other clinical data

SOFTWARE TO PROCESS DIFFUSION IMAGING



www.mrtrix.org

MRtrix3

Advanced tools for the analysis of diffusion MRI data

Download Use MRtrix3



FSL
 Diffusion
 Toolbox

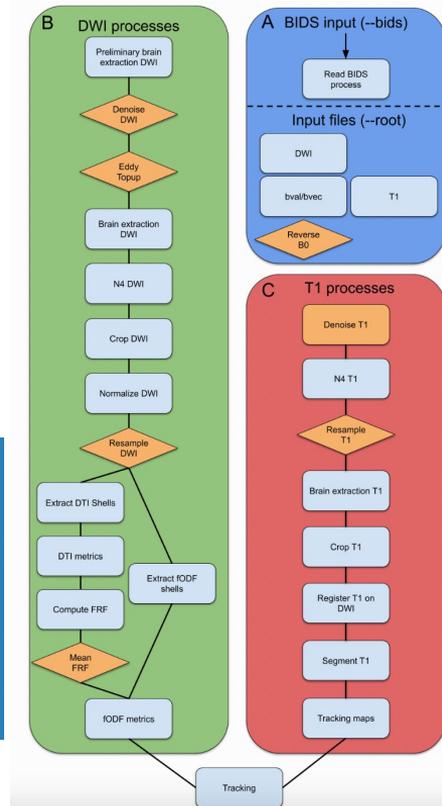
dMRIprep

<https://www.nipreps.org/dmriprep/index.html>

Basic preprocessing steps such as head-motion correction, susceptibility-derived distortion correction, eddy current correction, etc. providing outputs that can be easily submitted to a variety of diffusion models.

DIPY

Diffusion in Python
dipy.org





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<https://training.alz.org/Research-Webinars>

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Neuroimaging PIA: Advanced MR imaging in Alzheimer's disease and related dementias

Recorded: April 21, 2022

The basic principles and latest advances in arterial spin labelling and diffusion weighted imaging discussed in the context of neurodegenerative diseases. The application of these techniques, focusing on scientific research is covered.

Neuroimaging PIA: Getting started with connectomics analysis

Recorded: May 12, 2022

This webinar is a primer on how to setup, generate, review and interpret structural and functional connectivity analyses. Speakers go over best practices and what pitfalls to watch out for with their work.

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AAIC>23

POP QUIZ!

What is the signal that is acquired on a diffusion MRI scan?

- a) The orientation of white matter fibers
- b) The movement of water molecules
- c) The white matter density
- d) All of the above

What is the signal that is acquired on a diffusion MRI scan?

- a) The orientation of white matter fibers
- b) The movement of water molecules**
- c) The white matter density
- d) All of the above

What do the main preprocessing steps in diffusion MRI try to account for?

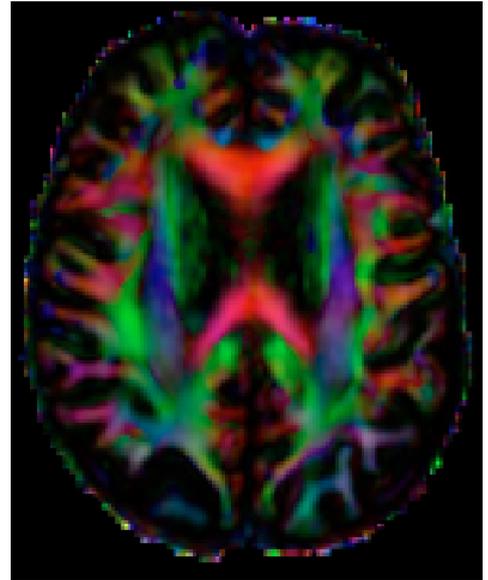
- a) Eddy currents, movement, and registration
- b) Susceptibility-induced distortions, movement and registration
- c) Susceptibility-induced distortions, eddy currents and movement

What do the main preprocessing steps in diffusion MRI try to account for?

- a) Eddy currents, movement, and registration
- b) Susceptibility-induced distortions, movement and registration
- c) Susceptibility-induced distortions, eddy currents and movement**

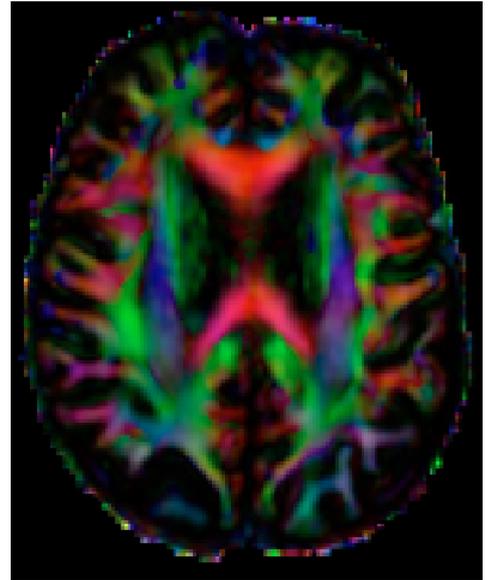
On this map, what do the colors represent?

- a) The strength of structural connectivity
- b) Mean diffusivity
- c) Length of the white matter fibers
- d) The main direction of the white matter fibers



On this map, what do the colors represent?

- a) The strength of structural connectivity
- b) Mean diffusivity
- c) Length of the white matter fibers
- d) The main direction of the white matter fibers**



If we compare cognitively unimpaired participants to patients with AD dementia, the fractional anisotropy will be:

- a) Lower in the AD patients
- b) Lower in the cognitively unimpaired participants
- c) Unchanged between the two groups

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THANK YOU!

On demand at <https://training.alz.org/>:

Basics of Neuroimaging: Data structure and formats by Ludovica Griffanti

Basics of Neuroimaging: Structural Magnetic Resonance Imaging (MRI) by David Cash

Basics of Neuroimaging: Positron emission tomography (PET) by Tobey Betthausen

Next up!

Basics of Neuroimaging: Functional Magnetic Resonance Imaging (fMRI) by Luigi Lorenzini

26 April, 2023; 10AM – 11AM CT

GETTING STARTED WITH NEUROIMAGING WORKSHOP Friday July 14 8:00-12:00 Amsterdam



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