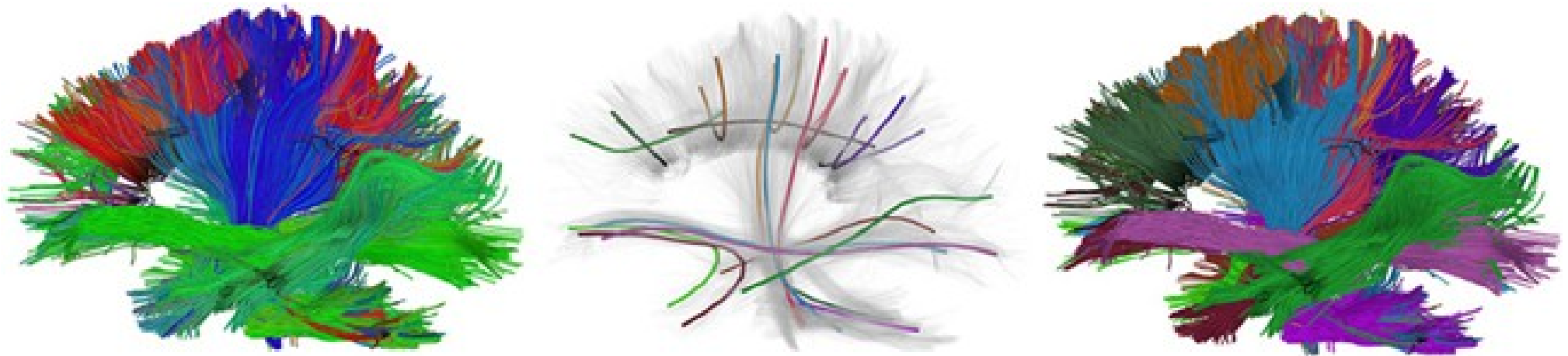
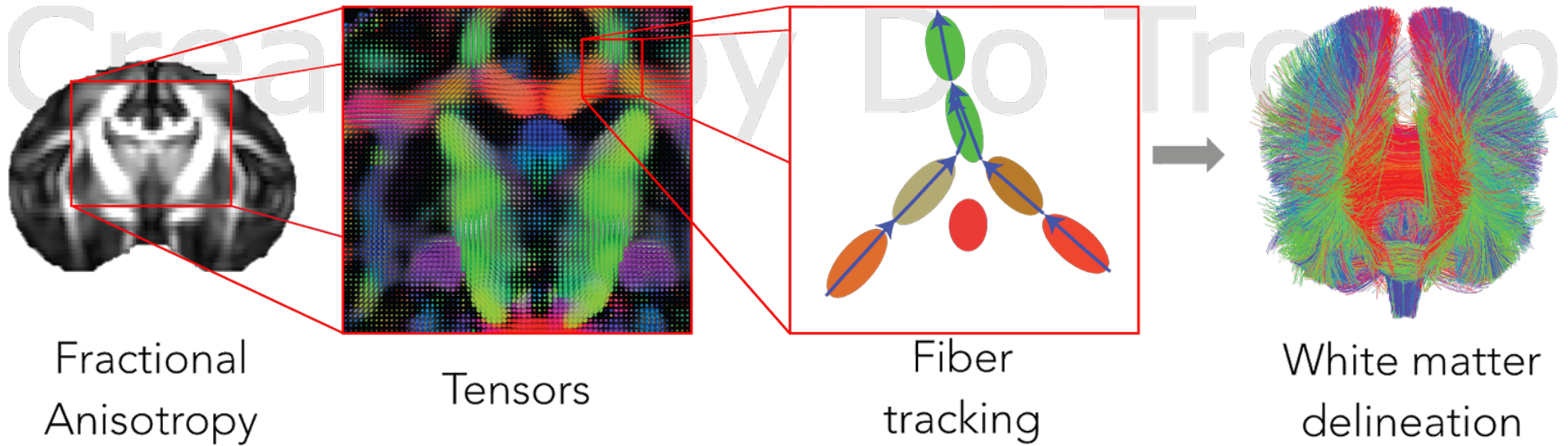


Tractography tutorial

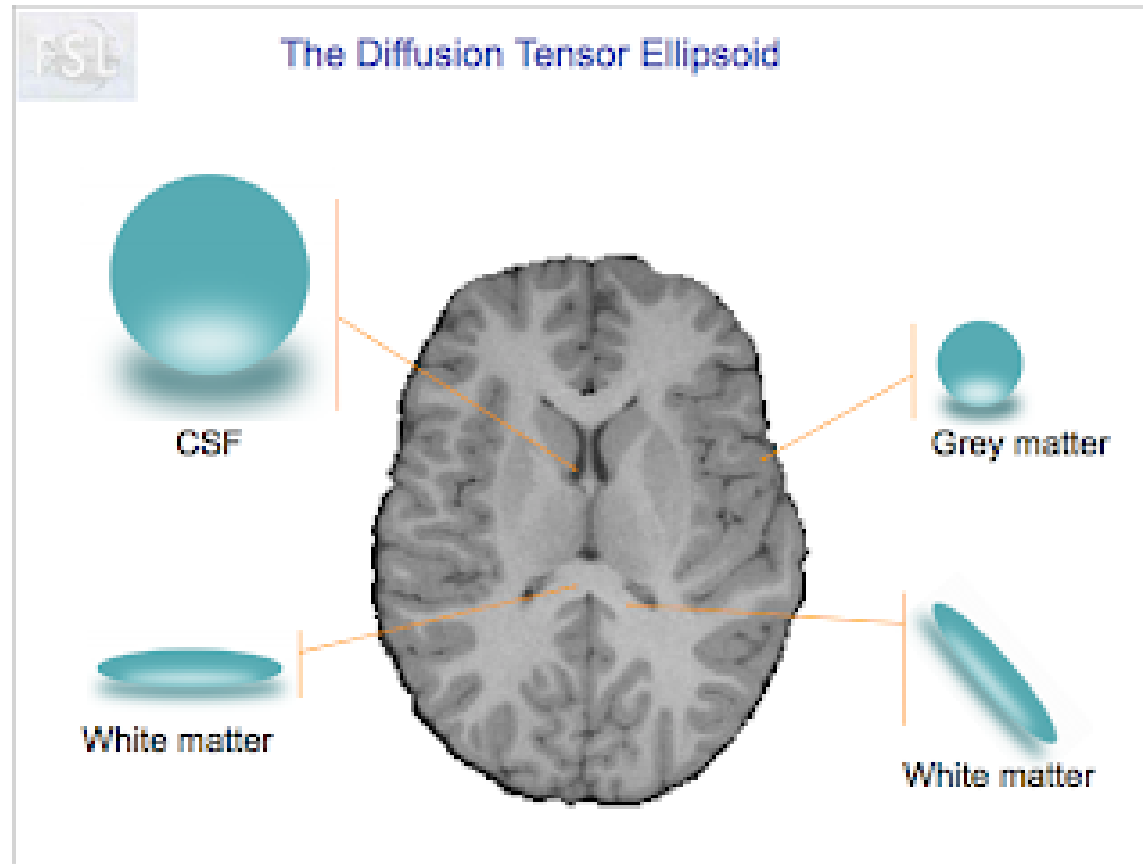


Dr. Alessandro Crimi

Overview



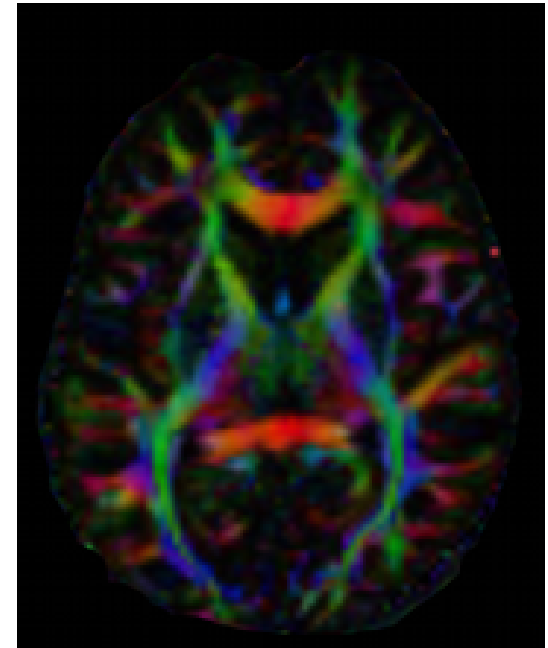
Axons is what we want



$$\underline{D} = \begin{bmatrix} \text{axial} & \text{axial} & \text{axial} \\ \text{sagittal} & \text{sagittal} & \text{sagittal} \\ \text{coronal} & \text{coronal} & \text{coronal} \end{bmatrix} = \begin{bmatrix} D_{xx} & D_{xy} & D_{xz} \\ D_{yx} & D_{yy} & D_{yz} \\ D_{zx} & D_{zy} & D_{zz} \end{bmatrix} \rightarrow \text{Ellipsoid}$$

Diffusion files

- 1) A dwidata file.
- 2) A bvals files contains a scalar value for each applied gradient, corresponding to the respective b-value.
- 3) A bvecs contains a 3x1 vector for each gradient, indicating the gradient direction. The entries in bvals and bvecs are as many as the number of volumes in the dwidata file. The i th volume in the data corresponds to a measurement obtained after applying a diffusion-sensitising gradient with a b-value given by the i th entry in bvals and a gradient direction given by the i th vector in bvecs.



Jupyter Notebooks

First, install Ipython:

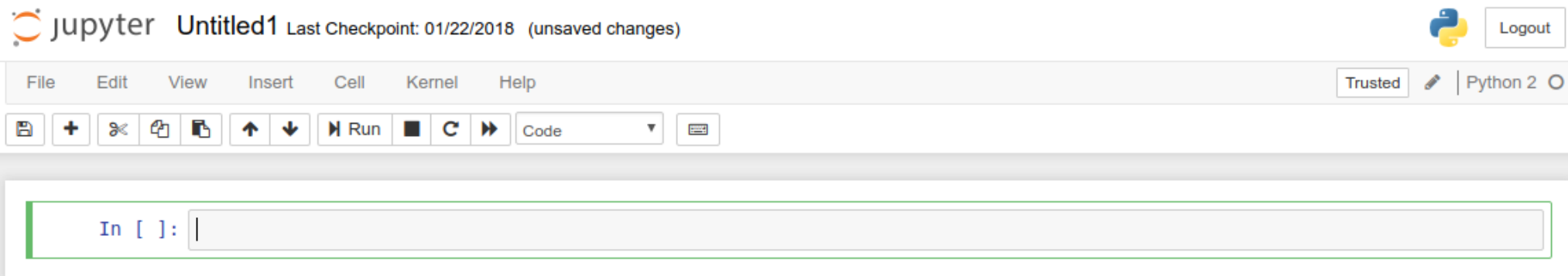
```
sudo apt-get -y install ipython ipython-notebook
```

Now we can move on to installing Jupyter Notebook:

```
sudo -H pip install jupyter
```

Running Jupyter Notebook

```
jupyter notebook
```



Load data

```
import dipy
import numpy as np
import nibabel as nib

fimg = "dti_fm_bet.nii.gz"
img = nib.load(fimg)
data = img.get_data()
affine = img.get_affine()
header = img.get_header()
voxel_size = header.get_zooms()[:3]
mask, S0_mask = median_otsu(data[:, :, :, 0])
fbval = "../.../bvals"
fbvec = "../.../bvecs"

bvals, bvecs = read_bvals_bvecs(fbval, fbvec)
gtab = gradient_table(bvals, bvecs)
```

We use the fractional anisotropy (FA) of the DTI model to build a tissue classifier.

-
- `ten_model = TensorModel(gtab)`
- `ten_fit = ten_model.fit(data, mask)`
-
- `fa = fractional_anisotropy(ten_fit.evals)`
- `cfa = color_fa(fa, ten_fit.evecs)`
-

Constant Solid Angle

The first thing we need to begin fiber tracking is a way of getting directions from this diffusion data set. In order to do that, we can fit the data to a Constant Solid Angle ODF Model. This model will estimate the Orientation Distribution Function (ODF) at each voxel. The ODF is the distribution of water diffusion as a function of direction. The peaks of an ODF are good estimates for the orientation of tract segments at a point in the image.

- `csamodel = CsaOdfModel(gtab, 6)`
- `sphere = get_sphere('symmetric724')`
- `sphere = get_sphere('symmetric724')`
-
- `pmd = peaks_from_model(model=csamodel,`
- `data=data,`
- `sphere=sphere,`
- `relative_peak_threshold=.5,`
- `min_separation_angle=25,`
- `mask=mask,`
- `return_odf=False)`
-

Perform Tractography

- `#Deterministic tractography`
- `eu = EuDX(a=fa, ind=pmd.peak_indices[..., 0], seeds=2000000, odf_vertices=sphere.vertices, a_low=0.01)`
- `affine = eu.affine`
- `csd_streamlines= list(eu)`
-
- `#Remove tracts shorter than 30mm`
- `#print np.shape(csd_streamlines)`
- `from dipy.tracking.utils import length`
- `csd_streamlines=[t for t in csd_streamlines if length(t)>30]`
-

Save as a trackvis

- `#Trackvis`
- `hdr = nib.trackvis.empty_header()`
- `hdr['voxel_size'] = img.get_header().get_zooms()[:3]`
- `hdr['voxel_order'] = 'LAS'`
- `hdr['dim'] = fa.shape`
- `tensor_streamlines_trk = ((sl, None, None) for sl in csd_streamlines)`
- `ten_sl_fname = 'tensor_streamlines.trk'`
- `nib.trackvis.write(ten_sl_fname, tensor_streamlines_trk, hdr, points_space='voxel')`
-

Trackvis

