Diffusion MRI: Principles and Applications

Outline

Basics of diffusion

- What is "diffusion" and why it is important
- Diffusion MRI in a nutshell

Main applications

- Characterization of *tissue properties*
- Estimation of tissue structure
- Reconstruction of *neuronal fiber tracts*
- Estimation of structural brain connectivity

Intro to local reconstruction techniques

- Diffusion Spectrum Imaging (DSI)
- Diffusion Tensor Imaging (DTI)
- ► Q-BALL Imaging (QBI)
- Constrained Spherical Deconvolution (CSD)

What is diffusion?

Random movement of molecules due to thermal agitation from regions of high to regions of low concentration

EXAMPLE: in a glass of water, molecules diffuse randomly and freely, only constrained by the boundaries of the container



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First noted by Robert Brown in 1828

"...random motion without any apparent cause..."

Formally described by **Albert Einstein** in 1905



What happens in the brain?

Rough **brain anatomy**:

- ► *gray-matter* : neuronal cell bodies
- white-matter : mainly myelinated tracts
- ► Cerebrospinal fluid (CSF)



Metaphor: "connectivity" as "water supply network"





Cerebrospinal Fluid (CSF)



- ► Displacements are *isotropic*
- Variance depends on the fluid's properties



Neuronal tracts (white-matter)





- Diffusion more *restricted* perpendicular to the tracts
- Degree of restriction depends on tissue properties



Diffusion MRI in a nutshell

MRI sequences are sensitive to diffusion by inserting two additional magnetic field gradient pulses

- ► The goal is to **change phase** of moving molecules
- Movement
 phase differences
 signal cancels out/drops



Diffusion MRI in a nutshell

MRI sequences are **sensitive to diffusion** by inserting two additional *magnetic field gradient pulses*

- ► The goal is to **change phase** of moving molecules
- Movement \Rightarrow phase differences \Rightarrow signal cancels out/drops

signal $\propto e^{-bD}$

Signal decays as:

- ► D : diffusion coefficient of the tissue
- \blacktriangleright b : diffusion weighting/contrast of the images (δ , Δ , $|\mathbf{G}|$)

NOTE: signal <u>strongly depends</u> on

- ▶ the *b*-value (b)
- ► the diffusion coefficient (D)
- ► the gradient direction (G)







increasing /





(1/3)

Scalar maps

Estimate *local features* of the tissue



diffusion coefficient



diffusion anisotropy



axonal density



axonal dispersion

Intra-voxel fiber structure

Estimate the *number and orientation of fiber populations* in each voxel







Fiber-tracking

▶ Infer *axonal trajectories* by exploiting the diffusion information in each voxel



Connectivity analysis

► In-vivo and non-invasive assessment of structural wiring of the brain



fiber-tracking



cortical segmentation





connectivity matrix











EAP, ODF and fODF

EAP (Ensemble Average Propagator)

- In each voxel, 3D PDF giving the *probability of water displacements* → diffusion MRI is a 6D modality
- ► Related to the *signal attenuation* by a **3D FFT**:

$$\mathbf{P}(\vec{r}) = \int_{\mathbb{R}^3} E(\vec{q}) e^{-2\pi i \vec{q} \cdot \vec{r}} d\vec{q}$$
q-space



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ODF Or **dODF** (Orientation Distribution Function)

Probability of diffusion along a given direction:

$$\text{ODF}(\hat{r}) = \int_{\mathbb{R}_+} P(r, \hat{r}) \, r^2 \, dr$$

Function on the sphere

$$\int_{\mathbb{R}_+} P(r,\hat{r}) \, r^2 \, dr$$



FOD Or **fODF** (fiber ODF)

- Probability of having a fiber population along a given direction
- Function on the sphere





Diffusion SPECTRUM Imaging (DSI)

Exploits the **3D Fourier Transform relationship** between the *MR signal* and the *displacement distribution* (EAP)

 $\mathbf{P}(\vec{r}) = \int_{\mathbb{R}^3} E(\vec{q}) e^{-2\pi i \vec{q} \cdot \vec{r}} d\vec{q}$

SIGNAL E(q)



The q-space must be properly sampled

- Data must be sampled in a dense 3D cartesian grid
- Usual protocol: 515 samples with $b_{max} \approx 8000 \text{ s/mm}^2$





Measures directly water displacements making "almost" no assumptions

"almost" = short pulse condition is required



Advantages:

- Model free
- Complex fiber configurations recovered
- Recovers the EAP (even though radial information is usually ignored!!!)

Limitations:

- Long acquisitions (≈ 30-40 min)
- Motion sensitive
- ► High b-values + long TE ➡ low SNR
- Short pulses never met = smooth EAP
- No useful maps from the EAP
- Severe truncation artifacts
 - Inherent to FFT and "relatively low" b-values
 - Hanning filter mitigates but introduces blurring



0.005

8000



0.2



(2/2)

Diffusion TENSOR Imaging

Assumption: displacements of water molecules follow a multivariate gaussian distribution

- Process fully characterized by its covariance matrix (3x3 symmetric positive semi-definite matrix)
- Usually represented as an ellipsoid
- 6 degrees of freedom (3 rotations + 3 variances)



estimated acquiring 6+1 DWI images (at least)

represents one 3D DWI image

• Usual protocol: 6-32 directions with $b \approx 1000 \text{ s/mm}^2$





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Diffusion TENSOR Imaging

Advantages:

- Fast acquisitions (≈ 4-5 min) ⇒ clinically feasible
- Does not require special hardware
- ► Useful *scalar maps*, e.g.
 - Mean Diffusivity (MD) :

$$\bar{\lambda} = \frac{\lambda_1 + \lambda_2 + \lambda_3}{3}$$

- Fractional Anisotropy (FA) :

$$FA = \sqrt{\frac{3}{2}} \sqrt{\frac{(\lambda_1 - \bar{\lambda})^2 + (\lambda_2 - \bar{\lambda})^2 + (\lambda_3 - \bar{\lambda})^2}{\lambda_1^2 + \lambda_2^2 + \lambda_3^2}}$$

Limitation(s):

Biomedical Image Processing

- Complex fiber configurations cannot be modeled
 - majority of voxels in the brain
 - Enough for characterization of major bundles, but inadequate for whole brain connectivity analyses
 - NB: acquiring more data does overcome this limitation!



Mean Diffusivity: high values = fast diffusion

Fractional Anisotropy: high values = fiber bundles



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HARDI: overview

High angular resolution diffusion imaging (HARDI):

- ► Focus on *angular information*:
 - The radial component is discarded/averaged
 - No access to tissue micro-structural features, e.g. axonal diameter and density
- Usually based on *spherical sampling* (at least 60 samples in q-space)

Vast literature of methods:

- Multi-Tensor fitting (Tuch et al, 2002)
- ► Q-BALL (QBI) (Tuch, 2004)
- Q-BALL in Constant Solid Angle (QBI_{CSA}) (Aganj et al, 2010)
- Constrained Spherical Deconvolution (CSD) (Tournier et al, 2004)
- Diffusion Orientation Transform (DOT) (Ozarslan et al, 2006)

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IDEA: data is acquired on a single shell and the ODF is approximated by means of Funk-Radon Transform (FRT)



Samples distributed on a single-shell Same idea behind CT, but on the sphere

ODF are smooth



Notes:

- Model free
- ► FRT induces blurring
- ODF is only approximated (missing r² term in the integral)
- Usual protocol: >=60 directions with $b \approx 3000 \text{ s/mm}^2$



HARDI: constrained spherical deconvolution (3/3)

Assumption: ODF can be seen as a convolution on the sphere

ODF kernel FOD The kernel characterizes the diffusion response-function of a single fiber; can be estimated from the data:

- Identify known areas with only one fiber population
- ► Fit a tensor in each voxel and average

Notes:

- Model based
- ► High angular accuracy (i.e. sharper profiles), but sensitive to noise
- ► Assumes the *same diffusion properties* across the whole brain
- ▶ Usual protocol: <=60 directions with b ≈ 3000 s/mm²





