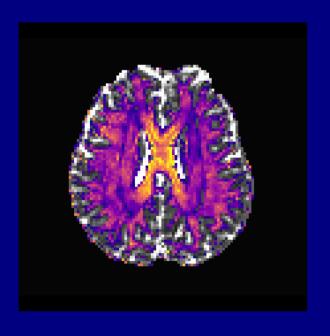
Introduction to: DWI + DTI

AFNI Bootcamp (SSCC, NIMH, NIH)







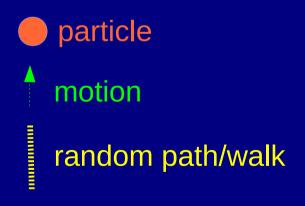
Outline

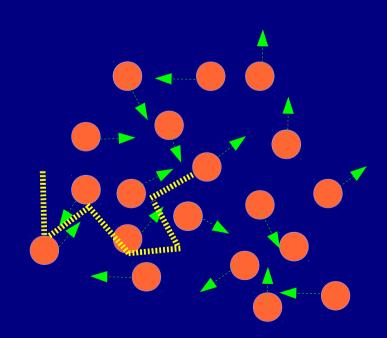
- + DWI and DTI
 - Concepts behind diffusion imaging
 - Diffusion imaging basics in brief
 - Connecting DTI parameters and geometry
 - Role of noise+distortion → DTI parameter uncertainty

DTI is a particular kind of magnetic resonance imaging (MRI)

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Diffusion: random motion of particles, tending to spread out \rightarrow here, hydrogen atoms in aqueous brain tissue





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Tensor: a mathematical object (a matrix) to store information \rightarrow here, quantifying particle spread in all directions

$$D = \begin{pmatrix} D_{11} & D_{12} & D_{13} \\ D_{21} & D_{22} & D_{23} \\ D_{31} & D_{32} & D_{33} \end{pmatrix}$$

DTI is a particular kind of magnetic resonance imaging (MRI)

Diffusion: random motion of particles, tending to spread out

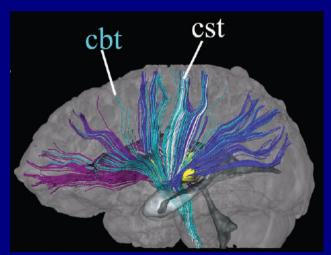
→ here, hydrogen atoms in aqueous brain tissue

Tensor: a mathematical object (a matrix) to store information

→ here, quantifying particle spread in all directions

Imaging: quantifying brain properties

→ here, esp. for white matter



The DTI model: Assumptions and relation to WM properties

Diffusion: random (Brownian) motion of particles → mixing or spreading





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Empty cup, no structure:
Atoms have equal probability of movement any direction

→ spherical spread of concentration

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But in the presence of structures:

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Ex: unstirred, steeping tea (in a large cup):



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But in the presence of structures: Unequal probabilities of moving in different directions

→ *non*spherical spread

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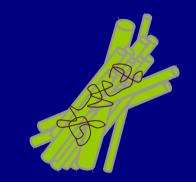
But in the presence of structures: Unequal probabilities of moving in different directions

→ *non*spherical spread

→ Diffusion shape tells of structure presence and spatial orientation

(In brief)

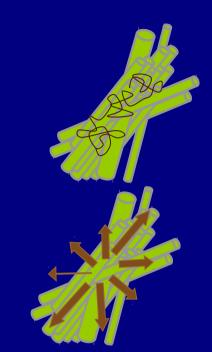
1) Random motion of molecules affected by local structures



(In brief)

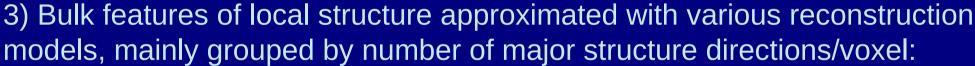
1) Random motion of molecules affected by local structures

2) Statistical motion measured using diffusion weighted MRI

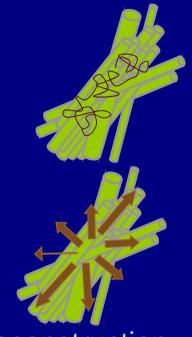


(In brief)

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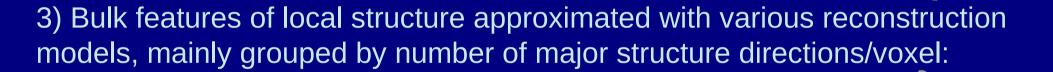


+ one direction:DTI (Diffusion Tensor Imaging)



(In brief)

- 1) Random motion of molecules affected by local structures
- 2) Statistical motion measured using diffusion weighted MRI



- + one direction:DTI (Diffusion Tensor Imaging)
- + >=1 direction:
 HARDI (High Angular Resolution Diffusion Imaging)
 Qball, DSI, ODFs, ball-and-stick, multi-tensor, CSD, ...

Diffusion in MRI

Mathematical properties of the matrix/tensor:

$$\mathbf{D} = \begin{pmatrix} D_{11} & D_{12} & D_{13} \\ D_{21} & D_{22} & D_{23} \\ D_{31} & D_{32} & D_{33} \end{pmatrix}$$

Having: 3 eigenvectors: \mathbf{e}_i 3 eigenvalues: λ_i

- Real-valued
- Positive definite $(\mathbf{r}^{\mathsf{T}}\mathbf{Dr} > 0)$

$$\mathbf{D}\mathbf{e}_{i} = \lambda \lambda_{i} \mathbf{e}_{i}, \quad \lambda_{i} > 0$$

- Symmetric ($D_{12} = D_{21}$, etc) 6 independent values

Diffusion in MRI

Mathematical properties of the matrix/tensor:



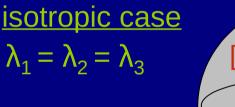
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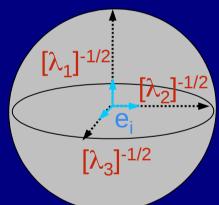
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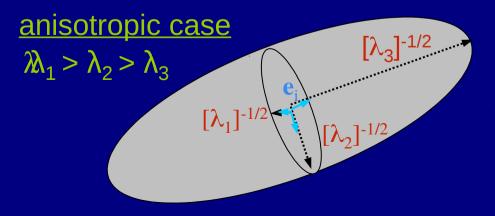
- Real-valued
- Positive definite $(\mathbf{r}^T \mathbf{D} \mathbf{r} > 0)$ $\mathbf{D} \mathbf{e}_i = \lambda \lambda_i \mathbf{e}_i, \quad \lambda_i > 0$
- Symmetric ($D_{12} = D_{21}$, etc) 6 independent values

Geometrically, this describes an ellipsoid surface:

$$C = D_{11}x^2 + D_{22}y^2 + D_{33}z^2 + 2(D_{12}xy + D_{13}xz + D_{23}yz)$$



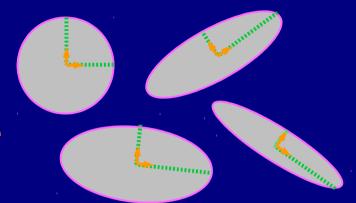




DTI: ellipsoids

Important mathematical properties of the diffusion tensor:

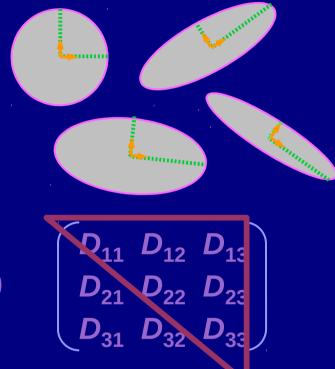
+ Help to picture diffusion model:
 tensor D → ellipsoid surface
 eigenvectors e_i → orientation in space
 eigenvalues λ_i → 'pointiness' + 'size'



DTI: ellipsoids

Important mathematical properties of the diffusion tensor:

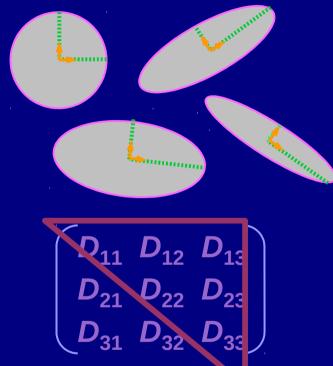
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- + Determine the minimum number of DWIs measures needed (6 + baseline)



DTI: ellipsoids

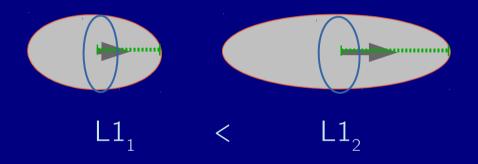
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- + Help to picture diffusion model:
 tensor D → ellipsoid surface
 eigenvectors e_i → orientation in space
 eigenvalues λ_i → 'pointiness' + 'size'
- + Determine the minimum number of DWIs measures needed (6 + baseline)
- + Determine much of the processing and noise minimization steps



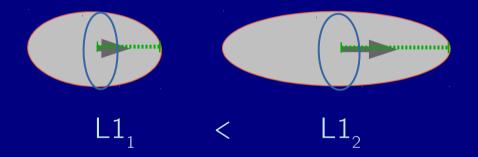
Main quantities of diffusion (motion) surface

first eigenvalue, L1 $(= \lambda_1$, parallel/axial diffusivity, AD)

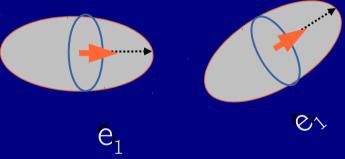


Main quantities of diffusion (motion) surface

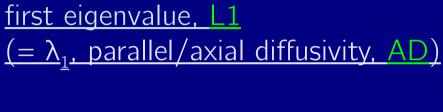
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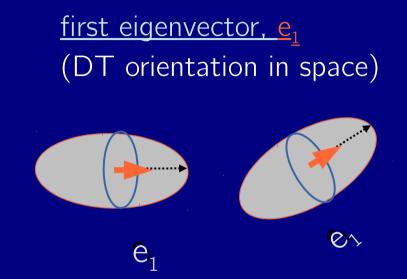
first eigenvector, e₁
(DT orientation in space)



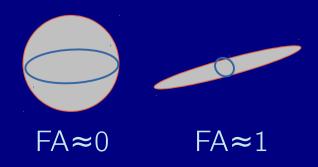
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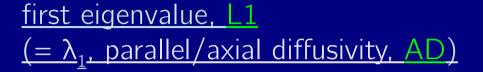


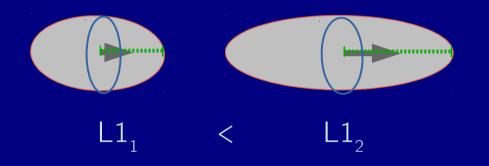


Fractional anisotropy, FA (stdev of eigenvalues)

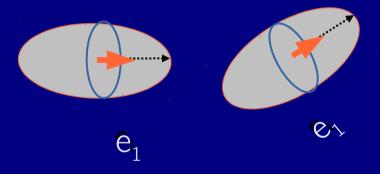


Main quantities of diffusion (motion) surface

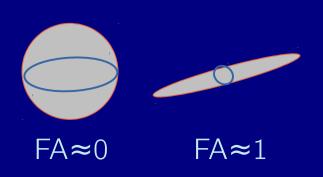




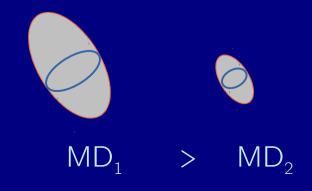
first eigenvector, e₁ (DT orientation in space)



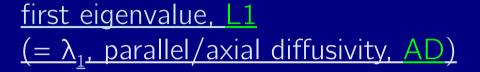
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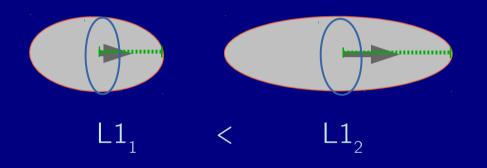


Mean diffusivity, MD (mean of eigenvalues)

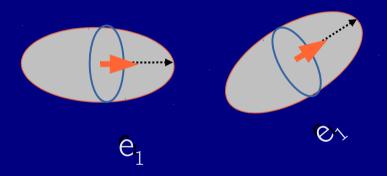


Main quantities of diffusion (motion) surface

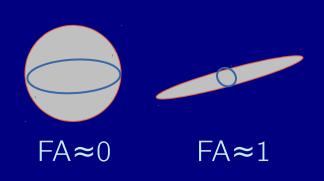




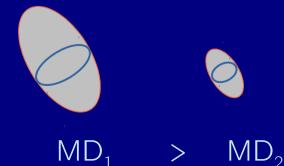
first eigenvector, e₁
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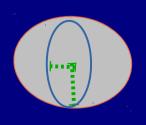
Fractional anisotropy, FA (stdev of eigenvalues)

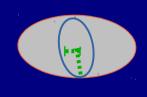


Mean diffusivity, MD (mean of eigenvalues)



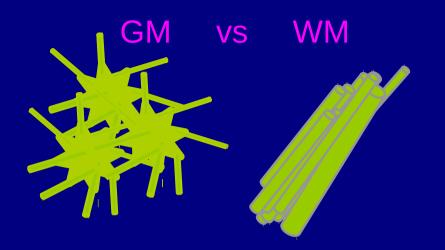
Radial diffusivity, RD $(= (\lambda_2 + \lambda_3)/2)$

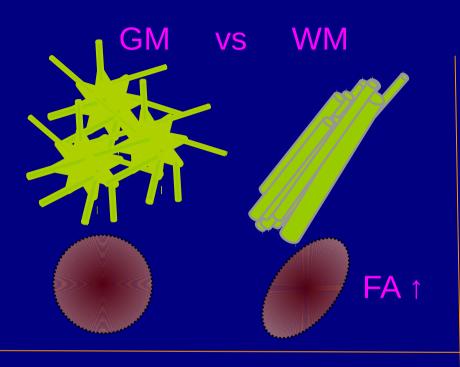


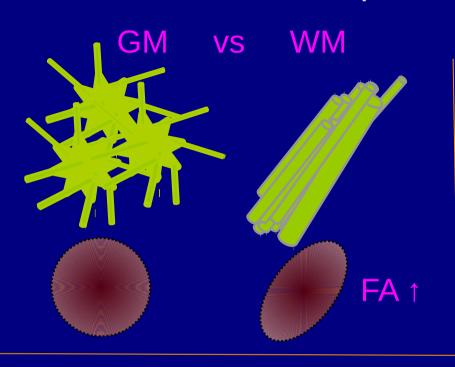


 RD_1

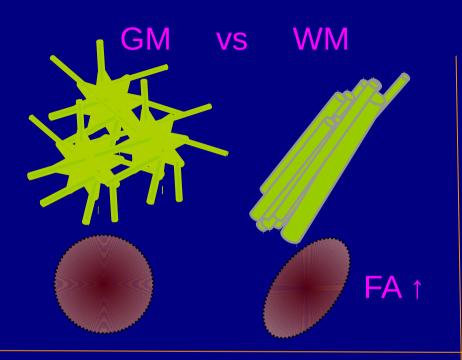
 $> RD_2$



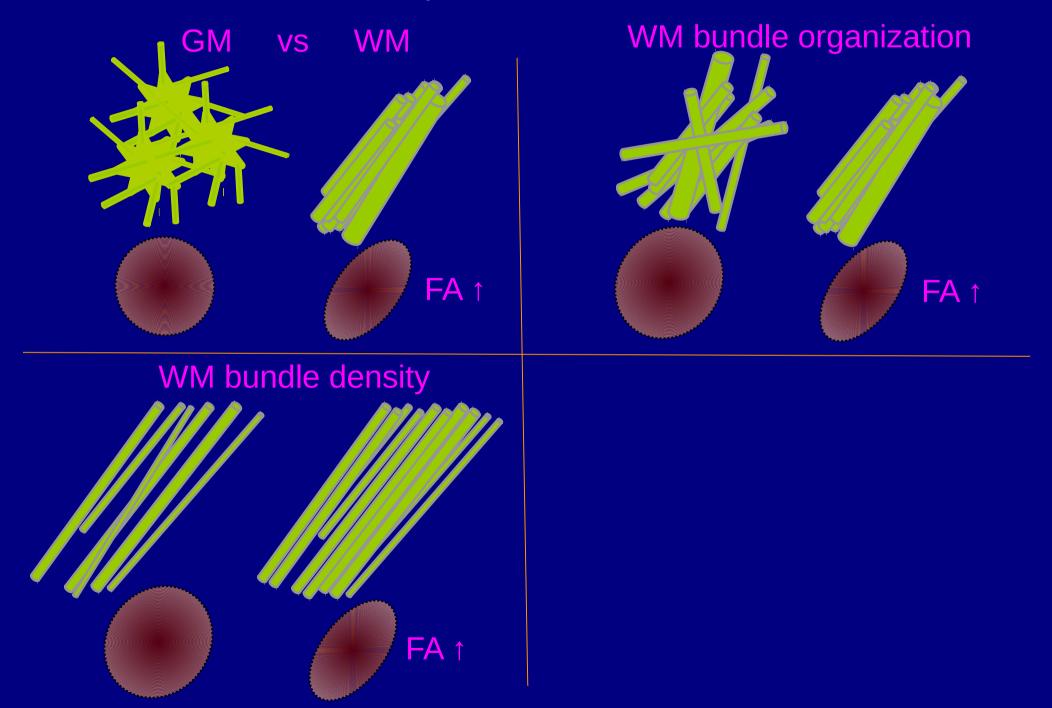


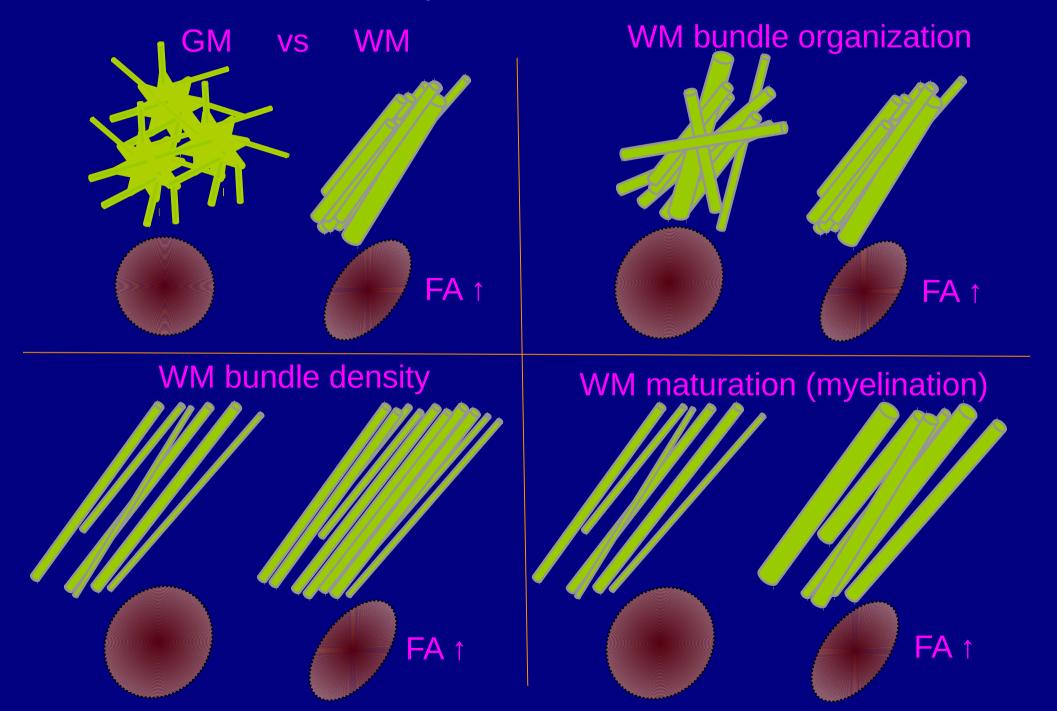












Interpreting DTI parameters

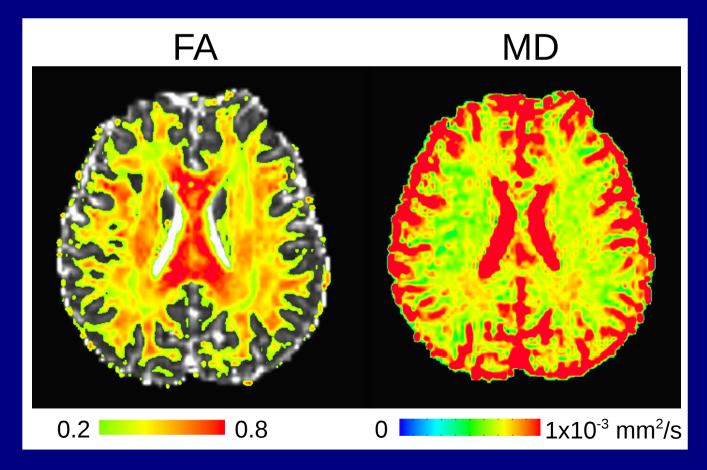
General literature:

FA: measure of fiber bundle coherence and myelination

- in adults, FA>0.2 is proxy for WM

MD, L1, RD: local density of structure

e₁: orientation of major bundles



Interpreting DTI parameters

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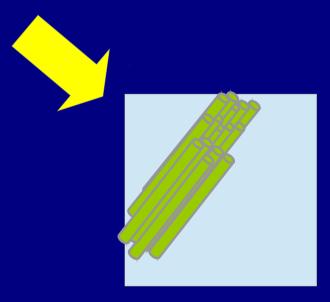
Cautionary notes:

- Degeneracies of structural interpretations
- Changes in myelination may have small effects on FA
- WM bundle diameter << voxel size
 - don't know location/multiplicity of underlying structures
- More to diffusion than structure-- e.g., fluid properties
- Noise, distortions, etc. in measures

Acquiring DTI data: diffusion weighted gradients in MRI

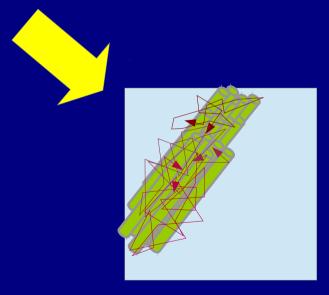
For a given voxel, observe relative diffusion along a given 3D spatial orientation (gradient)

DW gradient
$$\mathbf{g}_{i} = (g_{x'}, g_{y'}, g_{z})$$



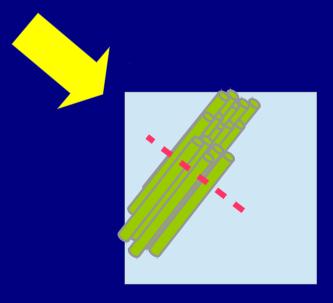
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DW gradient
$$\mathbf{g}_{i} = (g_{x'}, g_{y'}, g_{z})$$



MR signal is attenuated by diffusion throughout the voxel in that direction:

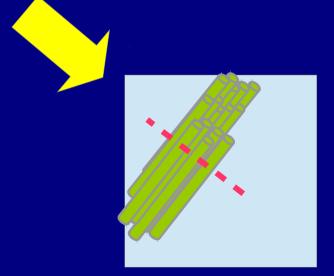
$$S_i = S_0 e^{-b g_i T D g_i}$$

→ ellipsoid equation of diffusion surface:

$$C = \mathbf{r}^T \mathbf{D}^{-1} \mathbf{r}$$
.

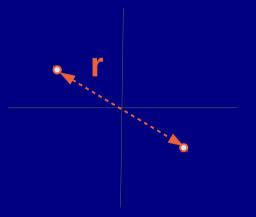
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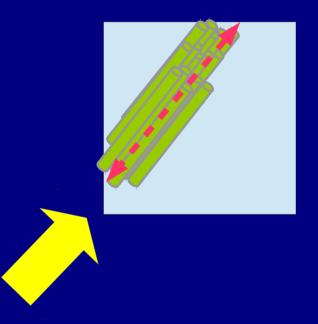
diffusion motion ellipsoid:

$$C_2 = \mathbf{r}^\mathsf{T} \mathbf{D}^{-1} \mathbf{r}$$
.



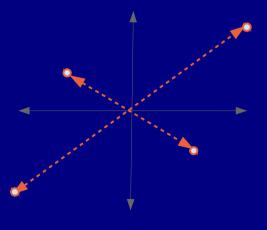
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.

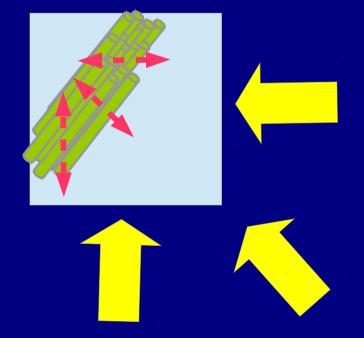


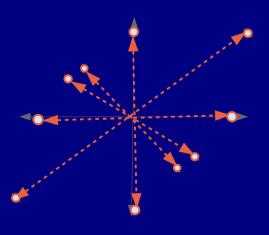
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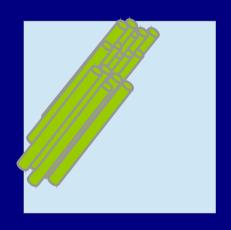
$$C_2 = r^T D^{-1} r$$
.





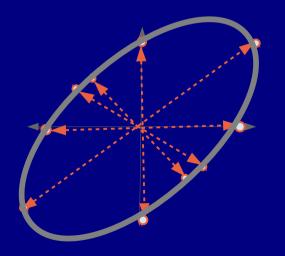
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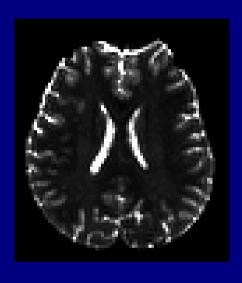


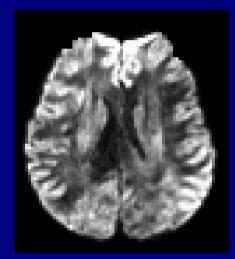
Individual points → Fit ellipsoid surface Individual signals → Solve for **D**

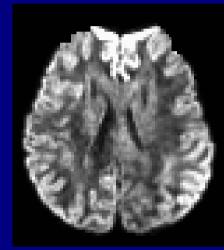
Sidenote: what DWIs look like

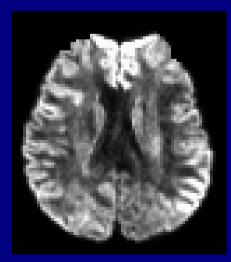
Unweighted reference b=0 s/mm²

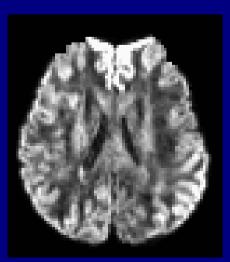
Diffusion weighted images (example: b=1000 s/mm²)

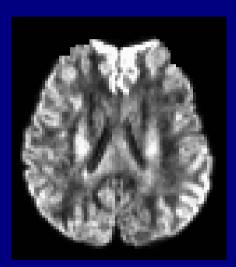


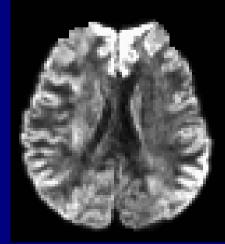








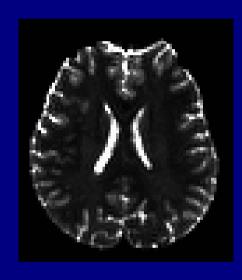


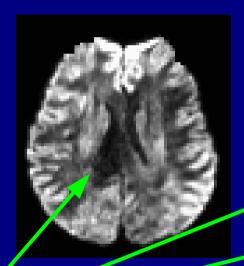


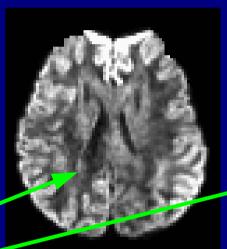
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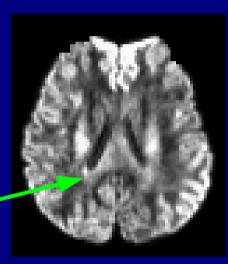
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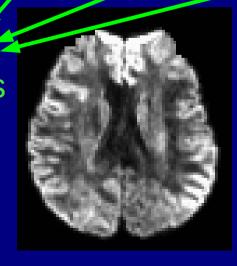


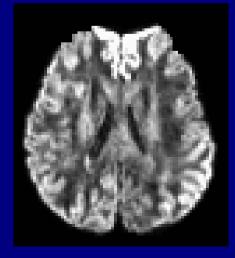


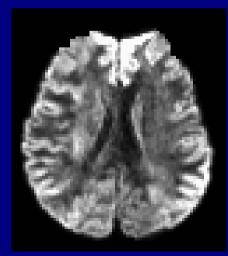




(Each DWI has a different brightness pattern: viewing structures from different angles.)







Noise in DW signals

MRI signals have additive noise

$$S_i = S_0 e^{-b g_i^T D g_i} + \varepsilon,$$

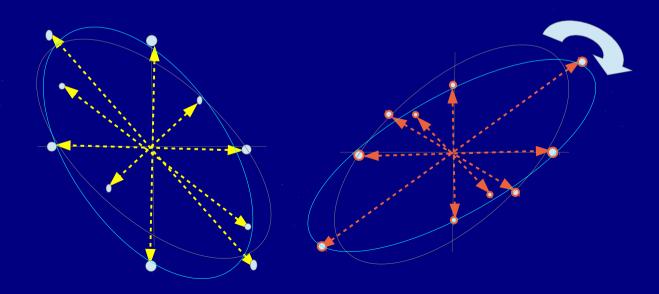
where ϵ is (Rician) noise.

Noise in DW signals

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$$S_i = S_0 e^{-b g_i^T \mathbf{p} g_i} + \varepsilon$$
,
where ε is (Rician) noise.

→ Leads to errors in surface fit, equivalent to rotations and rescalings of ellipsoids:



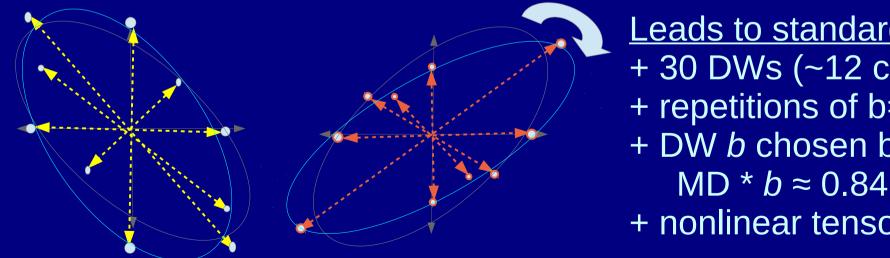
'Un-noisy' vs perturbed/noisy fit

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'Un-noisy' vs perturbed/noisy fit

Leads to standard:

- + 30 DWs (~12 clinical)
- + repetitions of b=0
- + DW b chosen by:
- + nonlinear tensor fitting

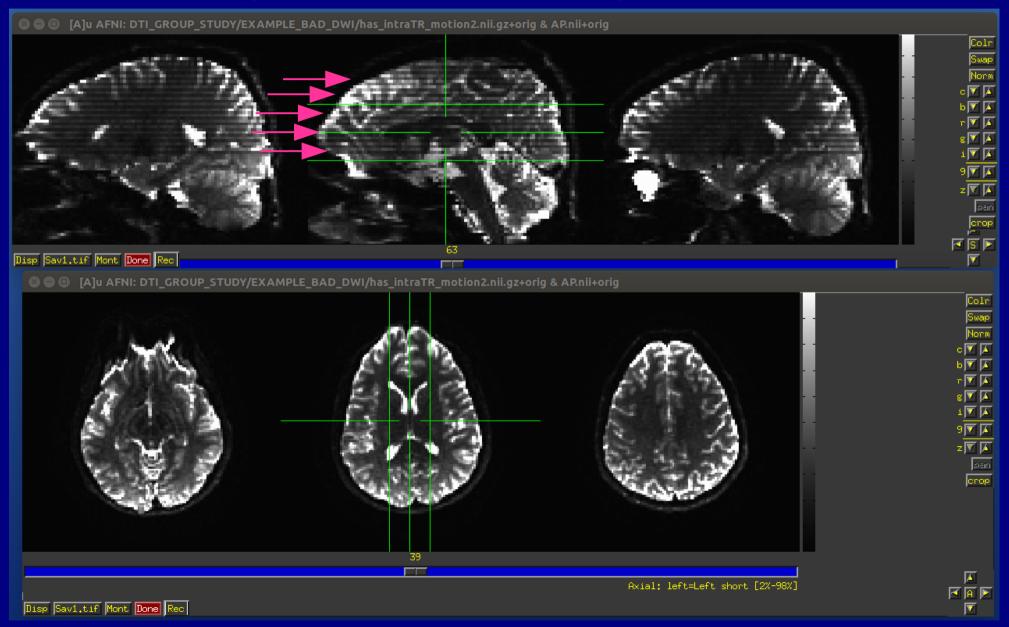
Distortions in DWI volumes

There are also **serious** sources of distortion when acquiring DWIs:

- + Subject motion
 - due to movement during/between volume acq. -> signal loss/overlap
- + Eddy current distortion
 - due to rapid switching of gradients -> nonlinear/geometric distortions
- + EPI distortion
 - due to B0 inhomogeneity -> geometric distortions along phase encoding dir, signal pileup or attenuation
- ---> And effects combine! Need careful acquisition (sometimes perhaps even **re**acquisitions) and post-processing.

Distortions in DWI volumes

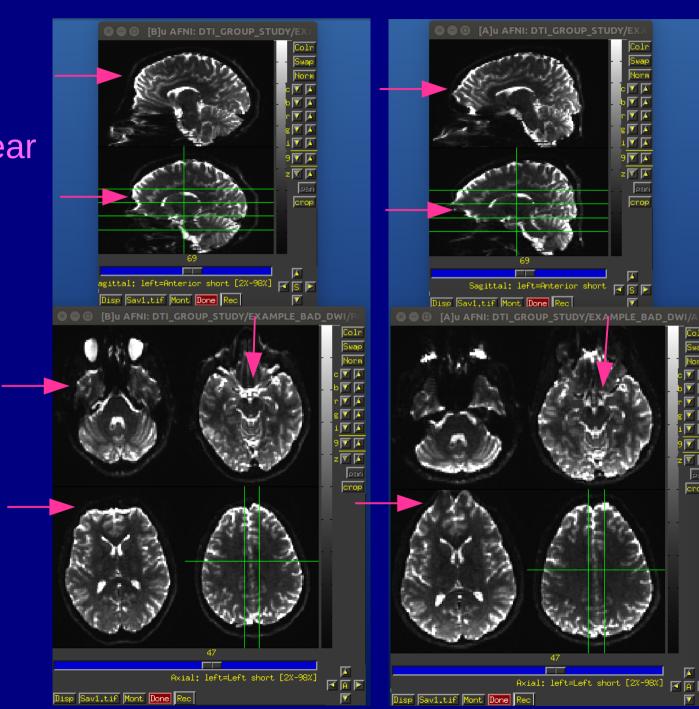
From subj motion: interleaved brightness distortions



Distortions in DWI volumes

From eddy and EPI distortions:

- + geometric/nonlinear warping
- + signal pileup and attenuation



SUMMARY

- + Diffusion-based MRI uses application of magnetic field gradients to probe the relative diffusivity of molecules along different directions.
- + DTI combines that information into a simple shape family, spheroids, to summarize the diffusivity.
- + From the DT, several useful properties are described in terms of scalar (e.g., FA, MD, L1) and vector (e.g., V1) parameters.
- + Many "standard" interpretations of DTI parameters exist (i.e., higher FA = "better" WM), but we must be cautious.
- + Distortions and noise affect all DTI estimates, and we must consider the consequences of these in all analyses.



