

Diffusion Tensor Imaging quality control : artifacts assessment and correction

A. Coste, S. Gouttard, C. Vachet, G. Gerig

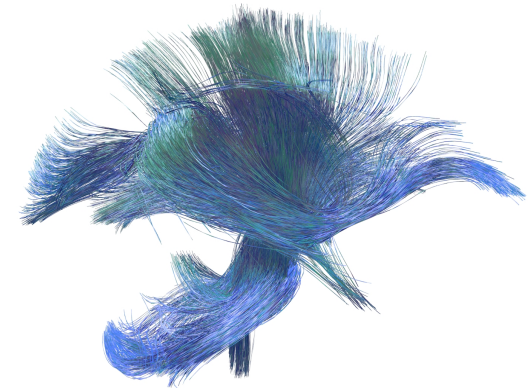


Medical Imaging Seminar

- Introduction
- DWI – DTI
- Artifact Assessment
- Artifact Consequences
- Artifact Detection
- Possible Correction
- Conclusion

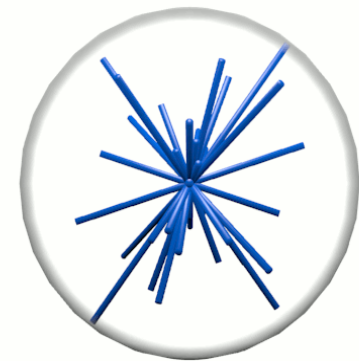
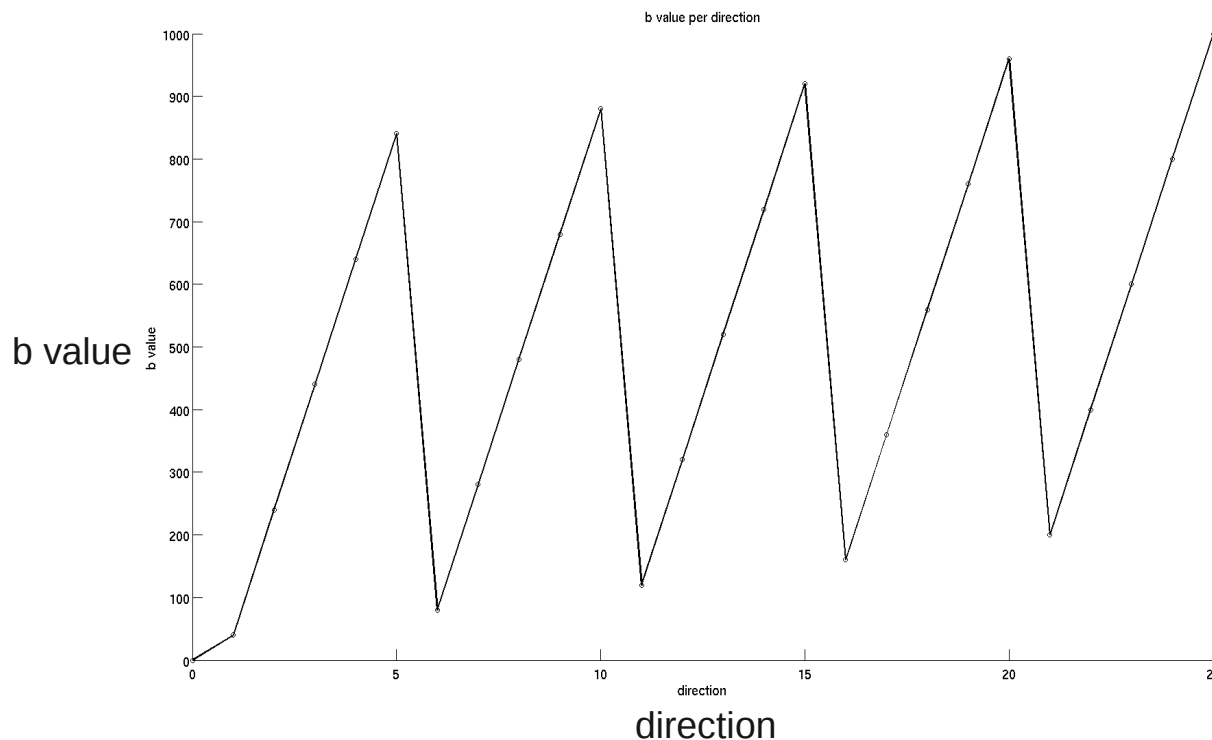
Introduction

- DTI is a powerful technique giving insights into white matter organization
- Allows Tractography
- Increased Artifacts
- Reduced Signal to Noise Ratio
- Need of Quality Control
- Need of new processing methods to process DTI due to new kind of artifacts



Diffusion Weighted Imaging

- Acquisition of multiple set of MRI with several values of the exciting magnetic field and with different orientation.
- In our protocol we use 25 different directions



Diffusion Tensor Imaging

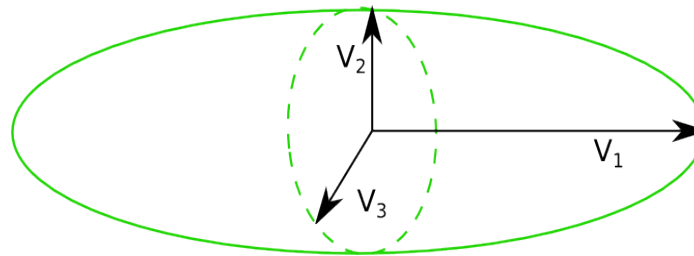
- DTI is computed with an overdetermined system based on all images of the DWI set.
- Each voxel contains a second order diffusion Tensor :

$$D = \begin{bmatrix} D_{XX} & D_{XY} & D_{XZ} \\ D_{YX} & D_{YY} & D_{YZ} \\ D_{ZX} & D_{ZY} & D_{ZZ} \end{bmatrix}$$

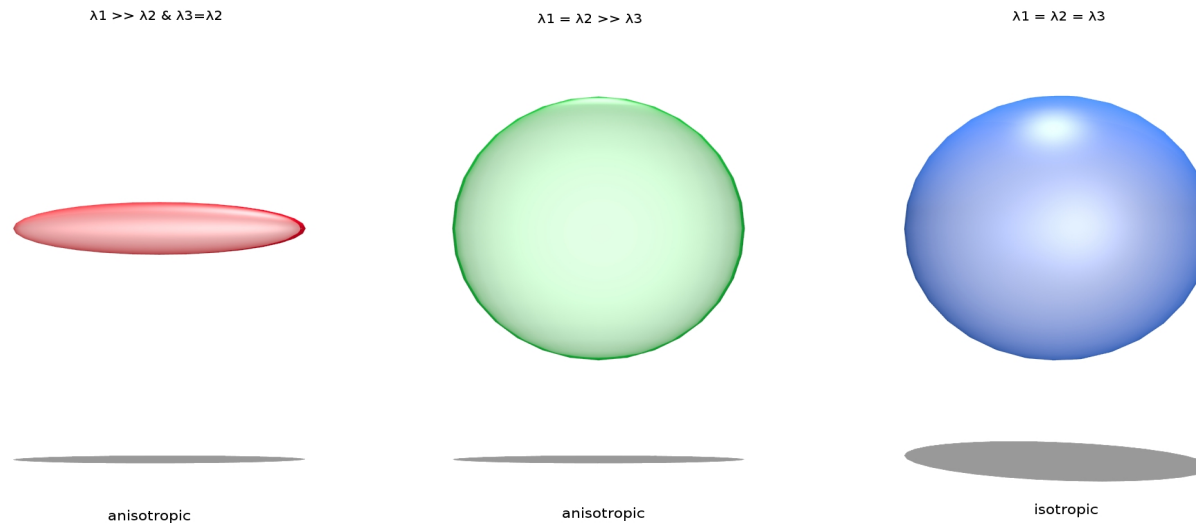
- D is symmetric and positive definite
- To visualize the tensor we use Eigen decomposition to get the local diffusion characteristics

DTI Visualization

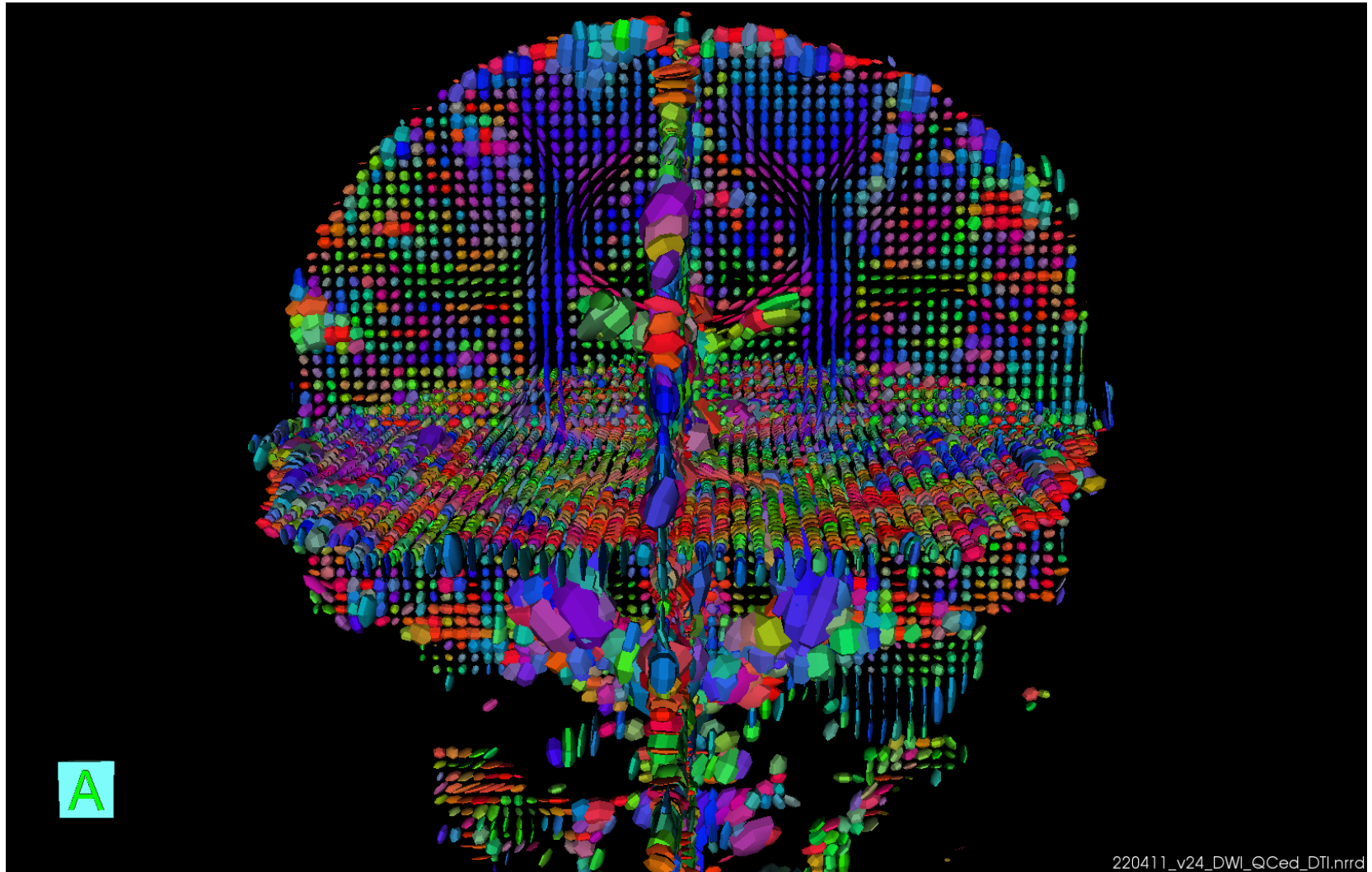
- Ellipsoid parametrization with eigen vectors and eigenvalues



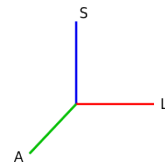
- Interpretation of diffusion properties



DTI Visualization



09/10/12

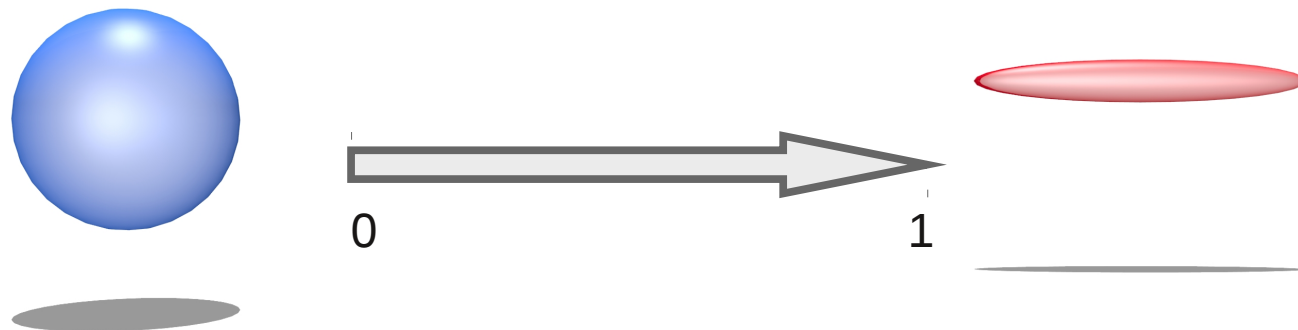


Fractional Anisotropy

$$D = \begin{bmatrix} D_{XX} & D_{XY} & D_{XZ} \\ D_{YX} & D_{YY} & D_{YZ} \\ D_{ZX} & D_{ZY} & D_{ZZ} \end{bmatrix}$$

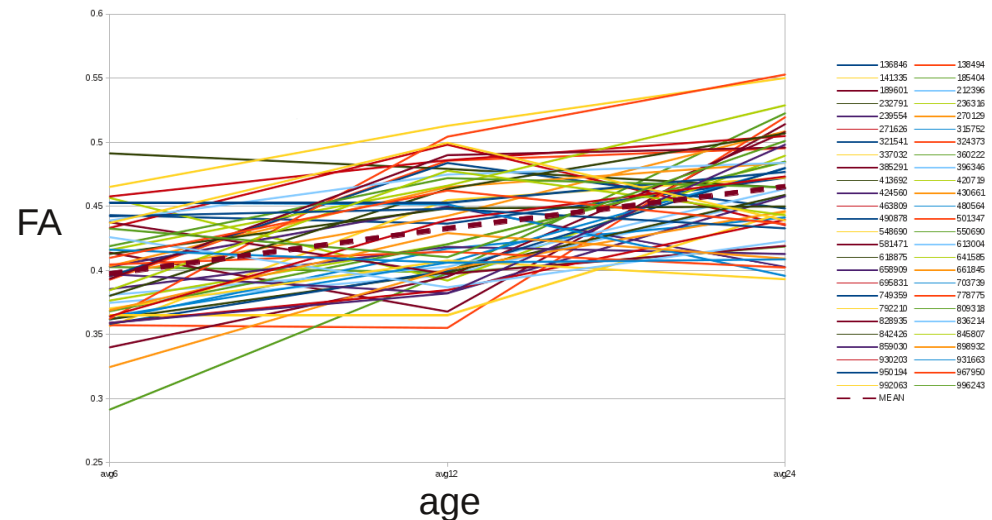
- Fractional Anisotropy (FA)

$$FA = \sqrt{\frac{1}{2} \frac{\sqrt{(\lambda_1 - \lambda_2)^2 (\lambda_1 - \lambda_3)^2 (\lambda_3 - \lambda_1)^2}}{\sqrt{\lambda_1^2 + \lambda_2^2 + \lambda_3^2}}}$$



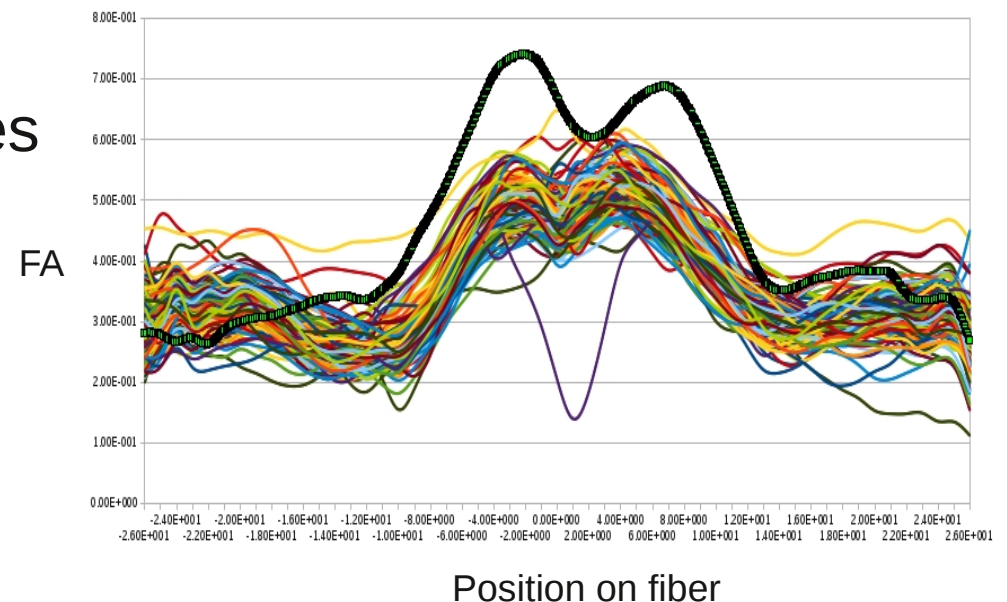
Artifact Assessment

- Quantitative assessment
 - Unexpected evolution of FA over time
 - Diffusion evolution inconsistent
- Qualitative assessment
 - “blurry” FA images
 - Red Color FA images
 - “Red” Tensors



Artifact Assessment

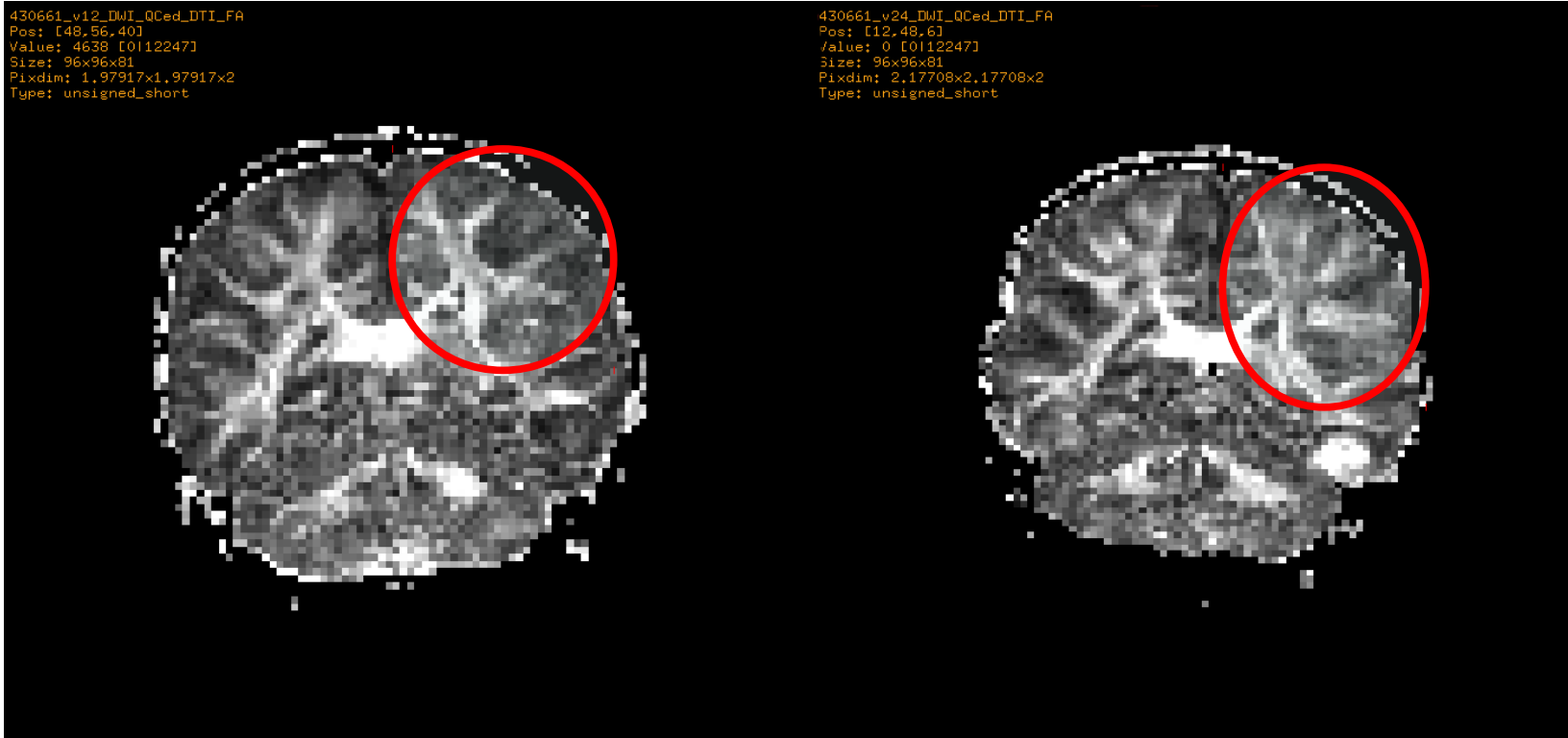
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Artifact Assessment

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$$FA = \sqrt{\frac{1}{2} \frac{\sqrt{(\lambda_1 - \lambda_2)^2 (\lambda_1 - \lambda_3)^2 (\lambda_3 - \lambda_1)^2}}{\sqrt{\lambda_1^2 + \lambda_2^2 + \lambda_3^2}}}$$

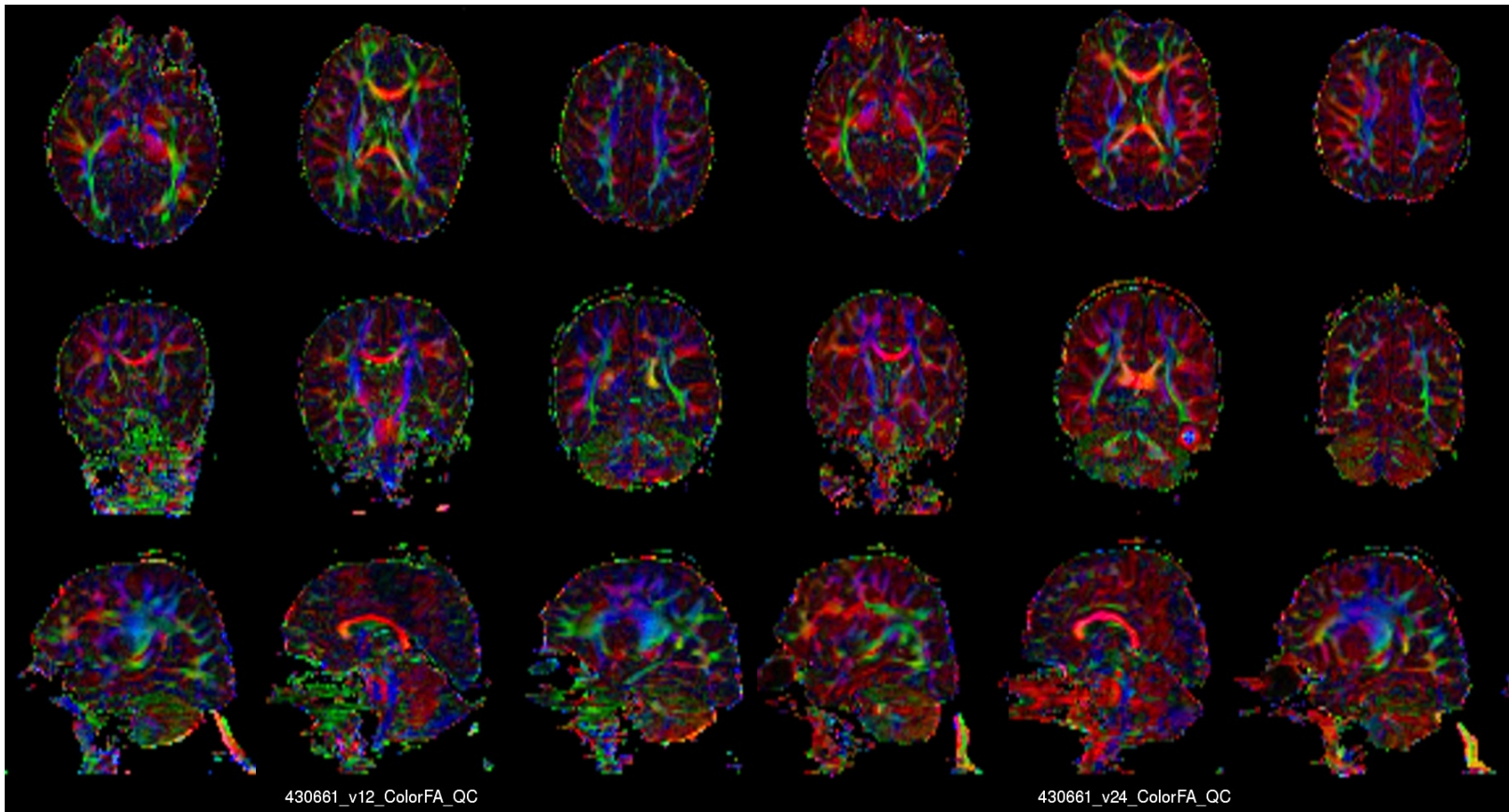
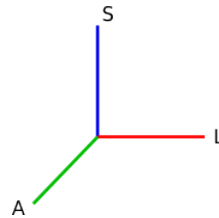


Good case

Bad case

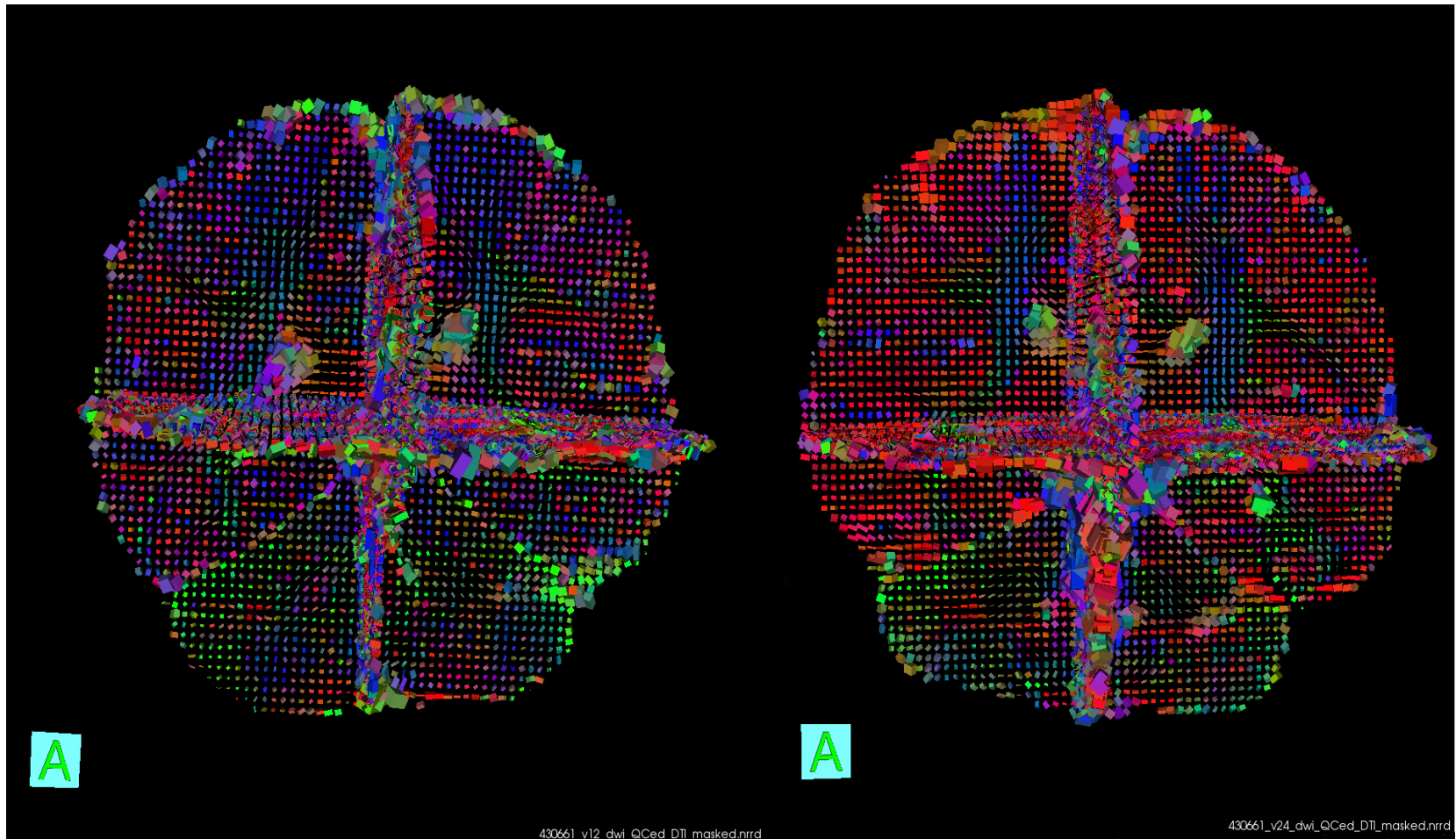
Artifact Assessment

- Red ColorFA images



Artifact Assessment

“Red” Tensors



09/10/12

Good case

Bad case

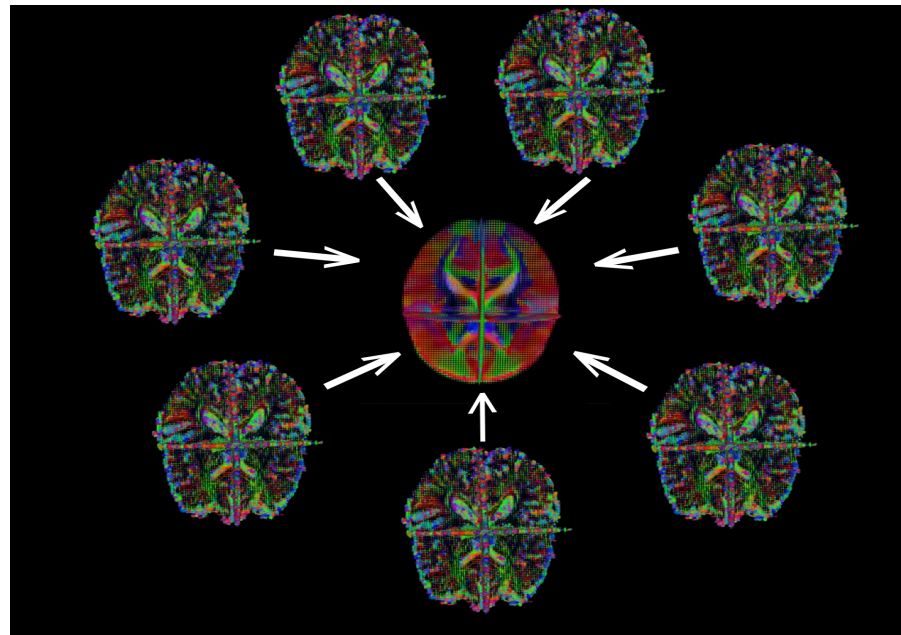
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Artifact Assessment

- Quantitative assessment
 - Unexpected evolution of FA over time
 - Diffusion evolution inconsistent
- Qualitative assessment
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 - Red Color FA images
 - “Red” Tensors
- ⇒ Something is wrong but what ?
 - Processing
 - Acquisition

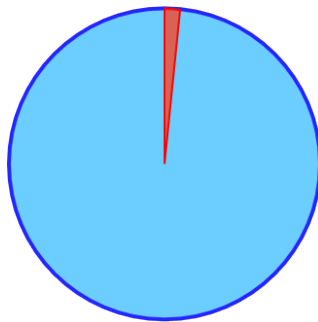
Artifact Consequences

- Wrong calculation of diffusion characteristics
- Poor quality tensors used in atlas building
- Unknown influence of it in atlas building
- Reliability of conclusions based on these data

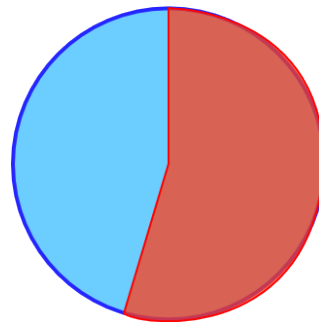


Statistical Assessment

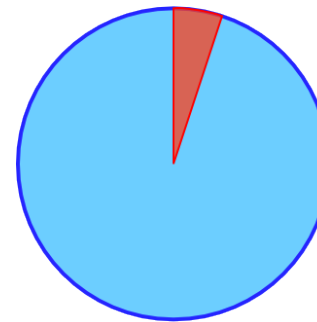
- Our data set is composed of 1200+ DWI sets
- 4 acquisition sites across the US : ACE / IBIS Network
 - University of North Carolina at Chapel Hill
 - University of Washington in Seattle
 - Washington University in St. Louis
 - Children Hospital of Pennsylvania in Philadelphia
- Artifact ratio within the data set :



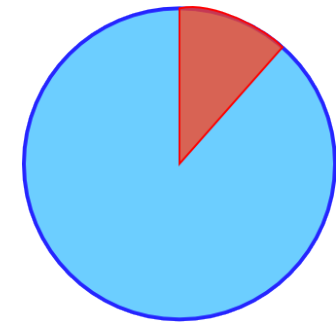
UNC : 1%



SEA : 54%



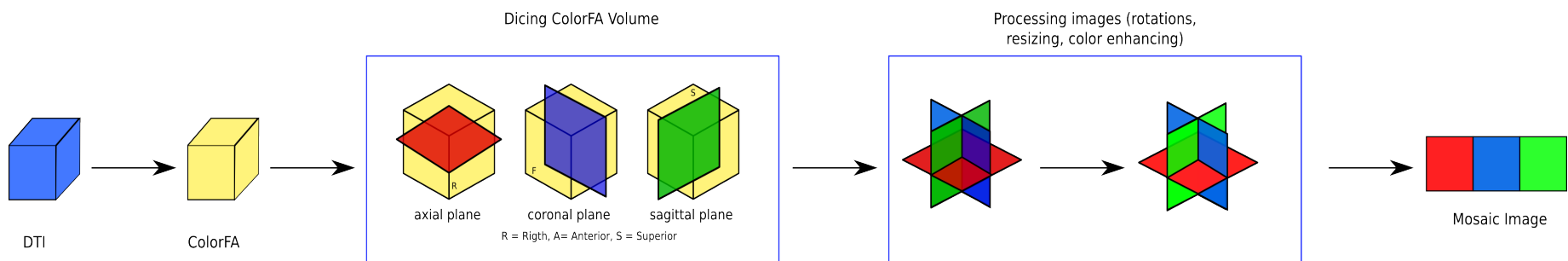
STL : 5%



PHI : 12%

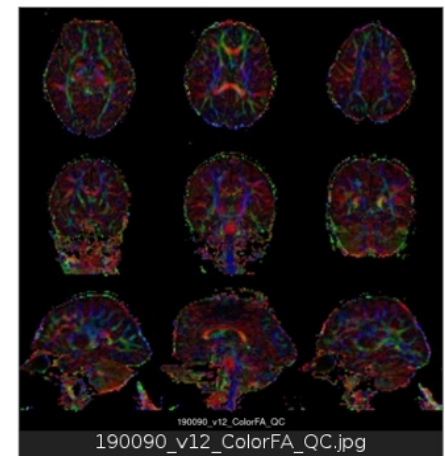
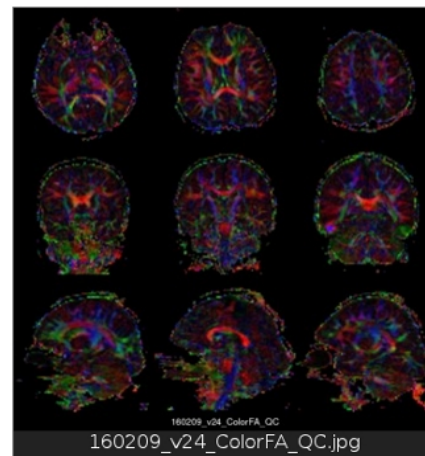
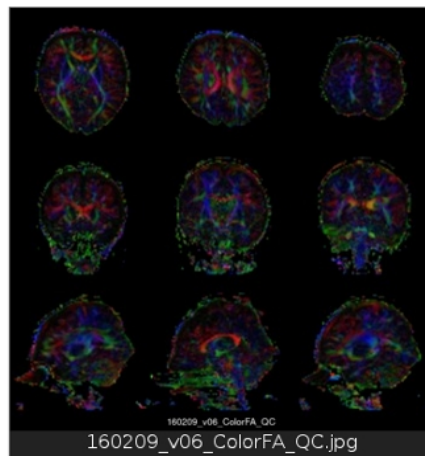
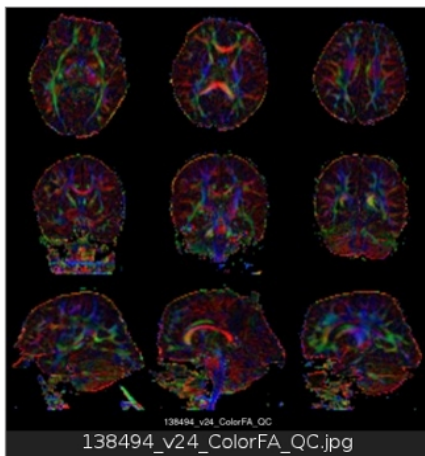
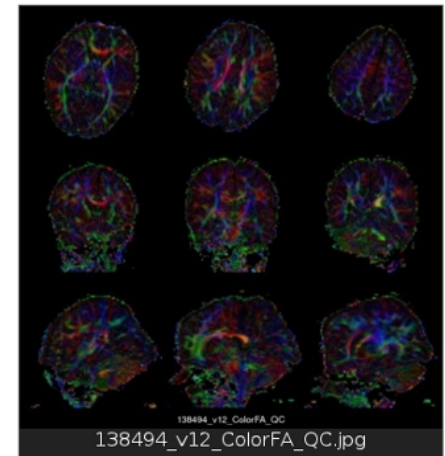
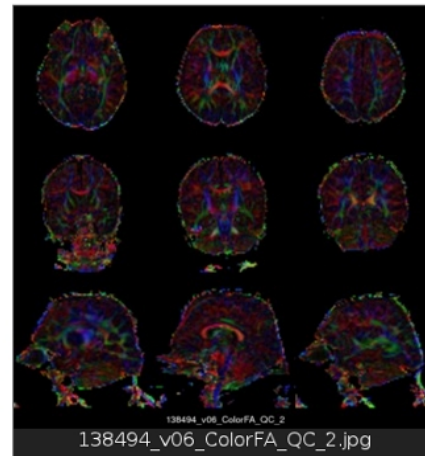
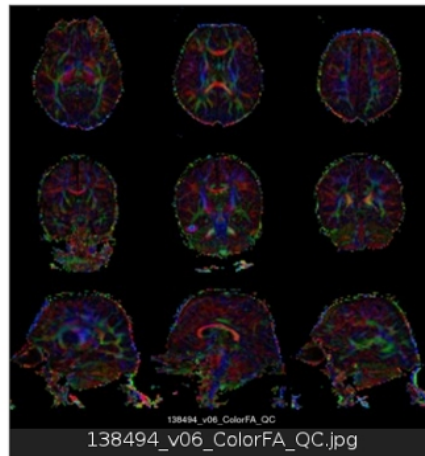
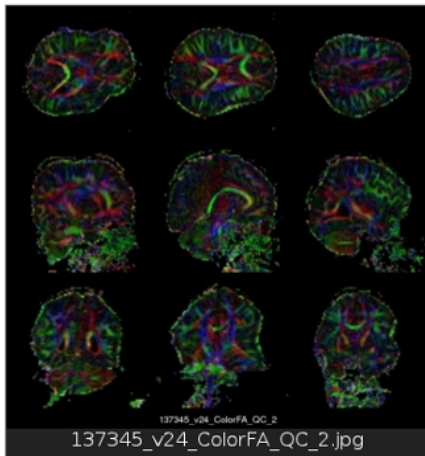
Artifact Detection

- Visual QC
 - ColorFA-OnlineQuickViewer



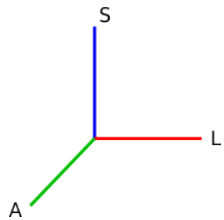
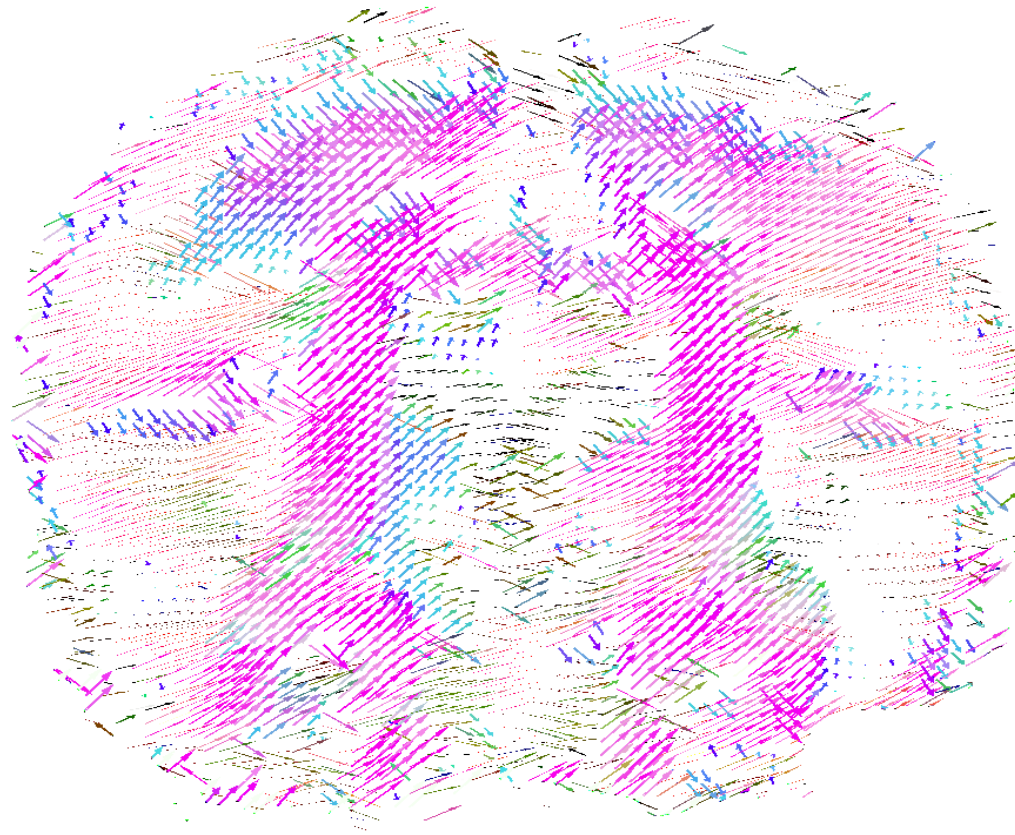
Artifact Detection

- Visual Quality Control



Artifact Detection

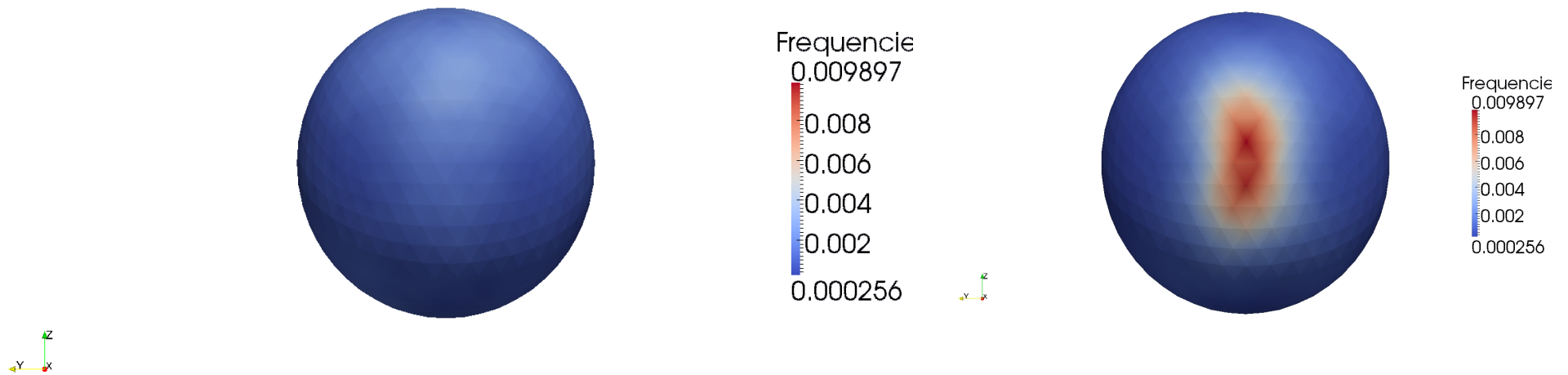
- Principal Direction Space



Principal Eigen Vector Field

Artifact Detection

- Principal Direction Spherical Histogram

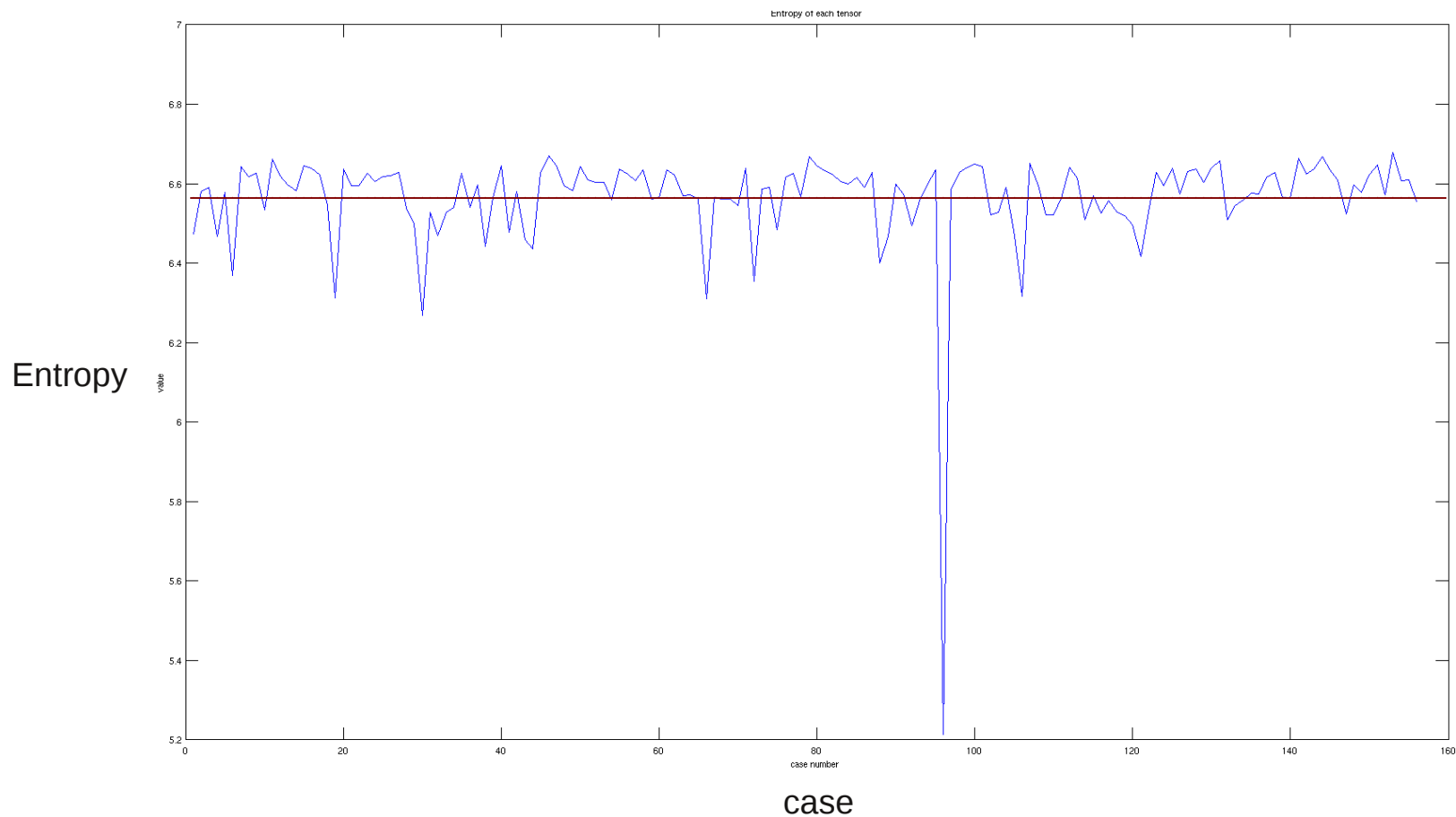


Entropy Based Diffusion Imaging Quality Control via Principal Orientation
Distribution : M. Farzinfar et al

Artifact Detection

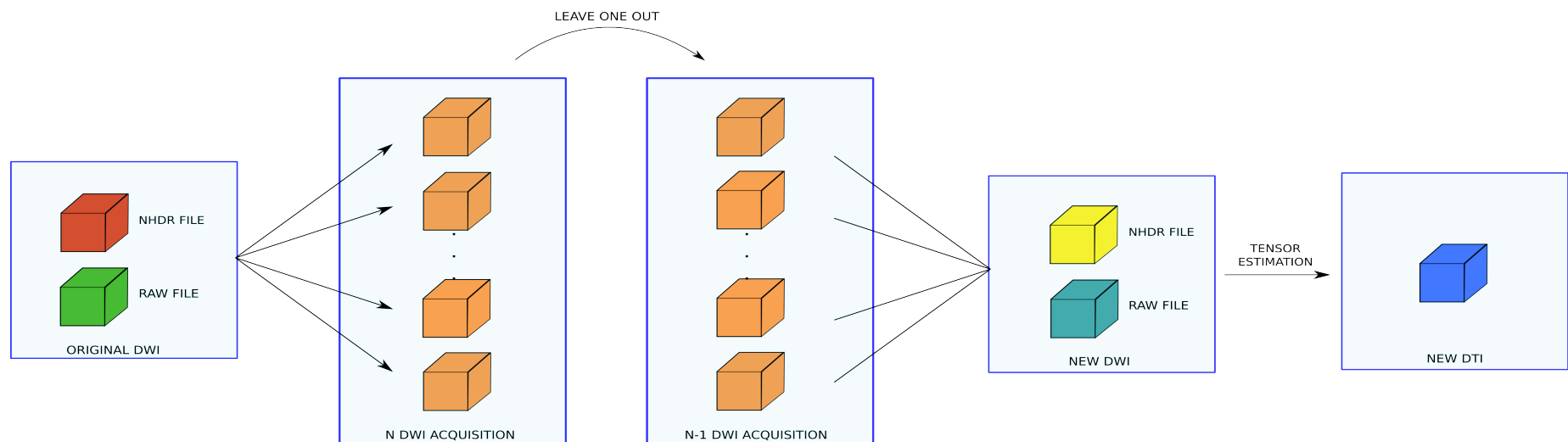
- Entropy

$$E = - \sum_{i=1}^K p_i \log(p_i)$$



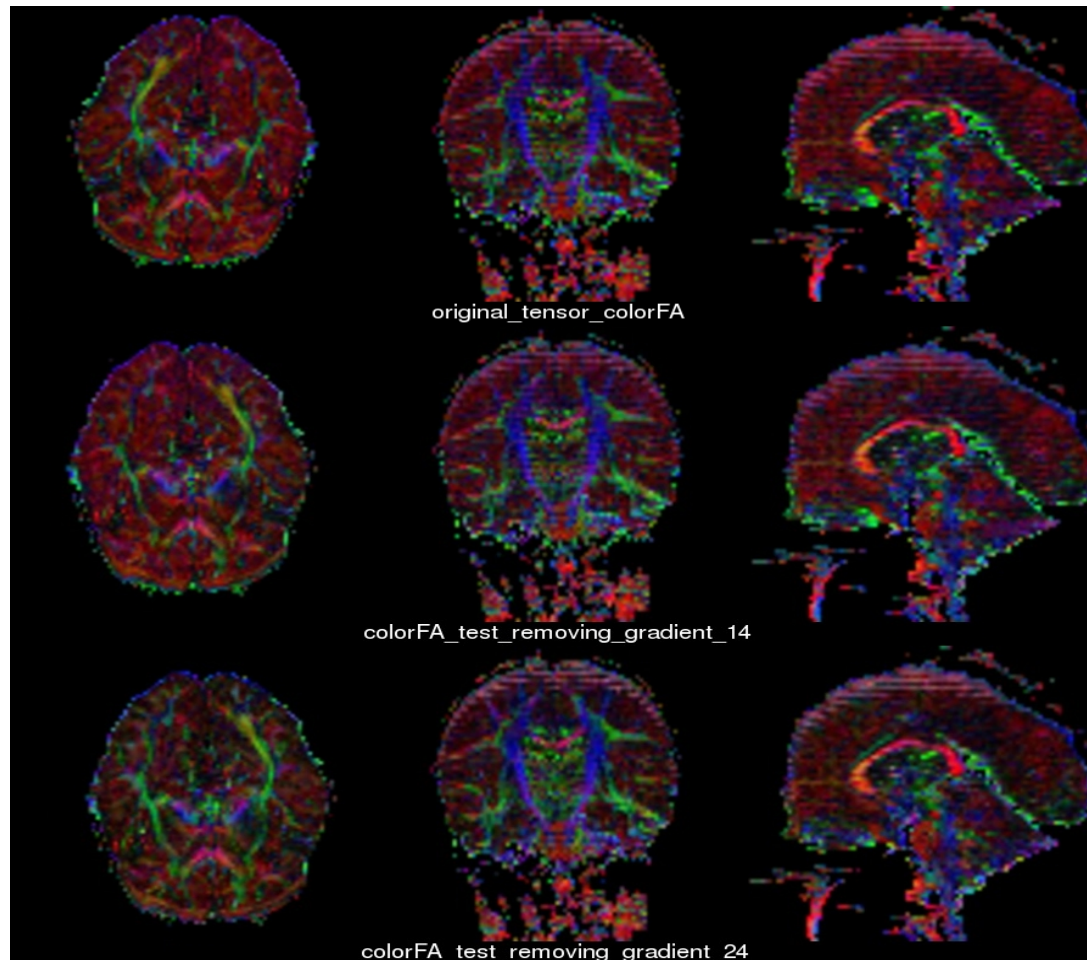
Possible Correction

- The Leave One Out Method (LOO) on DWI set



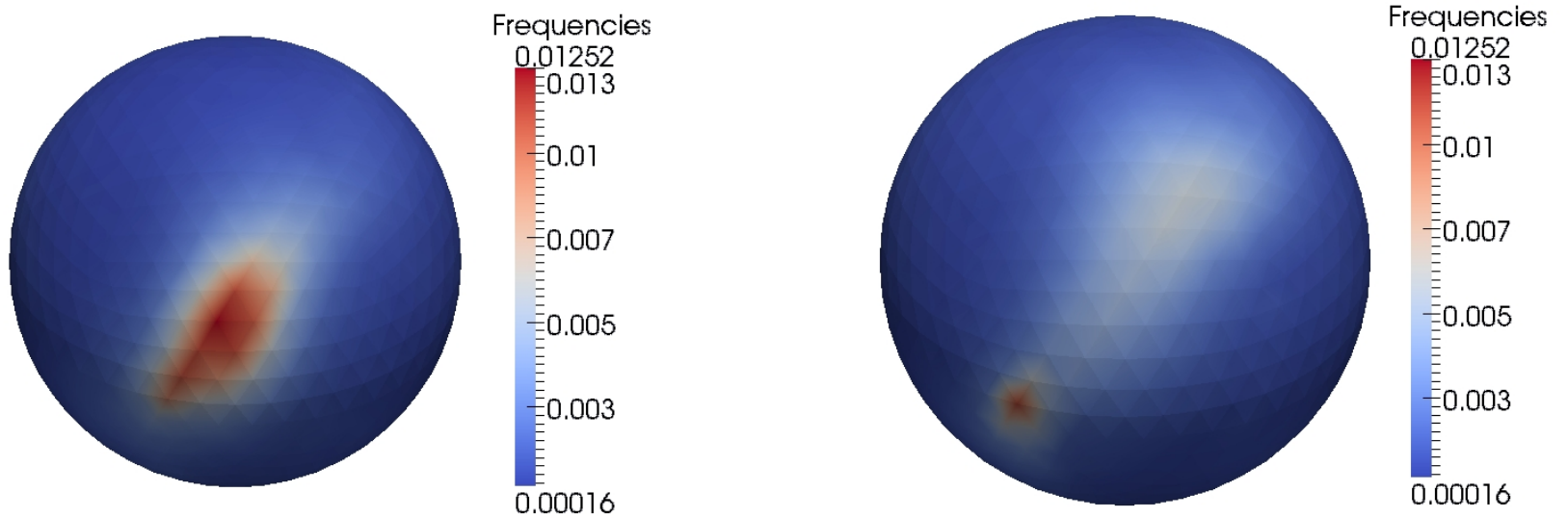
Possible Correction

- Leave One Out improved results :



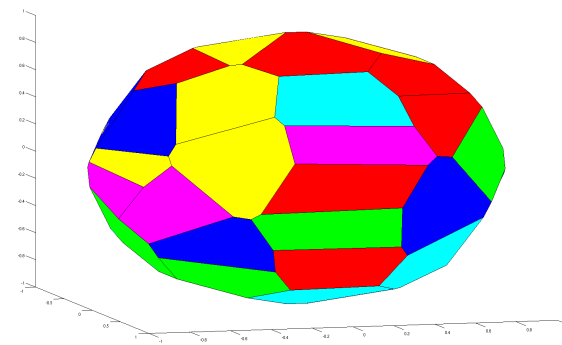
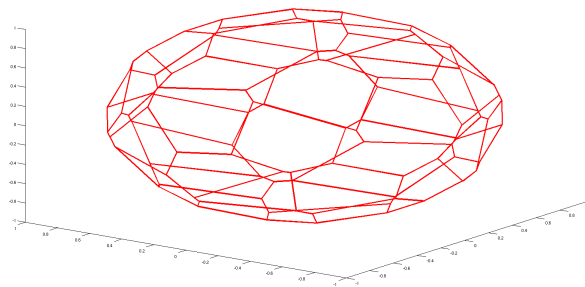
Possible Correction

- Analysis with Principle Direction Distribution



Possible Correction

- Leave Multiple Out Method (LMO)
 - LOO is not generalizable
- Ongoing research
 - Combination of directions depending on patient position
 - Influence of each gradient direction
 - Computation of spherical Voronoi Tessellation



Conclusion

- Identify and characterize directional artifact
- Possible automated detection via entropy
- Possible correction with Leave One / Multiple Out
- Further investigation on acquisition
- Overall Objective: prevent the artifact occurrence

Acknowledgment

- I would like to thank :
 - Dr Guido Gerig
 - Sylvain Gouttard
 - Clement Vachet
 - People in Utah, University of North Carolina and at the Montreal Neurological Institute
 - Everyone for your attention

Questions ?

Tensor Estimation Method

$$S_k = S_0 e^{-b g_k^T \mathbf{D} g_k}$$

$$\ln\left(\frac{S_k}{S_0}\right) = -b g_k^T \mathbf{D} g_k$$

