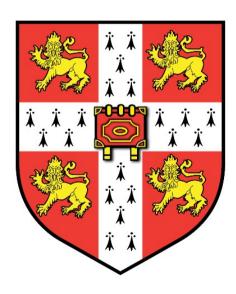
## Cartesian grid q-space reconstruction

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\* University of Cambridge

† MRC Cognition and Brain Sciences Unit



Non-parametric!!! Mostly ...

ODF calculated from Diffusion Spectrum Imaging (Wedeen et al.)

$$\psi_{DSI}(\hat{\mathbf{u}}) = \int_0^\infty P(r\hat{\mathbf{u}}) r^2 dr$$

ODF calculated from Generalized Q-Sampling Imaging (Yeh et al.)

$$\psi_{GQI}(\mathbf{\hat{u}}) = \int_{0}^{\lambda} Q(r\mathbf{\hat{u}})dr$$

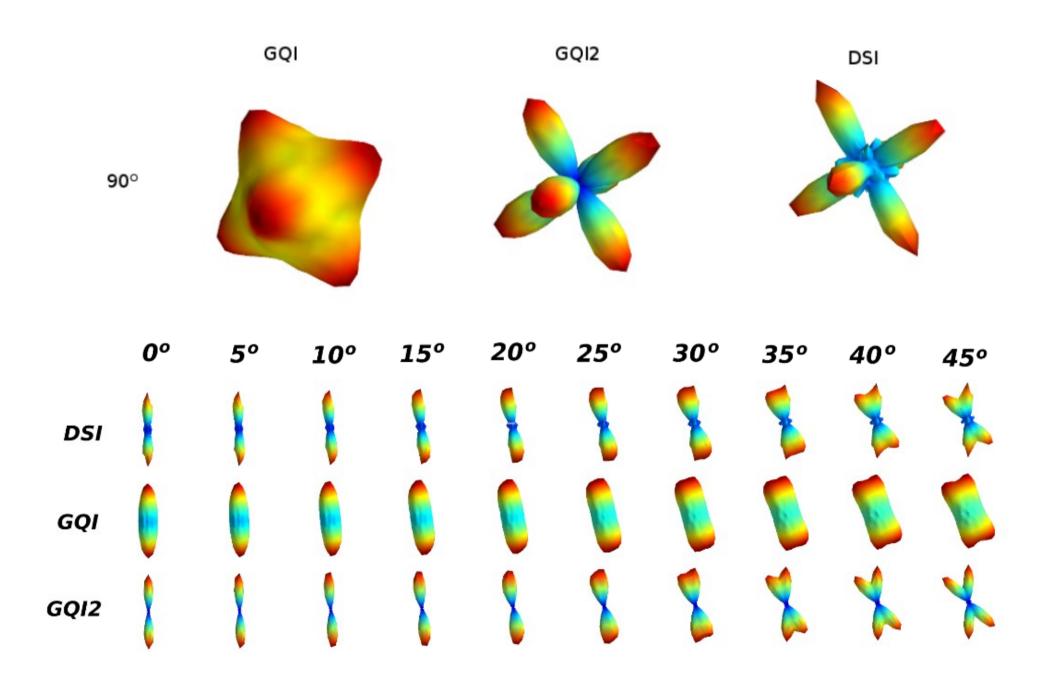
$$= \lambda \int S(\mathbf{q}) \operatorname{sinc}(2\pi r\mathbf{q} \cdot \mathbf{\hat{u}})d\mathbf{q}$$

ODF calculated from GQI with consideration of radial projection (Yeh et al.). We call this QGI2 (Garyfallidis thesis).

$$\psi_{GQI2}(\hat{\mathbf{u}}) = \int_{0}^{\lambda} Q(r\hat{\mathbf{u}})r^{2}dr$$

$$= \lambda^{3} \int S(\mathbf{q})H(2\pi r\mathbf{q} \cdot \hat{\mathbf{u}})d\mathbf{q}$$
where  $H(x) = \begin{cases} \frac{2\cos(x)}{x^{2}} & +\frac{(x^{2}-2)\sin(x)}{x^{3}}, x \neq 0\\ 1/3, x = 0 \end{cases}$ .

Simple analytical form:  $\psi_{GQI2} = \mathbf{s} \cdot \mathbf{H}((6D.G \circ \mathbf{b} \circ \mathbb{1}) \cdot G)\lambda^3/\pi$ 



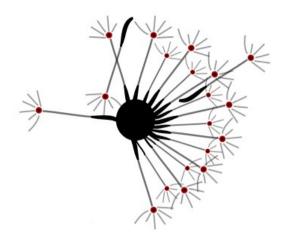
## Strategy for the workshop

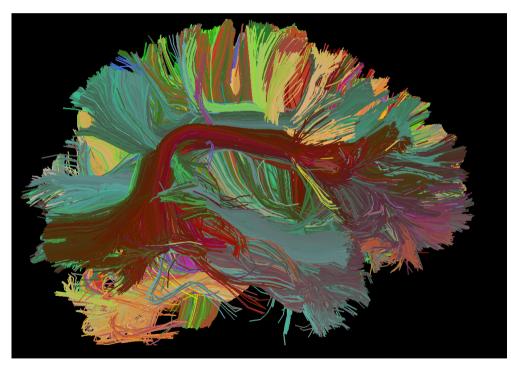
- Use GQI2 to calculate the ODFs of the data.
- Find the peaks of the ODFs.
- If a single peak then fit the Single Tensor (WLS)
   and return the analytical ODF for the Single Tensor.
- Else return the GQI2 ODFs (with no noise very close to the real ODFs).
  - In order to deal with the different SNRs we applied different sampling lengths for each SNR.

## Diffusion Imaging in Python



DIPY.org





Contribute, Share, Collaborate, Communicate Program in a language for human beings.

- Reconstruction
  - DTI, GQI, DSI, EIT
  - SHM, QBI (CSD)
- Propagation
  - Deterministic
    - Fact, Euler
  - Probabilistic
- Analysis
- Segmentation
  - QuickBundles

## Thank you...

