



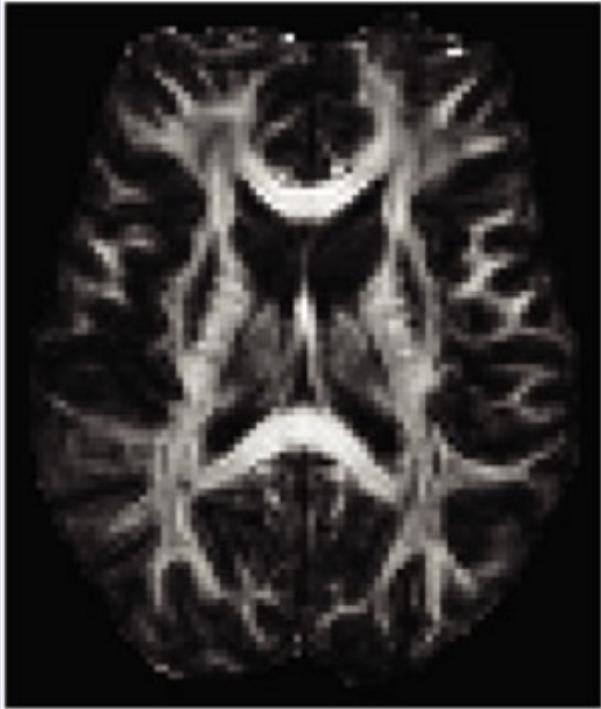
# Spherical-CNN based diffusion MRI parameter estimation is robust to gradient schemes and equivariant to rotation

Tobias Goodwin-Allcock,  
Jason McEwen,  
Robert Gray,  
Parashkev Nachev,  
Hui Zhang

# Machine learning has improved over model fitting

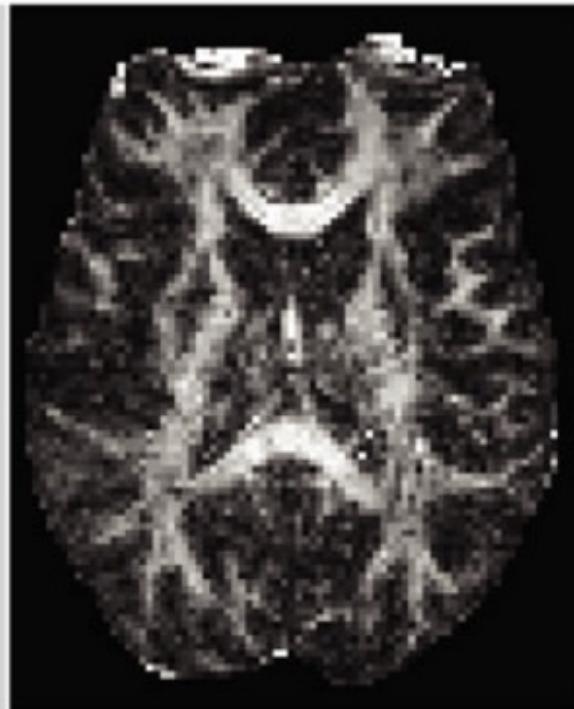
**90 DWI**

**Ground Truth**

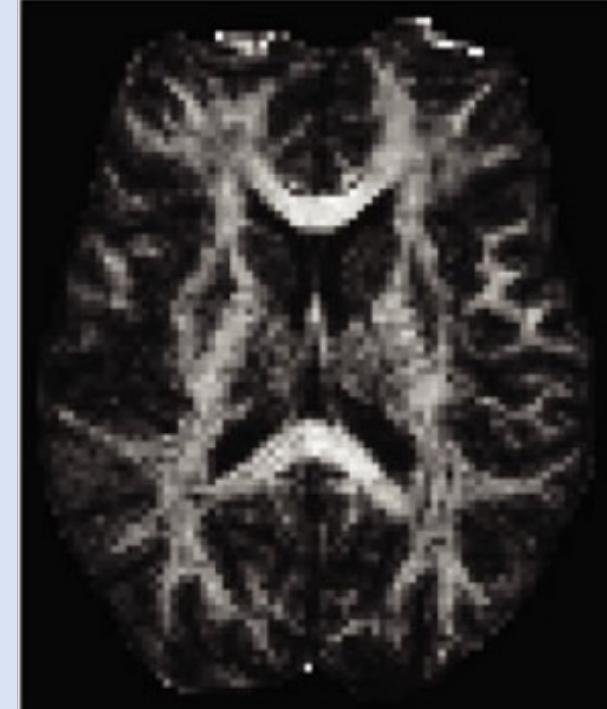


**6 DWI**

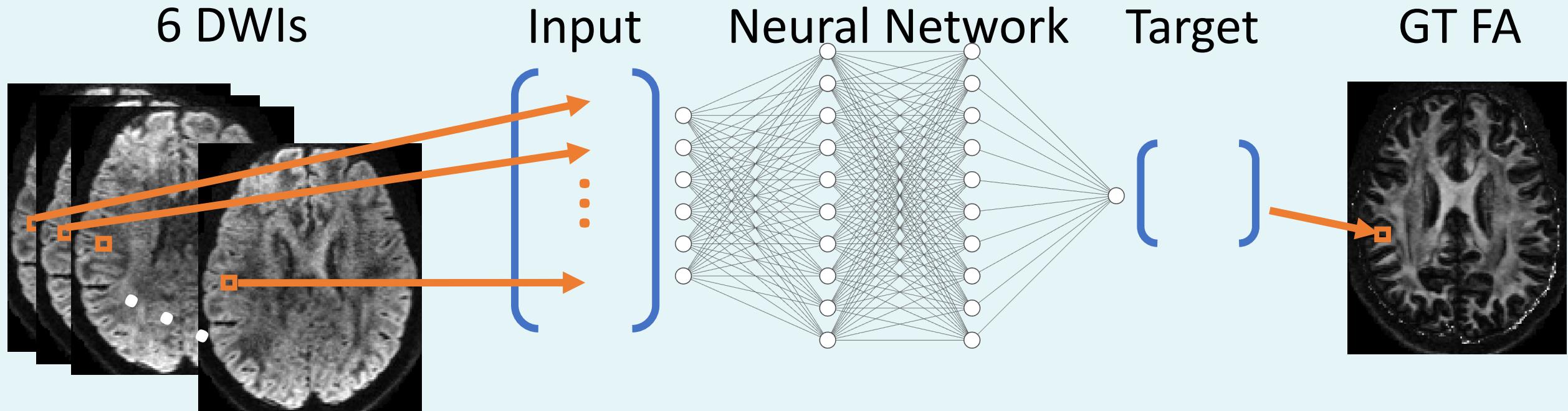
**LLS**



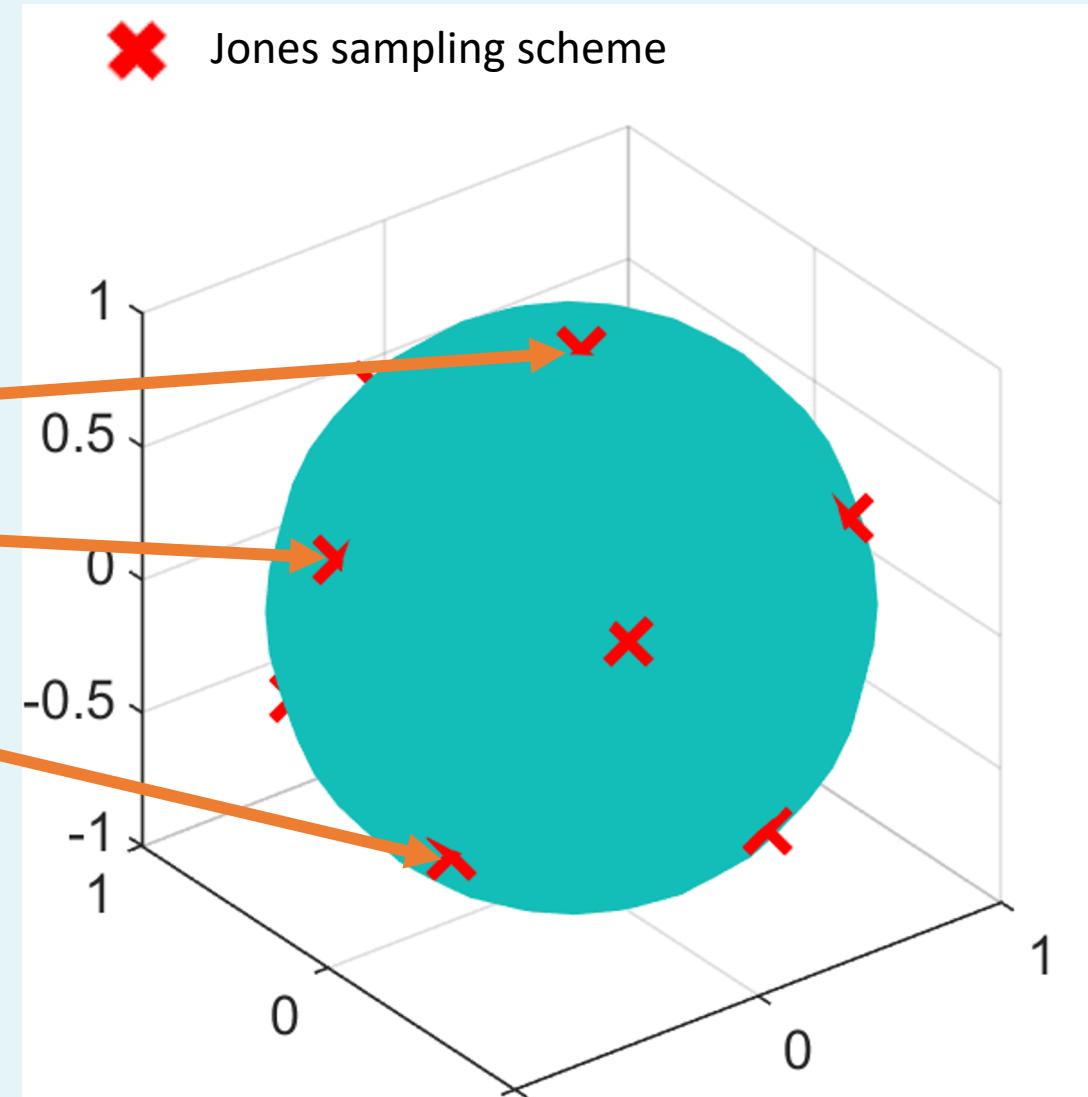
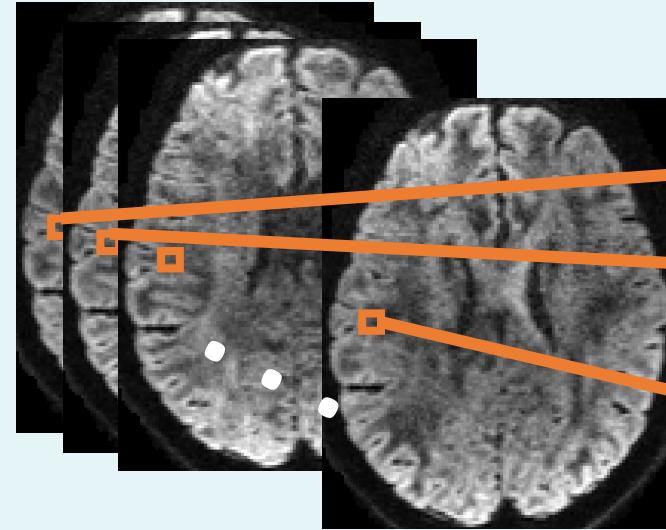
**DiffNet**



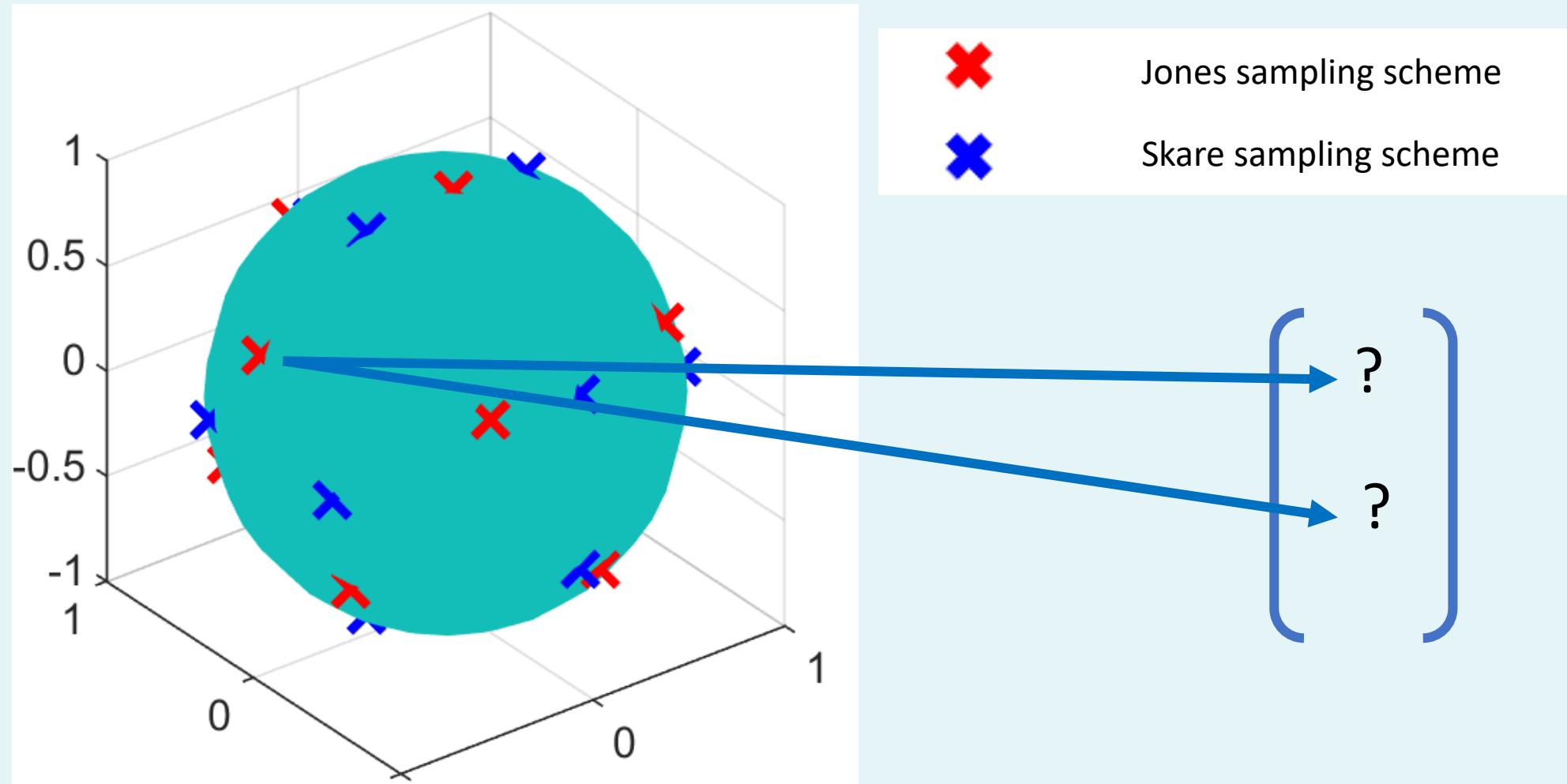
# An example of current ML



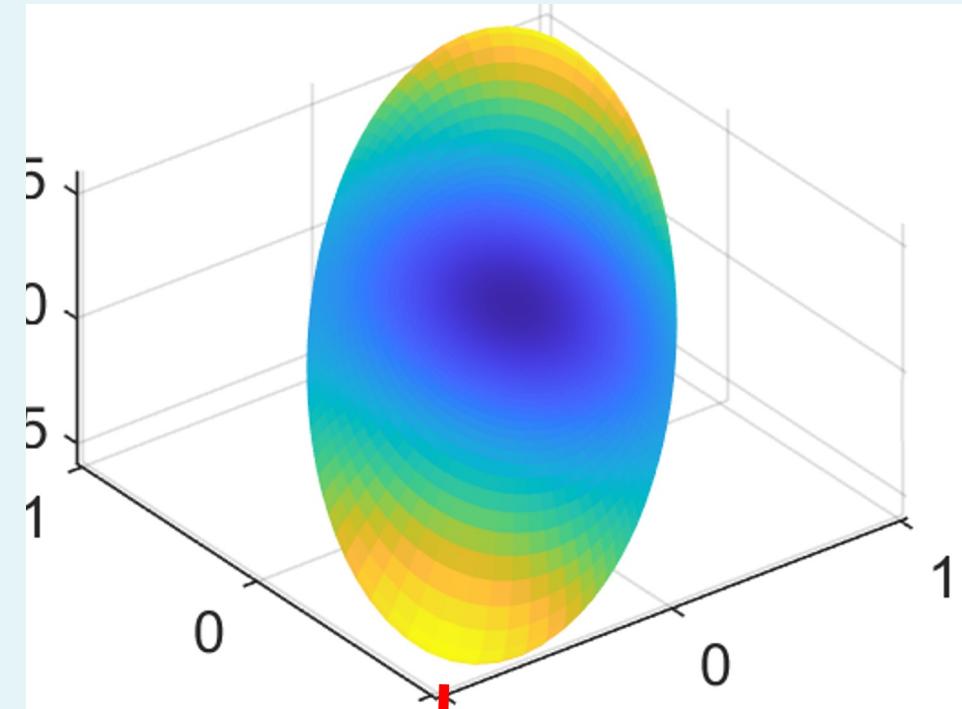
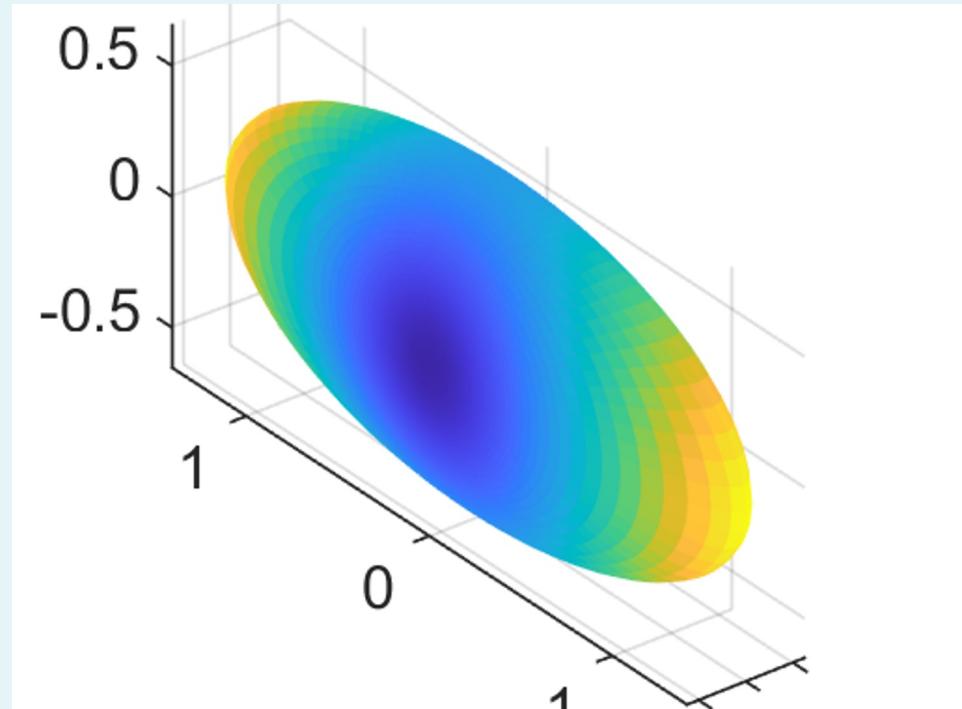
# The spherical geometry of diffusion weighted measures



# Issue 1: Not flexible to new sampling schemes



# Issue 2: Cannot generalise from limited fibre orientations



# Spherical CNNs are a possible solution



- Spherical CNNs still only see one voxel at a time<sup>[1]</sup>
- Spherical geometry of the input is inbuilt
- Can theoretically generalise to unseen fibre orientations due to rotational equivariance<sup>[2]</sup>

[1] Chen, 2020, MICCAI

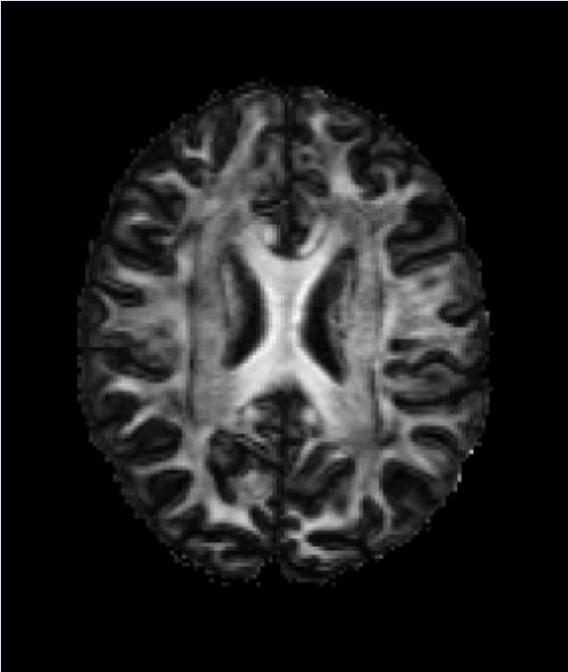
[2] Cobb, 2021, ICLR

# We show S-CNN robust to new sampling schemes

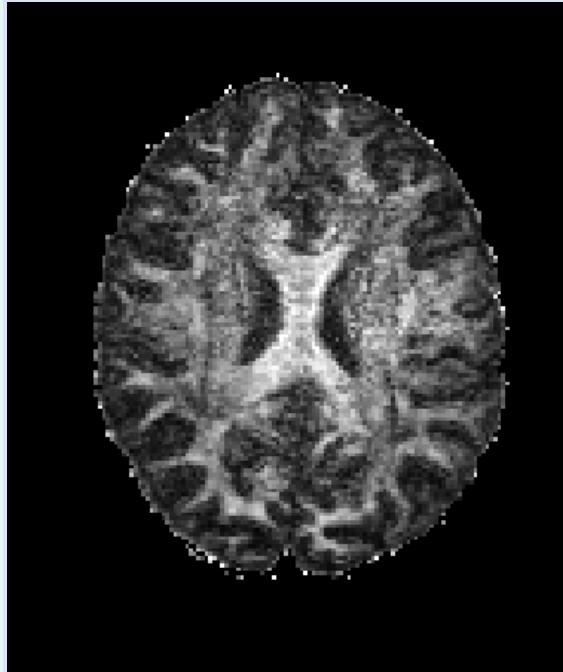
Train and test on a different sampling scheme

**90 DWI**

Ground Truth

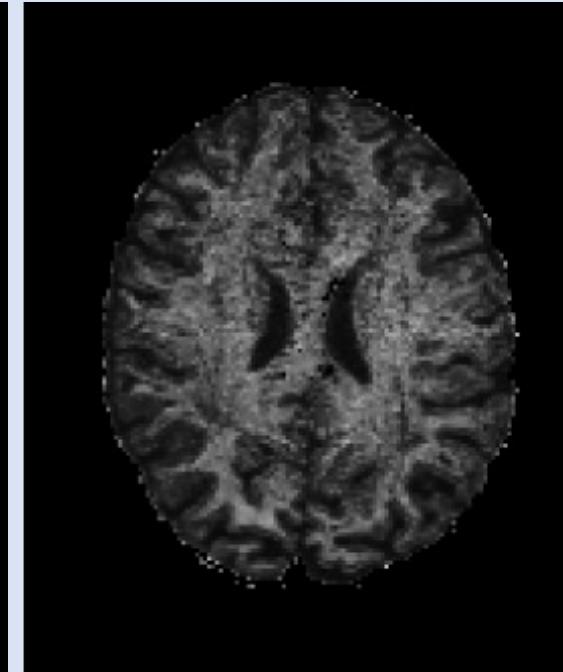


Model Fit

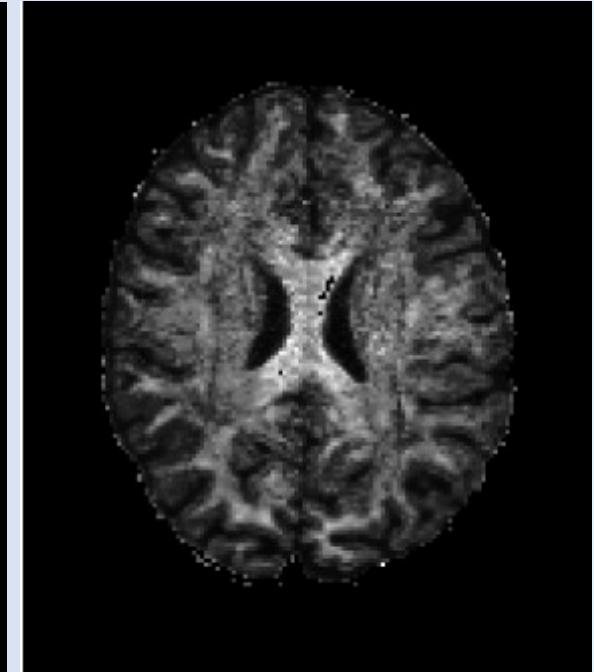


**6 DWI**

FCN



Spherical CNN



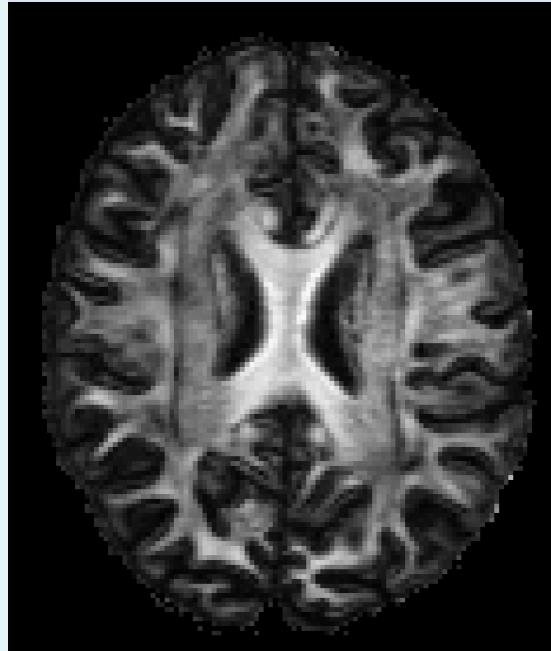
# We show spherical CNN robust to new orientations



Fibre orientation limited to AP direction in training dataset

**90 DWI**

Ground Truth

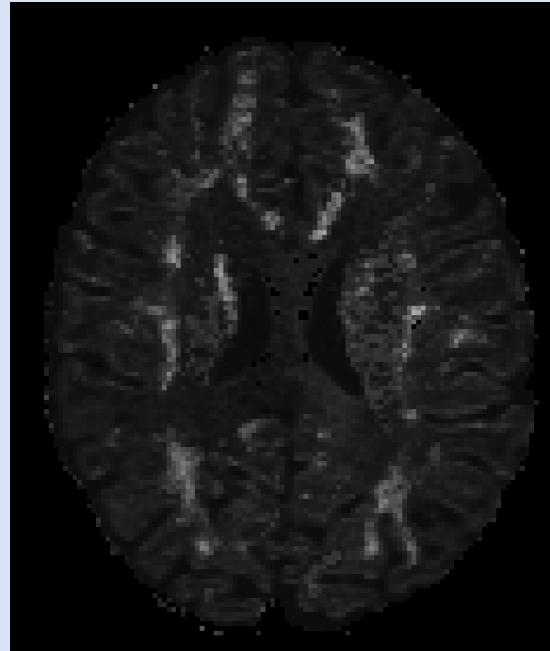


Model Fit

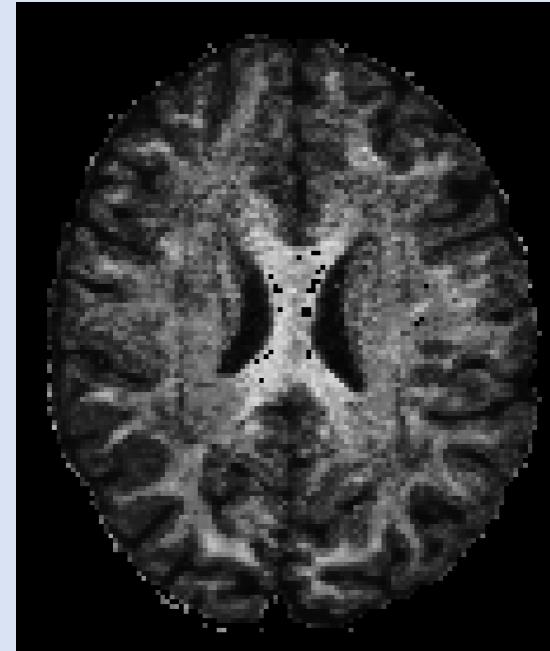


**6 DWI**

FCN



Spherical CNN



## Conclusions

Standard ML shown to lack beneficial properties that Spherical CNNs have:

- a) Robust to sampling scheme
- b) Robust to training set distribution of primary fibre orientation

## Come find me!!

Abstract Num	2605
Session	Quantitative Imaging II
Category	Acquisition & Analysis
Module	5: Machine Learning/Artificial Intelligence
Room	Exhibition Hall:S8 & S9
Computer	46
Day	Thursday
Time	10:15 AM - 11:15 AM



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