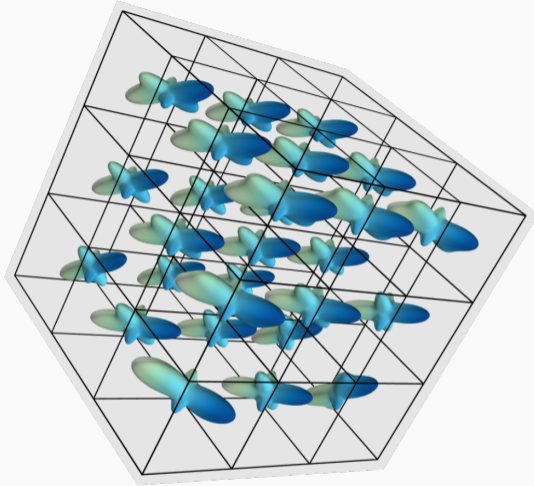




CIMAT



# Self-Supervised Deep Learning Methods for Intra-voxel Structure Analysis From DWI

XXIII Reunión de Neuroimagen

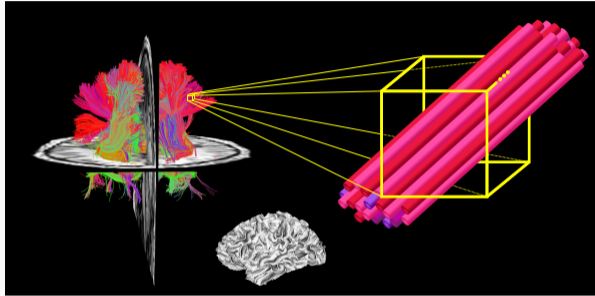
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Hanna Ehrlich

Mariano Rivera

October 8th, 2021

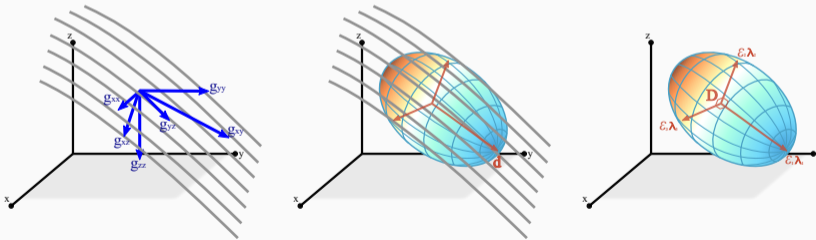
# Diffusion-Weighted Magnetic Resonance Imaging



DW-MRI is a technique that measures the diffusion of water molecules in the human body and generates contrast images. Intra-voxel structure can be inferred from DWI, which has made it possible to study the brain. Tractography algorithms allow the construction of 3D models of brain white matter.

# Diffusion Tensor Imaging

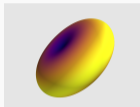
Diffusion is measured in the direction of the weighting gradient. Acquisitions from different gradients allows to compute the ellipsoid representing the diffusion in a region.



# Diffusion Tensor Models

## Diffusion Tensor

$$S_i(\mathbf{g}_i, b_i) = S_0 e^{-b_i \mathbf{g}_i^T \mathbf{D} \mathbf{g}_i}$$



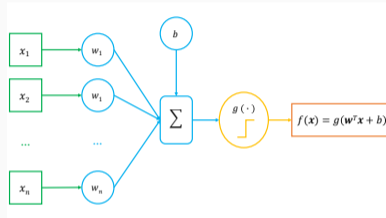
## Diffusion Multi-Tensor

$$S_i(\mathbf{g}_i, b_i) = S_0 \sum_{j=1}^t \alpha_j e^{-b_i \mathbf{g}_i^T \mathbf{D}_j \mathbf{g}_i}$$



# Objective

Estimate the tensor directions and volume fractions given a diffusion signal.



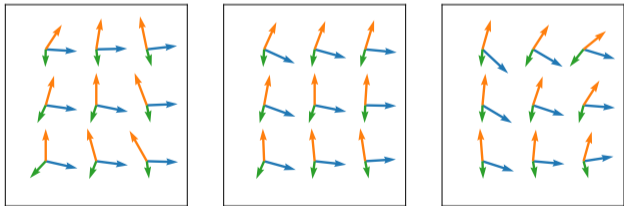
## Motivation

- Deep learning approach has shown excellent results on a wide variety of problems in multiple research areas.
- Designing an appropriate strategy for the data management and the information representation is essential to take off this approach.

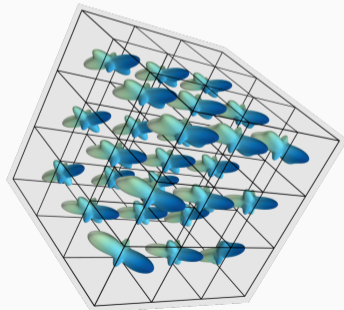
# Synthetic data

Diffusion Multi-tensor Model allows to generate synthetic datasets.

Parameters

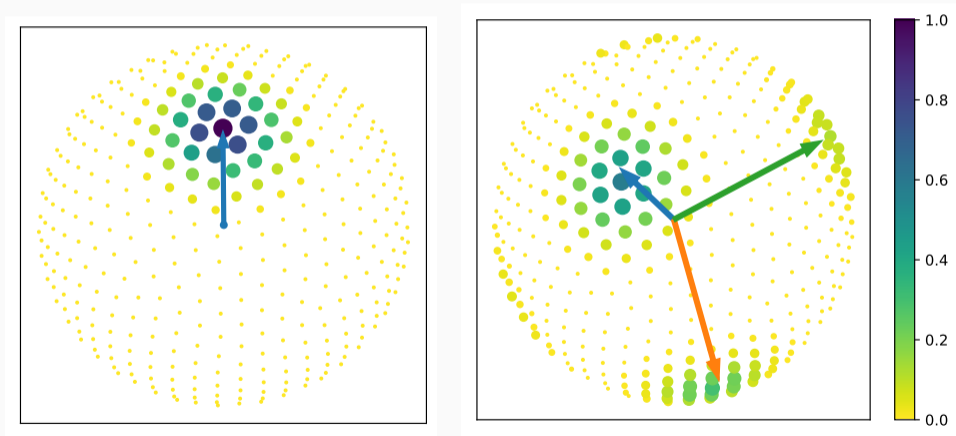


Neighborhood Signals



# Supervised Scheme

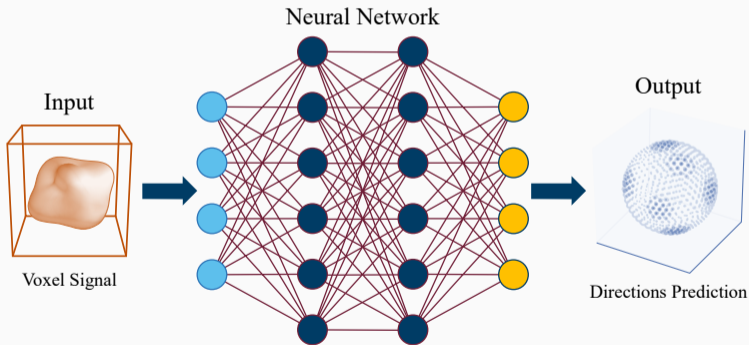
Nearest Elements (NE) and Gaussian Labels ( $G\sigma$ )



# Voxel Model (VOX)

Input : Single signal  $\mathbf{S} \in \mathbb{R}^n$  of a voxel.

Output :  $\mathbf{y} \in \mathbb{R}^m$  representation of the predicted directions.

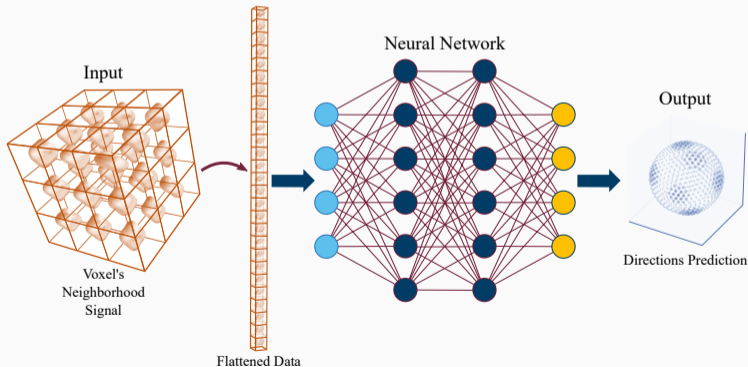




# Neighborhood Model (NBH)

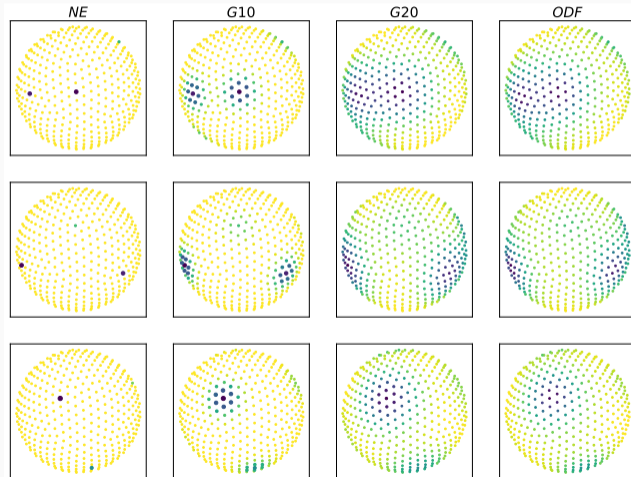
**Input :** Signal neighborhood  $[S] \in \mathbb{R}^{3 \times 3 \times 3 \times n}$  cast into a vector  $S \in \mathbb{R}^{27n}$

**Output :**  $y \in \mathbb{R}^m$  representation of the prediction for the center voxel



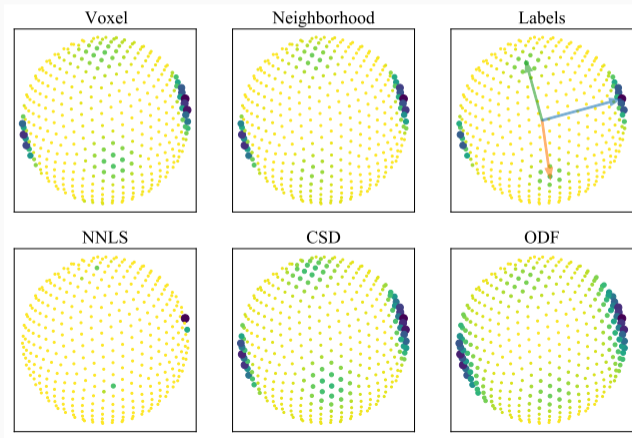
# Labels Construction

Nearest Element and Gaussian Labels were compared with the ODF

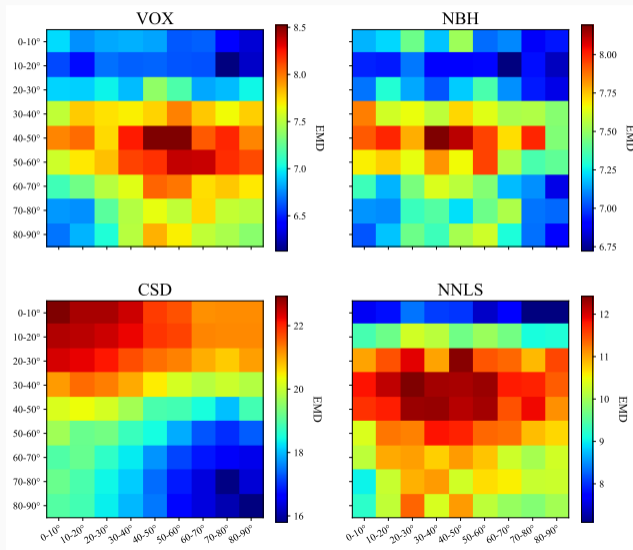


# Models Predictions

The output of the models was evaluated on the test dataset qualitatively and quantitatively.

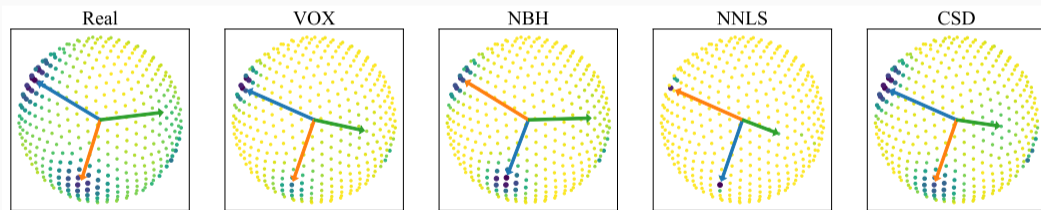


# Models Predictions Error Heat Maps

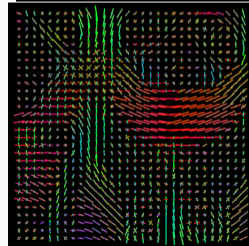
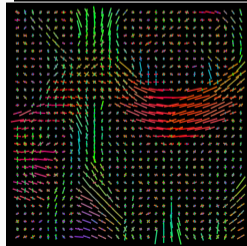
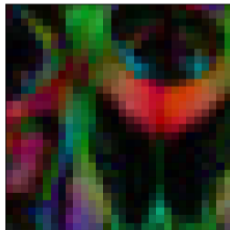
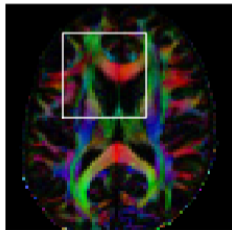
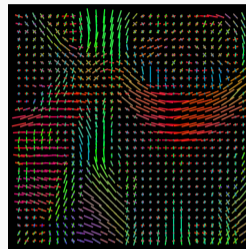
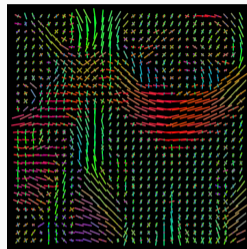
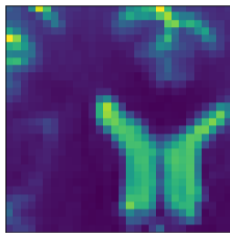
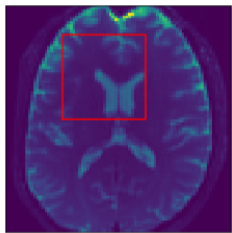


# Peaks Detection

The peaks detection from the models prediction was also evaluated on the test dataset qualitatively and quantitatively.

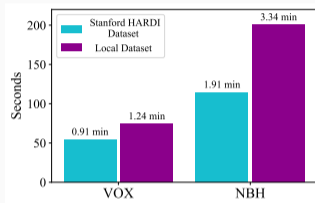


# Stanford HARDI axial plane result

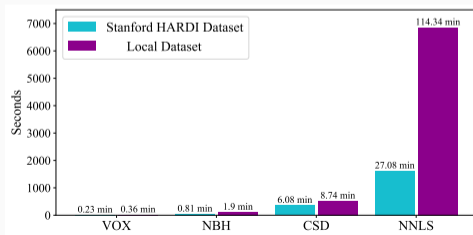


# Computational Times

Training Times:



Prediction Times:



# Conclusions

- It was introduced a self-supervised Deep Learning approach for DW-MRI intra-voxel structure analysis.
- Neural networks show potential to analyze diffusion-weighted signals information.
- Synthetic data is useful for training complex models with this aim.
- The proposed parameters encoding is helpful for the parameters prediction for this problem
- Voxel and Neighborhood proposed models results are comparable with the state of the art methods.



Thank you!