

RECENTRE

REal-time motion CorrEction in magneTic Resonance

PRIN PNRR

Prot. P202294JHK

Annovi Alberto²; **ARGIENTO Benedetta**^{1,2}; Capuani Silvia^{3,5}; Cacioppo Matteo³; Ciardiello Andrea^{2,6}; Coccurello Roberto⁵; Giagu Stefano^{2,3}; Giove Federico⁴; Lonardo Alessandro²; Lo Cicero Francesca²; Maiuro Alessandra^{2,3}; Mancini Terracciano Carlo²; Merola Mario^{1,2}; Montuori Marco⁵; Nistico Emilia³; Perticaroli Pierpaolo²; Rossi Biagio²; Rossi Cristian²; Rossi Elvira^{1,2}; Simula Francesco³; Voena Cecilia².

1 - Intent

Magnetic Resonance (MR) is a non-invasive modality for investigation of tissue structure and function, but hindered by susceptibility to motion artifacts.

The RECENTRE project introduces an innovative **real-time motion correction technique for MRS**, leveraging advanced deep learning methods incorporating high-speed algorithms from experimental high-energy physics.

3 - Framewise Displacement Approach

Framewise displacement (FD), a subject-specific time-series indexing an **overall estimate of movement over time**

$$FD_{total}(\text{mm}) = \sum |T_{i,t} - T_{i,t+1}| + 50 \cdot \frac{\pi}{180} \sum |R_{i,t} - R_{i,t+1}|$$

$$FD_{predicted}(\text{mm}) = \sum |\hat{T}_{i,t+1} - T_{i,t+1}| + 50 \cdot \frac{\pi}{180} \sum |\hat{R}_{i,t+1} - R_{i,t+1}|$$

$$FD_{gain} = \frac{FD_{total} - FD_{predicted}}{FD_{total}}$$

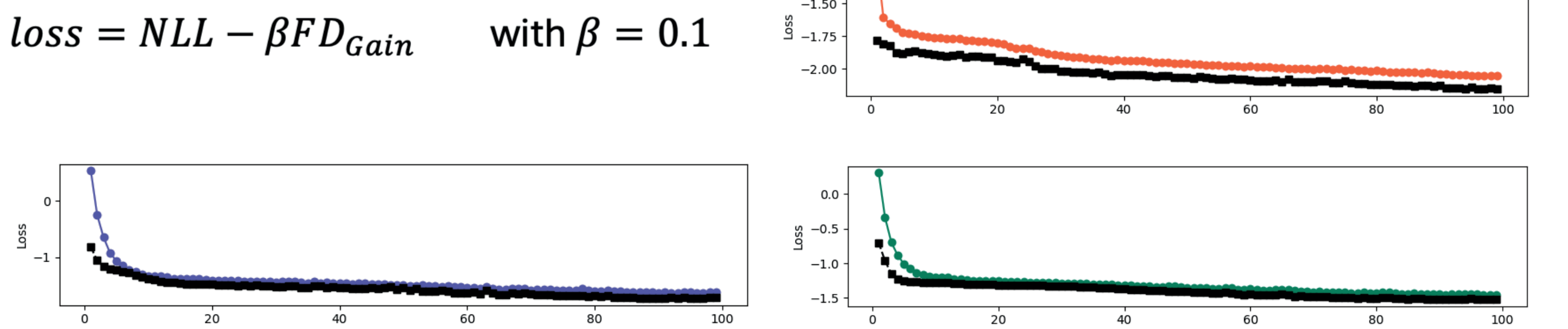
where $T_{i,t}/T_{i,t+1}$ and $R_{i,t}/R_{i,t+1}$ are 3 translation parameters (mm) and 3 rotation (rad) parameters and $\hat{T}_{i,t+1}$ and $\hat{R}_{i,t+1}$ are the predicted quantities.

4 - Gated Recurrent Unit (GRU)

- **Recurrent Neural Network (RNN)**
- Designed for sequential data (simpler variant of LSTM)
- Fewer parameters → faster training, less overfitting;
- Reduces vanishing gradient problem;
- Similar performance to **LSTM**, with lower computational cost.

Layer (type:depth-idx)	Output Shape	Param #
GRUModel	[64, 9]	--
GRU: 1-1	[64, 20, 128]	251,520
LayerNorm: 1-2	[64, 128]	256
ReLU: 1-3	[64, 128]	--
Linear: 1-4	[64, 128]	16,512
LayerNorm: 1-5	[64, 128]	256
ReLU: 1-6	[64, 128]	--
Dropout: 1-7	[64, 128]	--
Linear: 1-8	[64, 9]	1,161
Linear: 1-9	[64, 6]	774

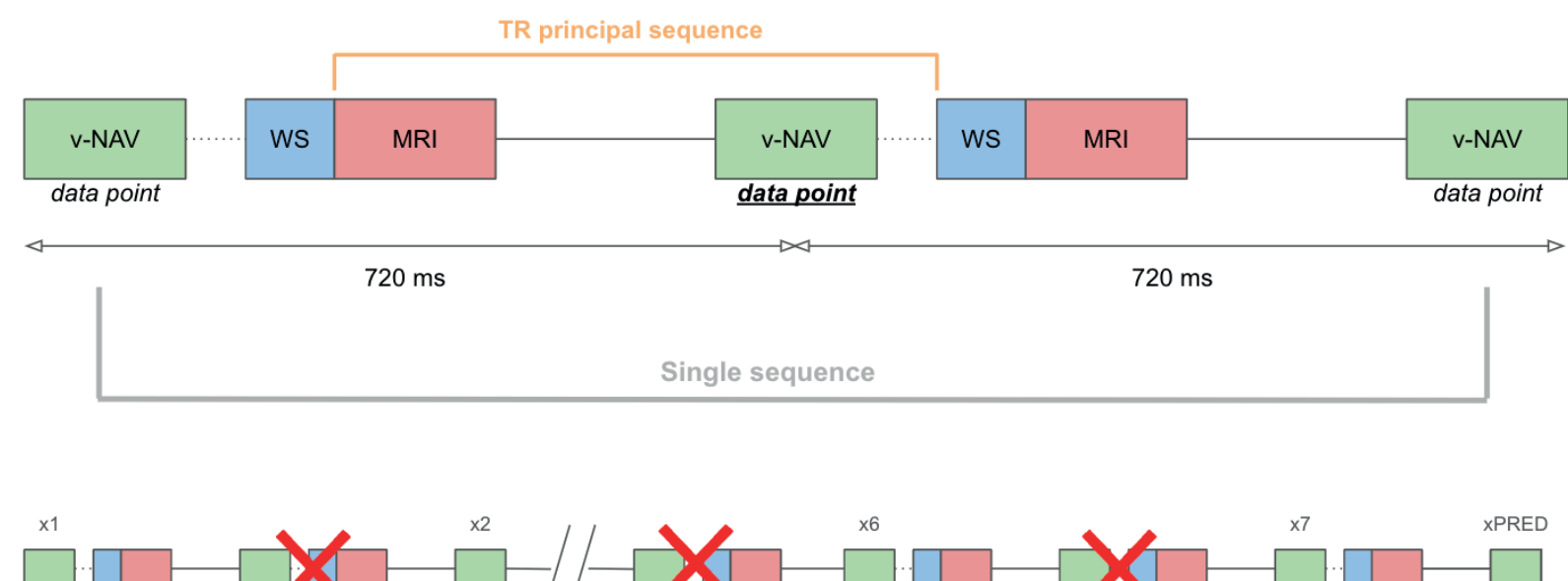
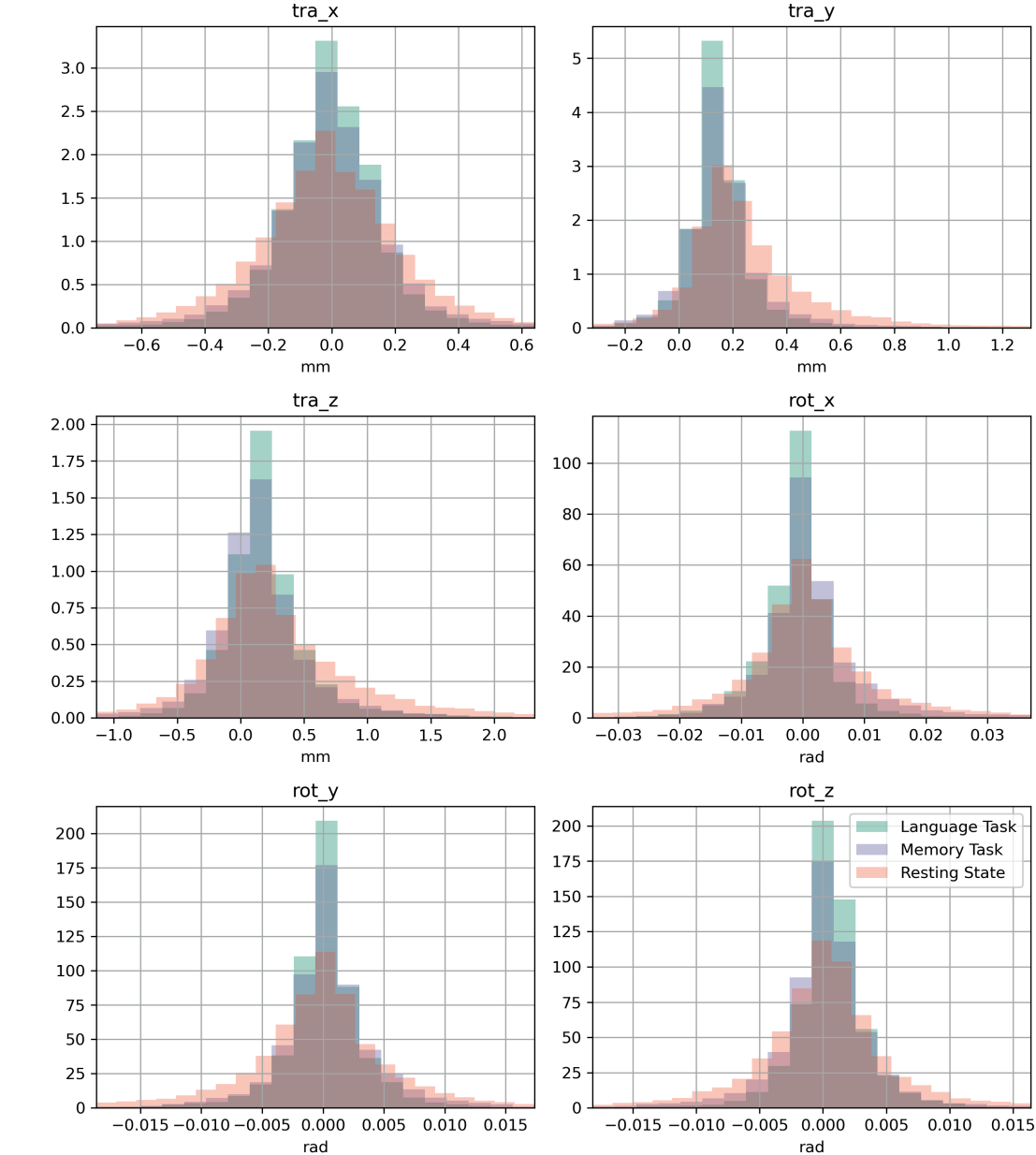
Total params: 270479



2 - Dataset

Human Connectome Project (HCP):

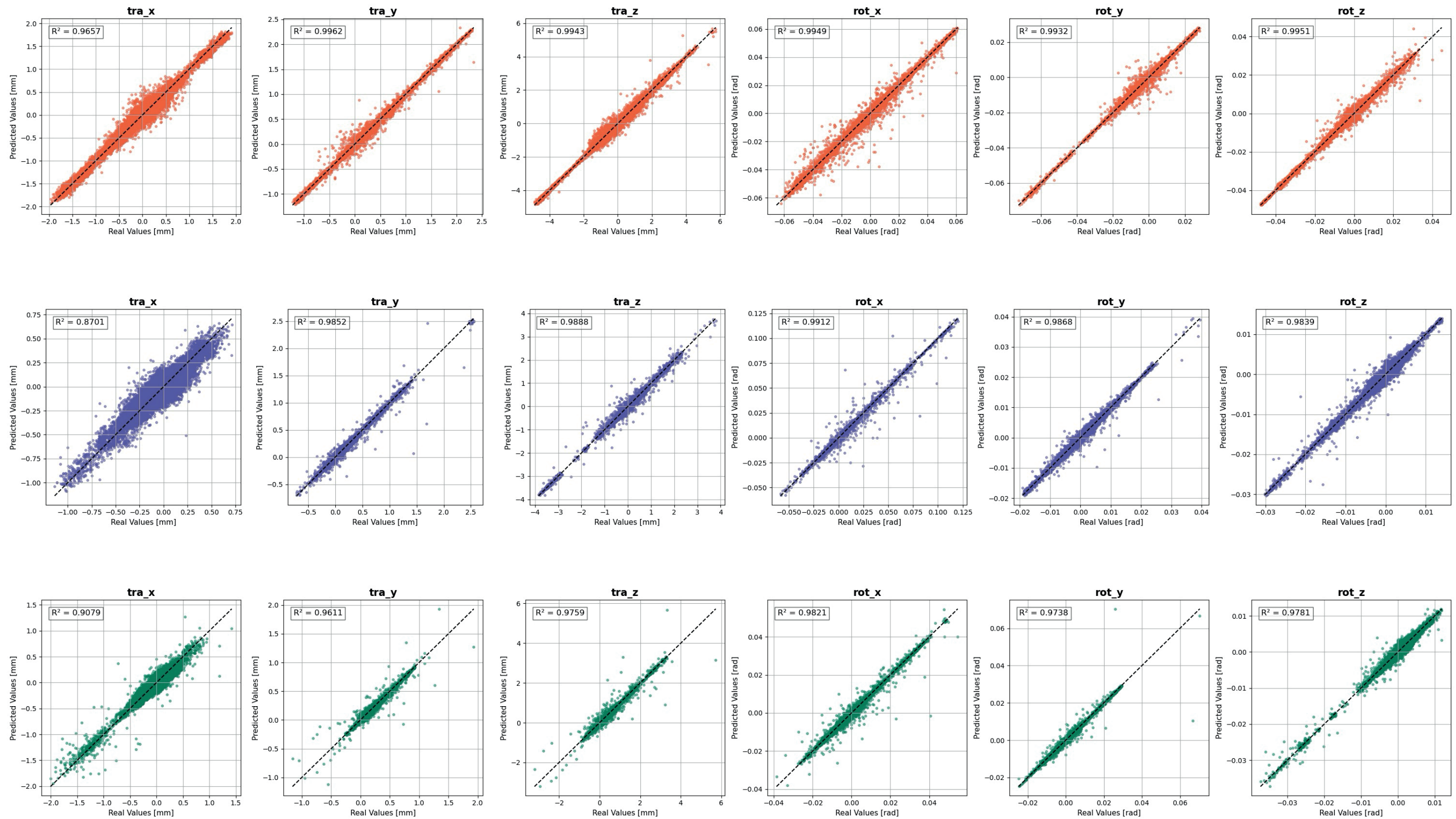
- 1113 Subjects with Simenes 3T MRI
- Repetition time (TR) of 720 ms
- echo time of 33.1 ms
- 3 different acquisition type:
 - **Resting State** (1200 frames),
 - **Working memory task** (316 frames),
 - **Language task** (405 frames)



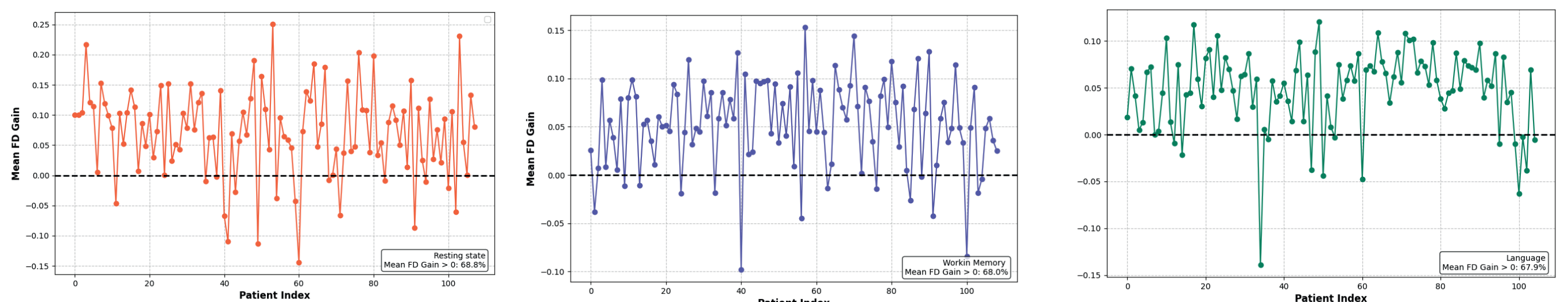
The motion parameters (3 translations and 3 rotations) used were obtained from **rigid realignment** during post-processing of the acquired MRI data.

Given the low repetition time (TR), a single sequence was defined as consisting of **two acquisition points**.

5 - Results



- **Clear linear trends** and consistently high R² values (all above 0.87)
- Very good motion prediction in all the dimension
- Capturing both large and small movements with reasonable accuracy



- Most values of FD gain are positive → NN predictions tends to underestimate the motion on average across patients
- Negative gains → NN may occasionally overestimate slight increases in estimated motion

Next steps

The GRU will be integrated in the workflow of the Siemens syngo MR MAGNETOM system, where the **Siemens Image Calculation Environment (ICE)** operates image reconstruction exploiting roto-translation correction parameters predicted by the the GRU NN model implemented on the **Siemens Framework for Image Reconstruction Environment (FIRE)**.

The FIRE server will be deployed onto the Siemens MARS computer leveraging the on-board GPU cards to enhance the performance of the NN inference task.

The net will be **validated with in-vivo** data taken with the Siemens 3T Prisma MR scanner installed at the **IRCSS Santa Lucia of Rome**.