

# Deep learning-based parameter mapping for joint relaxation and diffusion tensor MR Fingerprinting

**Carolyn M. Pirkel**<sup>\*1,2</sup>, Pedro A. Gómez<sup>\*1</sup>, Ilona Lipp<sup>3,4,5</sup>, Guido Buonincontri<sup>6,7</sup>, Miguel Molina-Romero<sup>1</sup>, Anjany Sekuboyina<sup>1,8</sup>, Diana Waldmannstetter<sup>1</sup>, Jonathan Dannenberg<sup>2,9</sup>, Sebastian Endt<sup>1,2</sup>, Alberto Merola<sup>3,5</sup>, Joseph R. Whittaker<sup>3,10</sup>, Valentina Tomassini<sup>3,4,11</sup>, Michela Tosetti<sup>6,7</sup>, Derek K. Jones<sup>3,12</sup>, Bjoern H. Menze<sup>+1,13,14</sup>, Marion I. Menzel<sup>+2,9</sup>

<sup>1</sup>Department of Informatics, Technical University of Munich, Garching, Germany

<sup>2</sup>GE Healthcare, Munich, Germany

<sup>3</sup>Cardiff University Brain Research Imaging Centre (CUBRIC), Cardiff University School of Psychology, Cardiff, United Kingdom

<sup>4</sup>Institute of Psychological Medicine and Clinical Neurosciences, Cardiff University School of Medicine, Cardiff, United Kingdom

<sup>5</sup>Max Planck Institute for Human Cognitive and Brain Sciences, Leipzig, Germany

<sup>6</sup>Fondazione Imago7, Pisa, Italy

<sup>7</sup>IRCCS Fondazione Stella Maris, Pisa, Italy

<sup>8</sup>Department of Neuroradiology, Klinikum rechts der Isar, Munich, Germany

<sup>9</sup>Department of Physics, Technical University of Munich, Garching, Germany

<sup>10</sup>Cardiff University School of Physics and Astronomy, Cardiff, United Kingdom

<sup>11</sup>Institute for Advanced Biomedical Technologies (ITAB), Department of Neurosciences, Imaging and Clinical Sciences, School of Medicine, University "G. d'Annunzio" of Chieti-Pescara, Chieti, Italy

<sup>12</sup>Mary McKillop Institute for Health Research, Faculty of Health Sciences, Australian Catholic University, Melbourne, Australia

<sup>13</sup>Center for Translational Cancer Research, Munich, Germany

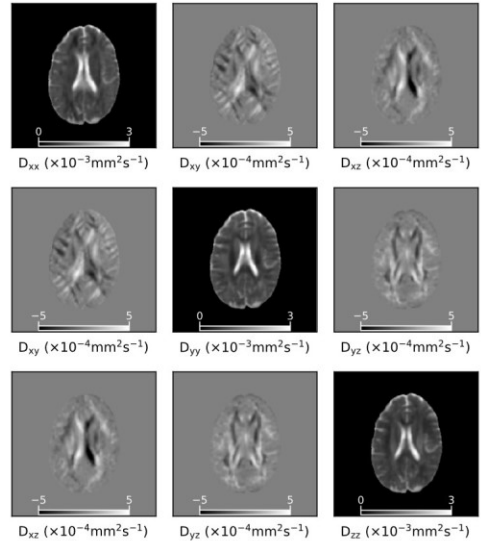
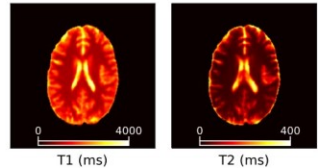
<sup>14</sup>Munich School of BioEngineering, Garching, Germany





MR Fingerprinting (MRF)

Ma et al. *Nature* (2013)  
Jiang et al. *Magn Reson Med* (2015)



Diffusion-weighted MRF

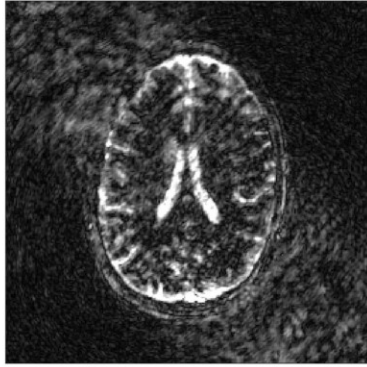
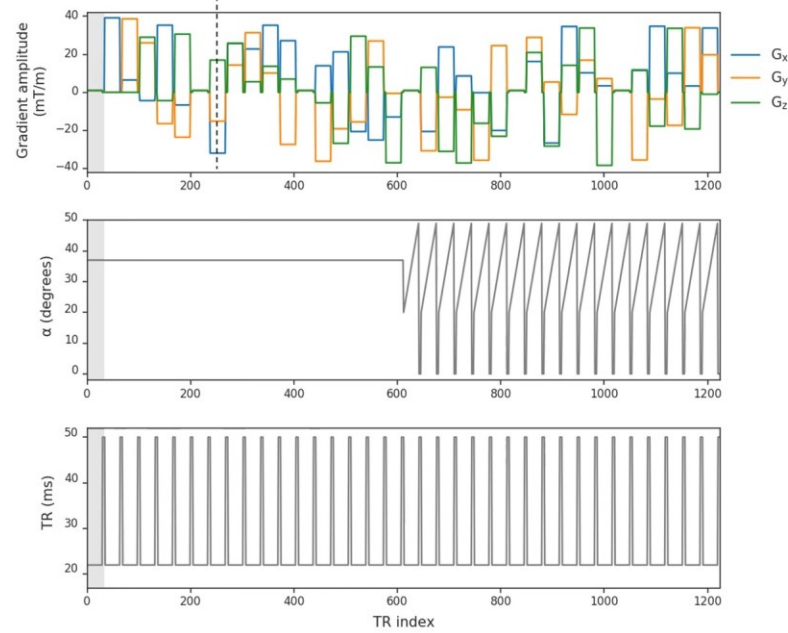
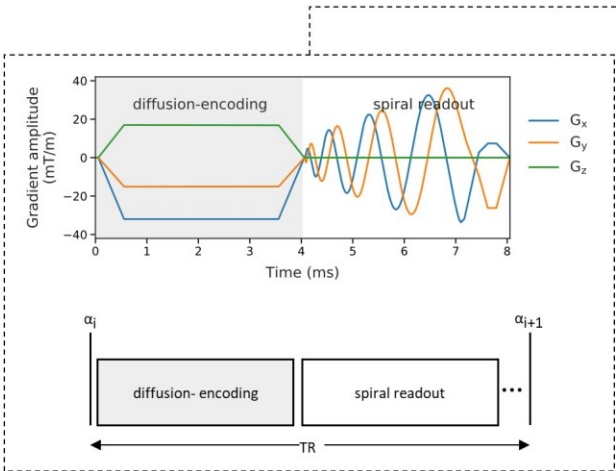
Jiang et al. *ISMRM* (2014, 2016, 2017)  
Cohen et al. *ISMRM* (2018)  
Rieger et al. *ISMRM* (2018)

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PirkI, Gómez et al. *MIDL* (2020)



# Methods | Diffusion-sensitized MRF sequence

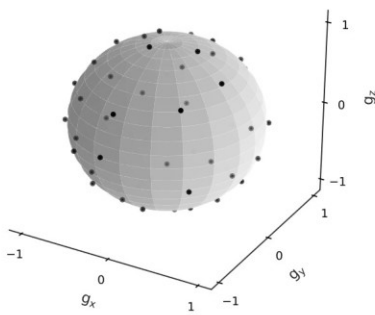


## Acquisition parameters

- $1.2 \times 1.2 \times 5 \text{ mm}^3$  resolution
- $22.5 \times 22.5 \text{ cm}^2$  FOV
- Variable density spiral sampling (34 interleaves)
- 30 diffusion encoding directions
- $T_I = 18\text{ms}$ ,  $T_E = 6\text{ms}$
- 32s / slice acquisition time

## Sliding window reconstruction

- Window size = 34
- Window stride = 34

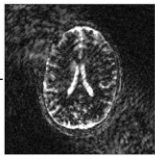




MRF image-series  $\mathbf{x}$

$$\mathbf{x} \in \mathbb{R}^{256 \times 256 \times T},$$

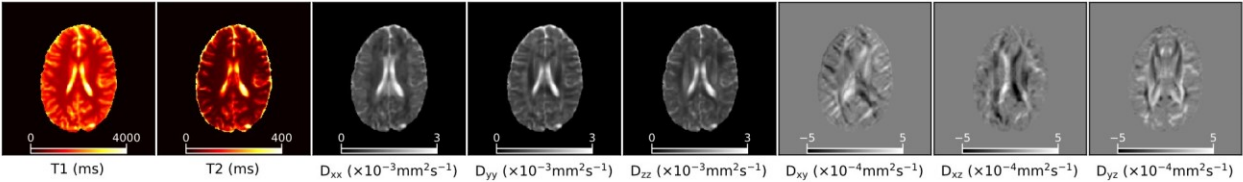
$T = 36$  temporal channels



Quantitative relaxation and diffusion tensor reference maps  $\mathbf{y}$

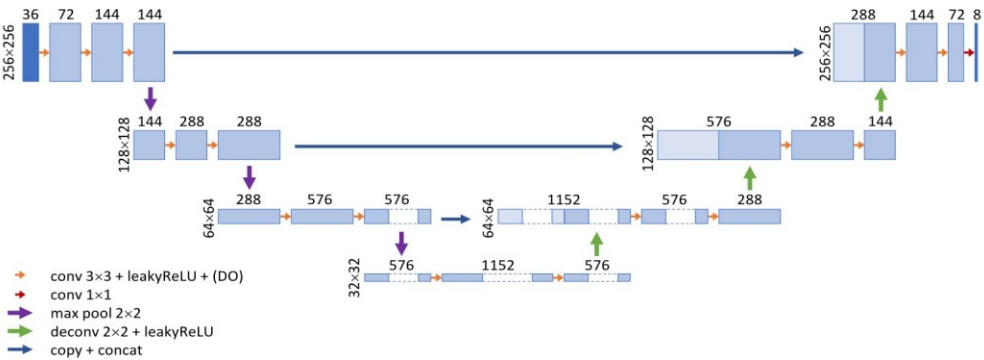
$$\mathbf{y} \in \mathbb{R}^{256 \times 256 \times Q},$$

$Q = 8$  quantitative channels



$$\mathbf{x}' = \frac{\mathbf{x} - \min(\mathbf{x})}{\max(\mathbf{x}) - \min(\mathbf{x})}$$

$$\mathbf{y}'_q = \frac{\mathbf{y}_q}{\max(|q_{min}|, |q_{max}|)}$$



Dataset

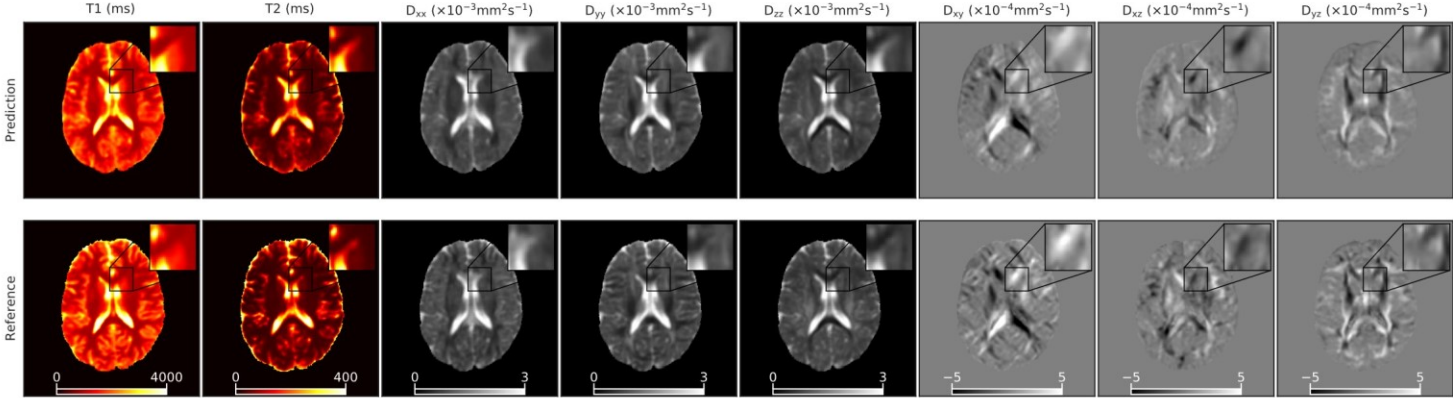
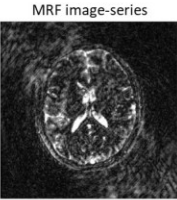
- 11 MS patients, 9 healthy volunteers
- 8-12 slices / subject

Experimental setup + CNN training

- 10-fold cross validation
- L1 loss
- Learning rate = 1e-4
- Dropout rate = 0.25
- Batch size = 5
- 400 epochs



## Results | Qualitative evaluation



### Relaxation and diffusion tensor maps

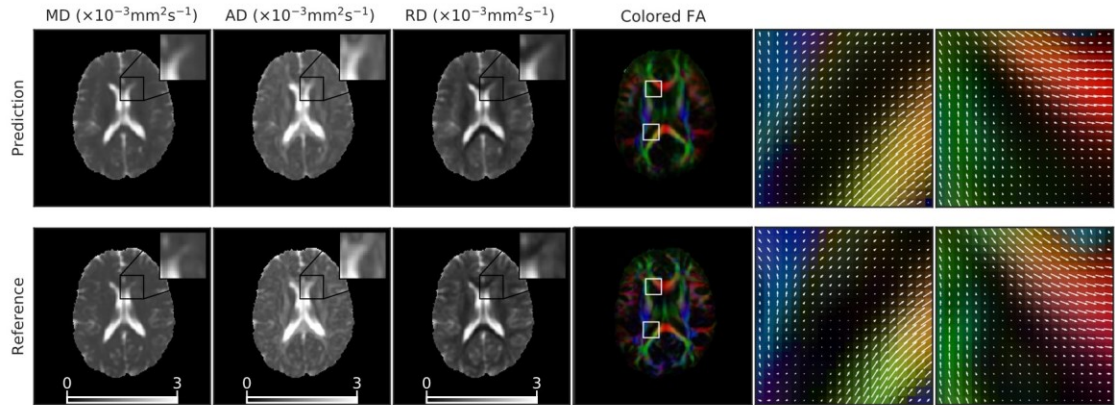
- High consistency between CNN prediction and state-of-the-art reference
- CNN reliably reconstructs relaxation and orientational diffusion information

### Scalar diffusion metrics

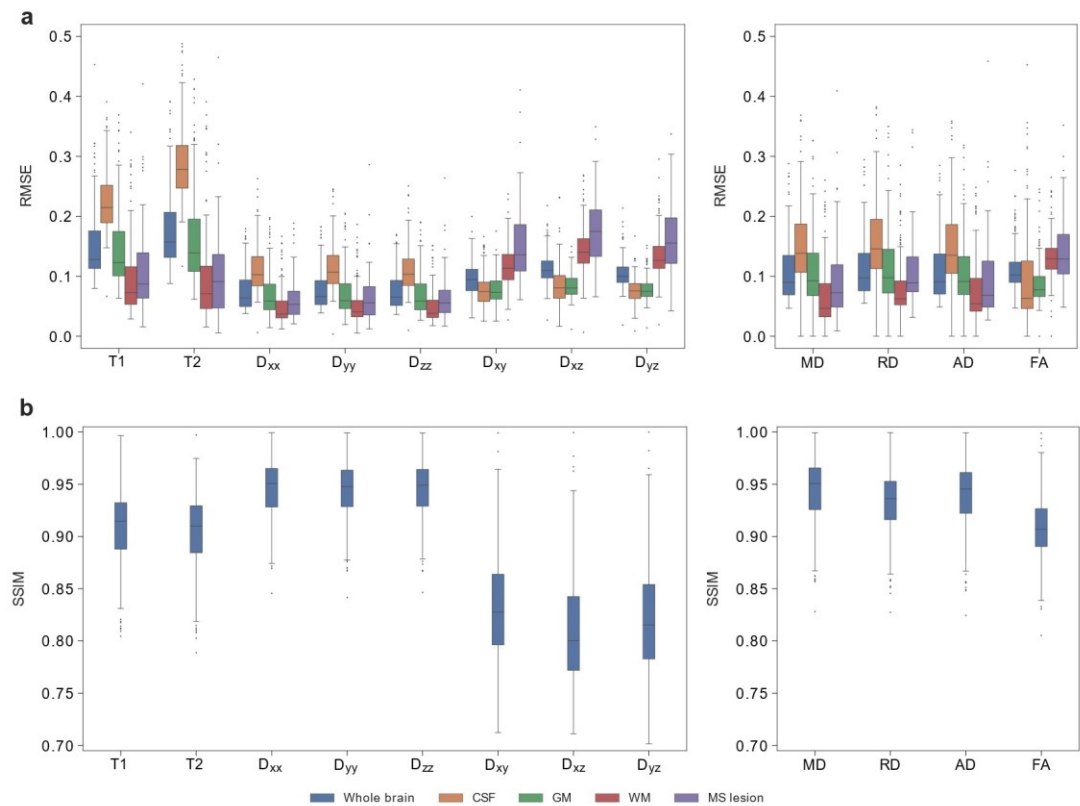
Good agreement with EPI-DTI reference

### Colored FA maps + primary diffusion eigenvectors

Characteristic fiber structure in WM is captured







**Quantitative evaluation substantiates qualitative findings**

- Reliably reconstruction of relaxation and orientational diffusion information, also in regions of diagnostic importance (MS lesions)  
→ Generalization capability
- Comparable reconstruction performance for T1 and T2 with respect to DESPOT1/2 methods
- Better agreement with EPI-DTI reference for diagonal diffusion tensor elements ( $D_{xx}$ ,  $D_{yy}$ ,  $D_{zz}$ ) than off-diagonal elements ( $D_{xy}$ ,  $D_{xz}$ ,  $D_{yz}$ )



Thank you!

**Relaxation and diffusion-sensitized MRF sequence**  
**CNN-based multivariate regression**



- ✓ Relax MR acquisition requirements
- ✓ Efficiently encode:
  - T1 and T2 relaxation times
  - Orientational diffusion information
- ✓ Bypass conventional dictionary matching

**Outlook: Improve on our baseline**

- More advanced deep learning approaches
- More efficient, motion-robust diffusion encoding scheme

**Major challenge: Severe head motion**

- Prospective and retrospective motion correction approaches
- Increase motion robustness of sequence design



IRCCS FONDAZIONE  
**STELLA MARIS**