

# MAGNETIC RESONANCE FINGERPRINTING RECONSTRUCTION VIA A SPATIOTEMPORAL NEURAL NETWORK

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## Introduction

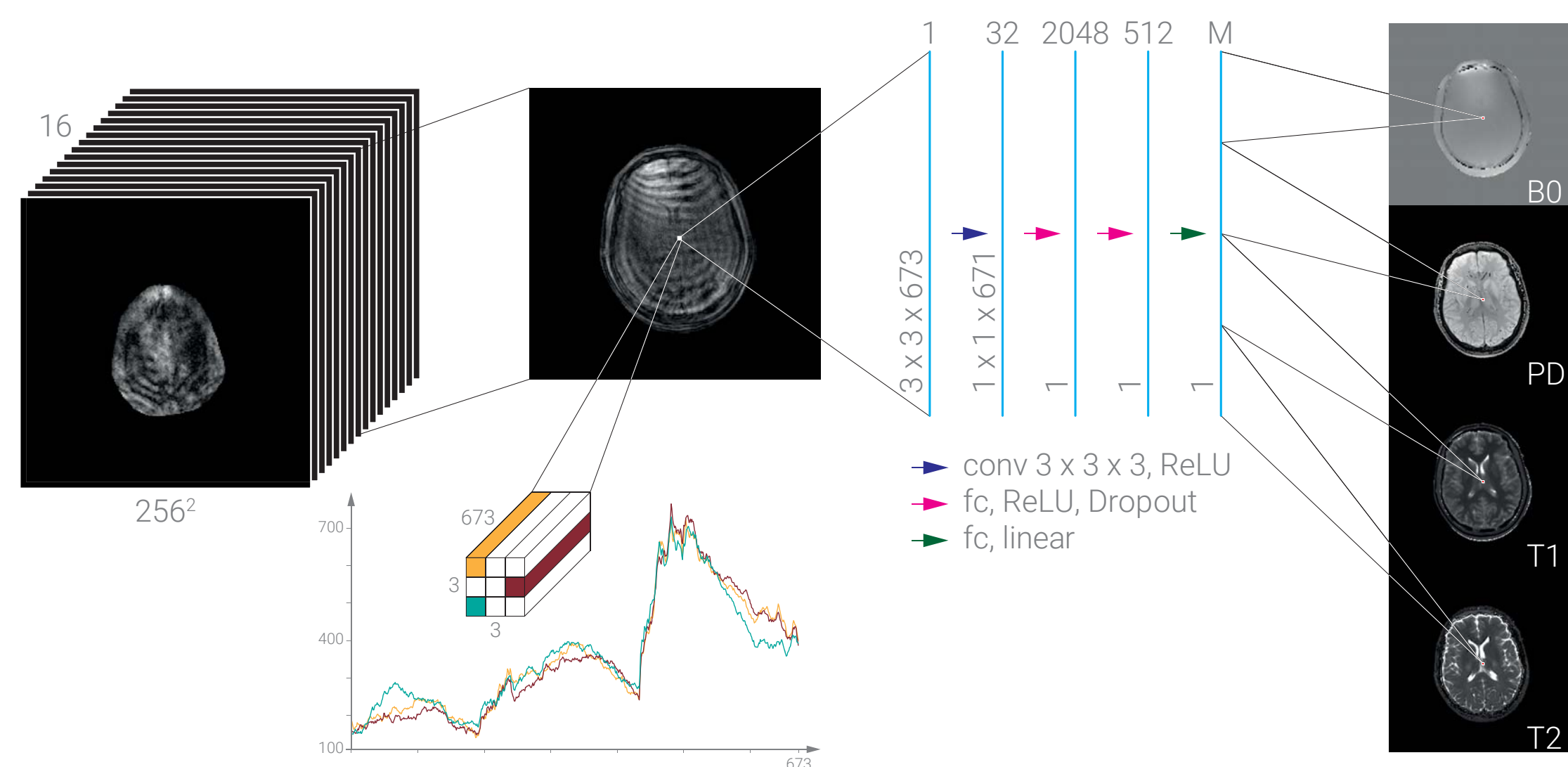
Magnetic resonance fingerprinting (MRF)<sup>1)</sup> quantifies multiple nuclear magnetic resonance parameters in one fast and motion robust acquisition. Standard MRF reconstructs the parametric tissue maps by matching the fingerprints of each voxel to a dictionary of pre-computed signals. This dictionary matching lacks scalability due to computational efficiency and introduces artefacts due to the under-sampled k-space during the acquisition. We propose a dictionary-free, spatiotemporal deep learning-based MRF reconstruction, motivated by the issues associated with dictionary-based reconstruction.

## Materials & Methods

We use a convolutional neural network (CNN) that exploits the spatiotemporal relationship between neighbouring MRF signal evolutions. The CNN consists of one three-dimensional convolutional layer and three fully-connected layers. We acquired four brain scans of healthy volunteers with four parametric maps: static magnetic field inhomogeneity (B<sub>0</sub>), proton density (PD), T<sub>1</sub> relaxation time (T<sub>1</sub>), and T<sub>2</sub> relaxation time (T<sub>2</sub>). We evaluate our approach using a leave-one-out cross-validation and compare to a dictionary-based<sup>2)</sup> and a dictionary-free<sup>3)</sup> reconstruction approaches, which represent the state-of-the-art reconstruction approaches.

## Results

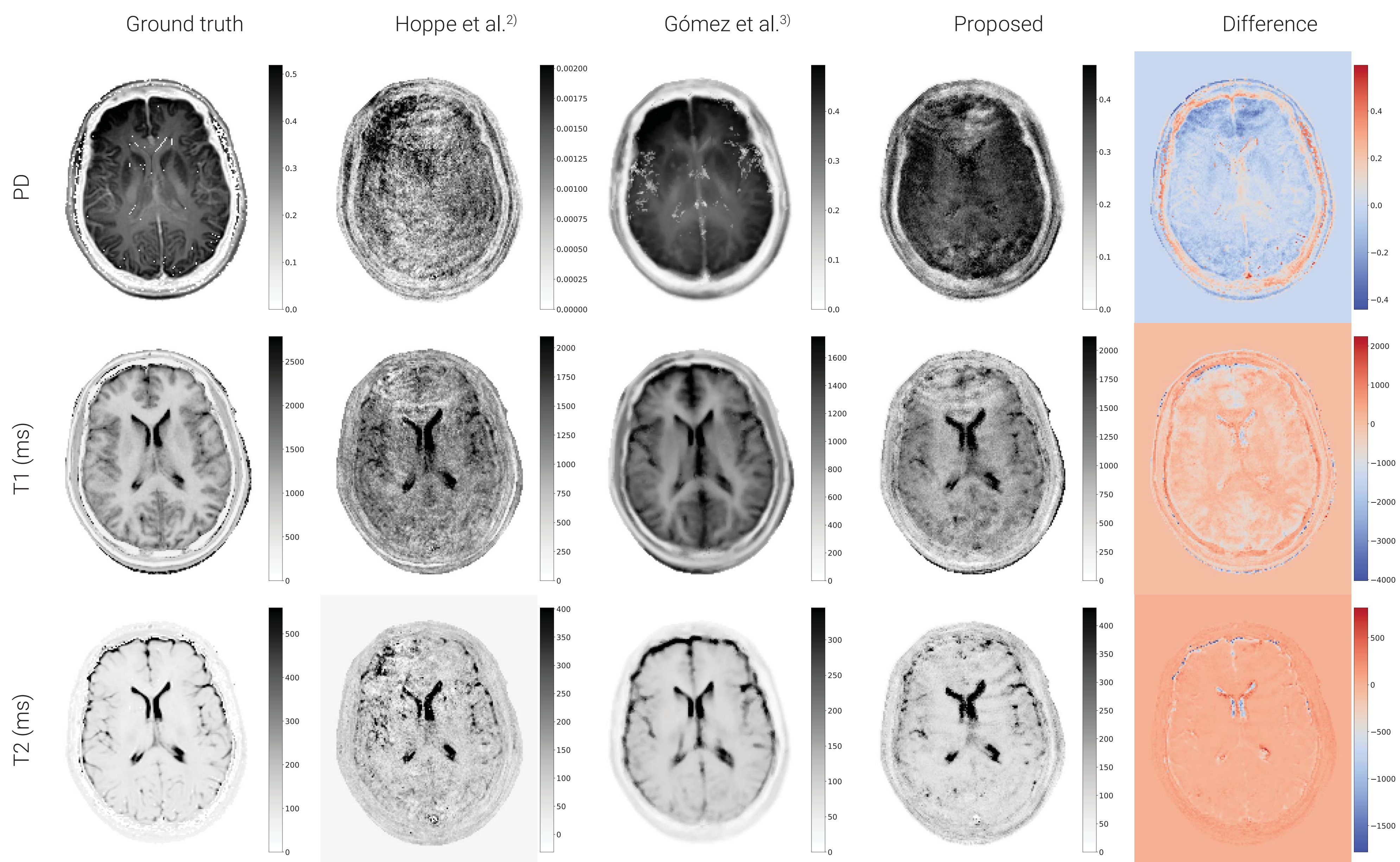
We achieve state-of-the-art reconstruction accuracy and robustness compared to the dictionary-free approach while the dictionary-based approach remains challenging to compete. The root mean square errors for B<sub>0</sub>, PD, T<sub>1</sub>, and T<sub>2</sub> reconstruction are 233.9 Hz, 0.058, 593.4 ms, and 144.8 ms. Compared to 79.2 Hz, 0.052, 485.0 ms, and 128.5 ms for the dictionary-based method.



The reconstruction of the four parametric maps requires several seconds with our method while the dictionary-based method requires several minutes.

## Discussion & Conclusion

We developed a dictionary-free and deep learning-based approach to reconstruct parametric maps from MRF. We exploit the spatiotemporal relationship between neighbouring fingerprints, which yields a more accurate and robust reconstruction compared to other dictionary-free reconstructions. Our approach reconstructs the parametric maps in significantly reduced time compared to the dictionary-based approach. These preliminary results suggest that dictionary-free MRF reconstruction is possible and might be a solution to a clinical application of MRF.



## References

- 1) Ma D. et al. Magnetic Resonance Fingerprinting. Nature. 2013;495(7440):187-192.
- 2) Hoppe E. et al. Deep Learning for Magnetic Resonance Fingerprinting: A New Approach for Predicting Quantitative Parameter Values from Time Series. 2017.
- 3) Gómez P.A. et al. Simultaneous Parameter Mapping, Modality Synthesis, and Anatomical Labeling of the Brain with MR Fingerprinting. MICCAI. 2016:579-586.

## Acknowledgments

This work was supported by the Swiss Foundation for Research on Muscle Diseases (ssem).  
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