

Enhancing the speed and robustness of magnetic resonance imaging applications using deep learning

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Centre for Innovation in Biomedical Imaging Technology





Content





Content



DATA AVAILABILITY STATEMENT

All data and code used in this paper can be found here: https://www.dropbox.com/c/iikxmoor/088e4iu/ matlab__________3. Sharing deep learning ________________zip?dl=0



What is the scientific problem?

• Is this lesion calcified or bleeding?

• On a standard MRI scan we only

see a black spot in both cases \otimes





Quantitative susceptibility mapping (QSM) can help!



QSM differentiates between blood products (Hyperintense=white) and calcifications (Hypointense=black).





Review: Deistung et al. NMR Biomed 2017; Schweser et al. Z Med Phys 2016





Neurolmage Volume 195, 15 July 2019, Pages 373-383



DeepQSM - using deep learning to solve the dipole inversion for quantitative susceptibility mapping

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Generate data and train to solve inverse problem





Test on simple shapes – similar to training data





Generalizes to in vivo brain data ③





25 min compute time 1 min compute time



Problem: Sensitivity to input data distribution

Adding noise to the input data leads to failure of network prediction





Variational Network: Combine traditional optimization + DL





Robustness to noise due to data consistency constrain:



Adding noise to the input data



Content



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MAGNETIC RESONANCE IN MEDICINE



FULL PAPER

Improving FLAIR SAR efficiency at 7T by adaptive tailoring of adiabatic pulse power through deep learning ^B⁺ estimation

Shahrokh Abbasi-Rad 🗙, Kieran O'Brien, Samuel Kelly, Viktor Vegh, Anders Rodell, Yasvir Tesiram, Jin Jin, Markus Barth, Steffen Bollmann

First published: 23 November 2020 | https://doi.org/10.1002/mrm.28590



Background and Motivation

In MRI we are limited by the **Specific Absorption Rate (**SAR)



scans take longer to limit heating effects





Background and Motivation

- with *a-priori* knowledge of the 3D B1 field, SAR could be reduced by saving SAR where we don't need it
- But: B1-map requires additional scan time (~ 2min)
- Can we predict an approximate B1-map from the localizer scan's B1 artifact?

Localizer/Scout 15s



?

B1-Map 2min





2min B1-Map

Methods

15s Localizer/Scout



Train Data: 7000 pairs of patches Test Data: 1000 pairs of patches Epochs: 200 Learning rate: 0.02



Results



not perfect, but good enough - 5 % error



Don't trust a prediction you haven't checked yourself...

We check if the prediction delivers good results in practice using the underlying physics relationship:





SAR Reduction Experiment

3 min 18 sec (FLAIR) + 1min 31 sec (Delay)

2 min (B1 map) + 3 min 18 sec (FLAIR) + 0s Delay

3 min 18 sec (FLAIR)



Non-Scaled (SAR = 140%) Scaled (Measured) (SAR = 97%) Scaled (CNN) (SAR = 97%)



Content



DATA AVAILABILITY STATEMENT

All data and code used in this paper can be found here: https://www.dropbox.com/s/iikymony088e4iu/ 3. Sharing deep learning models



Problem: Sharing models and code with readers



Why?

- Reviewers are sceptical about deep learning and would like to understand why it works and when it fails
- Readers of the paper should be able to reproduce the results in the paper and modify the presented concepts

BUT:

• Deep learning toolkits not easy to install and require GPUs ...



Download 2GB

file just to see

which functions

were used?

Open Science is easy, right? What if this free service stops? DATA AVAILABILITY STATEMENT All data and code used in this paper can be found here: https://www.dropbox.com/s/ijkxmopv088e4iu/ matlab_code_data_mri_paper2021.zip?dl=0. Will the code still I found a bug and work in the next need to update version? Does the this reader have a matlab license?

24



Data upload via command line.



osfclient



DATA AVAILABILITY STATEMENT

We facilitate the reproducibility of our study by providing an interactive version of our implementation on a publicly accessible cloud-based platform. The readers can explore the implementation of the model (neural network), train the model with different hyper-parameters and architectures, investigate the stability of the training process, and reproduce our results with the identical model used in this manuscript (https://github.com/sbollmannMRI/scout2B1, 320a6ab). We anonymized and stored the input data (localizer, SA2RAGE B_1^+) of 28 participants in OSF (OSF, Center for Open Science, Inc., Virginia, USA) accessible via https://osf.io/y5cq9/.



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Data download within Jupyter notebook.





GitHub

Provide source code in an easy accessible way

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This commit was used for the paper, but bug fixes possible



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Creating snapshots for version control.



Platform can be changed: link in OSF

Provide source code in an easy accessible way

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Interactively running in browser – no setup needed

This commit was used for the paper, but bug fixes possible

Data and links can be updated if bugs found or services move



MRM Society Journal Highlight



Home > Highlights

Reproducible Research Insights with Shahrokh Abbasi-Rad, Markus Barth, and Steffen Bollmann



April 30, 2021



By Mathieu Boudreau

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The April 2021 MRM Highlights Reproducible Research Insights interview is with Shahrokh Abbasi-Rad, Markus Barth and Steffen Bollmann, researchers at The University of Queensland, Brisbane, Australia. Their

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Thank you

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CRICOS code 00025B

