

Self-gated self-supervised ADMM unrolling enables mesoscale high-resolution motion-robust diffusion-weighted imaging

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Declaration of Financial Interests or Relationships

Z Tan:

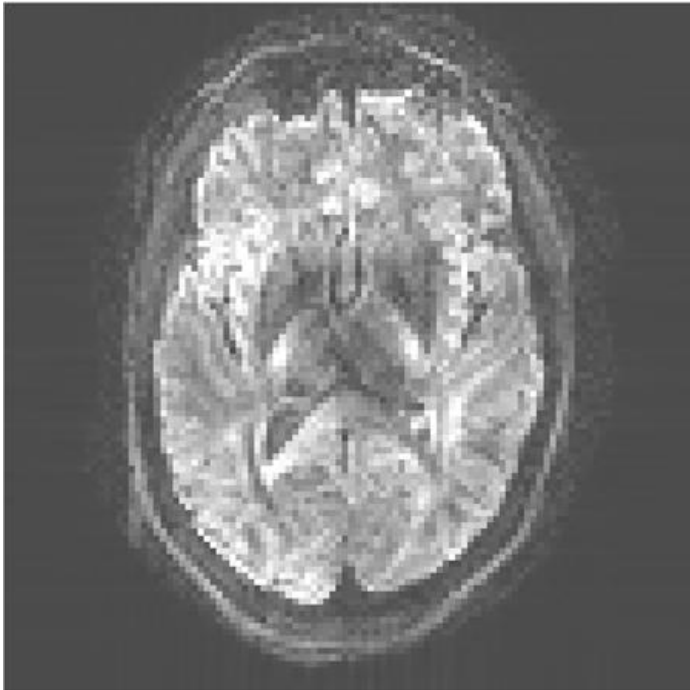
I have no financial interests or relationships to disclose with regard to the subject matter of this presentation.

INTRODUCTION

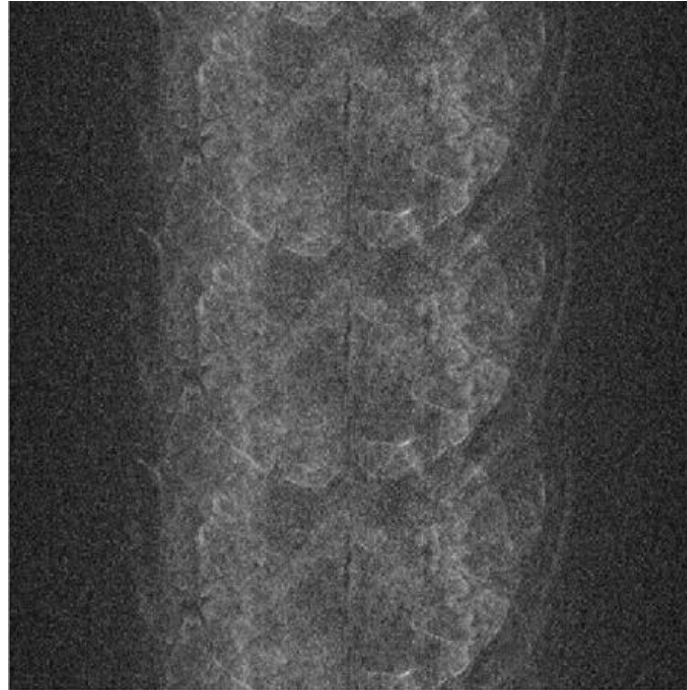
Diffusion-Weighted Imaging (DWI) ¹

- Powerful tool for acute stroke diagnosis, fiber tracking, cancer staging, ...
- Challenges:

Low spatial resolution



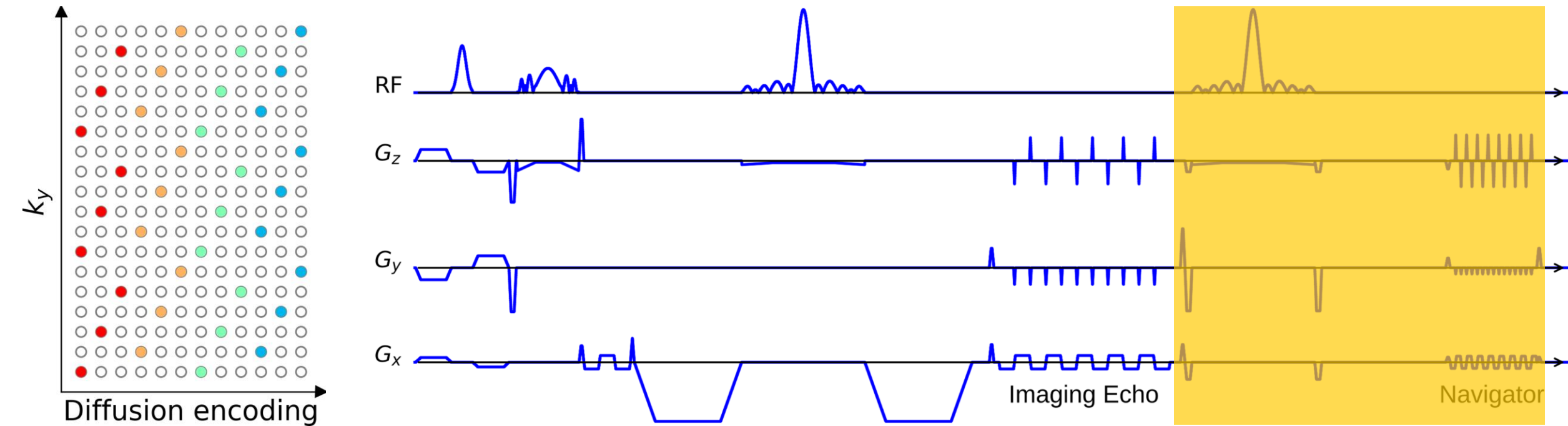
Shot-to-shot phase variation



Motion



Previous Work: Mesoscale High-Resolution DWI Acquisition with NAViEPI ²



Brain DWI Protocol: 0.7 mm ISO

Shots: 3; Acceleration: 2 x 2; Matrix size: 286 x 286 x 176; 1 x b0 & 20 x b1000; TEs/TR: 58/98/15000 ms; TA: 16 minutes.

TA reduced to 10 minutes w/o navigators.

Question: **Selfgated reconstruction for accelerated DWI?**

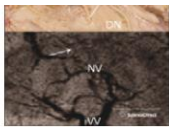
State-of-the-Art Selfgated DWI Reconstruction

- Multiplexed Sensitivity-Encoding (**MUSE**): simultaneous multi-slice (SMS) multi-shot SENSE



NeuroImage

journal homepage: www.elsevier.com/locate/ynimg



Technical Note

A robust multi-shot scan strategy for high-resolution diffusion weighted MRI enabled by multiplexed sensitivity-encoding (MUSE)

Nan-kuei Chen ^{*}, Arnaud Guidon, Hing-Chiu Chang, Allen W. Song

Brain Imaging and Analysis Center, Duke University Medical Center, Box 2737, Hock Plaza, Durham, NC 27710, USA

- **Low Rankness**

> **MUSSELS**: low-rankness among shots

Magnetic Resonance in Medicine 78:494–507 (2017)

Multi-Shot Sensitivity-Encoded Diffusion Data Recovery Using Structured Low-Rank Matrix Completion (MUSSELS)

Merry Mani,¹ Mathews Jacob,^{2*} Douglas Kelley,³ and Vincent Magnotta^{1,4,5}

> **LLR**: Locally low rank of spatial-diffusion matrices

Magnetic Resonance in Medicine

FULL PAPER

Multi-shot diffusion-weighted MRI reconstruction with magnitude-based spatial-angular locally low-rank regularization (SPA-LLR)

Yuxin Hu^{1,2} | Xiaole Wang¹ | Qiyuan Tian^{1,2} | Grant Yang^{1,2} |
Bruce Daniel^{2,3} | Jennifer McNab² | Brian Hargreaves^{1,2,3}

Inspiration: Self2Self & Zero-Shot Learning

- **Self2Self** ³

A denoising neural network can be learned from one single image.

- **Zero-shot learning** ^{4,5}

The regularization in iterative reconstructions can be learned without full-sampled k-space data.

- **Question: Self-supervised learning for high-resolution joint DWI reconstruction?**

3. Quan Y, et al. Self2Self with dropout: Learning self-supervised denoising from single image. *CVPR* (2020).

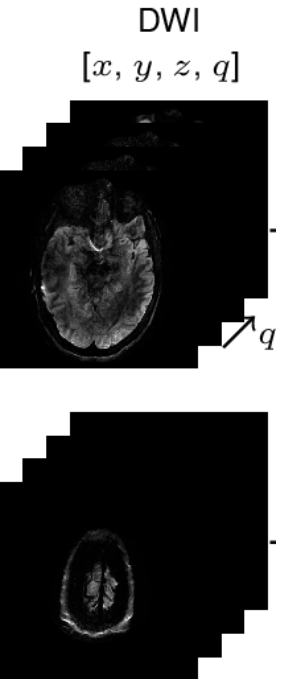
4. Yaman B, et al. Zero-shot self-supervised learning for MRI reconstruction. *ICLR* (2022).

5. Heydari A, et al. Joint MAPLE: Accelerated joint T1 and T2* mapping with scan specific self-supervised networks. *Magn Reson Med* (2024).

METHODS

Joint k-q-Slice Reconstruction

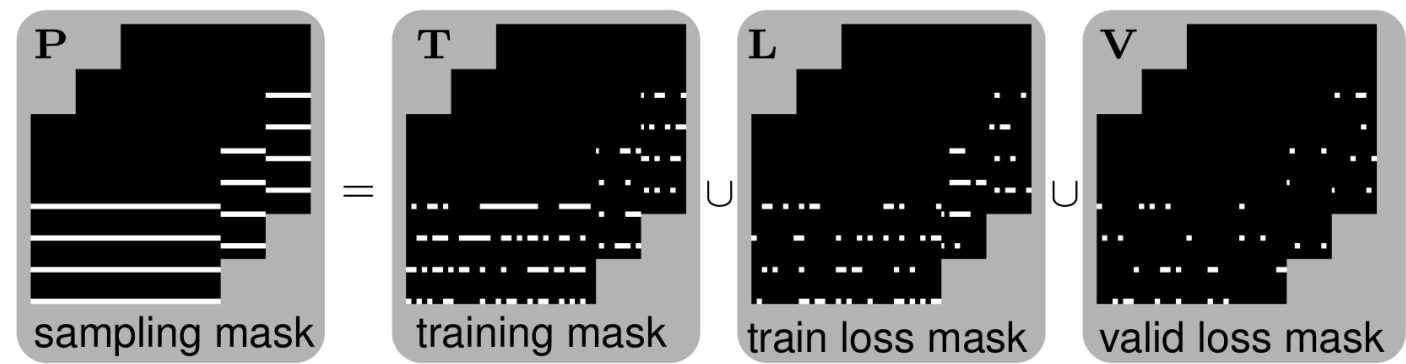
- Joint k-q-slice forward model (github.com/ZhengguoTan/NAViEPI)



$$\operatorname{argmin}_{\mathbf{x}} \|\mathbf{Ax} - \mathbf{y}\|_2^2 + \lambda \mathcal{D}_{\omega}(\mathbf{x})$$

Self-Supervised ADMM Unrolling Reconstruction

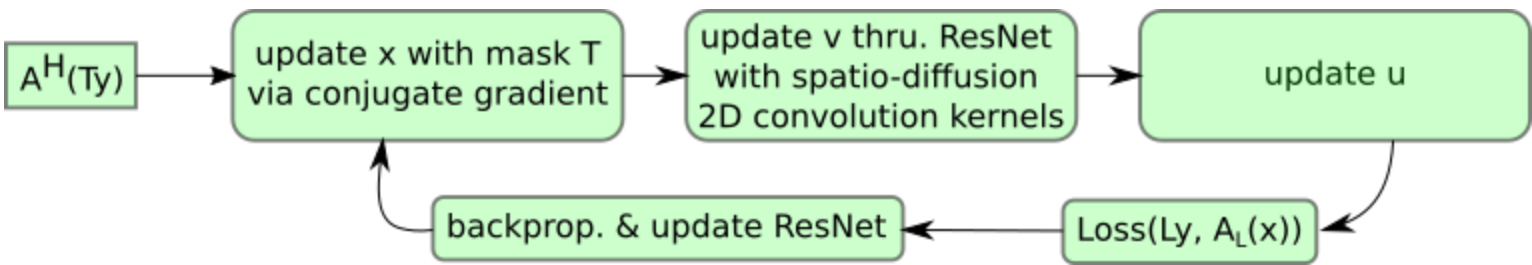
(a) sampling mask splitting



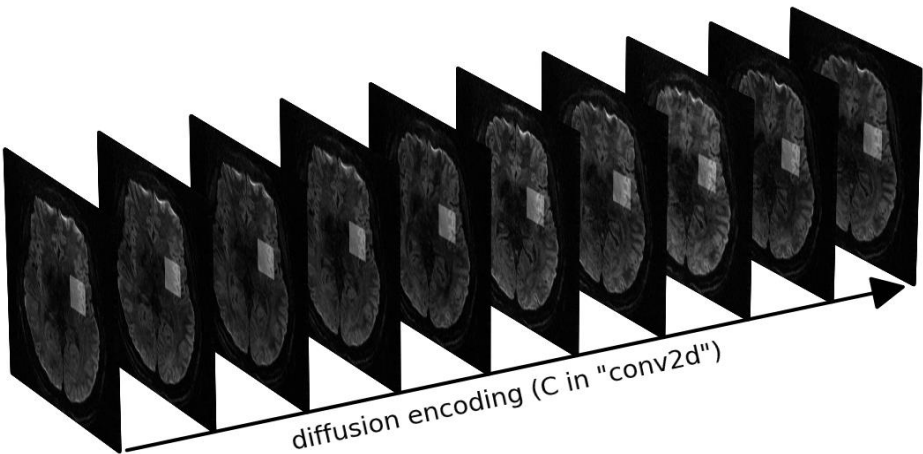
(b) ADMM ⁶ unrolling

$$\begin{cases} \mathbf{x}^{(k+1)} = \underset{\mathbf{x}^{(k)}}{\operatorname{argmin}} \left\| \mathbf{y} - \mathcal{A}(\mathbf{x}^{(k)}) \right\|_2^2 \\ \quad + \frac{\rho}{2} \left\| \mathbf{x}^{(k)} - \mathbf{v}^{(k)} + \mathbf{u}^{(k)} \right\|_2^2 \\ \mathbf{v}^{(k+1)} = (\lambda/\rho) \cdot \mathcal{D}_\omega(\mathbf{x}^{(k+1)} + \mathbf{u}^{(k)}) \\ \mathbf{u}^{(k+1)} = \mathbf{u}^{(k)} + \mathbf{x}^{(k+1)} - \mathbf{v}^{(k+1)} \end{cases}$$

(c) training



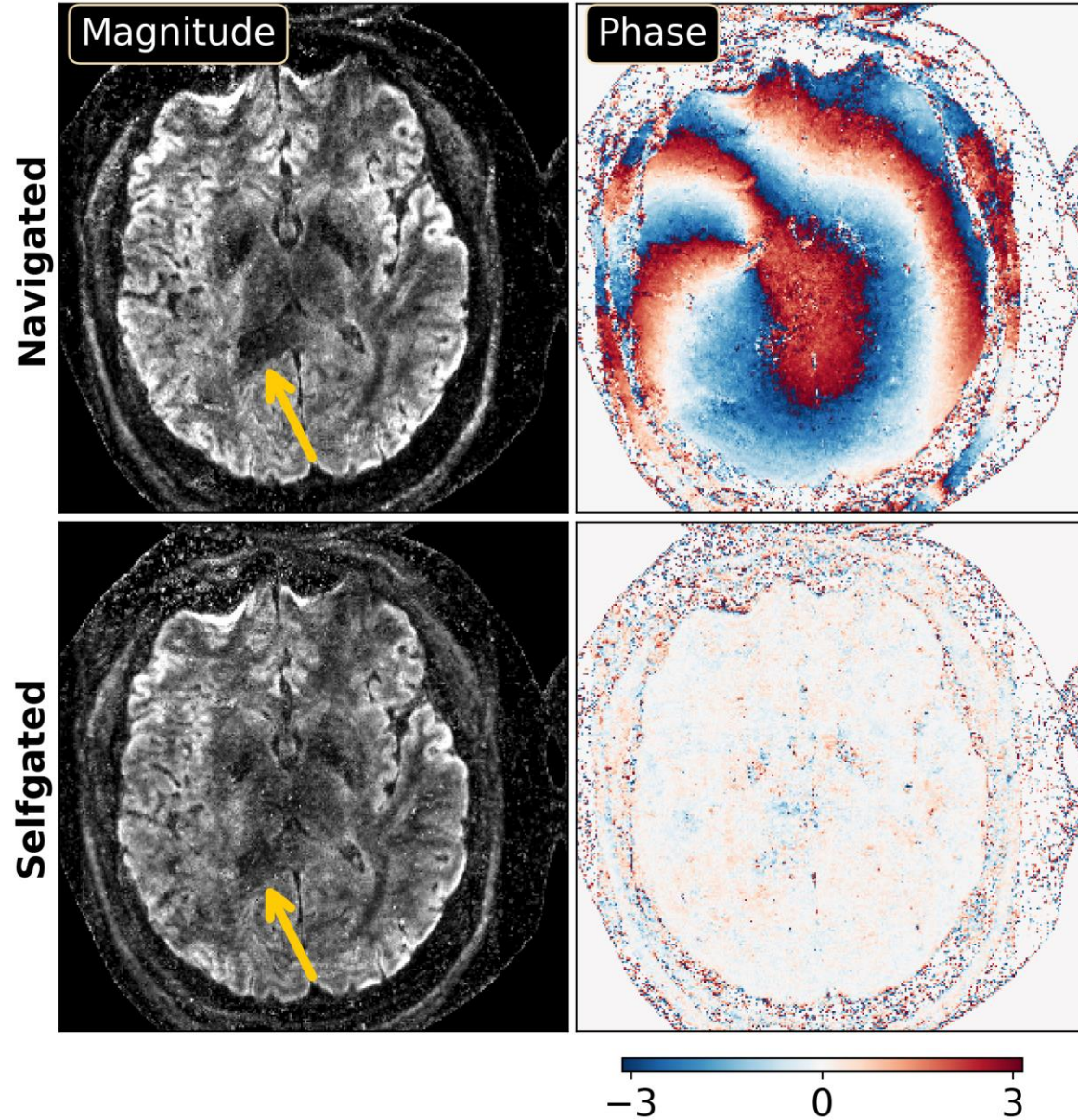
(d) spatial-diffusion convolution



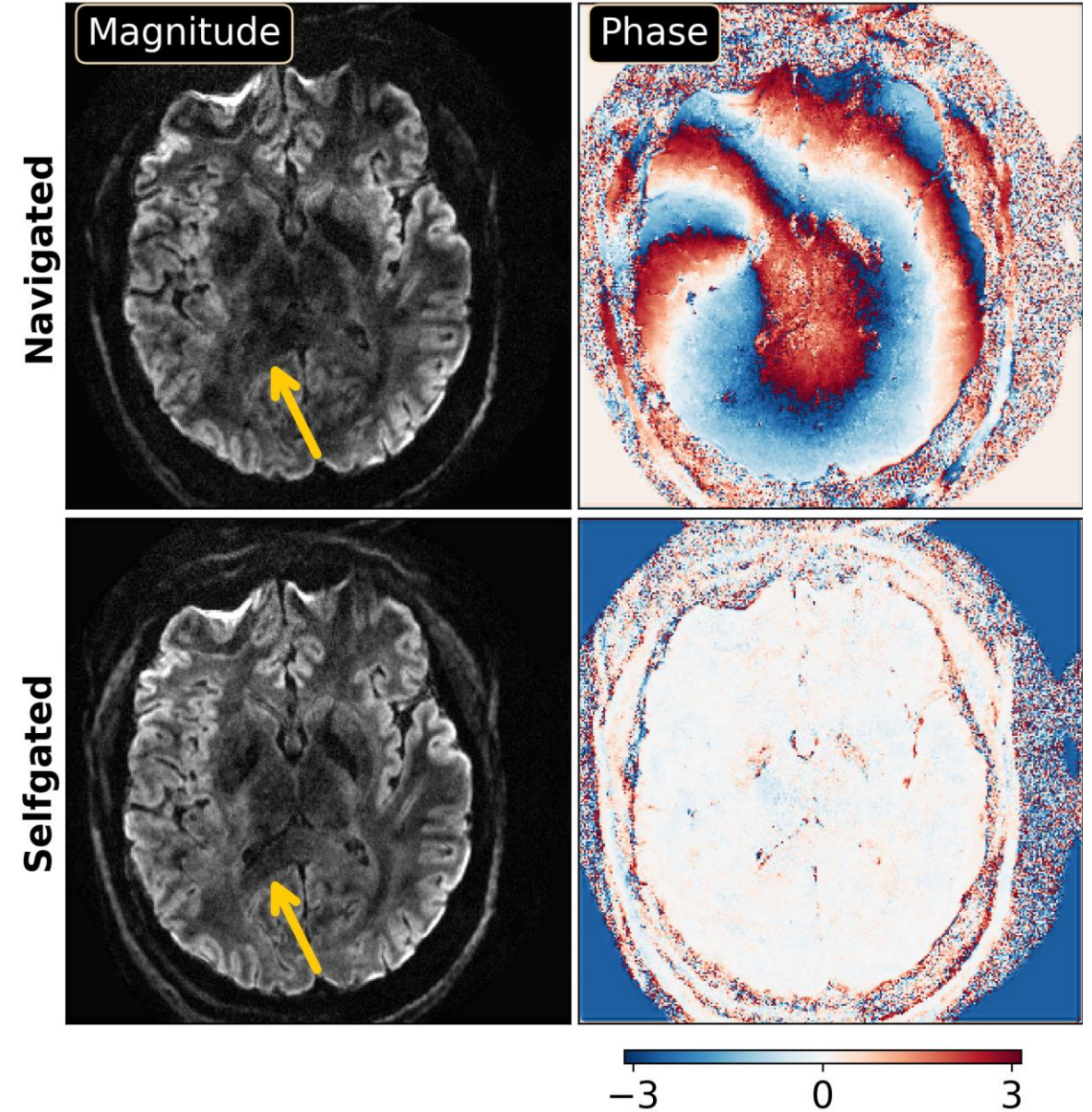
RESULTS

Joint k-q-Space ADMM Unroll Enables Selfgated DWI Reconstruction

MUSE + Denoiser

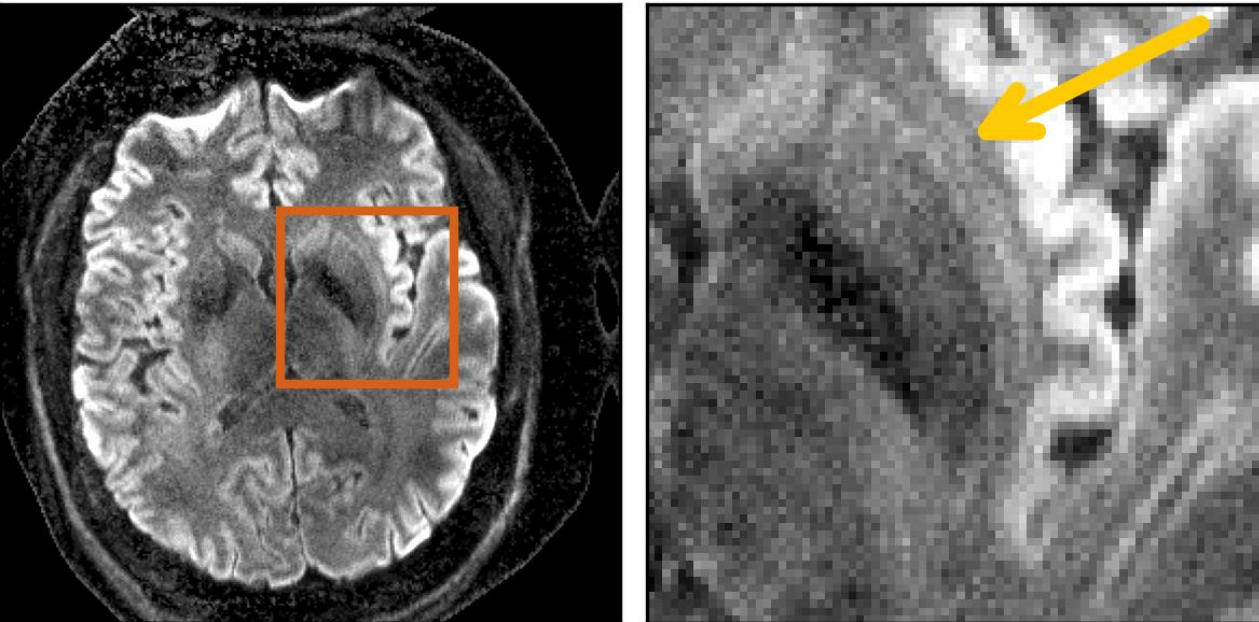


ADMM Unroll

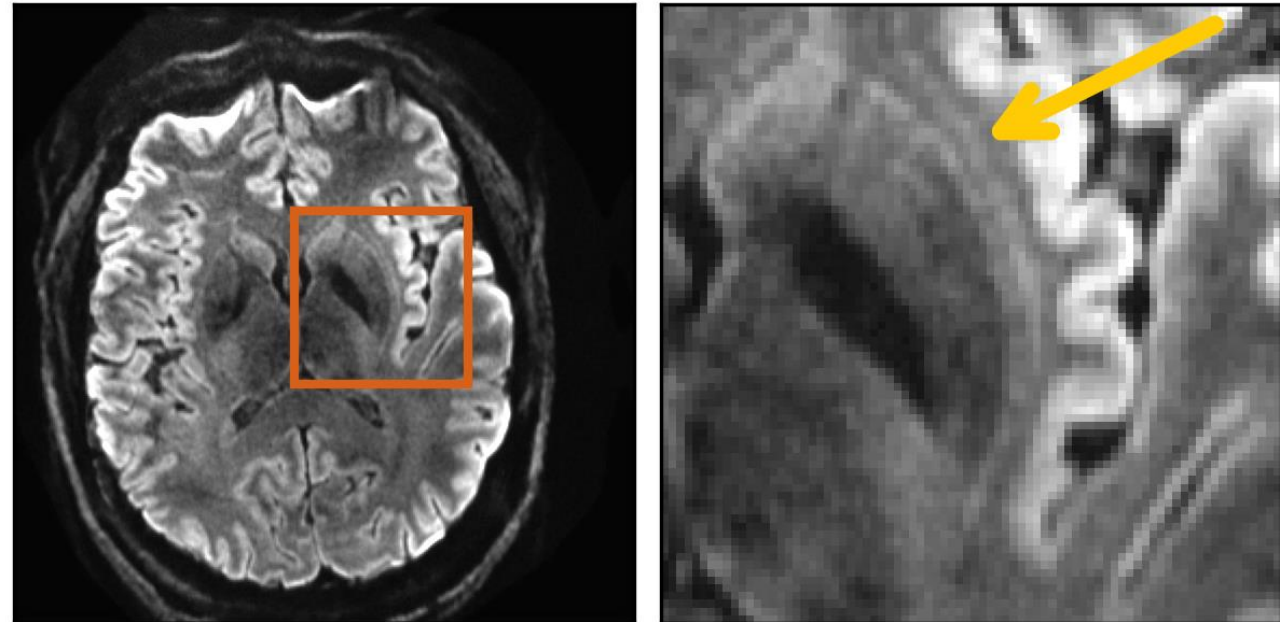


0.7 mm DWI Enables Visualization of the "Hidden-Away" Claustrum

Navigated MUSE + Denoiser



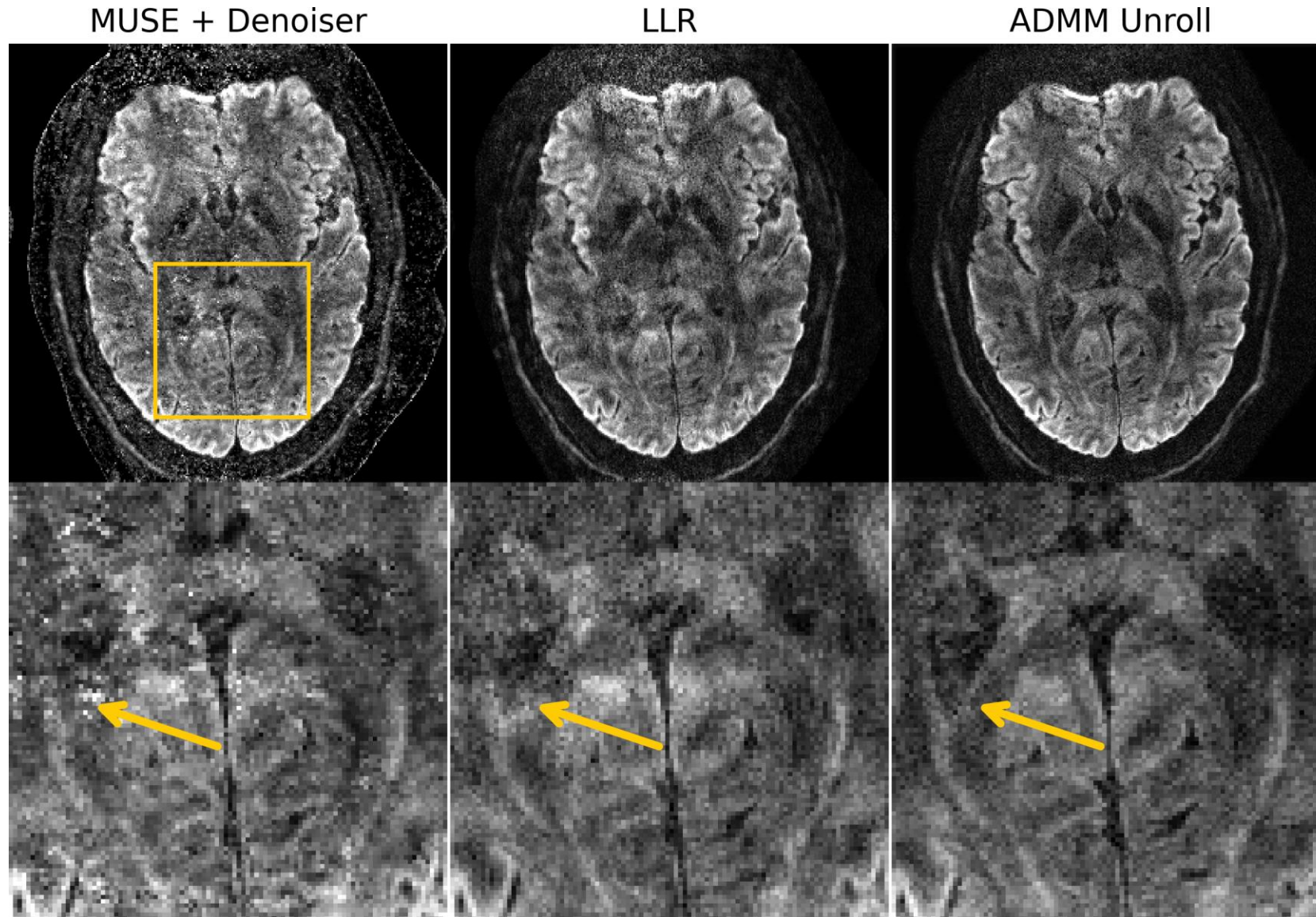
Selfgated ADMM Unroll



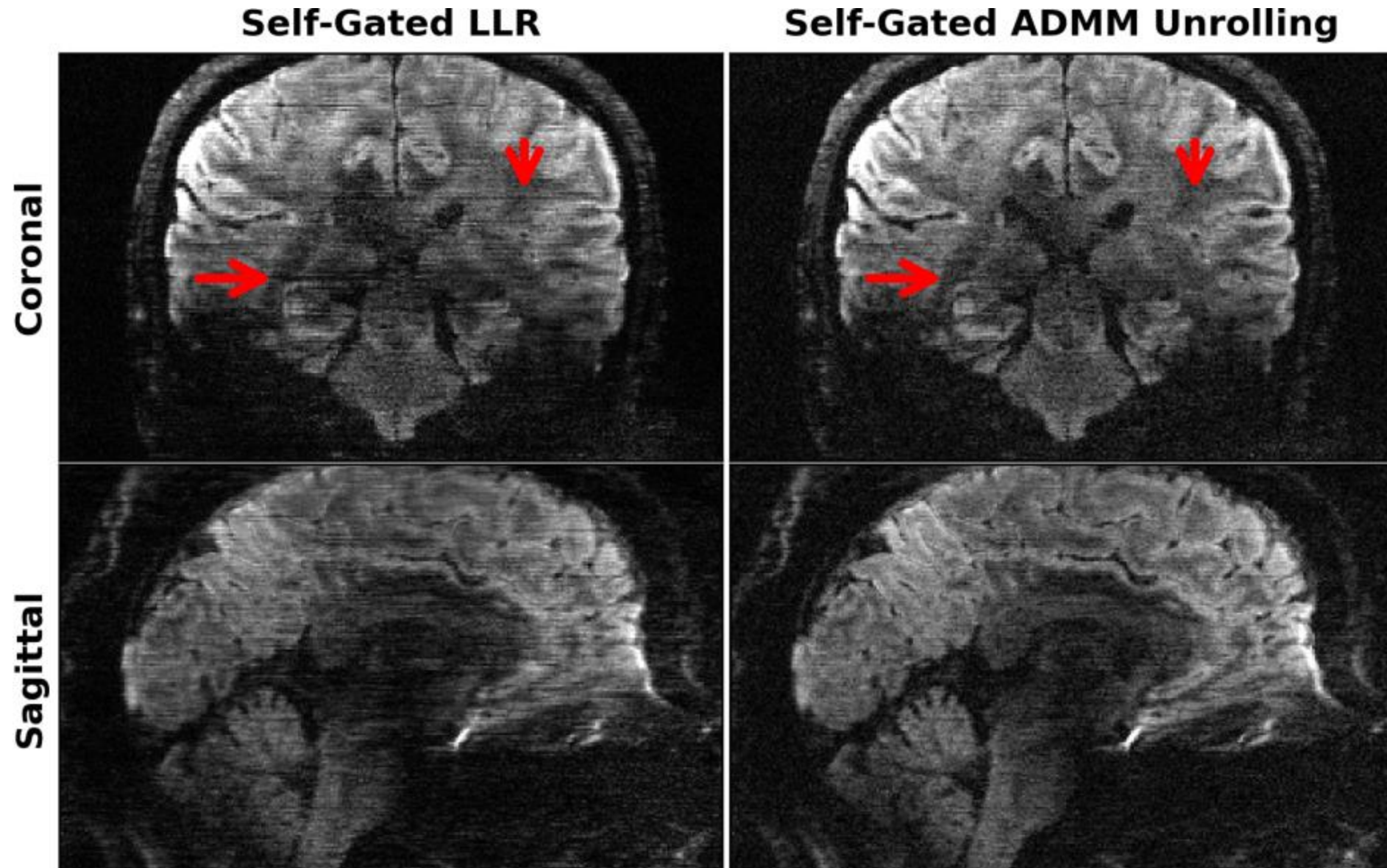
- Mean diffusion-weighted images from 20 directions at $b = 1000 \text{ s/mm}^2$ are displayed

Joint k-q-Space ADMM Unroll Enables Selfgated DWI Reconstruction

0.7 mm ISO with 150 slices and 21 diffusion volumes with $b = 1000 \text{ s/mm}^2$ at 10 minutes



Motion Robustness



SUMMARY

Summary

- **Self-supervised ADMM unrolling enables mesoscale DWI at short scan time.**

