

On the use of autoencoder to denoise diffusion MRI

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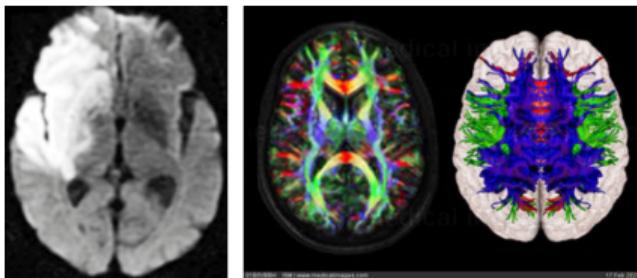
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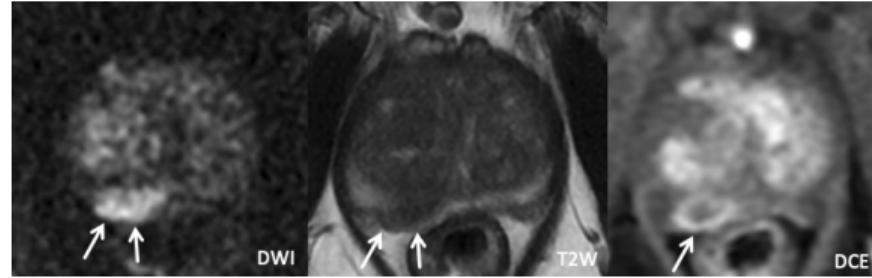
Motivation: In vivo High Resolution Diffusion MRI at 7 T

Various clinical applications of diffusion MRI ^{1, 2, 3, 4}

Brain: stroke & neurodegenerative diseases



Prostate cancer: detection and localization



¹Courtesy: ChatGPT

²<https://mriquestions.com/trace-vs-adc-map.html>

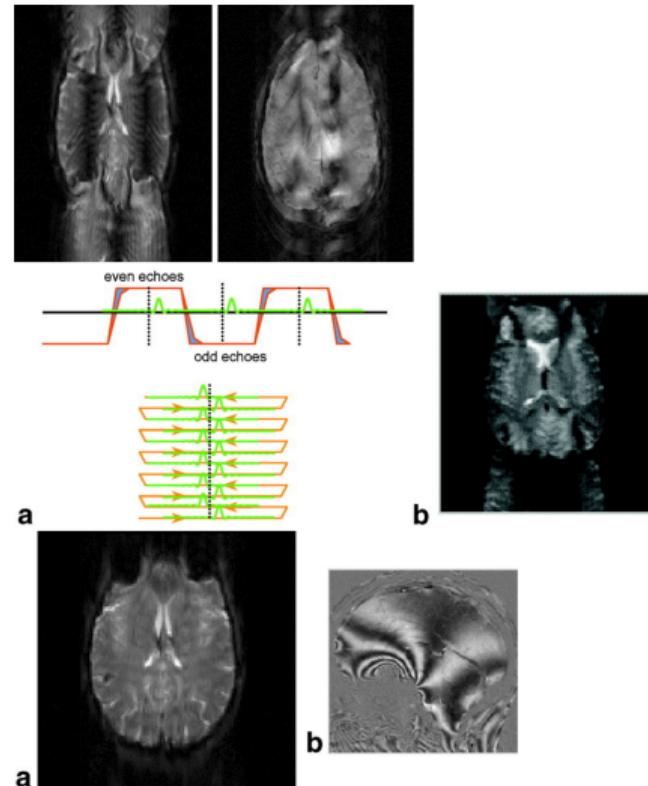
³<https://www.psychologytoday.com/gb/blog/adventures-in-cognition/202102/why-white-matter-matters-diffusion-tensor-imaging>

⁴<https://rad.bwh.harvard.edu/prostate-mri/>

Limitations in diffusion MRI⁵

- ▶ Common pitfalls:
 - ★ motion artifacts
 - ★ eddy currents
 - ★ geometric distortion and susceptibility artifacts

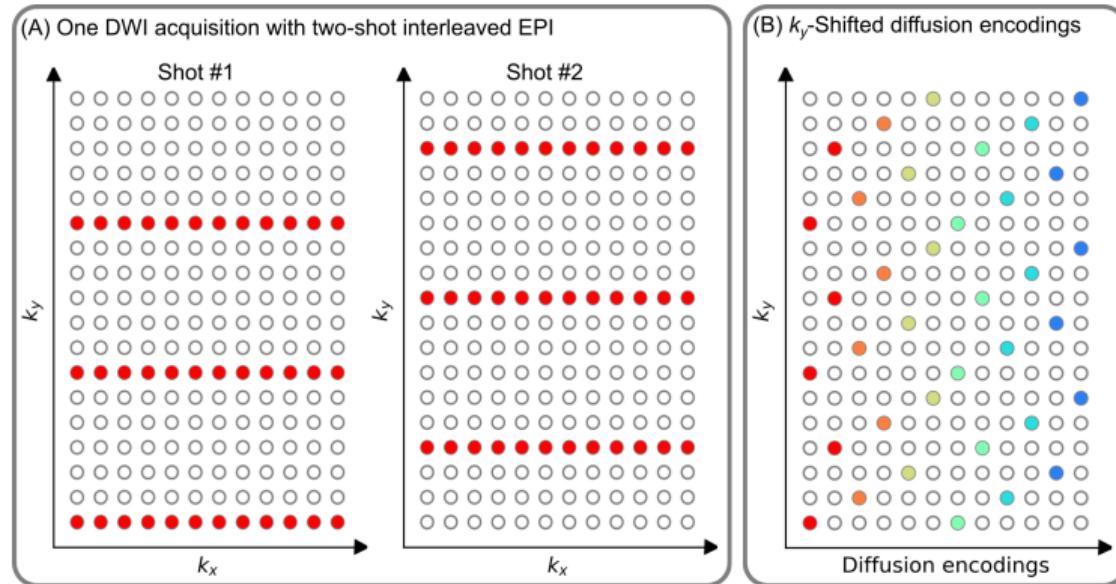
- ▶ Be aware that diffusion-weighted imaging (DWI),
 - ★ primarily uses single-shot echo planar imaging (EPI)
 - ★ clinical spatial resolution of about 2 mm
 - ★ suffers from low signal-to-noise ratio (SNR)



⁵Le Bihan D, et al. Artifacts and pitfalls in diffusion MRI. *J Magn Reson Imaging* (2006).

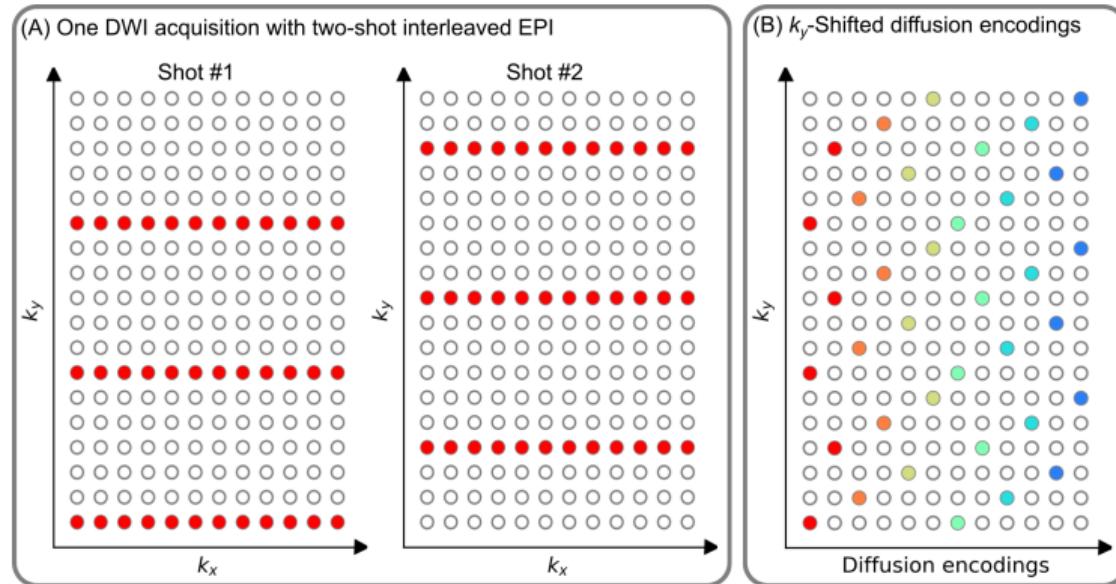
Method

Diffusion MRI Acquisition



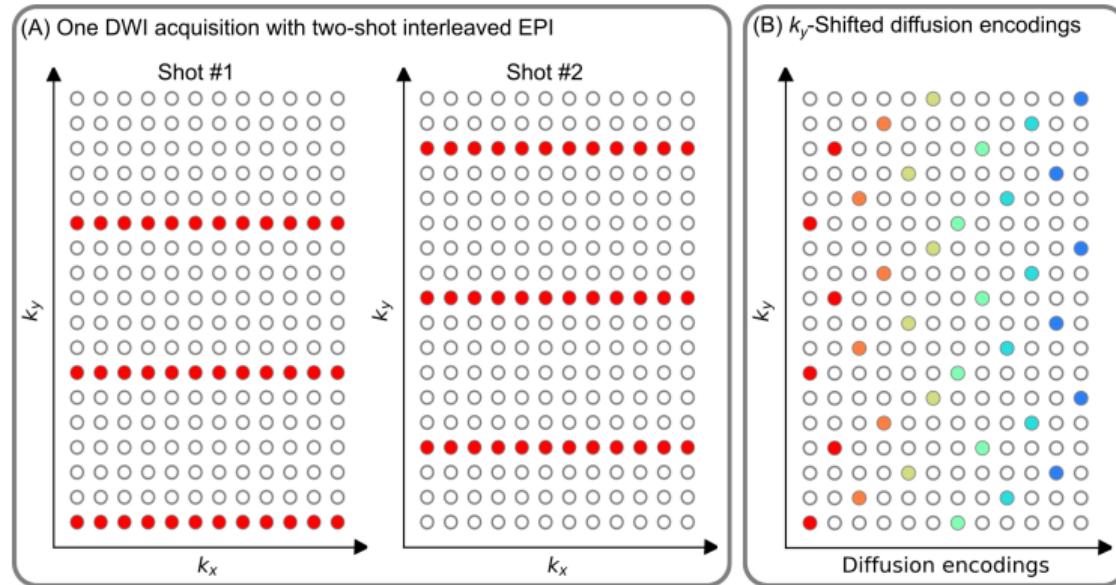
- ▶ 2-shot interleaved EPI with multi-band factor 2, yielding acceleration factor 6×2

Diffusion MRI Acquisition



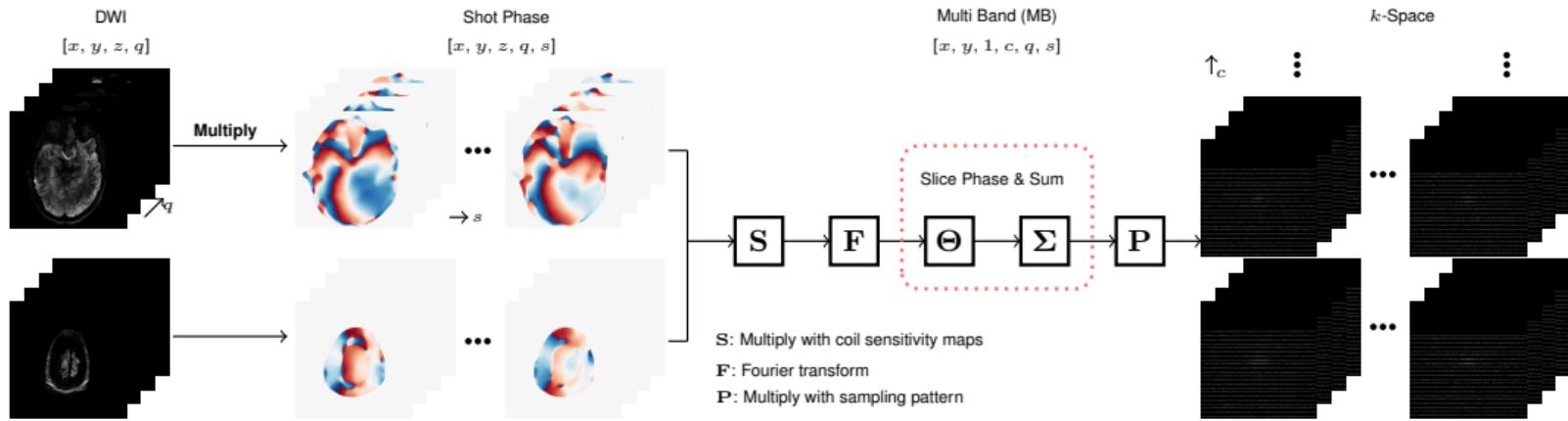
- ▶ 2-shot interleaved EPI with multi-band factor 2, yielding acceleration factor 6×2
- ▶ k_y -shift encoding to accomplish complementary k - q -space sampling

Diffusion MRI Acquisition



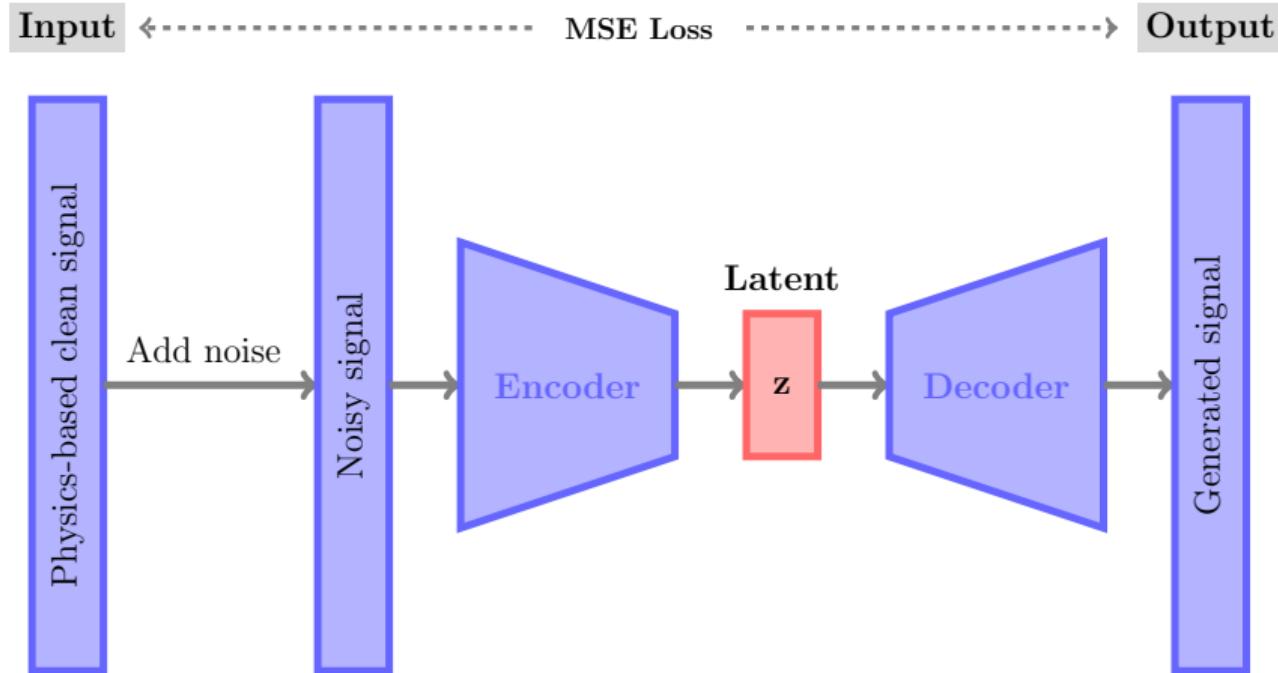
- ▶ 2-shot interleaved EPI with multi-band factor 2, yielding acceleration factor 6×2
- ▶ k_y -shift encoding to accomplish complementary k - q -space sampling
- ▶ 1.2 mm (isotropic), $2 \times b_0 + 30$ diffusion encodings with b -value 1000 s mm^{-2}

Diffusion MRI Forward Modeling and Image Reconstruction⁶



⁶Tan Z, et al. under review

Denoise Diffusion MRI Using Autoencoder^{7, 8, 9}



⁷Hinton GE, Salakhutdinov RR. Reducing the dimensionality of data with neural networks. *Science* (2006).

⁸Lam F, et al. Constrained magnetic resonance spectroscopic imaging by learning nonlinear low-dimensional models. *IEEE Trans Med Imaging* (2019).

⁹Mani M, et al. qModeL: A plug-and-play model-based reconstruction for highly accelerated multi-shot diffusion MRI using learned priors. *Magn Reson Med* (2021).

Denoise Diffusion MRI Using Autoencoder: How to Train the Model

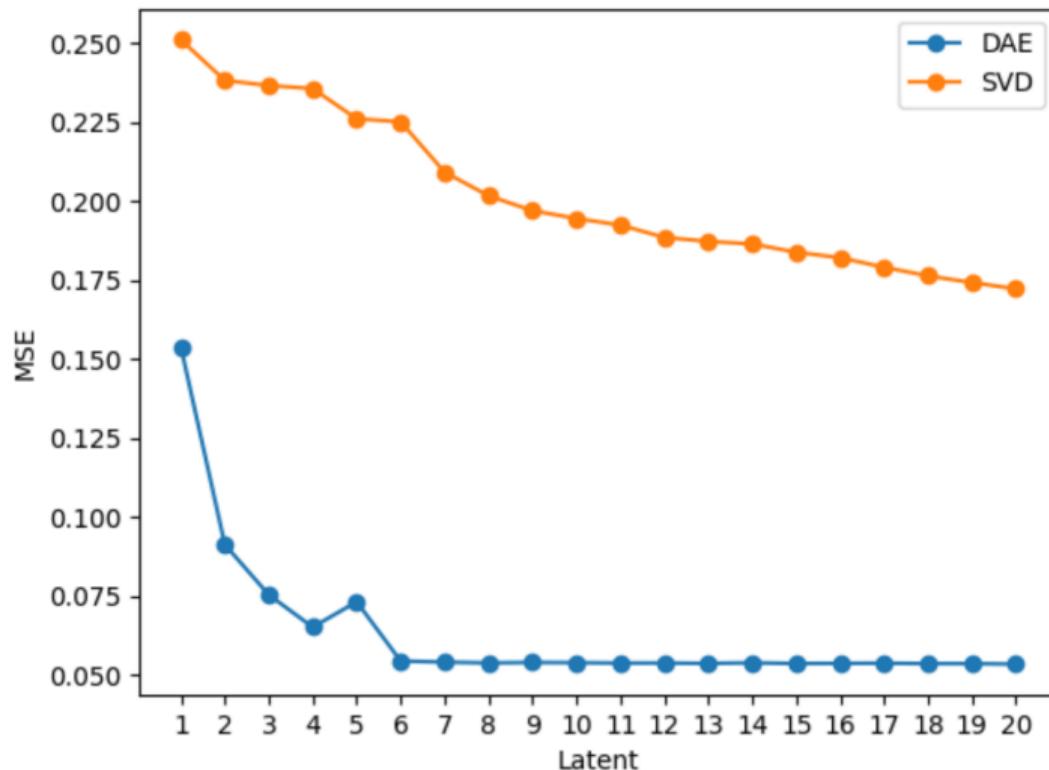
- ▶ Data generation:
 1. Simulation of DWI signal based on the DTI model ¹⁰
 2. Each tensor element contains 10 entries, resulting in 10^6 DWI signal entries
 3. Zero-mean Gaussian-distribution noise with 10 varying standard deviations are added to the simulated signal, yielding 10^7 DWI signal
 4. 80 % of the data for training, and 20 % for testing
- ▶ Training setup:
 1. Optimizer: ADAM ¹¹
 2. Learning rate: 0.0001
 3. Epochs: 800
 4. 4 fully-connected encoder/decoder layers with ReLU, whereas the output layer uses Sigmoid as the activation function
 5. Latent size: 10

¹⁰Basser PJ, et al. MR diffusion tensor spectroscopy and imaging. *Biophys J* (1994).

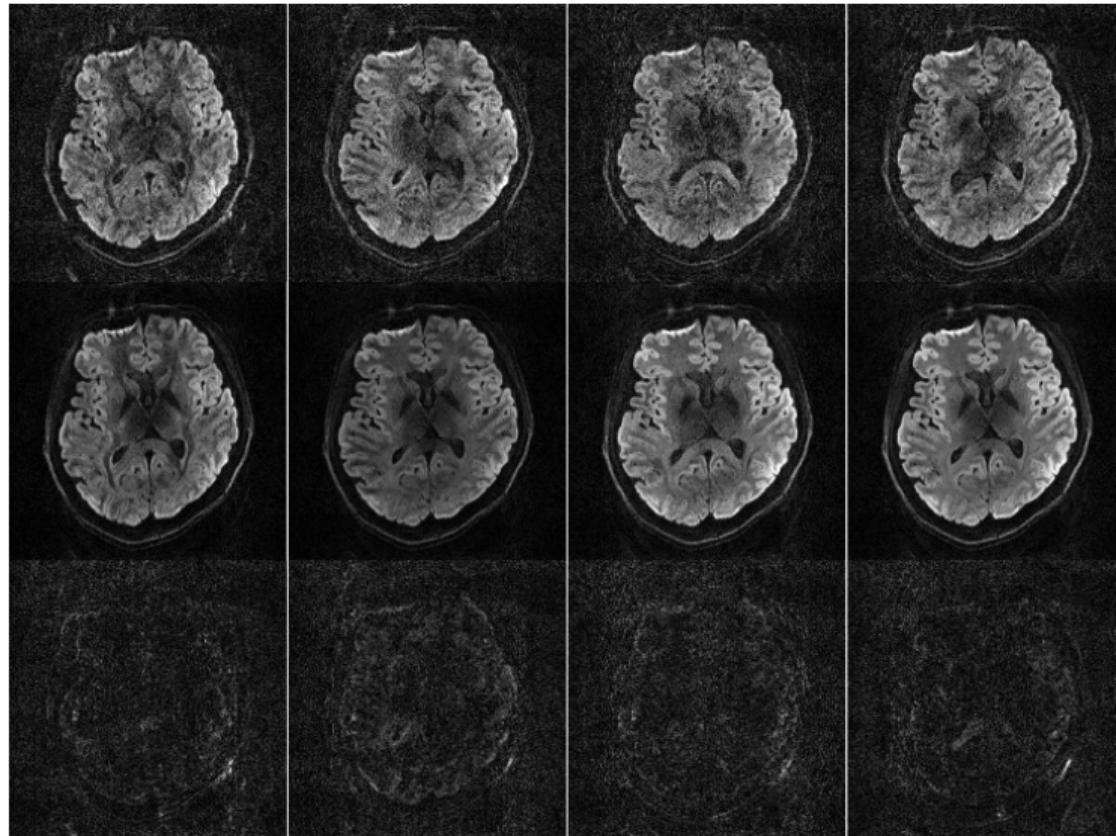
¹¹Kingma DP, et al. Adam: A Method For Stochastic Optimization. *ICLR* (2015).

Results

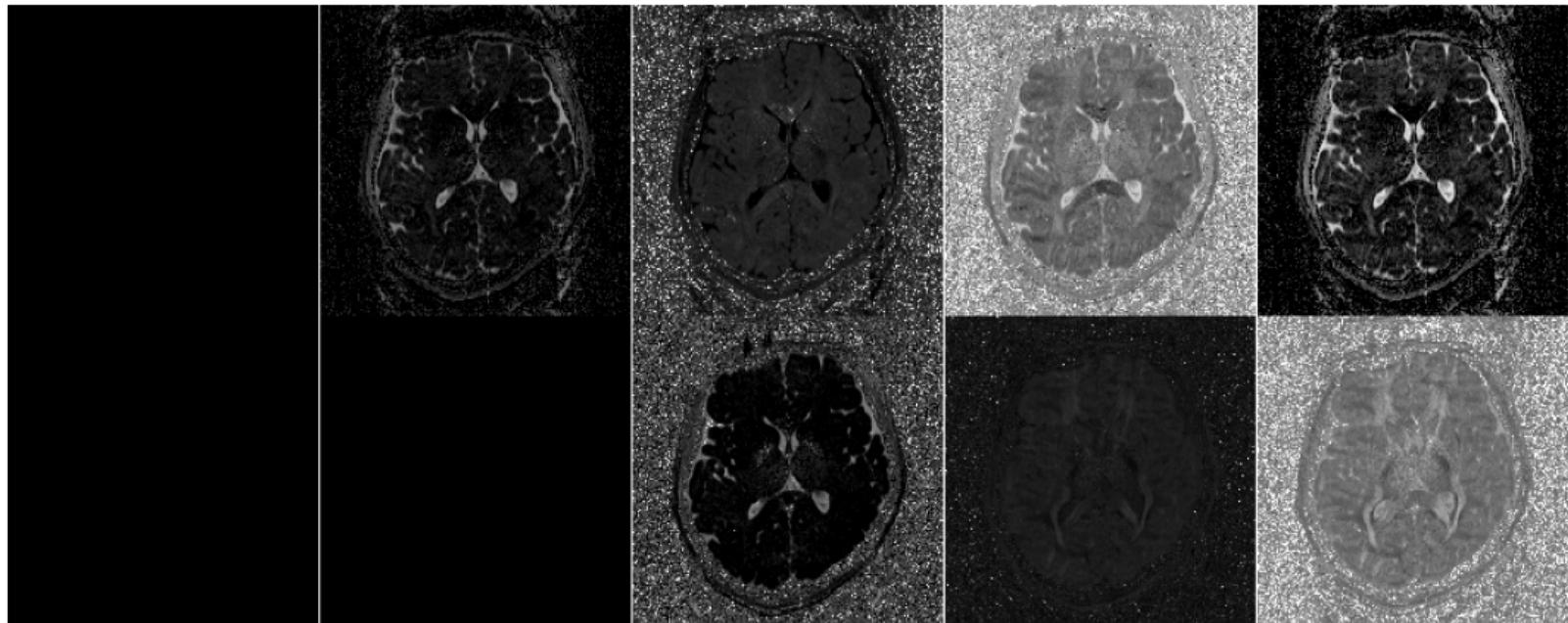
Comparison with the Linear Subspace Method



In vivo DWI Denoising



In vivo DWI Denoising: Learned Latent Signal



Conclusion

Conclusion

- ▶ Multi-shot EPI:
 - ★ enables high spatial resolution DWI
- ▶ we open source our DWI reconstruction techniques →
- ▶ Autoencoder:
 - ★ enables DWI denoising
 - ★ learns a nonlinear subspace

