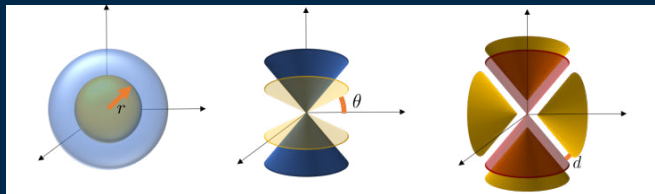
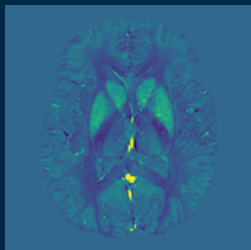


322 One-Dimensional k-Space Metrics on Cone Surfaces for Quantitative Susceptibility Mapping

Maximilian N. Diefenbach¹, Christof Böhm¹, Jakob Meineke², Chunlei Liu³, Dimitrios C. Karampinos¹



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³Department of Electrical Engineering and Computer Sciences, & Helen Wills Neuroscience Institute, University of California, Berkeley, Berkeley, CA, United States



ISMIRM 27TH ANNUAL MEETING & EXHIBITION

Palais des congrès de Montréal 🍁 Montréal, QC, Canada 🍁 11–16 May 2019

Declaration of Financial Interests or Relationships

Speaker Name: Maximilian N. Diefenbach

I have the following financial interest or relationship to disclose with regard to the subject matter of this presentation:

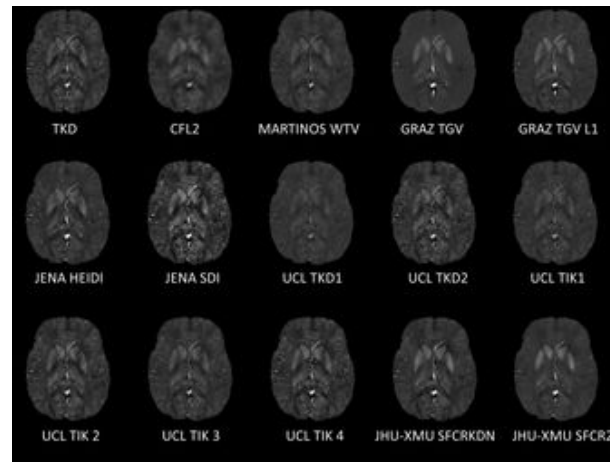
Company Name: Philips Healthcare

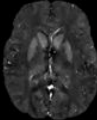
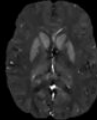
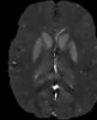
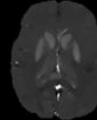
Type of Relationship: Grant Support

Motivation I

How to compare QSM
reconstruction results?

How to choose
hyper parameters?



				
$\alpha_0 =$	0.0005	0.001	0.002	0.004
<i>RMSE</i> =	125.8	97.9	81.2	77.5
<i>HFEN</i> =	121.8	94.9	78.2	74.2
<i>SSIM</i> =	91.3	93.5	94.2	92.9
<i>ROI Error</i> =	0.022	0.020	0.019	0.021

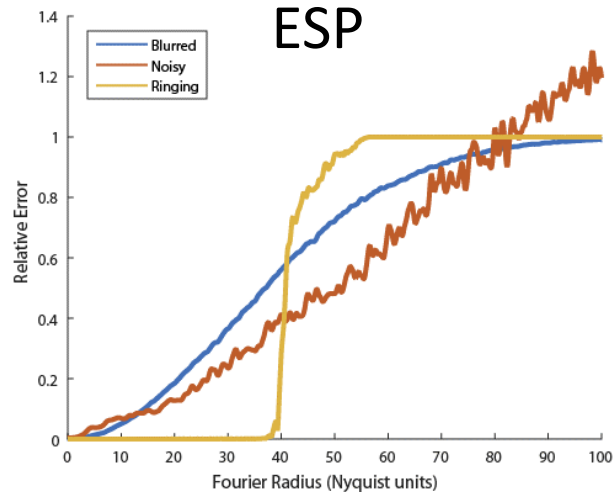
Quantitative Susceptibility Mapping:
Report from the 2016 reconstruction challenge,
MRM, 79(3), 1661–1673 (2017).

Motivation II

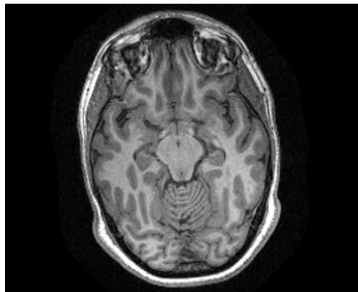
ISMRM 2018 Paris, #249

Kim, T. H., & Haldar, J. P.

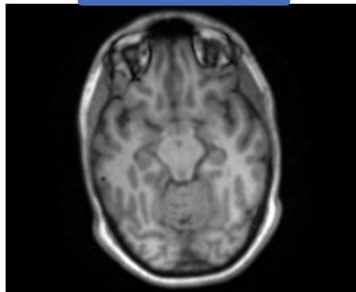
*"Assessing MR image reconstruction quality using the Fourier radial **error spectrum plot [(ESP)]"***



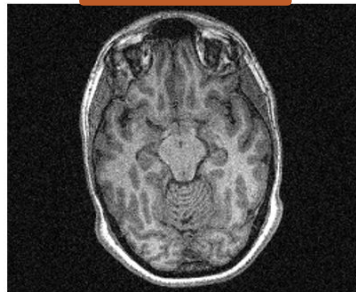
Gold Standard



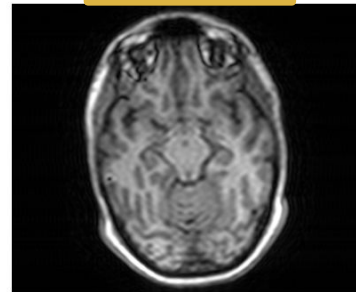
Blurred



Noisy



Ringing

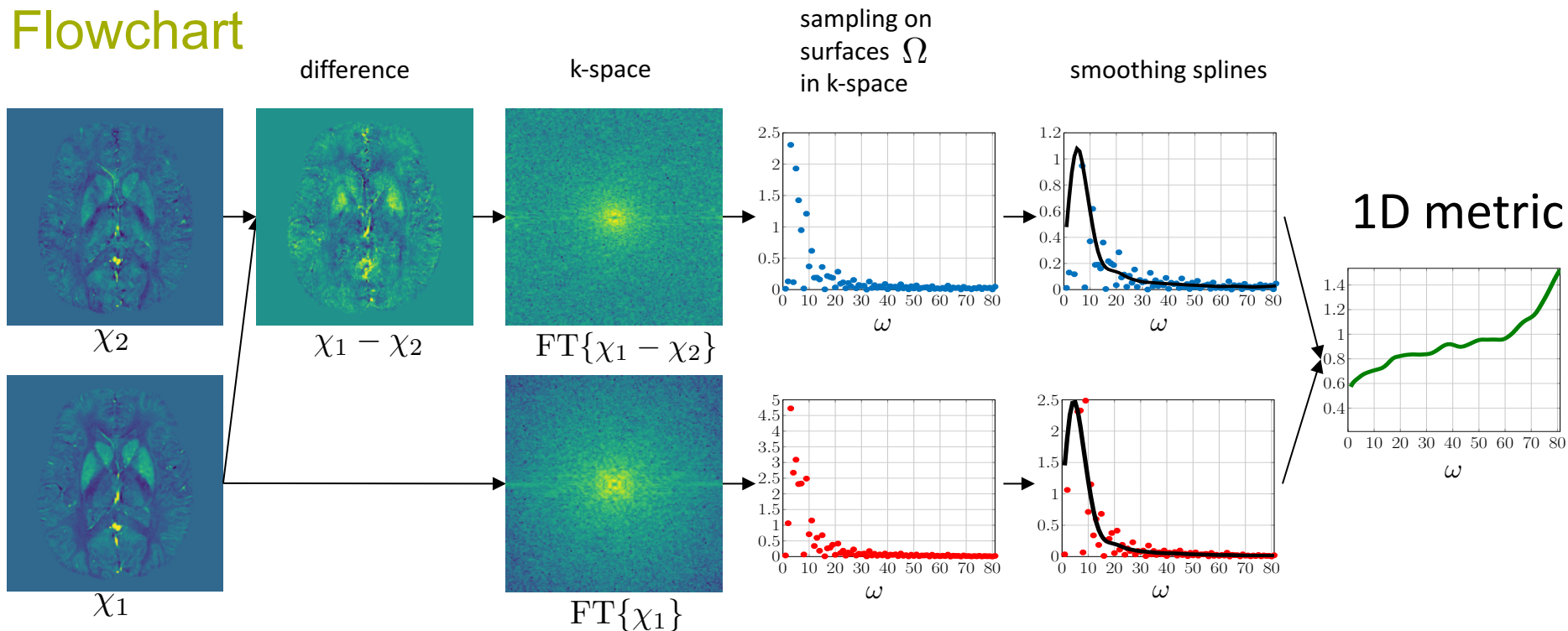


Purpose

to ...

1. Apply the **Error Spectrum Plot** concept in **QSM**
2. Explore **different parametrizations** suited to the intrinsic k-space geometries in QSM

Flowchart



$$E_{\Omega}(\omega(\vec{k}), \chi_1, \chi_2) = \sqrt{\frac{S(\|\text{FT}\{\chi_1(\omega) - \chi_2(\omega)\}\|_2)}{S(\|\text{FT}\{\chi_1(\omega)\}\|_2)}}$$

Parametrizations

$$E_{\Omega}(\omega(\vec{k}), \chi_1, \chi_2) = \sqrt{\frac{S(\|\text{FT}\{\chi_1(\omega) - \chi_2(\omega)\}\|_2)}{S(\|\text{FT}\{\chi_1(\omega)\}\|_2)}}, \quad \omega \in \Omega$$

$\vec{B}_0 \uparrow$

Recon challenge data

- tissue field after background field removal: **phs_tissue**
- reference: **chi_33**

Baseline reconstruction: $b = d * \chi \xrightarrow{\text{FFT}} B = DX$

- tkd** “Truncated k-space division” Shmueli et al., MRM, 62(6), 1510–1522 (2009). doi: 10.1002/mrm.22135

$$X = B/D' \quad D' = \text{sgn}(D(k)) \max(|D(k)|, \tau)$$

- L2** “close form L2 regularized solution” Bilgic et al., JMRI 40(1), 181–191 (2013). doi: 10.1002/jmri.24365

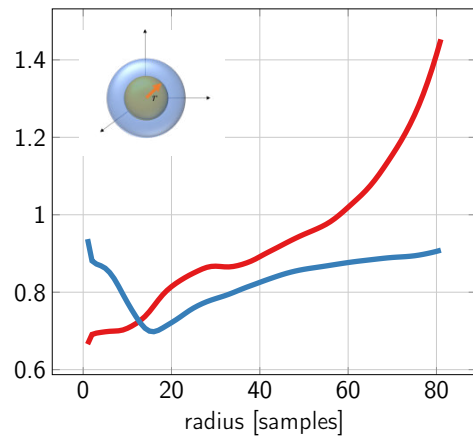
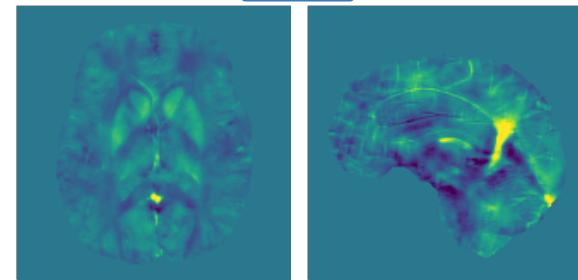
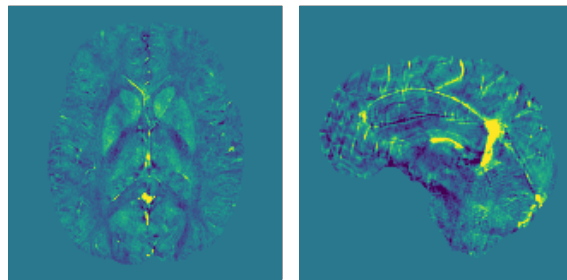
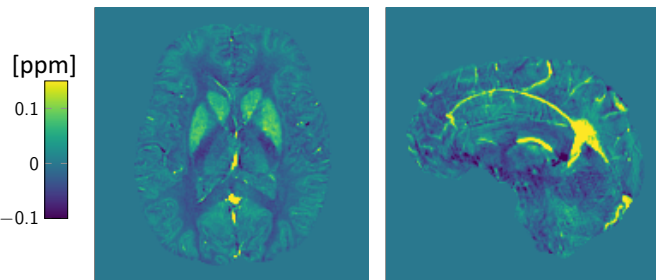
$$X = B/D'' \quad D'' = (D^2(k) + \lambda k^2)/D(k)$$

Comparison of QSM reconstruction results

chi 33

tkd

L2

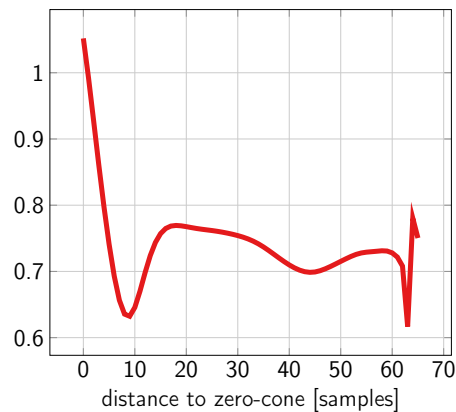
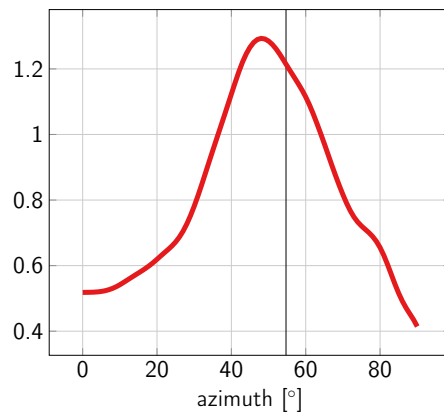
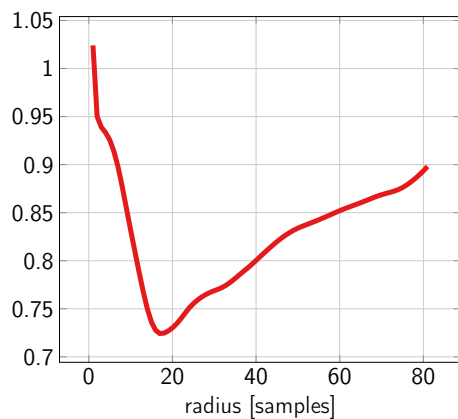
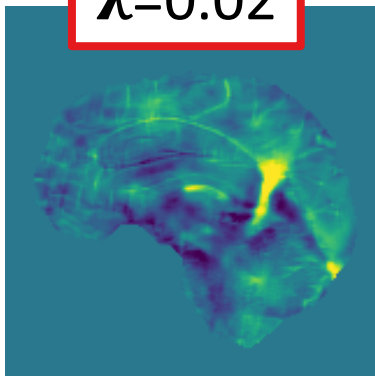
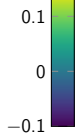


	rmse	hfen	ssim
tkd	86.50	81.97	0.77
L2	81.23	75.49	0.81

Selection of hyper parameters: L2 regularization parameter λ

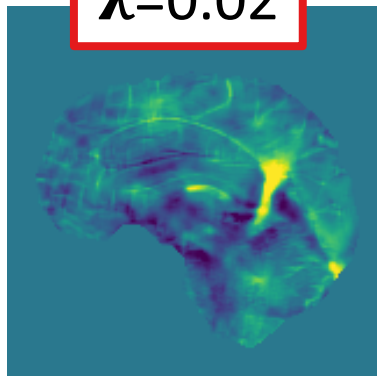
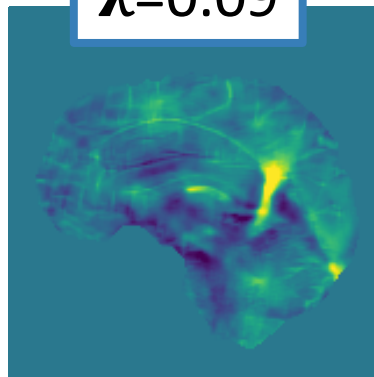
$\lambda=0.02$

[ppm]



λ	rmse	hfen	ssim
— 0.02	83.54	76.26	0.81

Selection of hyper parameters: L2 regularization parameter λ

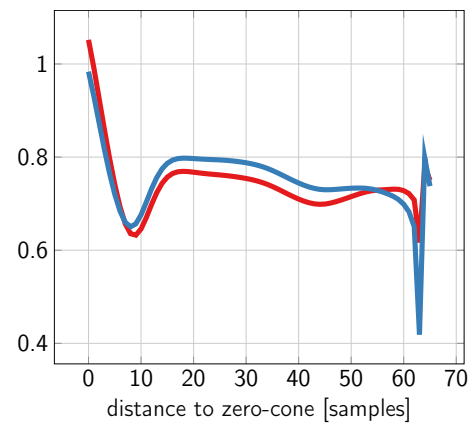
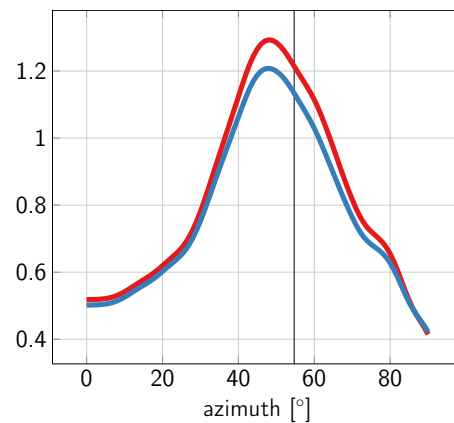
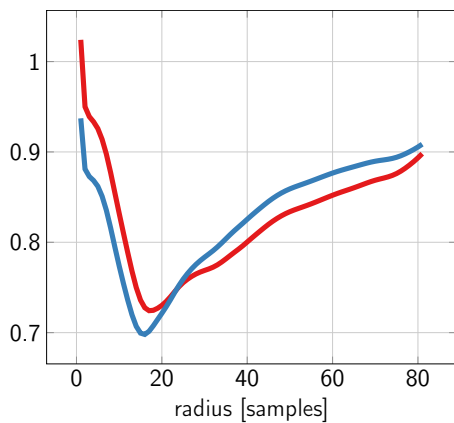
 $\lambda=0.02$  $\lambda=0.09$ 

[ppm]

0.1

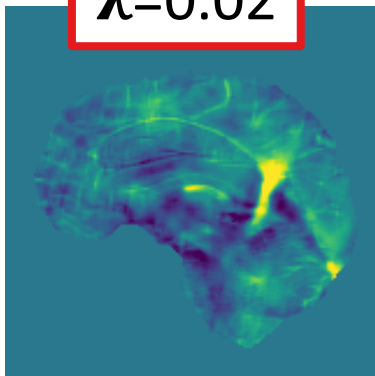
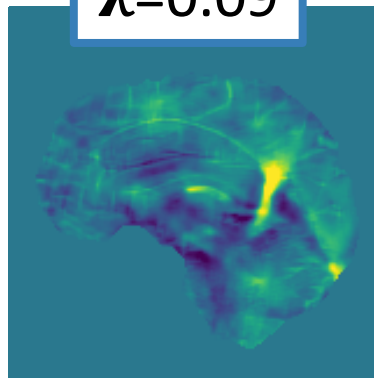
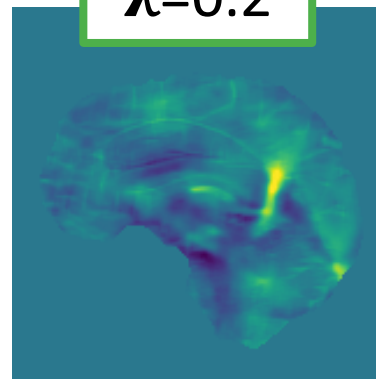
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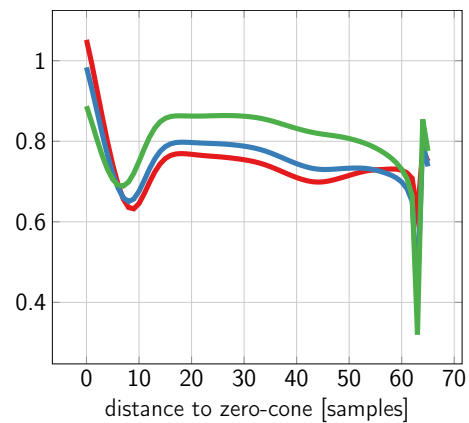
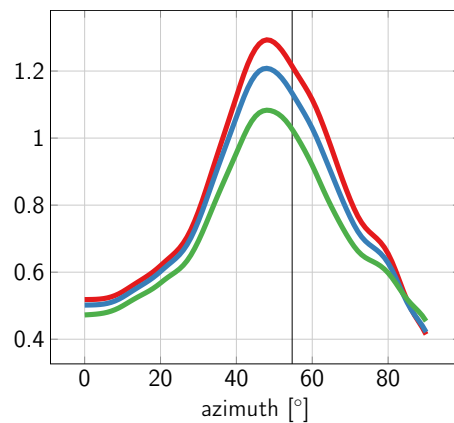
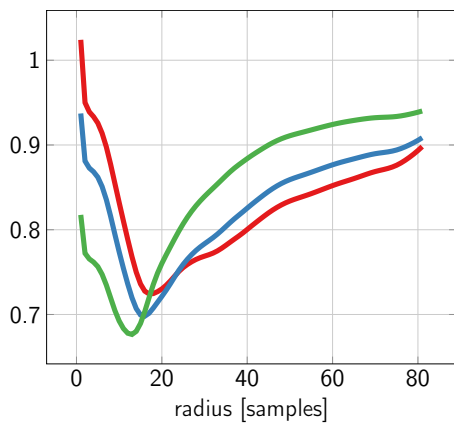


λ	rmse	hfen	ssim
0.02	83.54	76.26	0.81
0.09	81.23	75.49	0.81

Selection of hyper parameters: L2 regularization parameter λ

 $\lambda=0.02$  $\lambda=0.09$  $\lambda=0.2$ 

[ppm]
0.1
0
-0.1



λ	rmse	hfen	ssim
0.02	83.54	76.26	0.81
0.09	81.23	75.49	0.81
0.2	79.66	78.20	0.79

Fusion of QSM maps: $L2(\lambda_1) \Leftrightarrow L2(\lambda_2)$

 $\lambda_1 = 0.02$ $\lambda_2 = 0.2$

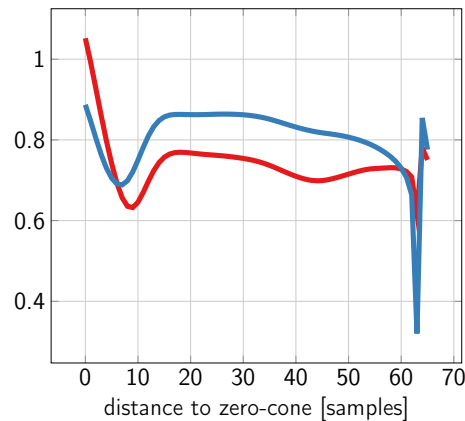
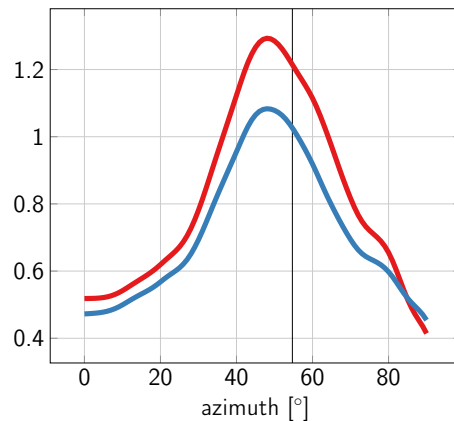
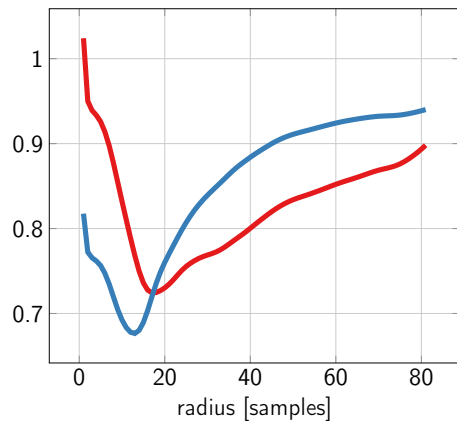
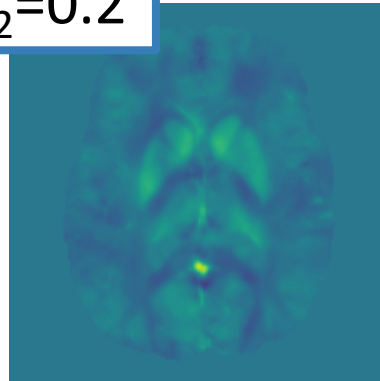
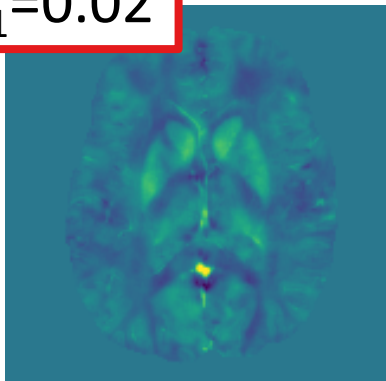
[ppm]



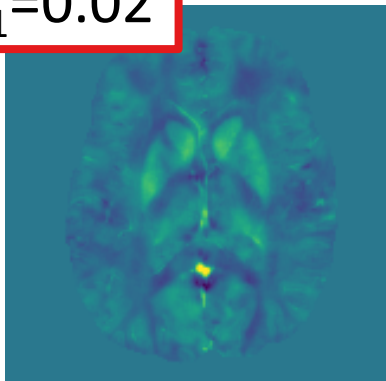
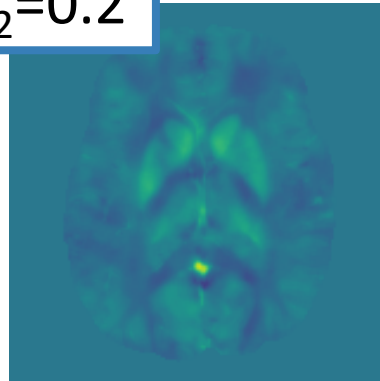
0.1

0

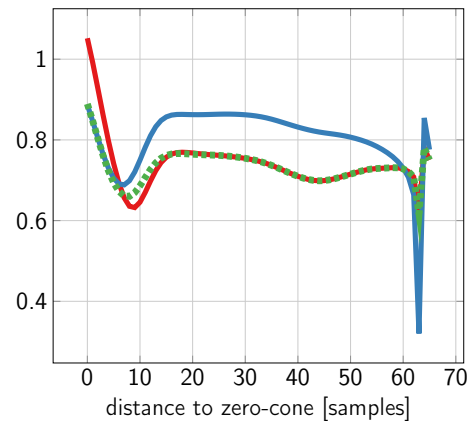
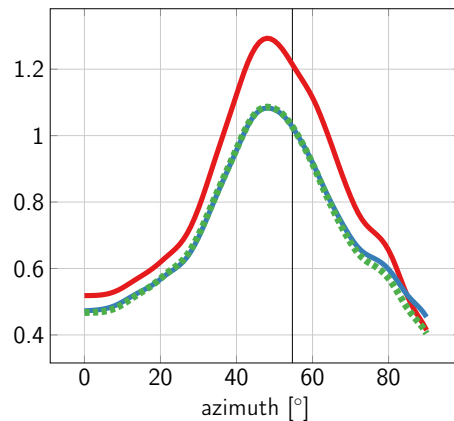
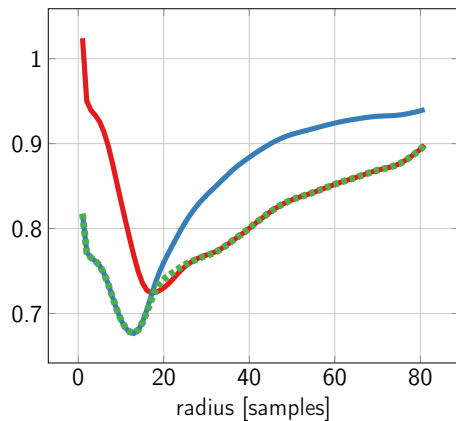
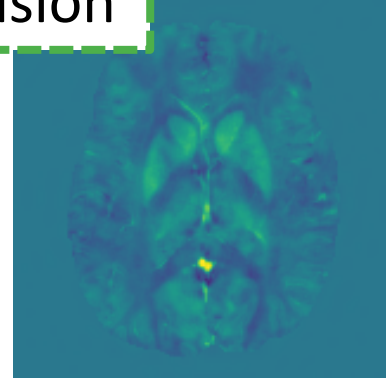
-0.1



Fusion of QSM maps: $L2(\lambda_1) \Leftrightarrow L2(\lambda_2)$

 $\lambda_1=0.02$  $\lambda_2=0.2$ 

fusion



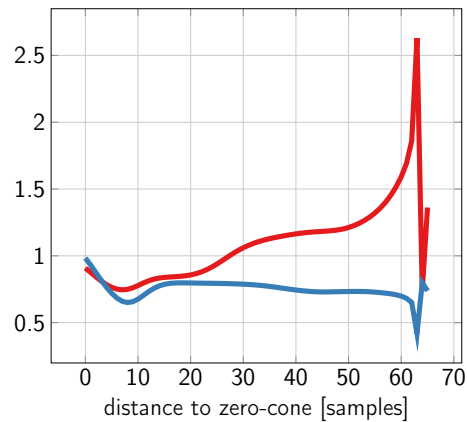
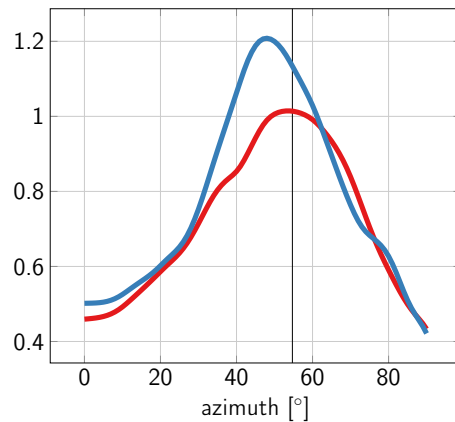
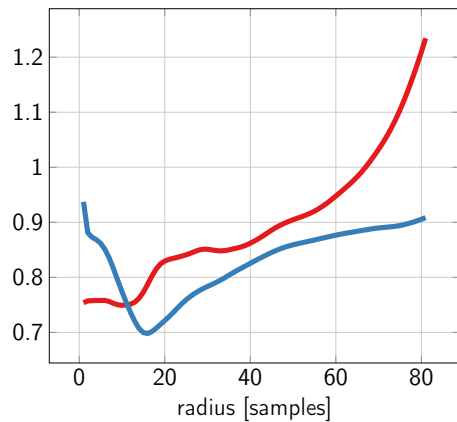
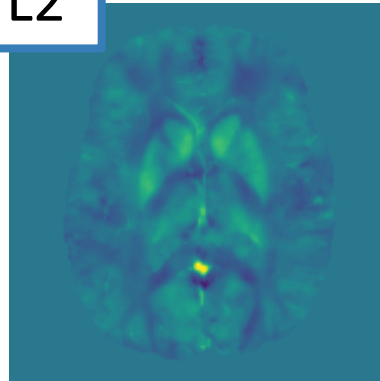
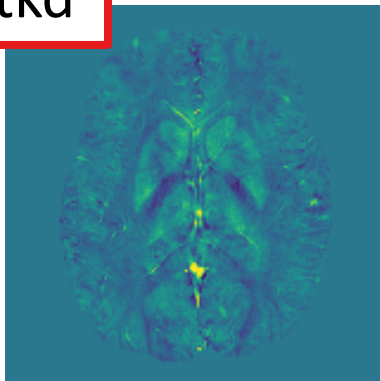
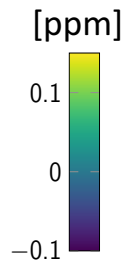
ssim
drops (?)



Fusion of QSM maps: tkd $\Leftrightarrow$ L2

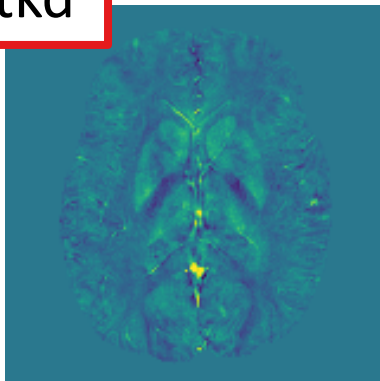
tkd

L2

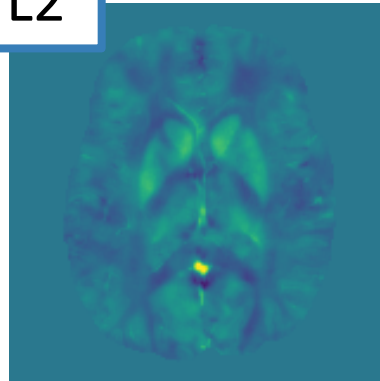


Fusion of QSM maps: tkd $\Leftrightarrow$ L2

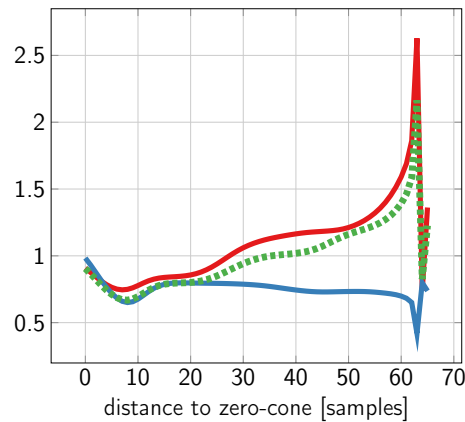
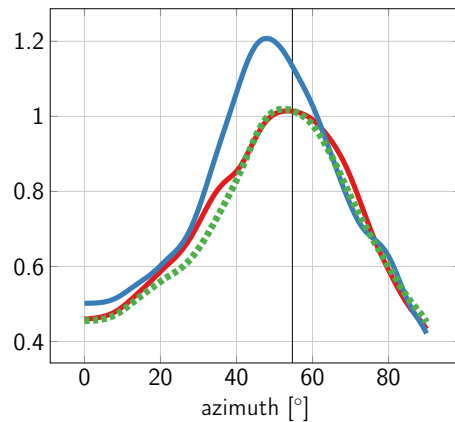
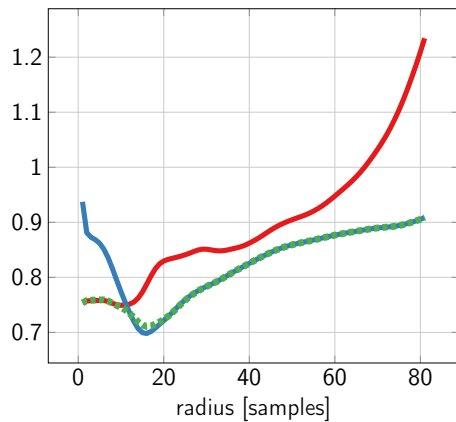
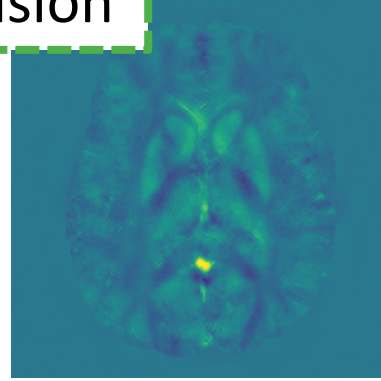
tkd



L2



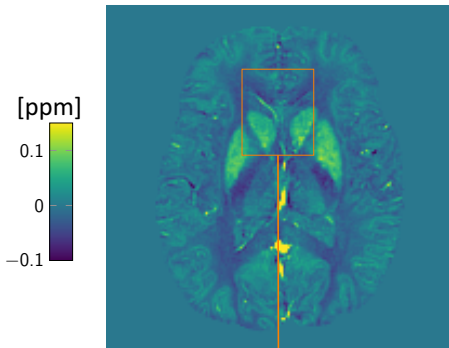
fusion



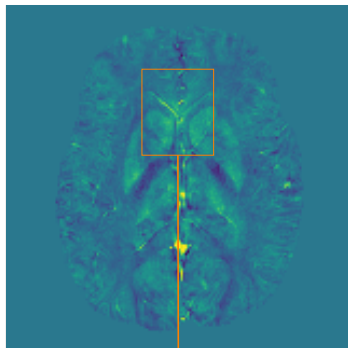
	rmse	hfen	ssim
tkd	84.52	82.28	0.79
L2	81.23	75.49	0.81
fusion	79.97	75.42	0.15

Fusion of QSM maps: tkd $\Leftrightarrow$ L2

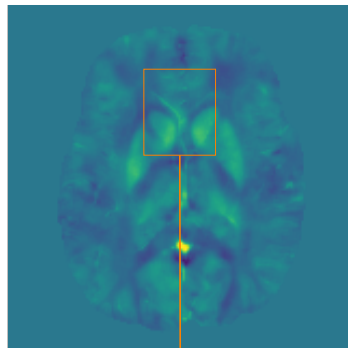
chi 33



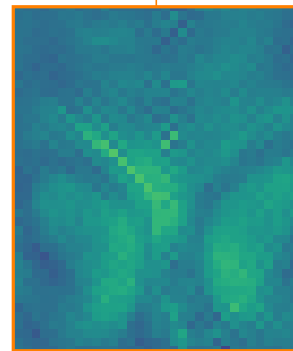
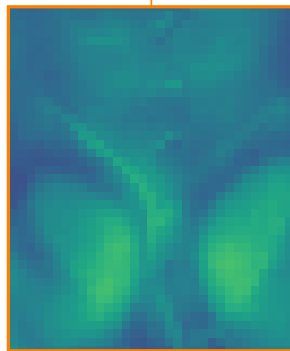
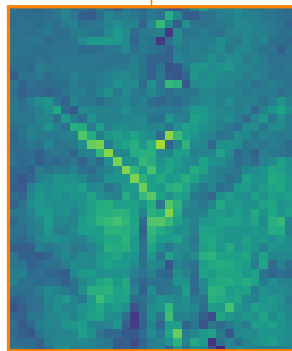
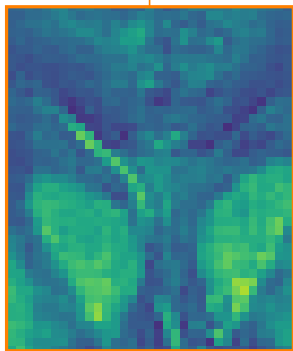
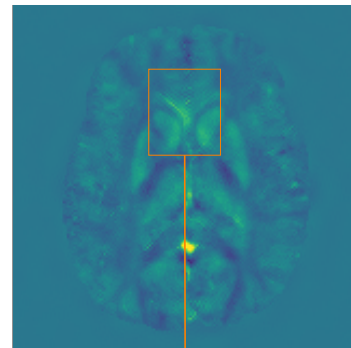
tkd



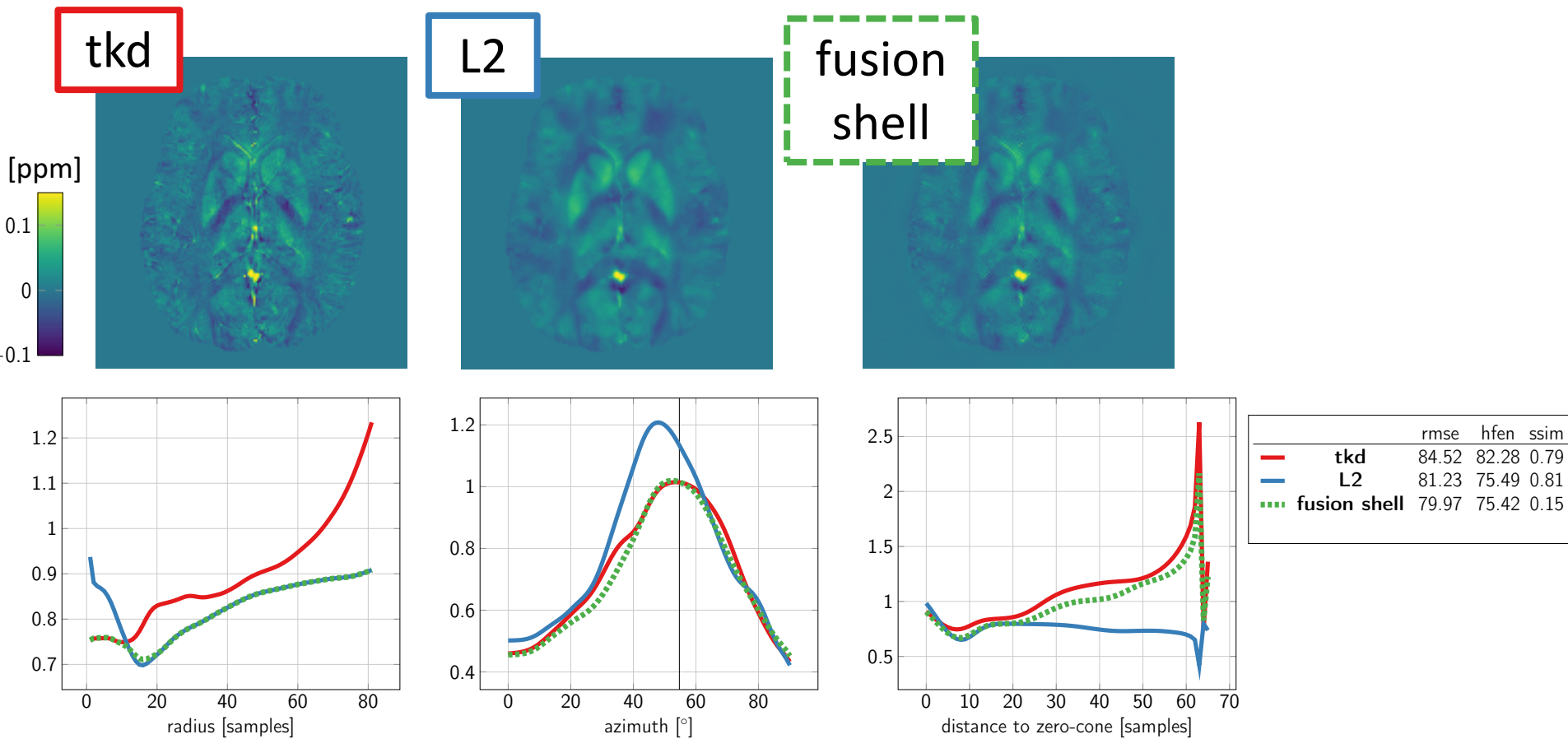
L2



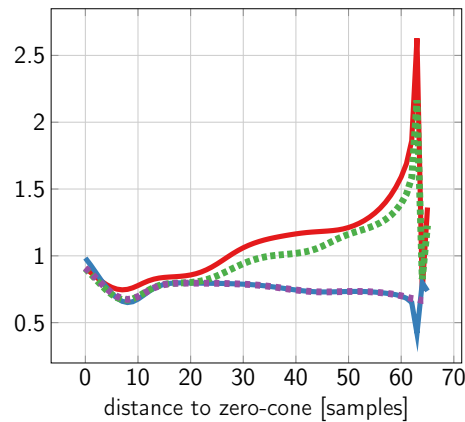
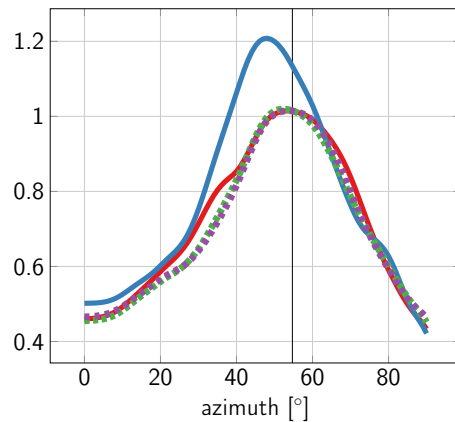
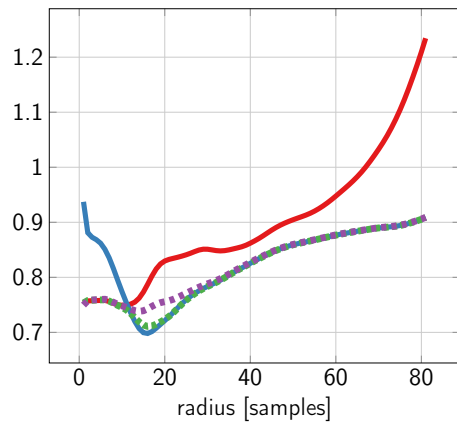
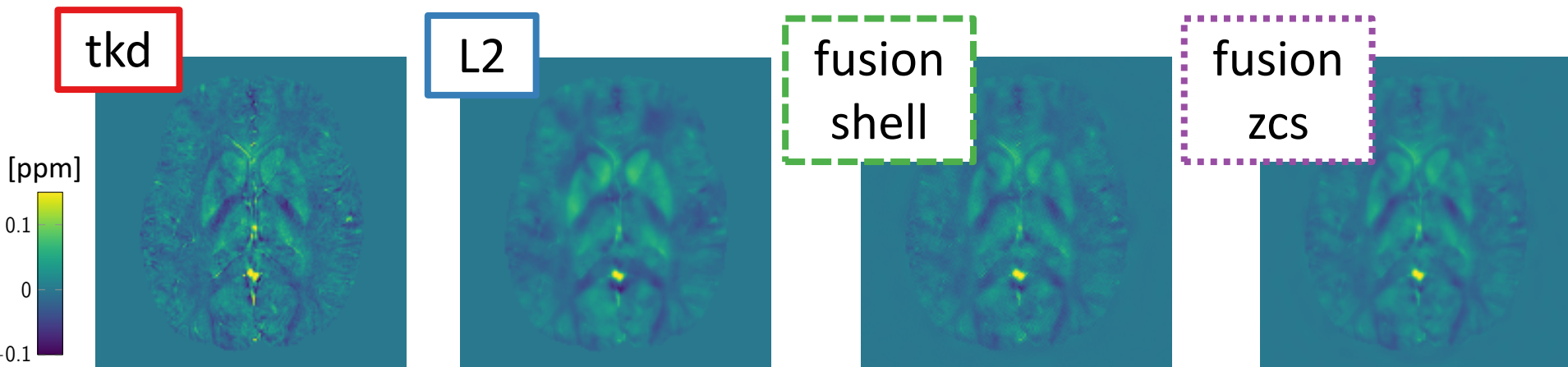
fusion



Fusion of QSM maps: $\text{tkd} \Leftrightarrow \text{L2}$, cone surfaces!

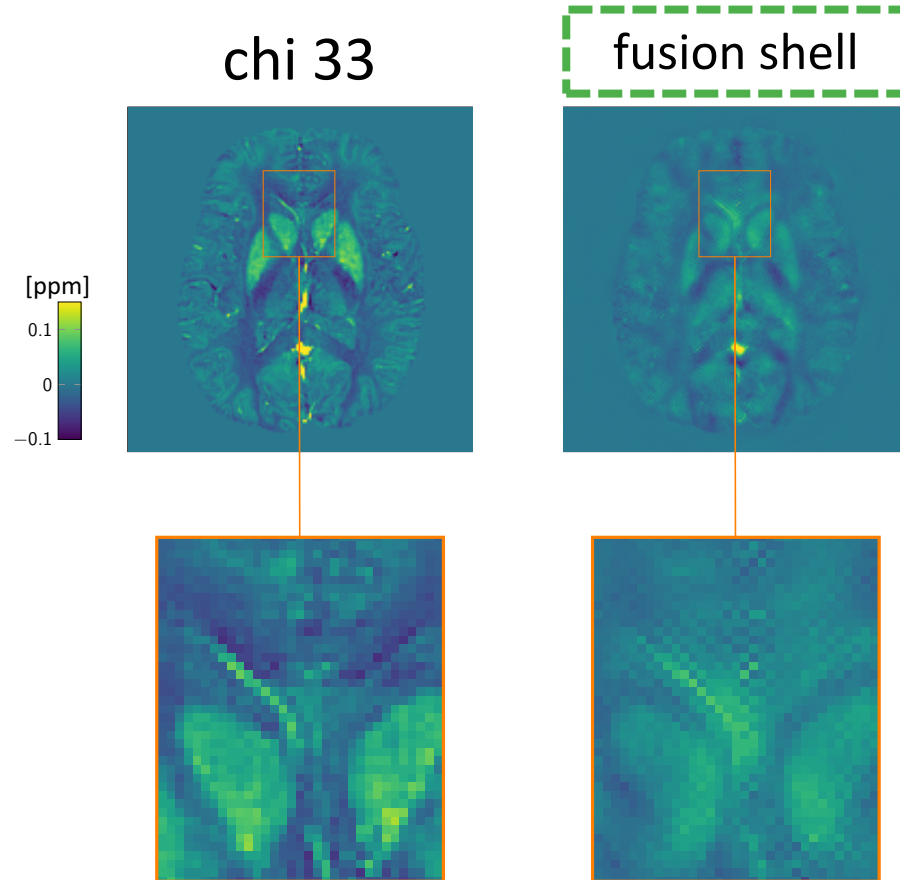


Fusion of QSM maps: tkd $\Leftrightarrow$ L2, cone surfaces!

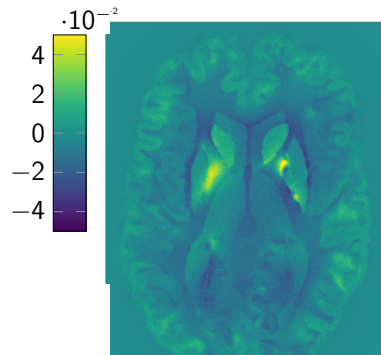


	rmse	hfen	ssim
tkd	84.52	82.28	0.79
L2	81.23	75.49	0.81
fusion shell	79.97	75.42	0.15
fusion zcs	79.09	76.97	0.15

Fusion of QSM maps: tkd $\Leftrightarrow$ L2, cone surfaces!



Validation numerical brain phantom



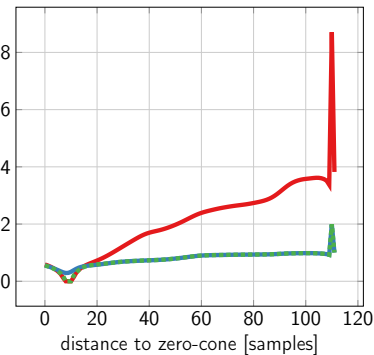
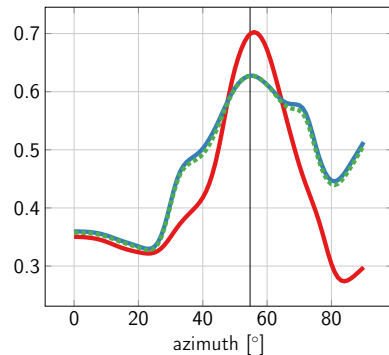
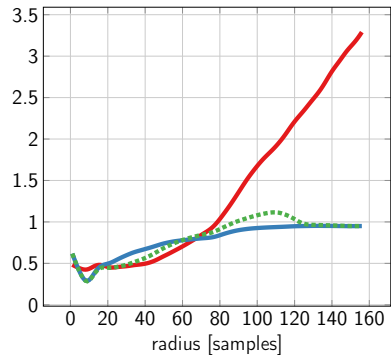
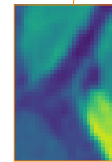
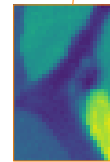
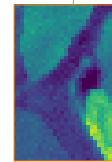
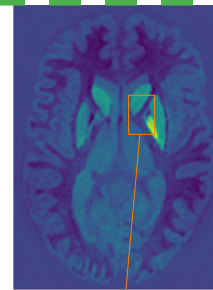
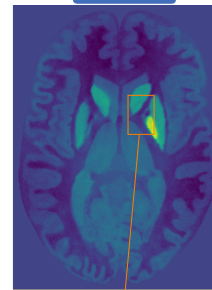
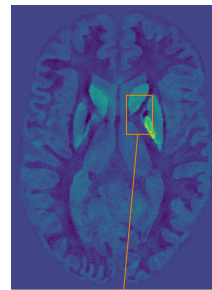
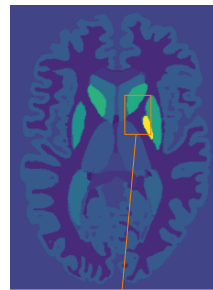
tissue field (SNR=100)

ref

tkd

L2

fusion zcs



	rmse	hfen	ssim
tkd	54.32	48.35	0.66
L2	49.10	56.64	0.74
fusion zcs	47.41	49.18	0.30

Limitations

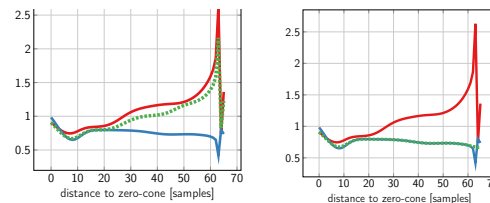
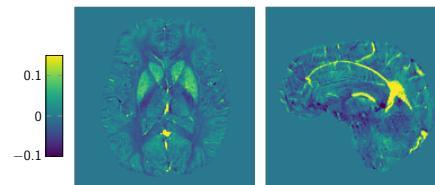
- Relation 1D to scalar metrics not clear (ssim)
- Focus on baseline methods (tkd, L2) not state-of-the-art
- 1D metrics still reference-based
- No clear prospective fusion strategy

	rmse	hfen	ssim
— tkd	84.52	82.28	0.79
— L2	81.23	75.49	0.81
.... fusion	79.97	75.42	0.15

$$D' = \text{sgn}(D(k)) \max(|D(k)|, \tau)$$

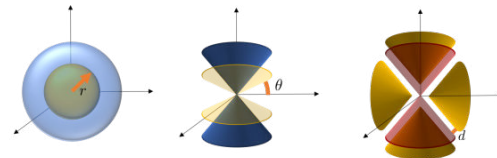
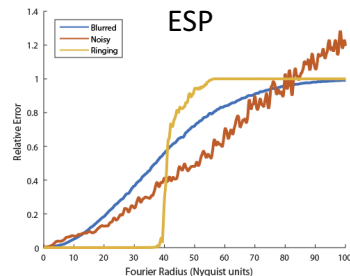
$$D'' = (D^2(k) + \lambda k^2) / D(k)$$

chi 33

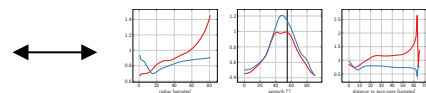


Conclusion

- **ESP** applicable in QSM
- **Conic parametrizations** allow visualization of differences between QSM methods
- **Complementary** to standard error metrics
- **Fusion** of different QSM methods can yield can improve results



	rmse	hfen	ssim
tkd	86.50	81.97	0.77
L2	81.23	75.49	0.81

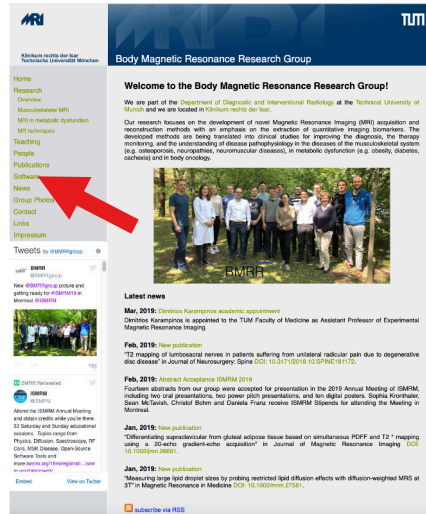


Software available

„qsm_esp“



<http://bmrr.de/software>



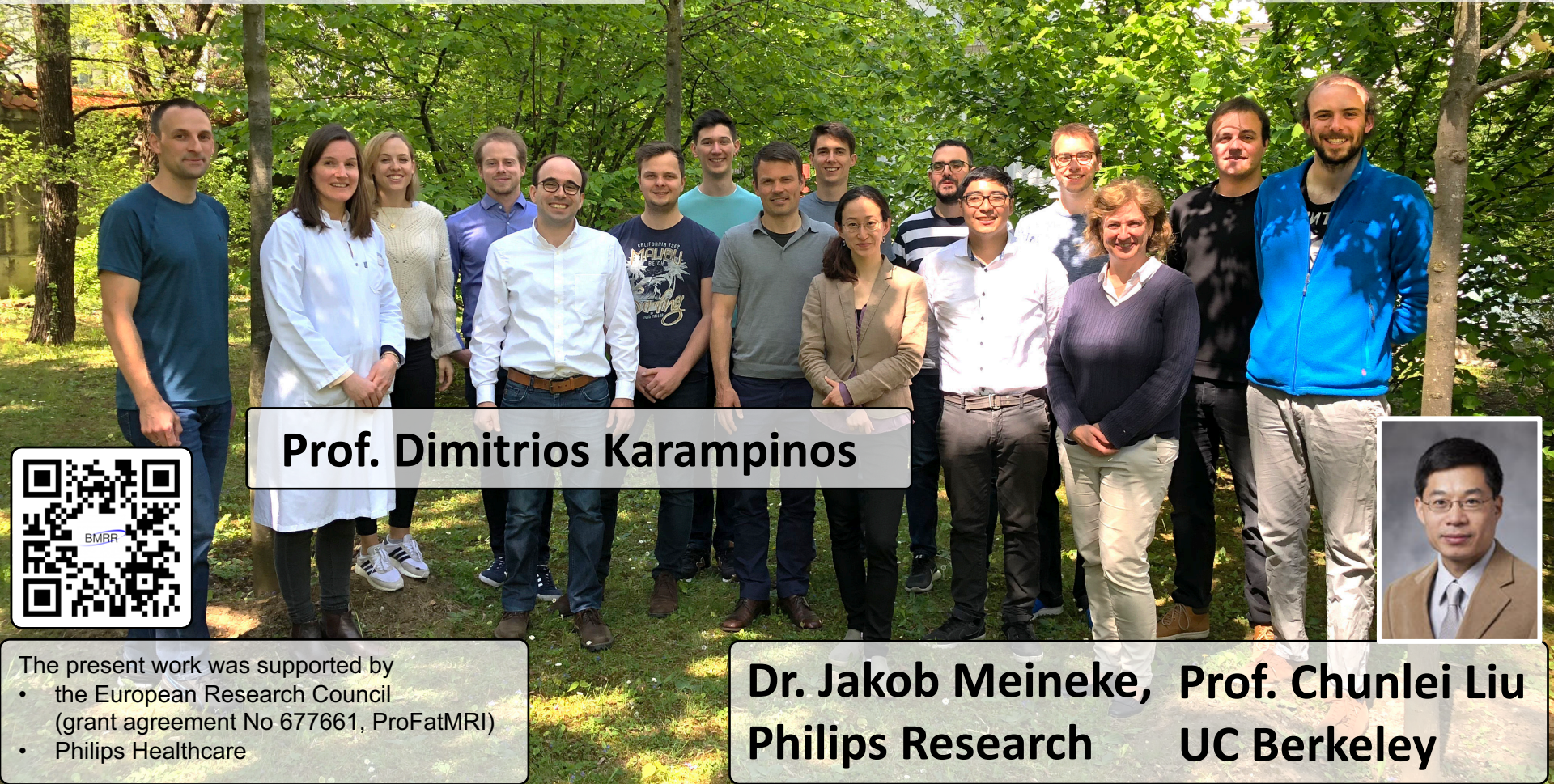
MATLAB (& python)

```
% addpath(genpath('qsm_esp/'))
```

```
metrics = compute_metrics(chi_ref, chi);  
plot_metrics(metrics);|
```

+ functions for fusion

Acknowledgements



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