

# Quantitative Susceptibility Mapping with Magnitude Prior

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- Quantitative Susceptibility Mapping (QSM) aims to quantify tissue magnetic susceptibility with applications such as,
  - Tissue contrast enhancement<sup>1</sup>
  - Estimation of venous blood oxygenation<sup>2</sup>
  - Quantification of tissue iron concentration<sup>3</sup>

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- Estimation of the susceptibility map  $\chi$  from the unwrapped phase  $\varphi$  involves solving an inverse problem,

$$\delta = \mathbf{F}^{-1}\mathbf{D}\mathbf{F}\chi$$

F: Discrete Fourier Transform matrix

**D**: susceptibility kernel in *k*-space

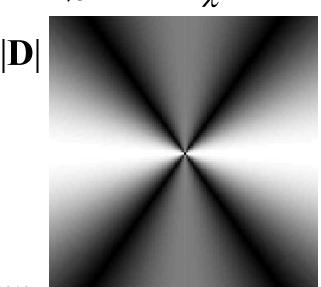
$$\delta = \frac{\varphi}{\gamma \cdot TE \cdot B_0}$$
: normalized field map

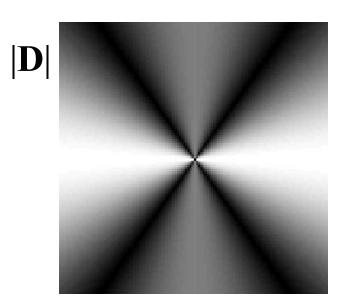
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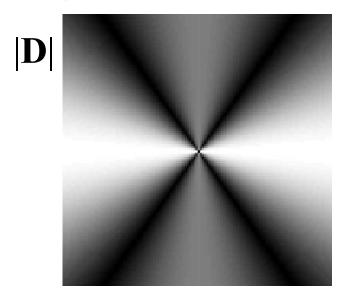
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- Estimation of the susceptibility map  $\chi$  from the unwrapped phase  $\varphi$  involves solving an inverse problem,  $\delta = \mathbf{F}^{-1}\mathbf{D}\mathbf{F}\chi$
- The inversion is made difficult by zeros on a conical surface in susceptibility kernel D

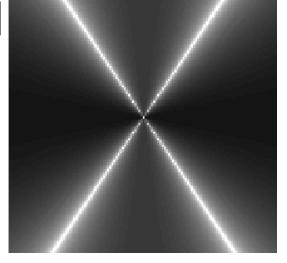
$$\mathbf{D} = \left(\frac{1}{3} - \frac{\mathbf{k}_z^2}{\mathbf{k}^2}\right)$$







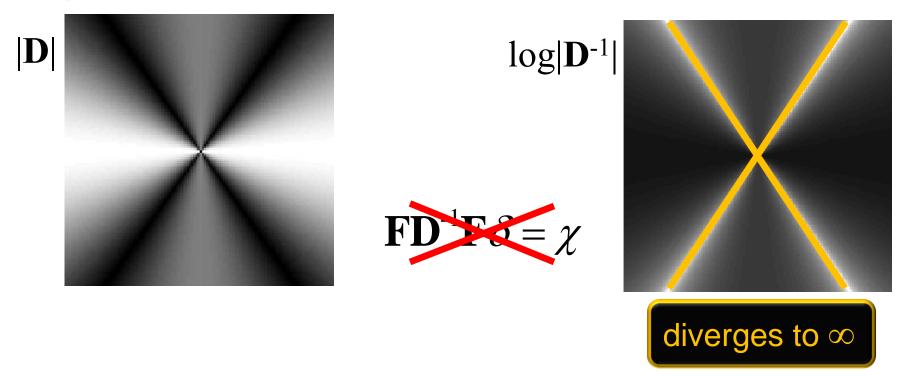




Solving for  $\chi$  by convolving with the inverse of **D** is not possible, as it diverges along the magic angle

 $\mathbf{F}\mathbf{D}^{-1}\mathbf{F}\delta = \chi$ 

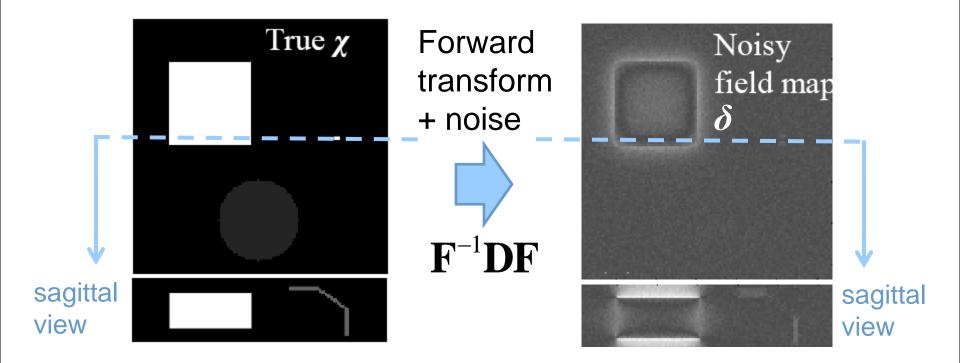




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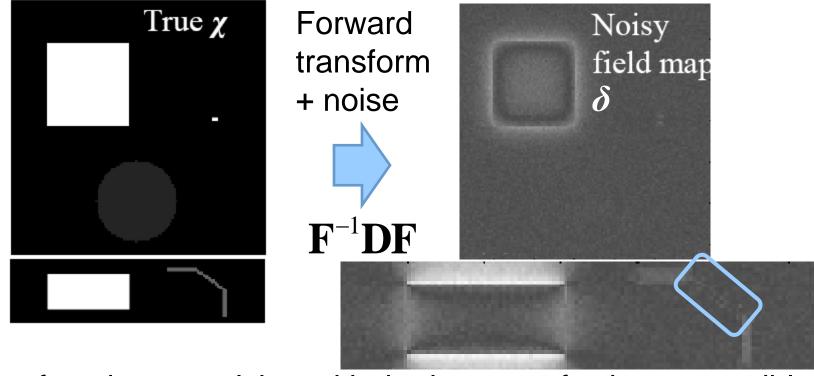




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- Spatial details that have frequency components at the magic angle lose conspicuity in the field map  $\delta$





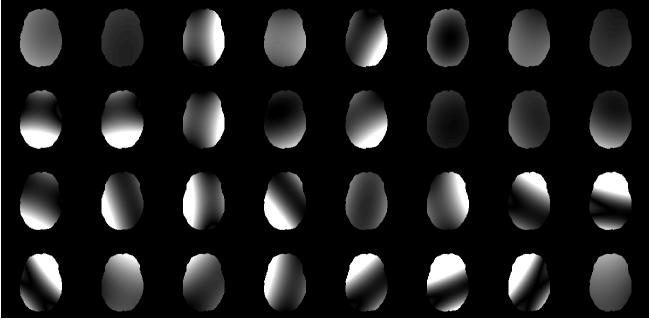


- Solving for  $\chi$  by convolving with the inverse of **D** is not possible, as it diverges along the magic angle
- Spatial details that have frequency components at the magic angle lose conspicuity in the field map  $\delta$
- We propose to use regularization to facilitate the inversion



- 3D GRE acquisition with phased array coils and body coil
- Normalize each channel image with the body coil

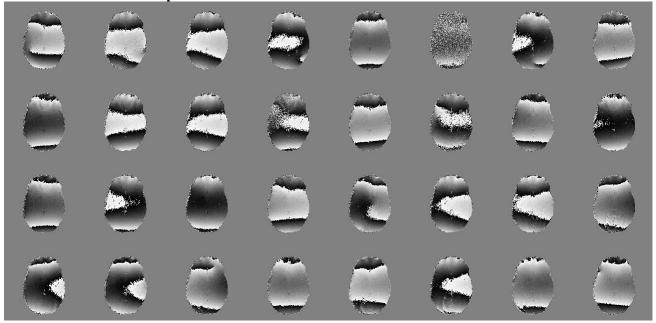
magnitudes of the coil sensitivities



- 3D GRE acquisition with phased array coils and body coil
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- Fit 2<sup>nd</sup> order polynomials to the magnitude of the normalized images → magnitude of the coil sensitivities



phase of the coil sensitivities

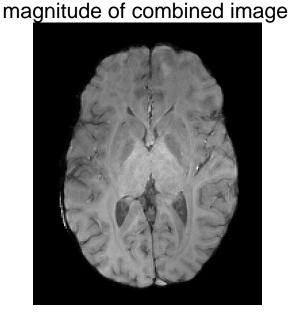


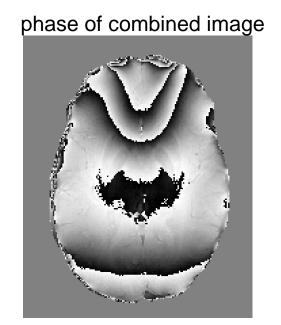
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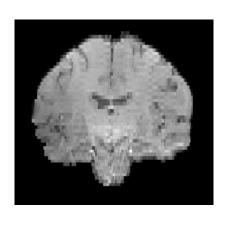


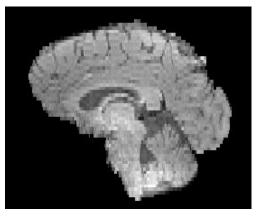


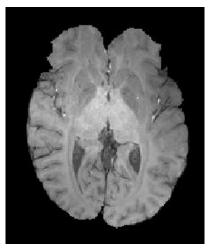
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- Phase of the normalized images → phase of the coil sensitivities
- Final image is obtained by least-squares coil combination

#### **Brain Mask Extraction & Phase Unwrapping**

Brain mask was generated with the FSL Brain Extraction Tool<sup>1</sup>

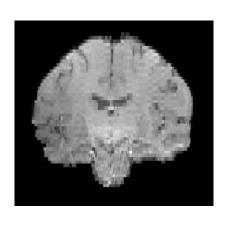


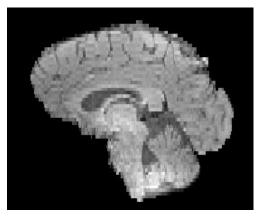




## **Brain Mask Extraction & Phase Unwrapping**

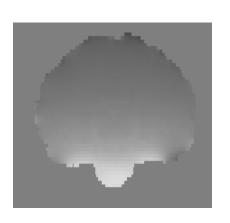
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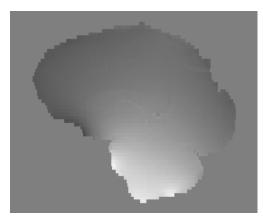






Phase unwrapping was done with the FSL PRELUDE<sup>2</sup>



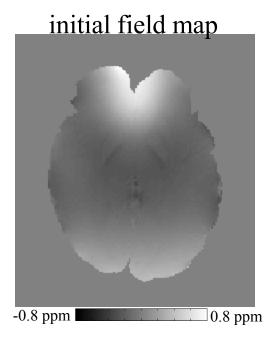


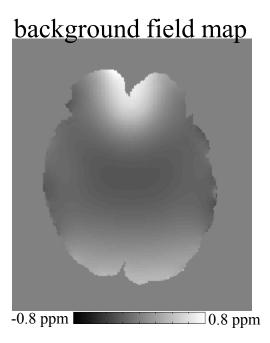


-30 rad

#### **Background Phase Removal**

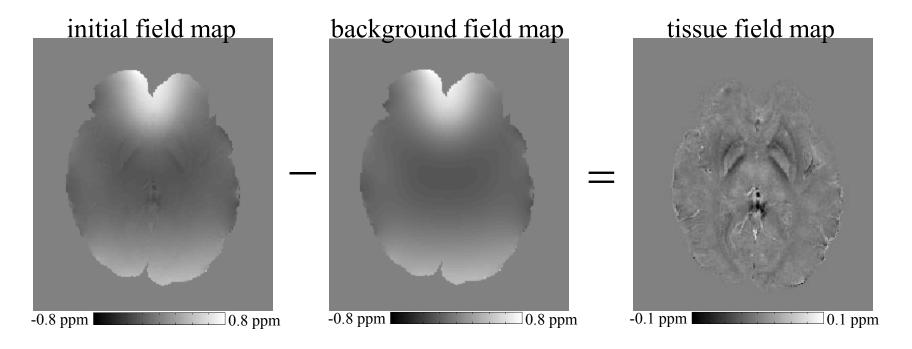
 The background phase was estimated with the Effective Dipole Fitting method¹





## **Background Phase Removal**

- The background phase was estimated with the Effective Dipole Fitting method¹
- Subtracting the estimated background from the initial field map gives the tissue field map



• The tissue field map  $\delta$  is related to the susceptibility distribution  $\chi$  via

$$\delta = \mathbf{F}^{-1}\mathbf{D}\mathbf{F}\chi$$





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Multiplying both sides with V<sub>x</sub>F

$$\mathbf{V}_{r}\mathbf{F}\delta = \mathbf{V}_{r}\mathbf{D}\mathbf{F}\chi$$

where  $\mathbf{V}_x$  is a diagonal matrix with  $\mathbf{V}_x(\omega,\omega) = \left(1 - e^{-2\pi j\omega/n}\right)$ 





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where  $V_x$  is a diagonal matrix with  $V_x(\omega,\omega) = (1 - e^{-2\pi j\omega/n})$ 

This corresponds to taking the spatial gradient along the x axis

$$\mathbf{F}(\partial_x \delta) = \mathbf{DF}(\partial_x \chi)$$

• The gradient of the tissue field map  $\delta$  is related to the gradient of the susceptibility distribution  $\chi$  via

$$\mathbf{F}(\partial_x \delta) = \mathbf{DF}(\partial_x \chi)$$

We solve for ∂<sub>x</sub> χ with the FOCUSS algorithm¹

at  $k^{\text{th}}$  iteration,

$$\mathbf{W}_{k} = \operatorname{diag}\left(\left|\partial_{x} \chi_{k-1}\right|^{1/2}\right)$$

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$$\mathbf{q}_{k} = \operatorname{argmin} \left\|\mathbf{F}\left(\partial_{x} \delta\right) - \mathbf{DFW}_{k} \mathbf{q}\right\|_{2}^{2} + \lambda \left\|\mathbf{q}\right\|_{2}^{2}$$

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$$\partial_{x}\chi_{k} = \mathbf{W}_{k}\mathbf{q}_{k}$$

 We expect the susceptibility distribution to share similar spatial gradients as the magnitude image.





- We expect the susceptibility distribution to share similar spatial gradients as the magnitude image.
- To impose this prior, we modify the update equations as,

$$\mathbf{W}_{prior} = \operatorname{diag}(\left|\partial_{x} \boldsymbol{m}\right|^{1/2}), \quad \boldsymbol{m}: \text{ magnitude image}$$
at  $k^{\text{th}}$  iteration,
$$\mathbf{W}_{k} = \operatorname{diag}(\left|\partial_{x} \chi_{k-1}\right|^{1/2})$$

$$\boldsymbol{q}_{k} = \underset{\boldsymbol{q}}{\operatorname{argmin}} \|\mathbf{F}(\partial_{x} \delta) - \mathbf{DFW}_{prior} \mathbf{W}_{k} \boldsymbol{q}\|_{2}^{2} + \lambda \|\boldsymbol{q}\|_{2}^{2}$$

$$\partial_{x} \chi_{k} = \mathbf{W}_{prior} \mathbf{W}_{k} \boldsymbol{q}_{k}$$



- We expect the susceptibility distribution to share similar spatial gradients as the magnitude image.
- Expressed in terms of  $\partial_{x}\chi$ ,

$$\mathbf{W}_{prior} = \operatorname{diag}(|\partial_x m|^{1/2}), \quad m$$
: magnitude image

$$\partial_{x} \chi_{k} = \underset{\partial}{\operatorname{argmin}} \left\| \mathbf{F} (\partial_{x} \delta) - \mathbf{D} \mathbf{F} (\partial_{x} \chi) \right\|_{2}^{2} + \lambda \left\| \mathbf{W}_{prior}^{-1} \mathbf{W}_{k}^{-1} (\partial_{x} \chi) \right\|_{2}^{2}$$

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if  $\partial_x m_i$  is small,  $\mathbf{W}_{prior}^{-1}(i,i)$  will be large and penalize  $\partial_x \chi_i$  more

 After estimating the spatial gradients along x, y and z axes, the susceptibility distribution that matches these is found by solving a least squares problem,

$$\chi = \underset{\theta}{\operatorname{argmin}} \sum_{r=x,y,z} \| \partial_r \theta - \partial_r \chi \|_2^2 + \beta \cdot \| \delta - \mathbf{F}^{-1} \mathbf{D} \mathbf{F} \theta \|_2^2$$

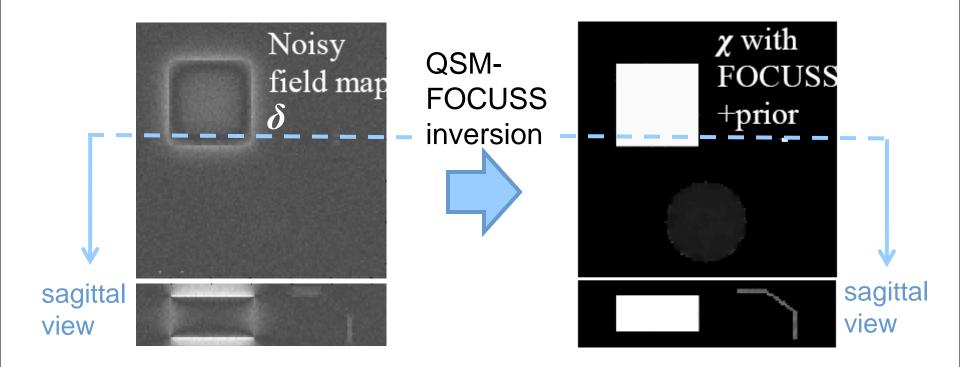




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$$\chi = \underset{\theta}{\operatorname{argmin}} \sum_{r=x,y,z} \left\| \partial_r \theta - \partial_r \chi \right\|_2^2 + \beta \cdot \left\| \delta - \mathbf{F}^{-1} \mathbf{D} \mathbf{F} \theta \right\|_2^2$$
matching gradients data consistency

## **QSM result: FOCUSS-QSM with magnitude prior**

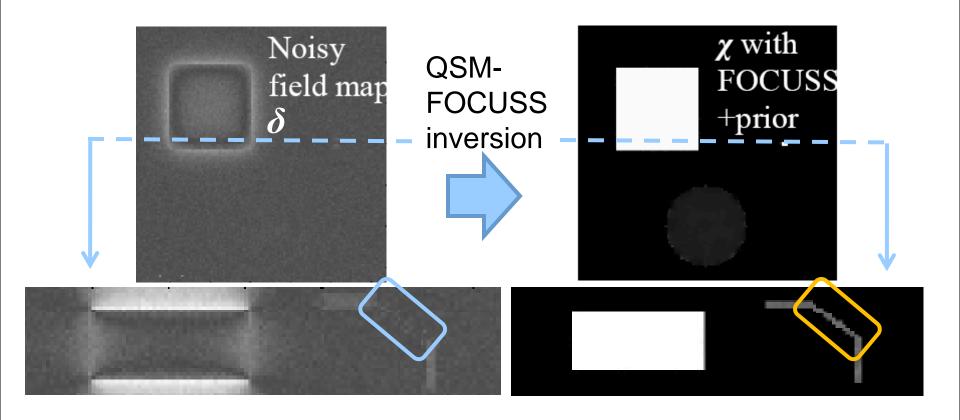


• Starting from the noisy field map  $\delta$ , FOCUSS-QSM with magnitude prior yielded a susceptibility map with 1.3 % RMSE relative to true  $\chi$ .





## **QSM result: FOCUSS-QSM with magnitude prior**

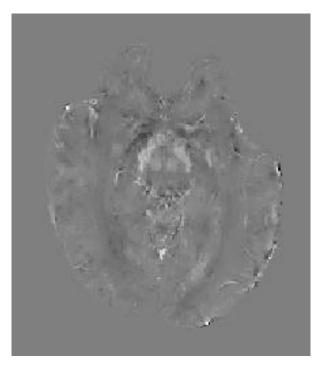


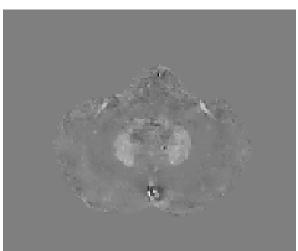
 The reconstructed susceptibility map managed to recover the vessel at the magic angle, which was virtually lost in the field map.



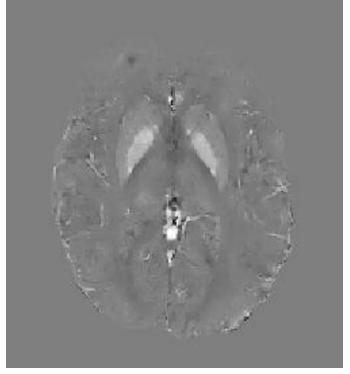


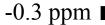
## In vivo QSM result: FOCUSS-QSM with magnitude prior





- 3D GRE acquisition at 3T
- 32 channel receive array
- 0.94x0.94x2.5 mm<sup>3</sup> resolution
- ❖ TE: 20 ms

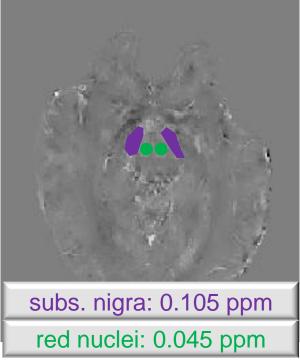


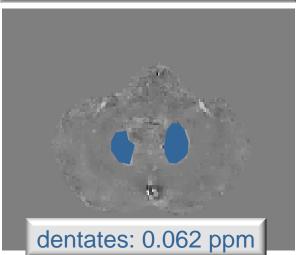






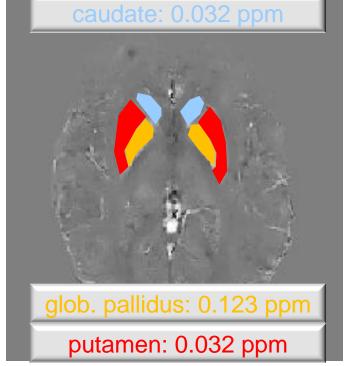
## In vivo QSM result: FOCUSS-QSM with magnitude prior





Structure	Δχ [ppm]
Globus Pallidus	12.3
Substantia Nigra	10.5
Dentate	6.2
Red Nucleus	4.5
Putamen	3.2
Caudate	2.3

x 0.01 ppm, relative to  $\chi_{CSF}$ 

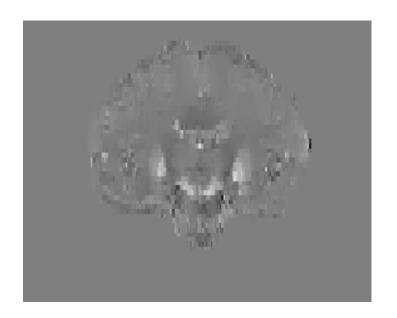


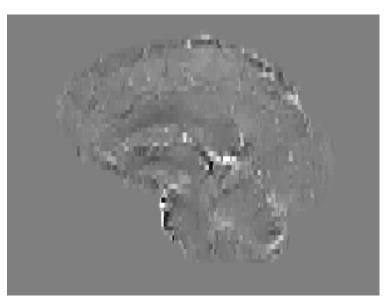


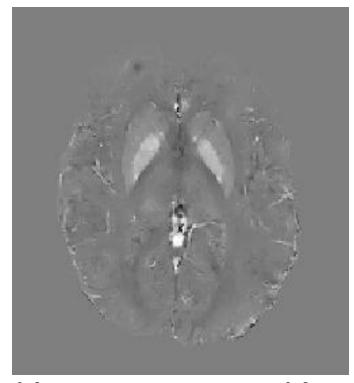




# In vivo QSM result: FOCUSS-QSM with magnitude prior



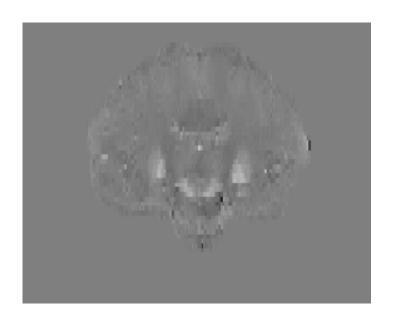




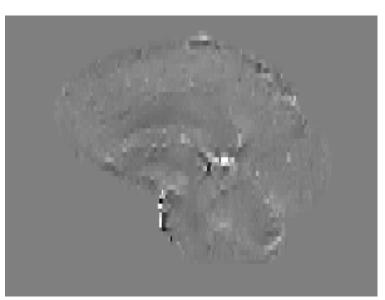


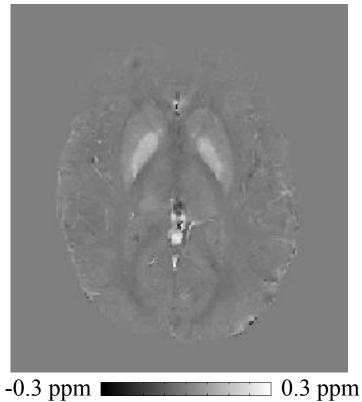


# In vivo QSM result: FOCUSS-QSM with a prior



Vessels are less apparent without the magnitude prior

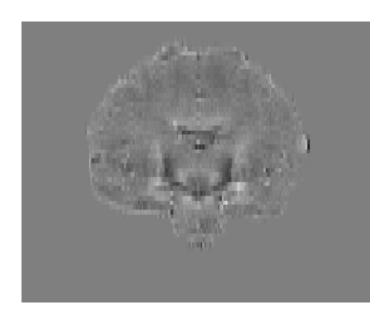


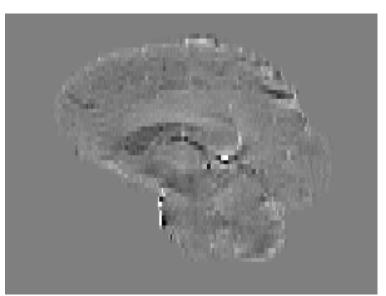


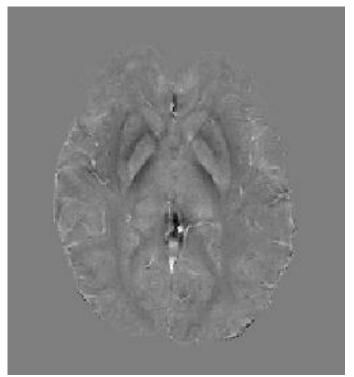




# **Corresponding Tissue Field Map:**





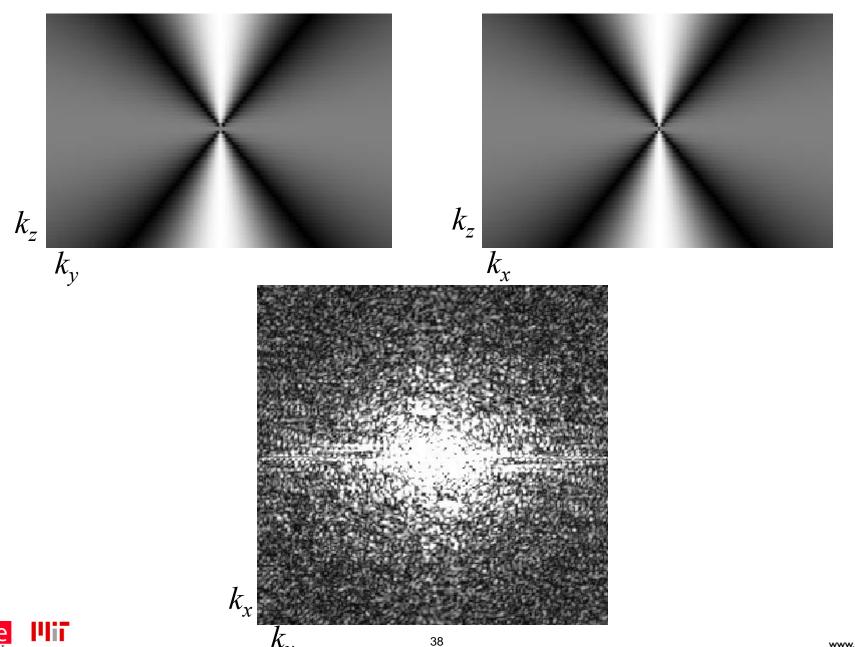






0.1 ppm -0.1 ppm I

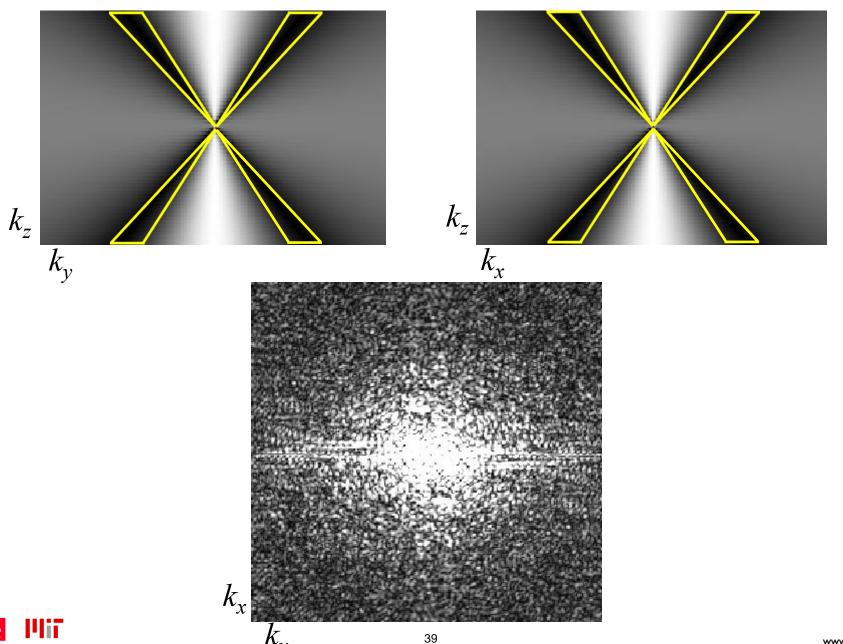
# In vivo QSM result with magnitude prior in k-space:







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#### **Potential drawbacks of FOCUSS-QSM**

- Computation time:
  - ❖ Dipole fitting for background removal ≈ 2 hours
  - FOCUSS-QSM ≈ 1 hours
  - ❖ Total processing time ≈ 3 hours for data of size [256x256x64]





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#### Solution:

Both algorithms solve Least Squares problems, Graphics Processing Card (GPU) implementation will greatly enhance the performance





#### **Conclusion**

- Starting with a multi-coil 3D GRE acquisition, we outlined coil combination and background phase elimination methods that yielded the tissue field map.
- We introduced a Quantitative Susceptibility Mapping algorithm that makes use of the magnitude image to facilitate the kernel inversion.



