



Wave-CAIPI: Highly Accelerated 3D Imaging with Reduced g-factor Penalty

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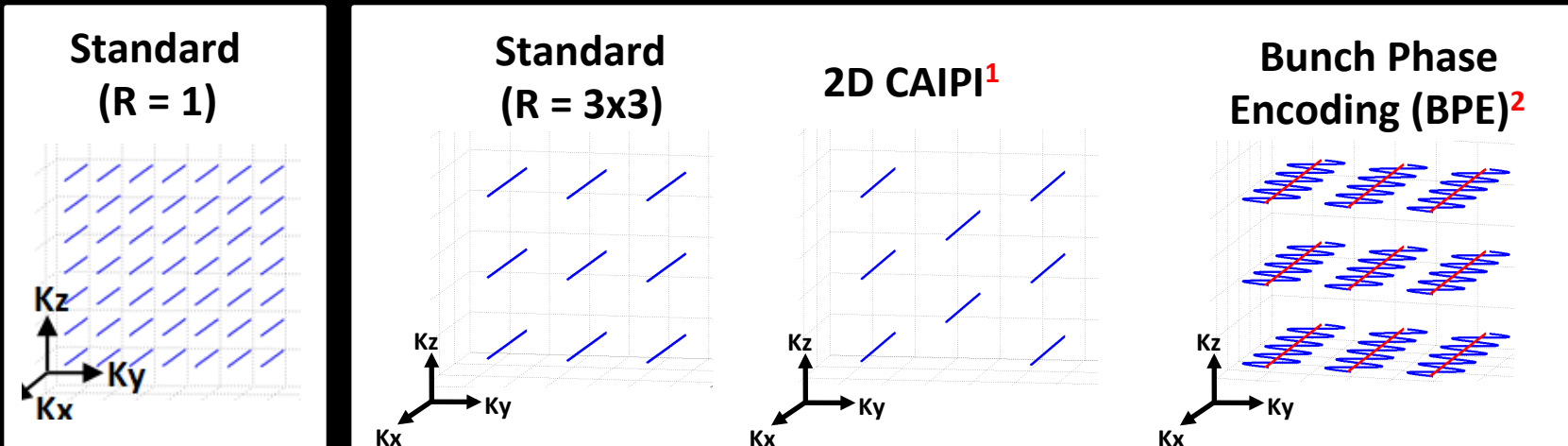
3 Boston Children's Hospital, Boston, MA

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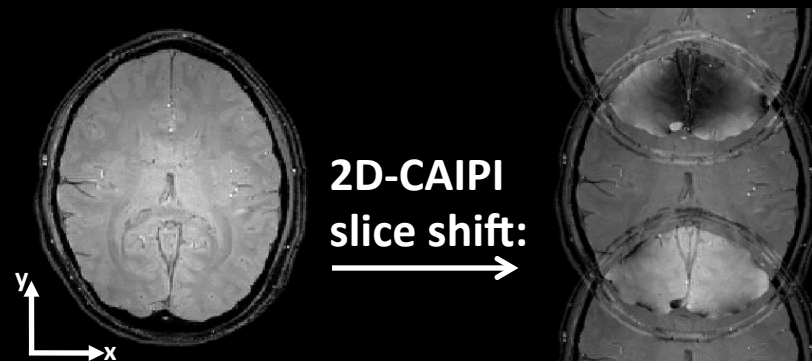
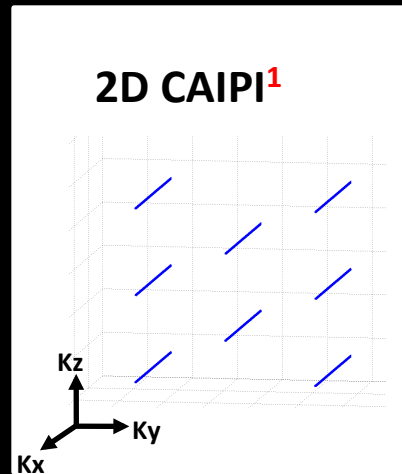
Introduction

- Recent modifications to rectilinear k-space sampling have provided more robust reconstructions of highly under-sampled datasets.



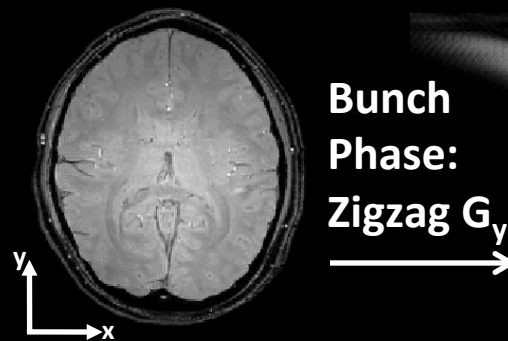
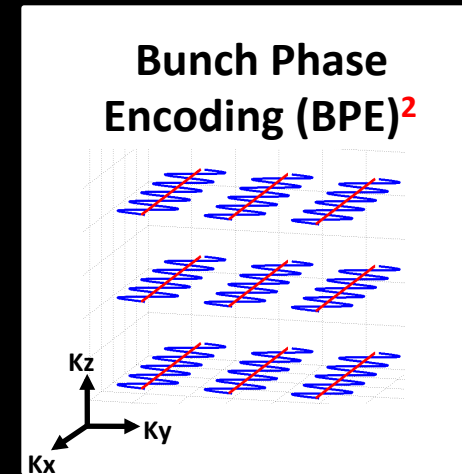
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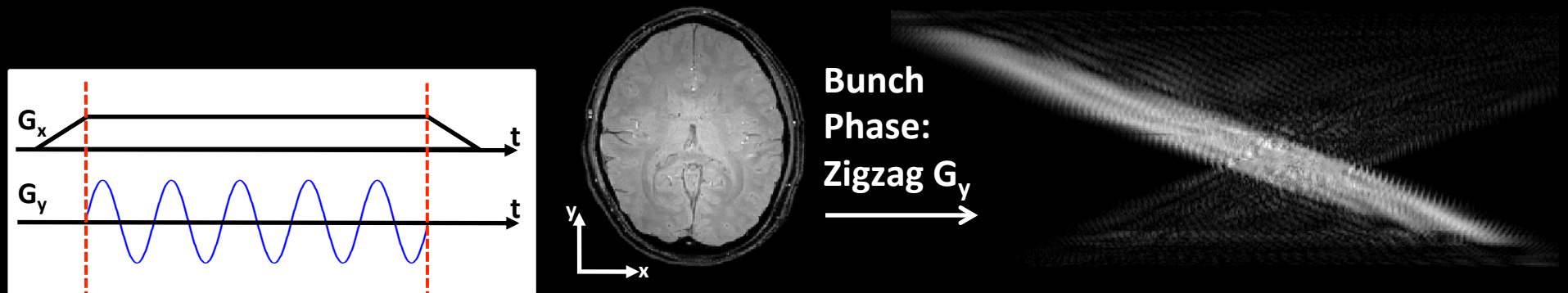
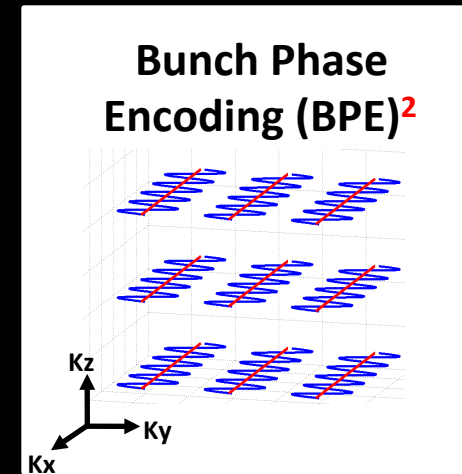
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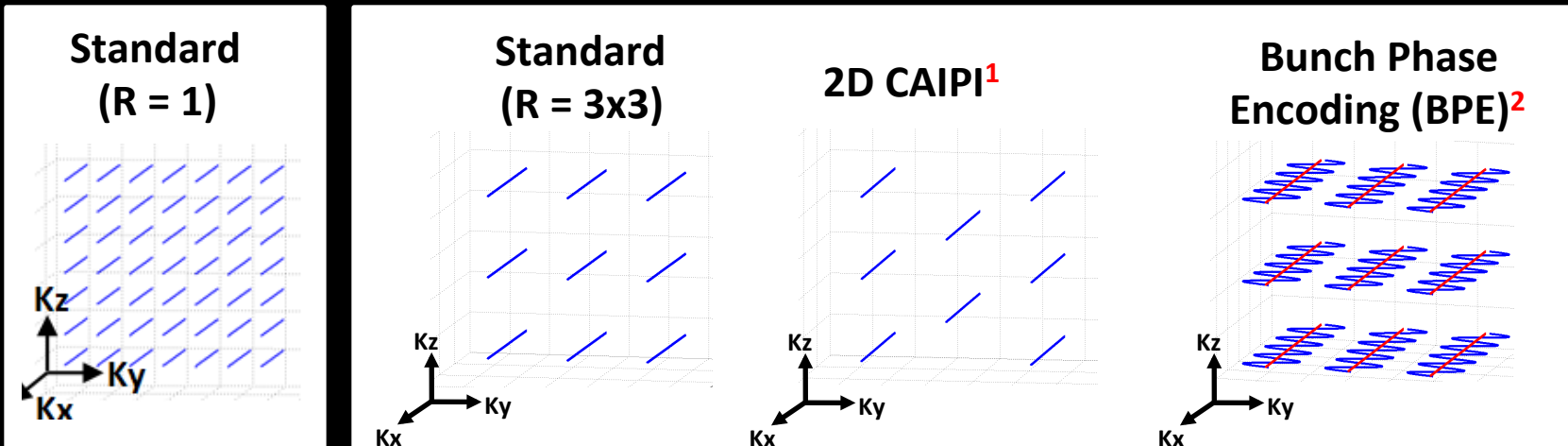
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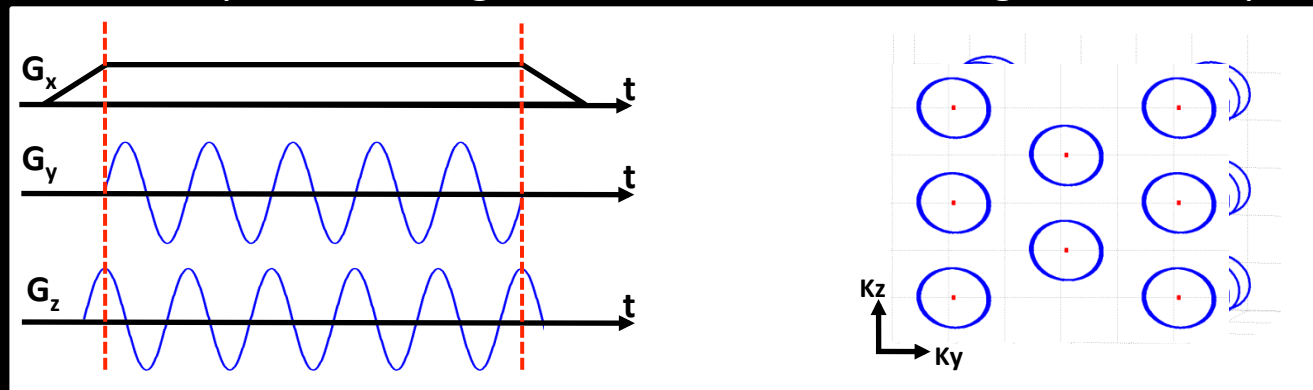


Introduction

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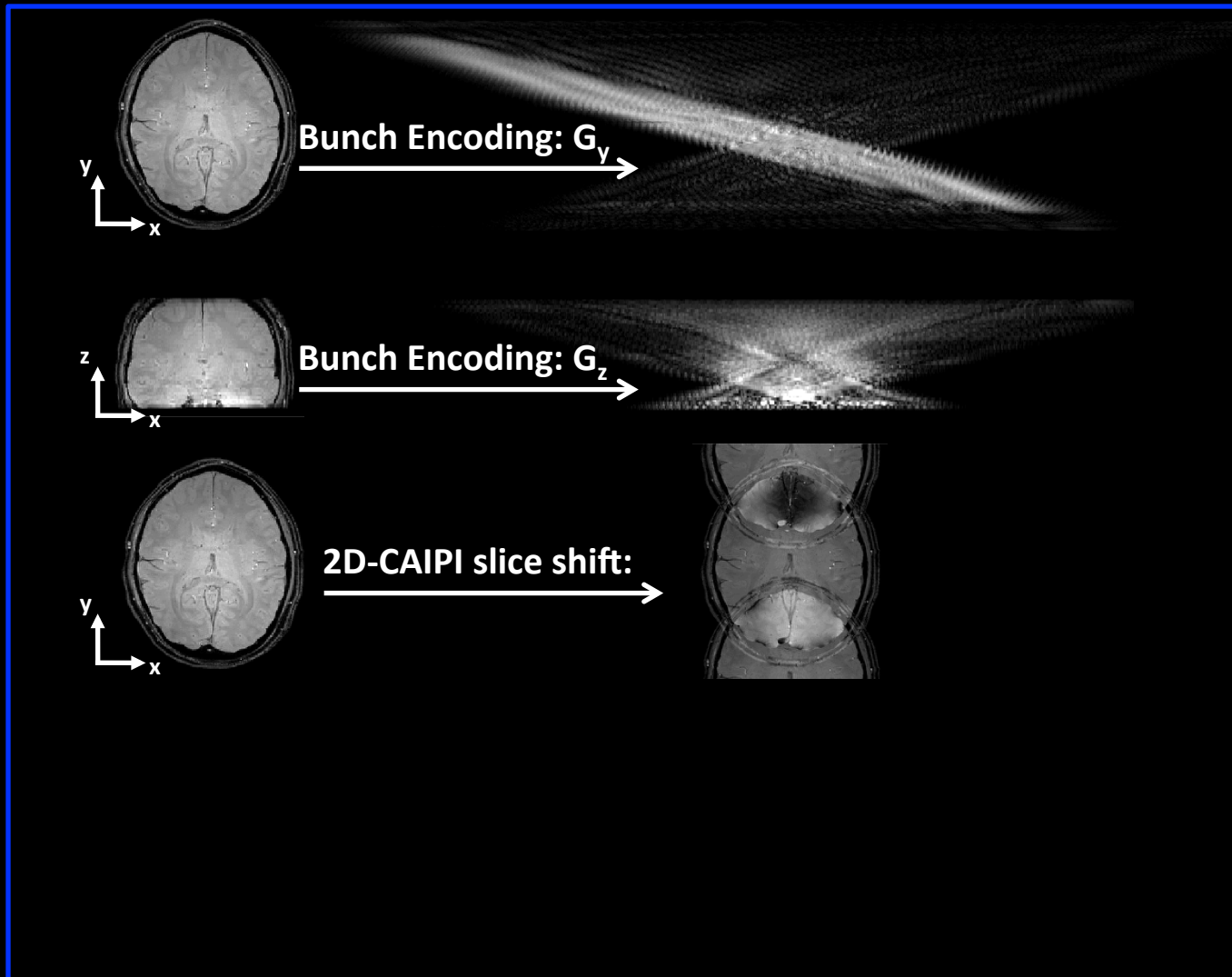


- Wave-CAIPI: 2D CAIPI + BPE in 2 direction
 - Spread aliasing in 3D to take full advantage of 3D coil profiles



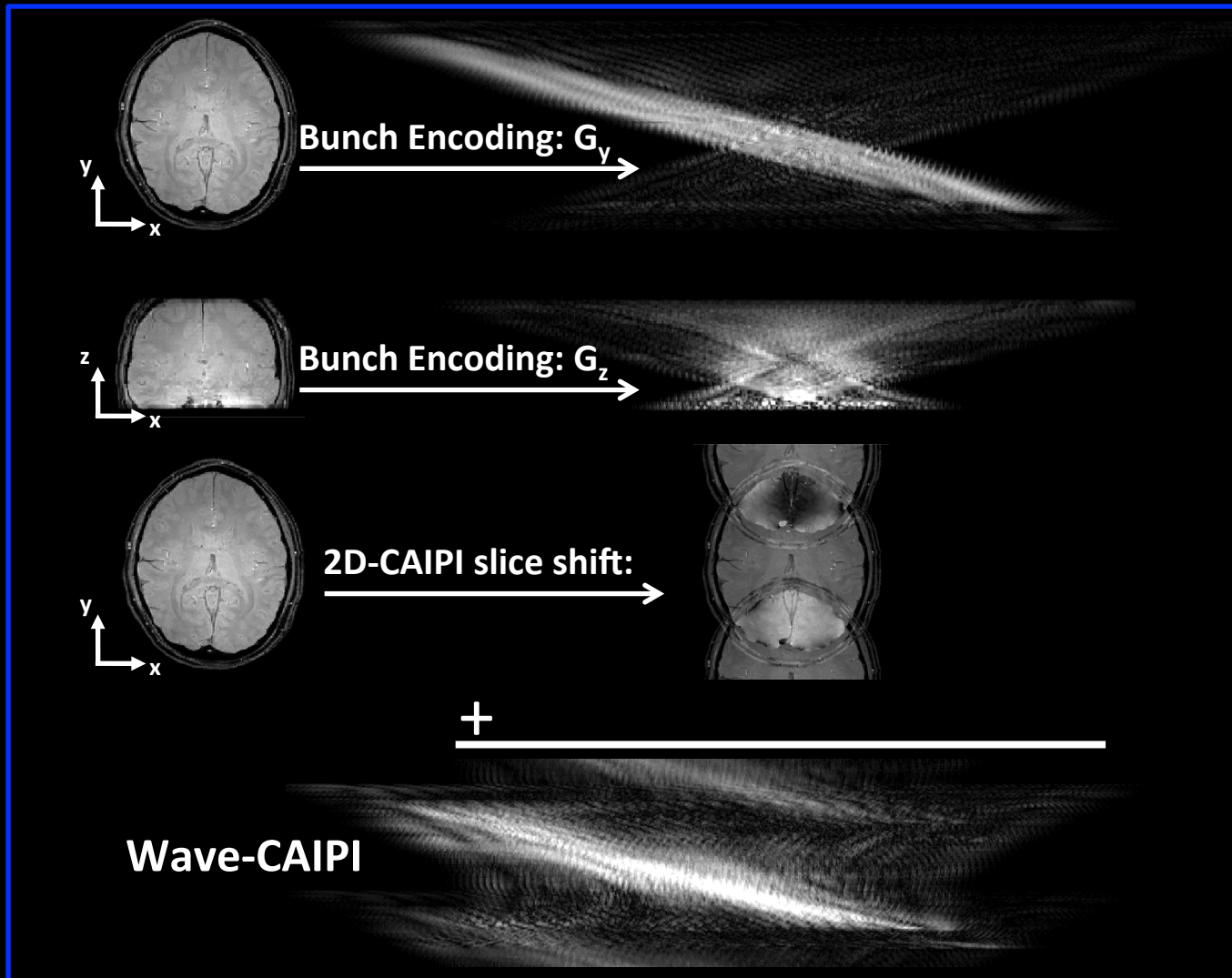
Effect of Wave Gradients

- Combination of G_y and G_z gradients with interslice shifts yield voxel spreading across three dimensions



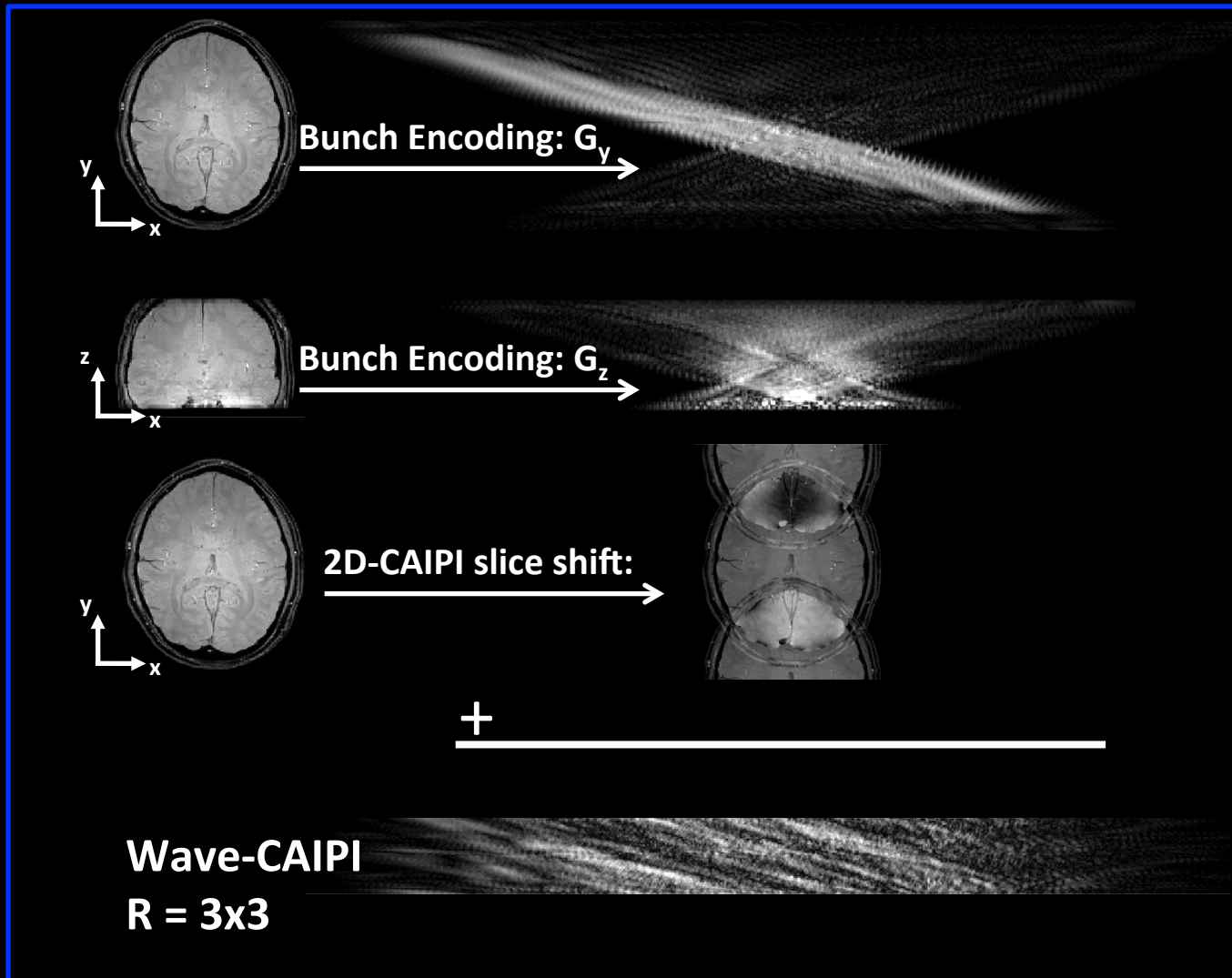
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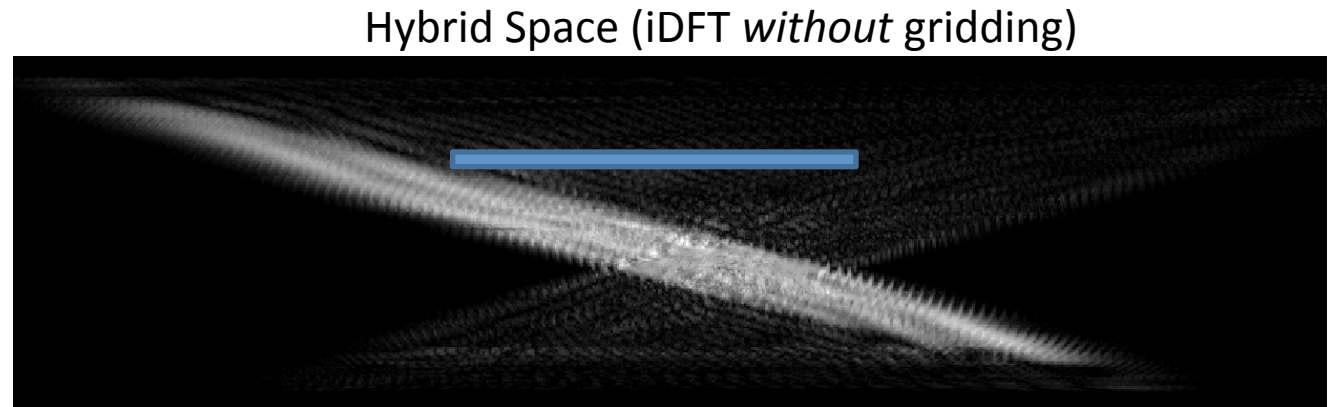
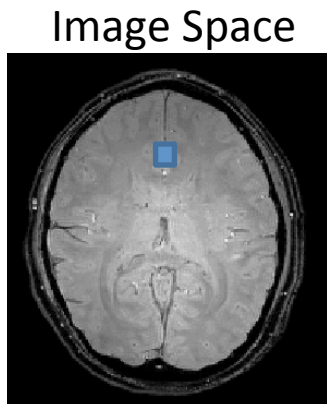
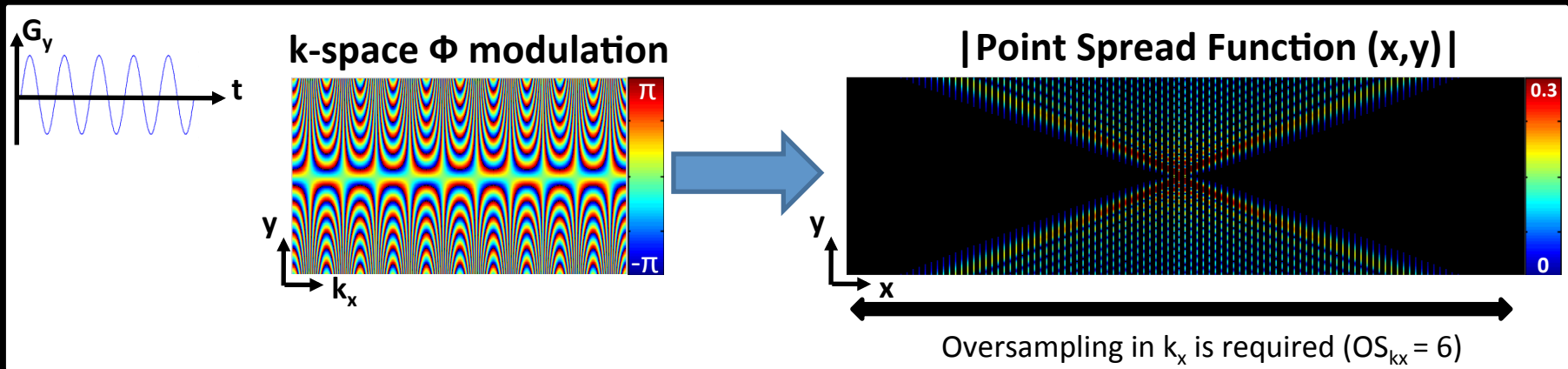
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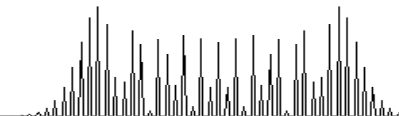


Point Spread Function

- Wave-CAIPI = BPE G_y + BPE G_z + CAIPI 2D
- View BPE G_y as extra modulation rather than modifying k-space traj.



$|\text{PSF}(x,y)|$



Point Spread Function

- From signal equation:

$$wave(x, y, z) = \sum_{k_x} e^{i2\pi x k_x / N} \cdot e^{-i2\pi W_y(k_x) y} \cdot \sum_x e^{-i2\pi x k_x / N} \cdot img(x, y, z)$$

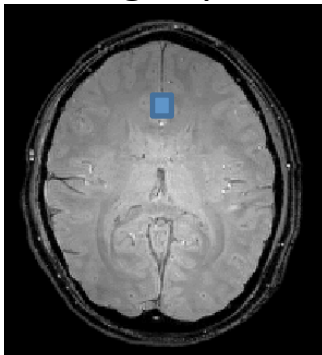
$wave(x, y, z)$ Wave image

$img(x, y, z)$ Underlying magnetization

$$W_y(k_x(t)) = \frac{\gamma}{2\pi} \int_0^t G_y(\tau) d\tau$$

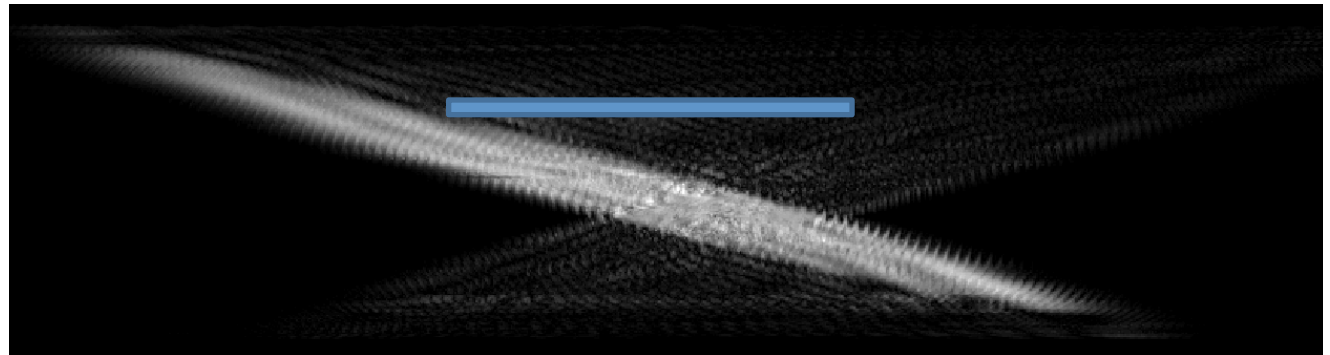
k-space trajectory

Image Space



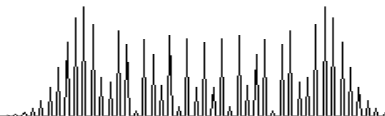
$img(x, y, z)$

Hybrid Space (iDFT *without* gridding)



$|PSF(x, y)|$

$wave(x, y, z)$

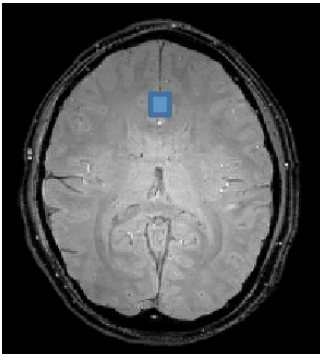


Point Spread Function

- From signal equation:

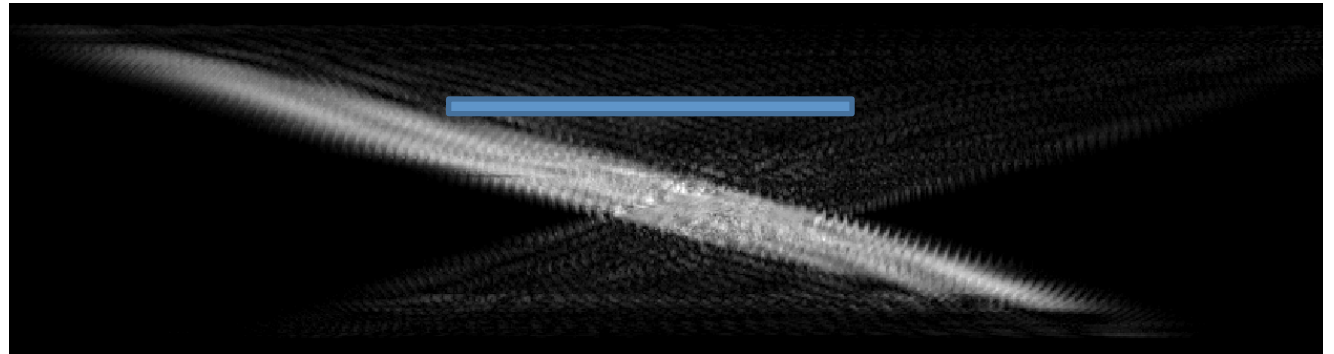
$$wave(x, y, z) = \underbrace{\sum_{k_x} e^{i2\pi x k_x / N}}_{\text{Inverse Discrete Fourier Transform}} \cdot e^{-i2\pi W_y(k_x) y} \cdot \underbrace{\sum_x e^{-i2\pi x k_x / N}}_{\text{Discrete Fourier Transform}} \cdot img(x, y, z)$$

Image Space



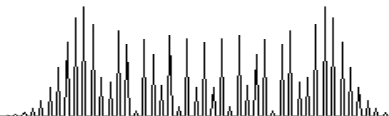
$img(x, y, z)$

Hybrid Space (iDFT *without* gridding)



$|PSF(x, y)|$

$wave(x, y, z)$



Point Spread Function

- From signal equation:

$$wave(x, y, z) = F^{-1} \cdot \underbrace{e^{-i2\pi W_y(k_x)y}}_{\text{Point Spread Function (PSF)}} \cdot F \cdot img(x, y, z)$$

Point Spread Function (PSF)

No need for gridding, simple DFT

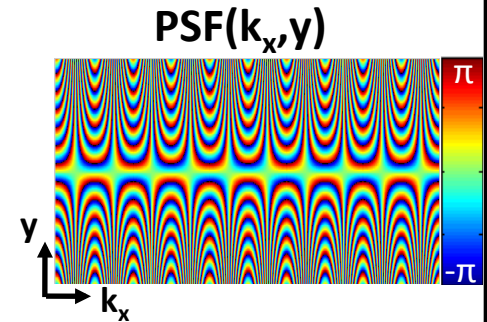
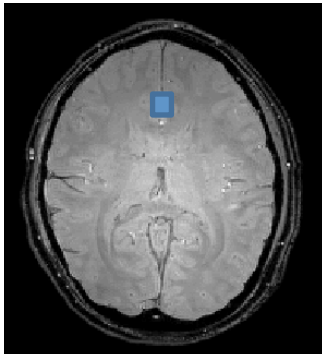
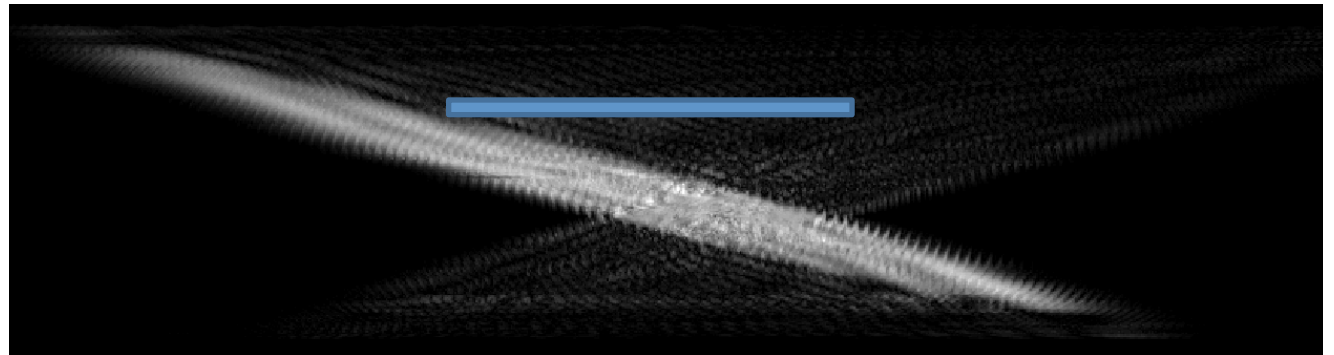


Image Space



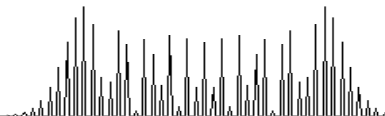
$img(x, y, z)$

Hybrid Space (iDFT *without* gridding)

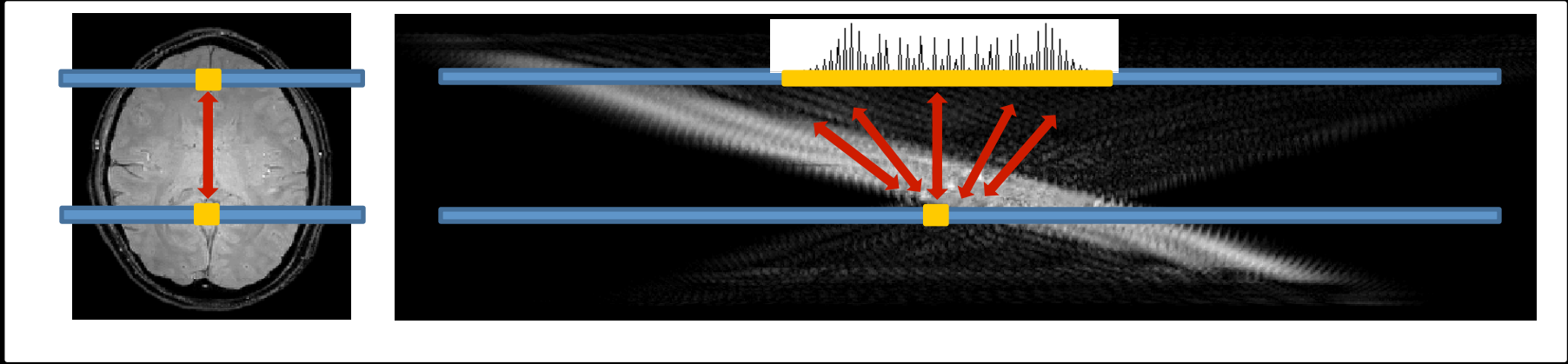


$|PSF(x, y)|$

$wave(x, y, z)$

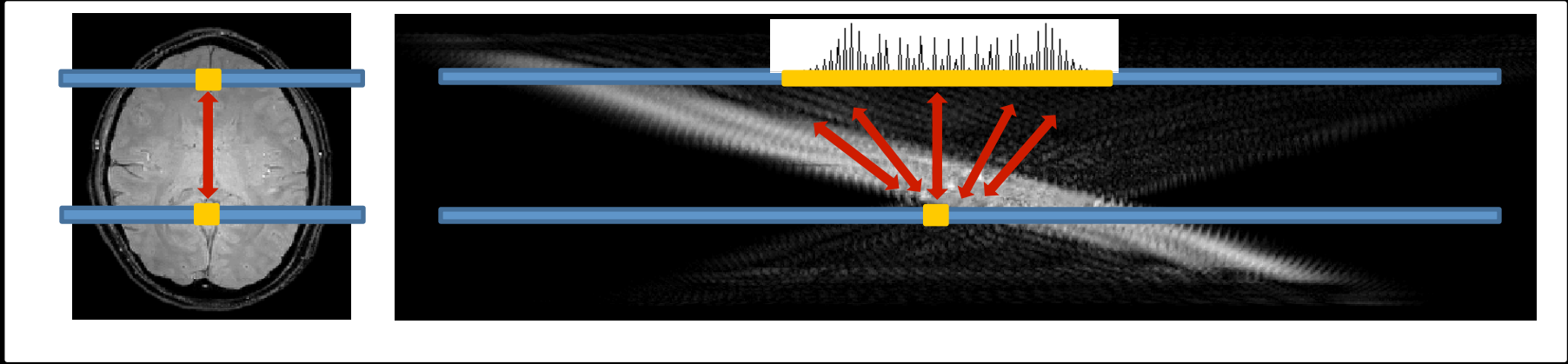


Reconstruction for accelerated acquisitions



- $R_{\text{inplane}} = 2$ => pair-wise aliasing of two rows of voxels
- => small Encoding matrix for each pair
- => separable and easy to solve
- => intuition on why Wave improves reconstruction

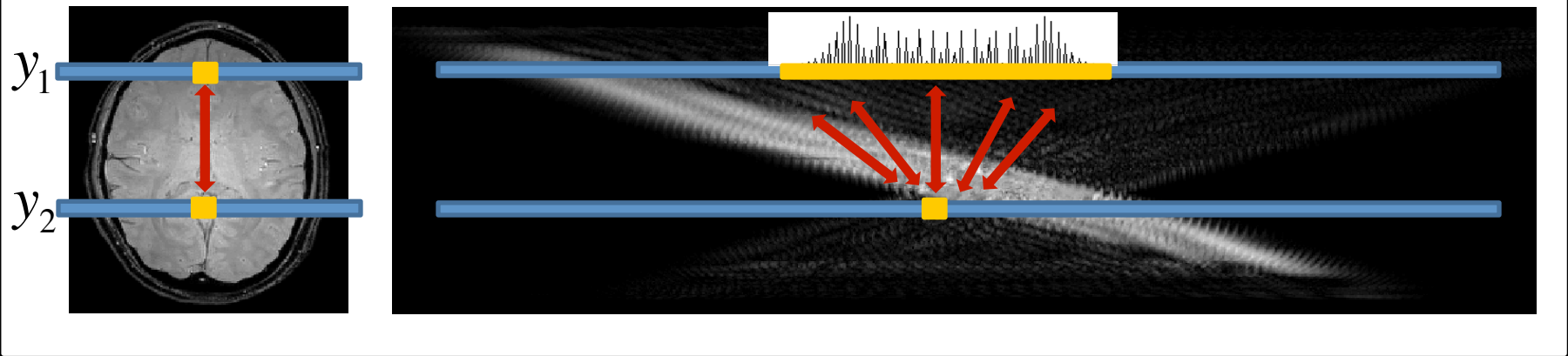
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Reconstruction for accelerated acquisitions

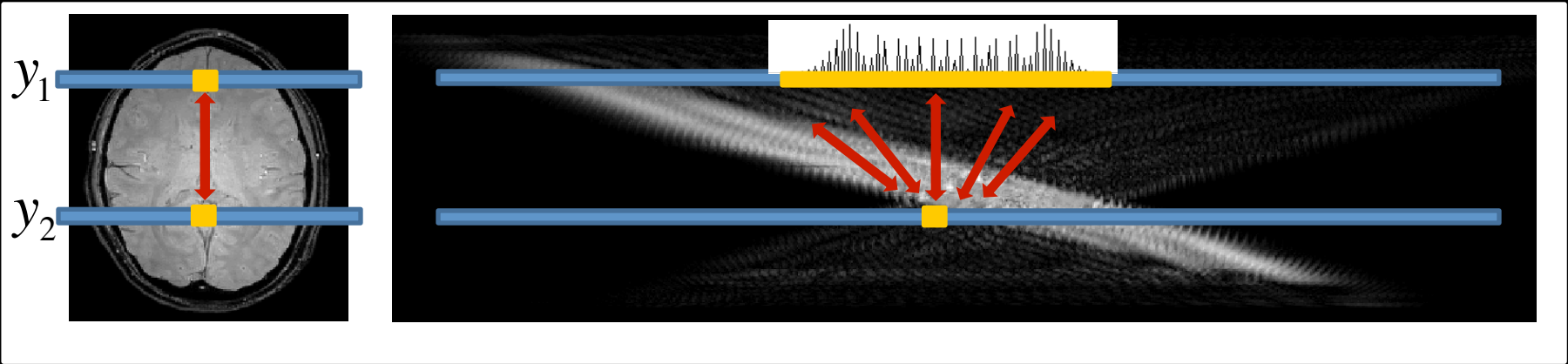


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$$\text{wave}(y) = F^{-1} \cdot \text{Psf}(y) \cdot F \cdot \text{row}(y)$$

$$\begin{bmatrix} F^{-1} \cdot \text{Psf}(y_1) \cdot F \\ F^{-1} \cdot \text{Psf}(y_2) \cdot F \end{bmatrix} \cdot \begin{bmatrix} \text{row}(y_1) \\ \text{row}(y_2) \end{bmatrix} = [\text{wave}]$$

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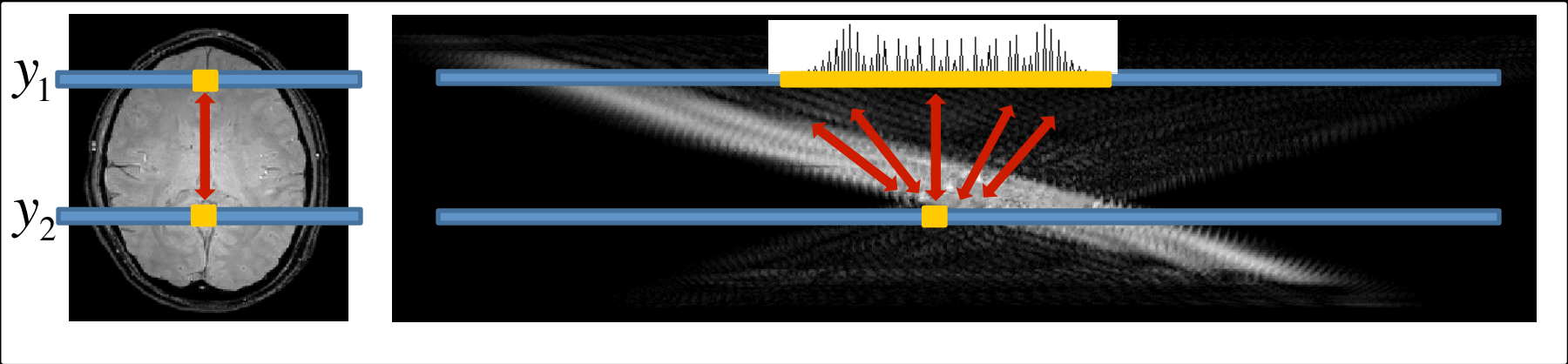


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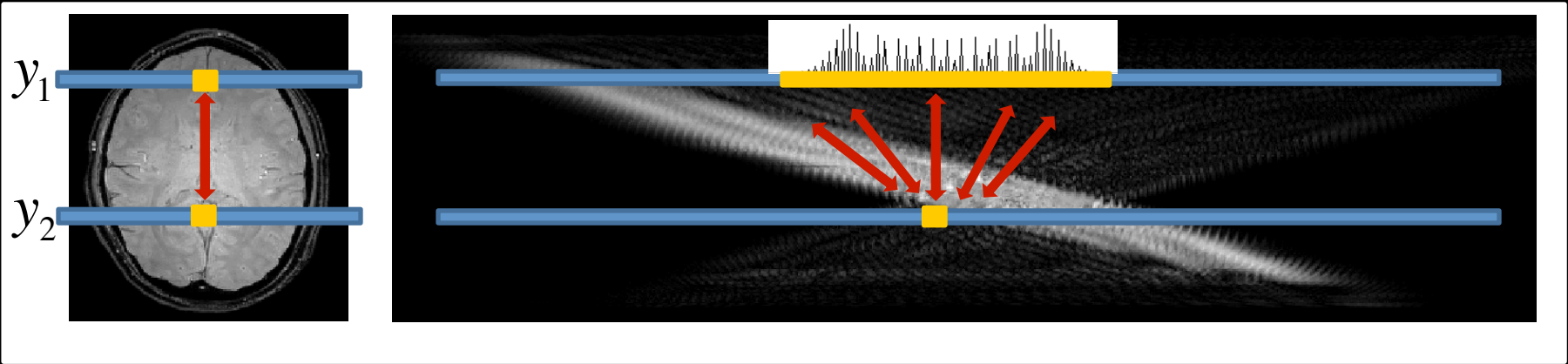
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$$\begin{aligned} & \text{wave}(y) = F^{-1} \cdot \text{Psf}(y) \cdot F \cdot \text{row}(y) \\ & \underbrace{\begin{bmatrix} F^{-1} \cdot \text{Psf}(y_1) \cdot F \cdot C_1(y_1) \\ \dots \\ F^{-1} \cdot \text{Psf}(y_2) \cdot F \cdot C_{32}(y_2) \end{bmatrix}}_{\text{Encoding matrix}} \cdot \begin{bmatrix} \text{row}(y_1) \\ \text{row}(y_2) \end{bmatrix} = \begin{bmatrix} \text{wave}_1 \\ \dots \\ \text{wave}_{32} \end{bmatrix} \end{aligned}$$

Reconstruction for accelerated acquisitions



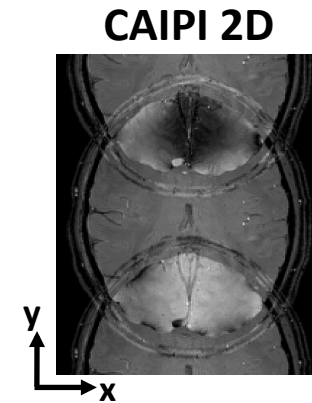
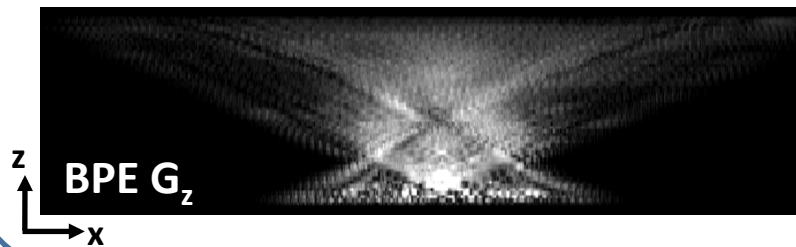
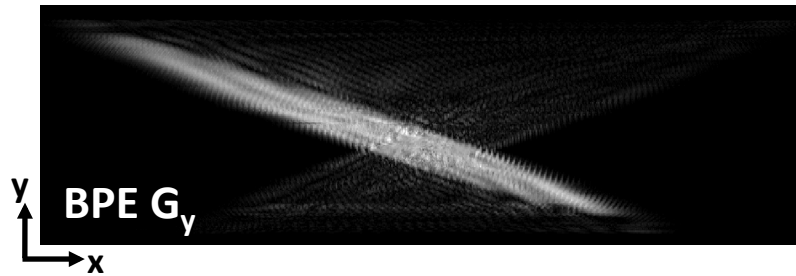
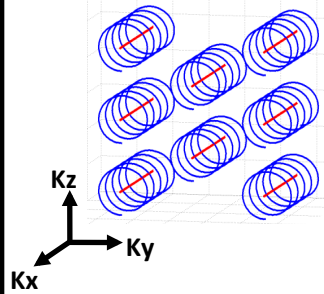
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- Solve for each set of collapsed rows iteratively using LSQR

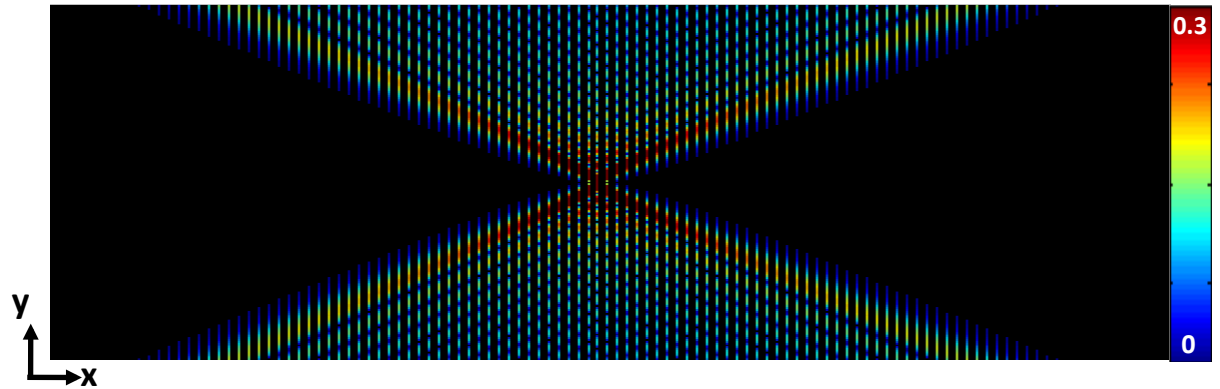
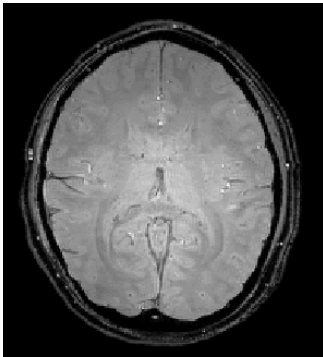
Wave-CAIPI reconstruction



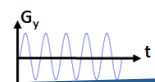
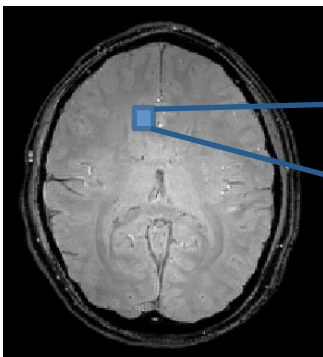
- ⇒ Wave gradients G_y and G_z create position dependent PSF
- ⇒ CAIPI 2D shift aliasing pattern
- ⇒ These are accounted for when generating the PSF-based Encoding matrices
 - ⇒ Ex: $R = 3 \times 3$
 - ⇒ each Encoding matrix corresponds to 9 rows of the image
 - ⇒ grouping of rows is determined by CAIPI 2D
 - ⇒ amount of spreading in each row determined by G_y and G_z

Artifact Quantification

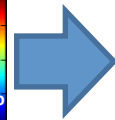
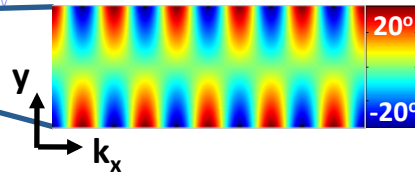
PSF over image FOV_y



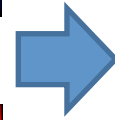
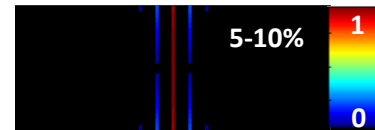
How about PSF over a voxel?



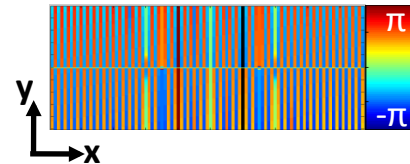
k-space differential Φ modulation



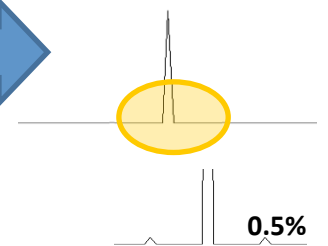
|PSF|



Φ (PSF)



PSF integrated across y



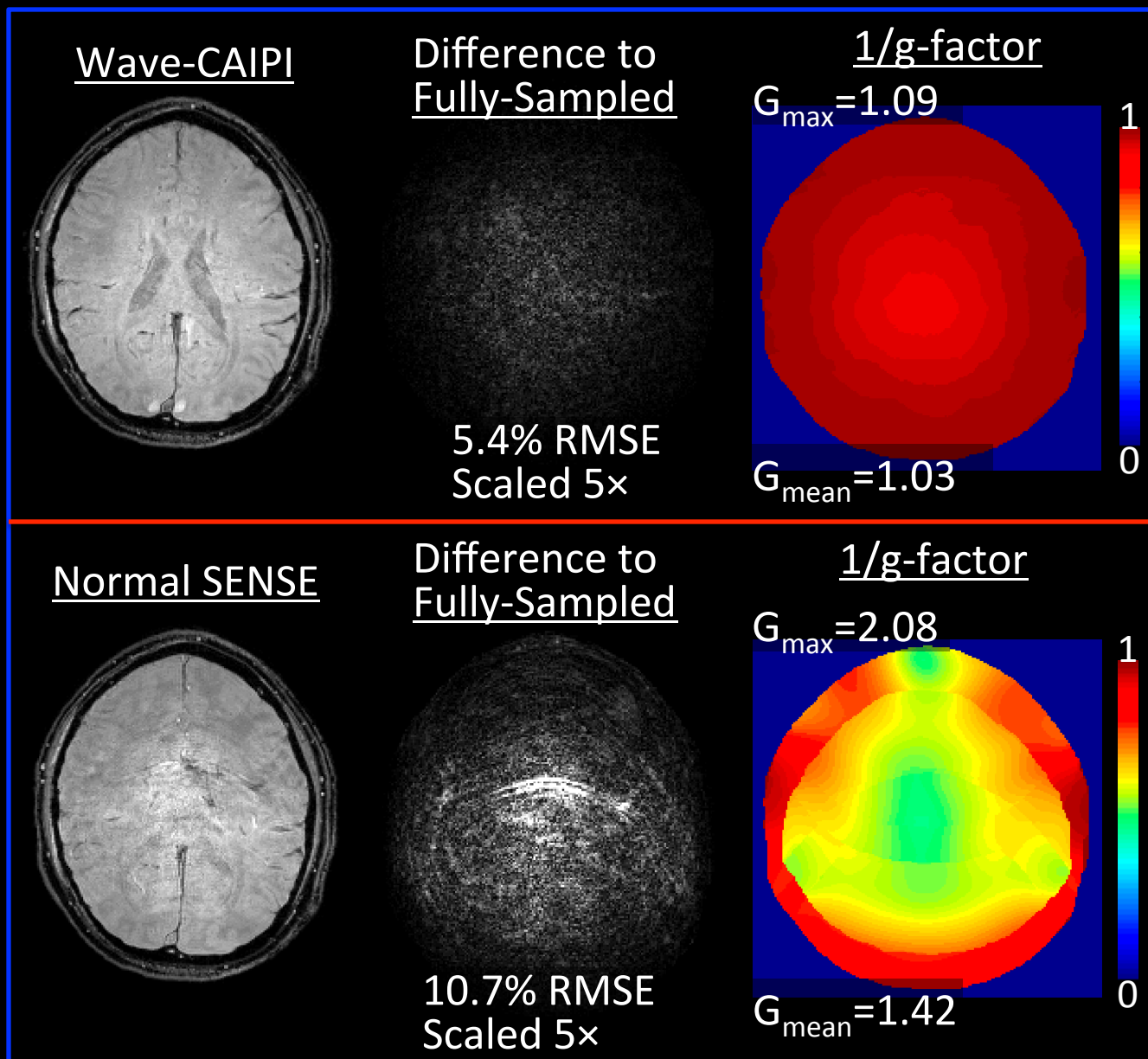
In Vivo Acquisition Comparison

- Compare Wave-CAIPI and conventional SENSE
- Acquire **fully-sampled** data, then accelerate by $R = 3 \times 3$
- Compute root-mean-square error (RMSE) and $1/g$ -factor maps (retained SNR)

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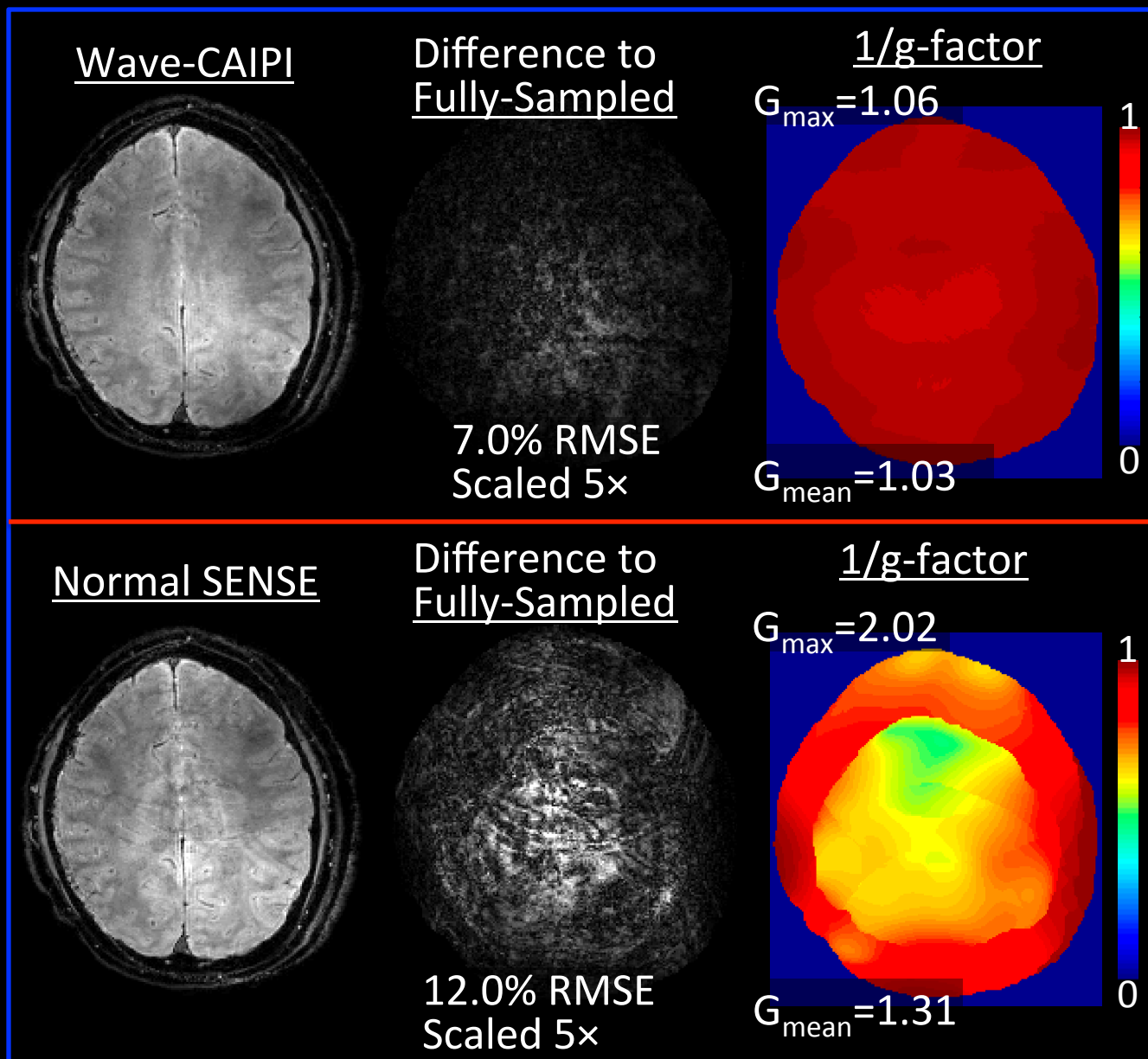
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- In vivo acquisitions:
 - At 3T and 7T
 - $1 \times 1 \times 2$ mm resolution
 - $224 \times 224 \times 120$ FOV

3 Tesla, R=3x3, 1x1x2 mm³, T_{acq}=38s



TR/TE = 26/13.3 ms

7 Tesla, R=3x3, 1x1x2 mm³, T_{acq}=40s



TR/TE = 27/10.9 ms

Accelerated Acquisition Comparison

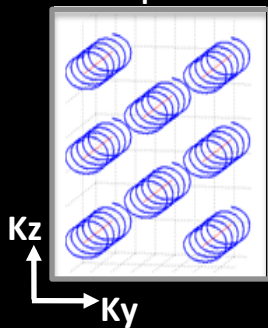
- Compare Wave-CAIPI, 2D-CAIPI¹ and Bunch Phase²
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- Compute 1/g-factor maps (retained SNR)

Accelerated Acquisition Comparison

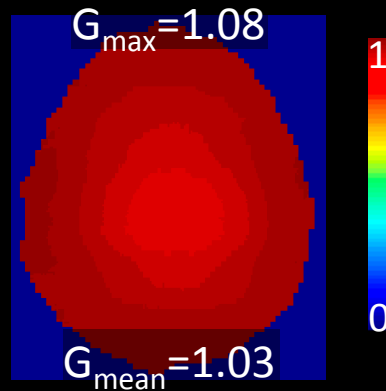
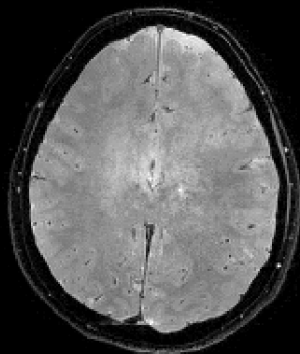
- Compare Wave-CAIPI, 2D-CAIPI¹ and Bunch Phase²
- Acquire R = 3x3 accelerated data
- Compute 1/g-factor maps (retained SNR)
- In vivo acquisitions:
 - At 3T and 7T
 - 1x1x1 mm isotropic resolution
 - Acquisition time: 2.3 min
 - 240x240x120 FOV

3 Tesla, R=3x3, 1x1x1 mm³, T_{acq}=2.3 min

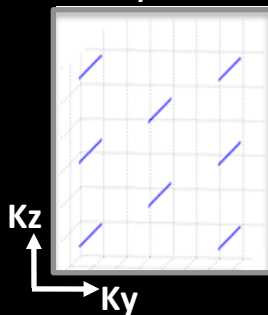
k-space



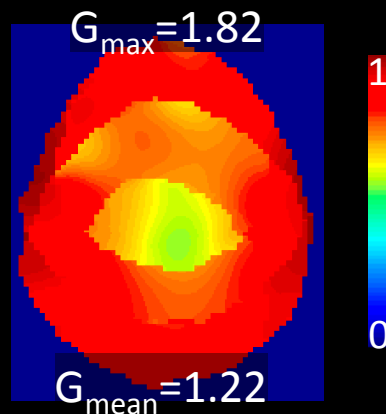
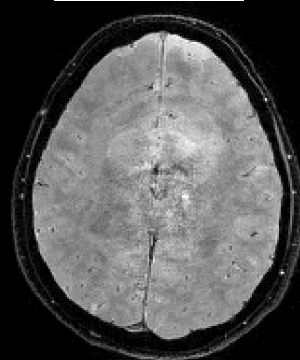
Wave-CAIPI



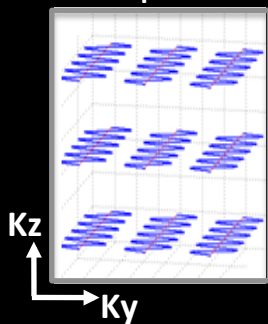
k-space



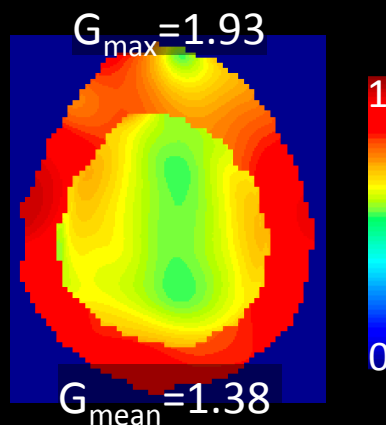
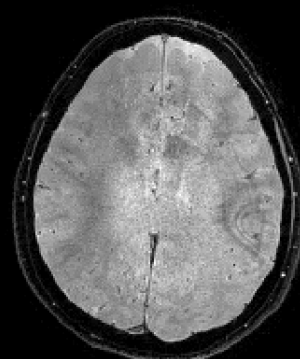
2D-CAIPI



k-space

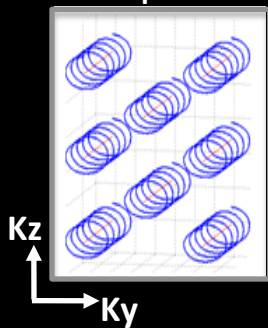


Bunch Encoding

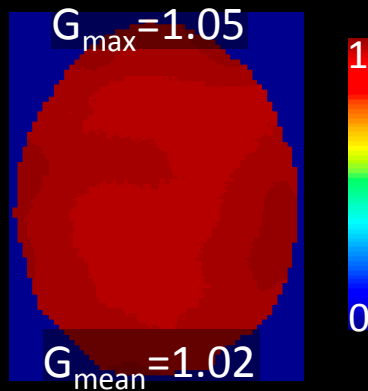
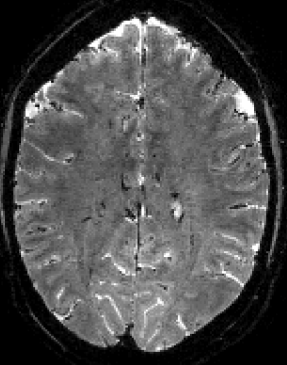


7 Tesla, R=3x3, 1x1x1 mm³, T_{acq}=2.3 min

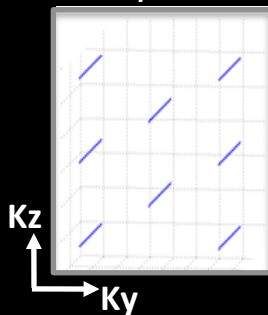
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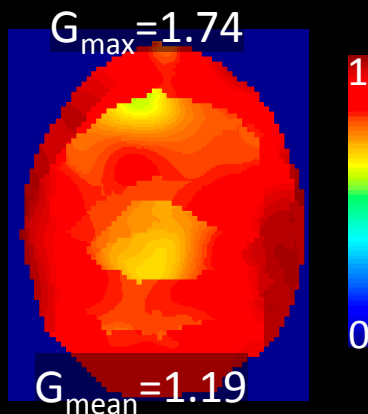
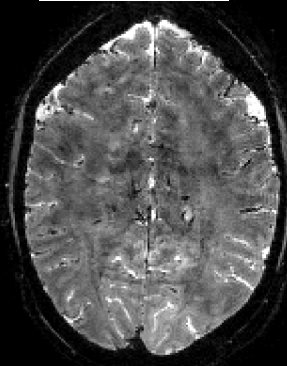
Wave-CAIPI



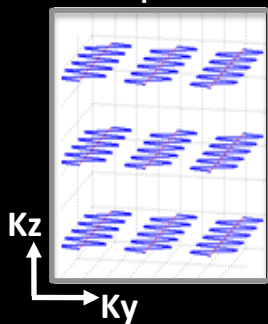
k-space



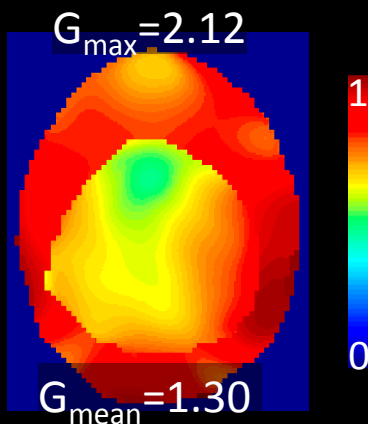
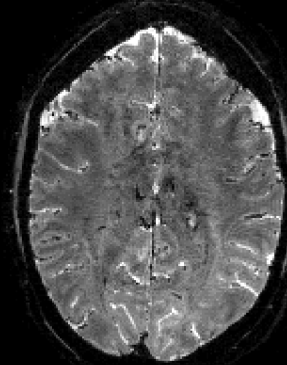
2D-CAIPI



k-space

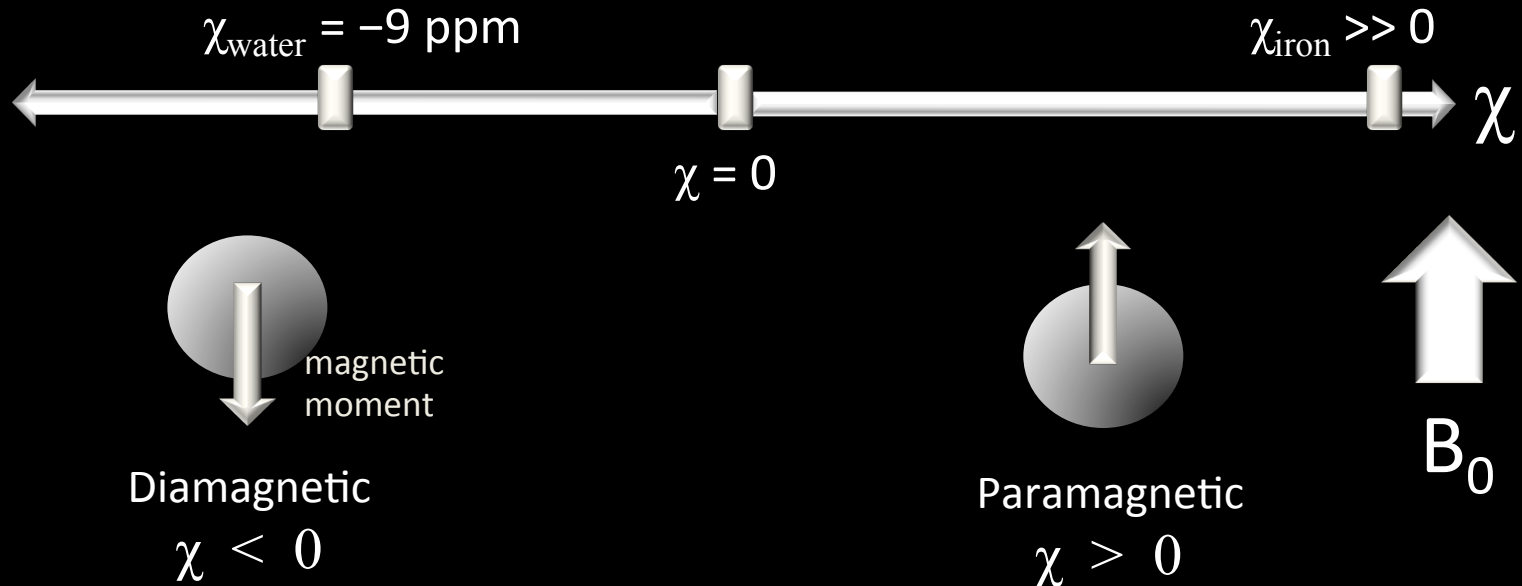


Bunch Encoding



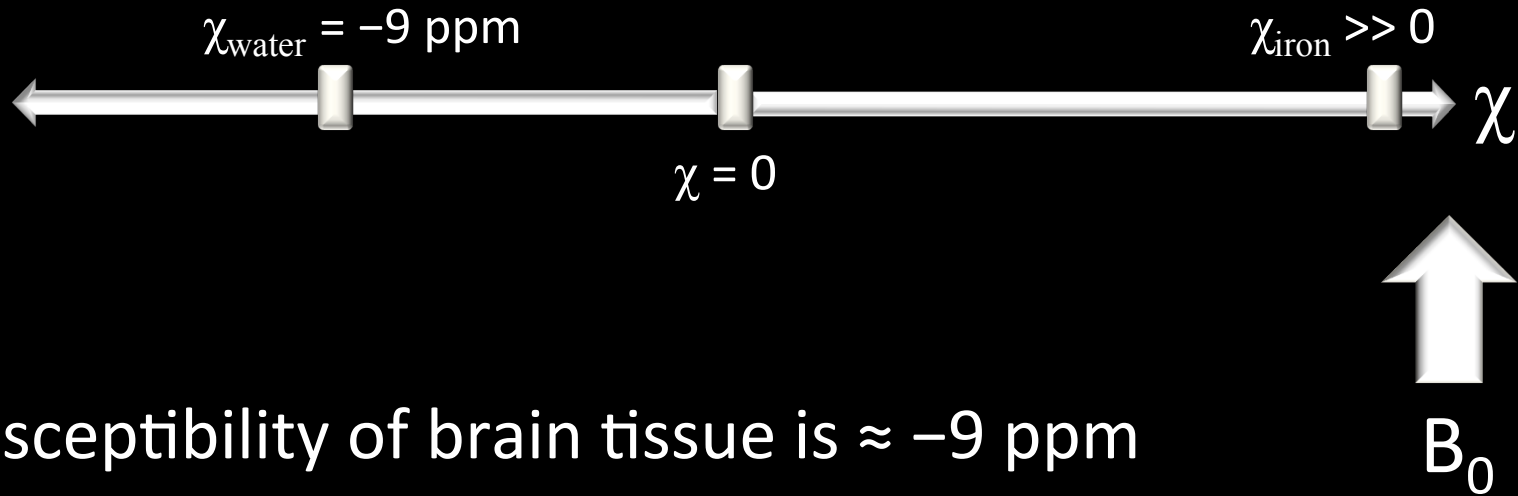
Susceptibility of Tissue

- Susceptibility χ : degree of magnetization of a material when placed in a magnetic field



Susceptibility of Tissue

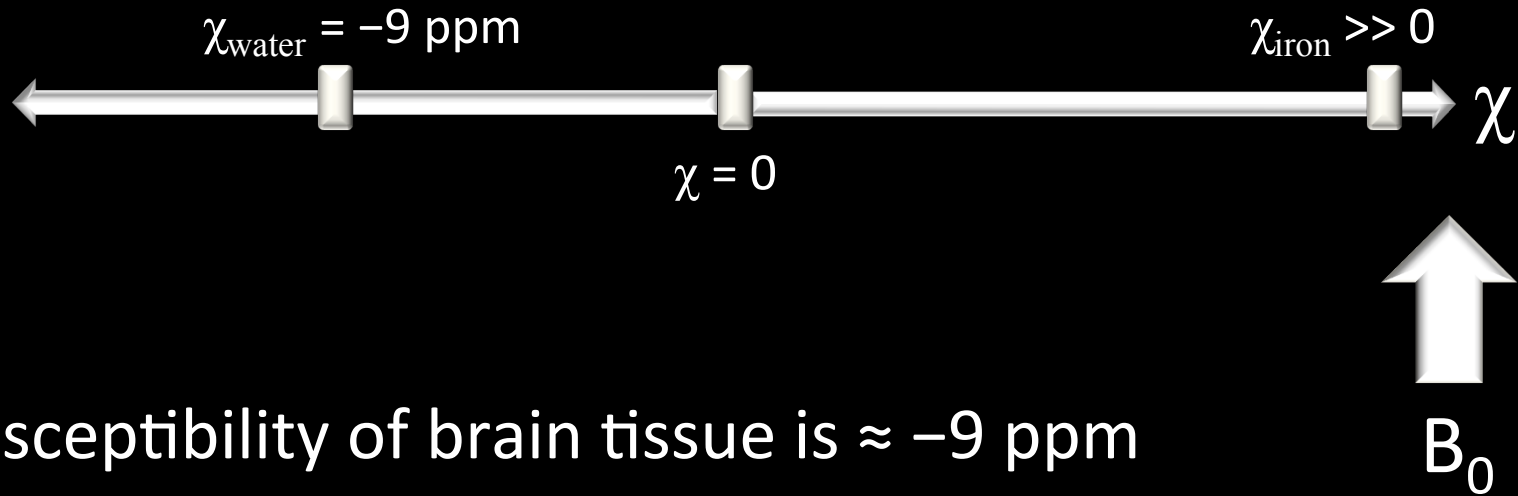
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- Susceptibility of brain tissue is $\approx -9 \text{ ppm}$
- Tissues with increased iron deposition are relatively paramagnetic $\rightarrow \chi$ is more positive

Susceptibility of Tissue

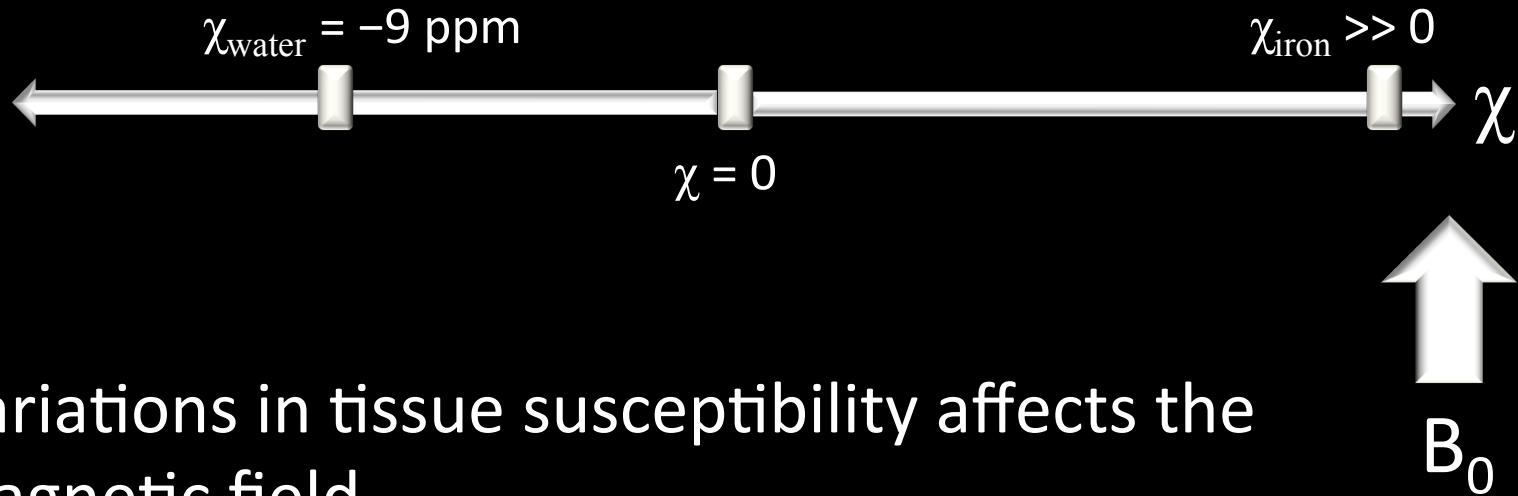
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- Susceptibility of brain tissue is $\approx -9 \text{ ppm}$
- Tissues with increased iron deposition are relatively paramagnetic $\rightarrow \chi$ is **more positive**
- Excessive iron concentration occurs in a variety of degenerative diseases,
e.g. Alzheimer's, multiple sclerosis, Parkinson's

Susceptibility of Tissue

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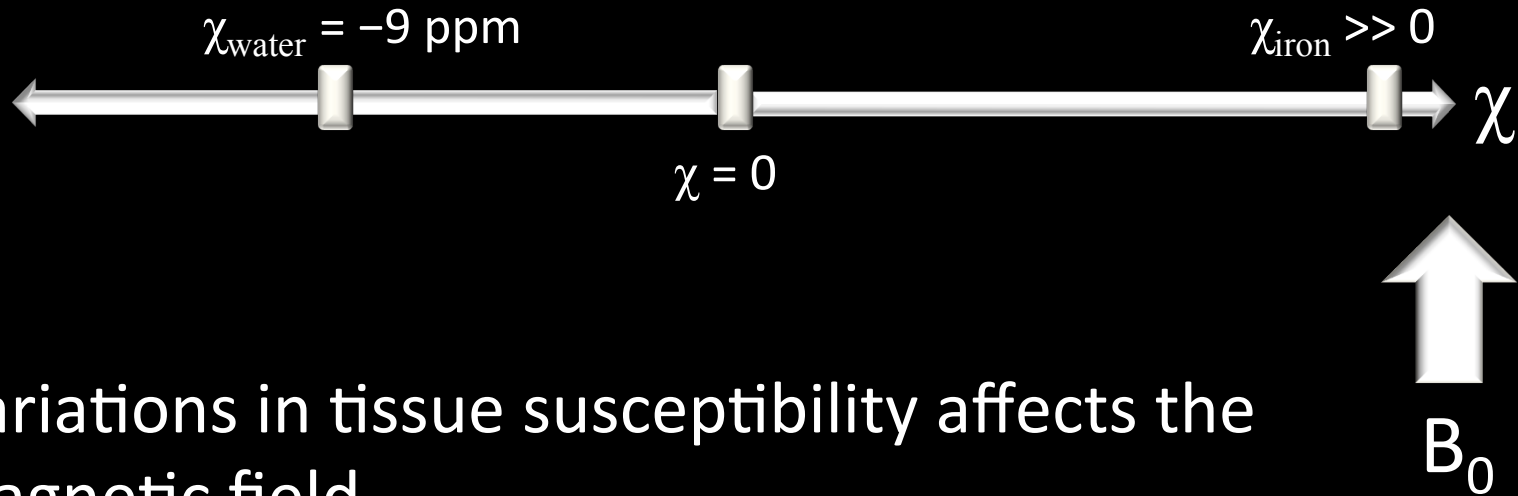


- Variations in tissue susceptibility affects the magnetic field

$\Delta\chi \rightarrow$ magnetic field perturbation

Susceptibility of Tissue

- Susceptibility χ : degree of magnetization of a material when placed in a magnetic field



- Variations in tissue susceptibility affects the magnetic field
- Field perturbation causes a change in MR signal phase



Quantitative Susceptibility Mapping (QSM)

- Quantitative Susceptibility Mapping (QSM) aims to quantify tissue magnetic susceptibility with applications such as,
 - ❖ Tissue contrast enhancement¹
 - ❖ Estimation of venous blood oxygenation²
 - ❖ Quantification of tissue iron concentration³

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- Estimation of the susceptibility map χ from the unwrapped phase φ involves solving an inverse problem,

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

\mathbf{F} : Discrete Fourier Transform

\mathbf{D} : susceptibility kernel

$\delta = \varphi / (\gamma \cdot TE \cdot B_0)$: normalized field map

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δ (measured) χ (estimate)

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Quantitative Susceptibility Mapping (QSM)

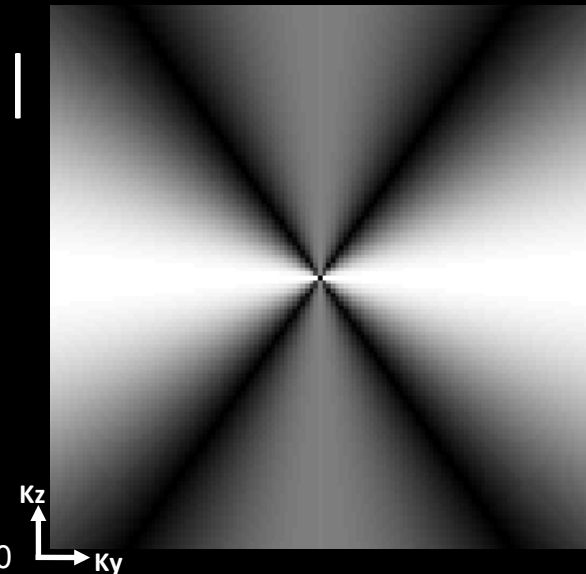
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$|\mathbf{D}|$

- The inversion is made difficult by zeros in susceptibility kernel \mathbf{D}

$$\mathbf{D} = \frac{1}{3} - \frac{k_z^2}{k_x^2 + k_y^2 + k_z^2}$$



¹ Duyn JH *et al.*, PNAS 2007

² Fan AP *et al.*, ISMRM 2010

³ Liu T *et al.*, ISMRM 2010

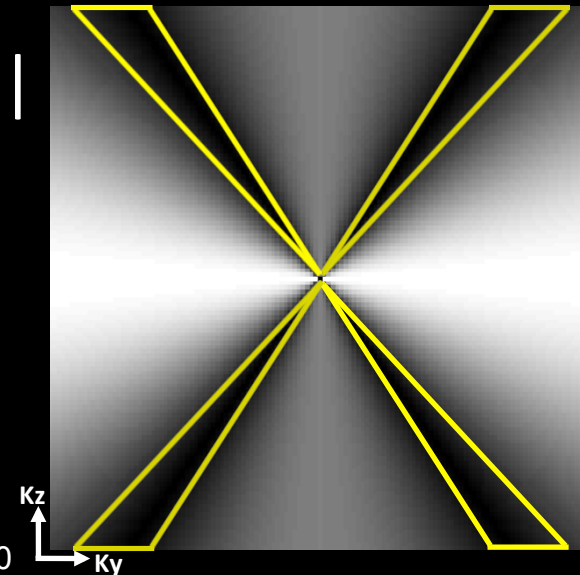
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- Undersampling is due to physics
Not in our control



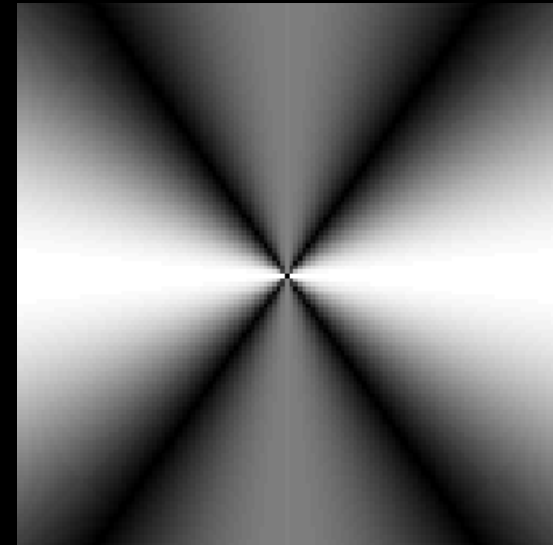
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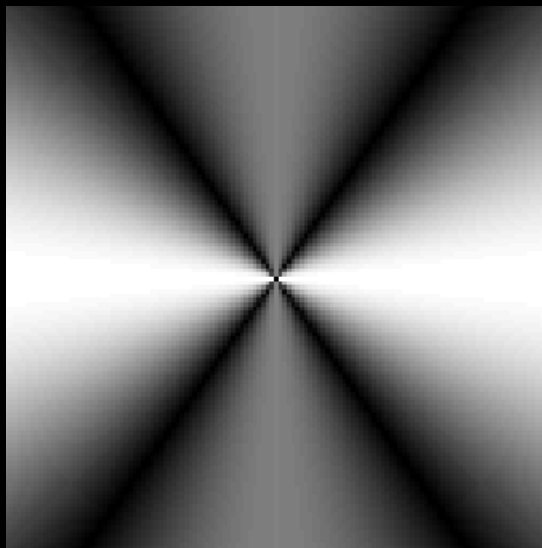
Regularized Inversion for QSM

$|D|$

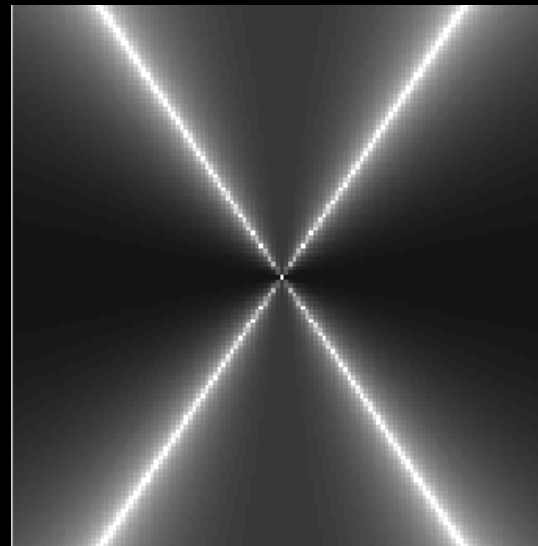


Regularized Inversion for QSM

$|\mathbf{D}|$



$\log|\mathbf{D}^{-1}|$



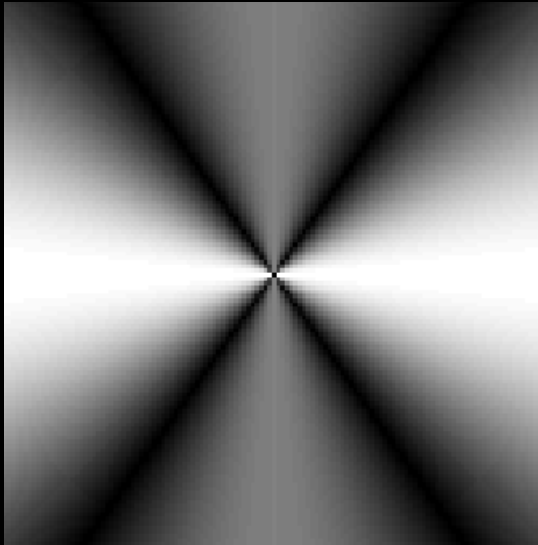
$$\delta = \mathbf{F}^{-1} \mathbf{D} \mathbf{F} \chi$$



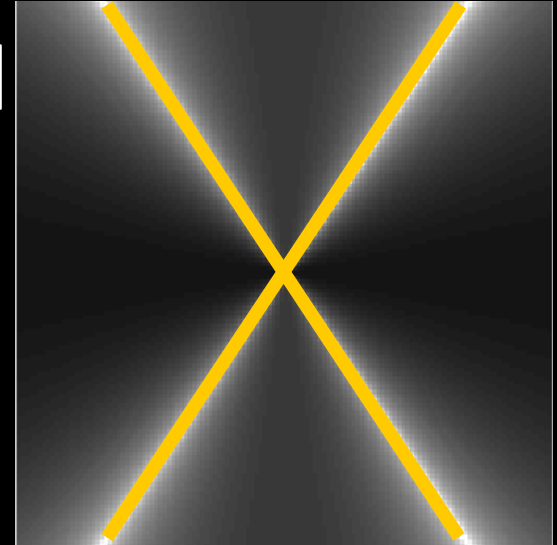
$$\chi = \mathbf{F}^{-1} \mathbf{D}^{-1} \mathbf{F} \delta$$

Regularized Inversion for QSM

$|\mathbf{D}|$



$\log |\mathbf{D}^{-1}|$



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



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

diverges to ∞

- Solving for χ by convolving with the inverse of \mathbf{D} is not possible, as it diverges along the magic angle
- Use inverse problem formulation, apply regularization

Phase Processing

- Several processing steps are required to obtain the tissue phase

Phase Processing

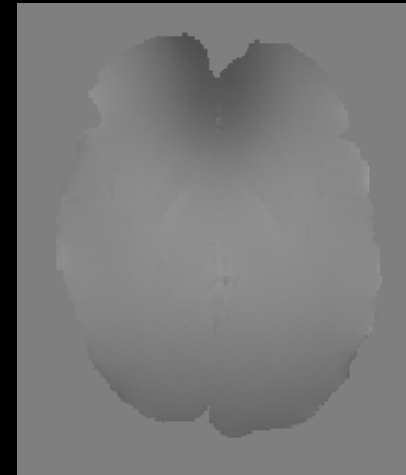
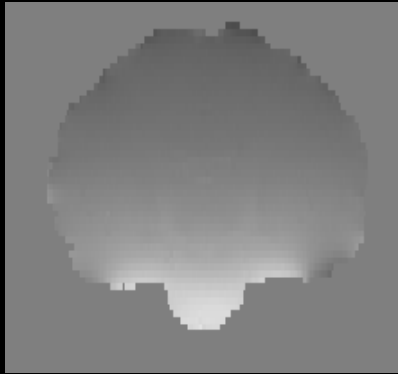
- Several processing steps are required to obtain the tissue phase
 - i. Mask out the skull



Using FSL Brain Extraction Tool¹

Phase Processing

- Several processing steps are required to obtain the tissue phase
 - i. Mask out the skull
 - ii. Unwrap the phase

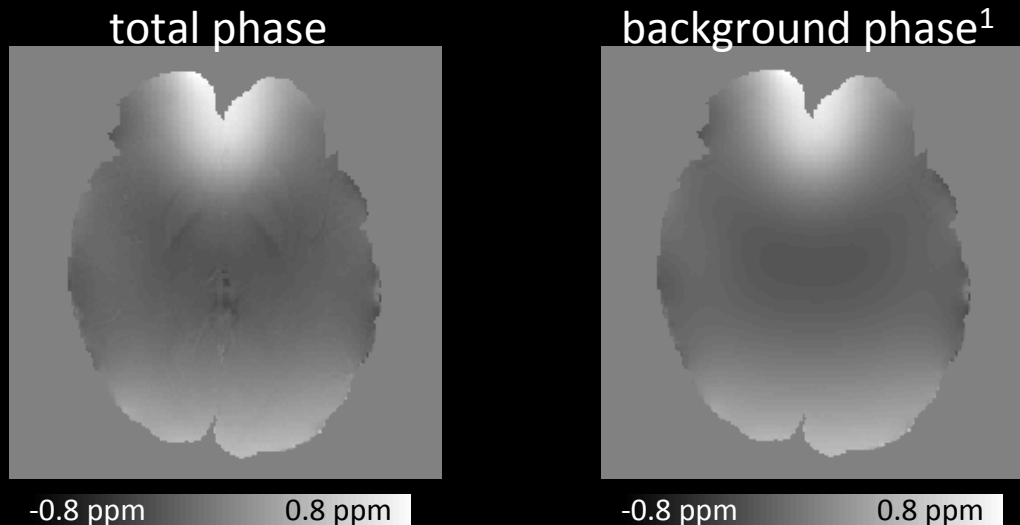


Using FSL PRELUDE¹

Phase Processing

- Several processing steps are required to obtain the tissue phase
 - i. Mask out the skull
 - ii. Unwrap the phase
 - iii. Remove background phase

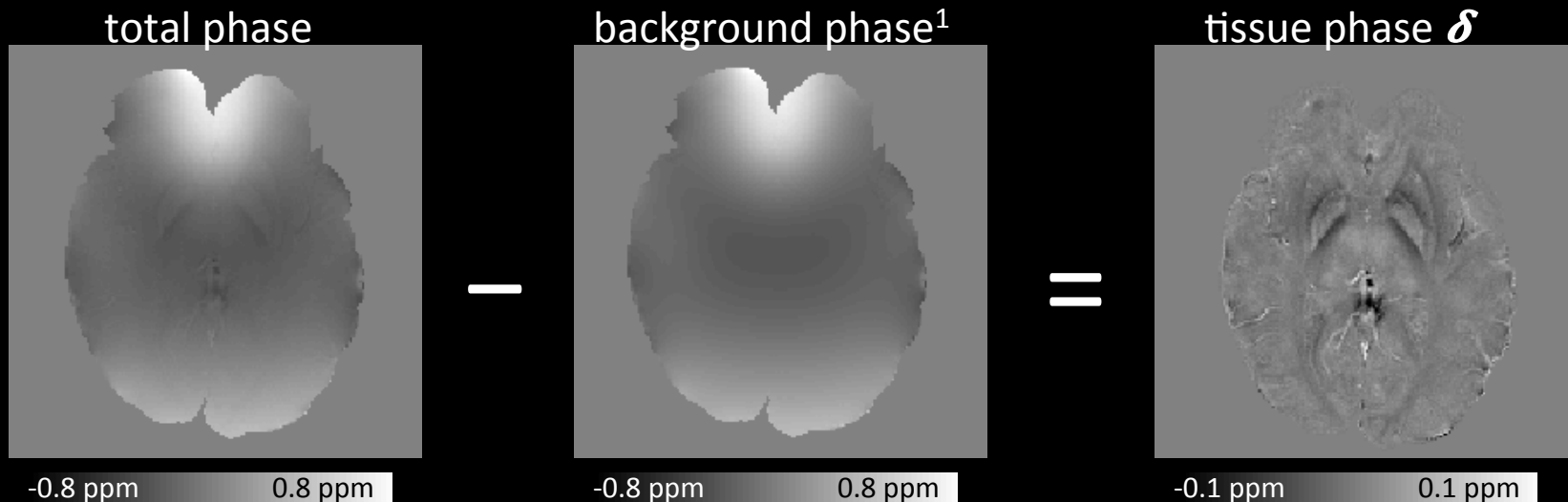
Phase accrued due to air-tissue interfaces needs to be removed
This background component is $\sim 10\times$ larger than tissue phase



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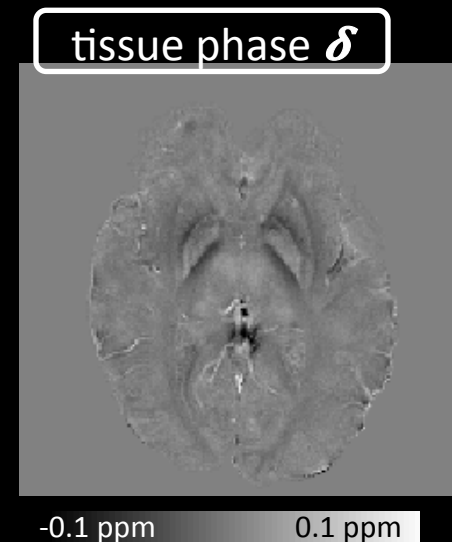
Phase Processing

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 - i. Mask out the skull
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Phase accrued due to air-tissue interfaces needs to be removed
This background component is $\sim 10\times$ larger than tissue phase

- Now we can solve for χ from tissue phase δ

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



L2 Regularized Susceptibility Inversion

- We seek the susceptibility map that matches the observed tissue phase,

$$\text{Find } \chi \text{ such that } \delta = \mathbf{F}^{-1} \mathbf{D} \mathbf{F} \chi$$

- **Prior:** Susceptibility is tied to the magnetic properties of the underlying tissue; hence it should vary smoothly within anatomical boundaries.

L2 Regularized Susceptibility Inversion

- We seek the susceptibility map that matches the observed tissue phase,

$$\text{Find } \chi \text{ such that } \delta = \mathbf{F}^{-1} \mathbf{D} \mathbf{F} \chi$$

- **Prior:** Susceptibility is tied to the magnetic properties of the underlying tissue; hence it should vary smoothly within anatomical boundaries.
- Employ regularization that encourages smoothness within tissues, but avoids smoothing across boundaries.

L2 Regularized Susceptibility Inversion

- We solve for the susceptibility distribution with a convex program,

$$\min \underbrace{\left\| \mathbf{F}^{-1} \mathbf{D} \mathbf{F} \boldsymbol{\chi} - \boldsymbol{\delta} \right\|_2^2}_{\text{Data consistency}} + \lambda \cdot \underbrace{\left\| \mathbf{M} \mathbf{G} \boldsymbol{\chi} \right\|_2^2}_{\text{Regularizer}}$$

L2 Regularized Susceptibility Inversion

- We solve for the susceptibility distribution with a convex program,

$$\min \left\| \mathbf{F}^{-1} \mathbf{D} \mathbf{F} \chi - \delta \right\|_2^2 + \lambda \cdot \left\| \mathbf{M} \mathbf{G} \chi \right\|_2^2$$

\mathbf{G} : Spatial gradient operator in 3D

\mathbf{M} : Binary mask derived from magnitude image, prevents smoothing across edges

λ : Determines the amount of smoothness

L2 Regularized Susceptibility Inversion

- We solve for the susceptibility distribution with a convex program,

$$\min \left\| \mathbf{F}^{-1} \mathbf{D} \mathbf{F} \boldsymbol{\chi} - \boldsymbol{\delta} \right\|_2^2 + \lambda \cdot \left\| \mathbf{M} \mathbf{G} \boldsymbol{\chi} \right\|_2^2$$

Optimizer given by the solution of:

$$(\mathbf{F}^{-1} \mathbf{D}^2 \mathbf{F} + \lambda \cdot \mathbf{G}^T \mathbf{M} \mathbf{G}) \boldsymbol{\chi} = \mathbf{F}^{-1} \mathbf{D}^T \mathbf{F} \boldsymbol{\delta}$$

**Large linear system, solve iteratively
with Conjugate Gradient**

L2 Regularized Susceptibility Inversion

- We solve for the susceptibility distribution with a convex program,

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Without magnitude weighting ($\mathbf{M} = \text{Identity}$), we proposed a closed-form solution:

$$\chi = \underbrace{(\mathbf{F}^{-1} \mathbf{D}^2 \mathbf{F} + \lambda \cdot \mathbf{G}^2)^{-1}}_{\text{Fast inversion with two DFTs}^1} \cdot \mathbf{F}^{-1} \mathbf{D}^T \mathbf{F} \delta$$

Fast inversion
with two DFTs¹

L2 Regularized Susceptibility Inversion

- We solve for the susceptibility distribution with a convex program,

$$\min \left\| \mathbf{F}^{-1} \mathbf{D} \mathbf{F} \chi - \delta \right\|_2^2 + \lambda \cdot \left\| \mathbf{M} \mathbf{G} \chi \right\|_2^2$$

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Without magnitude weighting ($\mathbf{M}=\text{Identity}$), we proposed a closed-form solution¹.

Using this inverse as **preconditioner** for Conjugate Gradient, we proposed a fast solution to the problem with magnitude weighting²

This improved computation speed **15-fold** relative to existing solvers

Wave-CAIPI accelerated QSM

- Wave-CAIPI enabled 3D GRE allows rapid QSM acquisition
- Compare in vivo phase and QSM from Wave-CAIPI, 2D-CAIPI and Bunch Phase Encoding:
 - At 3T and 7T
 - R = 3x3 acceleration, scan time = 2.3 min
 - 1 mm isotropic resolution
 - TE = 20 ms, TR = 40 ms
 - 240x240x120 FOV

Wave-CAIPI accelerated QSM

- Wave-CAIPI enabled 3D GRE allows rapid QSM acquisition
- Compare in vivo phase and QSM from Wave-CAIPI, 2D-CAIPI and Bunch Phase Encoding
- Phase Processing:
 - Laplacian unwrapping¹ and
 - SHARP filtering for background removal²

} **14 seconds**
- Susceptibility Inversion:
 - Fast L2-regularized inversion³

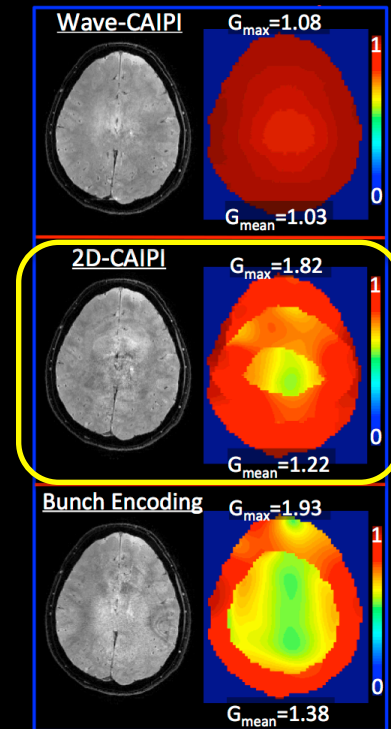
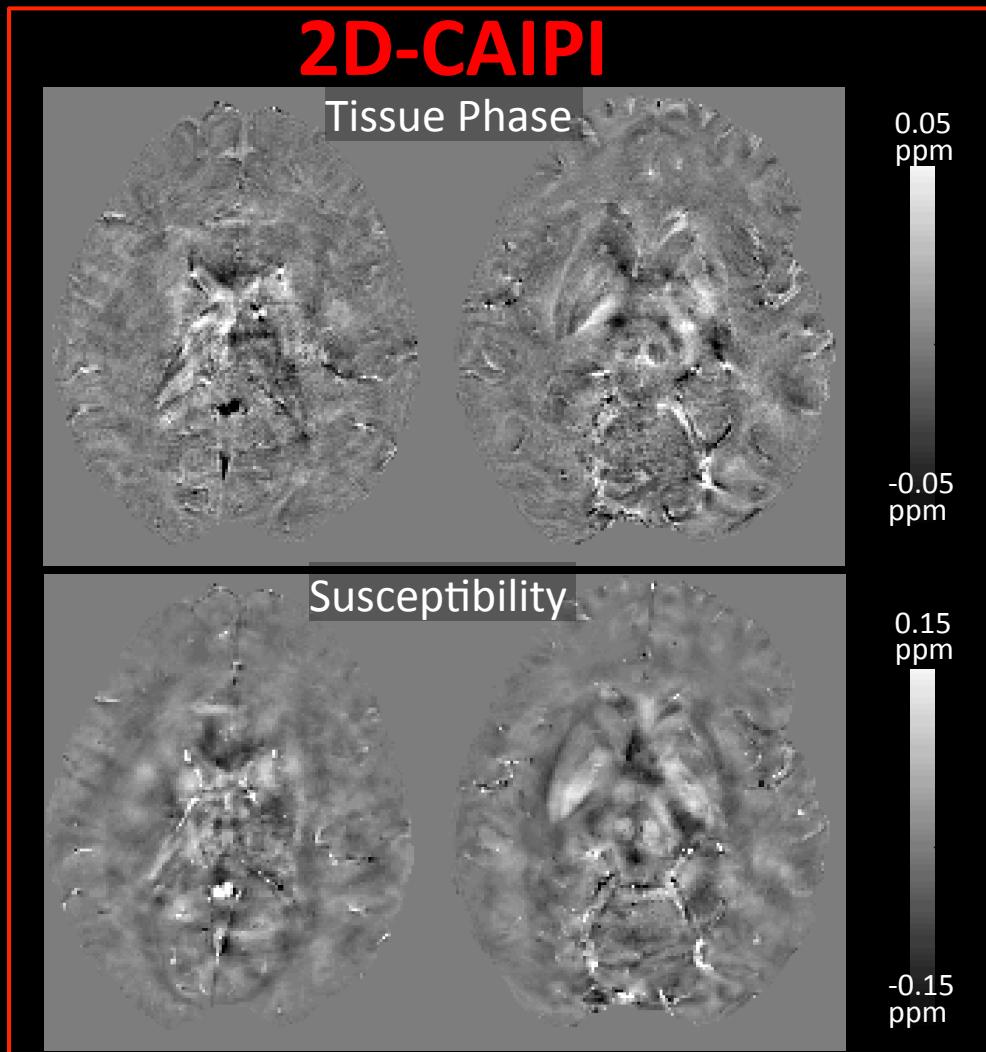
} **32 seconds**

¹ Li W, NIMG 2011

² Schweser F, NIMG 2011

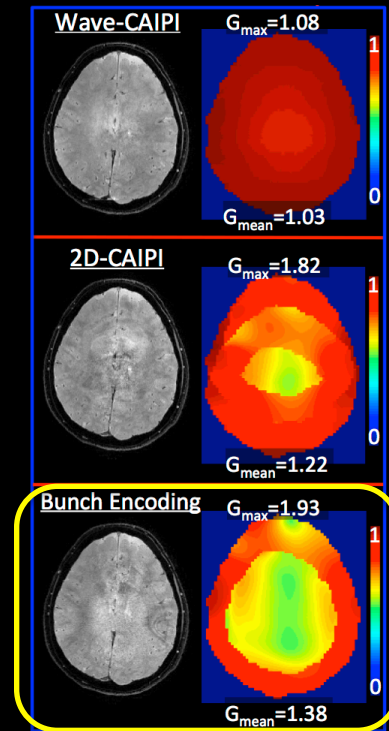
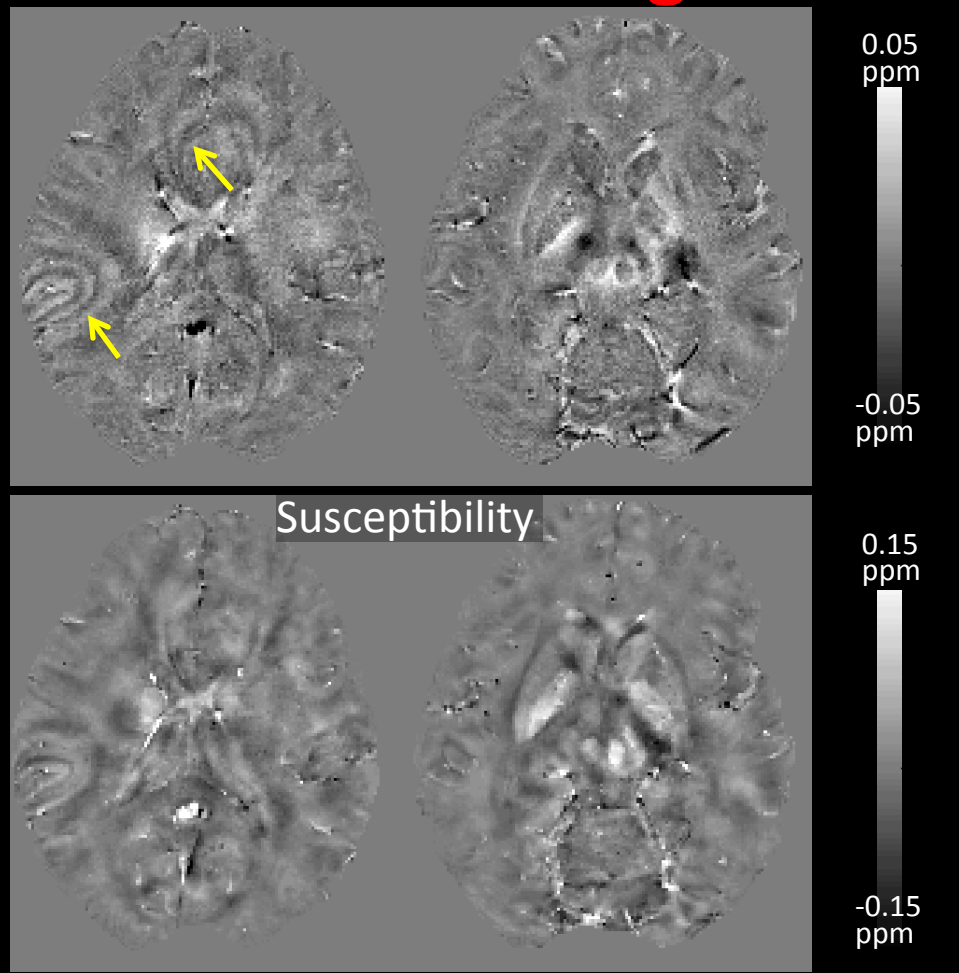
³ Bilgic B, MRM 2013

3 Tesla, R=3x3, 1 mm iso, 2.3 min acq



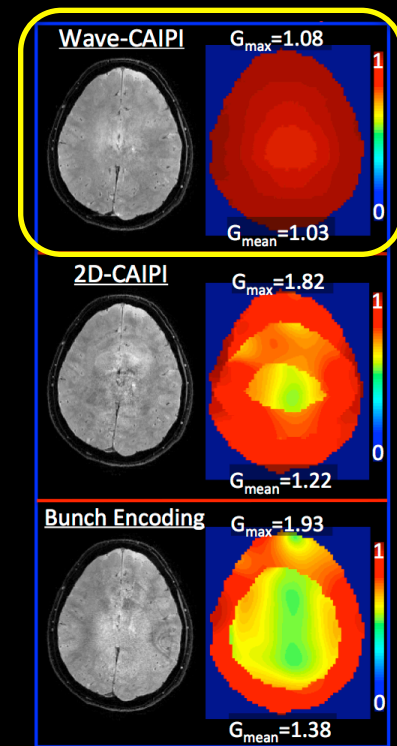
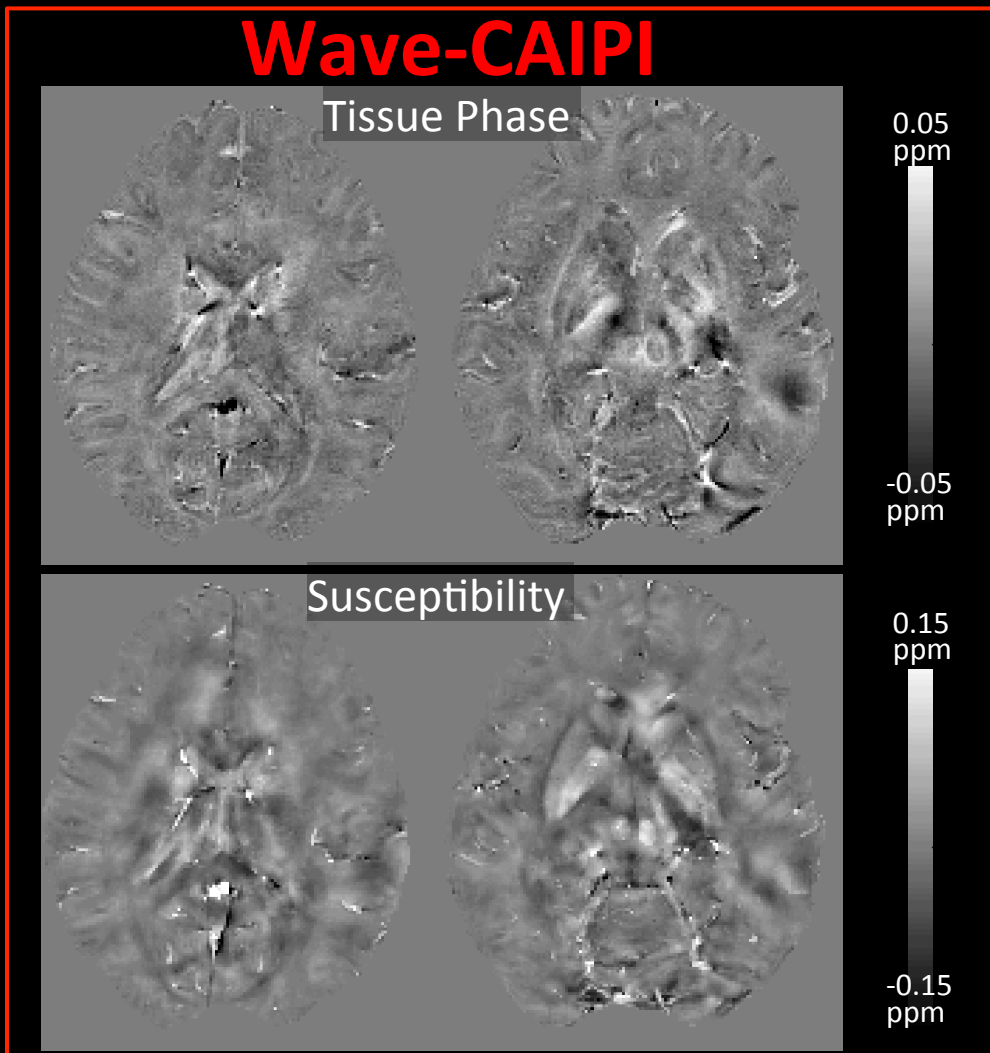
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Bunch Encoding

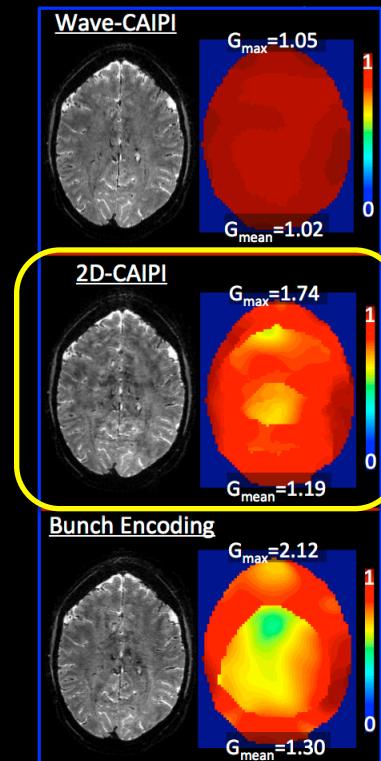
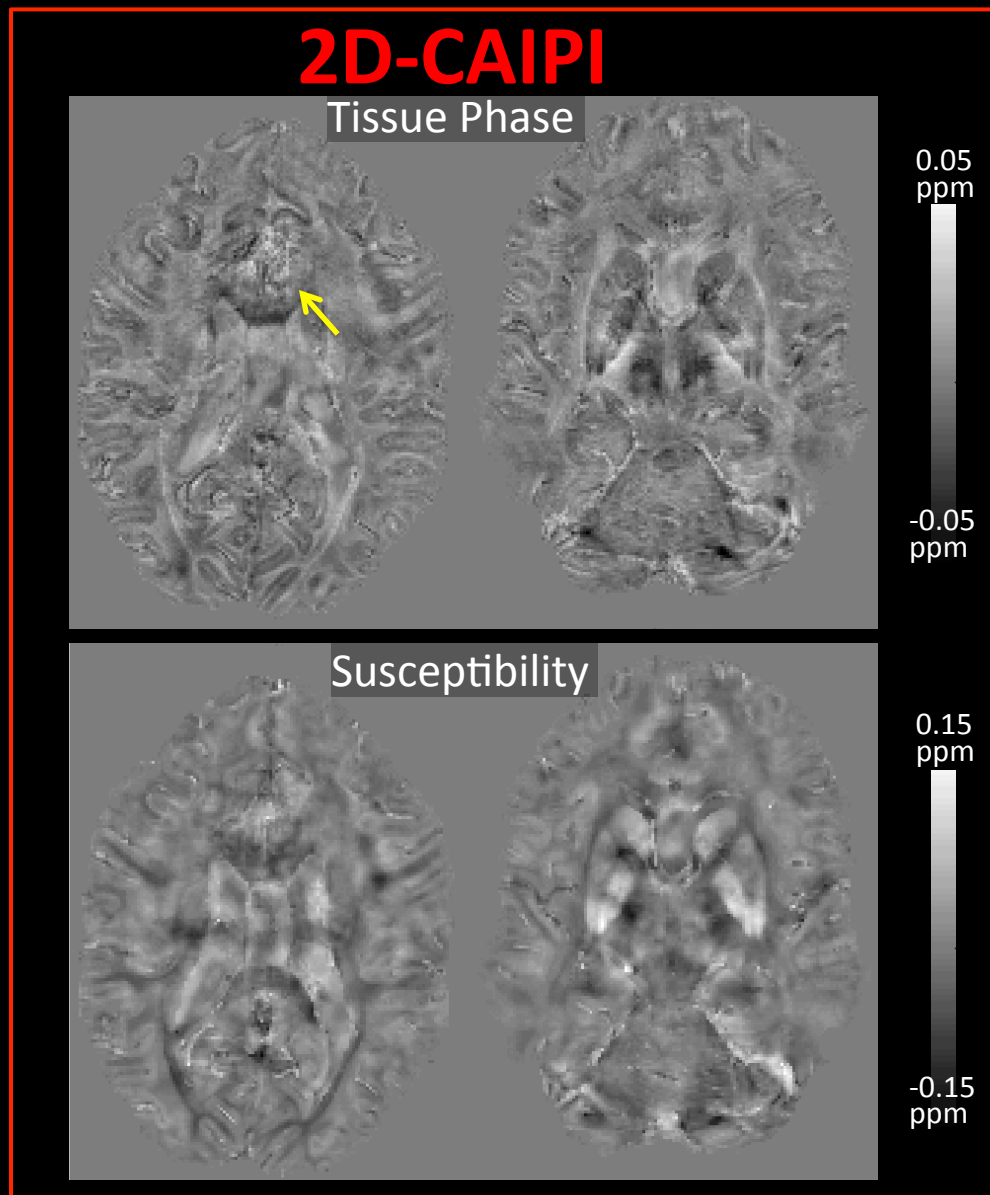


3 Tesla, R=3x3, 1 mm iso, 2.3 min acq

Wave-CAIPI

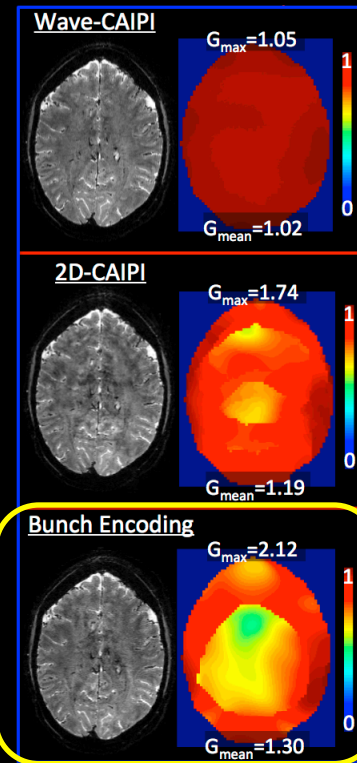
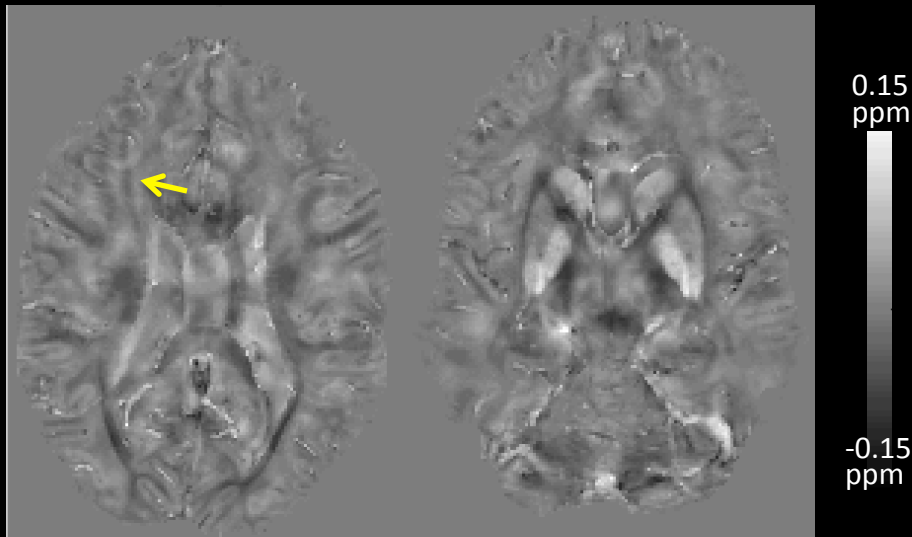
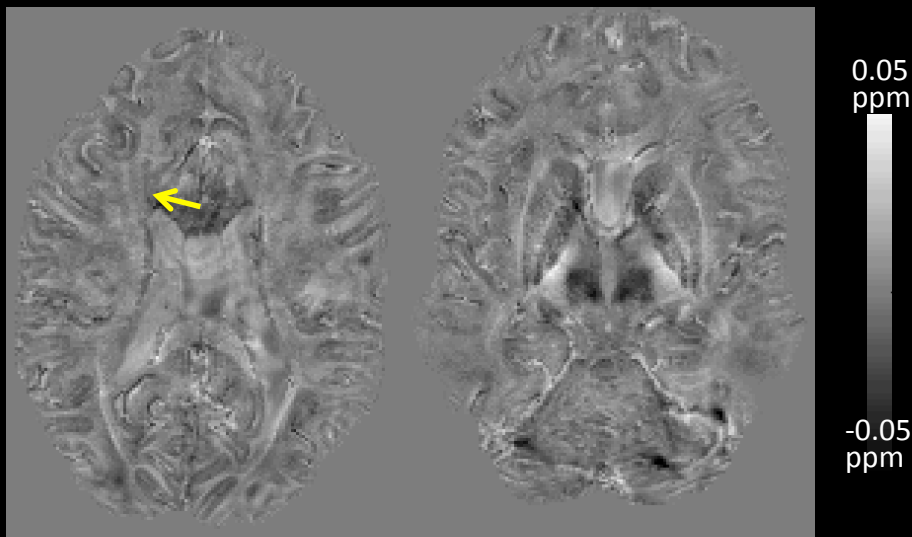


7 Tesla, R=3x3, 1 mm iso, 2.3 min acq



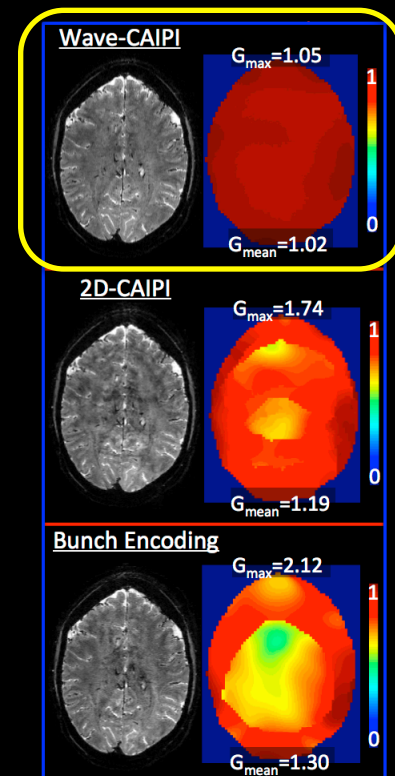
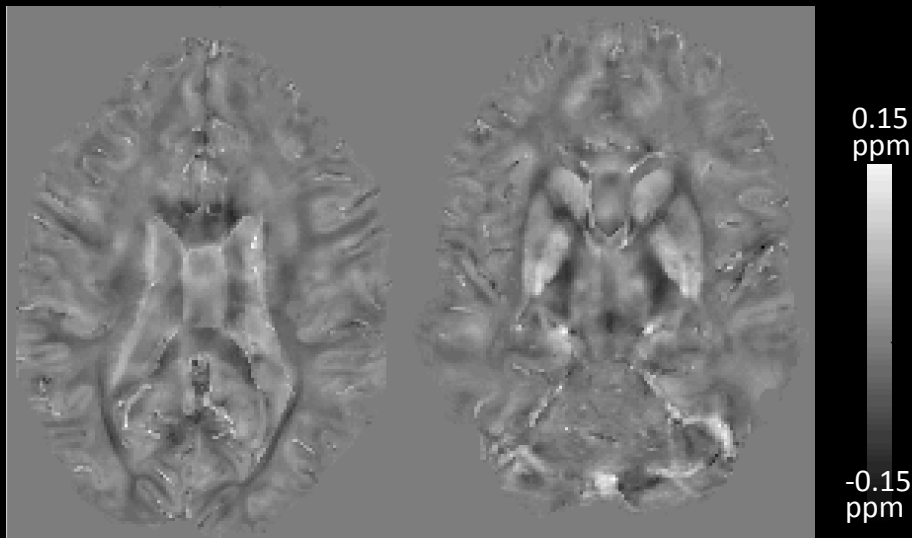
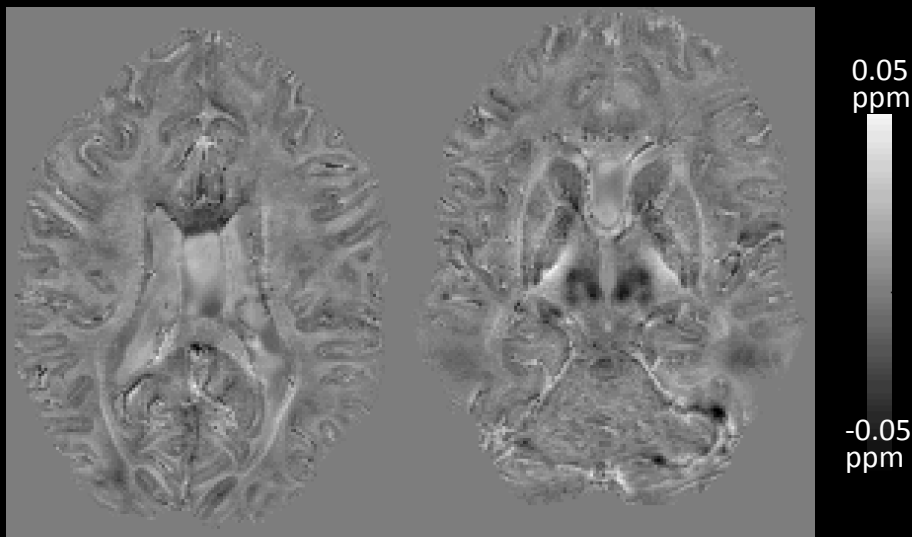
7 Tesla, R=3x3, 1 mm iso, 2.3 min acq

Bunch Phase



7 Tesla, R=3x3, 1 mm iso, 2.3 min acq

Wave-CAIPI



Summary

- Propose Wave-CAIPI acquisition/reconstruction scheme for highly accelerated 3D imaging
- Wave-CAIPI offers 2-fold improvement in g-factor and image artifact penalties compared to 2D-CAIPI and Bunch Phase Encoding

Summary

- Propose Wave-CAIPI acquisition/reconstruction scheme for highly accelerated 3D imaging
- Wave-CAIPI offers 2-fold improvement in g-factor and image artifact penalties compared to 2D-CAIPI and Bunch Phase Encoding
- Deployed in GRE imaging, Wave-CAIPI allows 9-fold acceleration with \sim perfect SNR retention at 3T and 7T
- Combined with fast phase and susceptibility processing methods, it enables QSM at 1 mm resolution in 2.3 min

Thank you for your attention