



Calibrationless Parallel Imaging in Multi Echo/Contrast Data

B Bilgic^{1,2}, B Zhao^{1,2}, I Chatnuntawech³, LL Wald^{1,2}, K Setsompop^{1,2}

1 Martinos Center, Charlestown, MA, USA

2 Harvard Medical School, Boston, MA, USA

3 National Nanotechnology Center, Pathum Thani, Thailand

— INTERNATIONAL SOCIETY FOR —
ISMRRM
MAGNETIC RESONANCE IN MEDICINE

 **ONE**
COMMUNITY
FOR CLINICIANS
AND SCIENTISTS

 **25TH Annual Meeting**

& Exhibition • 22–27 April 2017
SMRT 26th Annual Meeting • 22–24 April 2017

HONOLULU, HI, USA

www.ismrm.org • www.smrt.org



Declaration of Financial Interests or Relationships

Speaker Name: Kawin Setsompop

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

- Licensing agreements with Siemens, GE, Phillips and Samsung
- Research support from Siemens

Calibrationless Parallel Imaging

- Conventional parallel imaging requires auto-calibration signal (ACS) to estimate sensitivities or k-space kernels, which increases scan time
- Inconsistency due to motion or physiological noise in ACS may lead to artifacts [1]
- Dynamic, spectroscopic and non-Cartesian imaging, acquiring ACS can be challenging or not possible at all
- Emerging techniques use low-rank modeling [2-5] or jointly estimate sensitivities and images [6-8] to reduce/eliminate dependence on calibration.

[1] JR Polimeni, MRM'16

[3] PJ Shin, MRM'14

[5] KH Jin, IEEE TCI'16

[7] F Knoll, MRM'12

[2] JD Trzasko, ASILOMAR'11

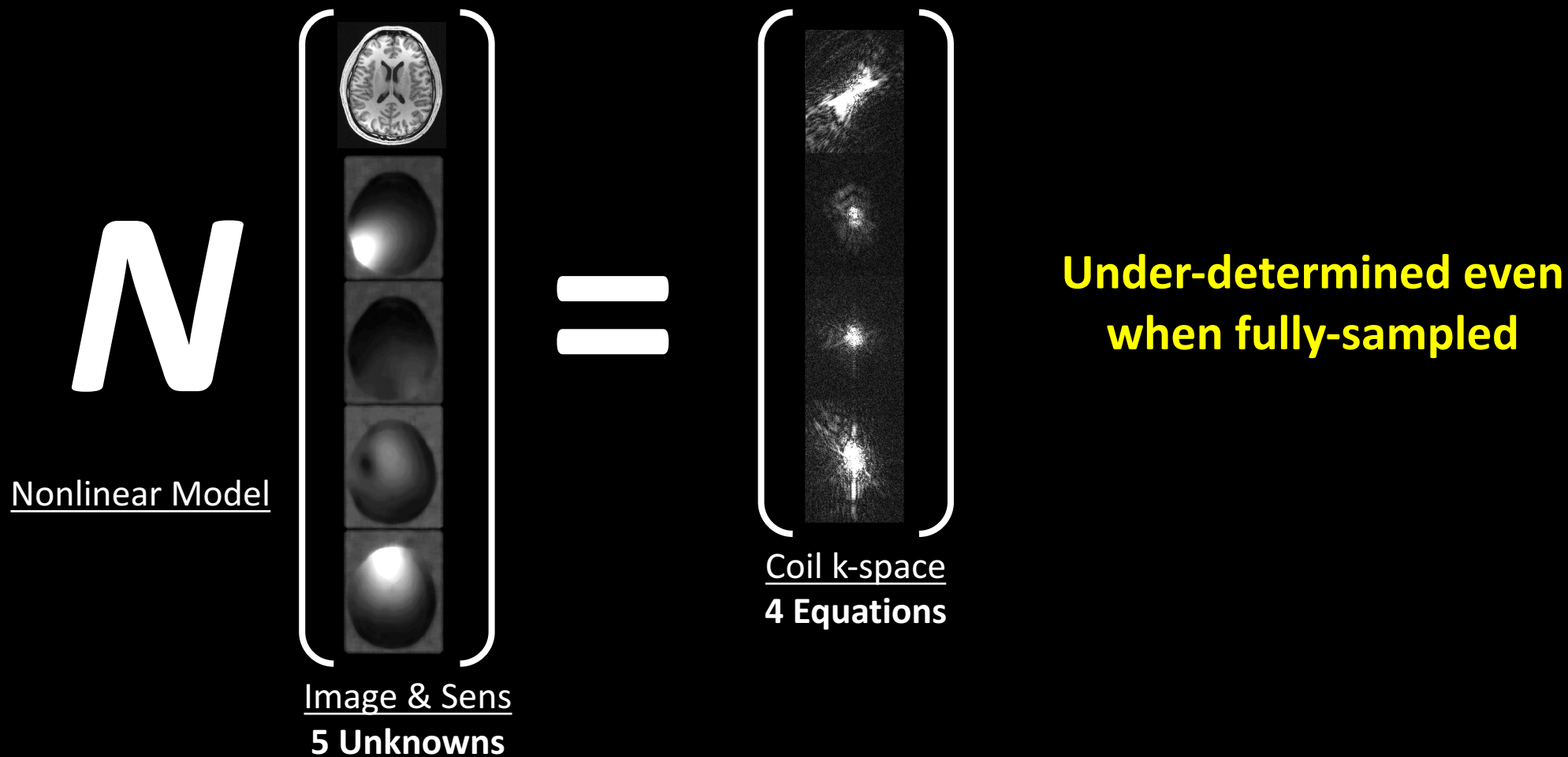
[4] JP Haldar, MRM'16

[6] M Uecker, MRM'08

[8] L Ying, MRM'07

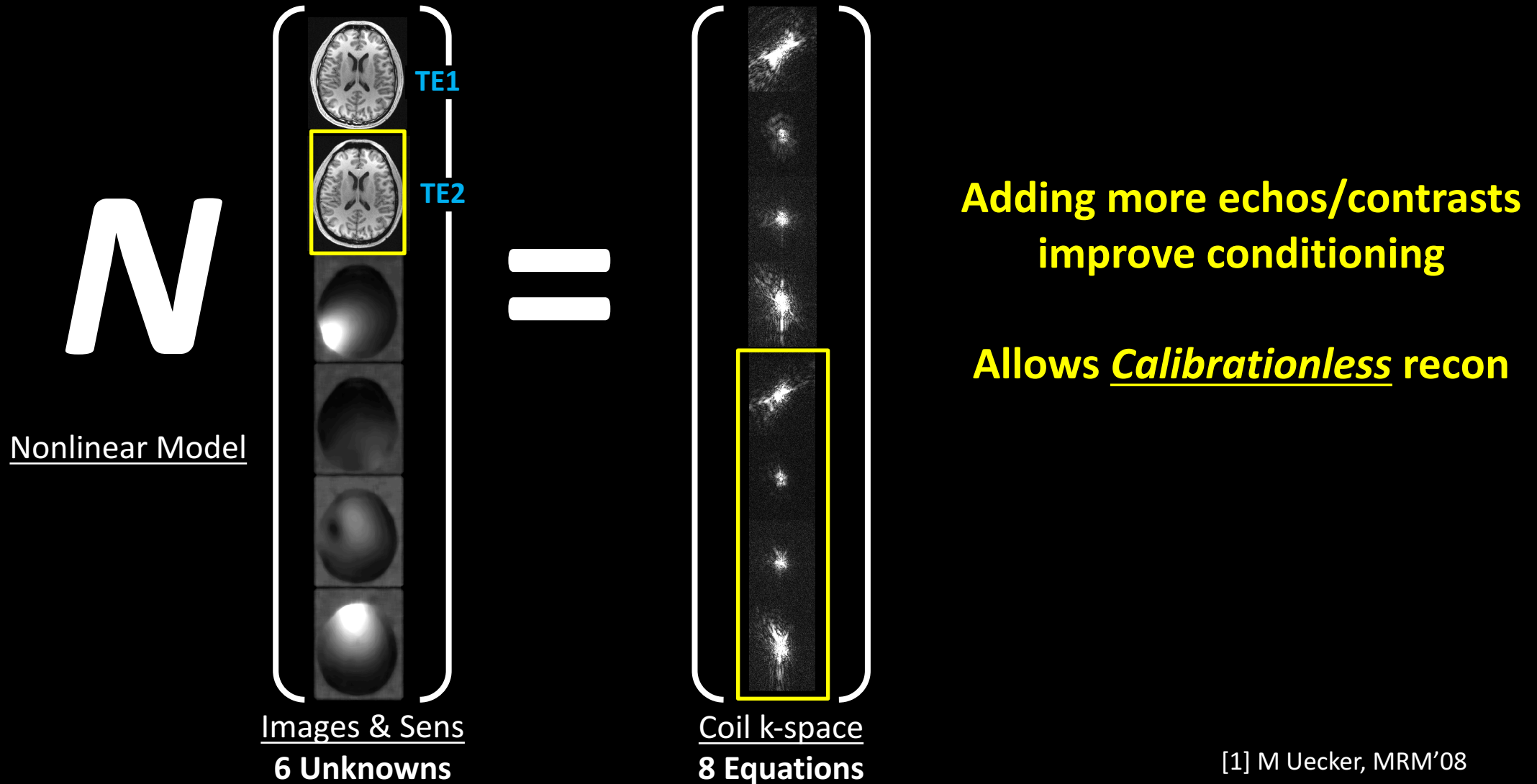
Nonlinear Inversion

- We jointly recon sensitivities & image content by extending Nonlinear INVersion (NLINV) [1,2]



Joint Nonlinear Inversion

- We jointly recon sensitivities & image content by extending Nonlinear INVersion (NLINV) [1,2]

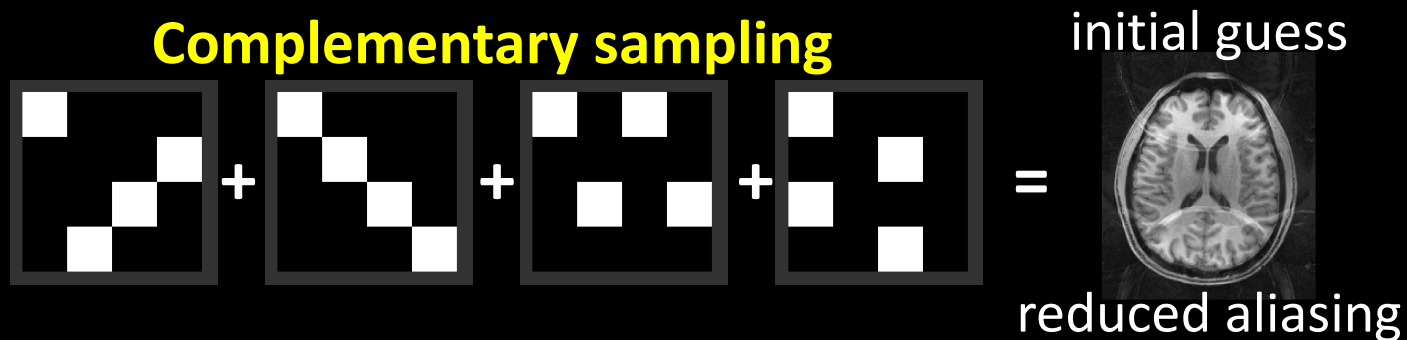
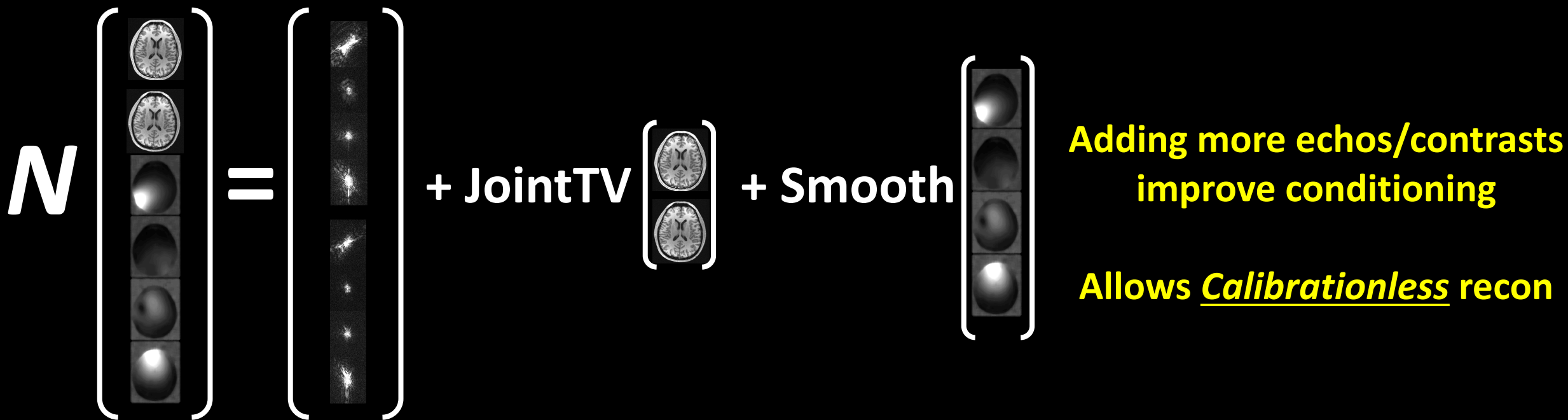


[1] M Uecker, MRM'08

[2] F Knoll, MRM'12

Joint Nonlinear Inversion

- We jointly recon sensitivities & image content by extending Nonlinear INVersion (NLINV) [1,2]



[1] M Uecker, MRM'08

[2] F Knoll, MRM'12

Calibrationless Joint NLINV

- Calibrationless recon is made possible by:
 - i. leveraging shared sensitivities across echos/contrasts to improve conditioning
 - ii. employing Joint TV to exploit shared edge structures and sparsity
 - iii. complementary sampling that improves collective k-space coverage and provides an image with *reduced aliasing* for initialization

Theory: Joint NLINV

■ Solve for images $\rho_1 \dots \rho_M$

and sensitivities $C_1 \dots C_K$

$$F_i \left(C_j \cdot \rho_i \right) = y_{ij}$$

undersampled DFT for image i sens j image i k-space image i coil j

Theory: Joint NLINV

■ Solve for images $\rho_1 \dots \rho_M$

and sensitivities $c_1 \dots c_K$

$$N(x) = y$$

nonlinear operator

all unknowns

$$\begin{bmatrix} \rho_1 \\ \vdots \\ \rho_M \\ c_1 \\ \vdots \\ c_K \end{bmatrix}$$

all k-space

$$\begin{bmatrix} y_{11} \\ \vdots \\ y_{MK} \end{bmatrix}$$

Theory: Joint NLINV

- Linearize the nonlinear operation around an initial guess x^n

$$N(x^n + dx) \approx N(x^n) + DN(x^n)dx = y$$

constant Jacobian

- Solve for the update $dx = \begin{bmatrix} d\rho_1 \\ \vdots \\ d\rho_M \\ dc_1 \\ \vdots \\ dc_K \end{bmatrix}$

Theory: Joint NLINV

- We constrain the sensitivities to be smooth via Sobolev norm [1]

$$\min_{dx} \underbrace{\|N(x^n) + DN(x^n)dx - y\|_2^2}_{\text{data consistency}}$$

$$W = (1 + k^2)^8 \cdot F \quad + \alpha^n \underbrace{\|W(c^n + dc)\|_2^2}_{\text{Sobolev norm}}$$

Penalize high freq in k-space

Sobolev norm

Theory: Joint NLINV

- We enforce joint sparsity in edge structures with ℓ_{21} penalty

$$\begin{aligned} \min_{dx} \quad & \|N(x^n) + DN(x^n)dx - y\|_2^2 \\ & + \alpha^n \|W(c^n + dc)\|_2^2 \\ & + \beta^n \underbrace{\|\nabla(\rho^n + d\rho)\|_{21}}_{\text{Joint TV}} \end{aligned}$$

Theory: Joint NLINV

- Each Gauss-Newton step is minimized using Nonlinear Conjugate Gradient [1]

$$\begin{aligned} \min_{dx} & \|N(x^n) + DN(x^n)dx - y\|_2^2 \\ & + \alpha^n \|W(c^n + dc)\|_2^2 \\ & + \beta^n \|\nabla(\rho^n + d\rho)\|_{21} \end{aligned}$$

- Reduce α^n and β^n 2-fold in each step to avoid over-smoothing

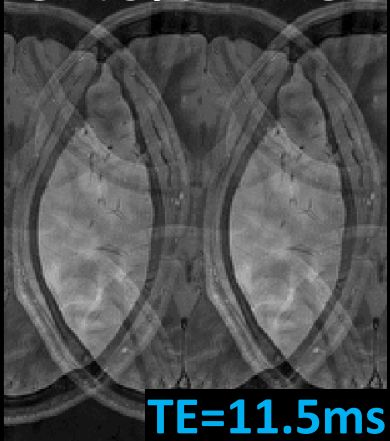
Calibrationless Joint NLINV

- We demonstrate calibrationless recon for:
 - i. Multi-echo spin-echo
 - ii. Multi-echo MPRAGE [1]
 - iii. Phase-cycled balanced SSFP
 - iv. Multi-contrast protocol

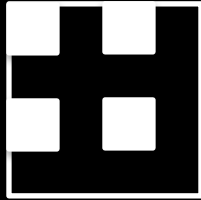
Multi-Echo Spin-Echo, Calibrationless, Acceleration R=2x2

NLINV

37.6% RMSE

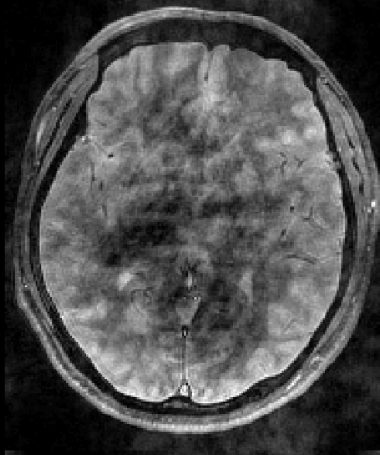


Uniform
sampling

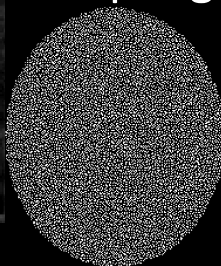


SAKE

28.5% RMSE



Poisson
sampling

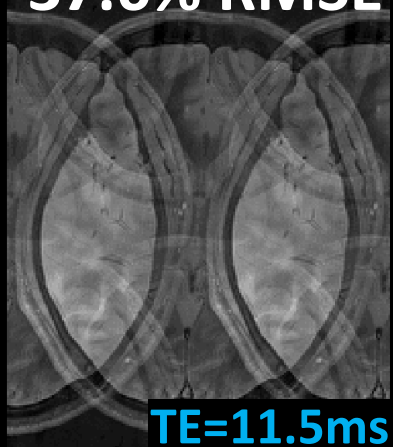


- ❖ 12-chan reception
- ❖ $\Delta TE = 11.5\text{ms}$, 8 echoes,
- ❖ FOV = $240 \times 180\text{mm}^2$, slice thickness = 3mm
- ❖ Mtx = 256×208

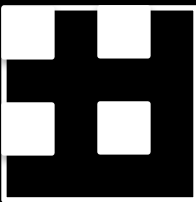
Multi-Echo Spin-Echo, Calibrationless, Acceleration R=2x2

NLINV
37.6% RMSE

Uniform sampling

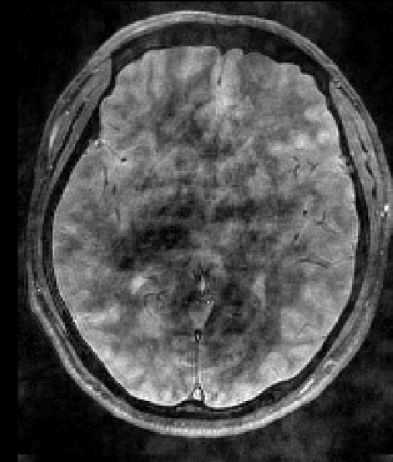


TE=11.5ms

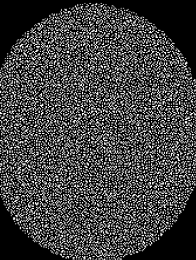


SAKE
28.5% RMSE

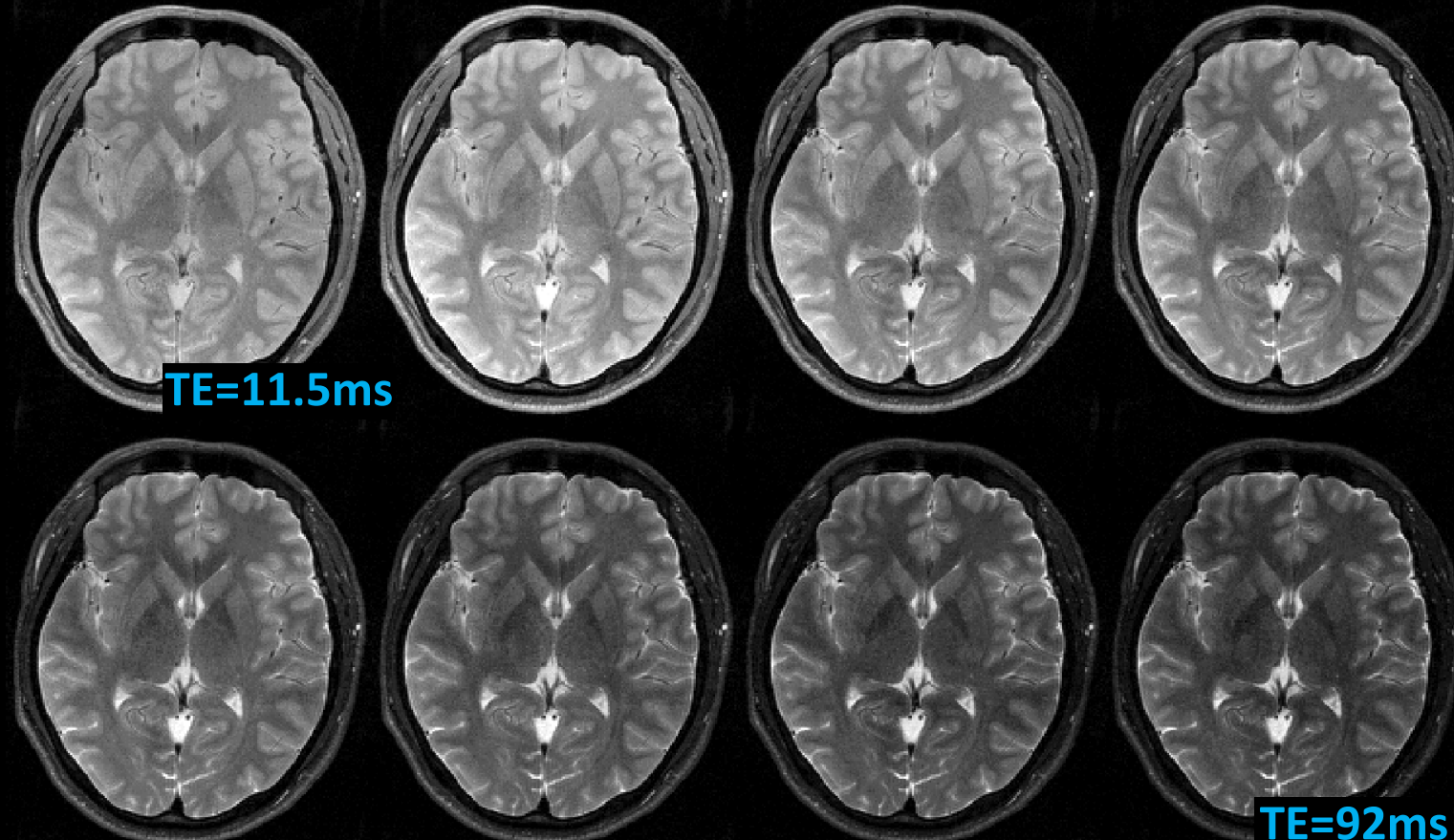
Poisson sampling



TE=11.5ms



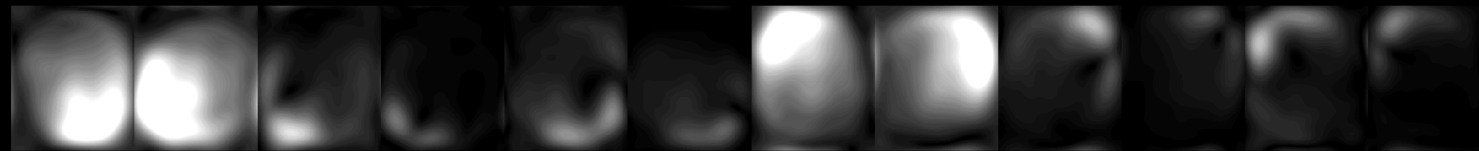
Proposed: Joint NLINV 6.0% RMSE



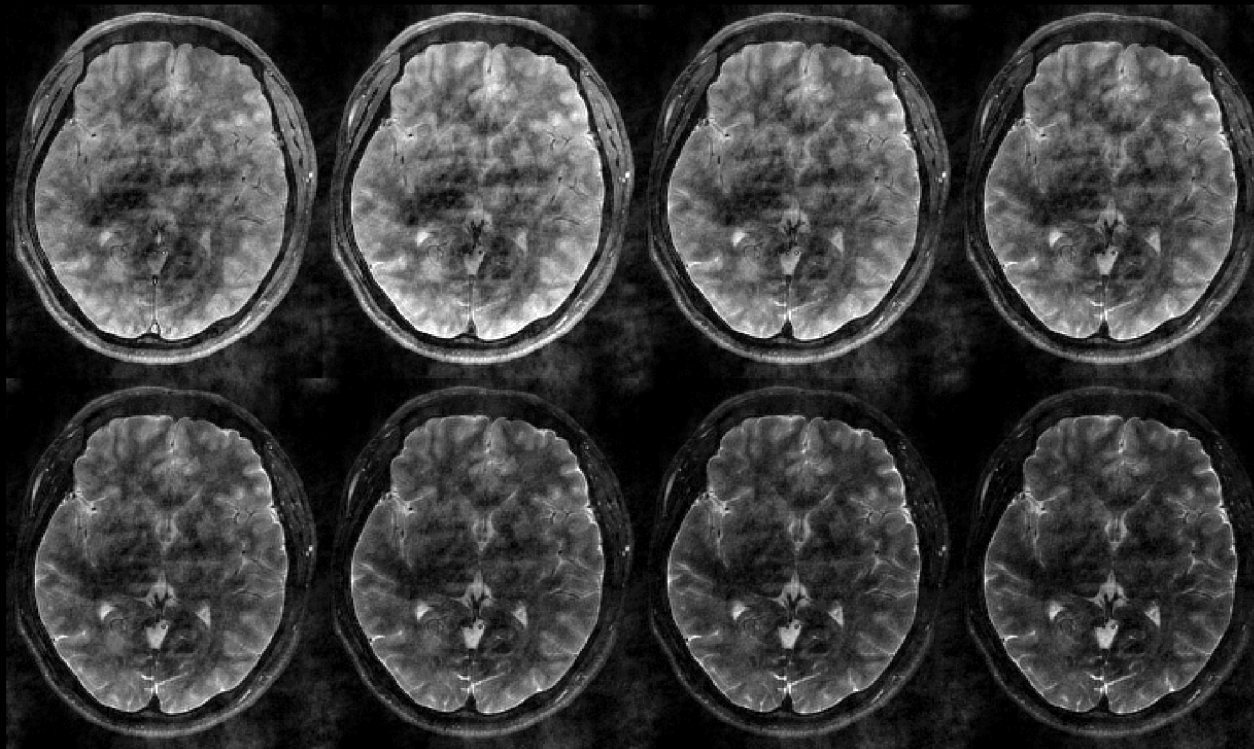
TE=11.5ms

TE=92ms

estimated 12 chan sensitivities

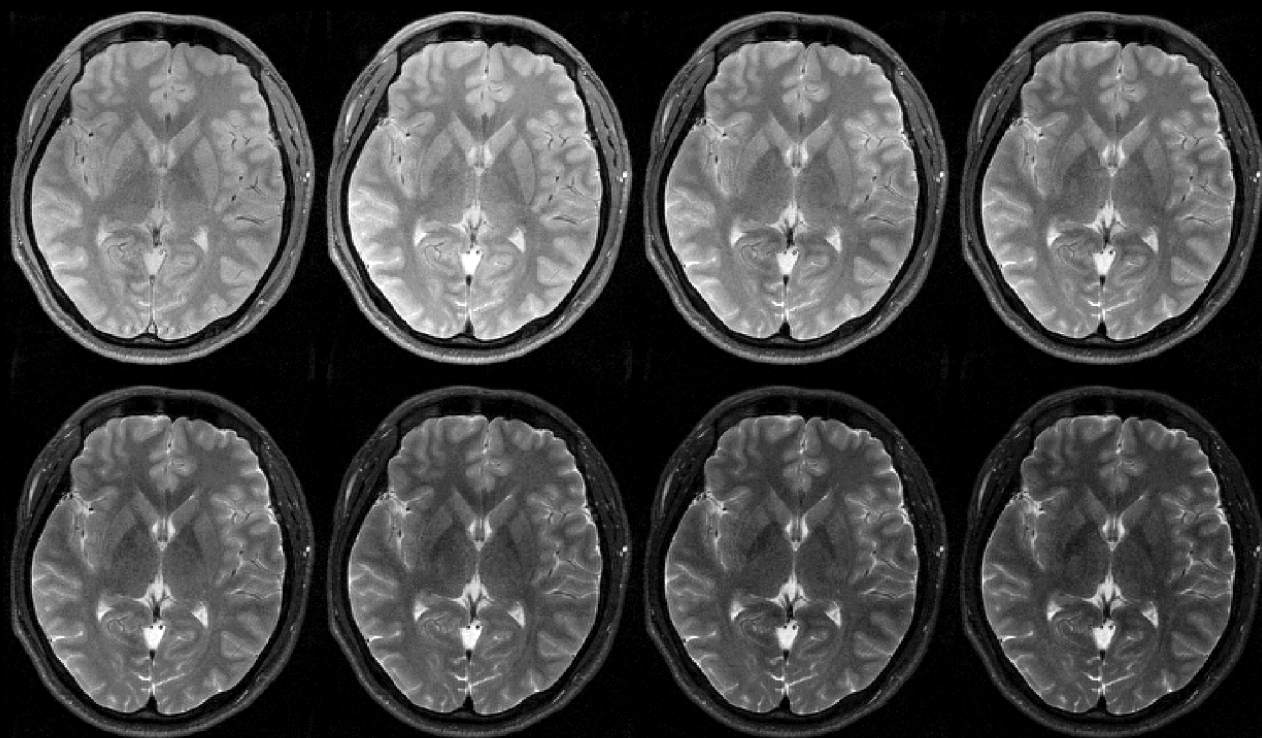


SAKE 28.5% RMSE

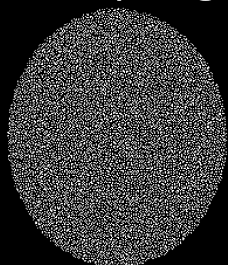


Proposed: Joint NLINV

6.0% RMSE



Poisson
sampling



Recon time: 73 min

100 iter

Window= 6x6

Threshold = 1.4

Uniform sampling



Recon time: 21 min

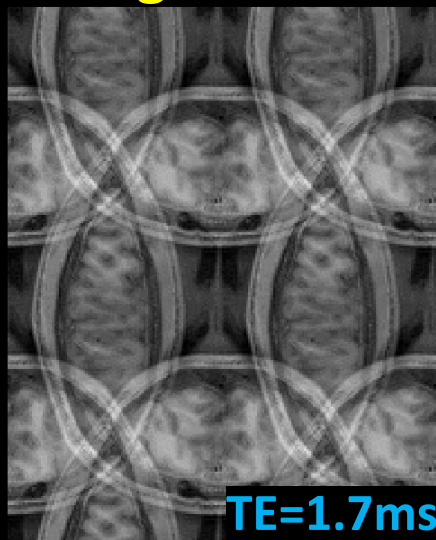
9 Newton iter

$\alpha_0 = 0.001$

$\beta_0 = 0.1$

Multi-Echo MPRAGE, Calibrationless, Acceleration R=2x2

**NLINV:
single echo**



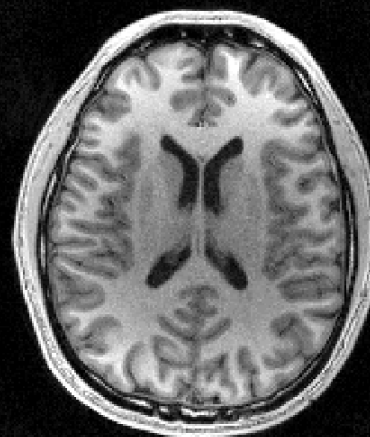
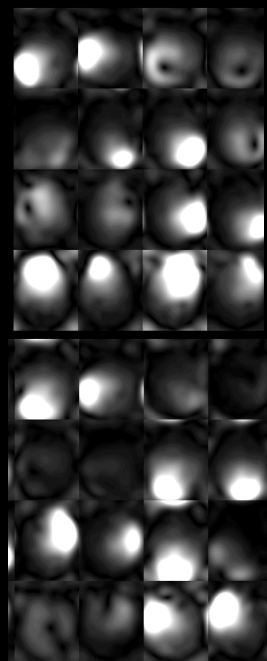
TE=1.7ms

67.4% RMSE

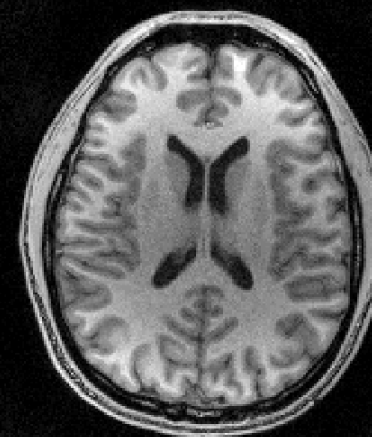
Proposed: Joint NLINV

4.9% RMSE

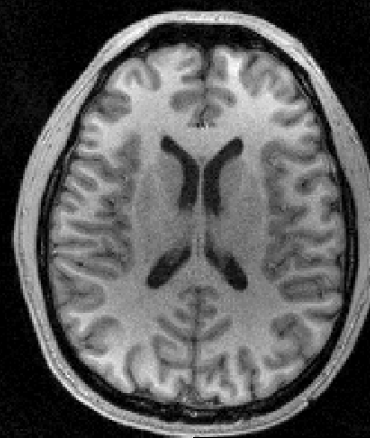
32 chan sens



TE=1.7ms



TE=3.6ms



TE=5.4ms

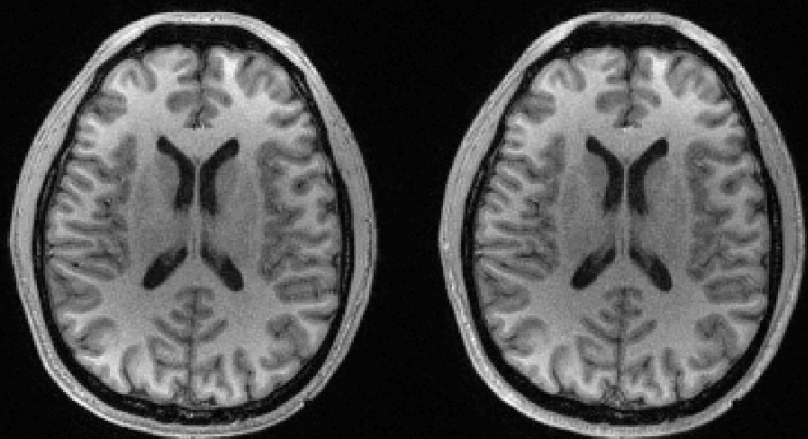
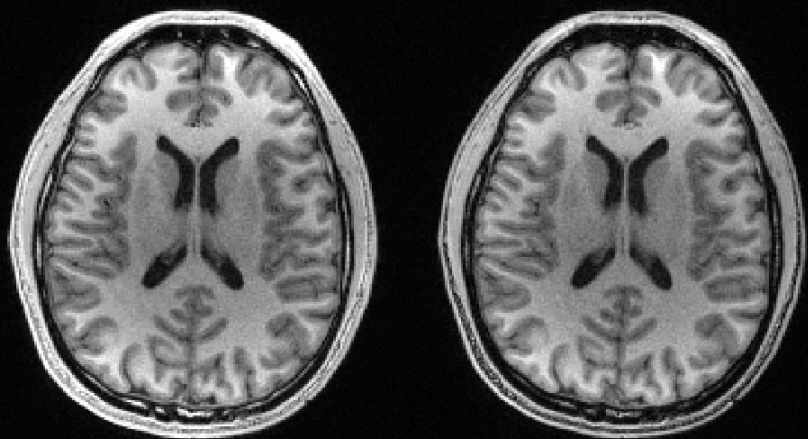


TE=7.3ms

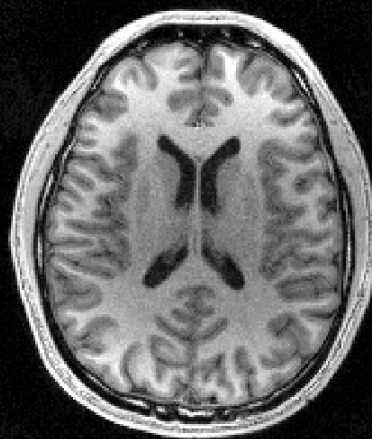
- ❖ 32-chan reception
- ❖ Mtx = 256x240x192

1 mm isotropic
BW = 651Hz/pixel

SAKE 6.4% RMSE



Proposed: Joint NLINV



4.9% RMSE



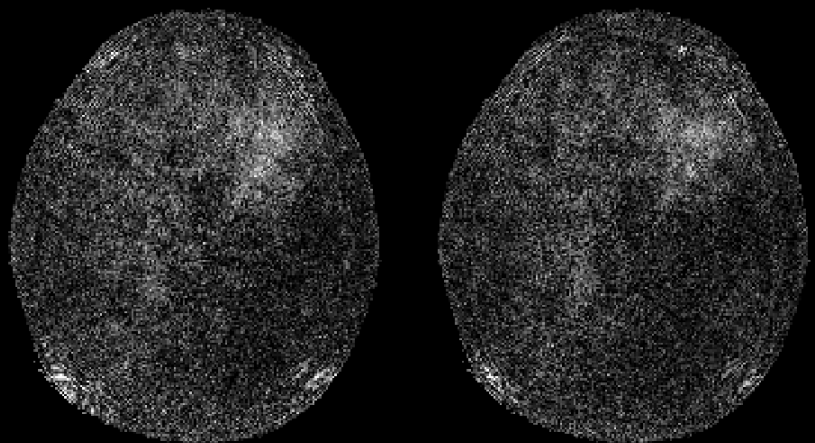
Recon time: 122 min
Window= 6×6

Poisson sampling
Threshold = 1.5

Recon time: 25 min
 $\alpha_0 = 0.001$

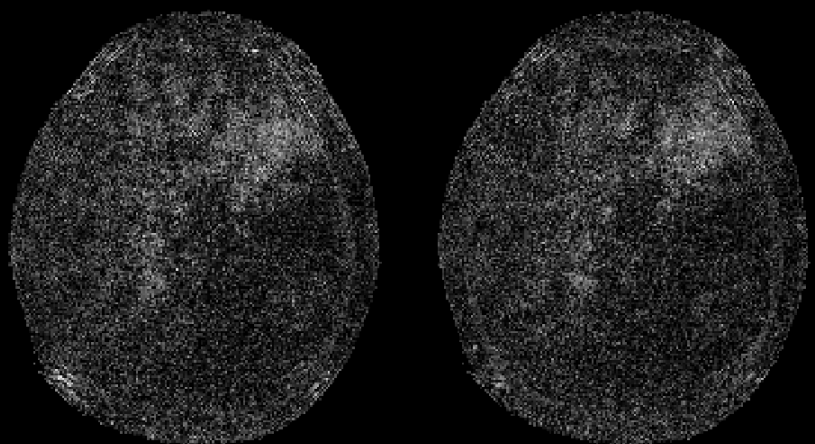
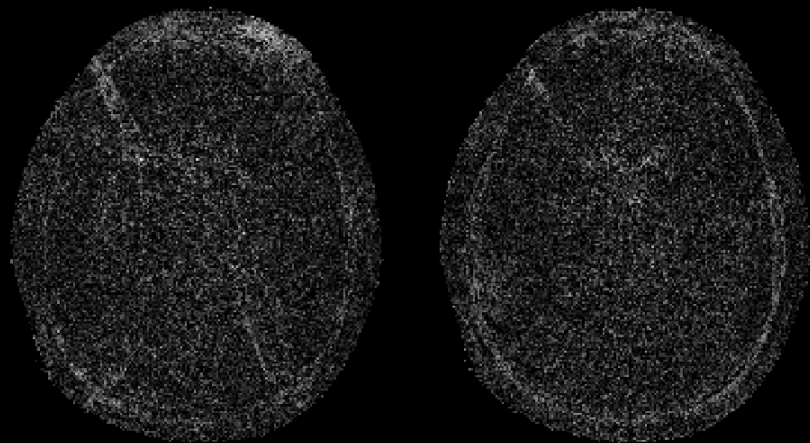
Uniform sampling
 $\beta_0 = 0.1$

SAKE 6.4% RMSE

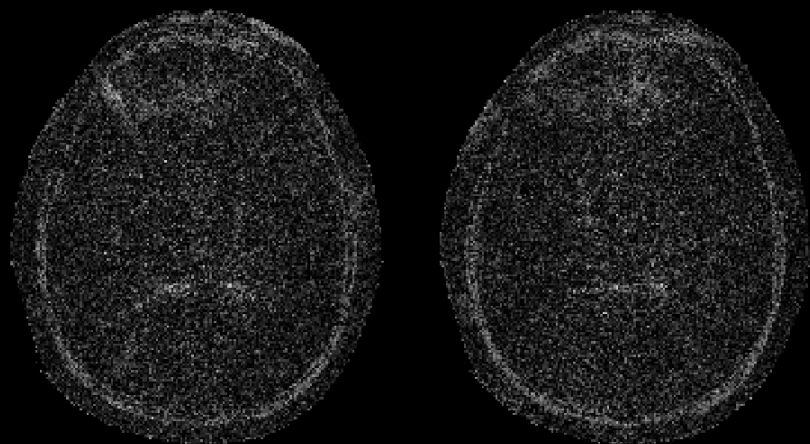


Proposed: Joint NLINV

4.9% RMSE



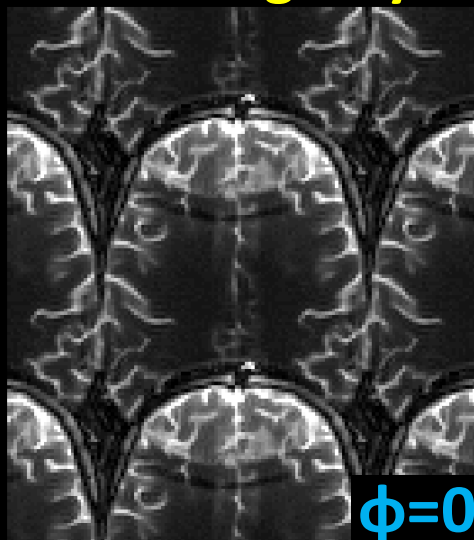
Error scaled 5x



Error scaled 5x

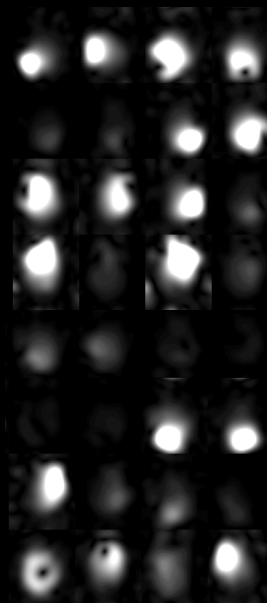
Phase-Cycled bSSFP, Calibrationless, Acceleration R=2x2

NLINV: single cycle



75.1% RMSE

32 chan sens



Proposed: Joint NLINV

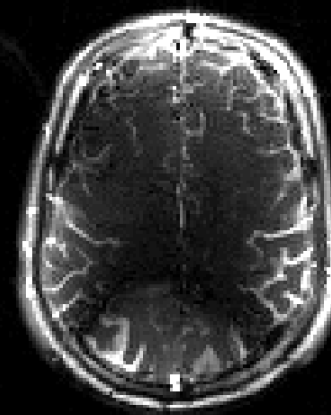
3.3% RMSE



phase-cycle $\phi=0$



$\phi=\pi/2$



$\phi=\pi$



$\phi=3\pi/2$

- ❖ 32-chan reception
- ❖ Mtx = 160x160
- ❖ FOV = 240x240

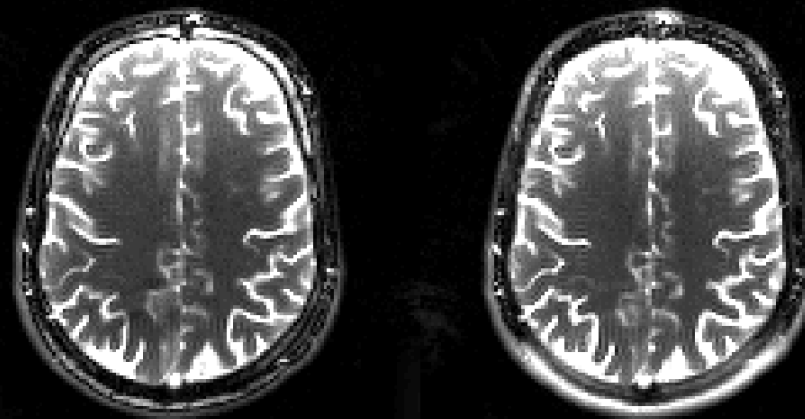
TR/TE = 3.37/1.57 ms
slice thickness = 4.5 mm
BW = 651Hz/pixel

Phase-Cycled bSSFP, Calibrationless, Acceleration R=2×2

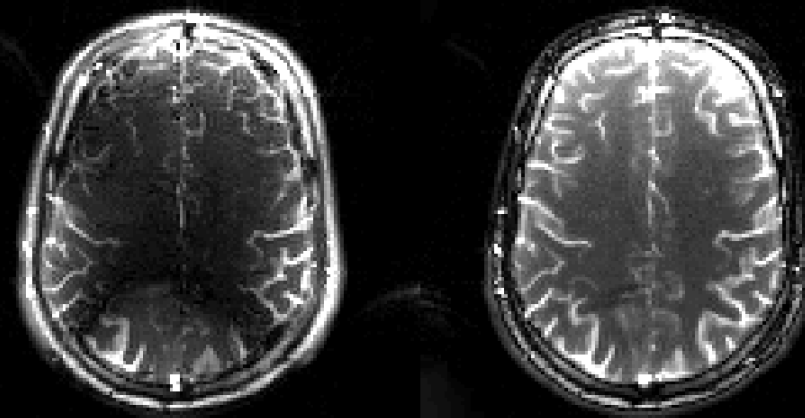
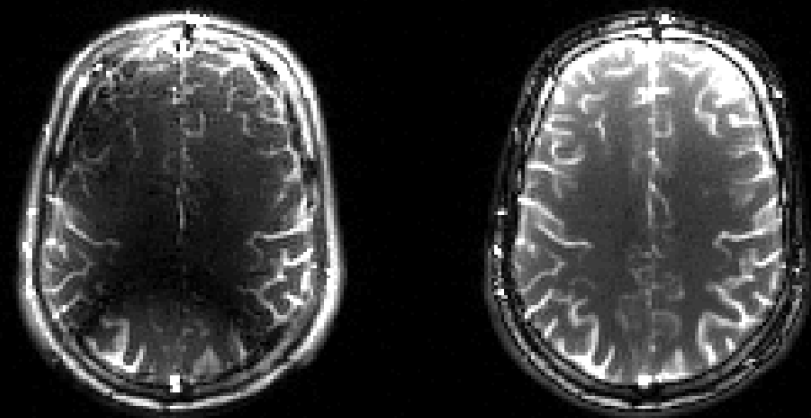
SAKE 5.2% RMSE



Proposed: Joint NLINV



3.3% RMSE



Recon time: 82 min
Window= 6×6

Poisson sampling
Threshold = 1.3

Recon time: 13 min
 $\alpha_0 = 0.01$

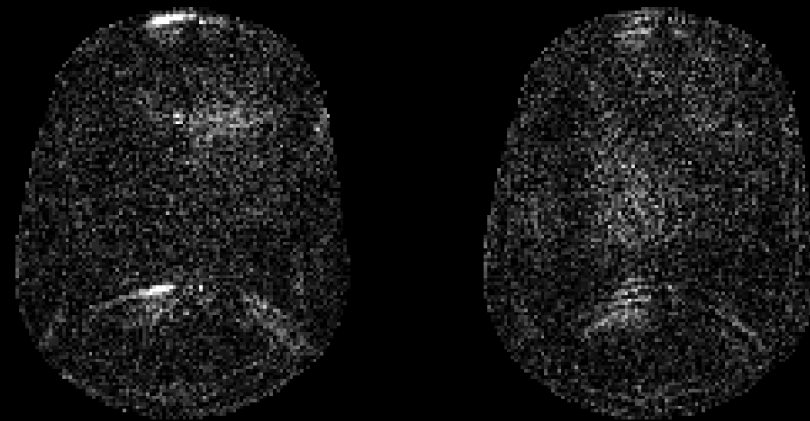
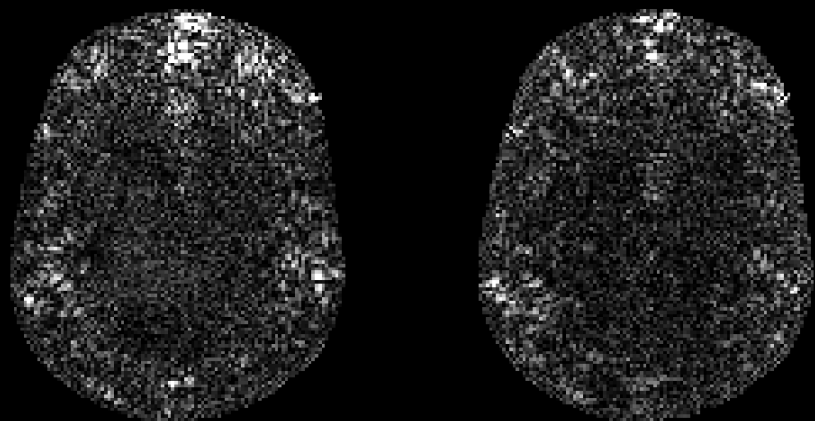
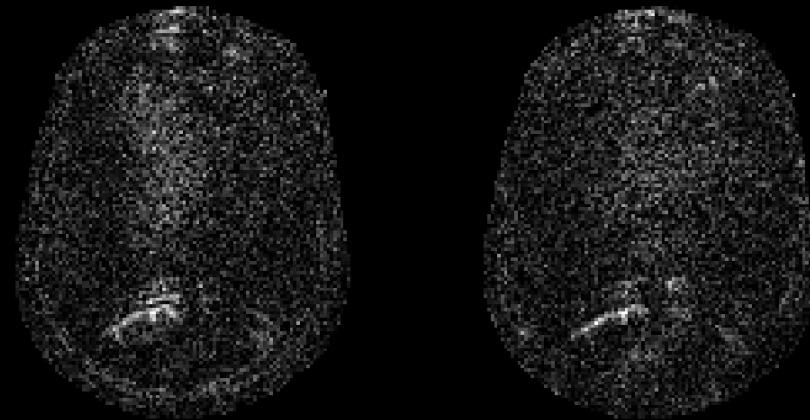
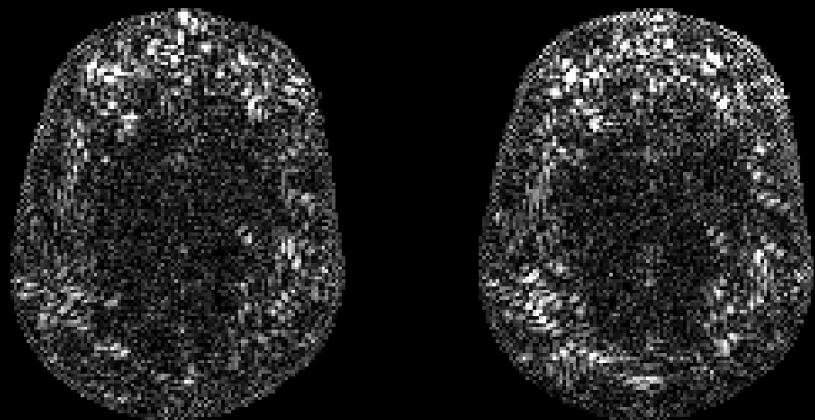
Uniform sampling
 $\beta_0 = 0.1$

Phase-Cycled bSSFP, Calibrationless, Acceleration $R=2\times 2$

SAKE 5.2% RMSE

Proposed: Joint NLINV

3.3% RMSE

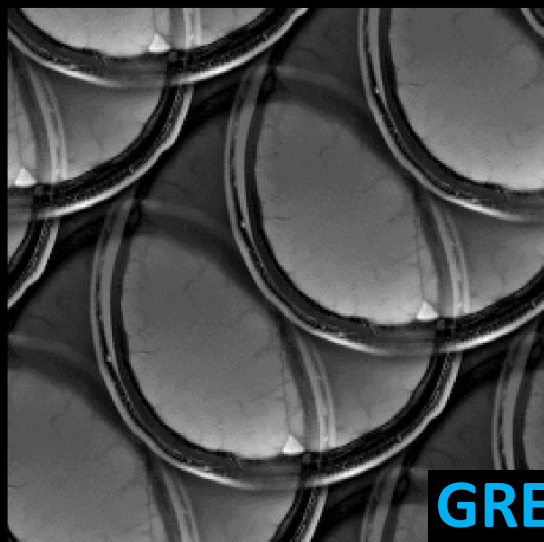


Error scaled 10x

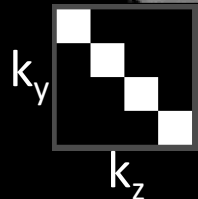
Error scaled 10x

Multi-Contrast, Calibrationless, Acceleration R=2x2

NLINV: single contrast



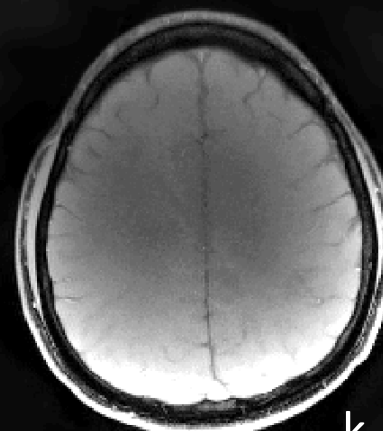
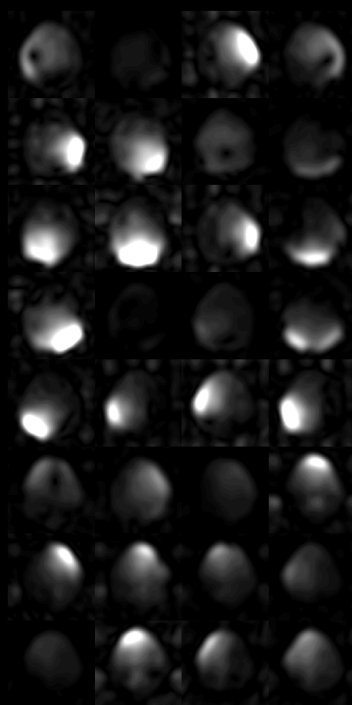
70.2% RMSE



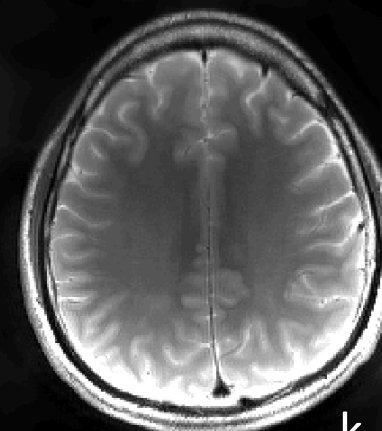
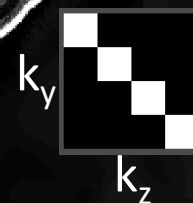
Proposed: Joint NLINV

5.2% RMSE

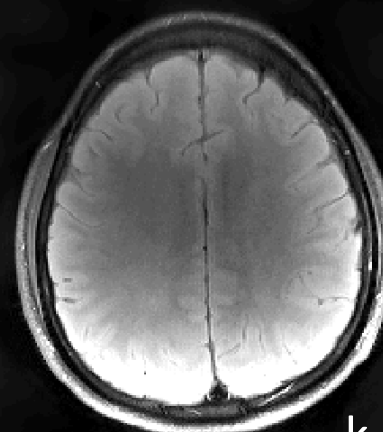
32 chan sens



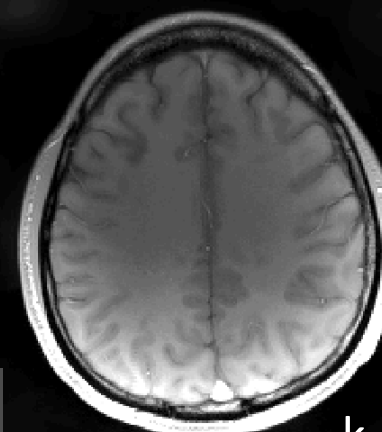
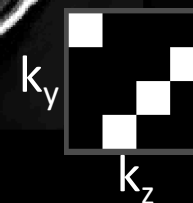
GRE



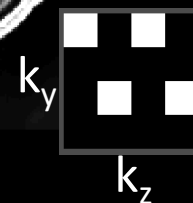
TSE



FLAIR



T1w

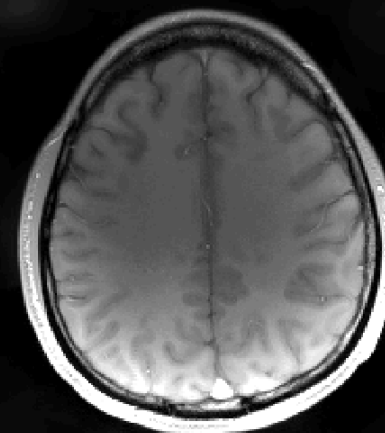
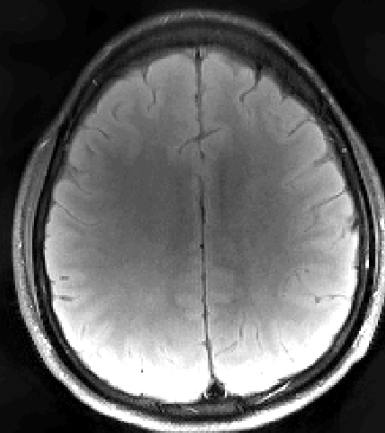
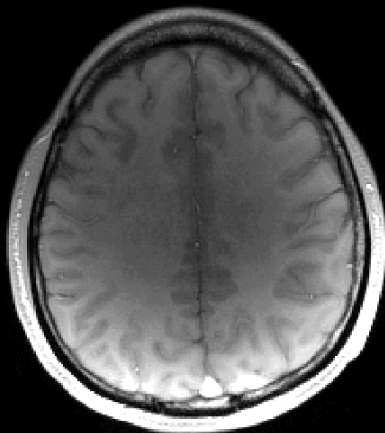
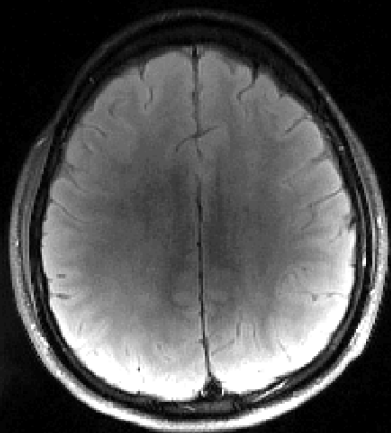
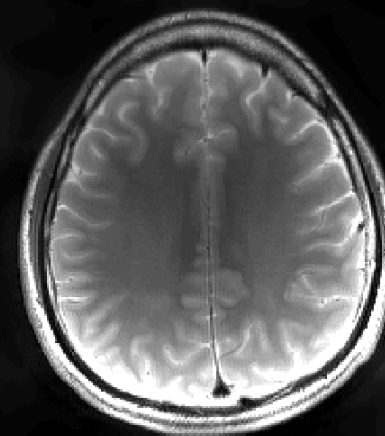
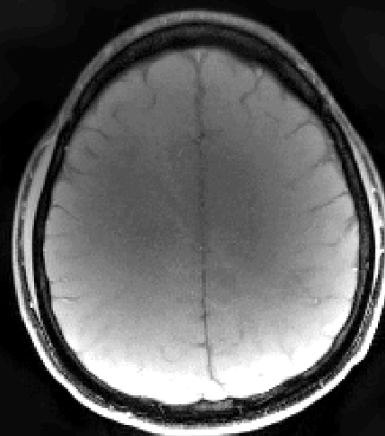
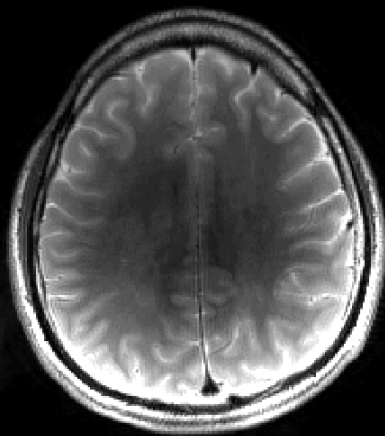
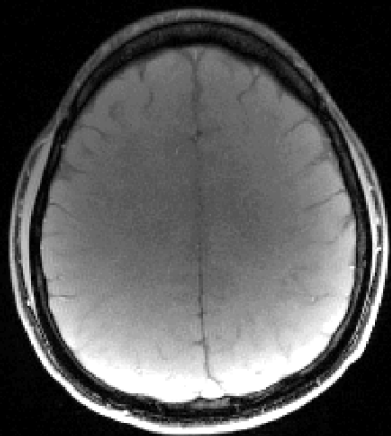


Multi-Contrast, Calibrationless, Acceleration R=2x2

SAKE 8.9% RMSE

Proposed: Joint NLINV

5.2% RMSE



Recon time: 171 min
Window= 6x6

Poisson sampling
Threshold = 1.3

Recon time: 31 min
 $\alpha_0 = 0.1$

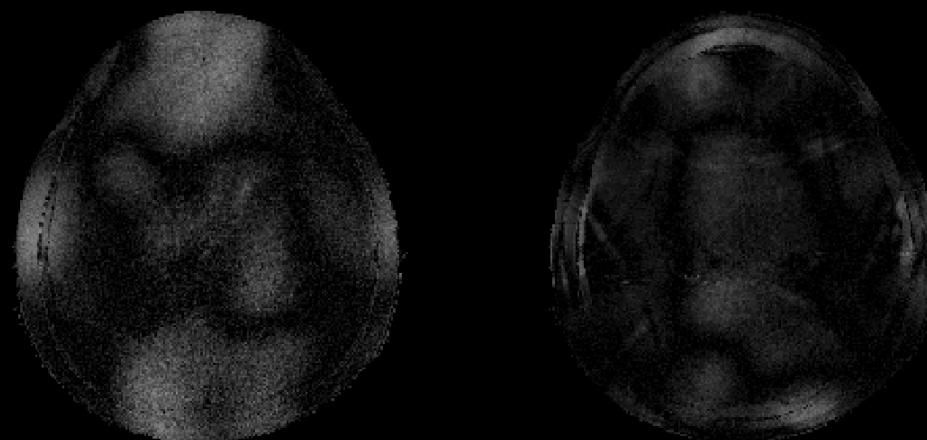
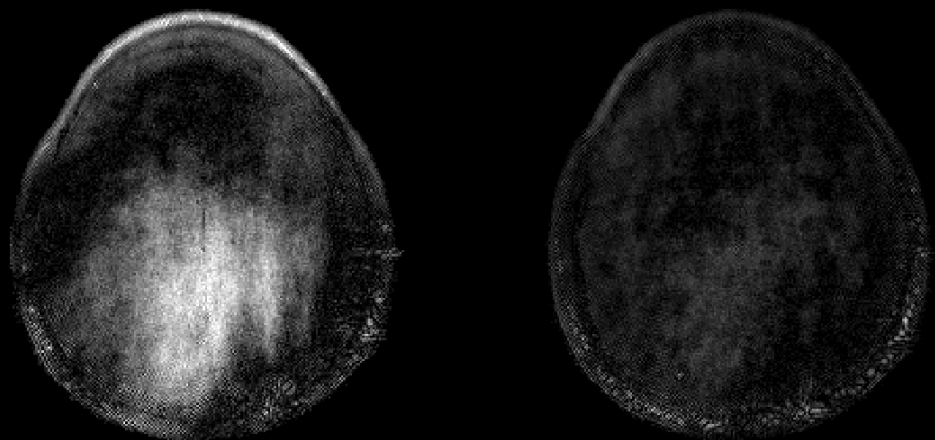
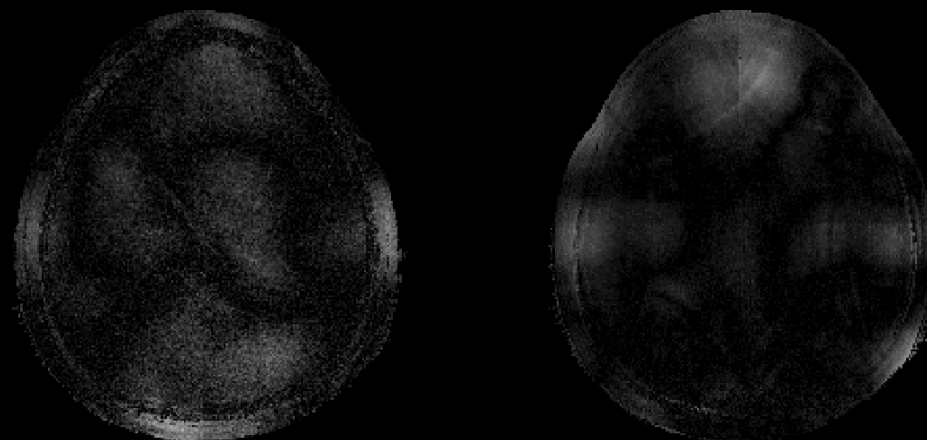
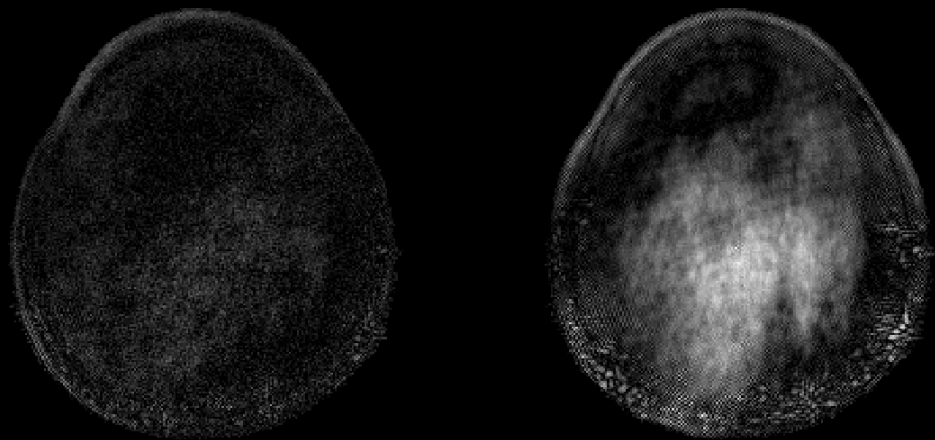
Uniform sampling
 $\beta_0 = 0.1$

Multi-Contrast, Calibrationless, Acceleration $R=2\times 2$

SAKE 8.9% RMSE

Proposed: Joint NLINV

5.2% RMSE



Error scaled 5x

Error scaled 5x

Discussion

- Standard NLINV without calibration is ill-conditioned and results in large errors
- Joint NLINV exploits shared sensitivities and joint sparsity across multiple echoes/contrasts to dramatically improve the recon
- This obviates the need for calibration, with >6-fold reduced RMSE over standard NLINV

Discussion

- Complementary sampling permits formation of a composite image by summing the k-space across echoes
- For spin-echo and MPRAGE where phase discrepancy between echoes is low, composite image has reduced aliasing and is used for initialization
- For phase-cycled bSSFP and multi-contrast, composite image contains significant aliasing and was not used
- This necessitated a larger starting parameter α_0 , which is reduced over iterations to result in negligible regularization in the final recon

Discussion

- SAKE: calibrationless from single image
elliptical Poisson sampling [1]
- Joint NLINV: allowed 30 – 70 % reduction in RMSE for 32-chan experiments
recon time reduced by 3× – 6× fold

[1] PJ Shin, MRM'14

Thanks!

Questions / Comments:

berkin@nmr.mgh.harvard.edu

martinos.org/~berkin

Support: NIH

R24 MH106096

R01 EB020613

R01 EB017337

U01 HD087211