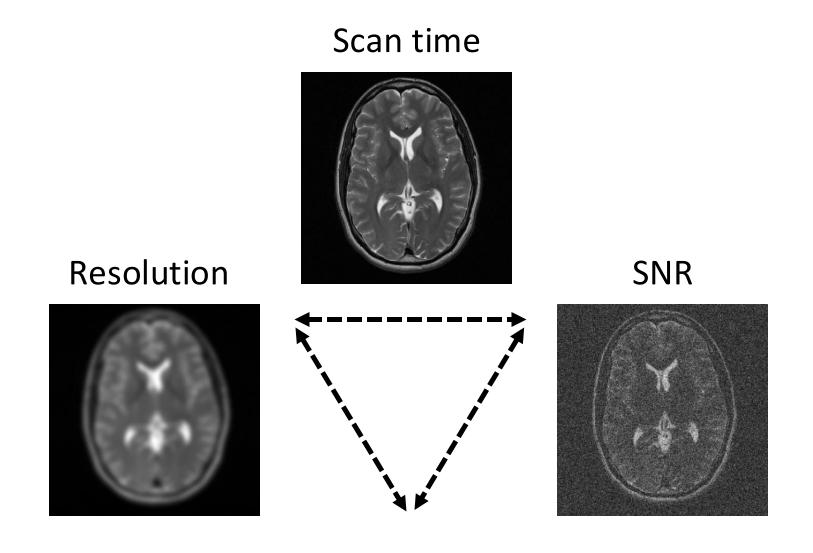
#### Computational MRI

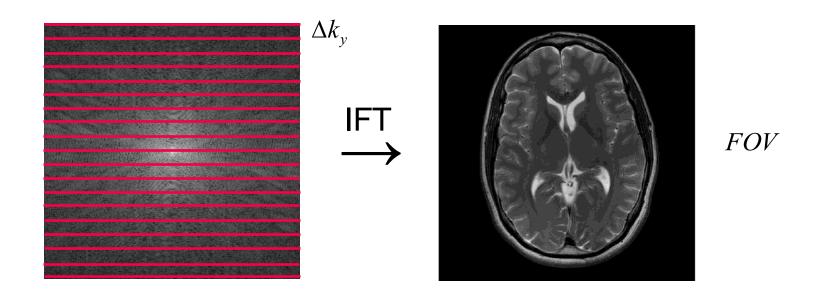
Parallel imaging I: Image-domain methods



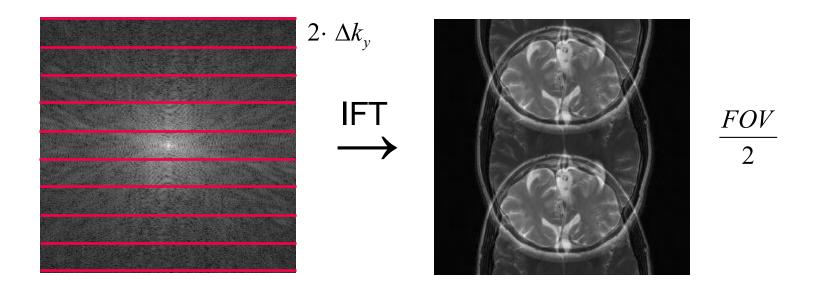


### MR imaging constraints: The triangle of fate





 Faster, no changes in gradient switching, but conventional Fourier reconstruction will result in aliasing artifacts



Question: Can we undersample in the readout dimension?

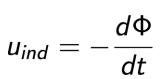
- Reconstruction
  - Exploit data redundancies!

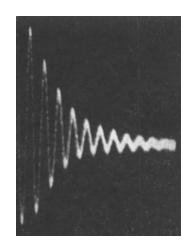
- Partial Fourier (constrained reconstruction), compressed sensing, machine learning,...
  - Image compressibility/sparsity (inherent redundancy)

- Parallel imaging
  - Multiple coils with different spatial sensitivities (real data redundancy)



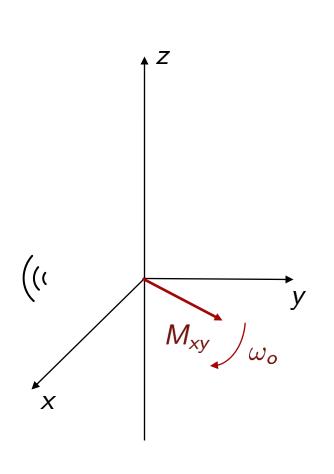
### Signal reception: MR receive coils





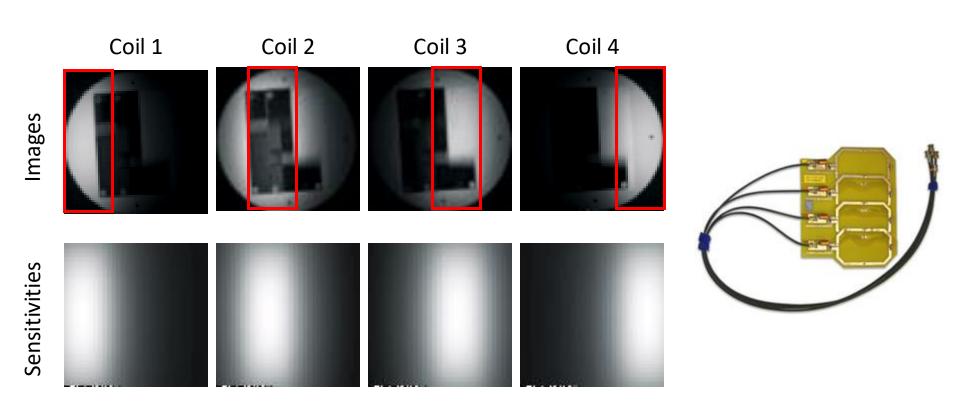
Hahn 1950



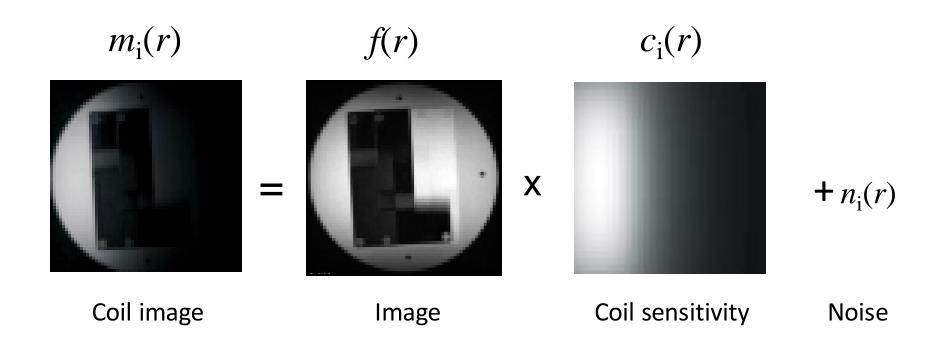


 $B_0$ 

Different spatial sensitivities



Sensitivity-encoding equation



- First used to improve SNR
  - What is the optimal coil combination?
- Matched-filter or least-squares combination

$$f(r) = \frac{\sum_{i=1}^{N_c} c_i^*(r) m_i(r)}{\sqrt{\sum_{i=1}^{N_c} |c_i(r)|^2}}$$

 $m_i(r)$ : single coil images

 $c_i(r)$ : coil sensitivities

Roemer FB et al. Magn Reson Med. 1990; 16(2):192-225.



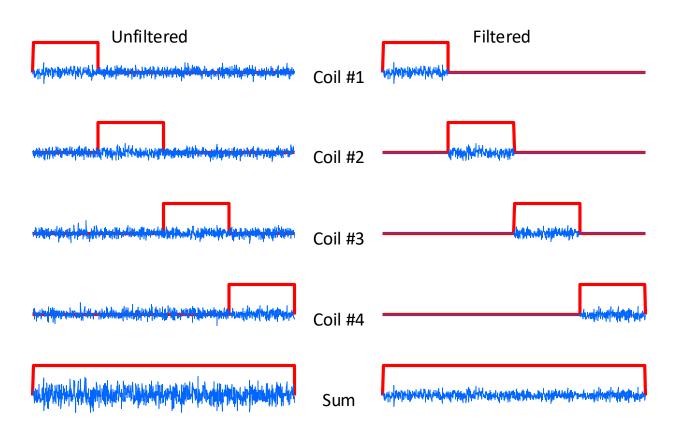
- Matched-filter or least-squares combination
  - In matrix form (for each pixel)

$$f = \left(\mathbf{C}^H \mathbf{C}\right)^{-1} \mathbf{C}^H \mathbf{m}$$

$$\mathbf{C} = \begin{pmatrix} c_1 \\ c_2 \\ \vdots \\ c_N \end{pmatrix} \qquad \mathbf{m} = \begin{pmatrix} m_1 \\ m_2 \\ \vdots \\ m_N \end{pmatrix}$$

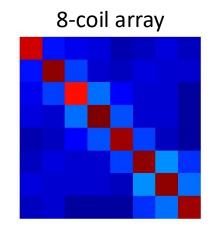
Roemer FB et al. Magn Reson Med. 1990; 16(2):192-225.

- Matched-filter or sensitivity-weighted combination
  - Effects on noise



Noise signals from different coils are correlated

Ψ: coil noise covariance matrix



Least-squares combination using the covariance matrix

$$f = \left(\mathbf{C}^H \mathbf{\Psi}^{-1} \mathbf{C}\right)^{-1} \mathbf{C}^H \mathbf{\Psi}^{-1} \mathbf{m}$$

- Pre-whitening
  - Virtual coils with uncorrelated noise

$$\mathbf{m}_{w} = \mathbf{\Psi}^{-\frac{1}{2}} \mathbf{m}$$
$$\mathbf{C}_{w} = \mathbf{\Psi}^{-\frac{1}{2}} \mathbf{C}$$

Solution

$$f = \left(\mathbf{C}_{w}^{H} \mathbf{C}_{w}\right)^{-1} \mathbf{C}_{w}^{H} \mathbf{m}_{w}$$

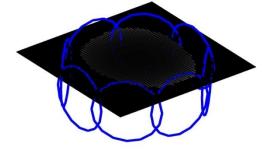
- Sum of squares
  - Approximation to the optimal combination
  - Images as coil sensitivities
  - SNR penalty of about 10%

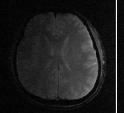
$$c_i(r) = m_i(r) \implies f(r) = \sqrt{\sum_{i=1}^{N_c} |m_i(r)|^2}$$

$$\mathbf{c} = \mathbf{m}$$
  $\Rightarrow$   $f = \sqrt{\mathbf{m}^H \mathbf{m}}$ 

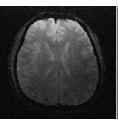
# Signal combination example

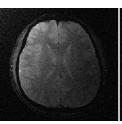
• 8-coil circular array

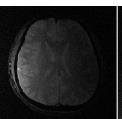




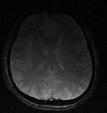






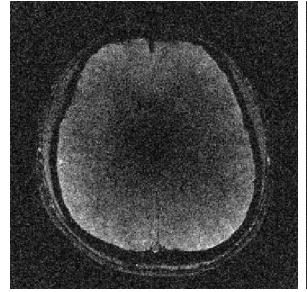




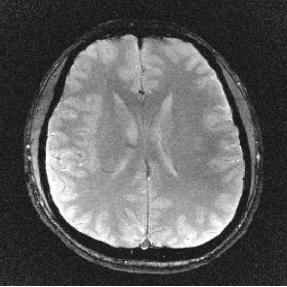




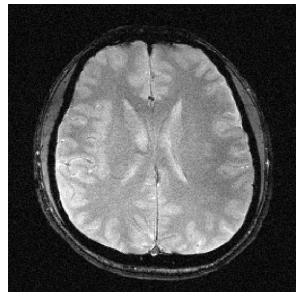
Complex sum



Sum of squares

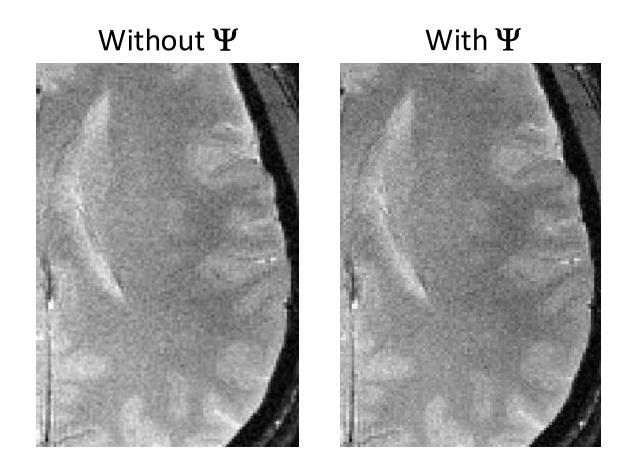


Least-squares (matched-filter)



## Signal combination example

• Matched-filter combination

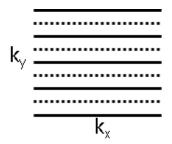


# Parallel imaging

### Parallel imaging

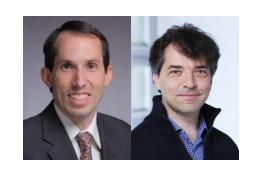
- Multiple coils enable acceleration of MRI data acquisition
  - Multiple coil data are redundant!

Regular k-space undersampling

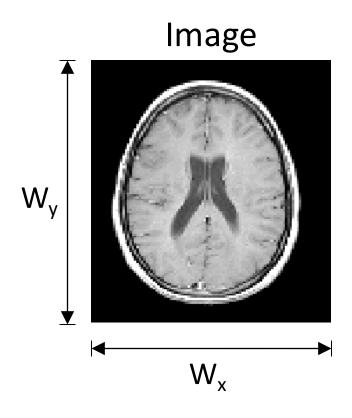


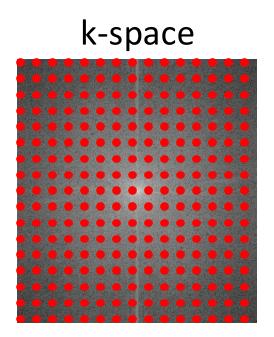
Reconstruction using matrix inversion

Sodickson DK, Manning WJ. Magn Reson Med. 1997; 38: 591-603 Pruessmann KP et al. Magn Reson Med 1999; 42: 952-962



## Recap: k-space sampling density and image FOV

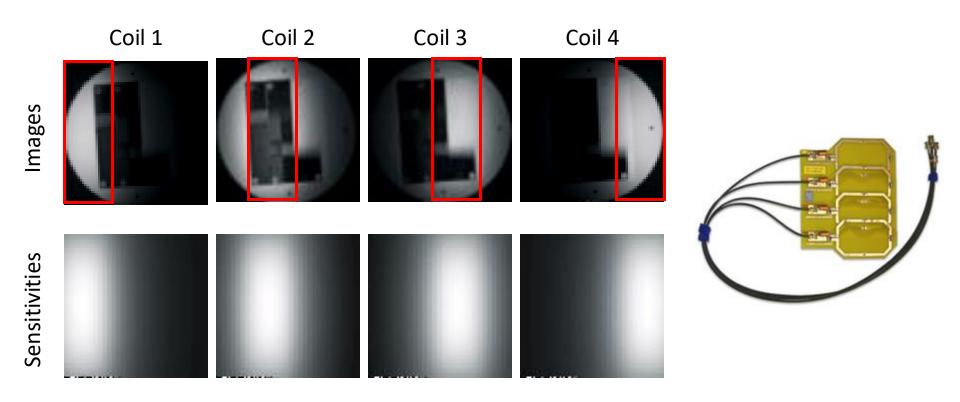




Nyquist rate:

$$\Delta k_x = \frac{1}{W_x}; \quad \Delta k_y = \frac{1}{W_y}$$

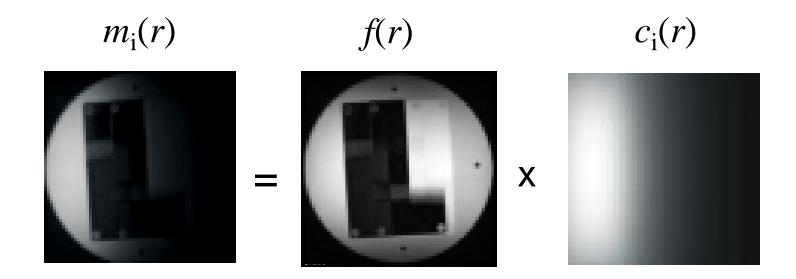
### Spatial encoding of receive coils



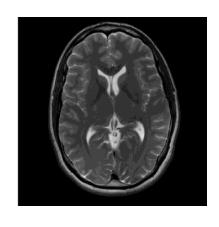
#### Coils also perform spatial encoding

- Pixels close to the coil are bright
- Pixels far from the coil are dark

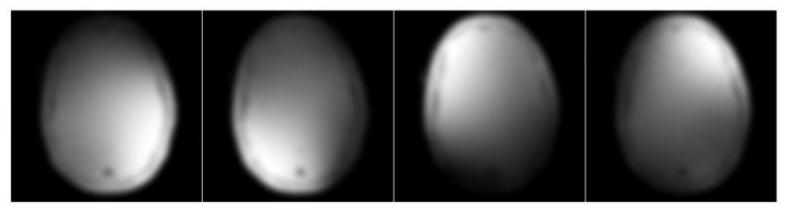
### Back to sensitivity-encoding equation



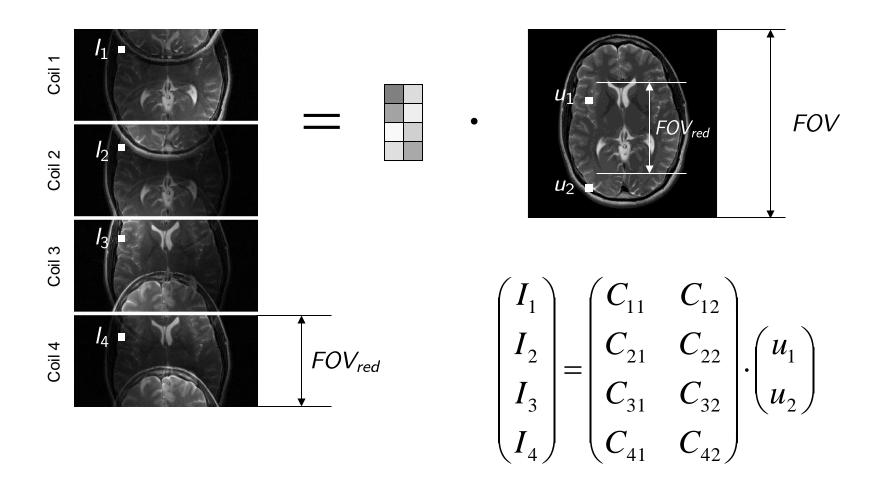
## Cartesian SENSE



- T2 weighted brain scan
- 4-Channel receive coil
- Sensitivities are known



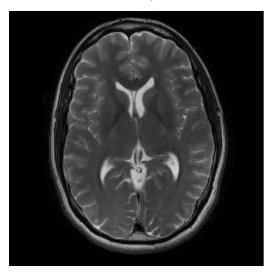
#### Cartesian SENSE



# Parallel Imaging: SENSE

$$u = (C^H C)^{-1} C^H I$$

SENSE, R=2



#### SNR penalty in parallel imaging

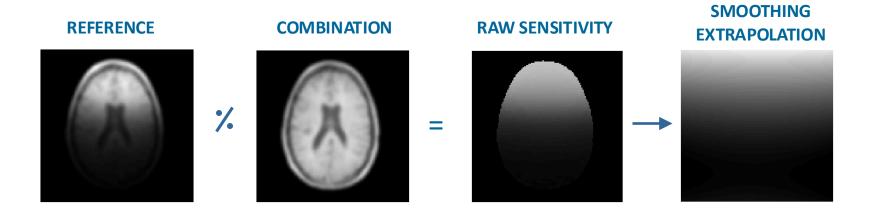
$$SNR_{acc} = \frac{SNR_{no-acc}}{g\sqrt{R}}$$

 g-factor: noise amplification due to ill-conditioning of the encoding matrix

$$g(r) = \sqrt{\left(\mathbf{E}^{H}\mathbf{\Psi}^{-1}\mathbf{E}\right)^{-1}\mathbf{E}^{H}\mathbf{\Psi}^{-1}\mathbf{E}}_{(r)}$$

#### Coil sensitivity estimation for SENSE

- Estimation of pure coil sensitivities (Pruessmann et al. MRM 1999).
  - Separate low resolution image for each coil.



#### How to reduce noise amplification?

Use more coils

Improve coil array design

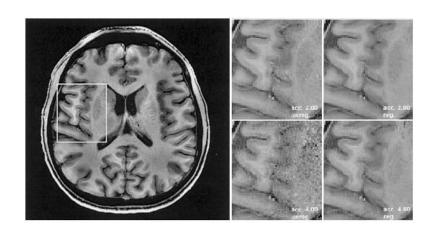
Regularization of the inverse reconstruction

2D acceleration instead of 1D acceleration (3D imaging)

#### Regularization of the inverse reconstruction

- Constrain the inverse problem to reduce noise amplification and control numerical instabilities
- Method 1: Tikhonov regularization
  - Constrain the power of the solution

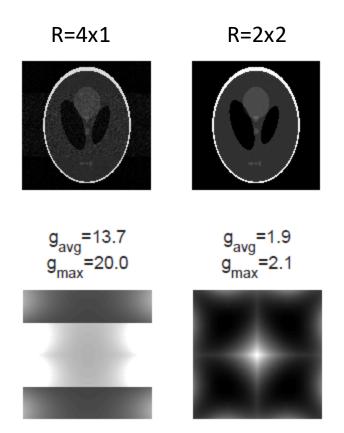
$$\hat{\mathbf{m}} = \min_{\mathbf{m}} \left\{ \left\| \mathbf{E} \mathbf{m} - \mathbf{s} \right\|_{2}^{2} + \lambda \left\| \mathbf{m} \right\|_{2} \right\} = \left( \mathbf{E}^{H} \mathbf{E} + \lambda \mathbf{I} \right)^{-1} \mathbf{E}^{H} \mathbf{s}$$

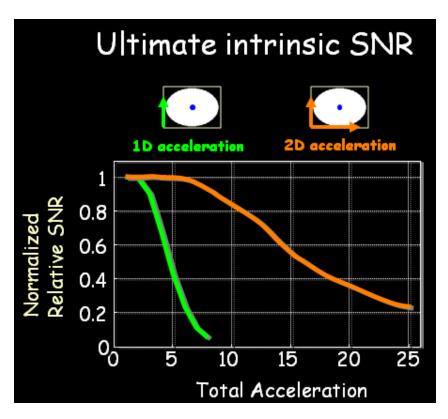


Lin FH et al, Magn Reson Med 2004; 51:559-567

#### 2D acceleration Vs. 1D acceleration

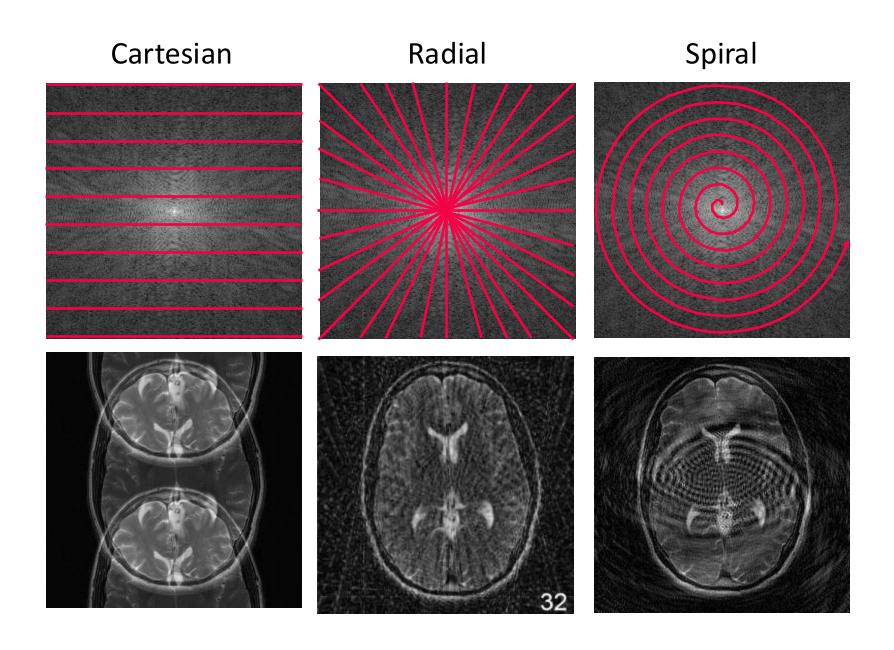
2D acceleration reduces g-factor





Ohliger MA et al. MRM 2003;50:1018-30

# Non-Cartesian undersampling



#### Non-Cartesian SENSE

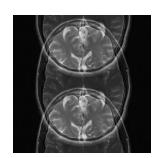
Decoupling is lost

- Each pixel is aliased with all other pixels
  - e.g. streaks in undersampled radial imaging

Need to invert the full encoding equation

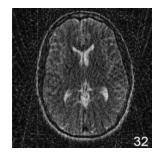
- Calls for an iterative algorithm
  - No explicit matrix inverse
  - Matrix-vector multiplications only

Cartesian





Radial



### Summary

- Fast, rapid or accelerated MRI
  - k-space undersampling
  - Reconstruction is more challenging, but more fun
  - Exploit redundancies in the acquired data

### Summary

- Parallel imaging
  - Exploit additional encoding provided by multiple receiver coils with different sensitivities
  - SNR penalty
  - SENSE (image-domain)
    - Unfolding images using coil sensitivities
    - Matrix inversion
  - SMASH, GRAPPA (k-space)
    - Next lecture