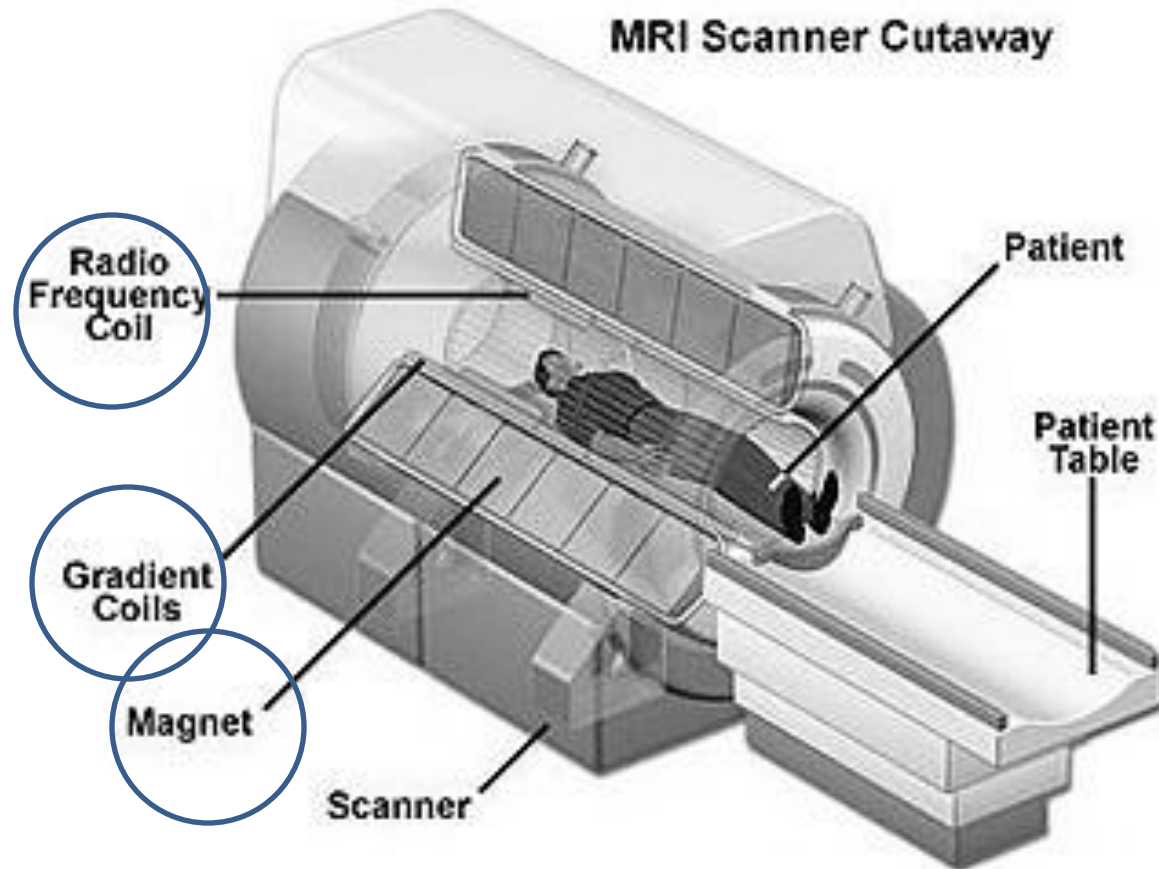


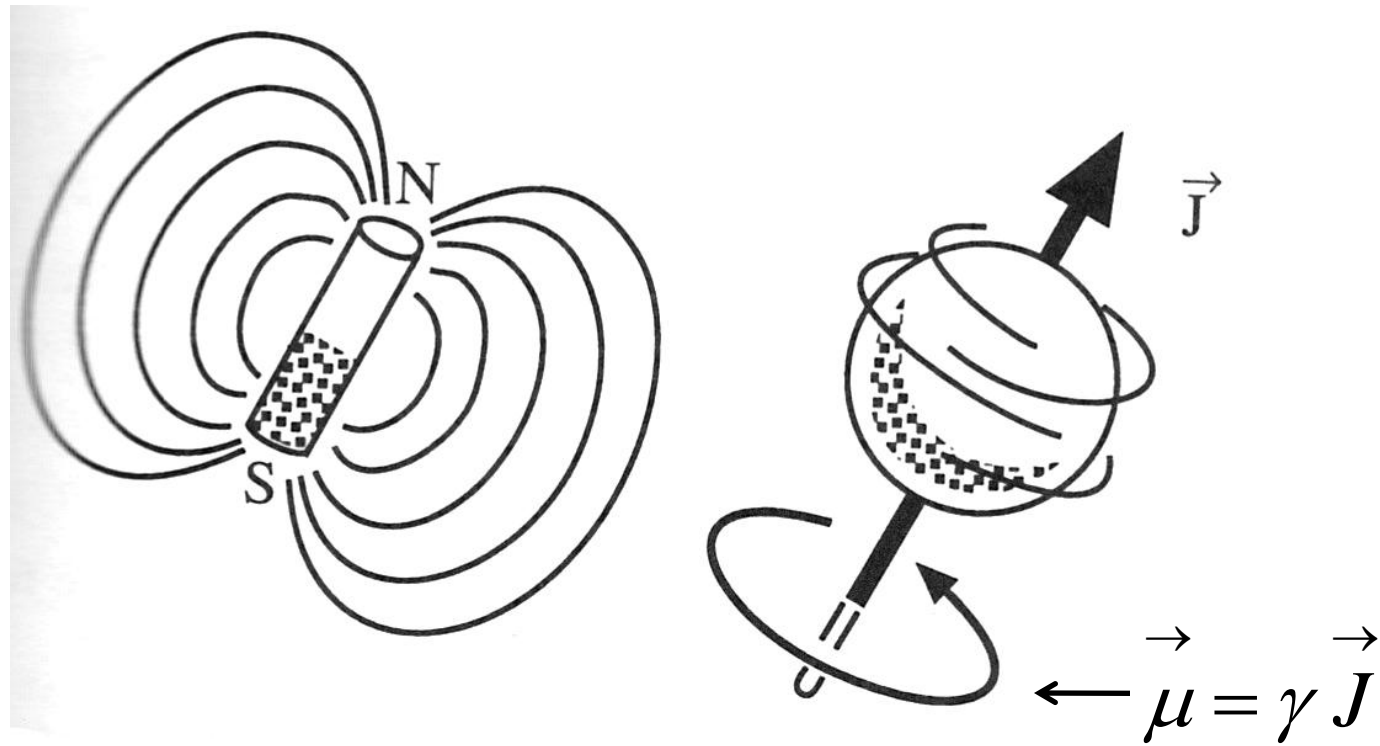
# Introduction to Accelerated Magnetic Resonance Imaging

Arjun Arunachalam  
Consultant, Philips

# A typical MRI Scanner



# The origin of MRI signal

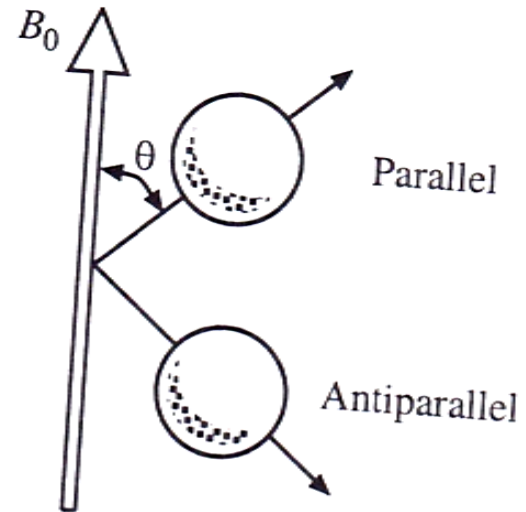
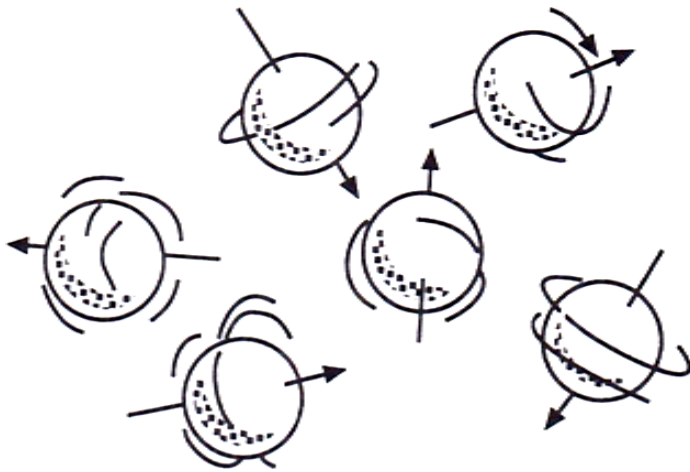


- Nuclei with odd atomic weights/numbers possess spin(J)
- Nuclear magnetism: Spin system in a magnetic field
  - Proton has electrical charges
  - Rotates about its own axis if it has non-zero spin

$\gamma$ - Gyromagnetic ratio  
 $\mu$ -magnetic moment

# Nuclear magnetic moments ( $\mu$ )

- A vector whose magnitude is given by  $\gamma = \sqrt{h(h+1)}$   
- $h$ -Planck's constant
- The direction of  $\mu$  is determined by the main magnetic field ( $B_0$ )



$$\vec{M} \propto (N_{\uparrow} - N_{\downarrow}) \approx N_s \frac{\lambda h B_0}{2KT_s}$$

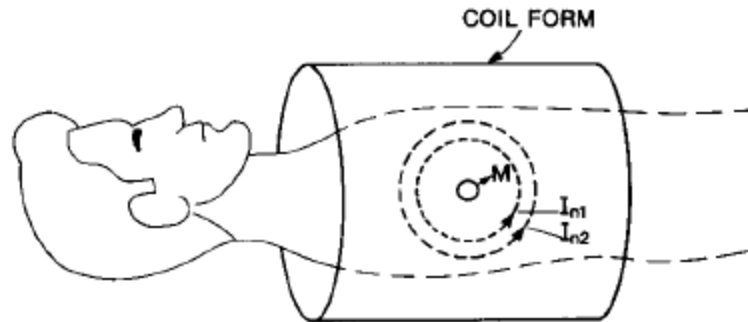
# Role of hardware in Imaging

Hardware	Imaging parameter
The magnet	Signal to Noise (SNR)
Gradient coils	Fourier data acquisition
RF Antenna	Role in data acquisition

- Magnet-Intrinsic SNR- Indispensable to sparse signal recovery
- Gradient coils- k-space acquisition-Limits acquisition speed
- RF Antennas- Important role in sparse signal recovery

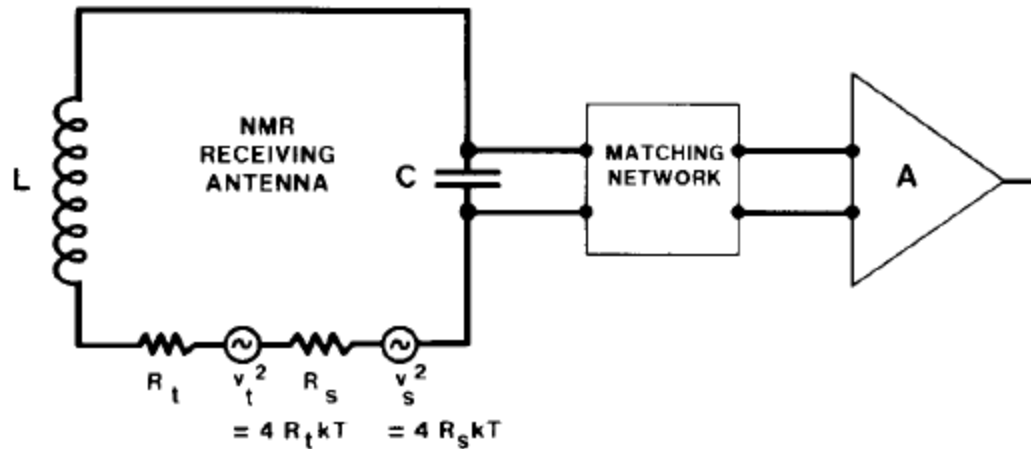
# Intrinsic SNR

- Intrinsic SNR is the ratio of the signal from the magnetization  $M$  to the signals produced by thermally generated noise currents



$$\vec{M} \propto B_0$$

# Intrinsic SNR



- $k$  - Boltzmann constant
- $T$  - Absolute temperature in degree Kelvin

$$S \propto \frac{dM}{dt} \propto \omega M \quad \text{and} \quad \omega = \gamma B_0, M \propto B_0$$

# Intrinsic SNR

Therefore,  $S \propto B_0^2$

But we also know<sup>1</sup>,  $v_s \propto B_0, v_t \propto \sqrt{B_0}$

Therefore at high field strengths, Intrinsic SNR  $\frac{S}{v_s} \propto B_0$

1. D. I. HOULT AND P. C. LAUTERBUR *J., Magn. Reson.* **34,425 (1 979)**.



The objective of Accelerated MRI is to minimize data acquisition time..

# The signal acquisition model

$$s(t, t_y) = \iint S_l(x, y) M(x, y) e^{-i\gamma G_y y t_y} e^{-i\gamma G_x x t} dx dy$$

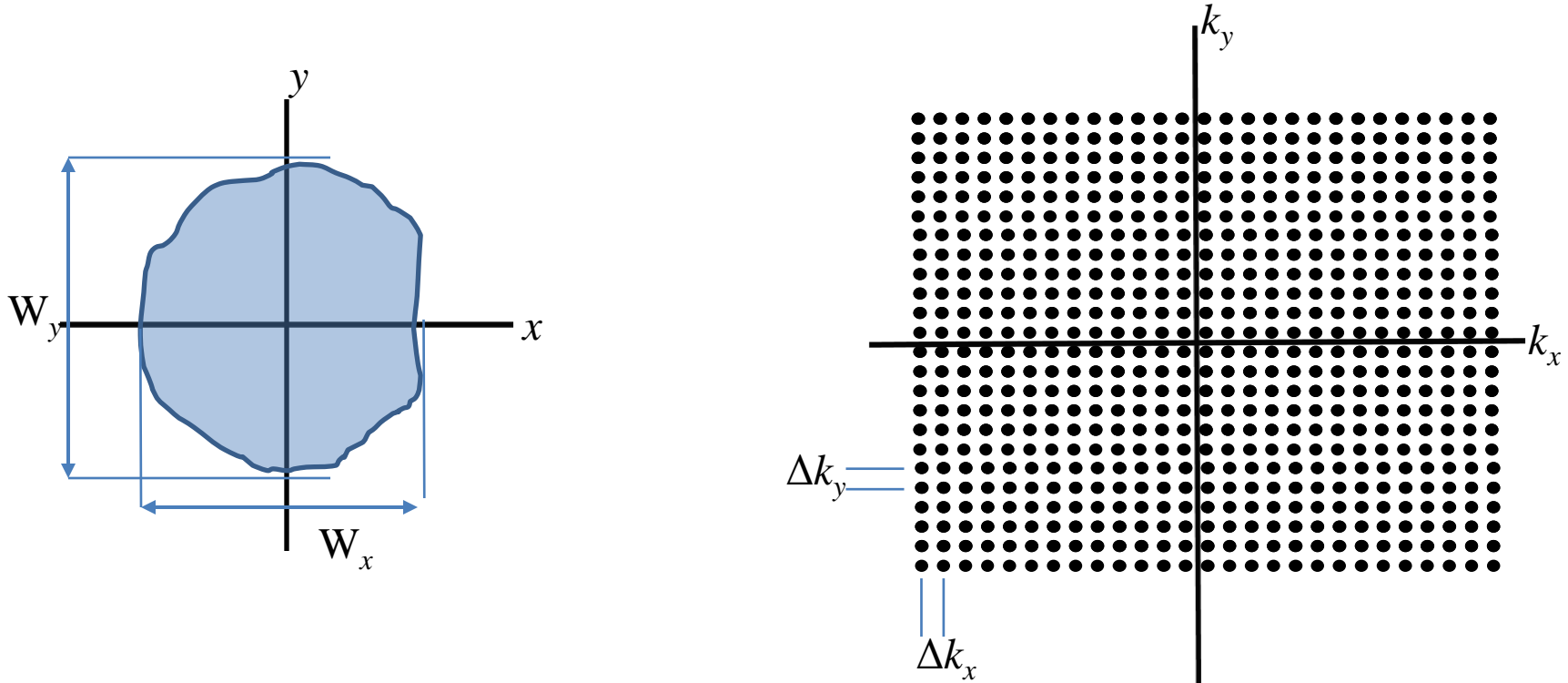
- $M$  – Magnetization distribution in x-y plane
- $t_y$  – phase encoding gradient on-time
- $t$  – Frequency encoding gradient on time
- $G_y$  - phase encoding gradient amplitude
- $G_x$  - Frequency encoding gradient amplitude
- $S_l$  - Signal weighting by the  $l^{\text{th}}$  RF antenna

$$F(k_x, k_y) = \iint S_l(x, y) M(x, y) e^{-i2\pi k_y y} e^{-i2\pi k_x x} dx dy$$

$$k_x = \left( \frac{\gamma}{2\pi} \right) G_x t$$

$$k_y = \left( \frac{\gamma}{2\pi} \right) G_y t_y$$

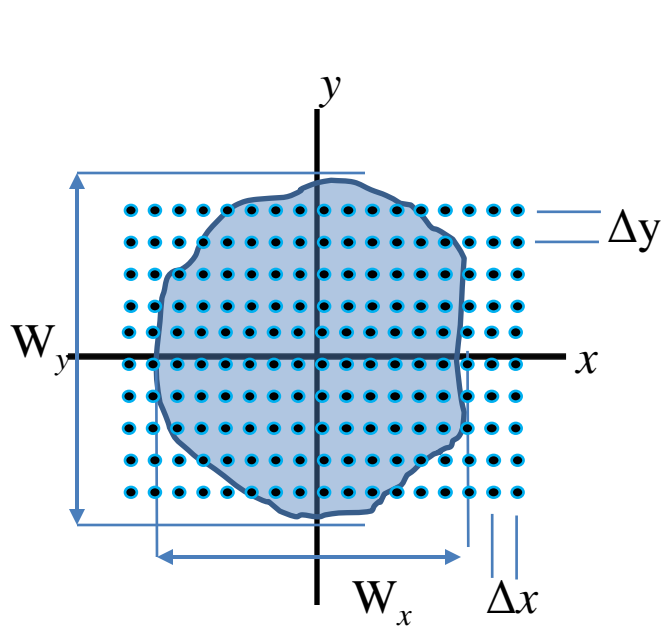
# Signal sampling requirements..



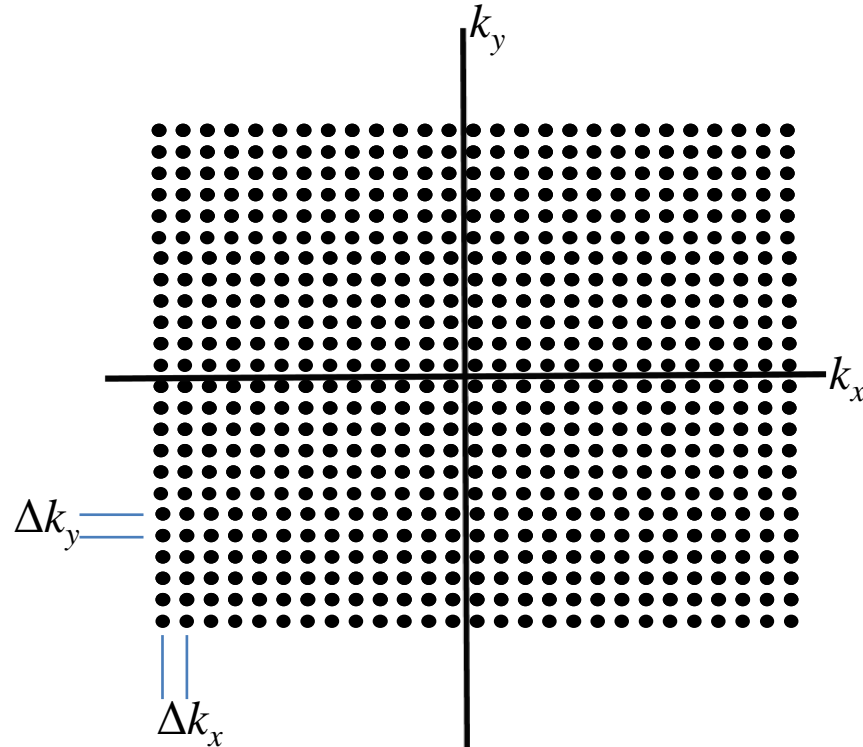
$$\Delta k_x = \left( \frac{\gamma}{2\pi} \right) G_x \Delta t \quad \Delta k_y = \left( \frac{\gamma}{2\pi} \right) \Delta G_y t_y$$

$$\Delta k_x \leq \frac{1}{W_x}, \Delta k_y \leq \frac{1}{W_y}$$

# Image resolution and sampling...



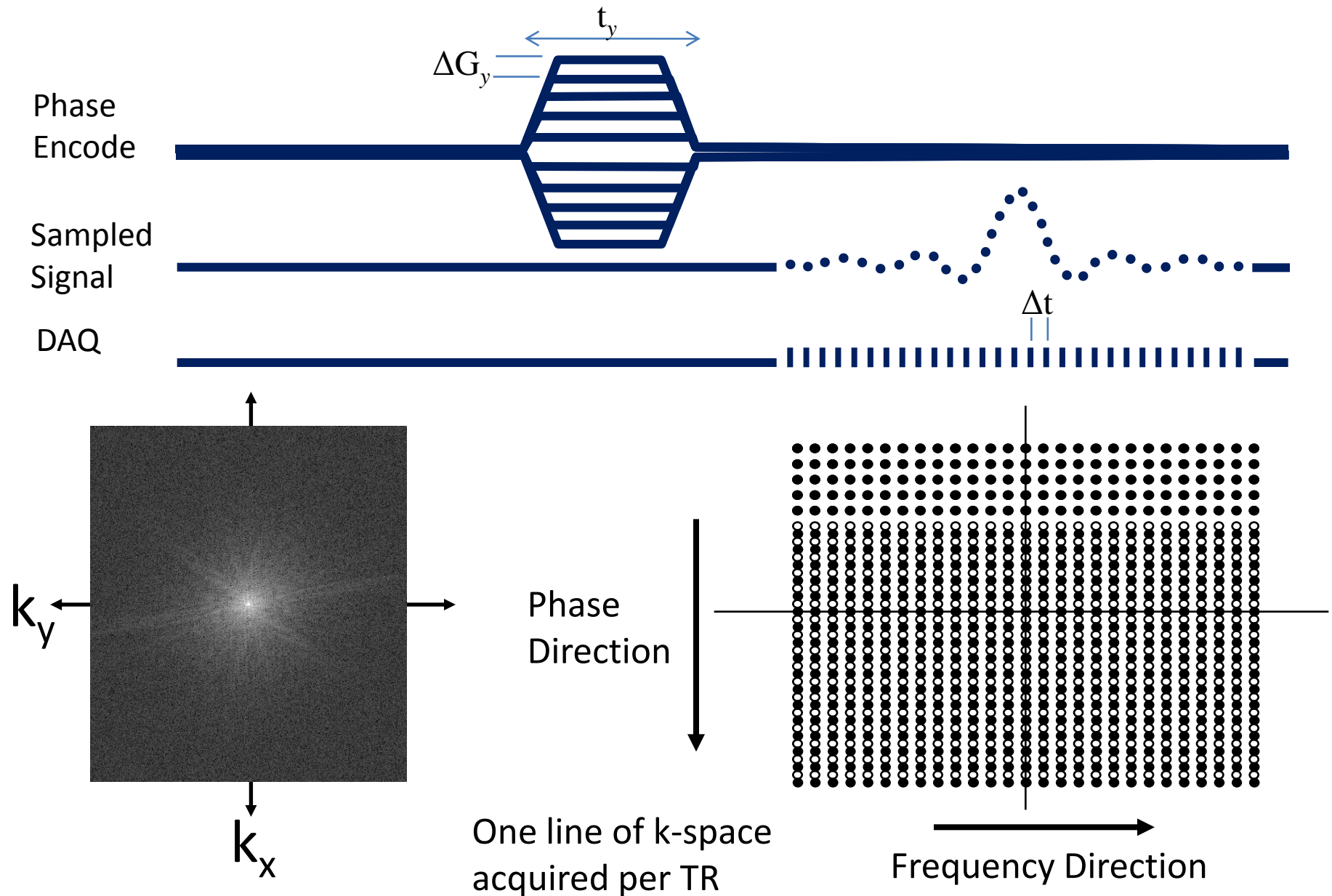
$$\Delta x = \frac{W_x}{N}, \Delta y = \frac{W_y}{N}$$



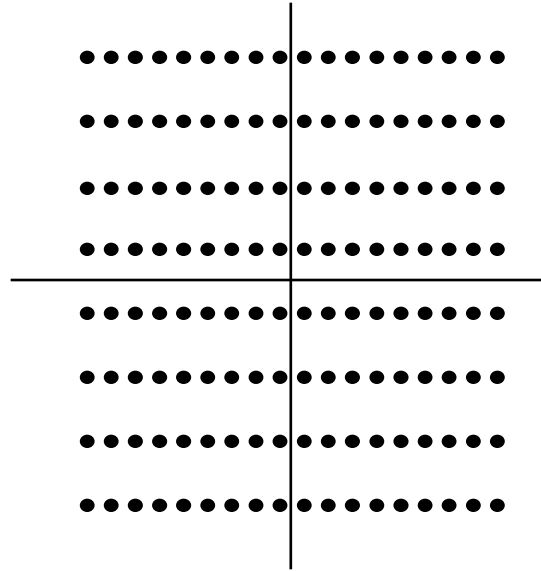
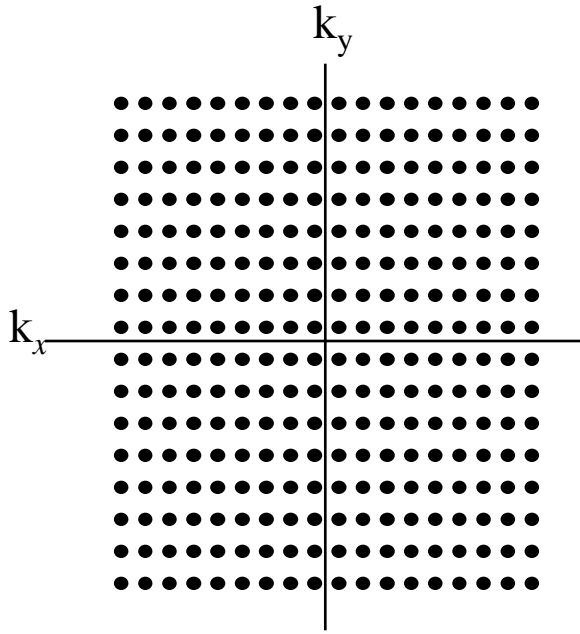
$$\Delta k_x \leq \frac{1}{W_x}, \Delta k_y \leq \frac{1}{W_y}$$

$$\Delta x = \frac{1}{N\Delta k_x}, \Delta y = \frac{1}{N\Delta k_y}$$

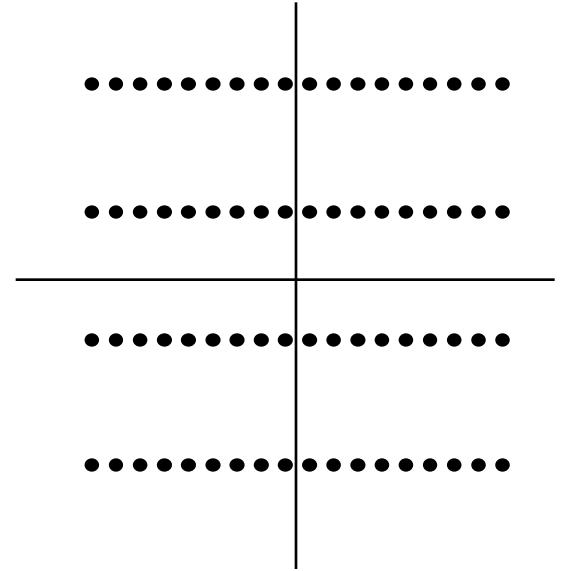
# Rectilinear acquisition



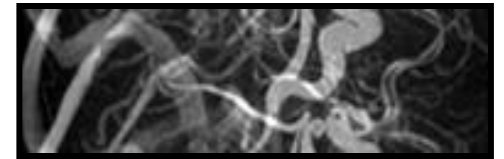
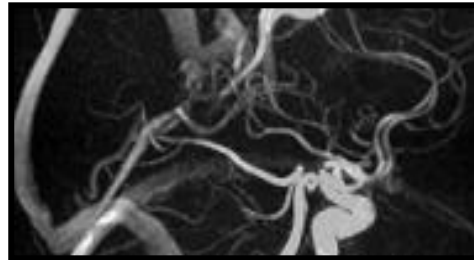
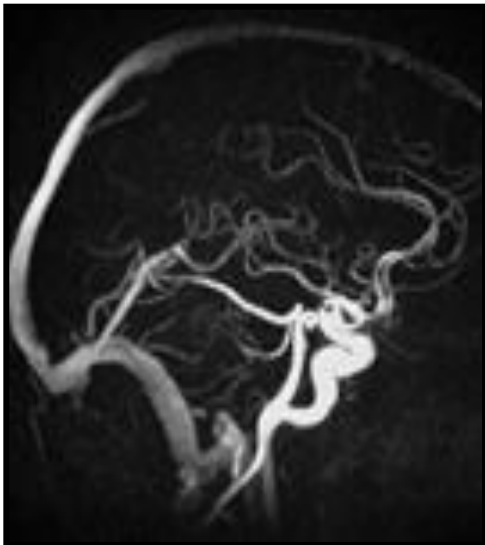
# Undersampling (Rectilinear Acquisition)



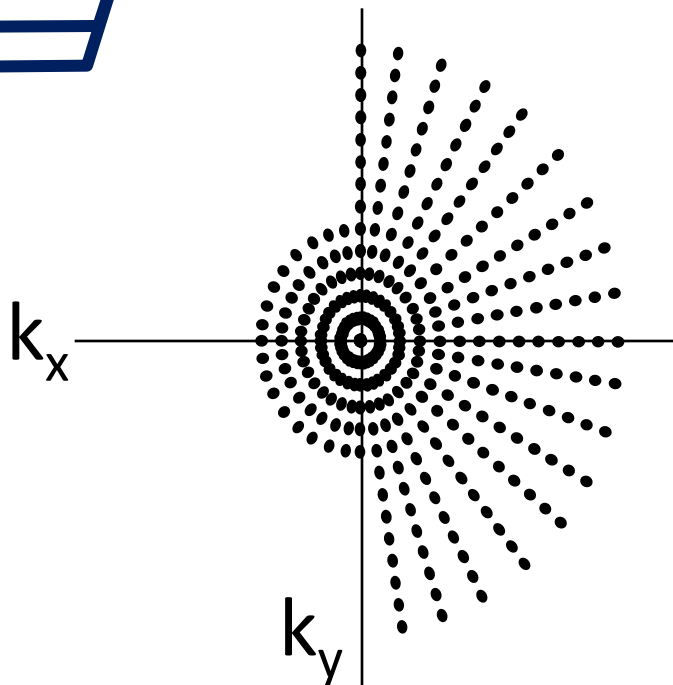
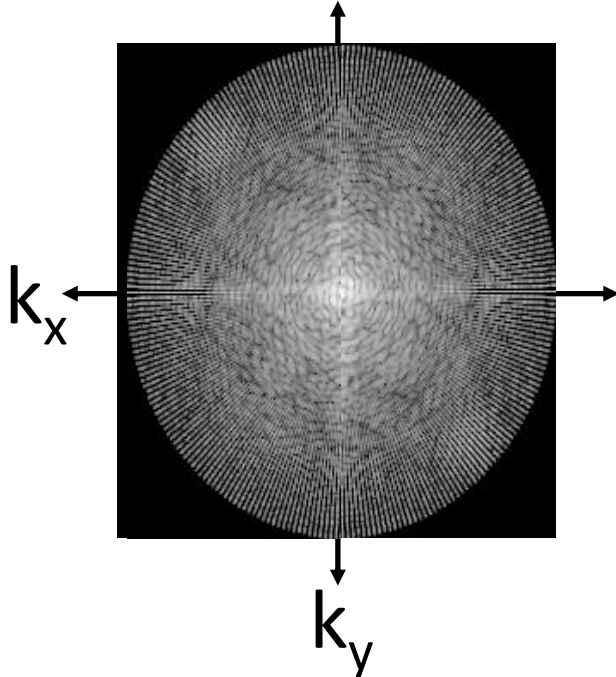
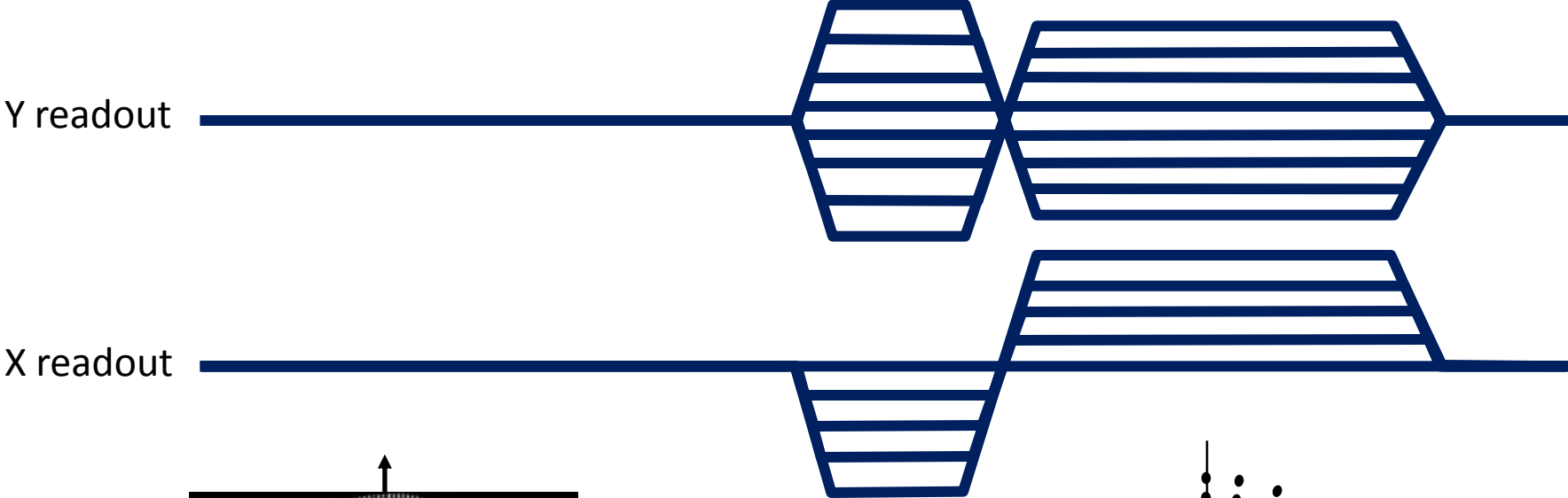
$1/2$  encodes



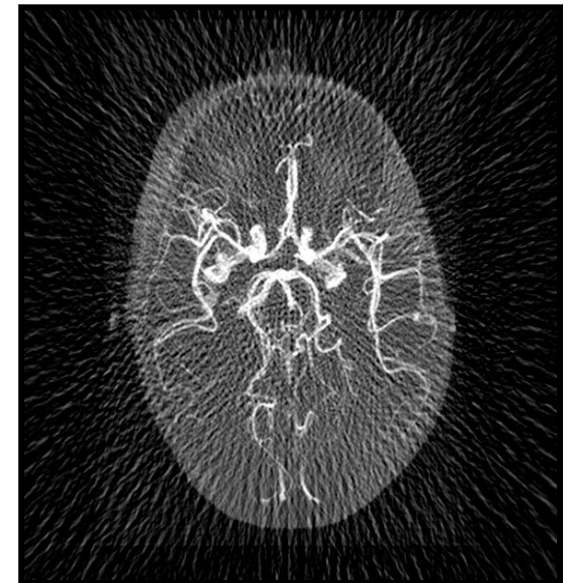
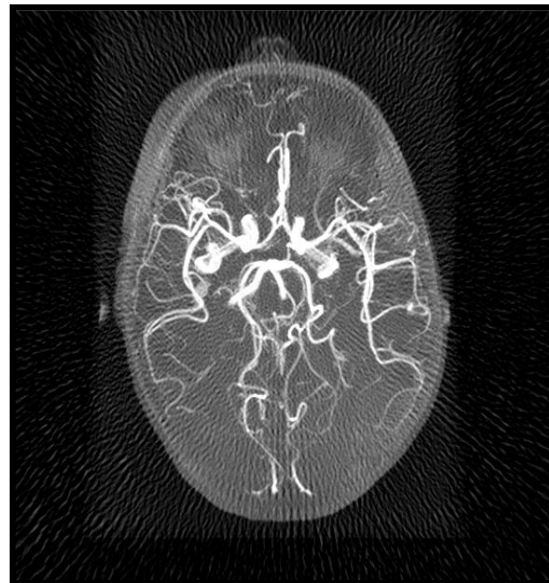
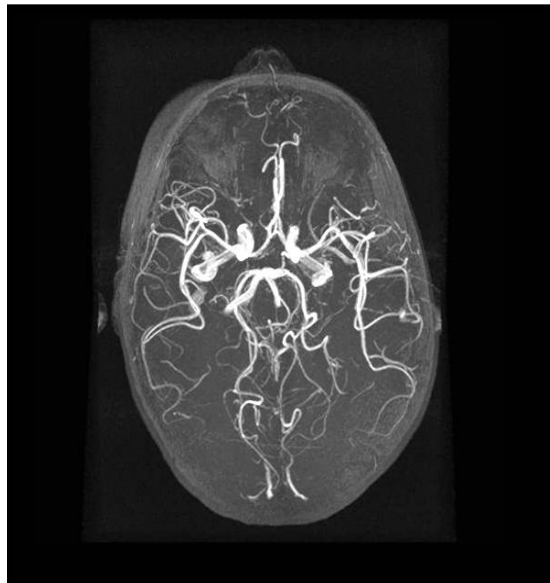
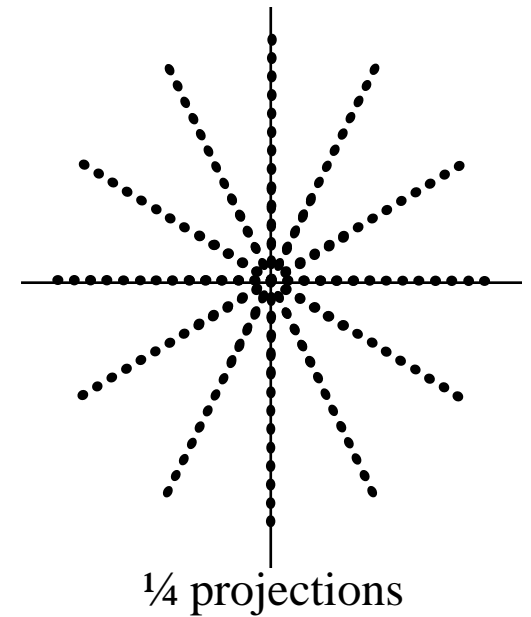
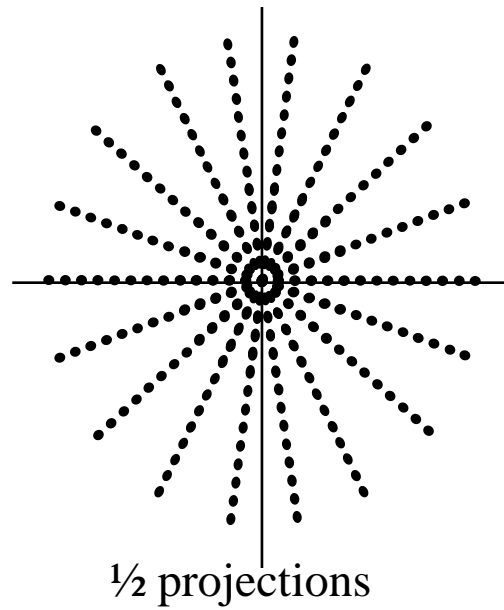
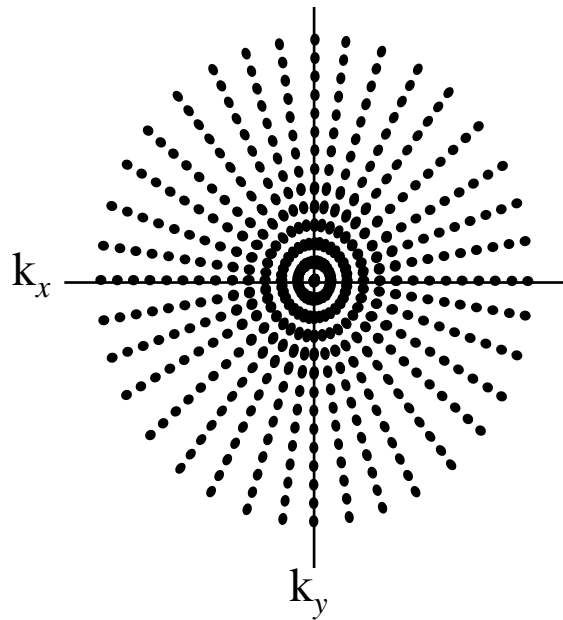
$1/4$  encodes



# k-Space Acquisition (Radial Sampling)



# Accelerated 2D Radial Sampling

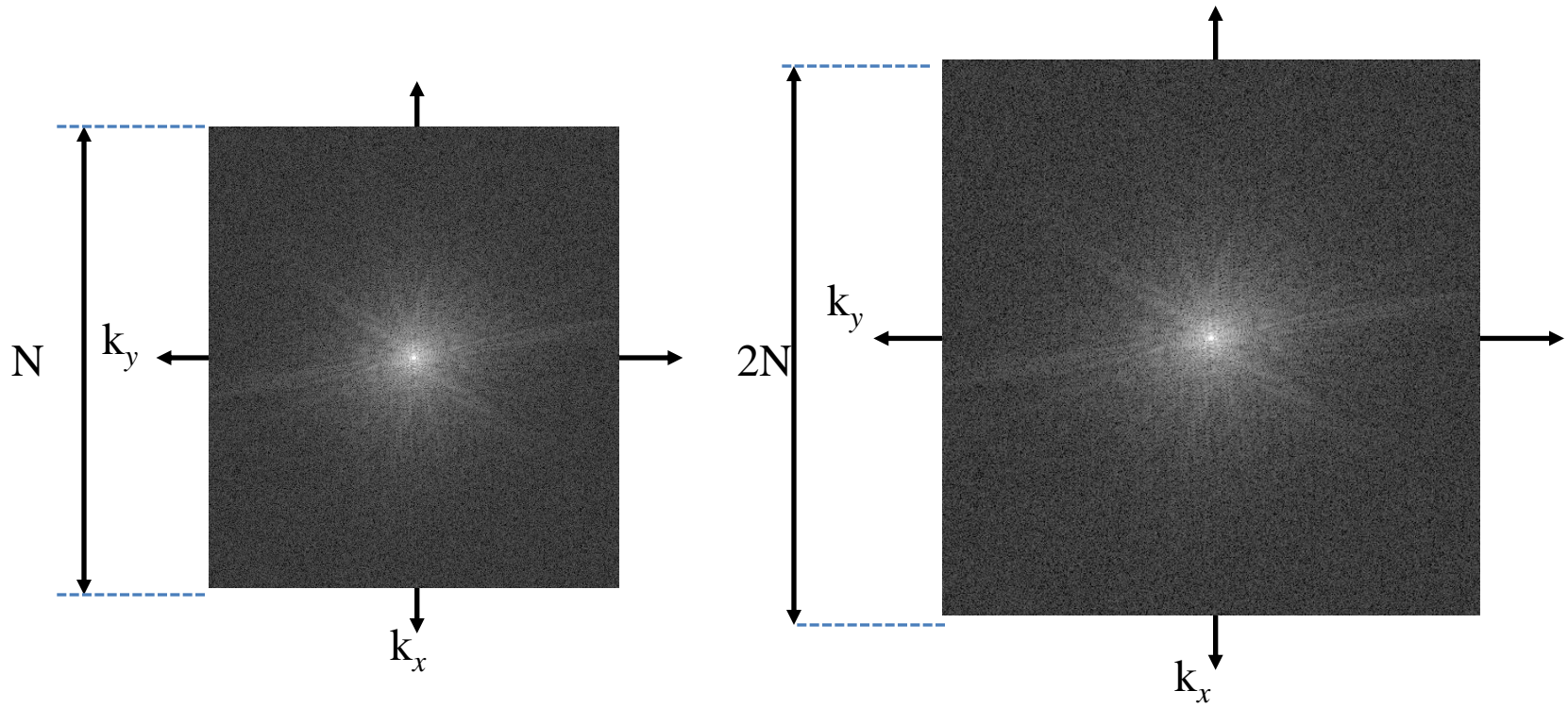




# Trade-offs

- Image Resolution
- Signal to Noise ratio (SNR)
- Acquisition Time

# Resolution Vs. Acquisition time



Double the resolution -> Double the time

# SNR-When resolution is fixed

- Consider an Impulse object centered at origin
  - Constant signal amplitude 'A' in k-space
  - Each sample contains independent noise with variance  $\sigma^2$

$$SNR = \frac{\sum_{j=1}^N A}{\sqrt{\sum_{j=1}^N \sigma_N^2}} = \frac{NA}{\sqrt{N\sigma_N^2}} = \frac{A\sqrt{N}}{\sigma_N}$$

SNR obtained  
By simply adding samples

- The effect of signal averaging-SNR improvement

$$SNR = \frac{\sum_{j=1}^N 2A}{\sqrt{\sum_{j=1}^N 2\sigma_N^2}} = \frac{2NA}{\sqrt{2N\sigma_N^2}} = \frac{A\sqrt{2N}}{\sigma_N}$$

$$SNR \propto \sqrt{N_{avg}}$$

# SNR-Accelerated scan-fixed resolution

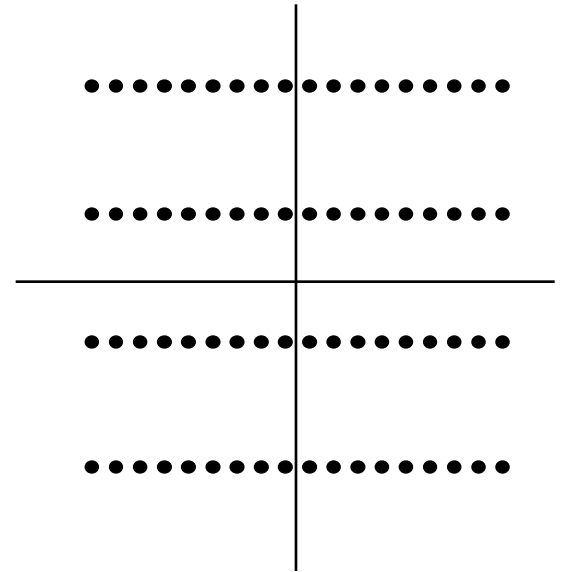
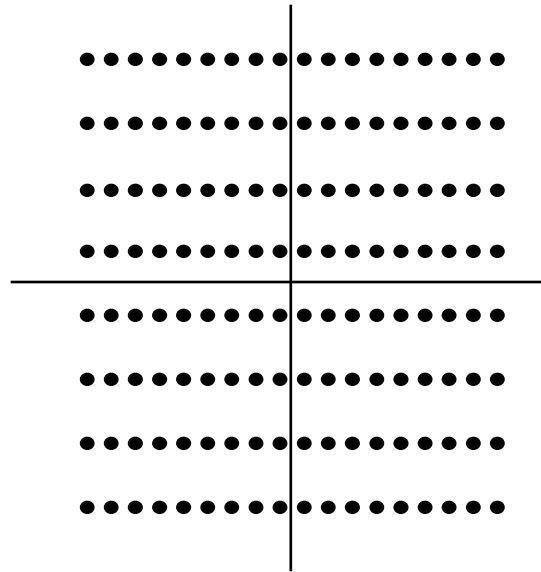
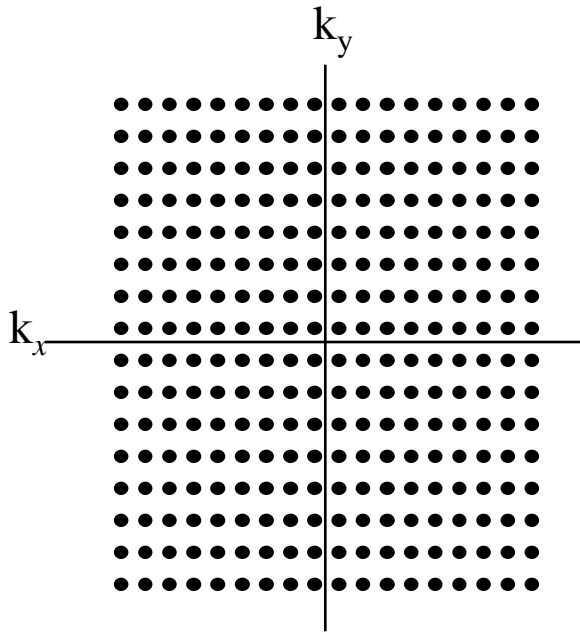
- Acceleration factor of 2-collect  $N/2$  samples

$$SNR = \frac{\sum_{j=1}^N A}{\sqrt{\sum_{j=1}^N \sigma_N^2}} = \frac{\frac{N}{2} A}{\sqrt{\frac{N}{2} \sigma_N^2}} = \frac{A}{\sigma_N} \sqrt{\frac{N}{2}}$$

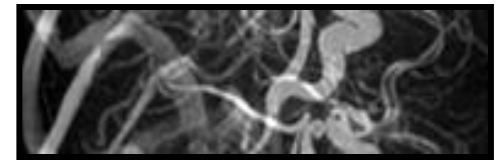
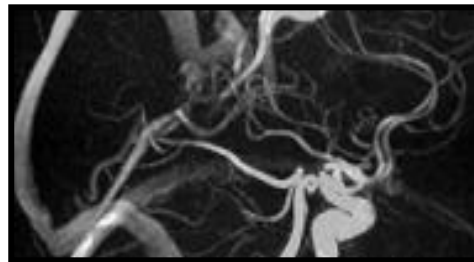
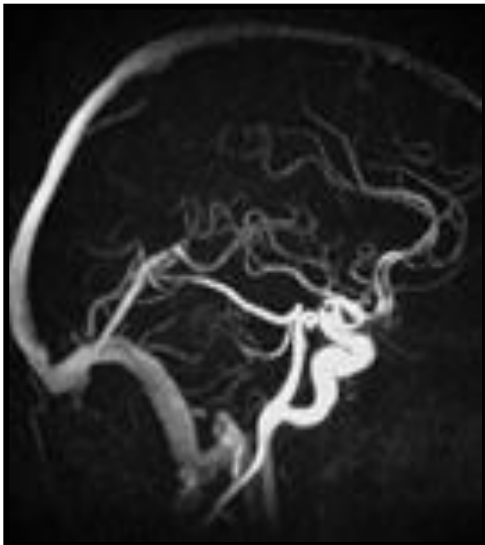
- In general- In an accelerated scan with image resolution fixed

$$SNR \propto \frac{1}{\sqrt{R}}$$

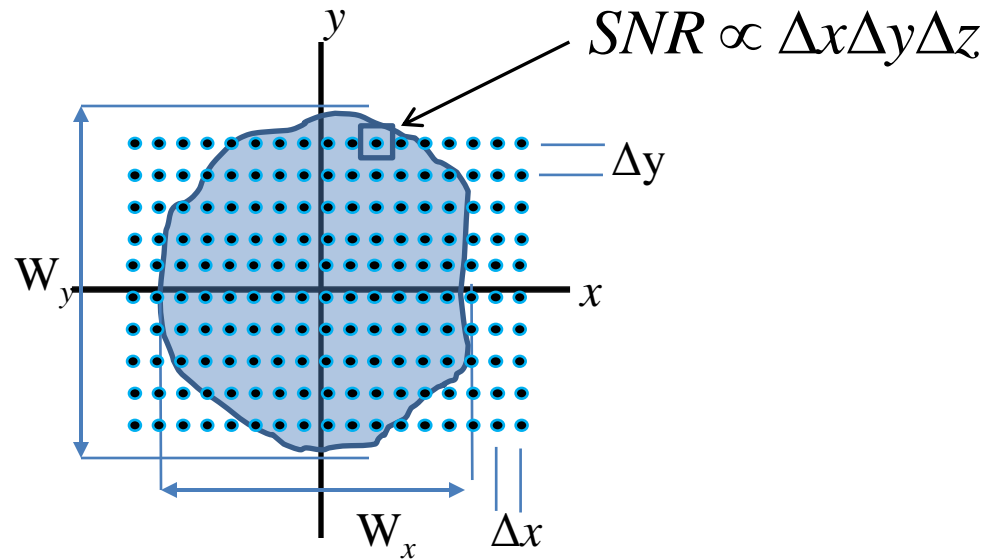
# Undersampling (Rectilinear Acquisition)



$1/2$  encodes – 30% lower SNR     $1/4$  encodes -50% lower SNR



# SNR -When Resolution is not fixed



$$SNR \propto \Delta x \Delta y \Delta z \sqrt{\frac{N}{R}} F(M(x, y, z), T1, T2)$$

SNR – Resolution Vs Acquired samples Vs Acceleration factor

# To summarize..

Acquired data	↓	Resolution	↔	SNR	↓	Acceleration	↑
Acquired data	↑	Resolution	↔	SNR	↑	Acceleration	↓
Acquired data	↑	Resolution	↑	SNR	↓	Acceleration	↓
Acquired data	↓	Resolution	↑	SNR	↓↓	Acceleration	↑
Acquired data	↓	Resolution	↓	SNR	↑	Acceleration	↑

- Increasing intrinsic SNR can allow for greater acceleration factor

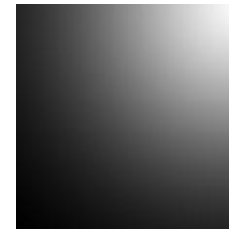
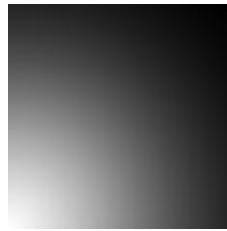
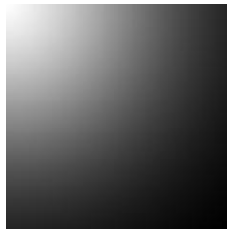
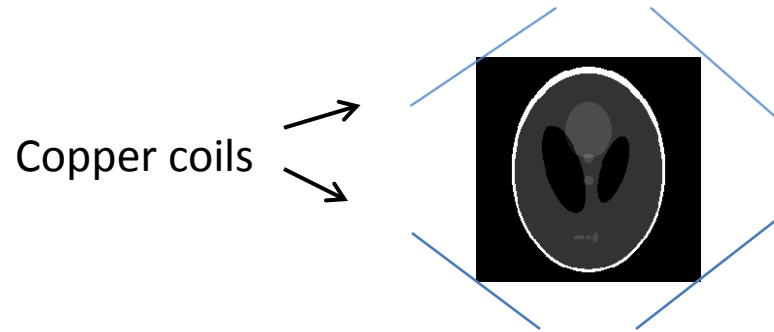
# Parallel Imaging, Compressed Sensing and Aliased k-space acquisitions



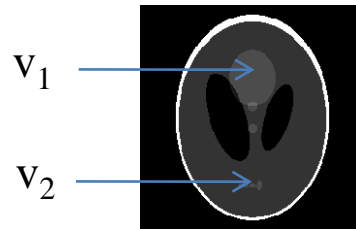
# Parallel Imaging

Acquired data	↓	Resolution	↔	SNR	↓	Acceleration	↑
Acquired data	↓	Resolution	↑	SNR	↓↓	Acceleration	↑

# Impact of coil sensitivity profiles



# Impact of coil sensitivity profiles



# Parallel Imaging-SENSE<sup>1</sup>

$$S_{(\gamma, \rho)} = s_{\gamma}(r_{\rho})$$

$\gamma^{\text{th}}$  Receiver sensitivity value at the  $r_{\rho}$  location

$$U = (S^H S)^{-1} S^H$$

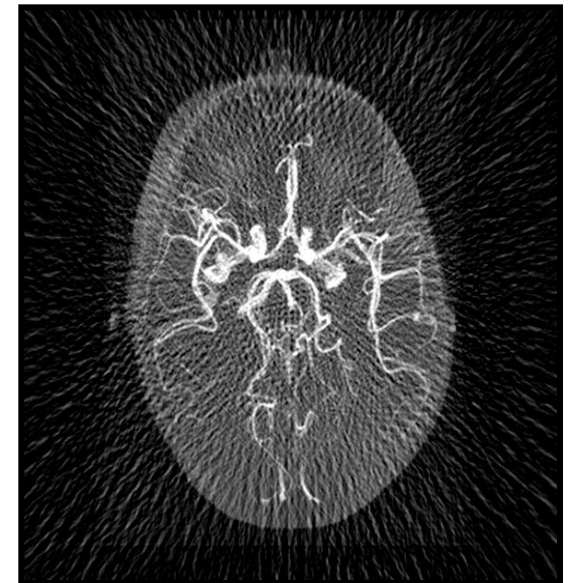
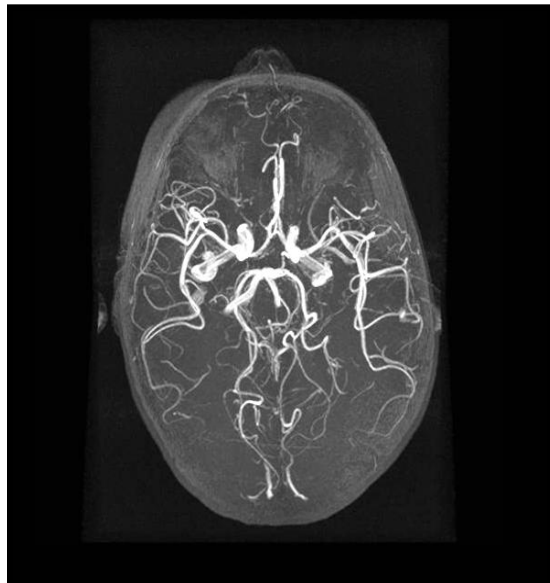
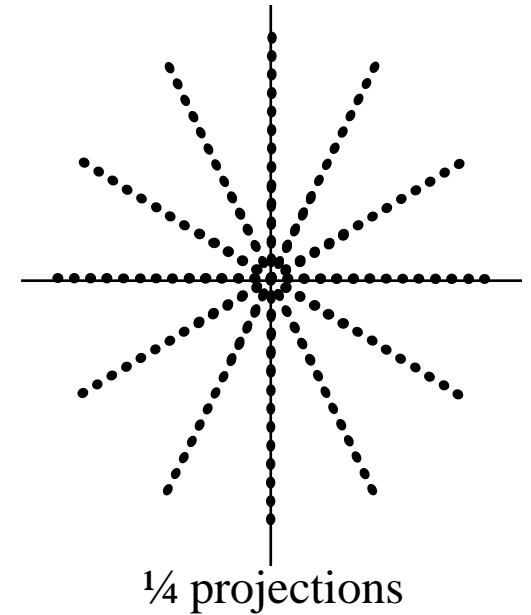
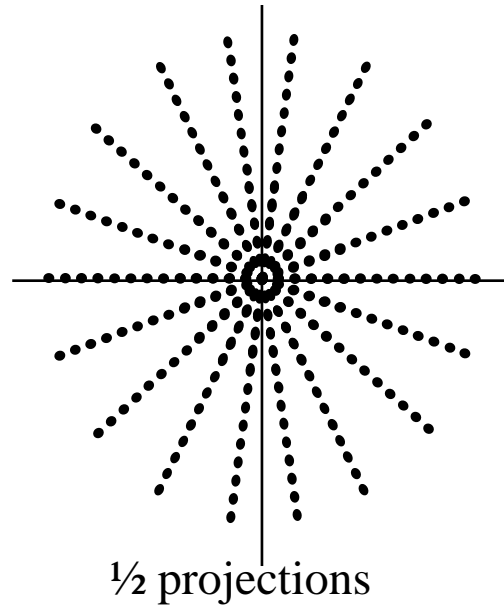
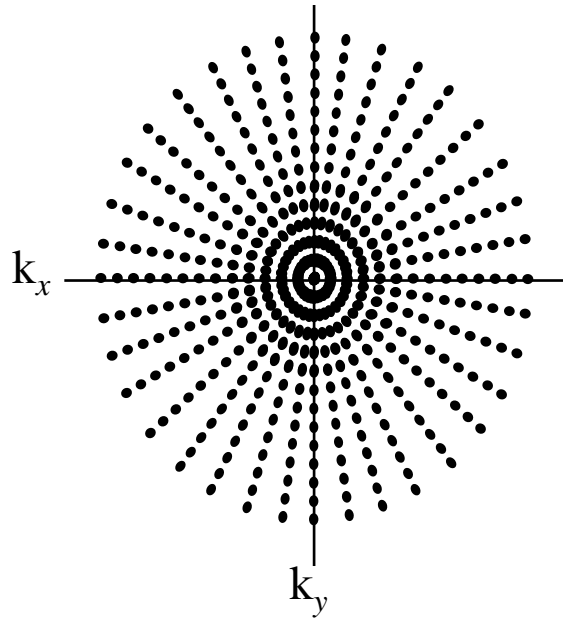
Matrix that can restore the original voxels

$$V = Ua$$

The equation to solve to restore original image

These equations work only for Rectilinear acquisitions!

# What about for Radial Sampling ?



# SENSE in more general terms..

$$F(k_x, k_y) = \iint M(x, y) \overbrace{S_l(x, y)} e^{-i2\pi k_y y} e^{-i2\pi k_x x} dx dy$$



$$E_{(l, \rho)} = e^{jk_k r_\rho} s_l(r_\rho)$$



$$E v = m$$

*m* - Undersampled k-space data



$$v = (E^h E) E^h m$$

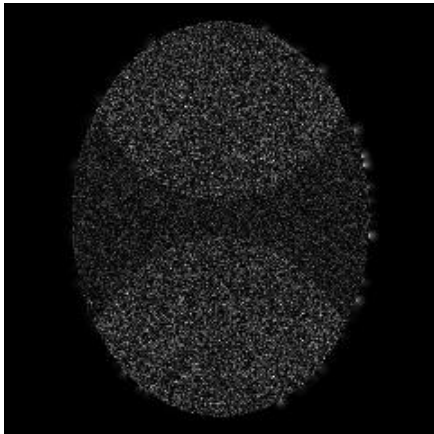
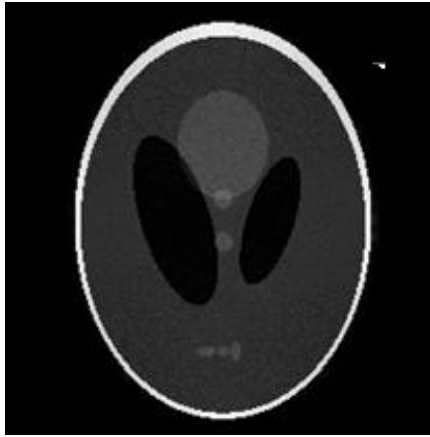
*v* - Unaliased image

# Noise propagation in SENSE

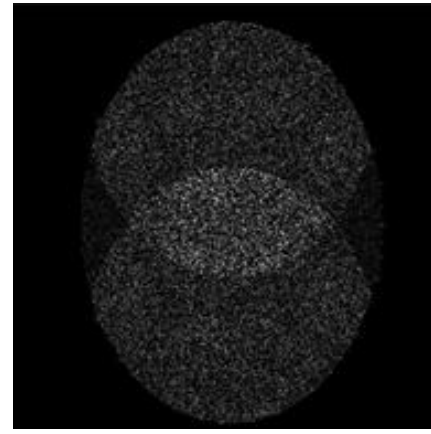
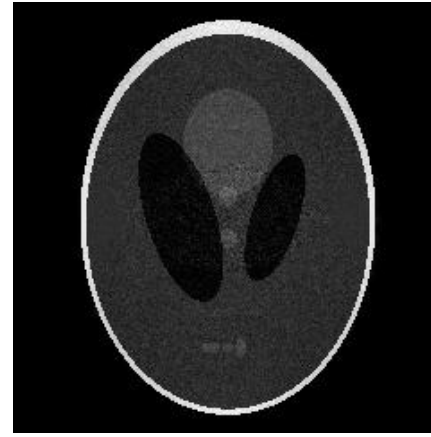
- SNR loss during acquisition  $SNR \propto \frac{1}{\sqrt{R}}$

During reconstruction...

R=2



R=3

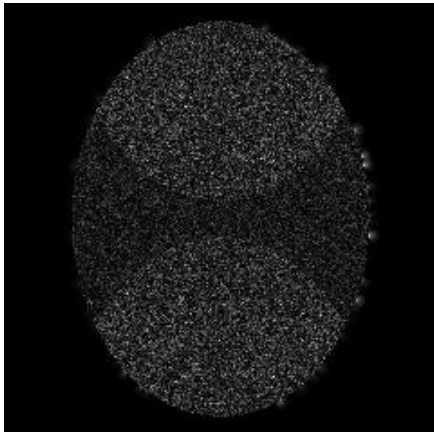
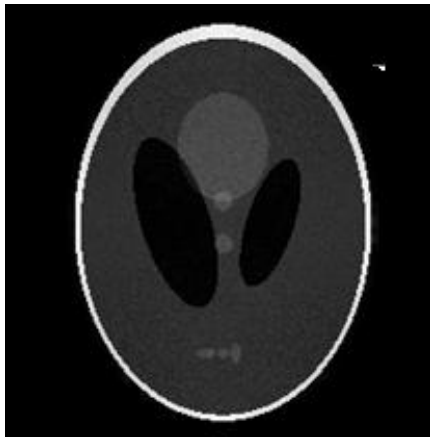


# Noise propagation in SENSE

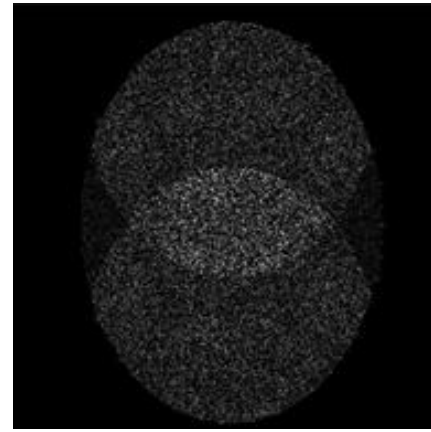
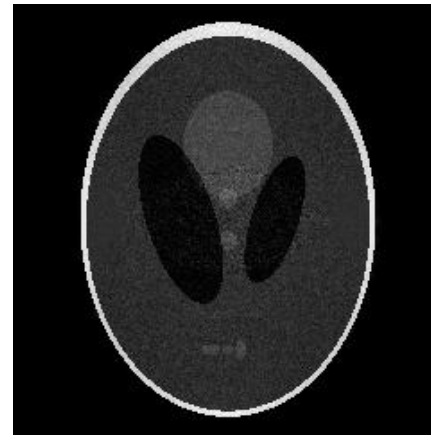
- SNR loss during acquisition  $SNR \propto \frac{1}{\sqrt{R}}$

During reconstruction...

R=2



R=3



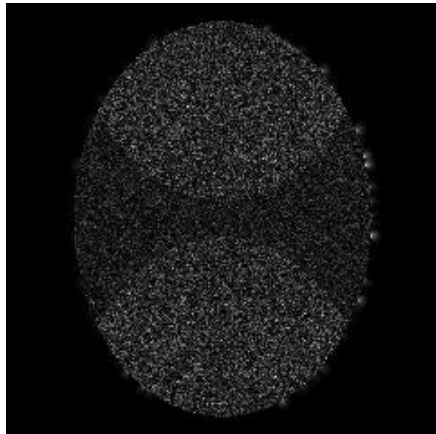
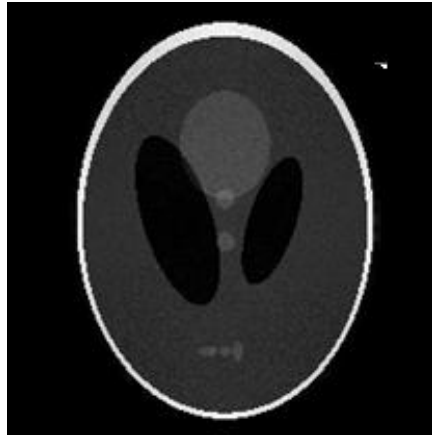


# Noise propagation in SENSE

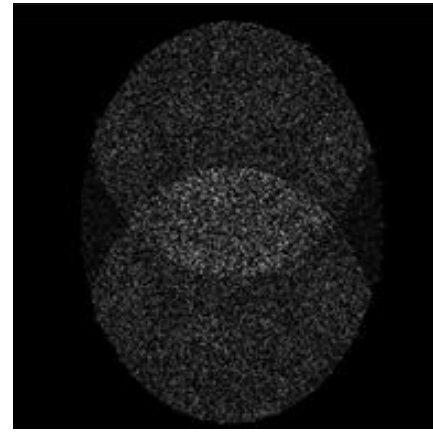
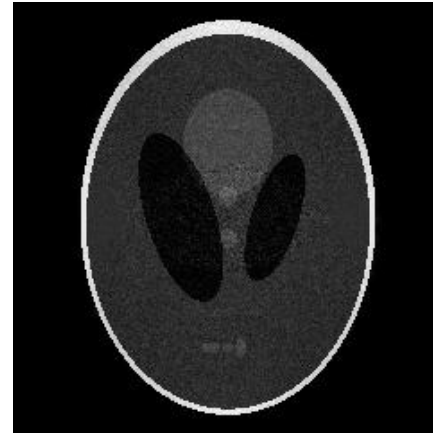
- SNR loss during acquisition  $SNR \propto \frac{1}{\sqrt{R}}$

During rectilinear data reconstruction...

R=2



R=3



# The noise propagation model

Noise in the  $\gamma^{\text{th}}$  channel:  $n_\gamma(t) = \sum_\tau \alpha_{\gamma,\tau} \xi_\tau(t)$       Modeled as weighted sum of individual noise sources

Noise variance in the  $\rho^{\text{th}}$  voxel in rectilinear acquisition:

$$\sigma_\rho^2 = \sum_\tau \sigma_\tau^2 \left| \sum_{l,k} U_{\rho,(l,k)} \alpha_{l,\tau} \right|^2 \quad \text{and} \quad U = (S^H S)^{-1} S^H$$

Rearranging the variance equation, we get the following noise matrix ( $X$ ):

$$X = U \psi U^H \quad \text{where} \quad \psi_{l,l'} = \sum_\tau \sigma_\tau^2 \alpha_{l,\tau} \alpha_{l',\tau}^*$$

Substitute  $U$  to get the following:  $X = (S^H \psi^{-1} S)^{-1}$

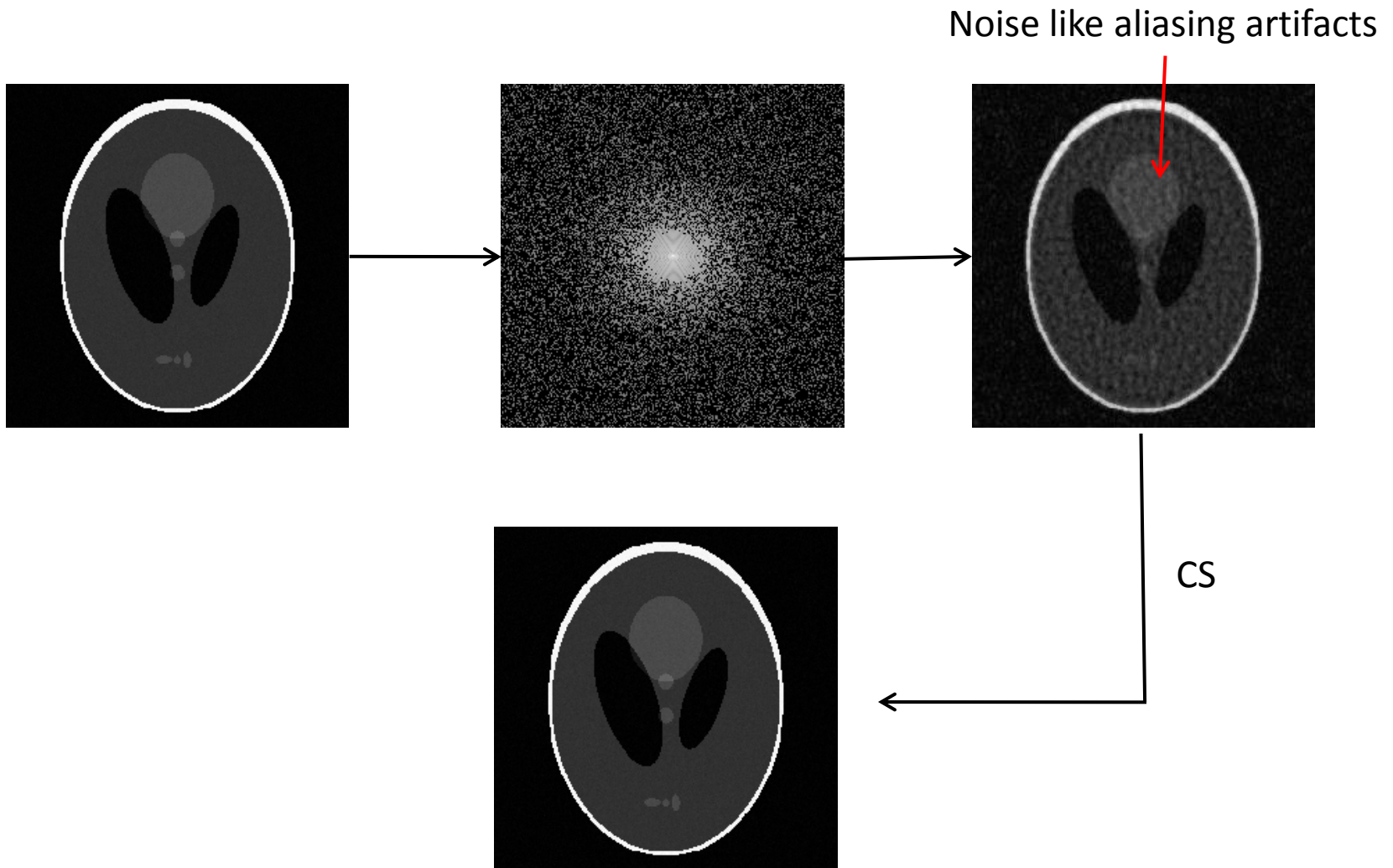
# Limitations of Parallel Imaging

- Guaranteed SNR loss during acquisition and reconstruction
- Limits on the maximum acceleration
- Demanding Clinical applications need more acceleration

# Compressive Sensing

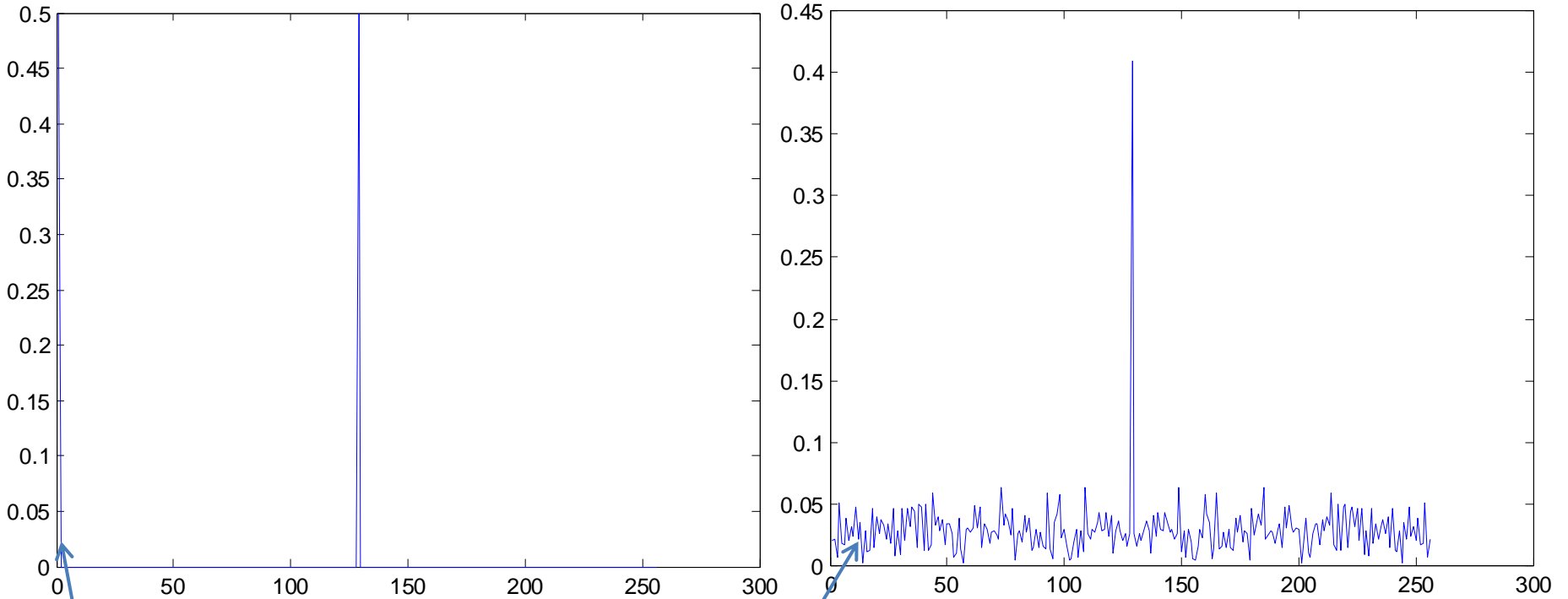
Acquired data	↓	Resolution	↔	SNR	↓	Acceleration	↑
Acquired data	↓	Resolution	↑	SNR	↓↓	Acceleration	↑

# Compressed Sensing example



# Idea behind random under-sampling?

1D Imaging example



Regular under-sampling factor of 2

Random under-sampling factor of 2

Regular aliasing artifacts

Noise like aliasing artifacts

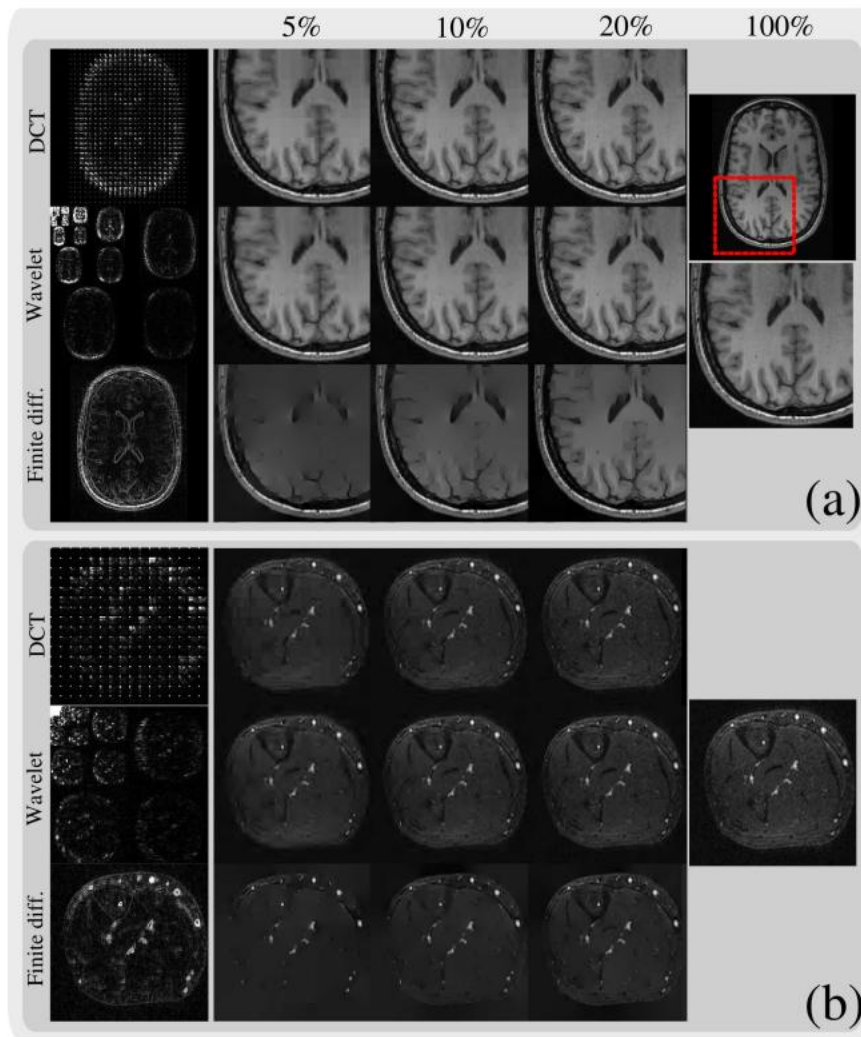
# CS reconstruction framework

$$\begin{array}{lll} \text{Minimize} & \|\Psi v\|_{l_1} & \Psi\text{-Sparse representation} \\ \text{s.t} & \|Ev - m\|_{l_2} \leq \varepsilon & \text{Data consistency} \end{array}$$

In MRI...TV transformation is always used..

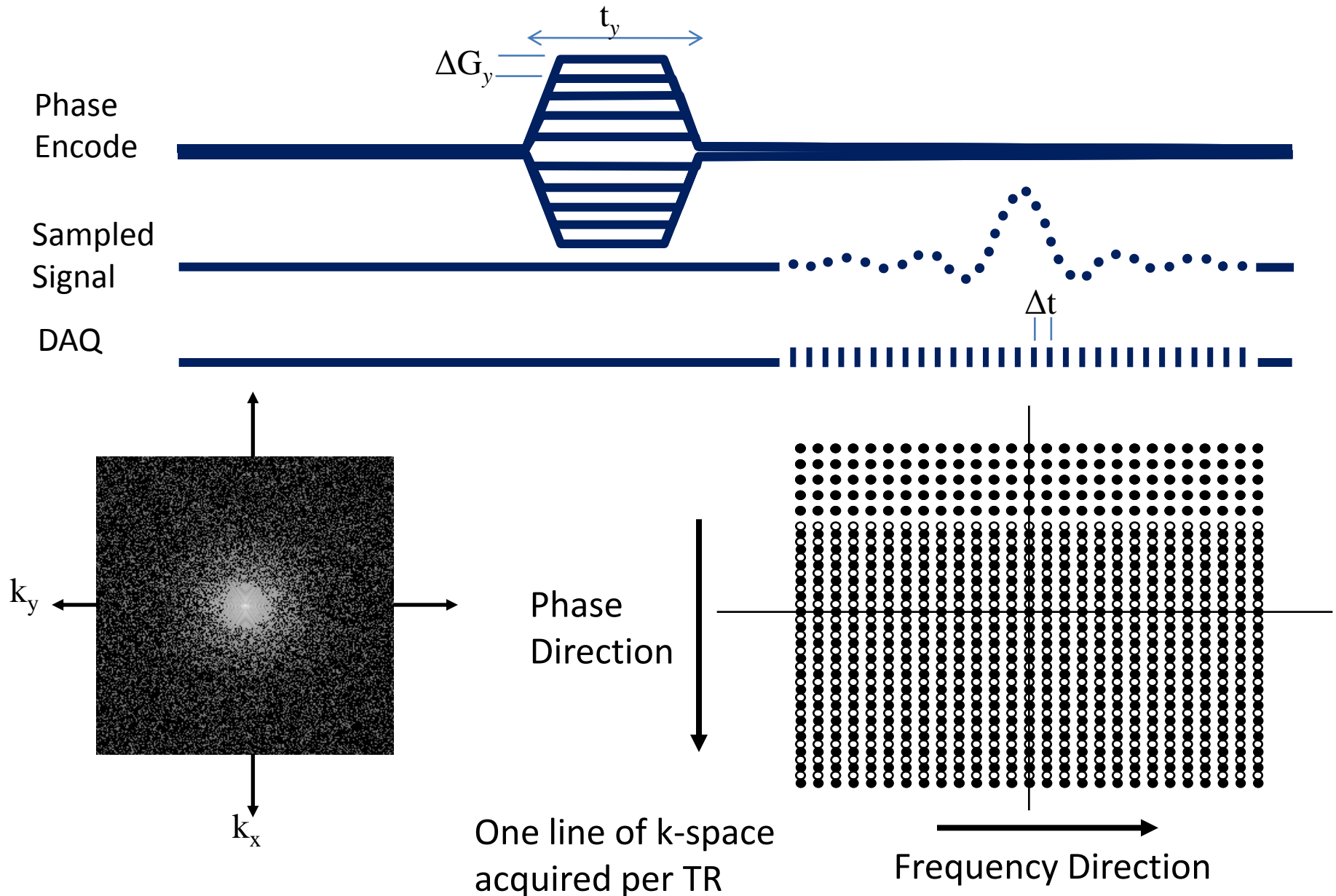
$$\begin{array}{lll} \text{Minimize} & \|\Psi v\|_{l_1} + \eta TV(v) & \text{TV-Total variation} \\ \text{s.t} & \|Ev - m\|_{l_2} \leq \varepsilon & \text{Data consistency} \end{array}$$

# Sparse representation of MRI Images<sup>3</sup>

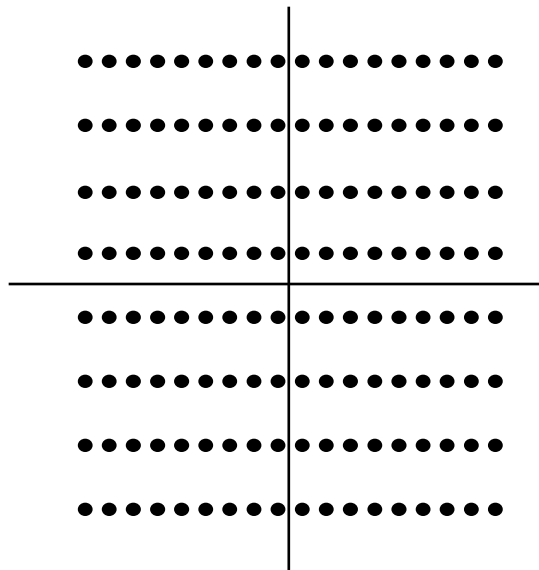
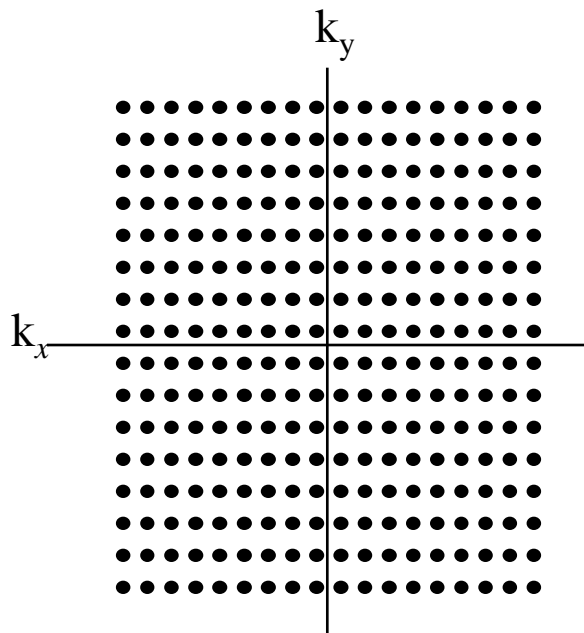




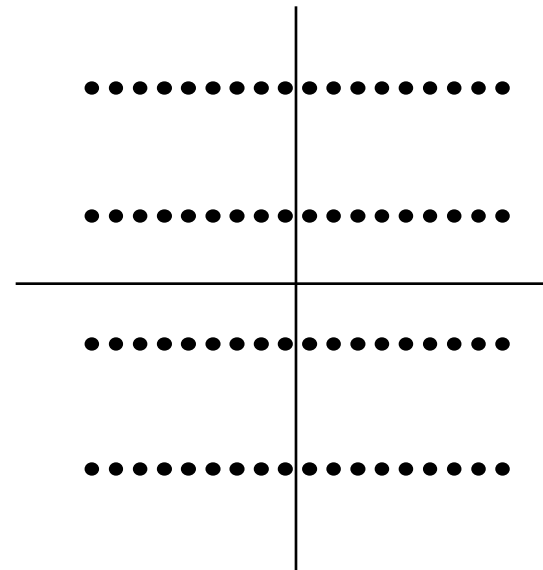
# 2D Rectilinear acquisition with CS?



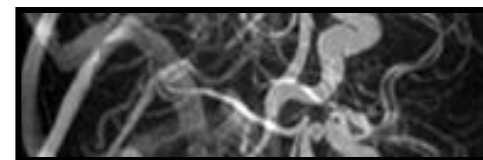
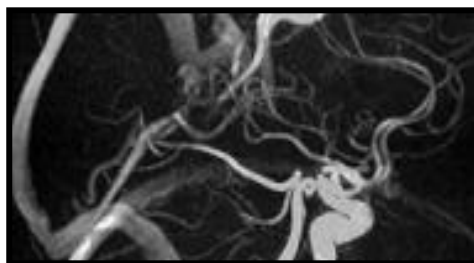
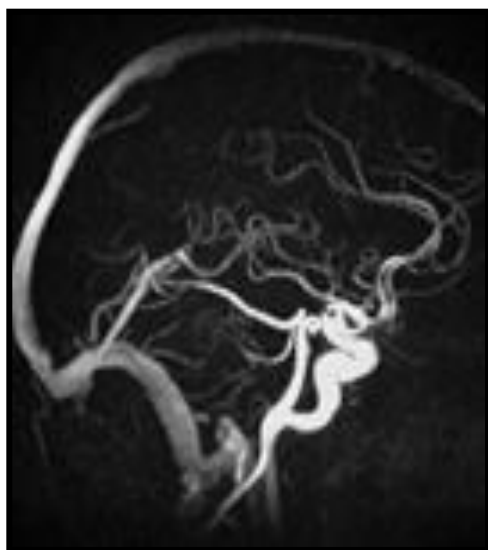
# Conventional Rectilinear undersampling?



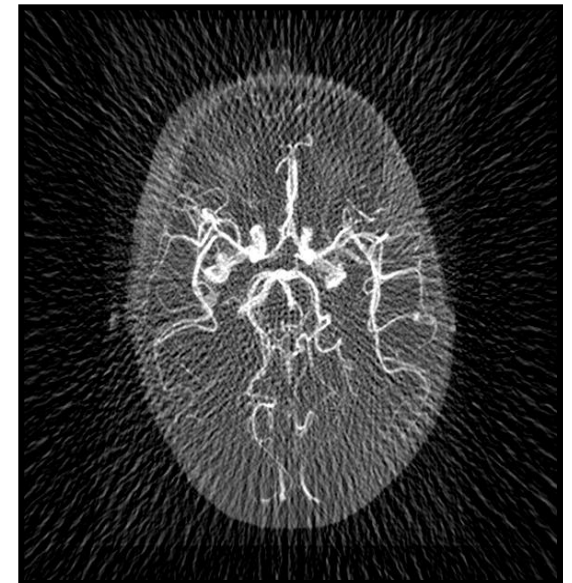
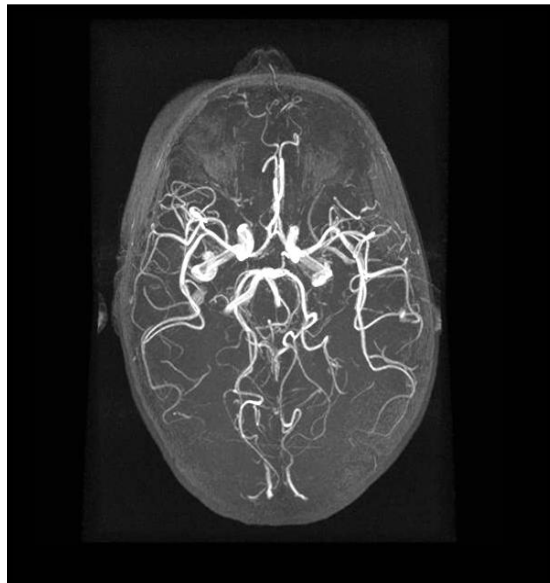
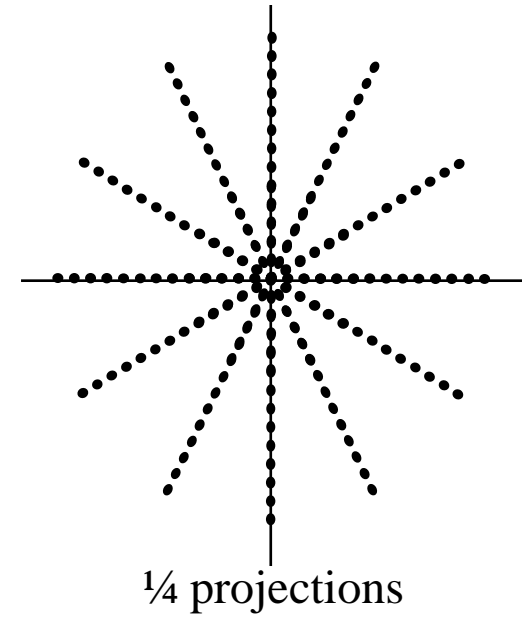
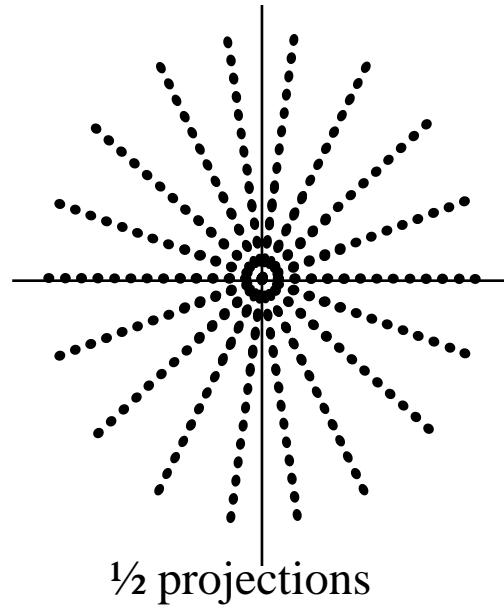
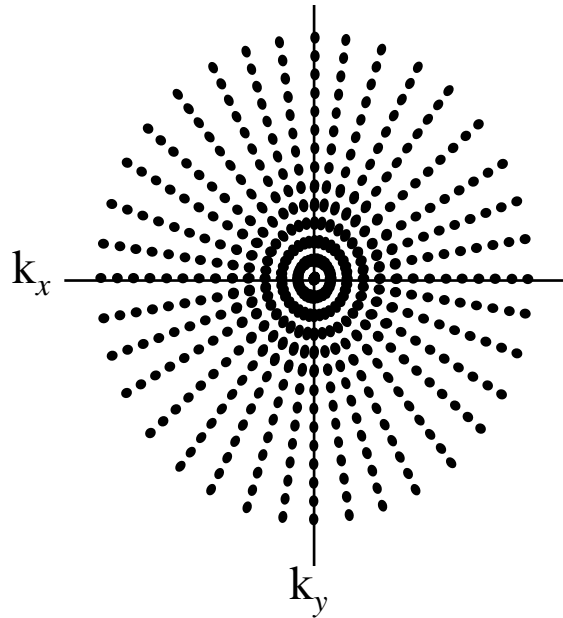
$\frac{1}{2}$  encodes



$\frac{1}{4}$  encodes



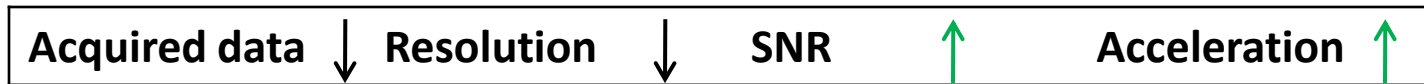
# What about for Radial Sampling ?



# Limitations of Compressed Sensing

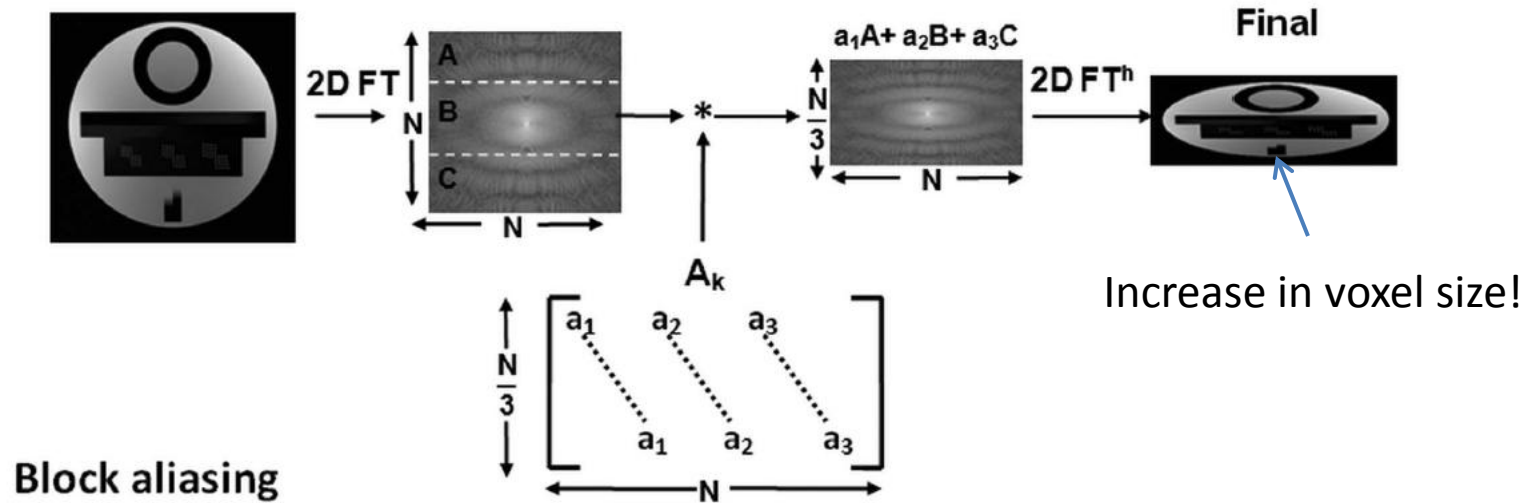
- Guaranteed SNR loss during acquisition
- Limits on the maximum acceleration
- Texture loss!
- Not applicable for all acquisitions (3D, Dynamic imaging most suitable)
- Compatibility with multi-receiver acquisitions not clear
- Visually improved or real restoration?

# Aliased k-space acquisitions



Can resolution be restored during reconstruction?

# An example of overlapping k-space<sup>4</sup>



- All data acquired through overlapping
- Acceleration achieved due to parallel acquisition of k-space data
- Acquisition SNR increase due to voxel size increase

# Aliased k-space acquisitions

$$F(k_x, k_y) = \iint M(x, y) S_l(x, y) \left( a_1 e^{-i2\pi k_{y1} y} + a_2 e^{-i2\pi k_{y2} y} + a_3 e^{-i2\pi k_{y3} y} \right) e^{-i2\pi k_x x} dx dy$$

$$E_{(l, \rho)} = \left( a_1 e^{jk_{k1} r_\rho} + a_{21} e^{jk_{k2} r_\rho} + a_3 e^{jk_{k3} r_\rho} \right) s_l(r_\rho)$$

$$E v = m \quad m - \text{Aliased } k\text{-space data}$$

$$v = \left( E^h E \right) E^h m \quad v - \text{Unaliased image}$$

# SNR gain in “aliased” acquisitions

- Consider an Impulse object centered at origin

$$SNR_{ref} = \frac{\sum_{j=1}^N A}{\sqrt{\sum_{j=1}^N \sigma_N^2}} = \frac{NA}{\sqrt{N\sigma_N^2}} = \frac{A\sqrt{N}}{\sigma_N}$$

SNR obtained  
By simply adding samples

---

$$SNR_k = \frac{|A| \sqrt{a_1^2 + a_2^2 + a_3^2} \sqrt{N} \sqrt{3}}{\sigma_N}$$

$$\frac{SNR_k}{SNR_{ref}} = \sqrt{a_1^2 + a_2^2 + a_3^2} \sqrt{3}$$



# Limitations of Aliased k-space acquisitions

- Poorly conditioned encoding matrix  $E$
- Sensitive to coil geometry
- Required several receivers
- Increase TR duration

# To summarize..

- Parallel Imaging in commercial scanners as a clinical imaging product
- Compressed Sensing – work in progress..
- Aliased k-space acquisitions – Undergoing clinical investigations..
- Use of receiver sensitivities proven technology
- Sparsity constraints unpredictable – Sampling patterns need development