

Lecture 11 Image Reconstruction

(Problem Definition)

Yuyao Zhang PhD

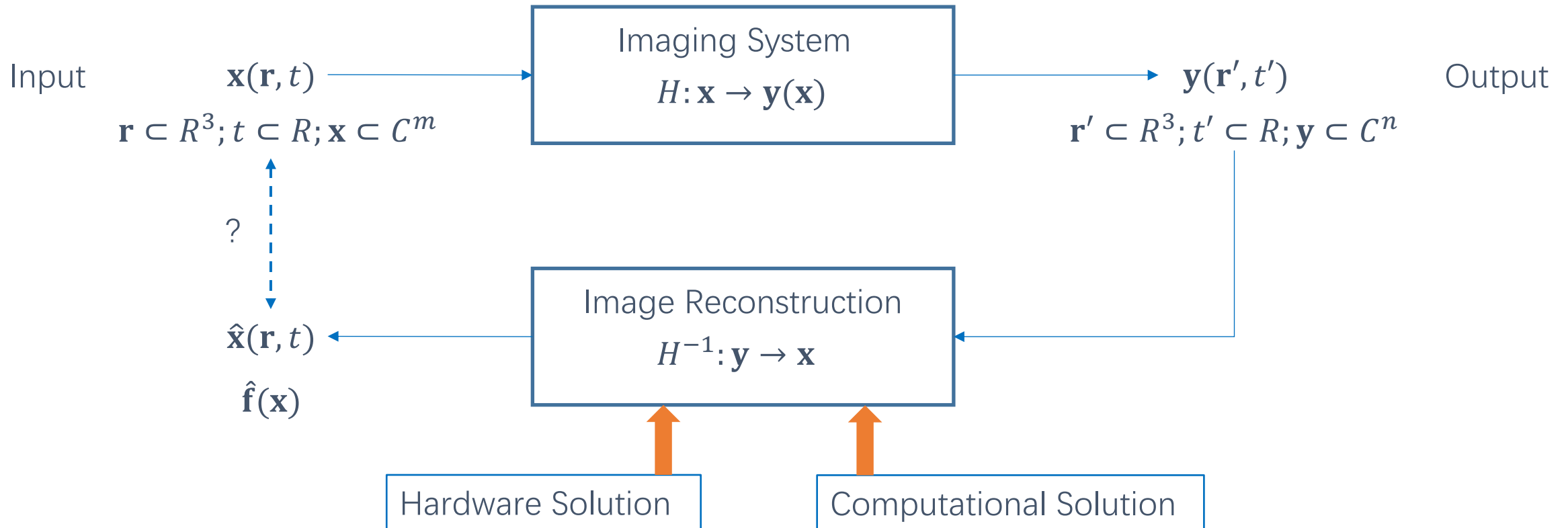
zhangyy8@shanghaitech.edu.cn

SIST Building-3 420

Outline

- Projection and back-projection
- Radon transform
- Fourier-Slice Theorem
- Filtered back-projection

Image Reconstruction: A System's View



1. Is H even known?

2. What's the evaluation criteria for reconstruction?



Reconstruction Evaluation Criteria

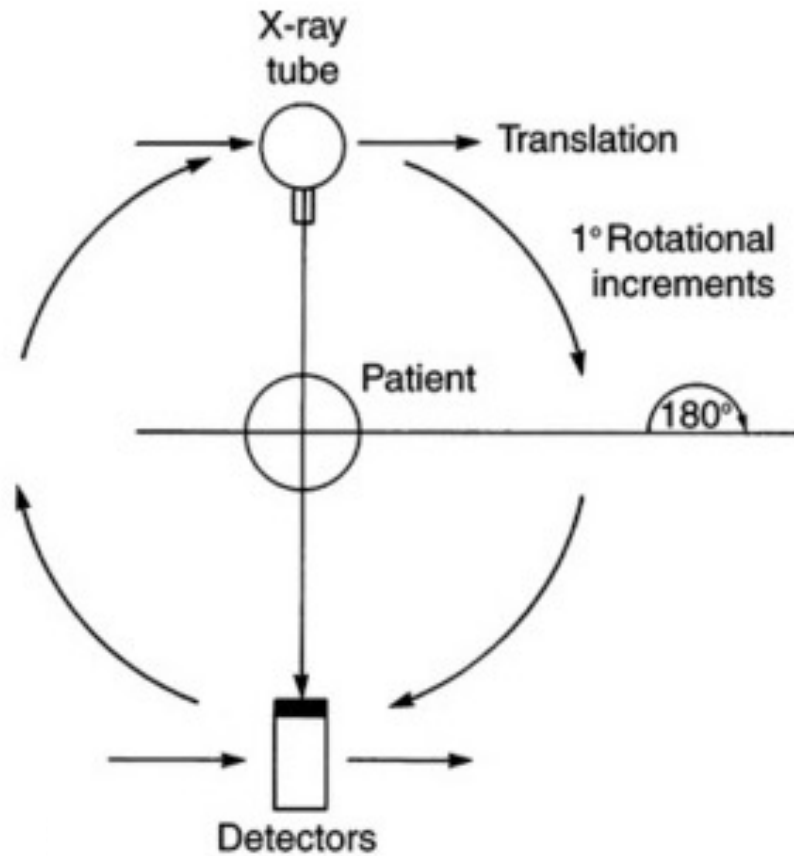
- Least Mean Square Errors: needs to know the ground truth. Typically done in numerical simulation;
- Point Spread Functions: evaluating image blurring;
- Signal-to-Noise (SNR): sensitivity to noise amplification;
- Contrast-to-Noise (CNR) ratio: e.g. sampling density correction in gridding;
- Ultimate test: real applications, e.g. by clinical practice.



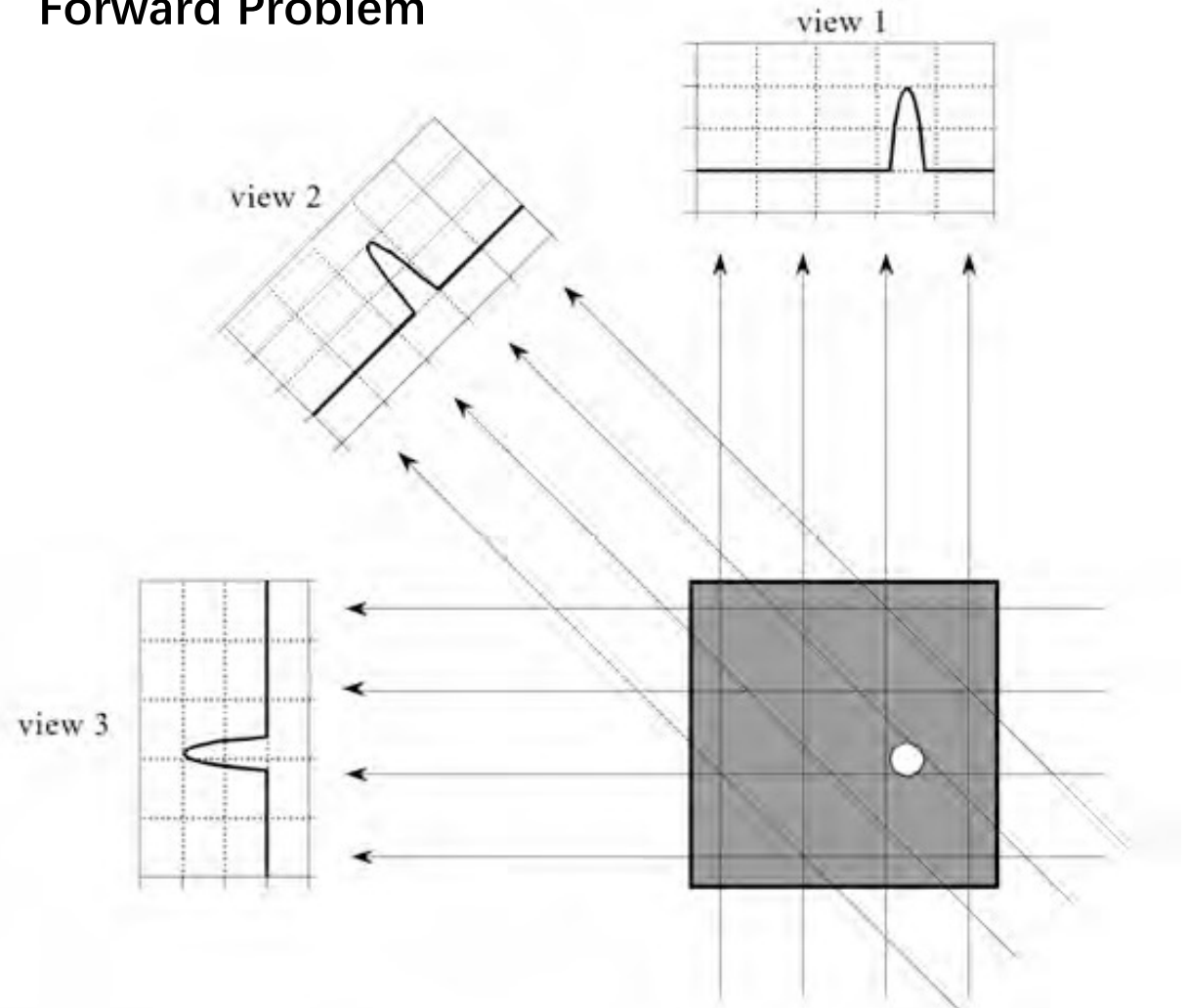
Computational Solution



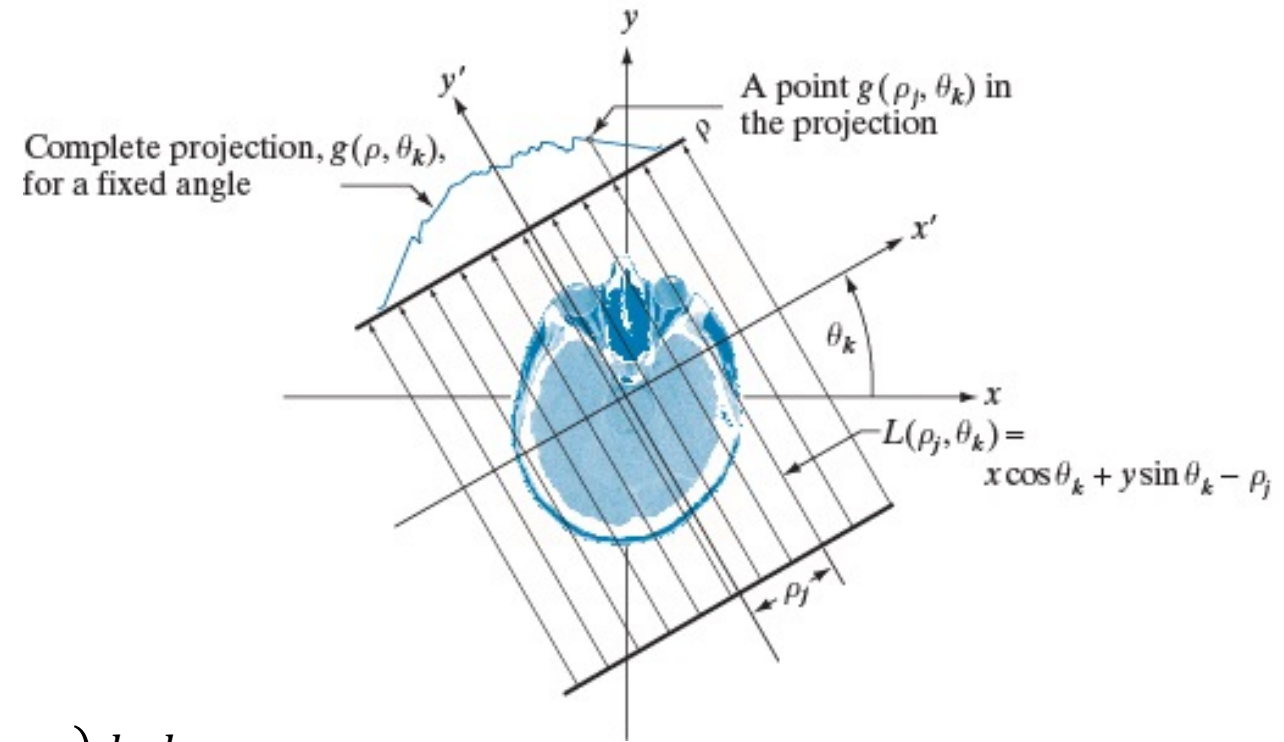
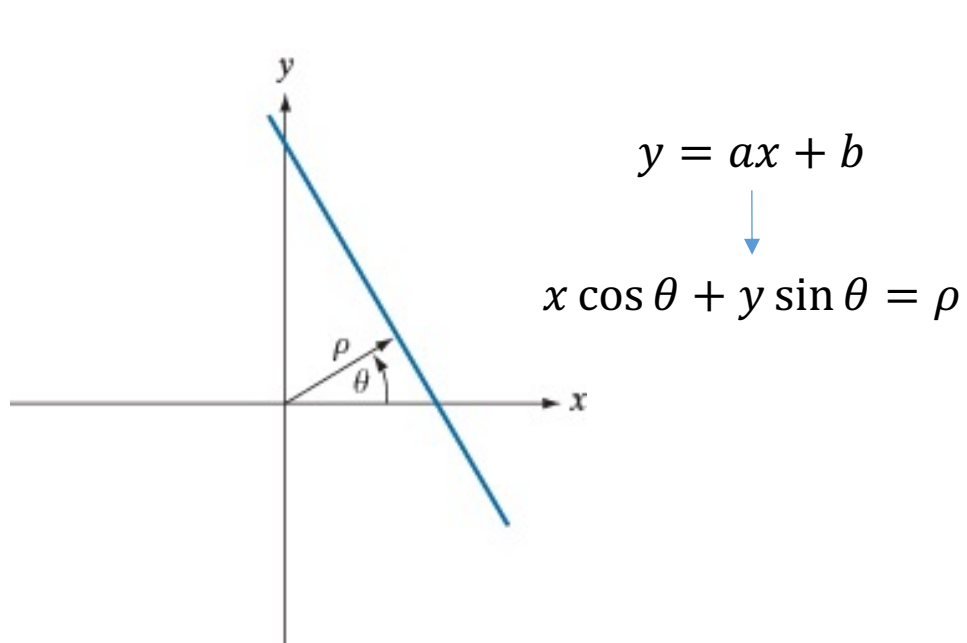
Computed Tomography



Forward Problem



Radon transform



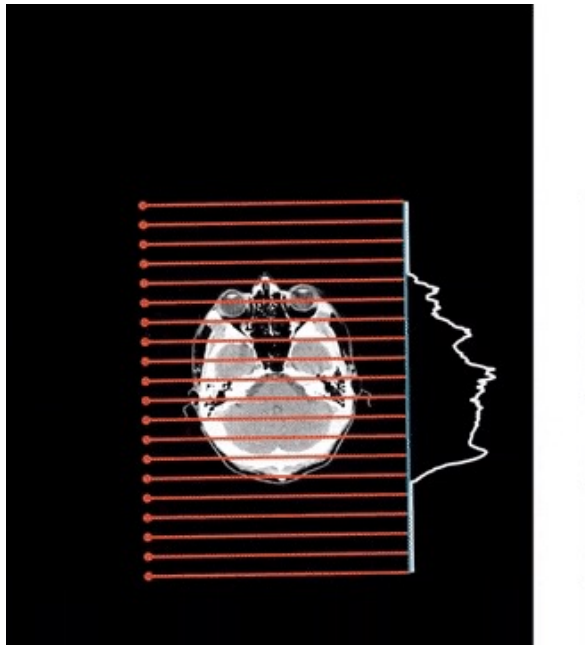
$$g(\rho_j, \theta_k) = \int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} f(x, y) \delta(x \cos \theta_k + y \sin \theta_k - \rho_j) dx dy$$

$$g(\rho, \theta) = \int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} f(x, y) \delta(x \cos \theta + y \sin \theta) dx dy$$

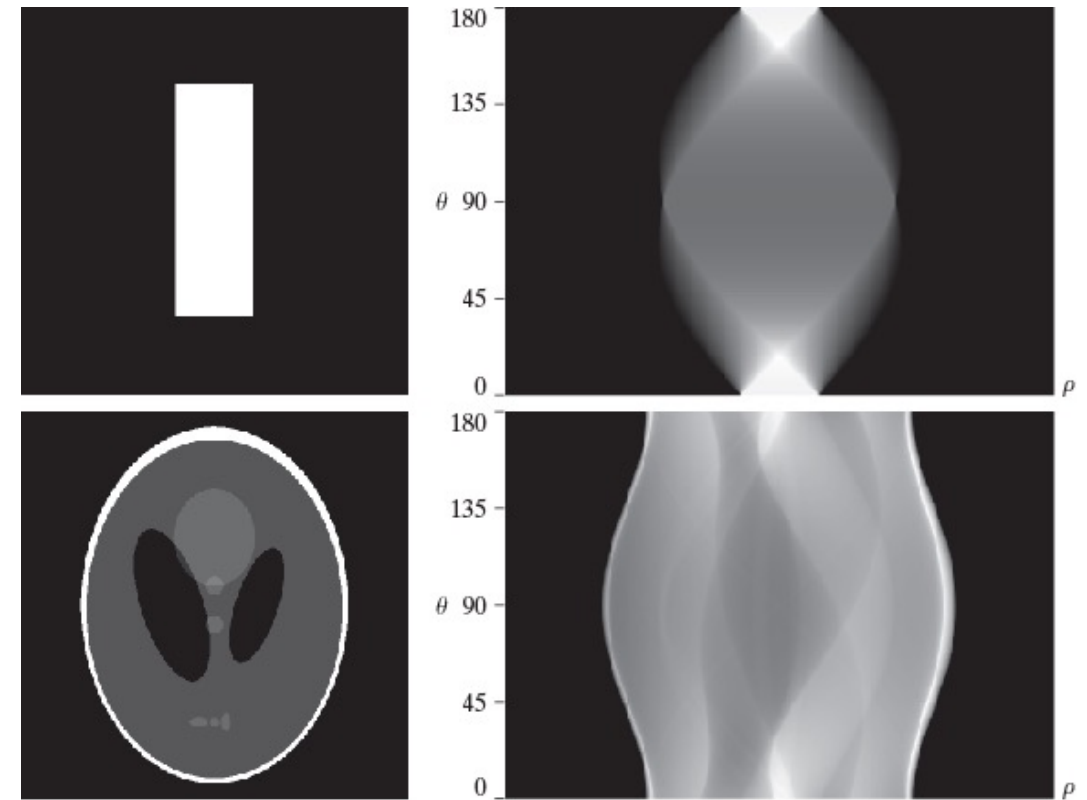
$$g(\rho, \theta) = \sum_{x=0}^{M-1} \sum_{y=0}^{N-1} f(x, y) \delta(x \cos \theta + y \sin \theta - \rho)$$

Sinogram

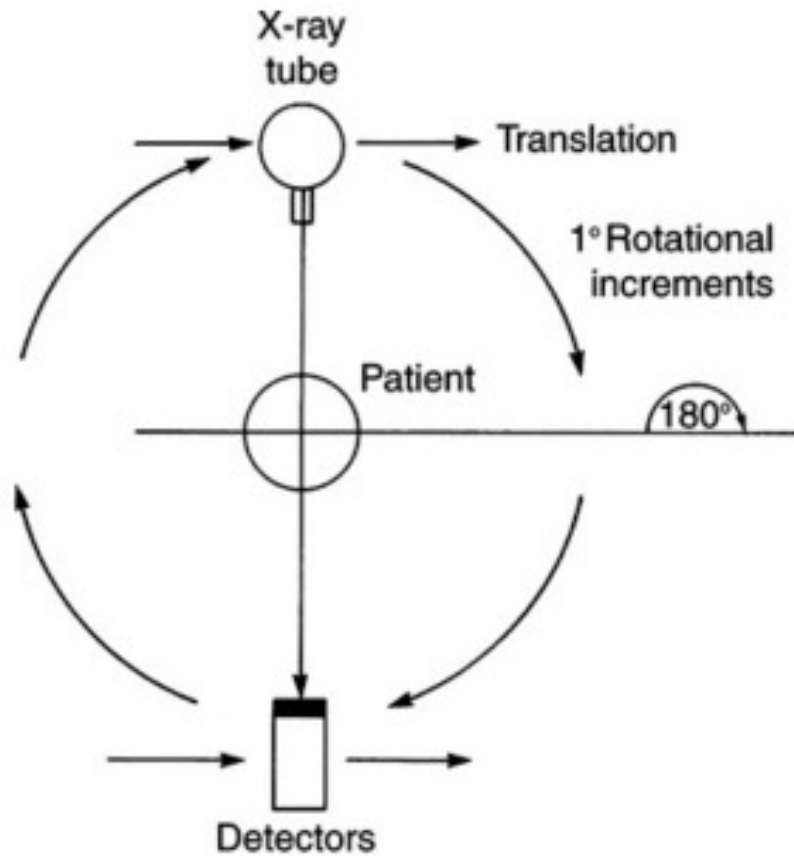
- Radon transform $g(\rho, \theta)$ is displayed as an image with ρ and θ as rectilinear coordinates



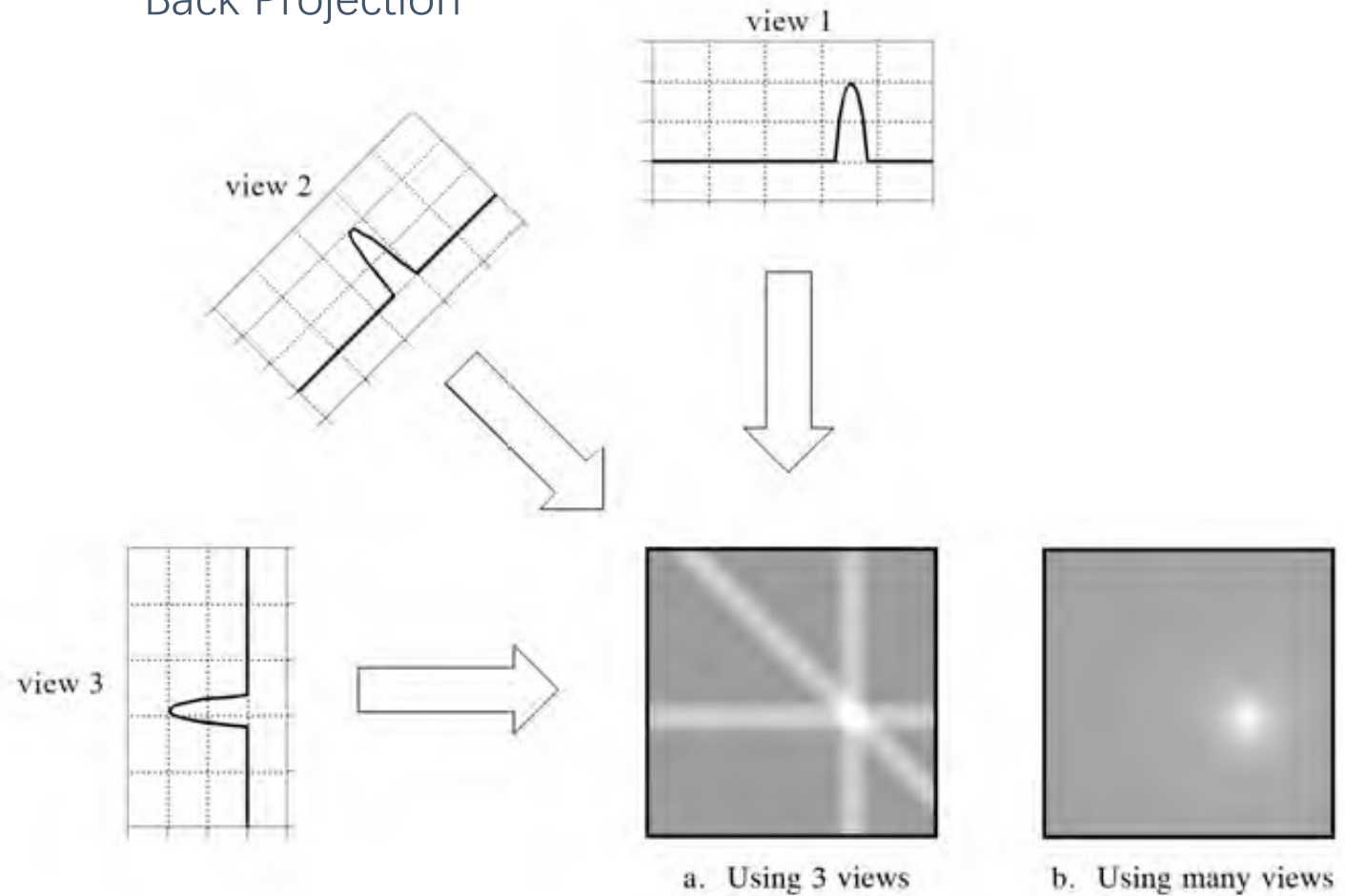
Radon Transform
 $f(x, y) \longrightarrow g(\rho, \theta)$ (Sinogram)



CT: Back Projection in Image Domain



Inverse Problem
Back Projection



Back-projection from Sinogram

- For a fixed value of rotation θ_k :

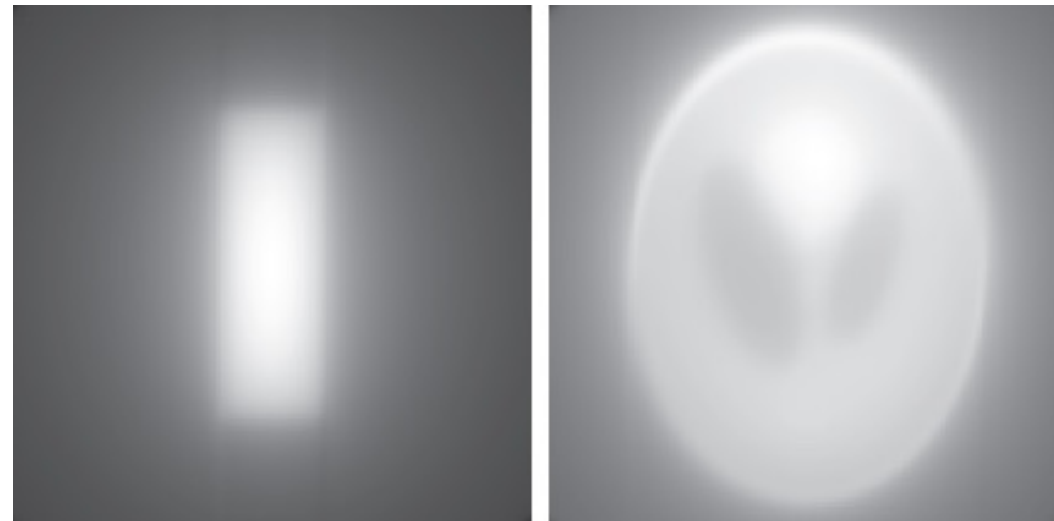
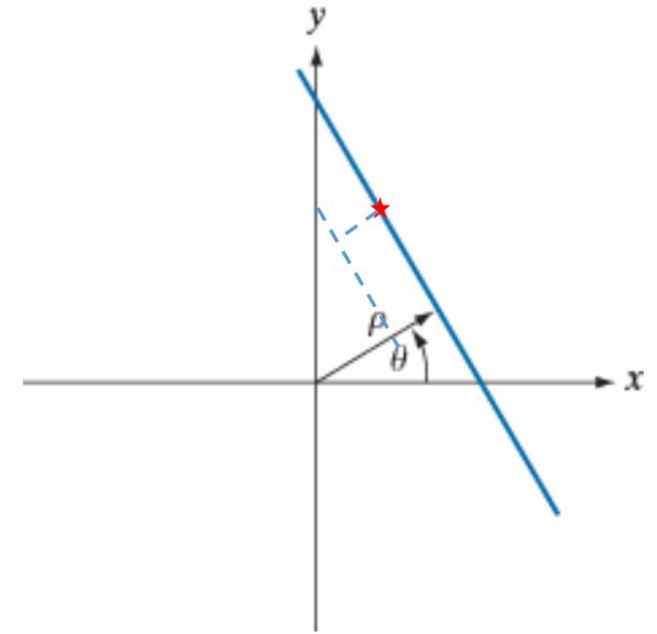
$$f_{\theta_k}(x, y) = g(\rho, \theta_k) = g(x \cos \theta_k + y \sin \theta_k, \theta_k)$$

- Then a single back-projection image obtained at an angle θ is :

$$f_{\theta}(x, y) = g(x \cos \theta + y \sin \theta, \theta)$$

- The reconstructed image is obtained by summing over all the back-projected images:

$$f(x, y) = \sum_{\theta=0}^{\pi} f_{\theta}(x, y)$$



Fourier-Slice Theorem

- The 1D FT of a projection with respect to ρ is:

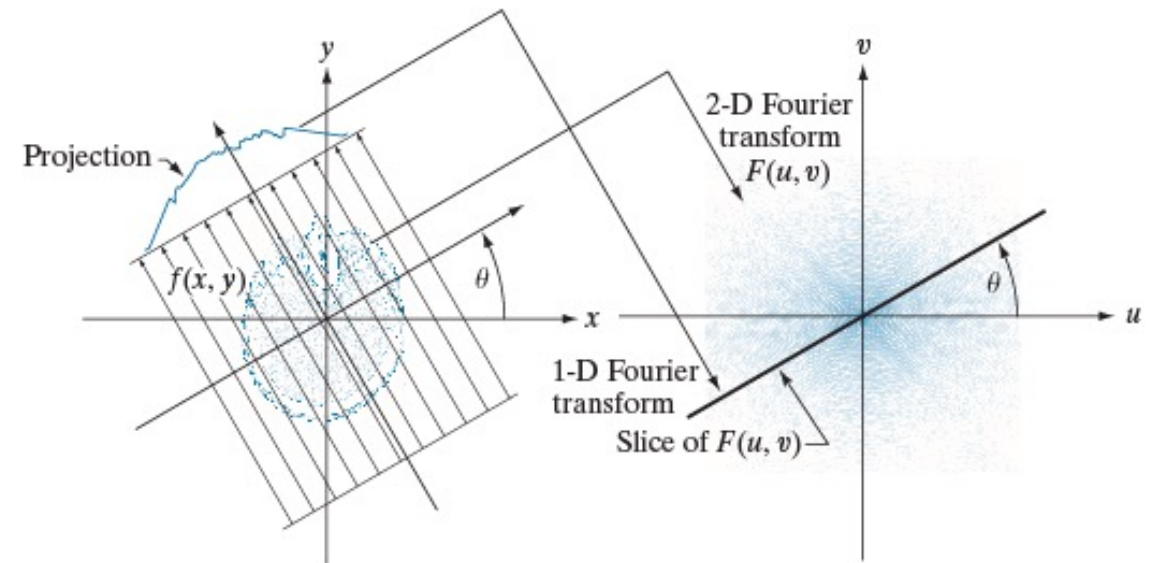
$$G(\omega, \theta) = \int_{-\infty}^{+\infty} g(\rho, \theta) e^{-j2\pi\omega\rho} d\rho$$

- Then

$$\begin{aligned} G(\omega, \theta) &= \int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} f(x, y) e^{-j2\pi\omega(x \cos \theta + y \sin \theta)} dx dy \\ &= \left[\int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} f(x, y) e^{-j2\pi(ux + vy)} dx dy \right]_{u=\omega \cos \theta; v=\omega \sin \theta} \end{aligned}$$

- Therefore

$$G(\omega, \theta) = [F(u, v)]_{u=\omega \cos \theta; v=\omega \sin \theta} = F(\omega \cos \theta, \omega \sin \theta)$$



Filtered back-projection

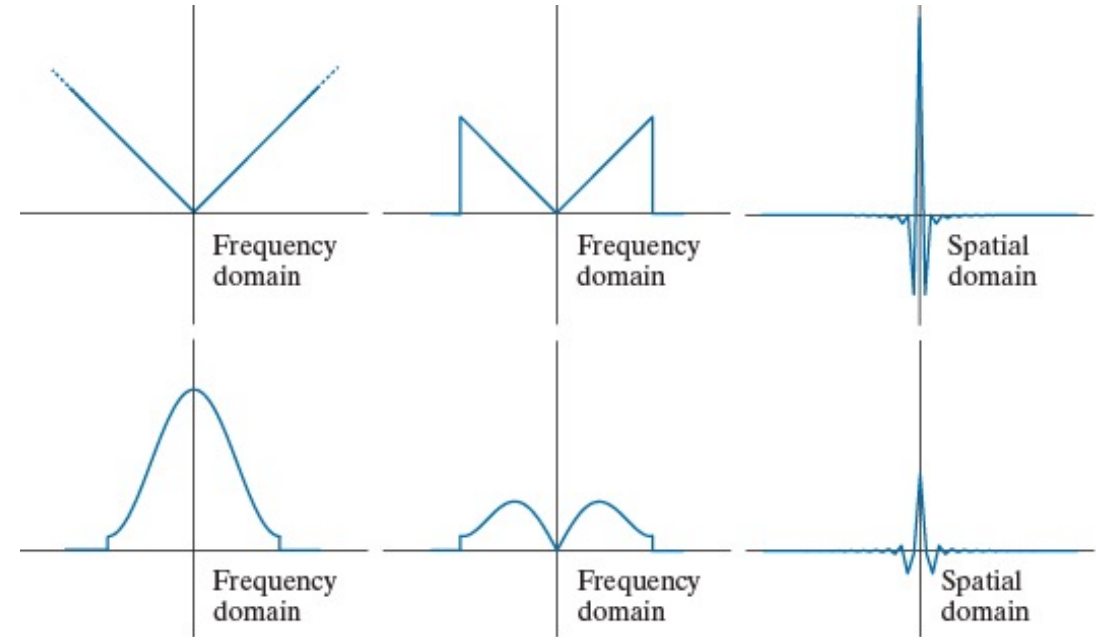
- The 2D IFT of $F(u, v)$ is:

$$\begin{aligned}
 f(x, y) &= \int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} F(u, v) e^{j2\pi(ux+vy)} du dv \\
 &= \int_0^{2\pi} \int_0^{\infty} G(\omega, \theta) e^{j2\pi\omega(x \cos \theta + y \sin \theta)} \omega d\omega d\theta \\
 &= \int_0^{\pi} \left[\int_{-\infty}^{+\infty} |\omega| G(\omega, \theta) e^{j2\pi\omega\rho} d\omega \right]_{\rho=x \cos \theta + y \sin \theta} d\theta
 \end{aligned}$$

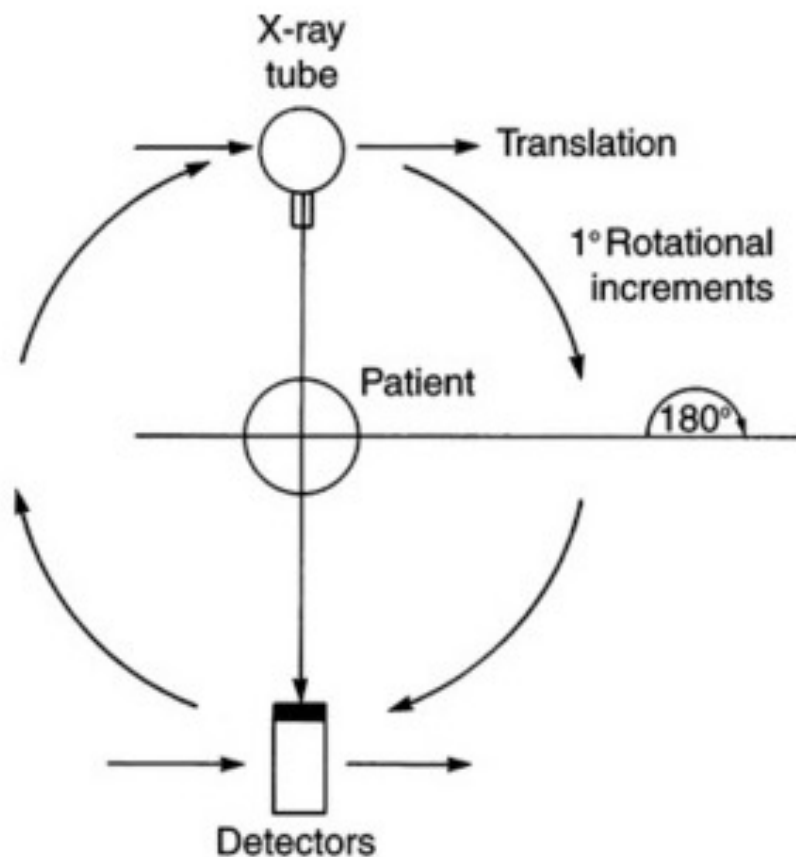
- Back-projection with convolution

$$f(x, y) = \int_0^{\pi} [s(\rho) \otimes g(\rho, \theta)]_{\rho=x \cos \theta + y \sin \theta} d\theta$$

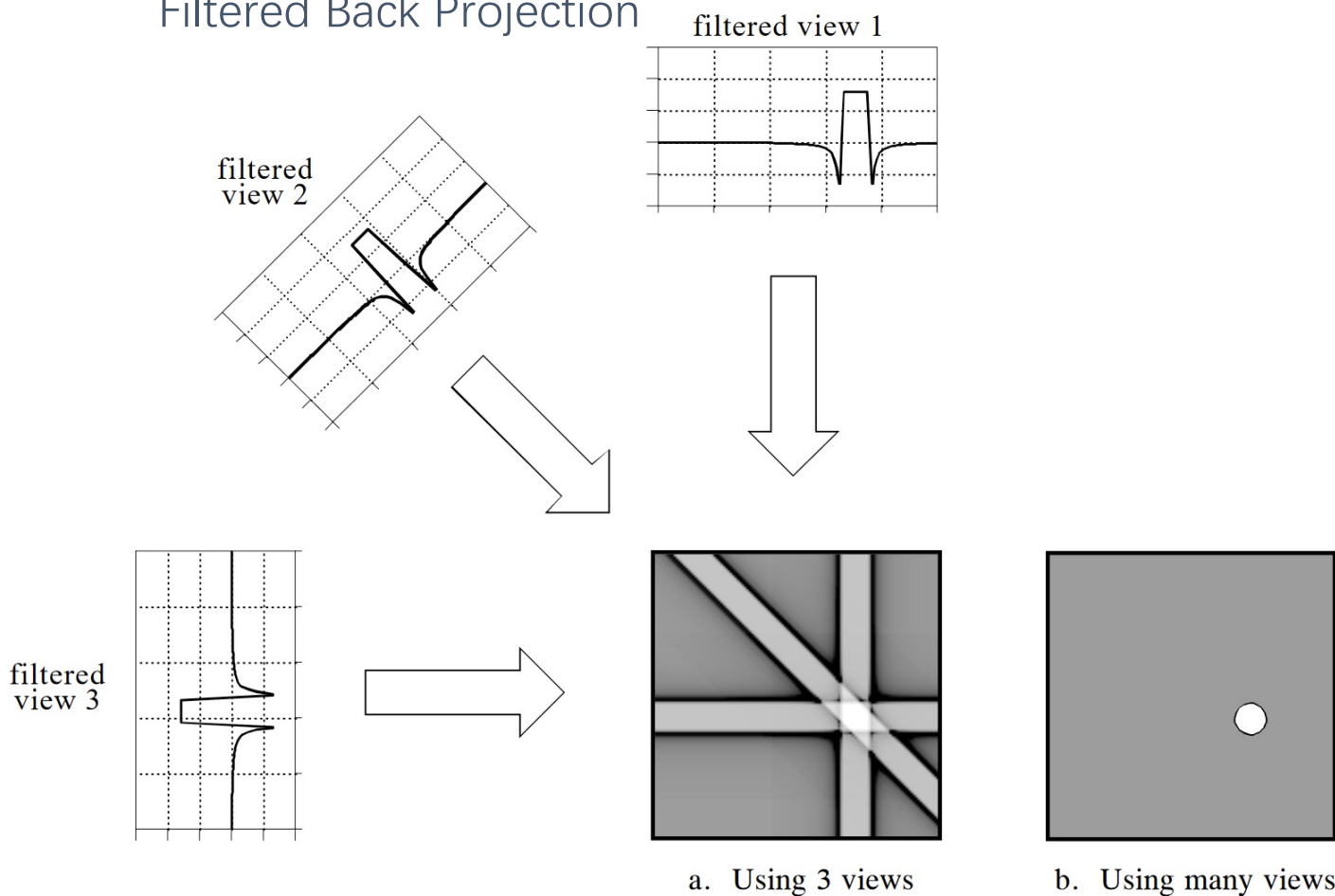
Where $s(\rho) = IFT(|\omega|)$, $g(\rho, \theta) = IFT(G(\omega, \theta))$



CT: Back Projection

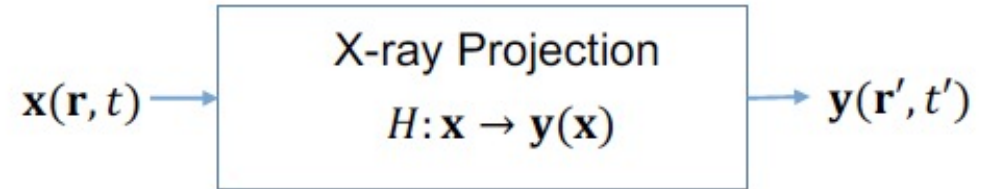
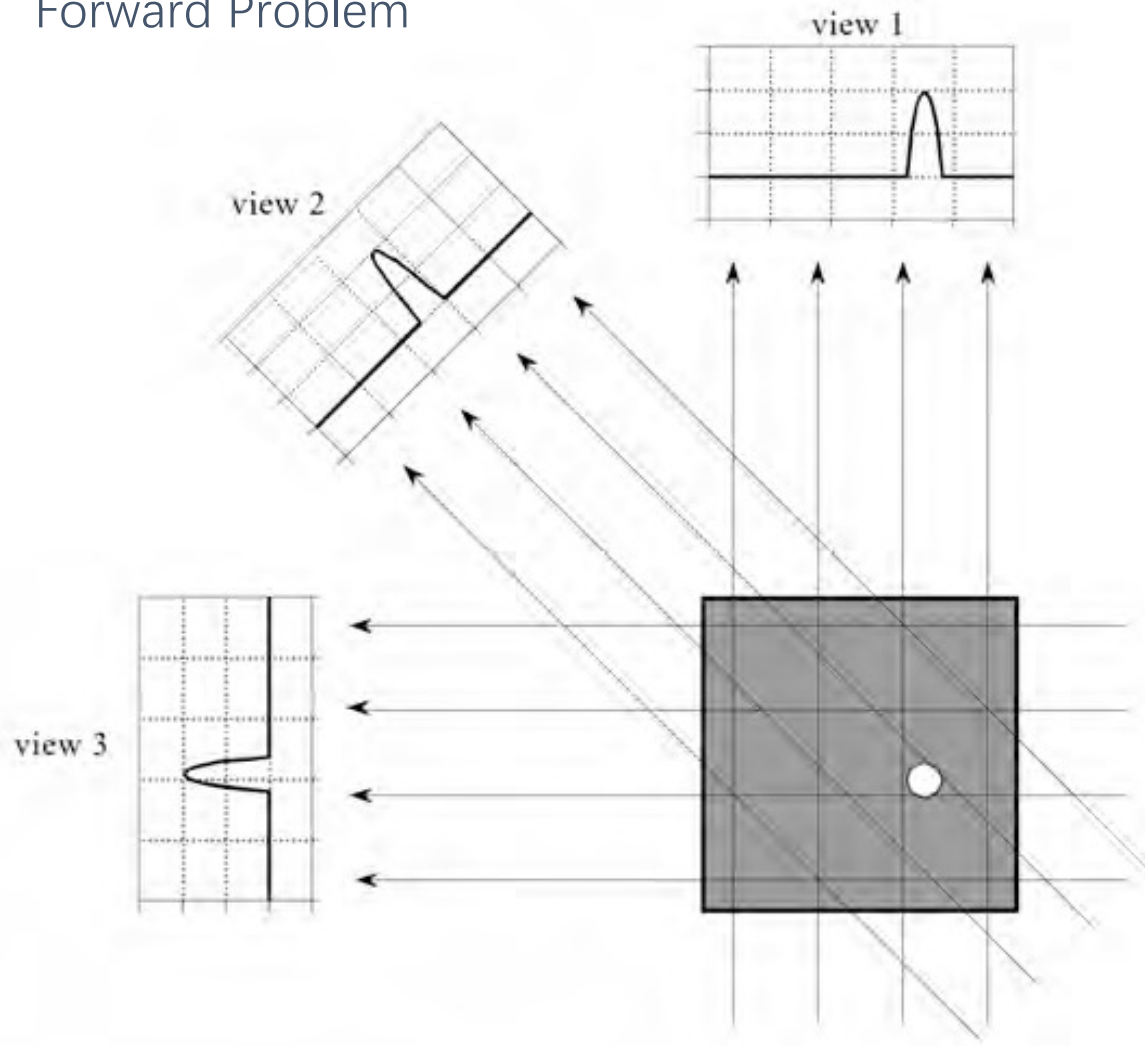


Inverse Problem
Filtered Back Projection



CT: Discrete Optimization Formulation

Forward Problem



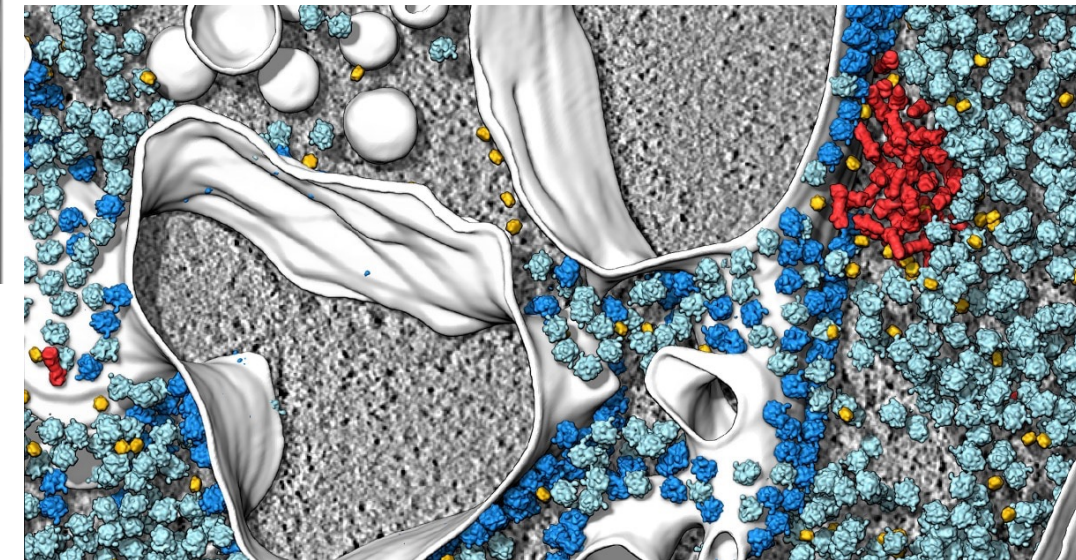
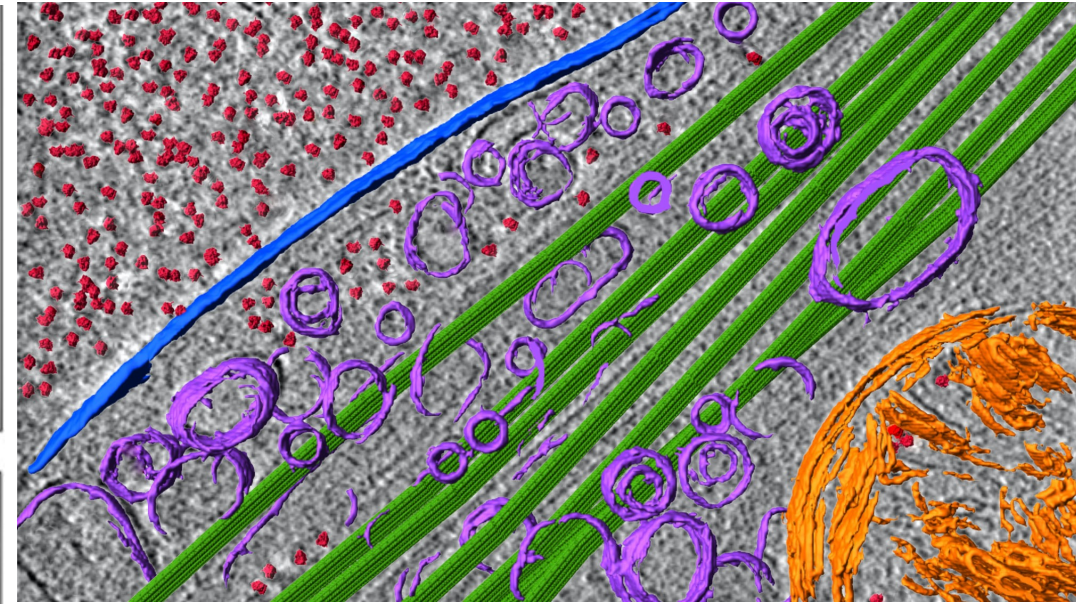
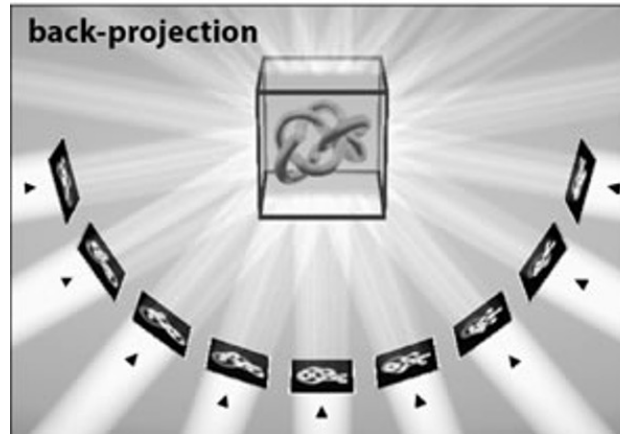
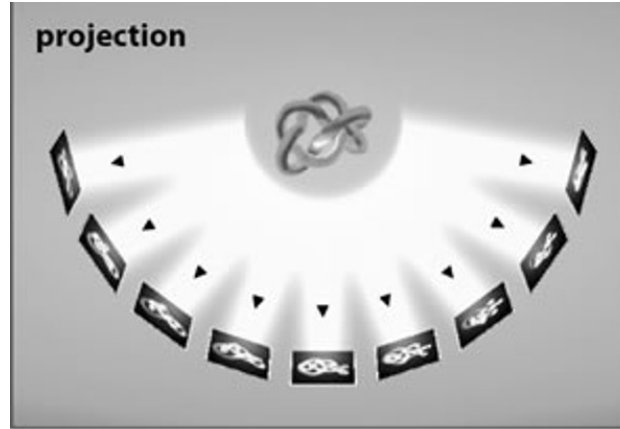
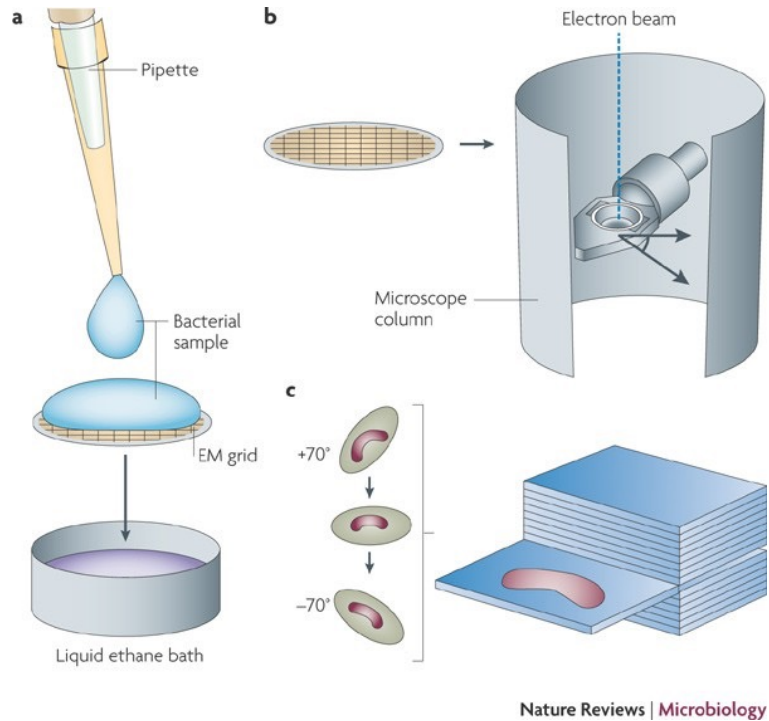
$$\mathbf{y} = \mathbf{A}\mathbf{x}$$

$$\hat{\mathbf{x}} = \underset{\mathbf{x}}{\operatorname{argmin}} \|\mathbf{y} - \mathbf{A}\mathbf{x}\|_2$$

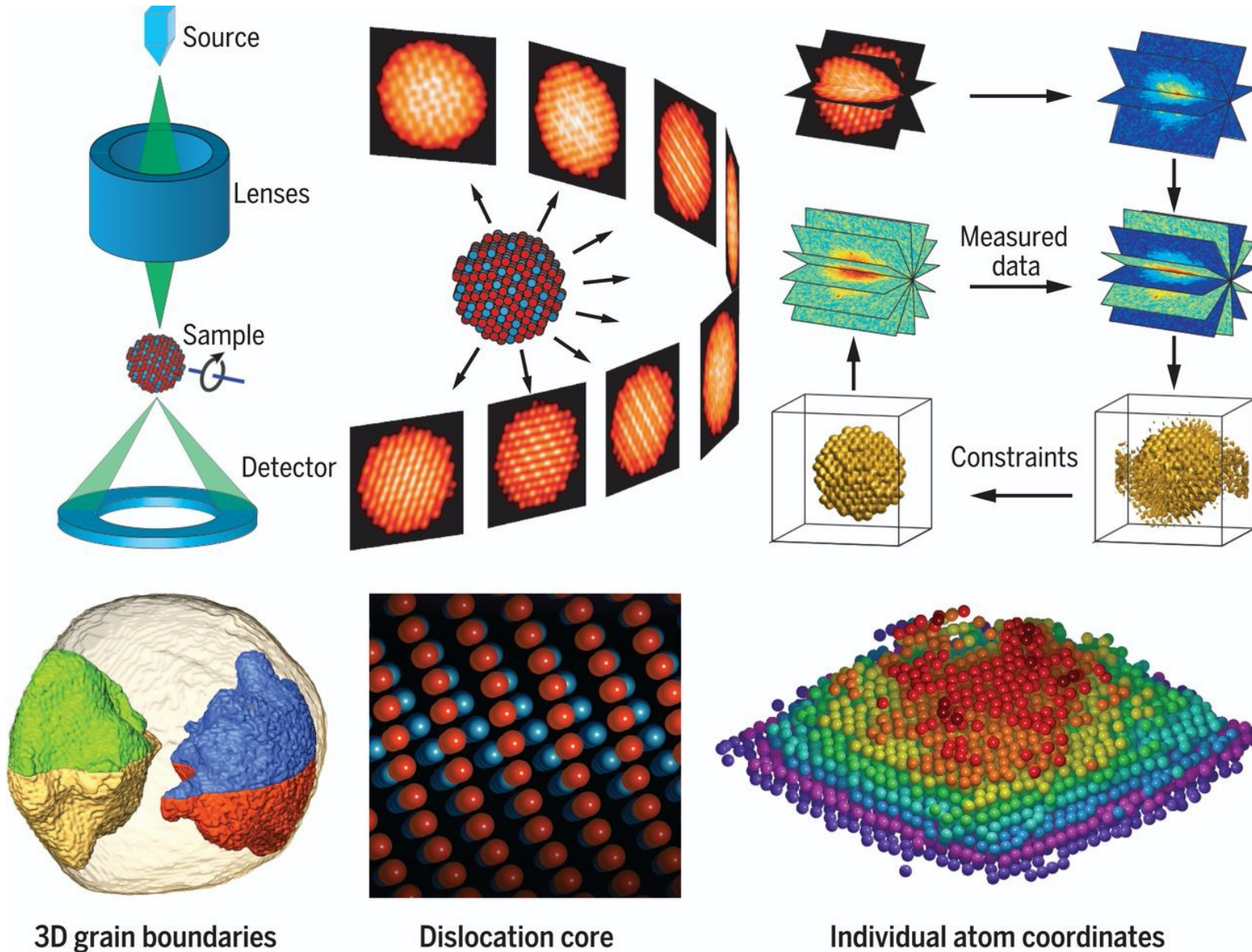
s. t. some constraints



Cryo Electron Tomography (冷冻电子断层扫描)



Atomic electron tomography (原子电子断层扫描)



Magnetic Resonance Imaging (MRI): Fourier Encoding

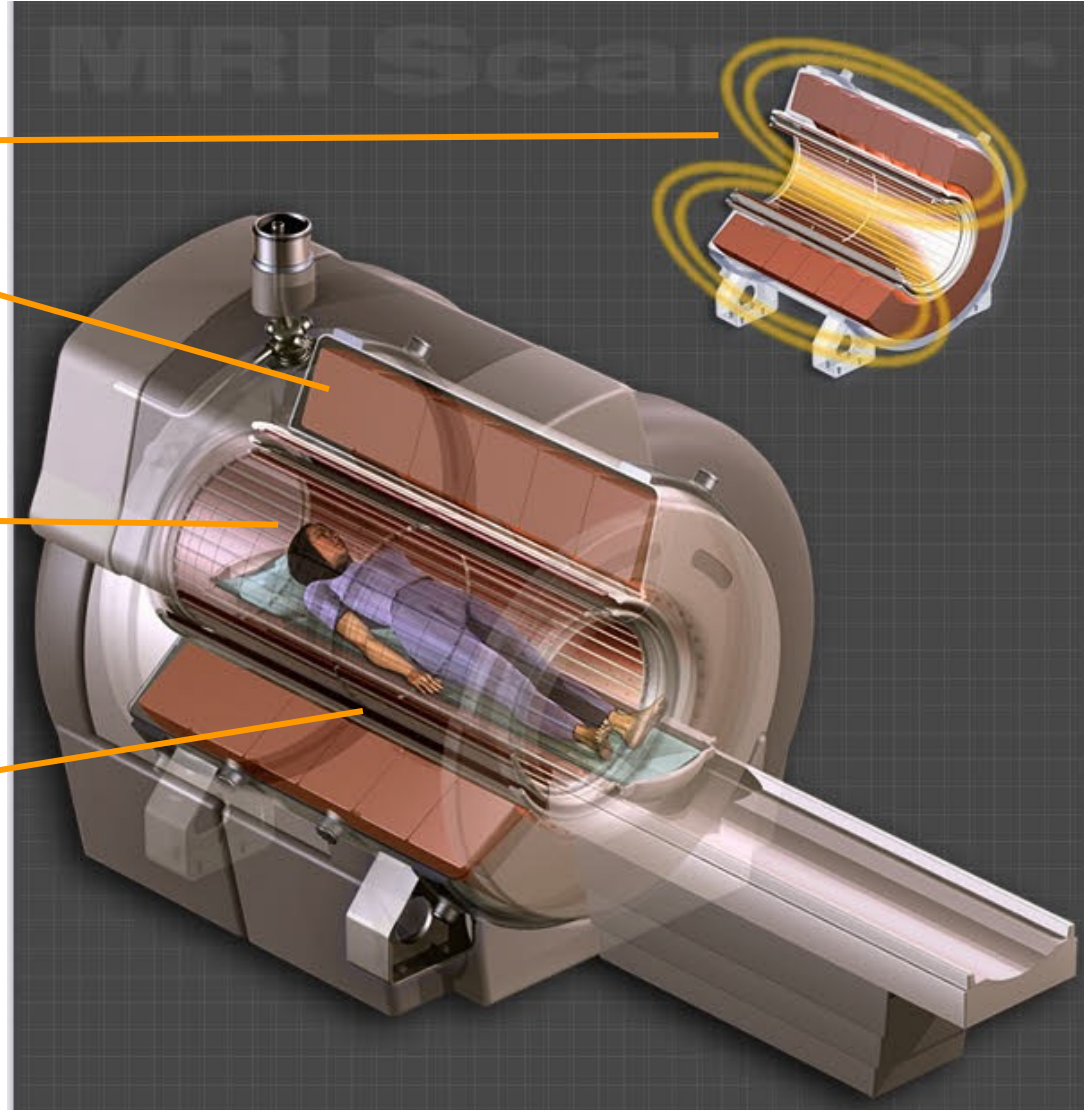


Basic Components of MRI System

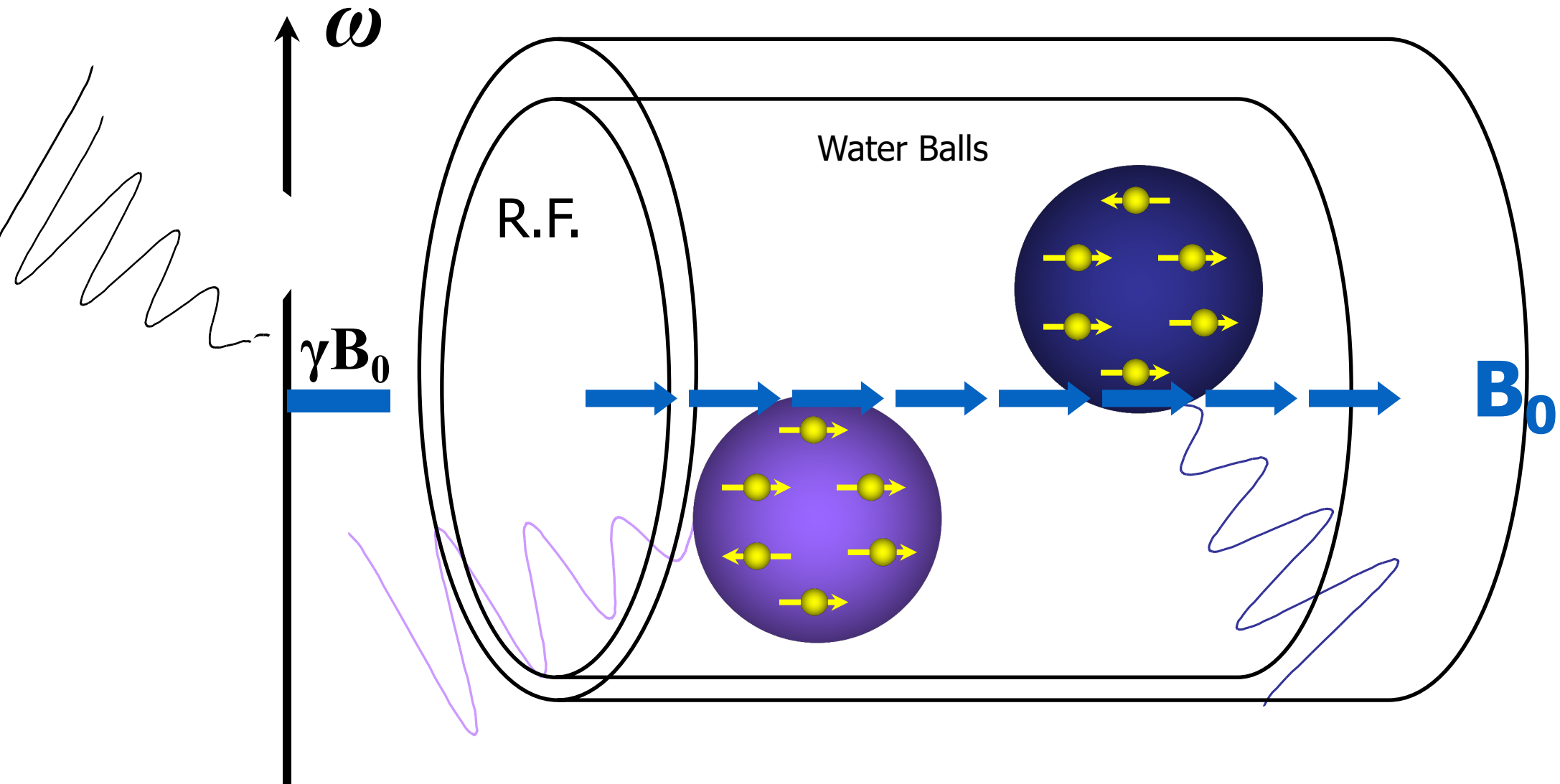
B_0 Field

RF Coil: B_1

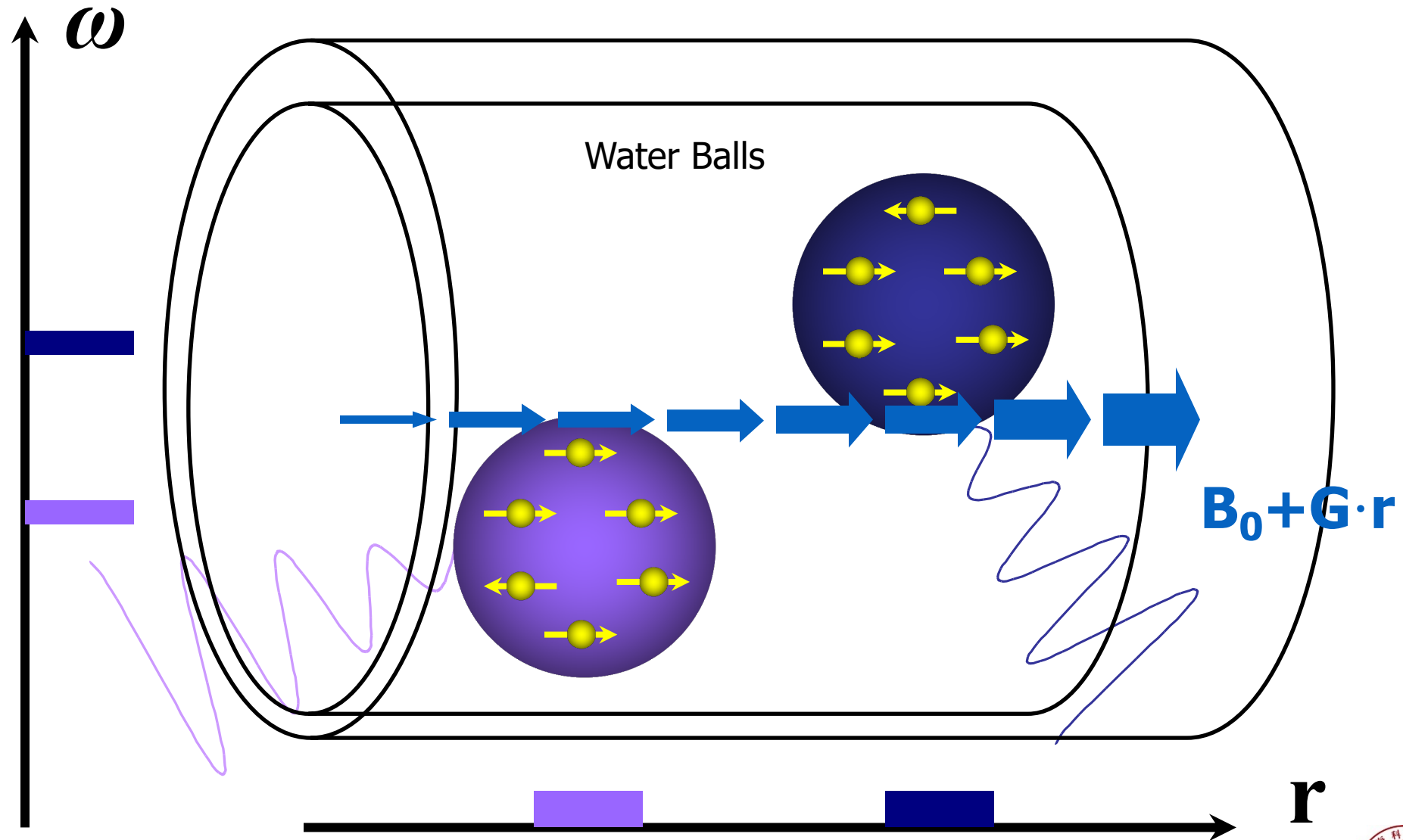
Gradient Coil: G



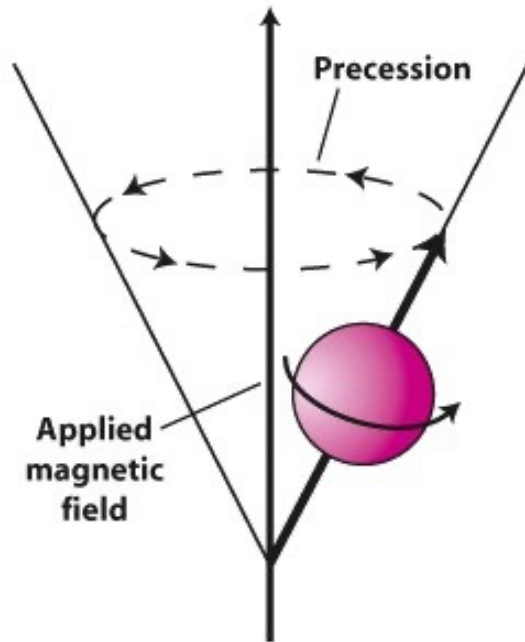
Magnetic Resonance Phenomenon



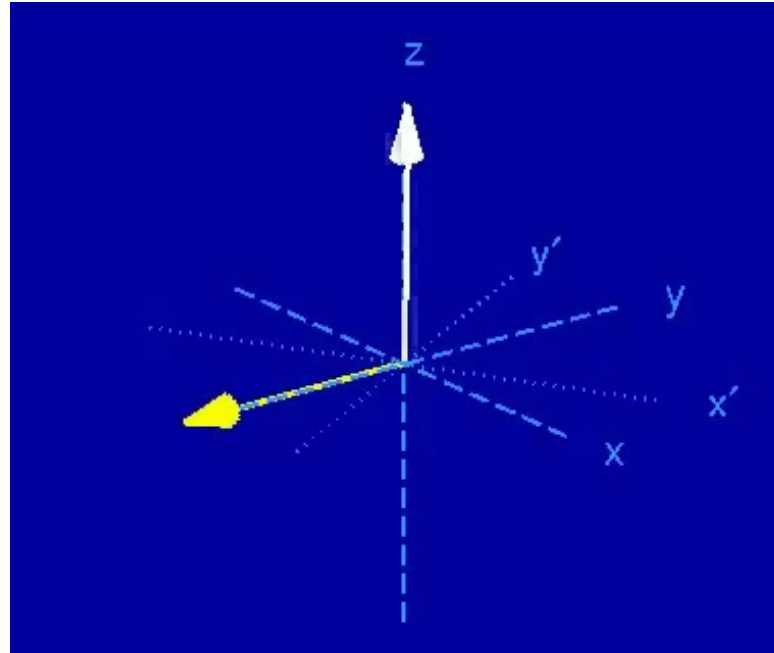
“Big” Idea: Magnetic Field Gradient



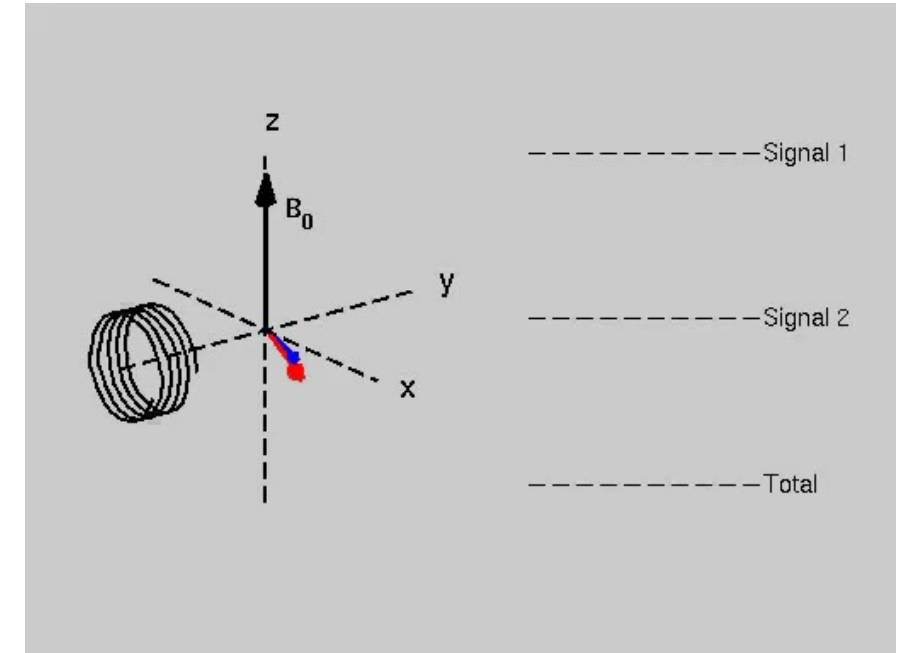
MR Signal Detection



B1 excitation



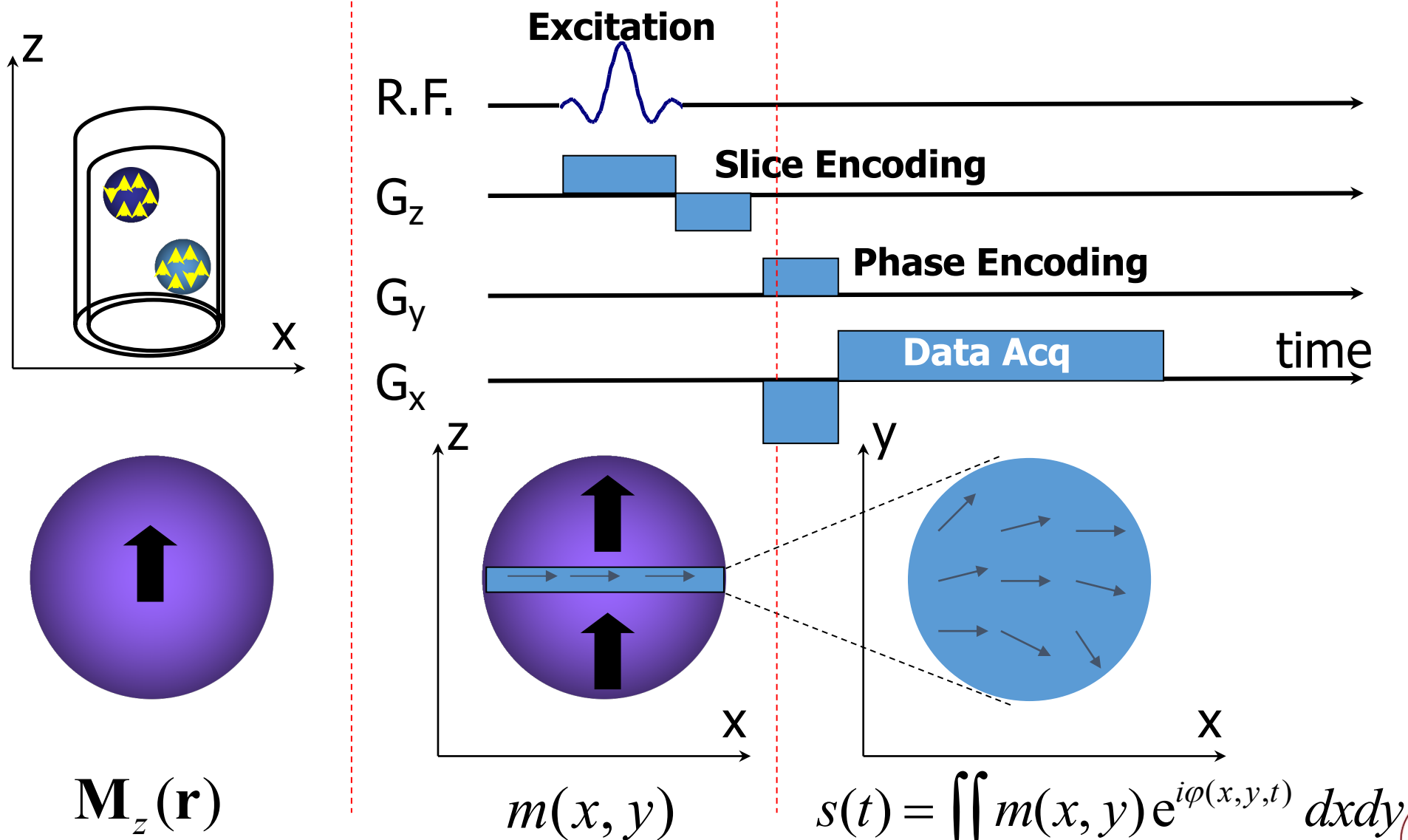
Spin dephasing



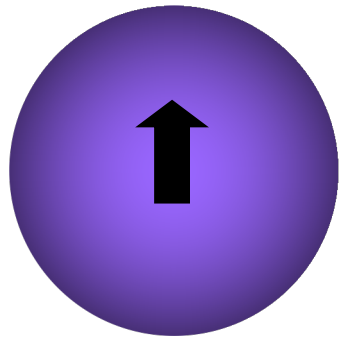
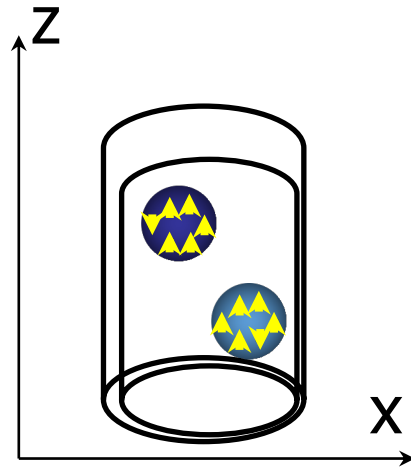
Q: 什么是FID信号,有何特点? 是否包含spin的位置信息?



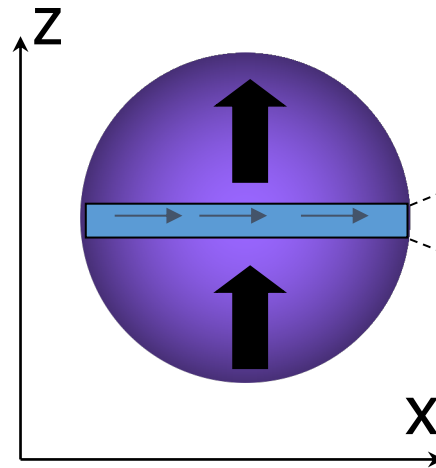
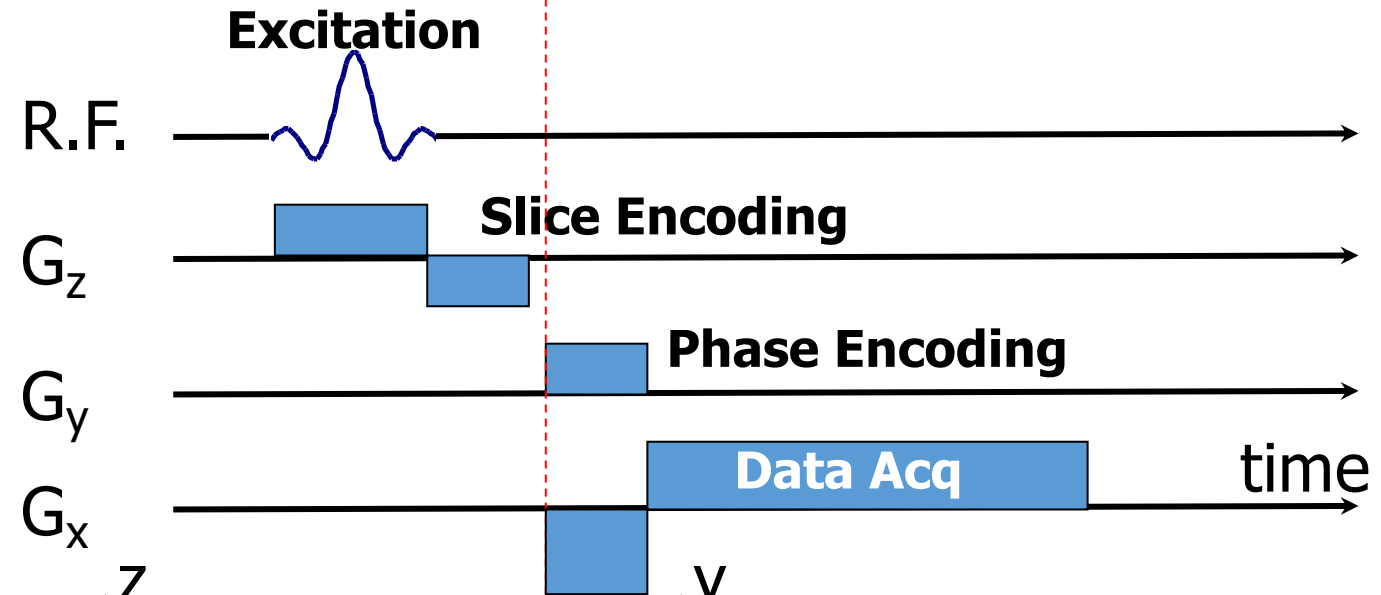
Pulse Sequence: Excitation, Encoding & Acquisition



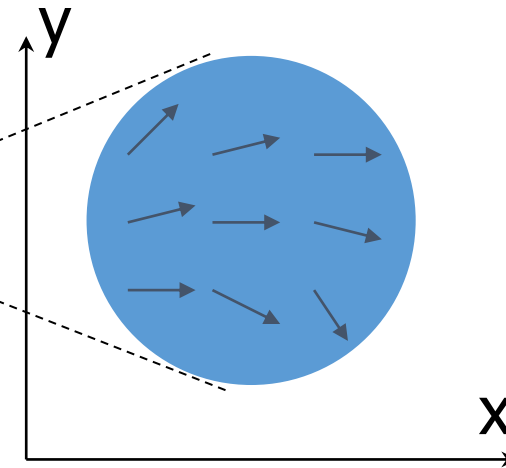
Pulse Sequence: Excitation, Encoding & Acquisition



$\mathbf{M}_z(\mathbf{r})$



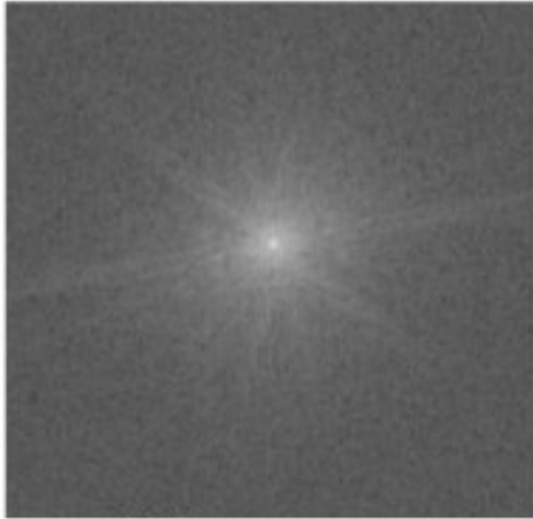
$m(x, y)$



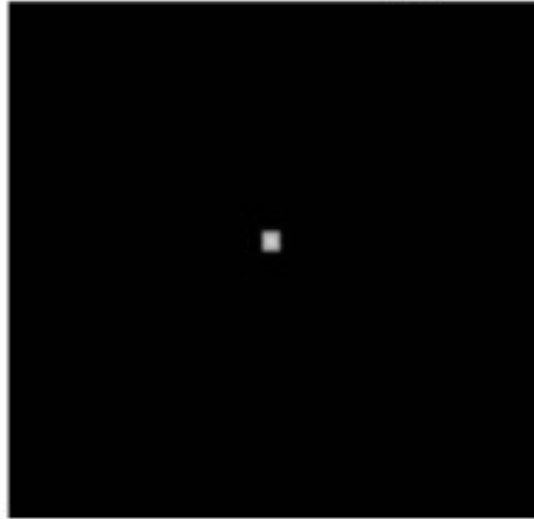
$$s(t) = \iint m(x, y) e^{i\varphi(x, y, t)} dx dy$$



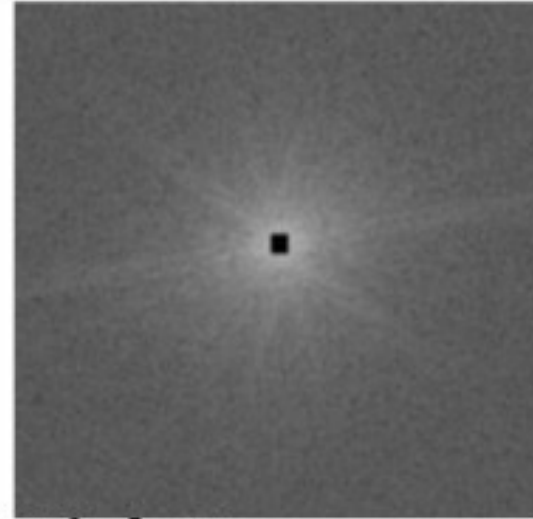
k-Space



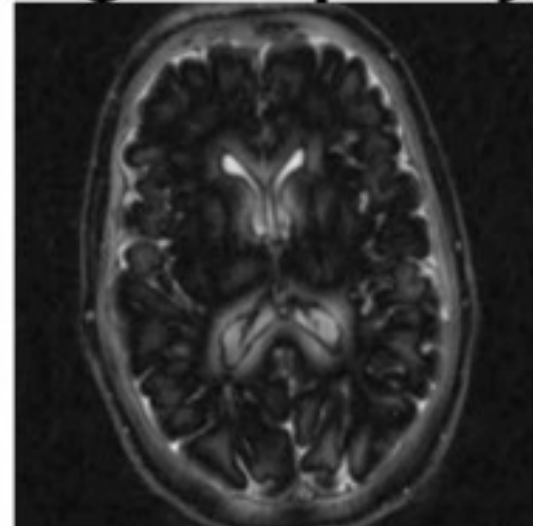
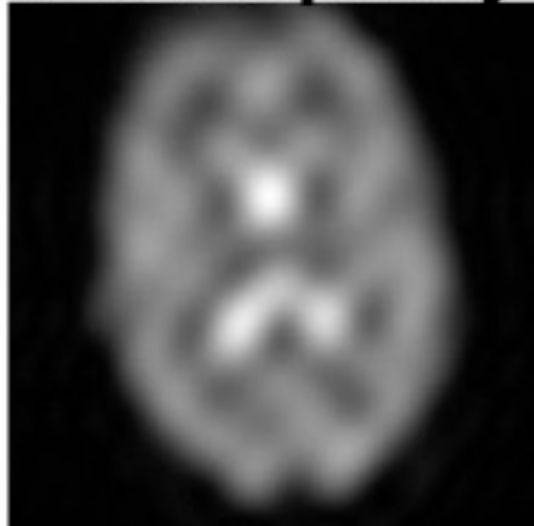
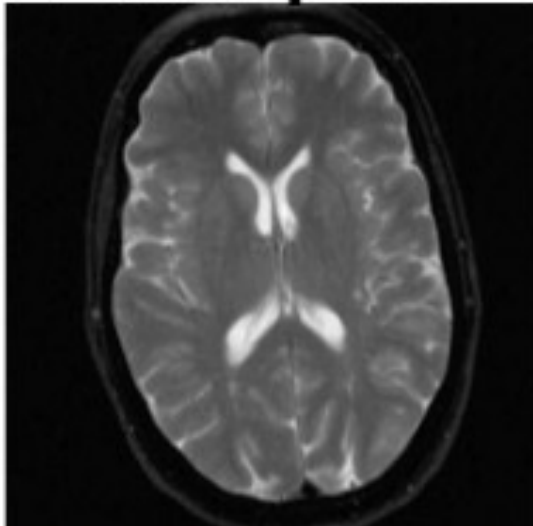
Full k-space



Low Frequency

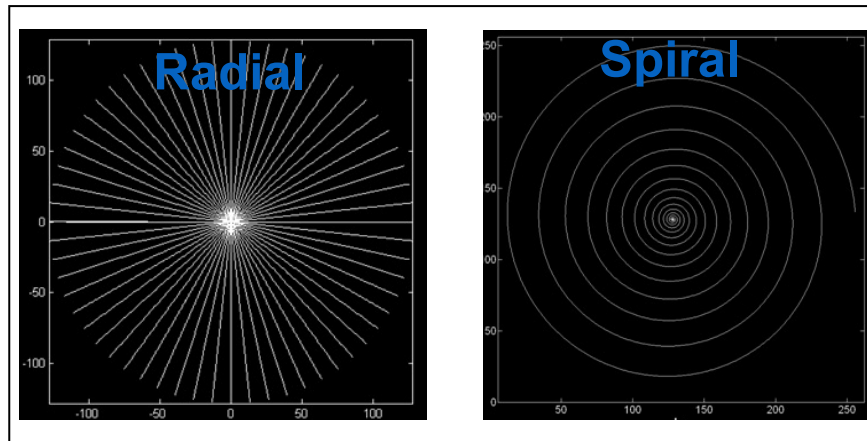
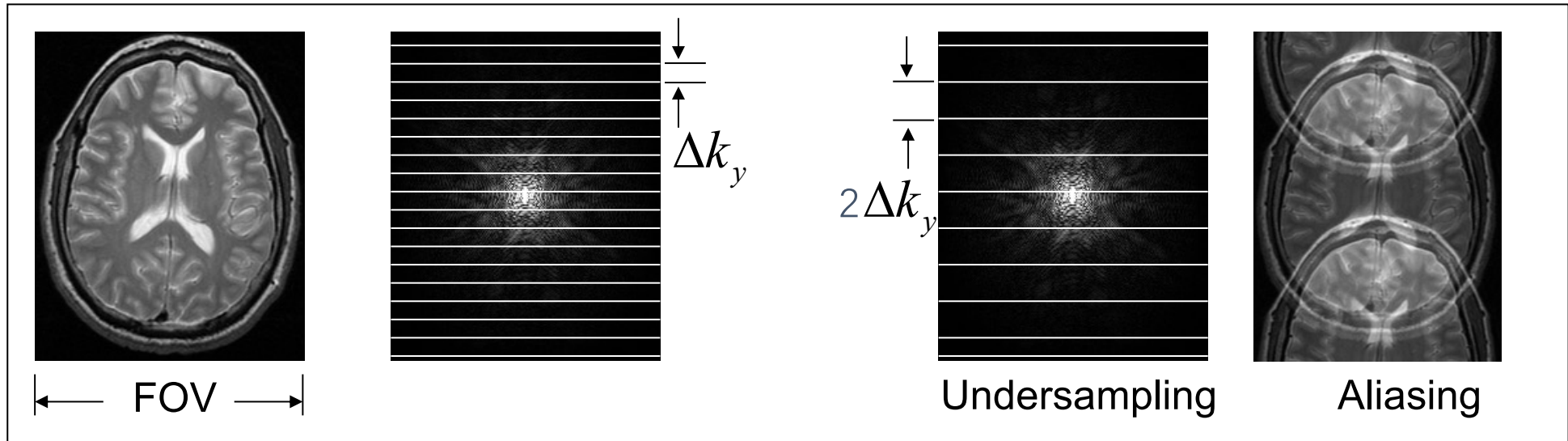


High Frequency



k-Space Sampling

Nyquist rate: $f = 2 \cdot Bw \rightarrow \Delta k_x = 1/FOV_x, \Delta k_y = 1/FOV_y$

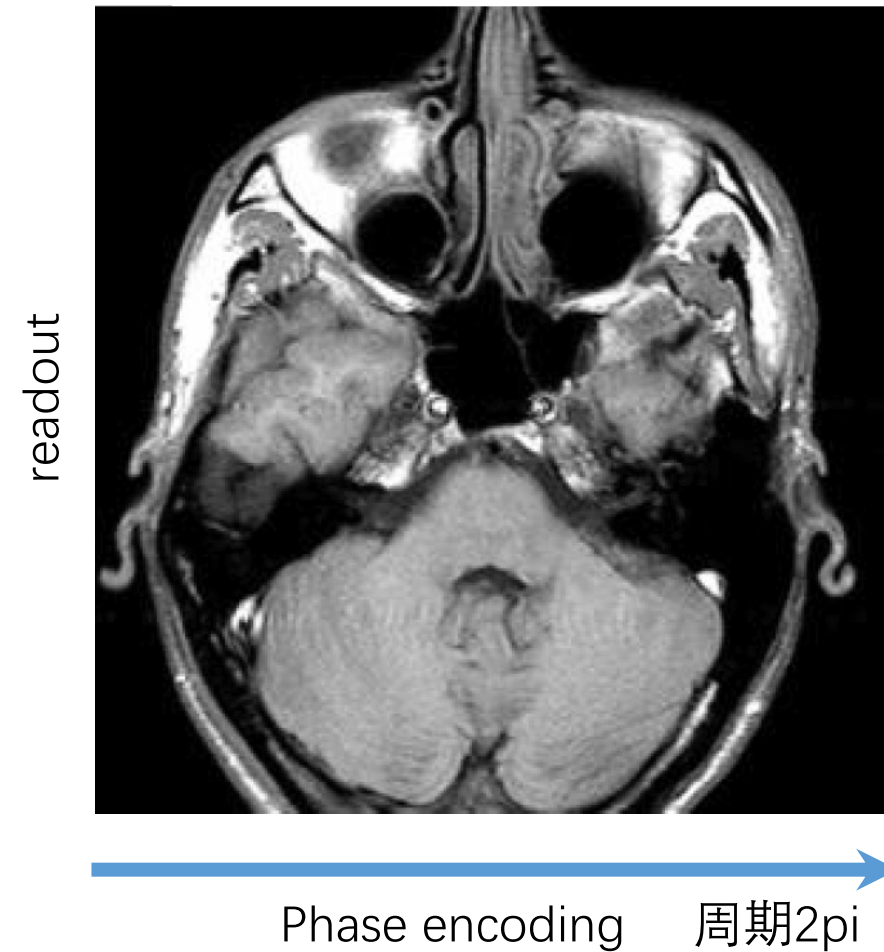
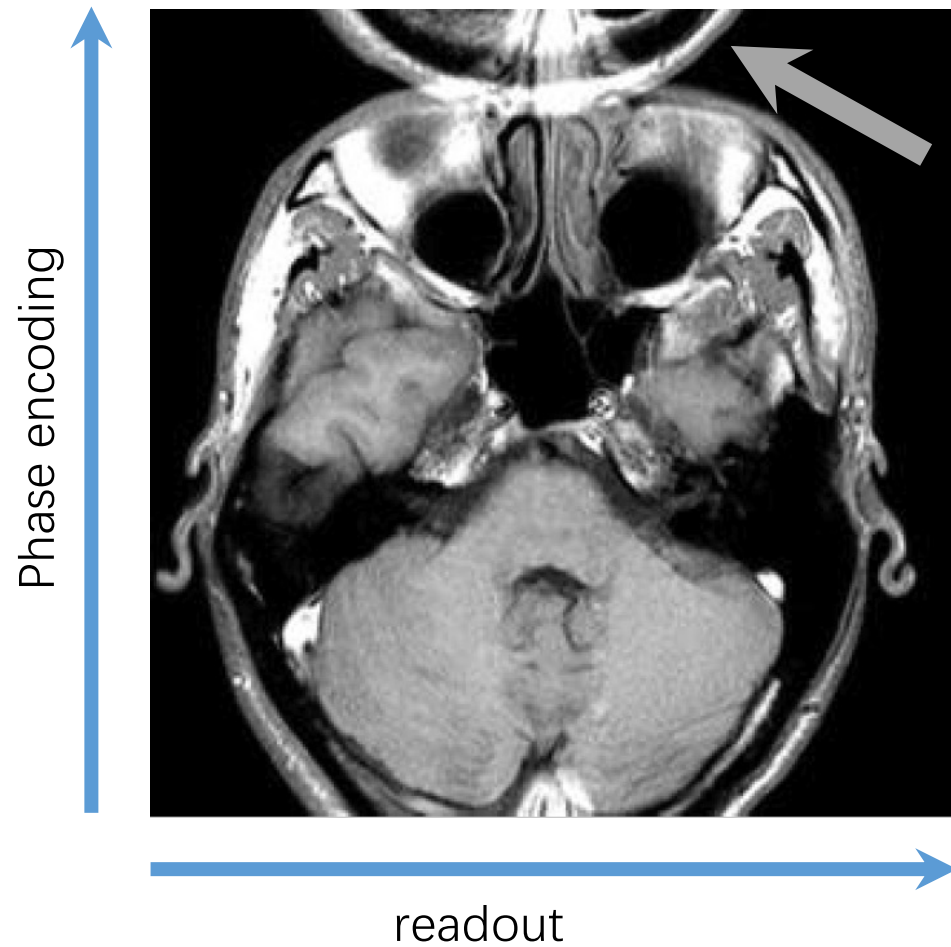


- Non-Cartesian Sampling
 - Design of gradient waveform
 - Image recon needs gridding



Aliasing

What is the difference in acquisition between the two images?



Take home message

- Tomography imaging reconstruction is based on accumulating back-projections data directive, while the reconstructed images are blurred.
- The 2D Fourier transform of an image for reconstruction can be obtained by accumulating the Fourier transform of the projections at different angles (Fourier-Slice Theorem).
- Key problem: how to use less projections to reconstruct clear images?