

Medical Imaging (RT516 – Spring 2021)

Lecture 5: Computed Tomography

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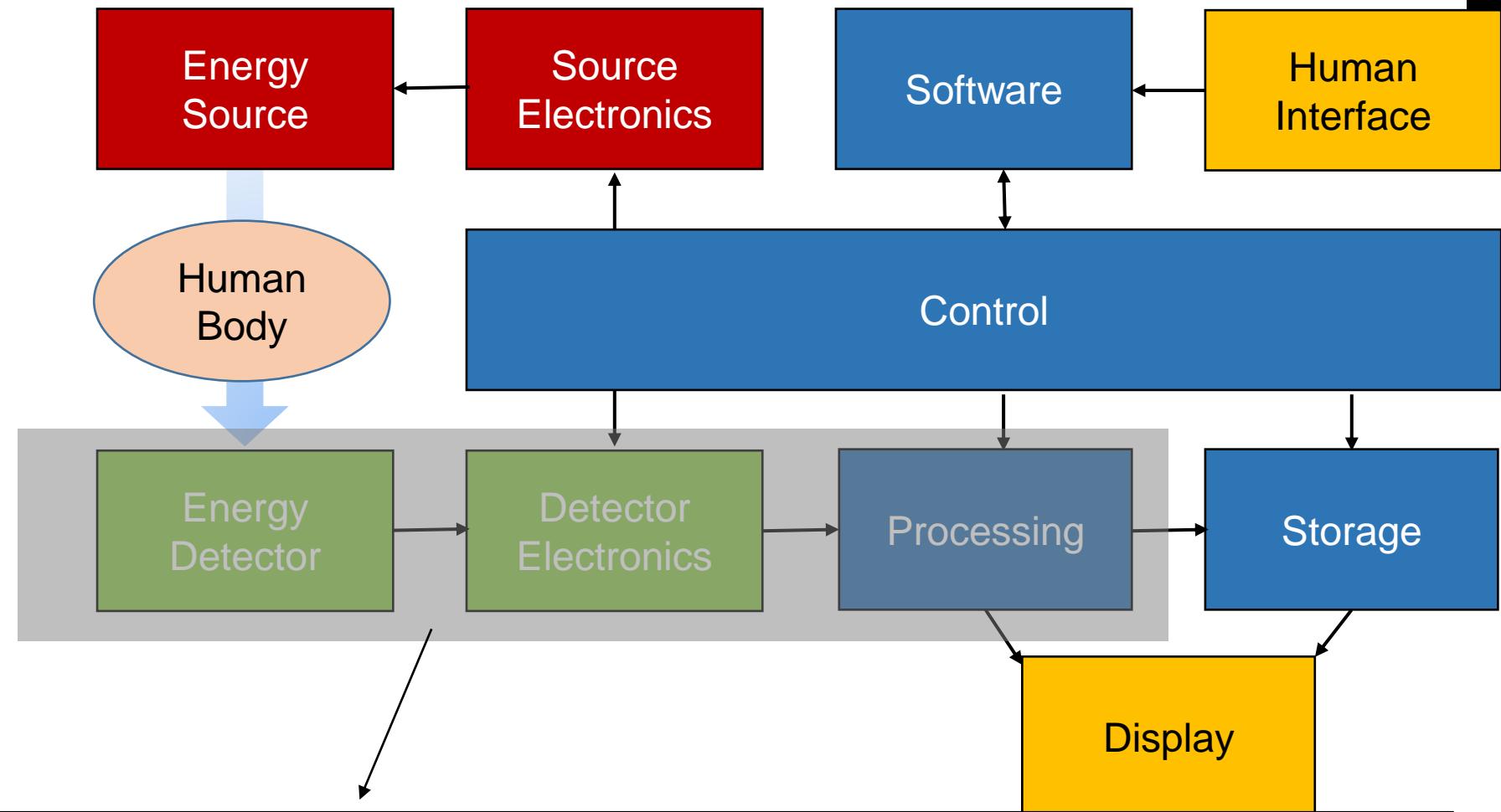
► Why Computed Tomography?

- Conventional Radiography (X-ray) is great, but... limited:
 - Imposition of a 3D subject onto a 2D plane (overlapping objects due to 'projection')
 - Limited differentiation of different densities (e.g., water and soft tissue look the same)
 - **Goal in this lecture:** Key will be understanding how to form images using acquired data.

► Computed Tomography

- Tomography refers to a **picture** (-graph) of a **slice** (tomo-)
- Truly “**Computed**”, i.e., impossible without computers
- Introduced in 1972, Last 40+ years, Improved over 7 generations
- Godfrey Hounsfield (UK) and Allan Cormack (U.S.) Nobel prize in medicine (1979)
- First CT scanner, EMI Mark 1: 80x80 pixels, 3-mm pixel, 4.5 minutes to acquire,
1.5 minute to reconstruct. Now, 1024x1024 pixels, 0.05-mm pixel, 130 images/s
- Original development: head imaging only. The capture/reconstruction were too slow for cardiac imaging.

► Computed Tomography (CT)



As CT utilized same energy source of the projection radiography, Our focus will be on
Instrumentation / Image reconstruction / Image quality / Applications

► Basic principles (Let's see the video, again)

► Basic principles

- Two orthogonal projections (views):

- 3-D localization

- CT: 1,000 views

- Data acquisition:

- Parallel beam geometry

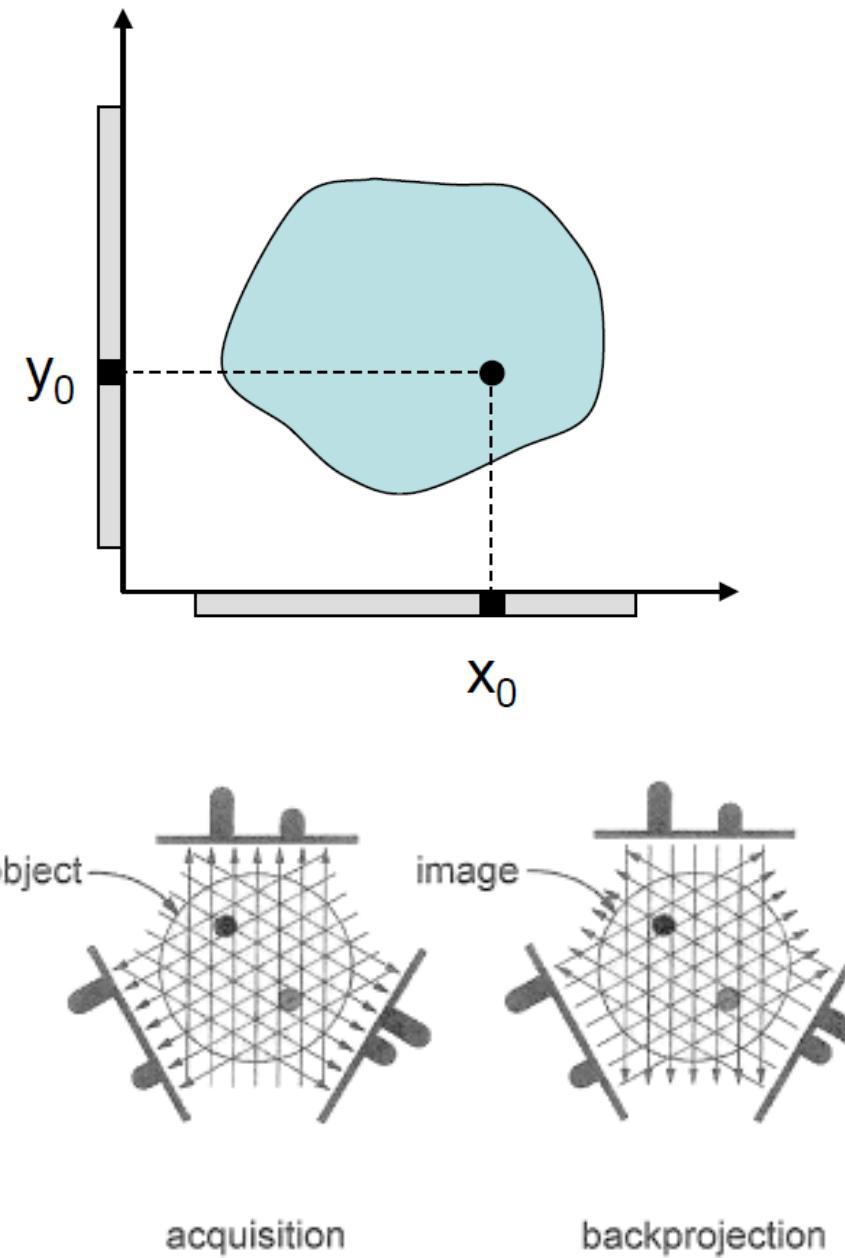
- Fan beam geometry

- 800 rays at 1,000 angles

- Reconstruction

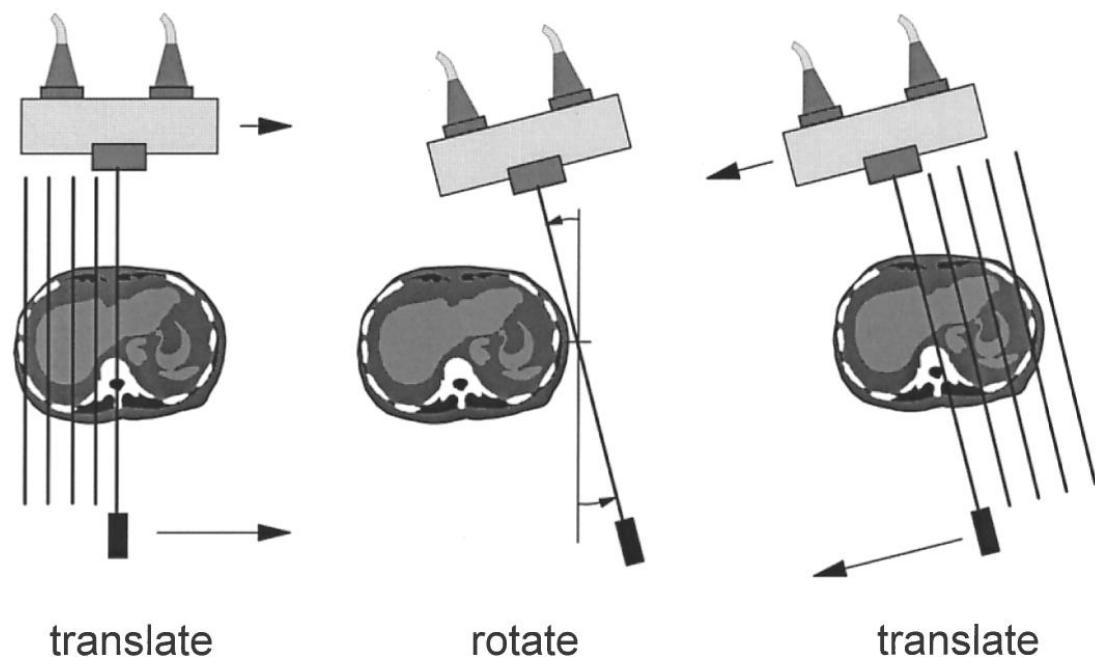
$$I = I_0 e^{-\mu r}$$

$$\ln \left(\frac{I_0}{I} \right) = \mu r$$



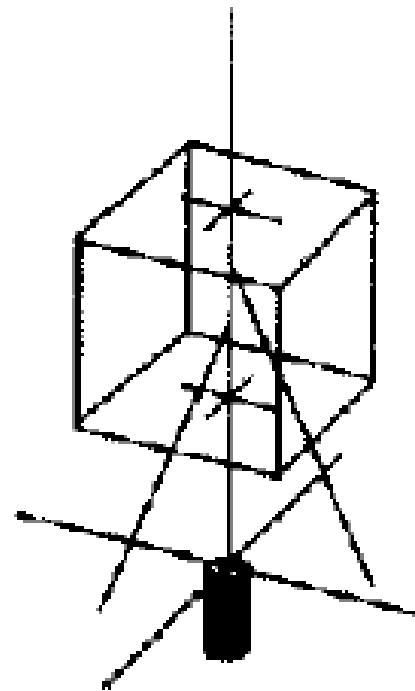
► First generation

- Rotate/Translate Pencil Beam
- One detector
- 160 parallel rays, 24 cm field of view
- 180 projections (1 degree rotation)
- Total of $180 \times 160 = 28,800$ rays were measured

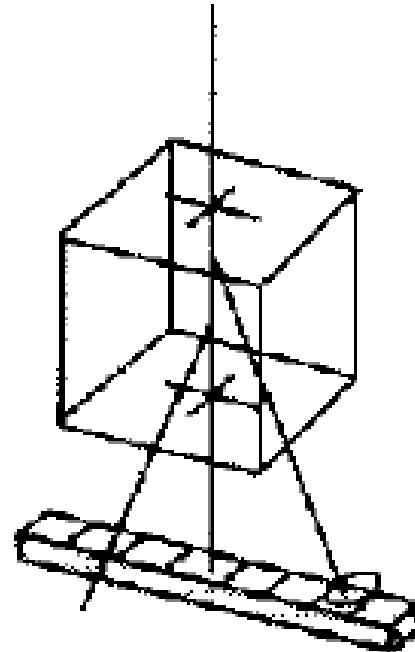


Pros. Scattering can be easily rejected
Cons. Motion artifact, Acquisition Time

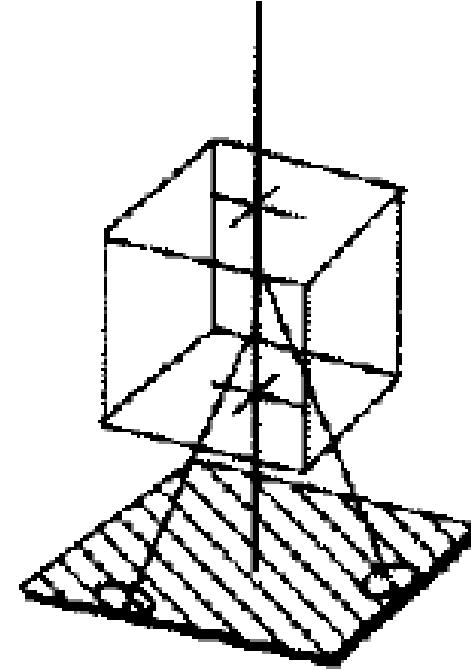
▶ FYI, Beam shape



Pencil beam



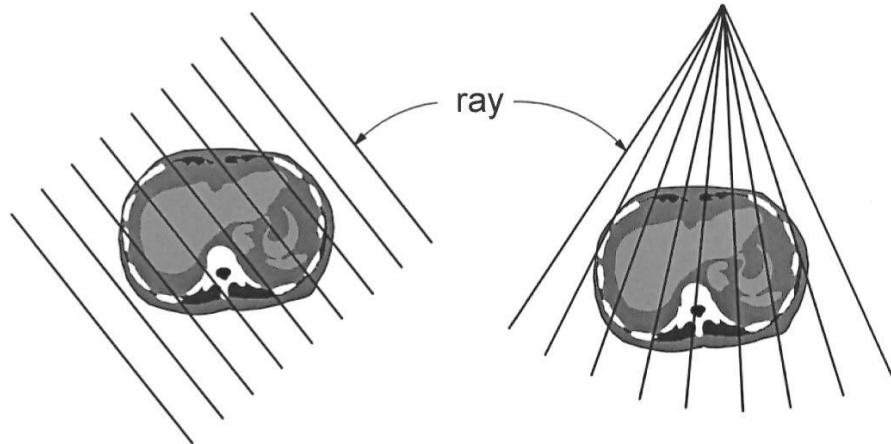
Fan beam



Open beam

► Second generation

- Rotate/Translate
- Narrow Fan Beam (10-degree)

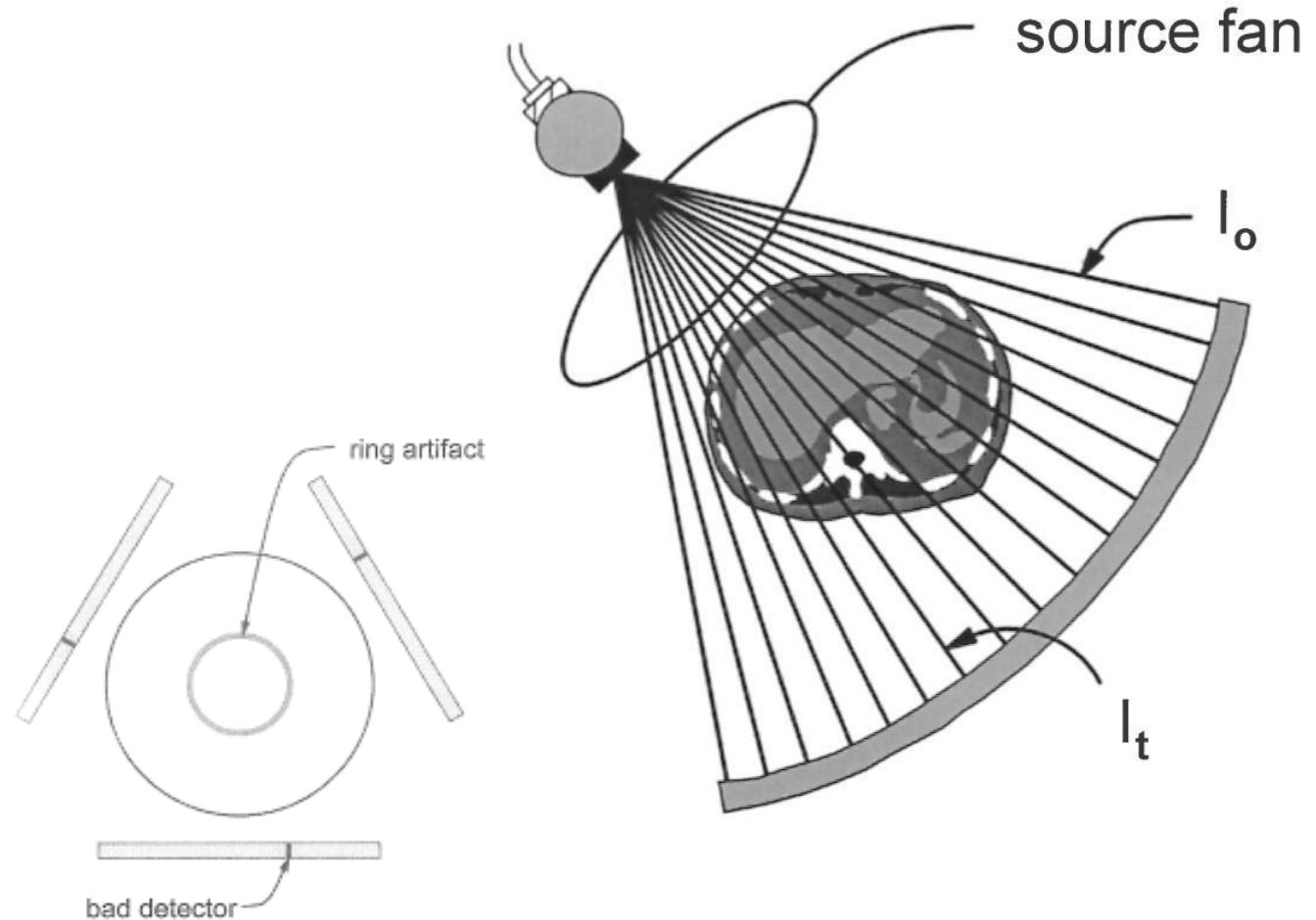


- 30 detectors (small!) / 600 rays / 540 views
- Total of $600 \times 540 = 324,000$ rays (data points) were measured

Pros. 15 times faster than the 1st generation (18s / slice)
Cons. Increased scattering

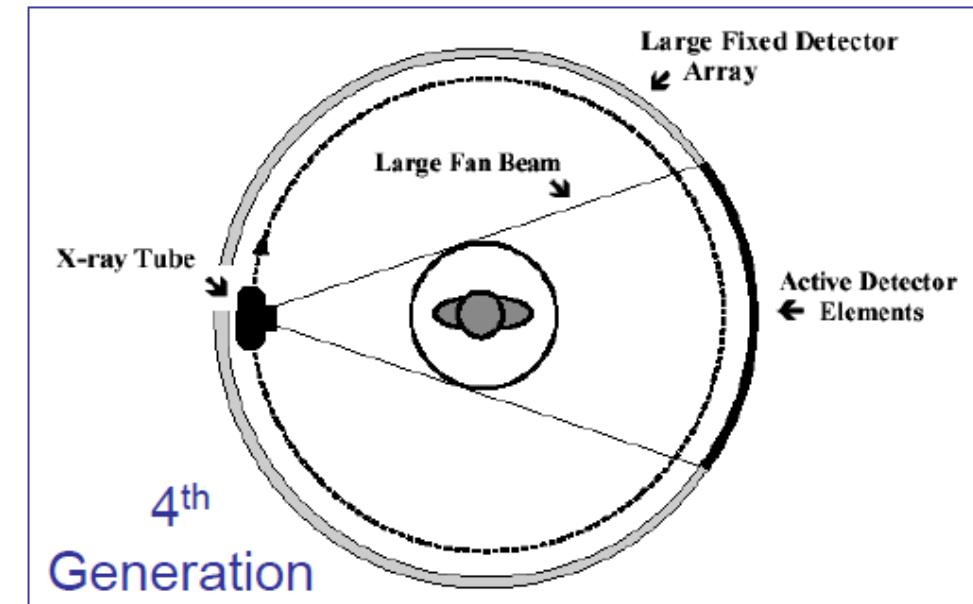
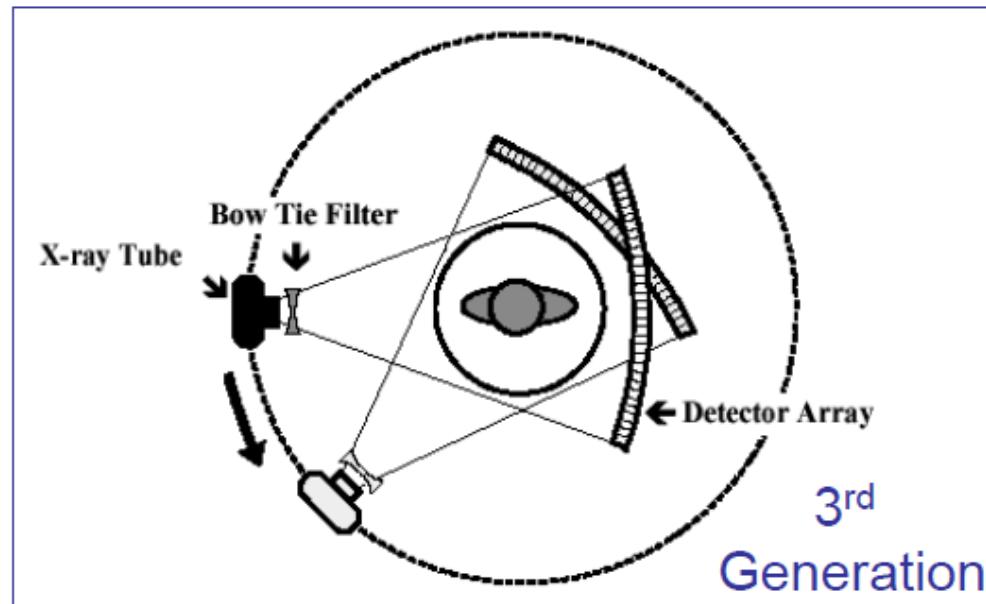
► Third generation

- Rotate/Rotate Wide Fan Beam
- > 800 detector
- 5 seconds per slice
- Major improvement in CT
- Artifacts (?)



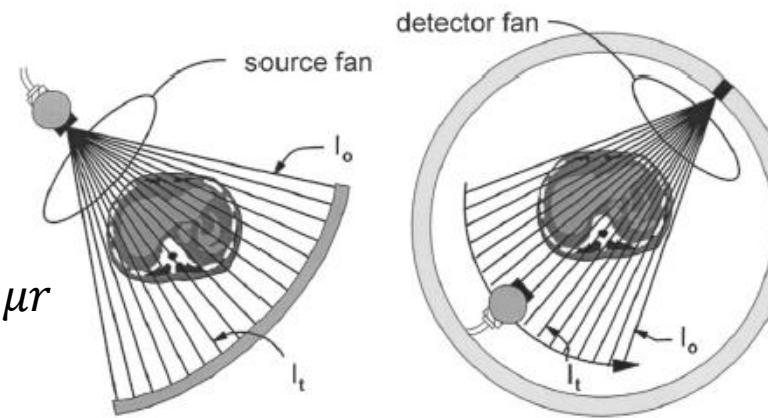
Pros. 3 times faster than the 2nd generation (5s / slice)
Cons. Scattering artifact

► Fourth generation (Modern CT)



- Rotate/Stationary Wide Fan Beam
- 360° stationary detectors
- Reduction of Ring artifacts

$$\ln\left(\frac{g_k I_0}{g_n I}\right) = \mu r$$

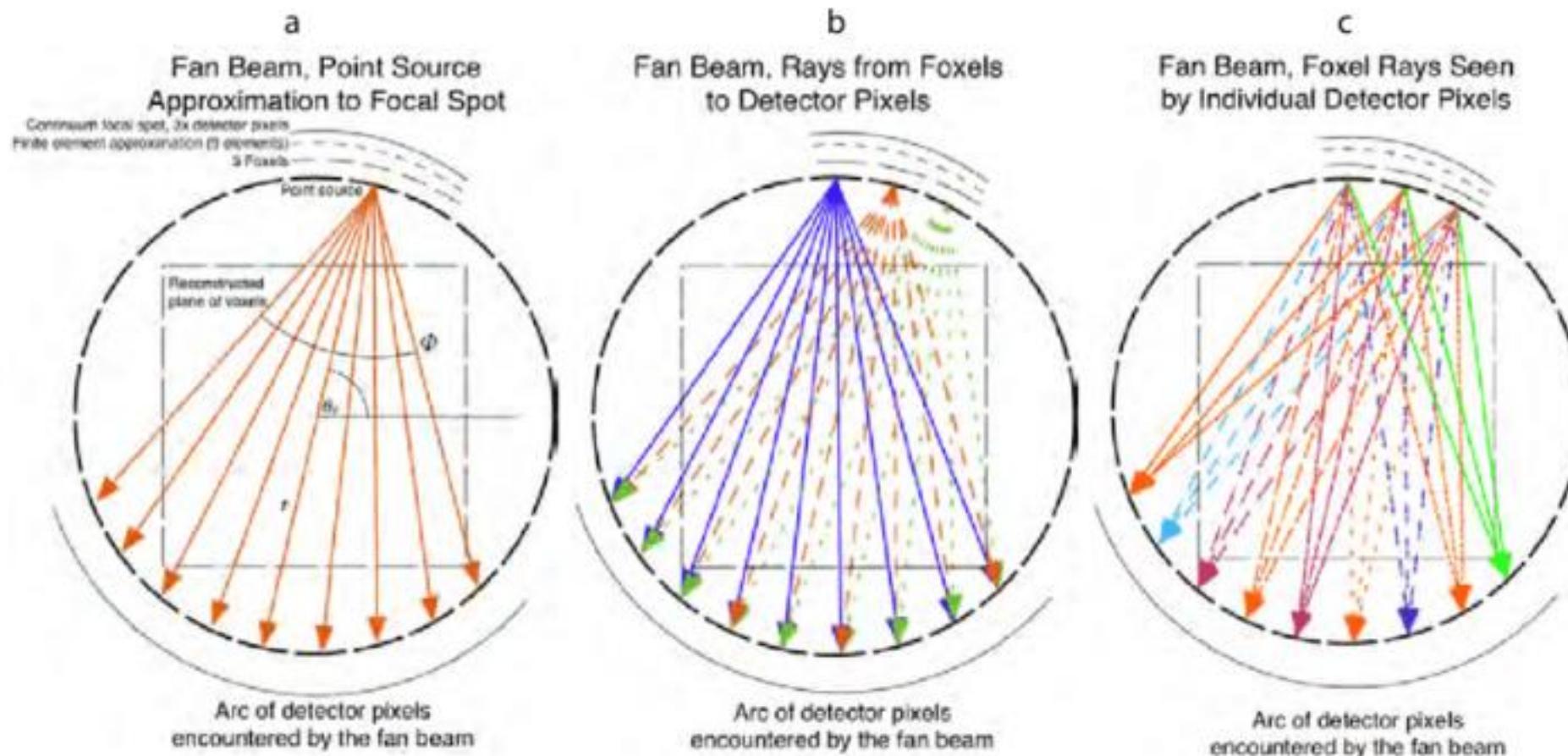


$$\ln\left(\frac{g_n I_0}{g_n I}\right) = \mu r$$

third generation

fourth generation

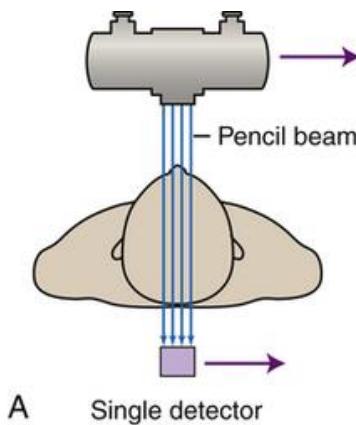
► Fourth generation (Modern CT)



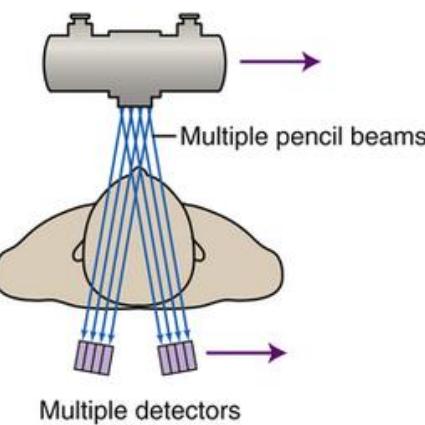
$$\ln\left(\frac{g_k I_0}{g_n I}\right) = \mu r$$

$$\boxed{\ln\left(\frac{g_n I_0}{g_n I}\right)} = \mu r$$

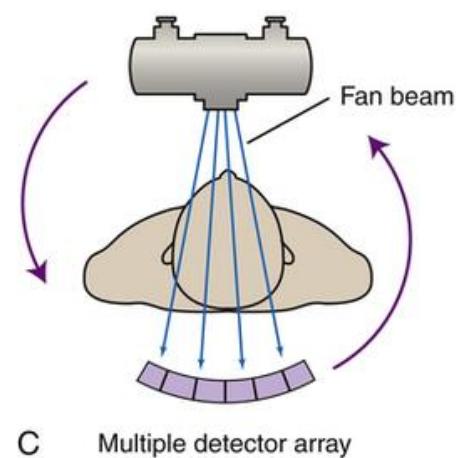
► Comparison



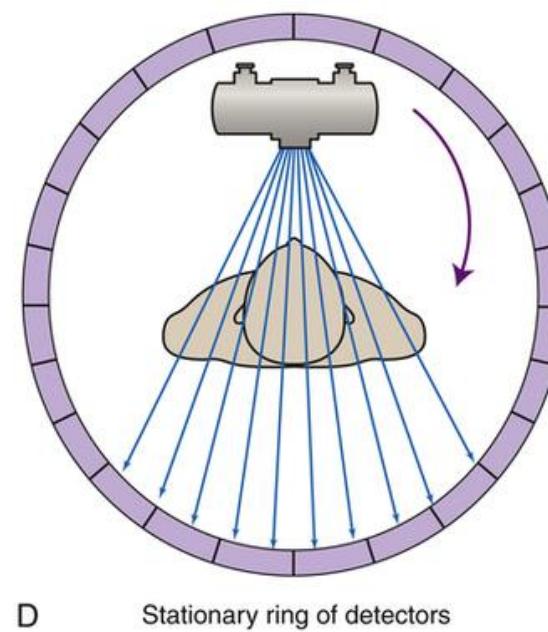
A Single detector



B Multiple detectors



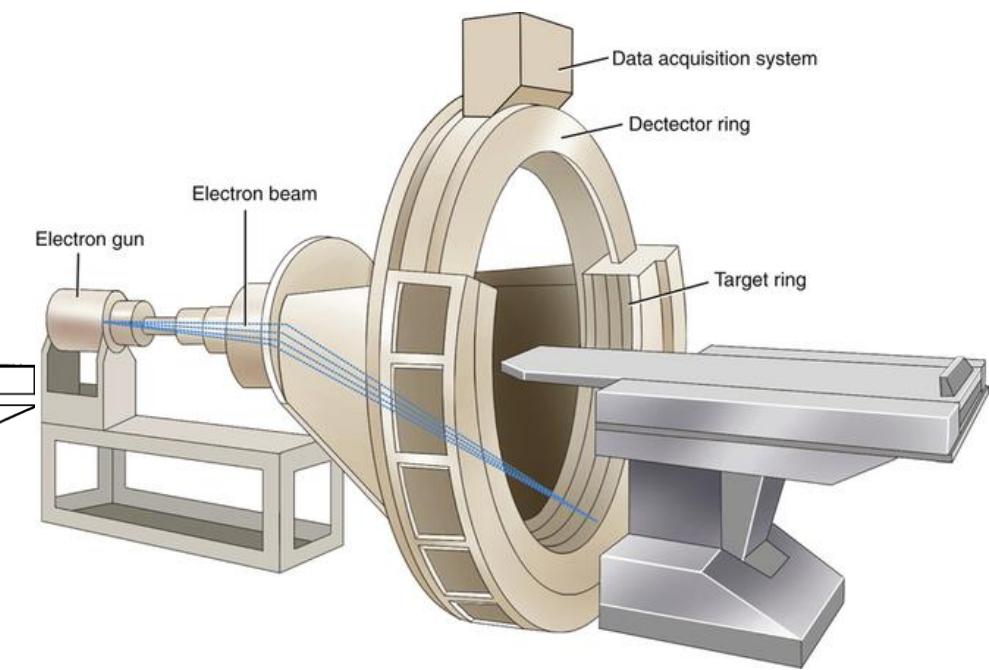
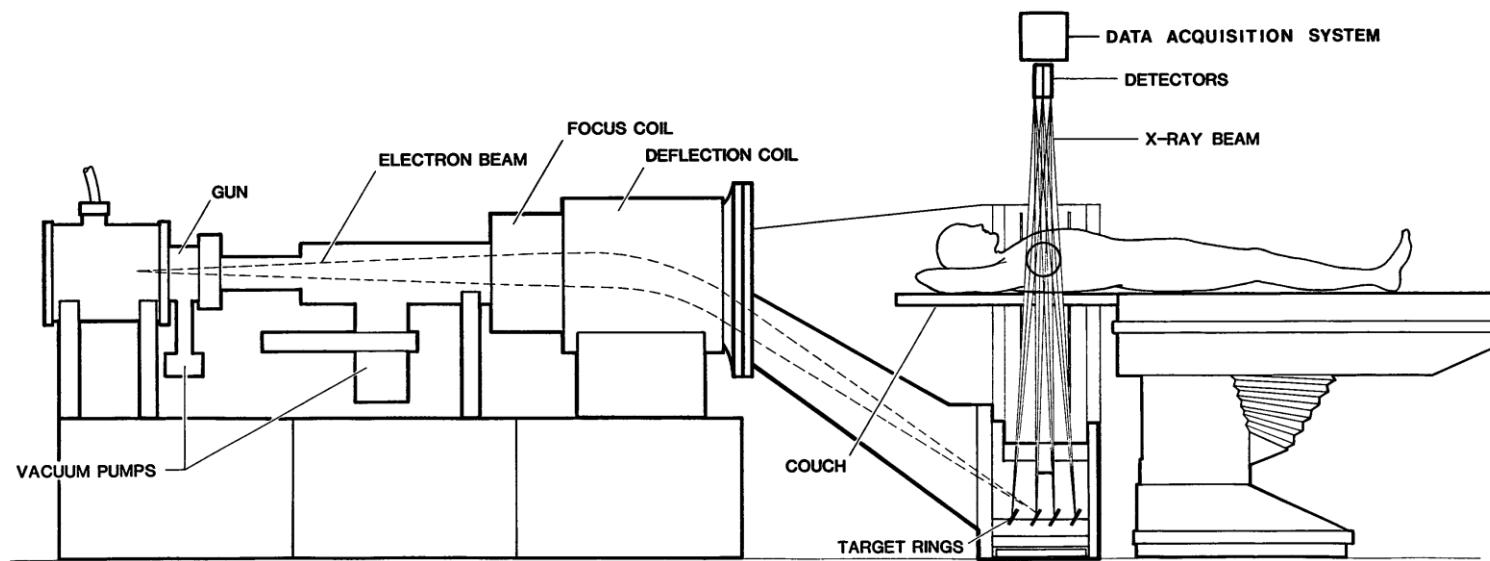
C Multiple detector array



D Stationary ring of detectors

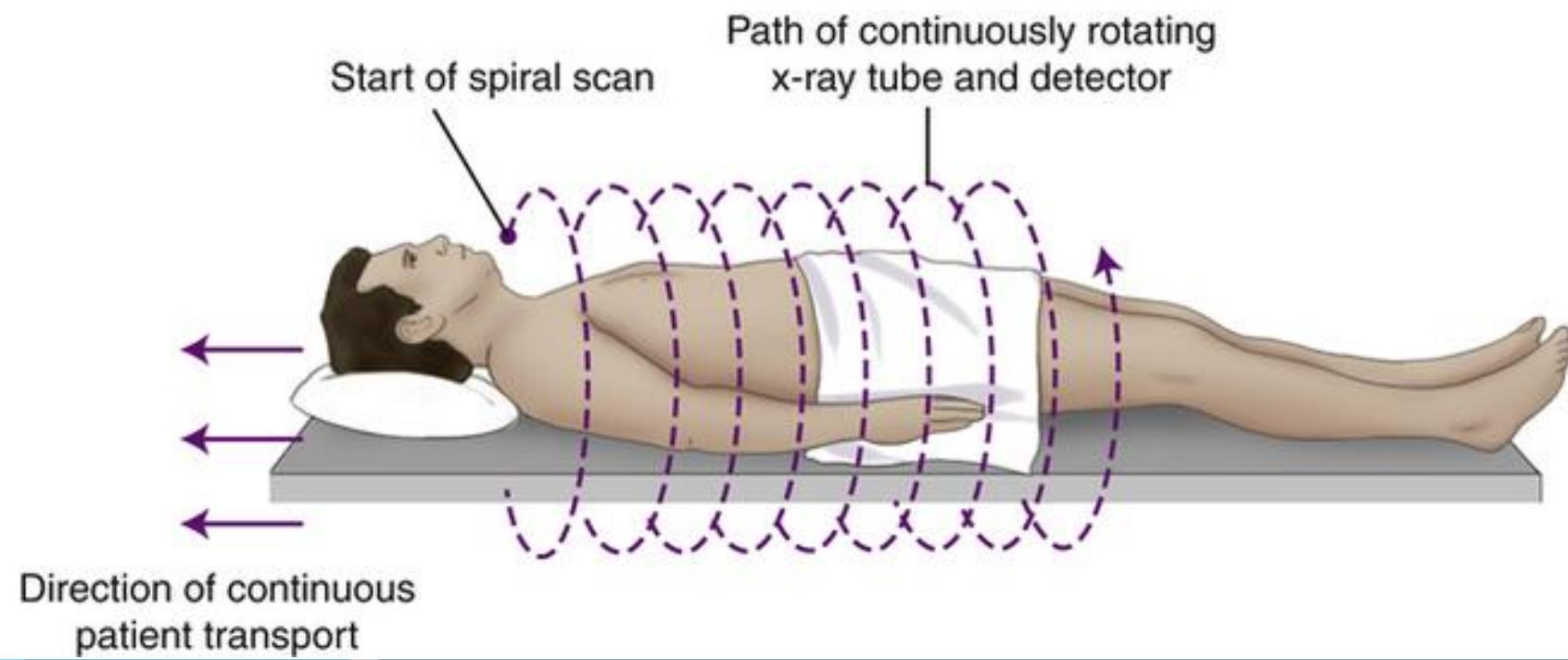
► Fifth generation

- Stationary/Stationary
- Tungsten arc source around patient, detector ring
- Faster! (50 ms scan time) – Cardiac applications available !



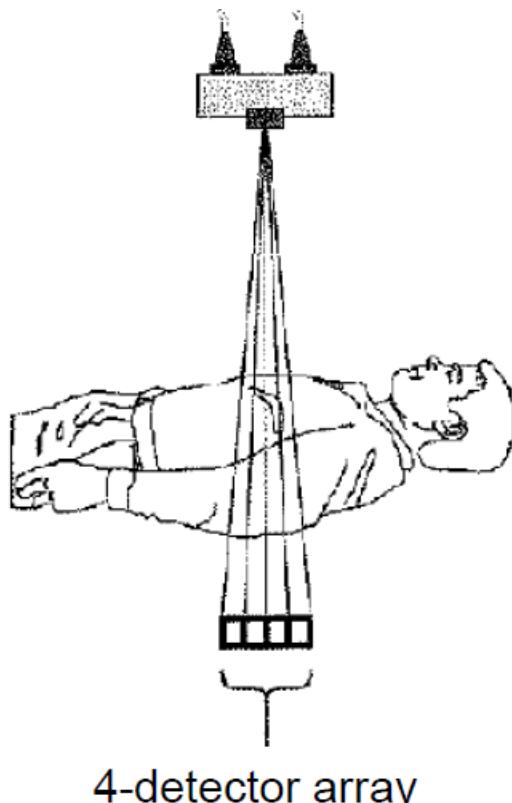
► Sixth generation

- Helical (spiral) scanning
 - Ribbon cables / Slip ring technology
- Increased scan time: (30 second scan for entire abdomen (60 cm))

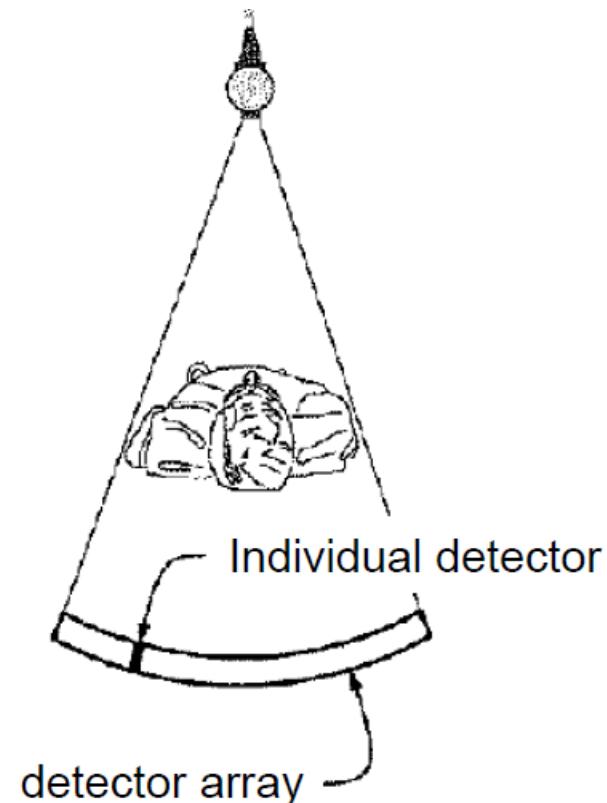


► Seventh generation

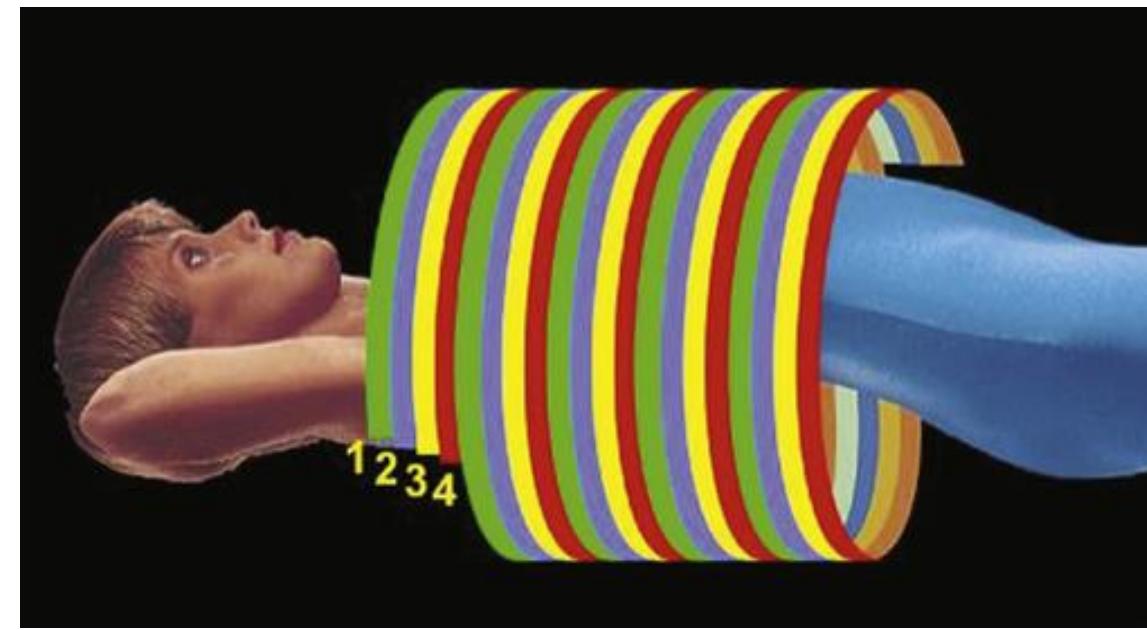
- Multiple detector array
- Better use of x-rays



4-detector array

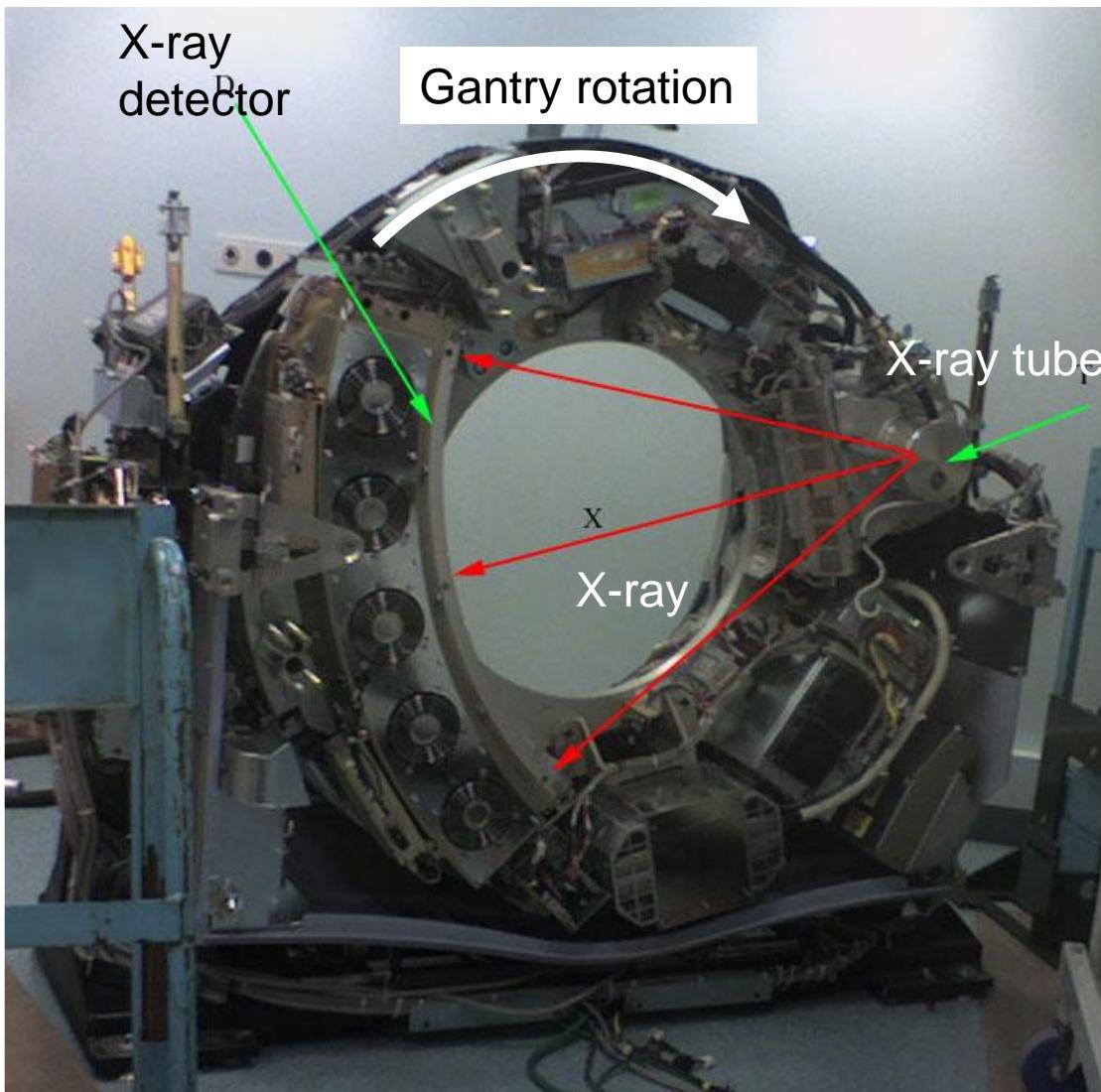


detector array



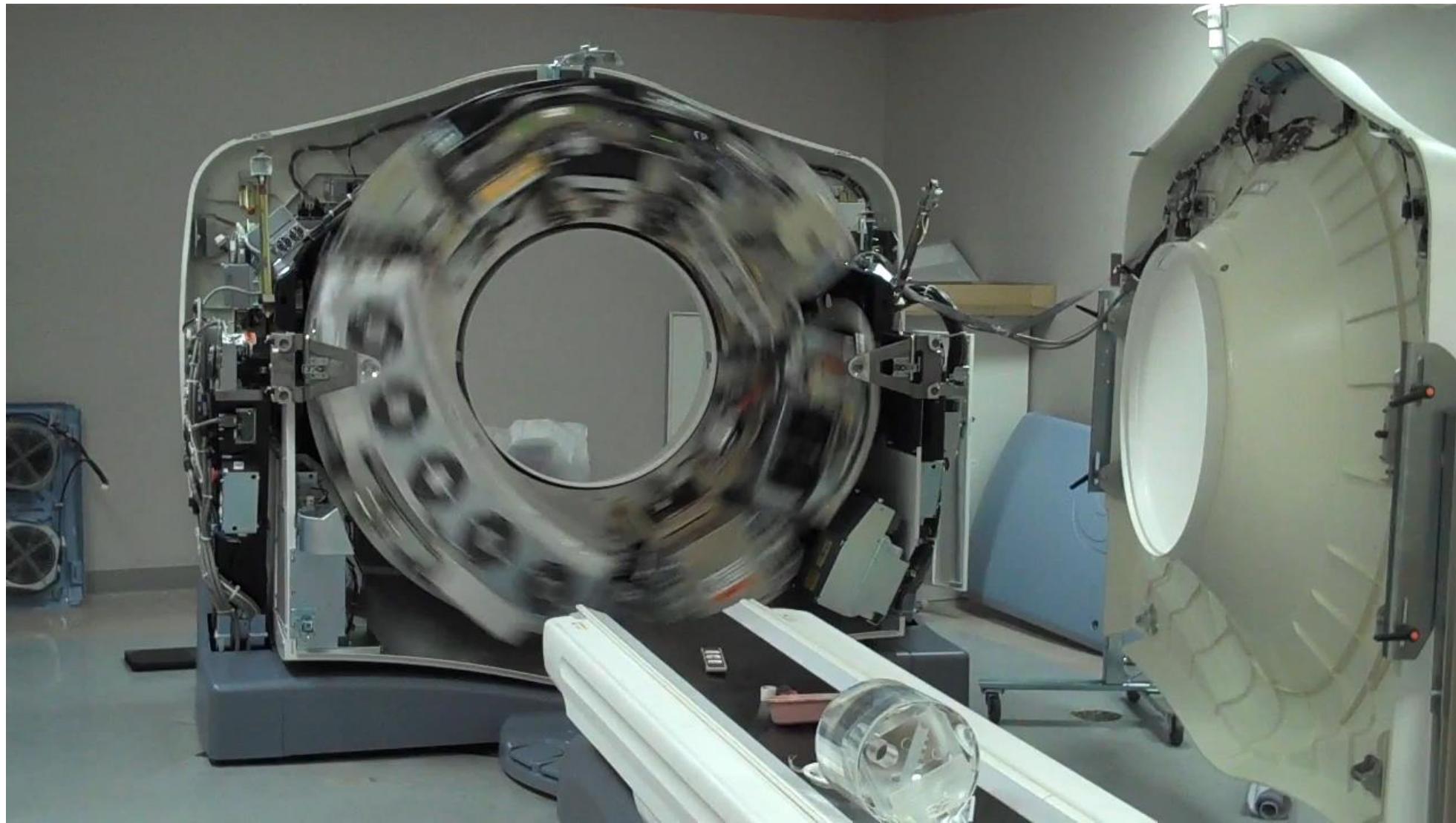
Multi-slice Spiral CT

► CT scanner



- A modern (2006) CT scanner with the cover removed, demonstrating the principle of operation.
- The X-ray tube and the detectors are mounted on a ring shaped gantry.
- The patient lies in the center of the gantry while the gantry rotates around them.
- This arrangement, a broad fan-shaped X-ray beam with rotating source and detectors, is the 'third-generation' configuration.
- This scanner is capable of helical scanning -the gantry is able to rotate freely while the patient moves continuously through the plane of the scan.
- Power (up to 150 kW) and data (up to 5 Gbps) are transferred to and from the gantry using slip rings (not shown).

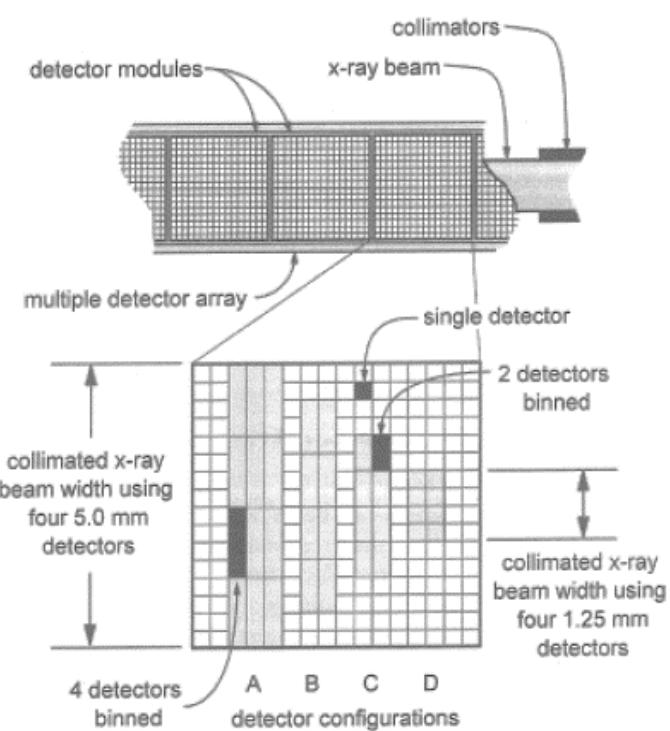
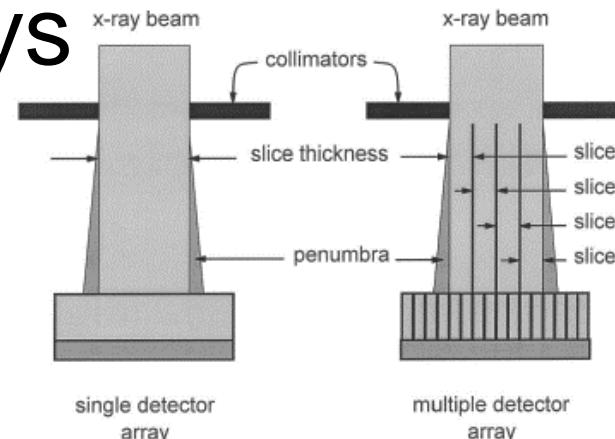
► CT scanner inside



► Multi Slice CT: Multiple Detector Arrays

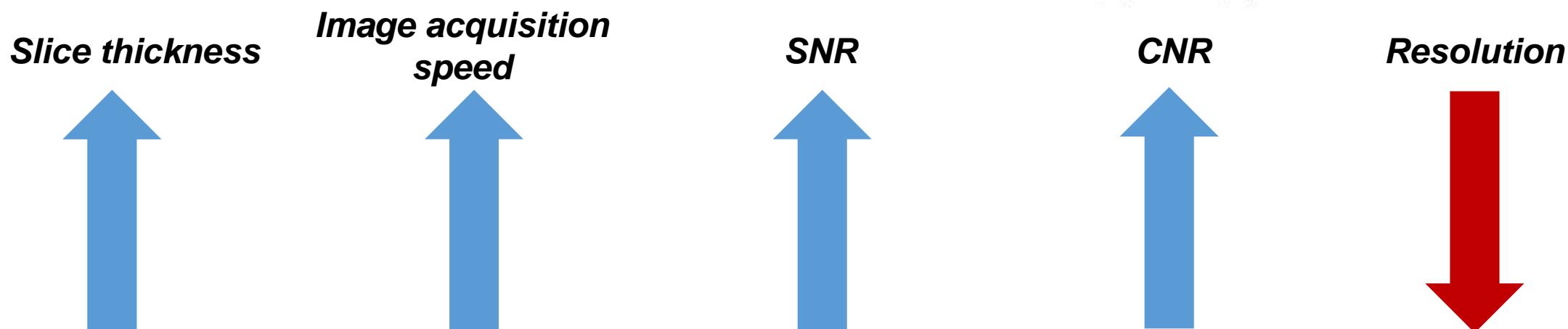
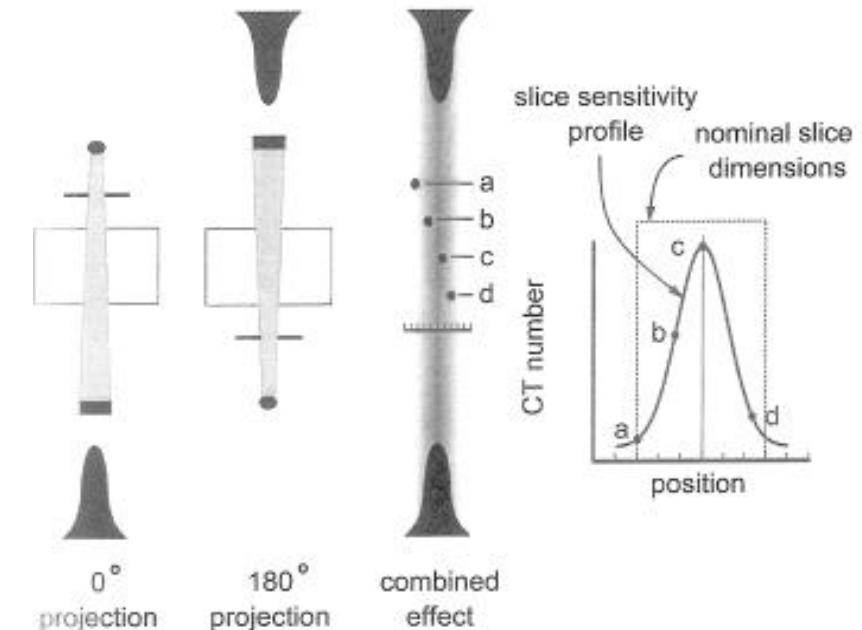
- Acquisition of multiple images per scan
- Electronic post-patient collimation:
slice thickness -detectors vs. collimator
- **Faster** volume acquisition times
- Better bolus tracking and thin slices for 3D
- Multi Slice CT uses 3rd generation platform

Detector	Beam thickness (mm)	Number of rotations	Total scan time (sec)
1 × 1.25	1.3	160	128
4 × 1.25	5	40	32
8 × 1.25	10	20	16
16 × 1.25	20	10	8

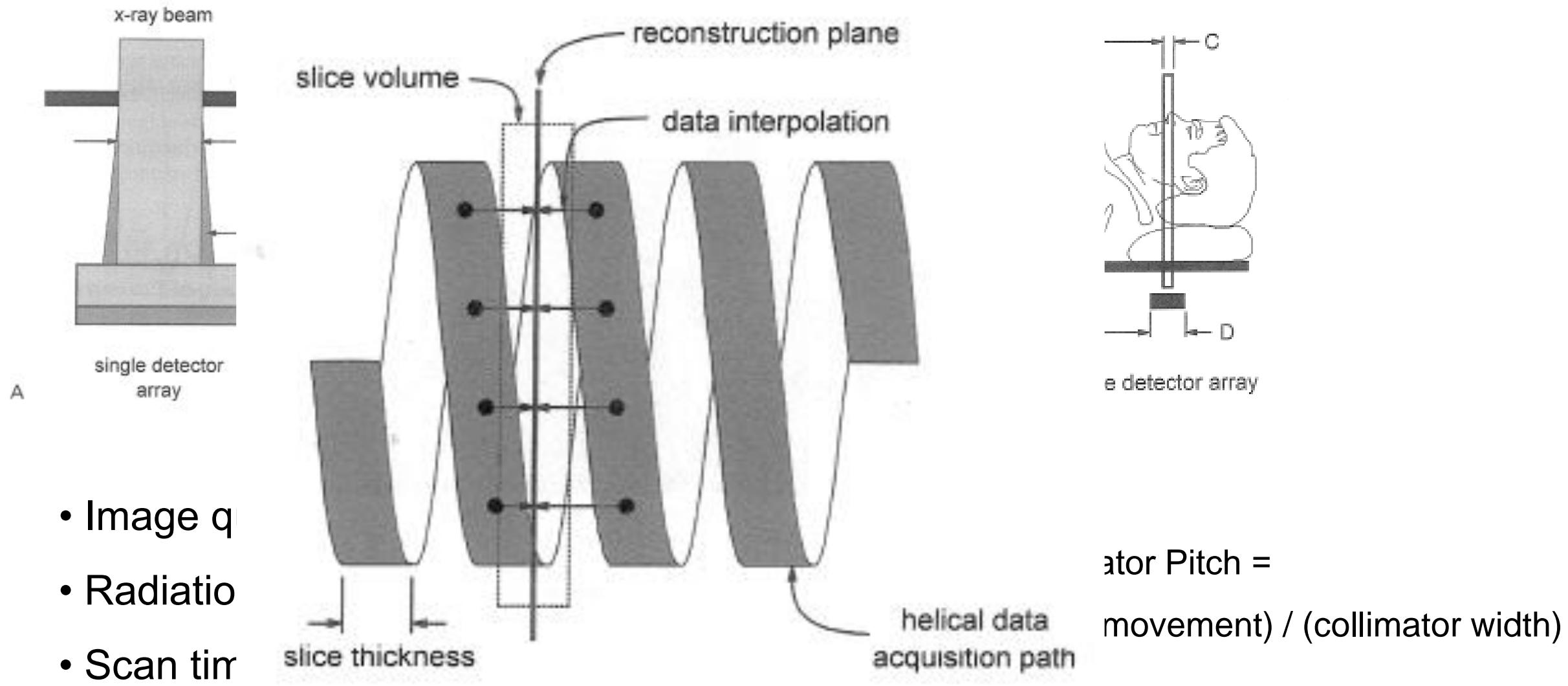


► Slice thickness, Detector and Collimator

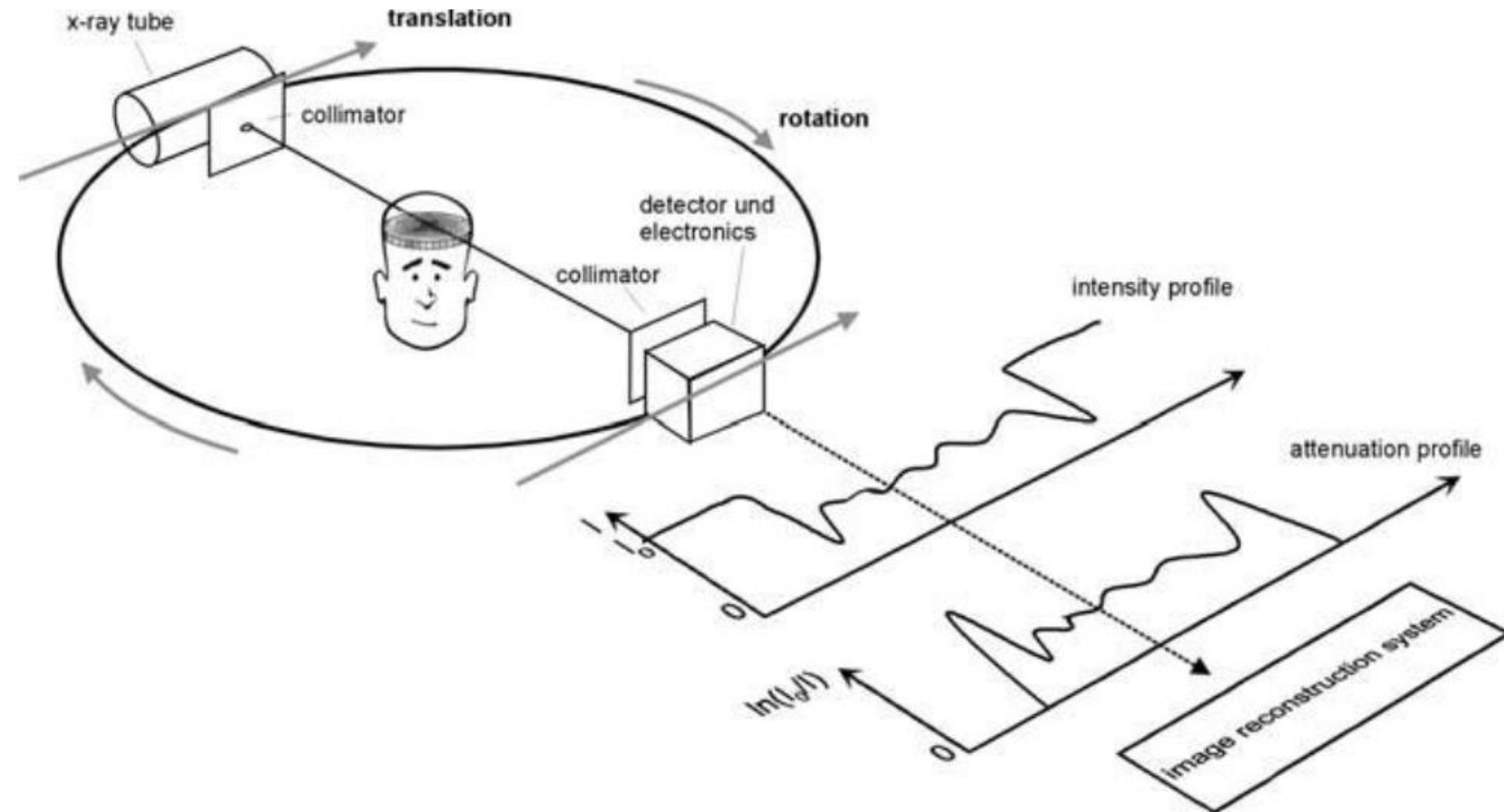
- Design parameters for CT scanner
- Slice thickness ~ number of x-ray photons
- $(\text{Number of x-ray photons})^{1/2} \sim \text{SNR}$
- SNR ~ CNR
- Slice thickness ~ 1 / Resolution



► Slice thickness, Detector and Collimator



► Image Reconstruction: Projection of Data



► Image Reconstruction: Projection of Data

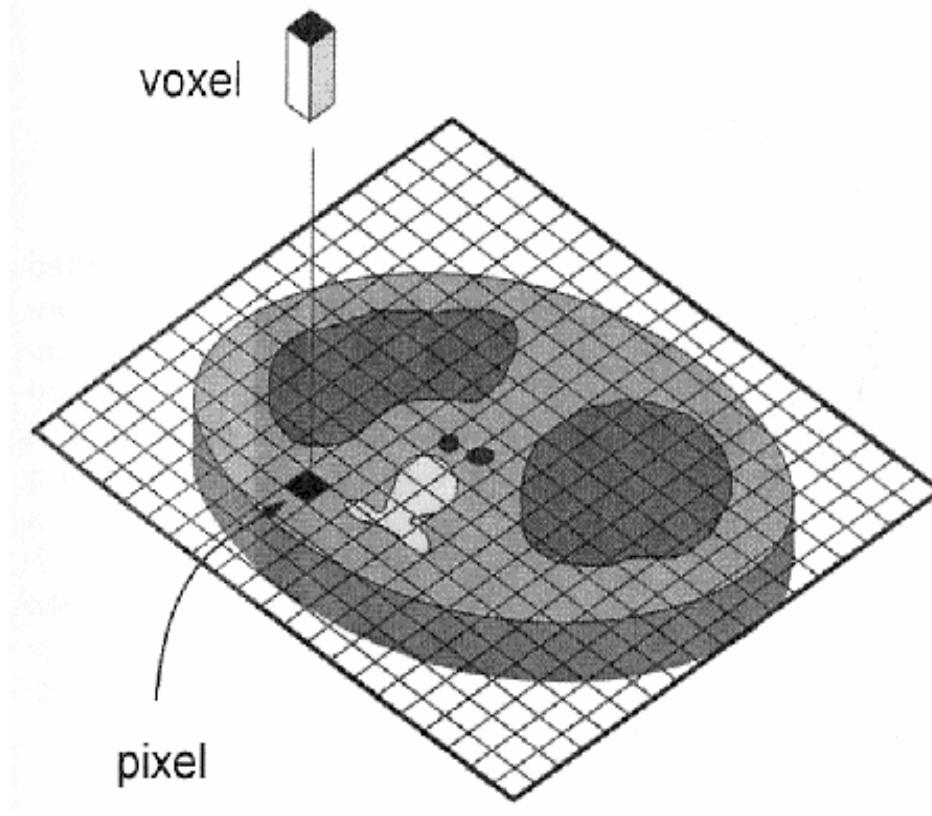
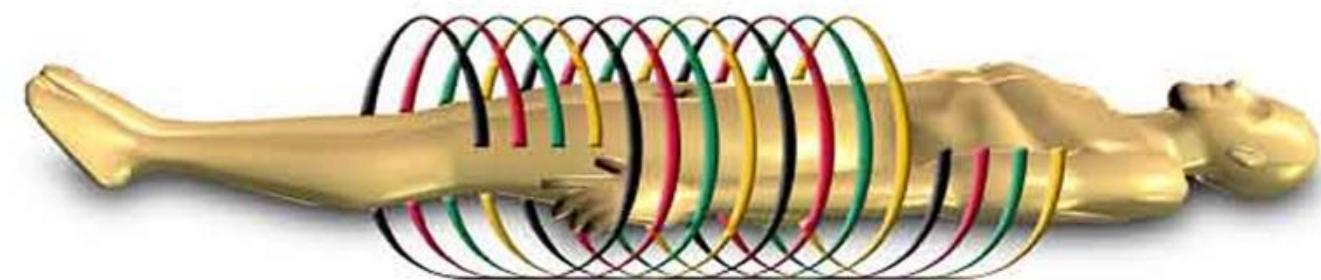


FIGURE 13-2. A pixel (picture element) is the basic two-dimensional element of a digital image. Computed tomographic (CT) images are typically square arrays containing 512×512 pixels, each pixel representing 4,096 possible shades of gray (12 bits). Each pixel in the CT image corresponds to a voxel (volume element) in the patient. The voxel has two dimensions equal to the pixel in the plane of the image, and the third dimension represents the slice thickness of the CT scan.

4,096 values
(12bits/voxel)



▶ Tomographic Reconstruction

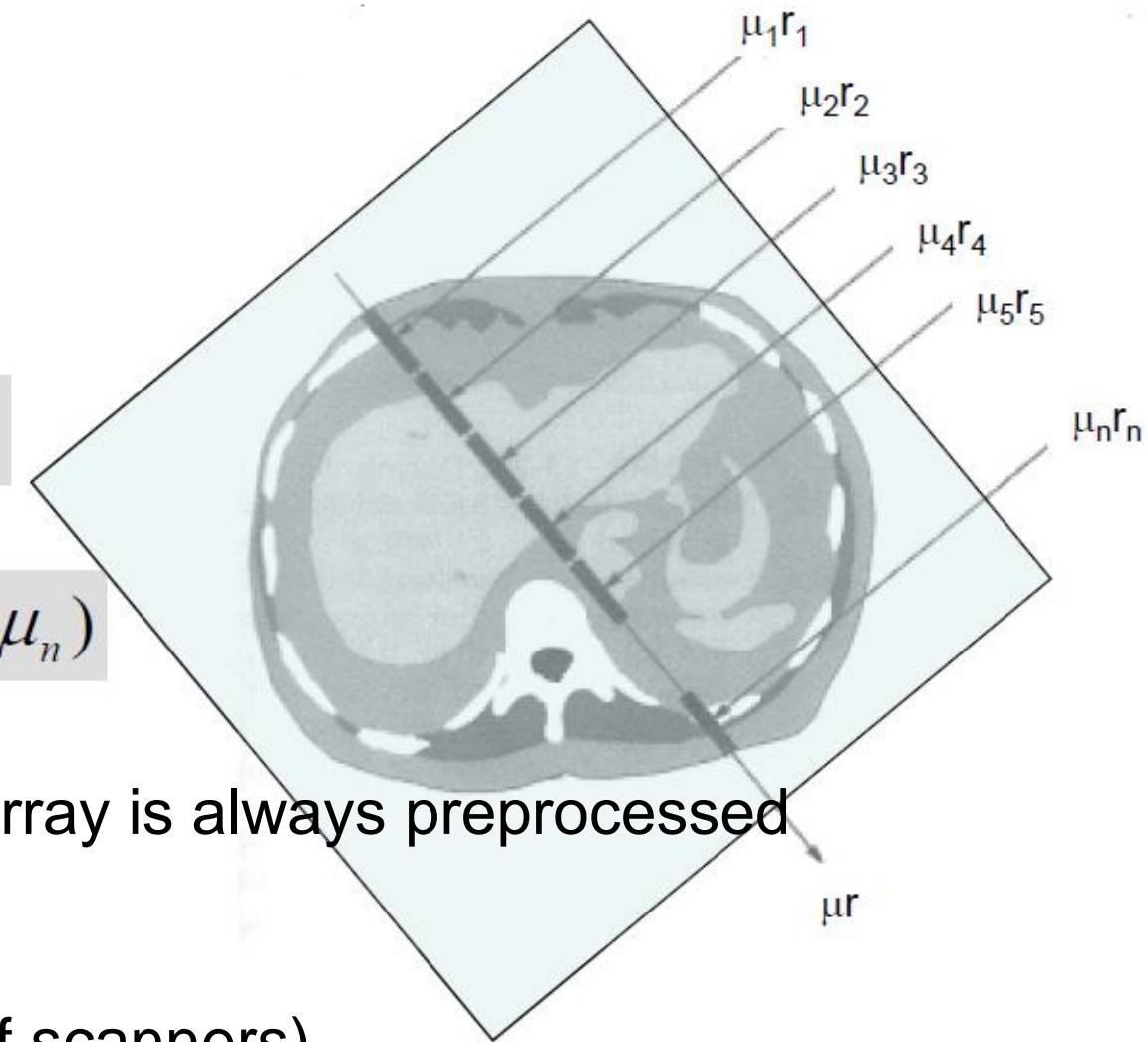
- Rays and Views: the sinogram
- Preprocessing the data
- Iterative reconstruction techniques
- Simple backprojection reconstruction
- Filtered backprojection reconstruction

► Step 1: Preprocessing

$$\ln\left(\frac{I_0}{I}\right) = \mu r$$

$$\mu r = \mu_1 r_1 + \mu_2 r_2 + \mu_3 r_3 + \dots + \mu_n r_n$$

$$\mu r = \Delta r (\mu_1 + \mu_2 + \mu_3 + \mu_4 + \dots + \mu_n)$$



- The data acquired by the detector array is always preprocessed
- Preprocessing requires calibration
(compare 3rd and 4th generation of scanners)
- CT number

► Reminder: Dirac Delta Function

- The delta function is defined through two integral relations

$$\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \delta(x - x_0, y - y_0) dx dy = 1$$

$$\delta(x - x_0, y - y_0) = 0 \text{ for } x \neq x_0, y \neq y_0 \quad (1)$$

$$g(x, y) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} g(\xi, \eta) \delta(x - \xi, y - \eta) d\xi d\eta \quad (2)$$

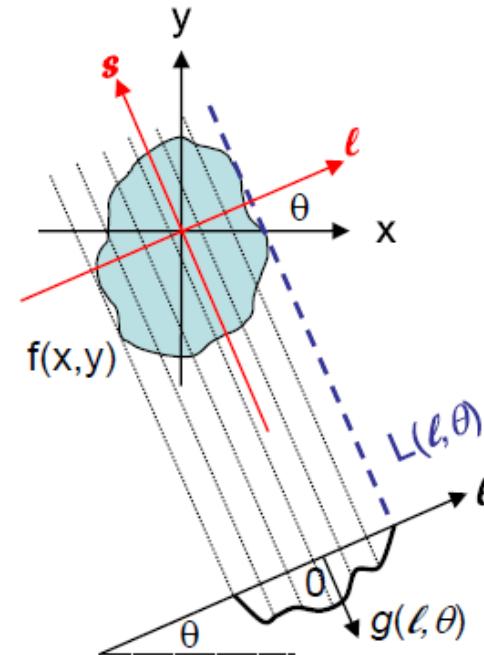
- The second property (sifting property) of the delta function is what makes it so powerful for math and engineering in general and imaging in particular.

▶ Step 1: Preprocessing

$$L(l, \theta) = \{(x, y) \mid x \cos \theta + y \sin \theta = l\}$$

$$L(l, \theta) = l - x \cos \theta - y \sin \theta$$

(or $L(l, \theta) = x \cos \theta + y \sin \theta - l$)



$$g(l, \theta) = \int_{-\infty}^{\infty} f(x(s), y(s)) ds$$

$$x(s) = l \cdot \cos \theta - s \cdot \sin \theta$$

$$y(s) = l \cdot \sin \theta + s \cdot \cos \theta$$

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

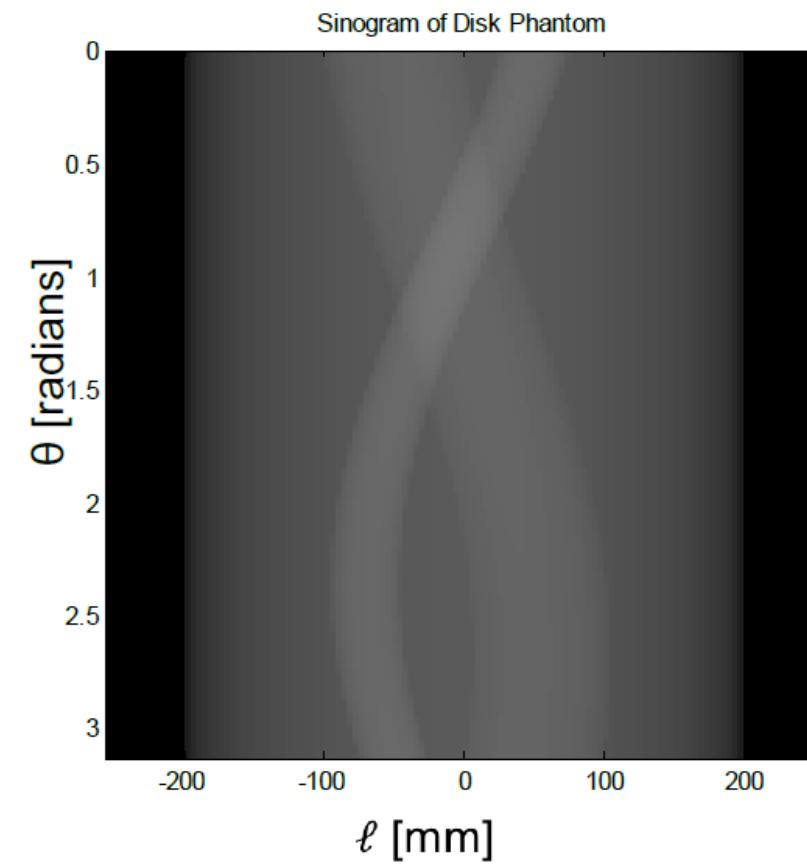
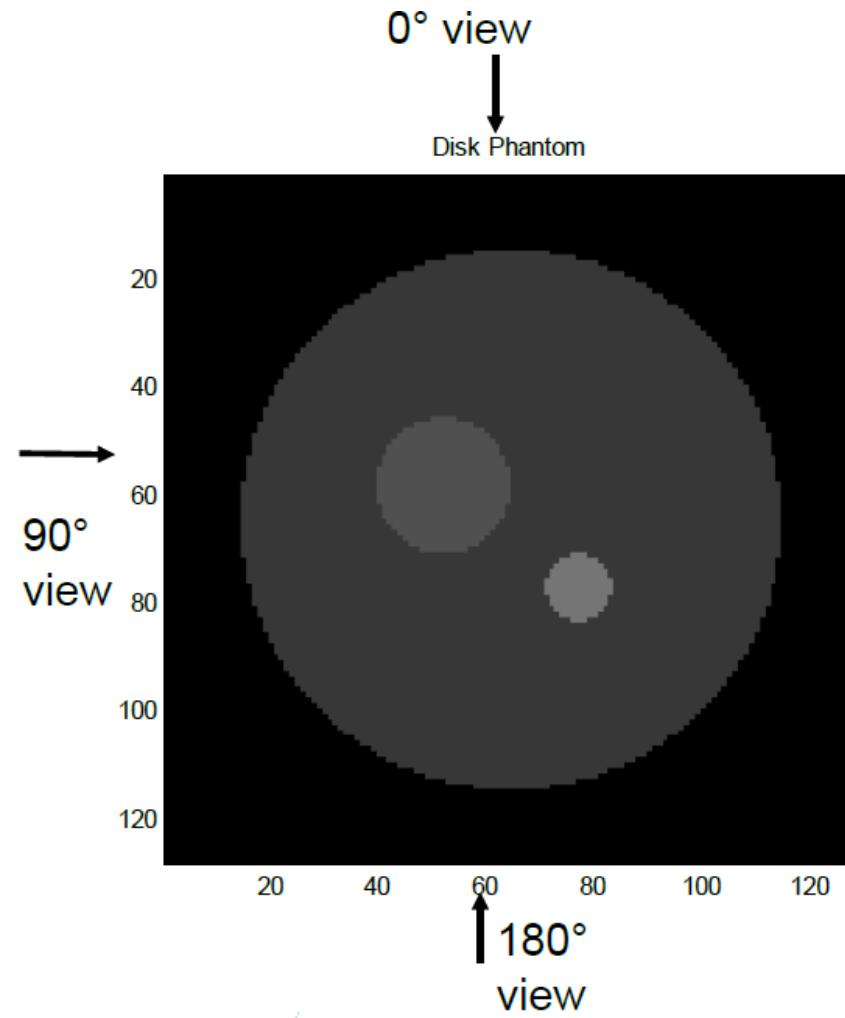
Rotate & Translate

$$f(x, y) = \mu(x, y, \bar{E})$$

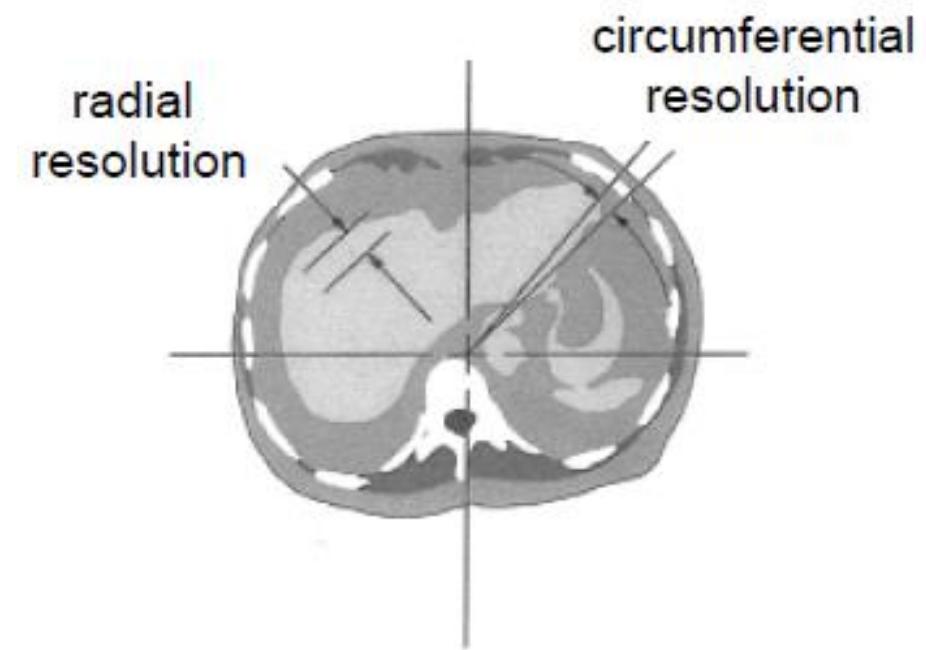
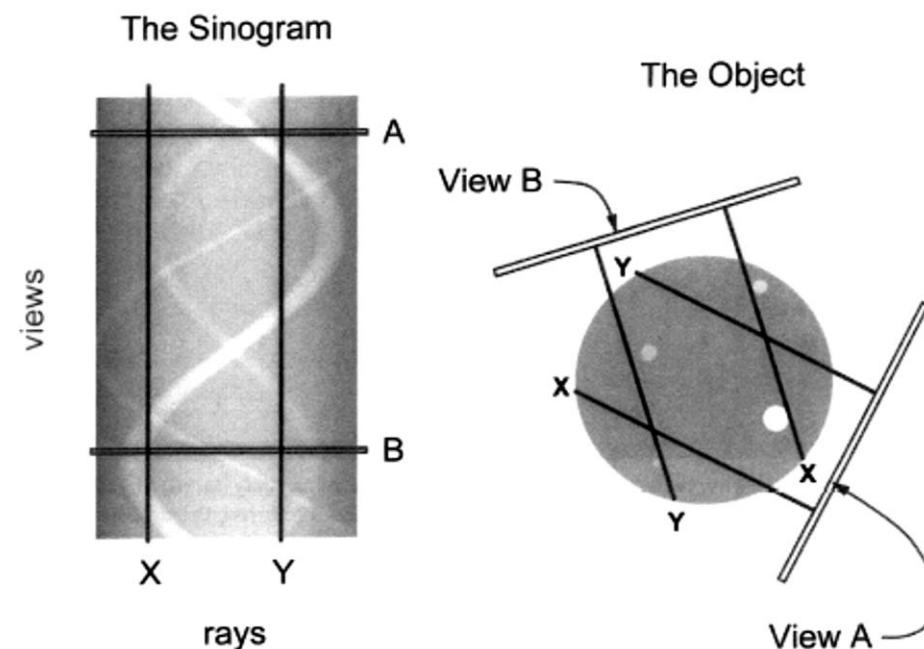
$$g(l, \theta) = -\ln\left(\frac{I_0}{I}\right)$$

► Step 2: Sinogram

- Raw data displayed before reconstruction.



► Step 2: Sinogram



- Data display before reconstruction
- Radial resolution: number of rays
 - “Ray” = 1 transmission measurement made through patient at 1 moment in time
- Circumferential resolution: number of views

► Reconstruction Techniques in CT

- Algebraic Reconstruction
- Fourier Reconstruction
- Projection Reconstruction
 - Backprojection
 - Filtered backprojection
 - Convolution + backprojection
- Further flavors
 - Parallel beam geometry
 - Fan beam geometry
 - Cone beam geometry
 - Helical CT reconstruction

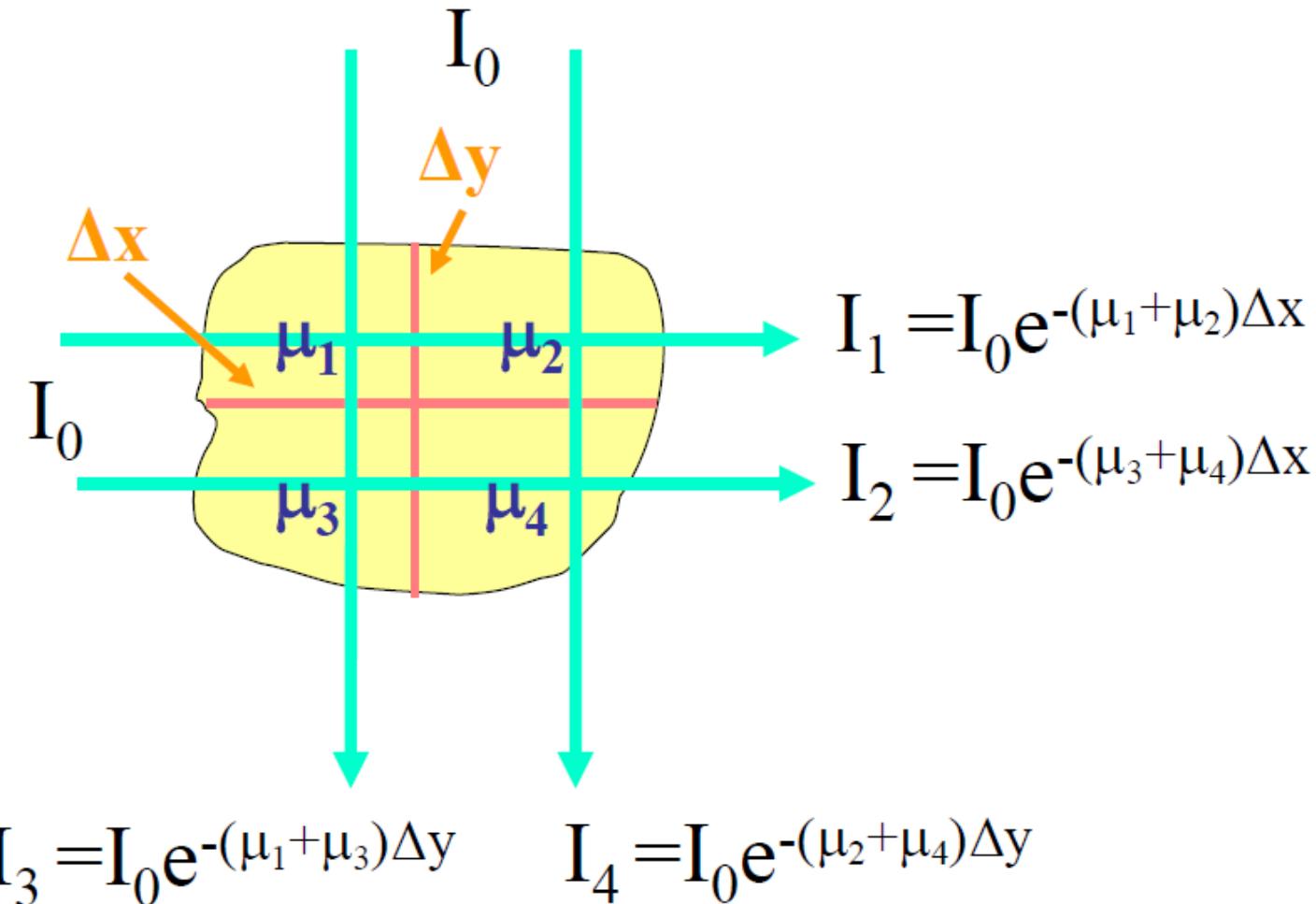
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► Direct Reconstruction

- Matrix Inversion
- Iterative Procedure
 - Algebraic Reconstruction Technique (ART)
- Best match of reconstruction and measured projections

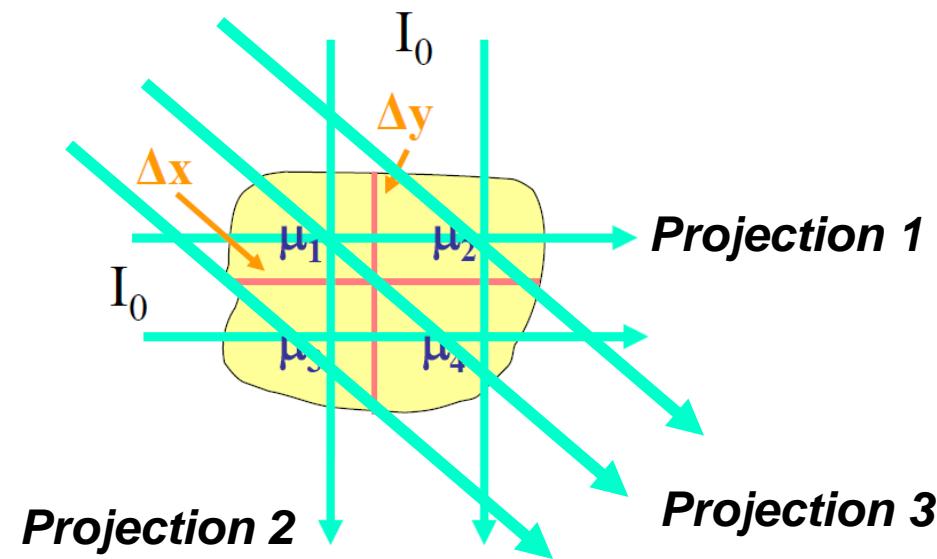
► Matrix Inversion



4 equations, 4 unknown

► Overdetermined system (#'s Equations > #'s variables)

- Each projection contains Q data points, and there are P rotations.
- For each projection, the signal intensity depends upon the composite attenuation coefficient (μ) of the tissue corresponding to the particular beam path.
- For the simple case shown below, a two-by-two matrix is imaged by projection 1 (I_1 and I_2), projection 2 (I_3 and I_4) and projection 3 (I_5 , I_6 and I_7).
- The intensities of the projections can be expressed in terms of the linear attenuation coefficients:



$I_1 = I_0 e^{-(\mu_1 + \mu_2)x}$	Projection 1
$I_2 = I_0 e^{-(\mu_3 + \mu_4)x}$	Projection 2
$I_3 = I_0 e^{-(\mu_1 + \mu_3)x}$	Projection 3
$I_4 = I_0 e^{-(\mu_2 + \mu_4)x}$	
$I_5 = I_0 e^{-(\mu_3 x \sqrt{2})}$	
$I_6 = I_0 e^{-(\mu_1 + \mu_4)x \sqrt{2}}$	
$I_7 = I_0 e^{-(\mu_2 x \sqrt{2})}$	

Projection 1

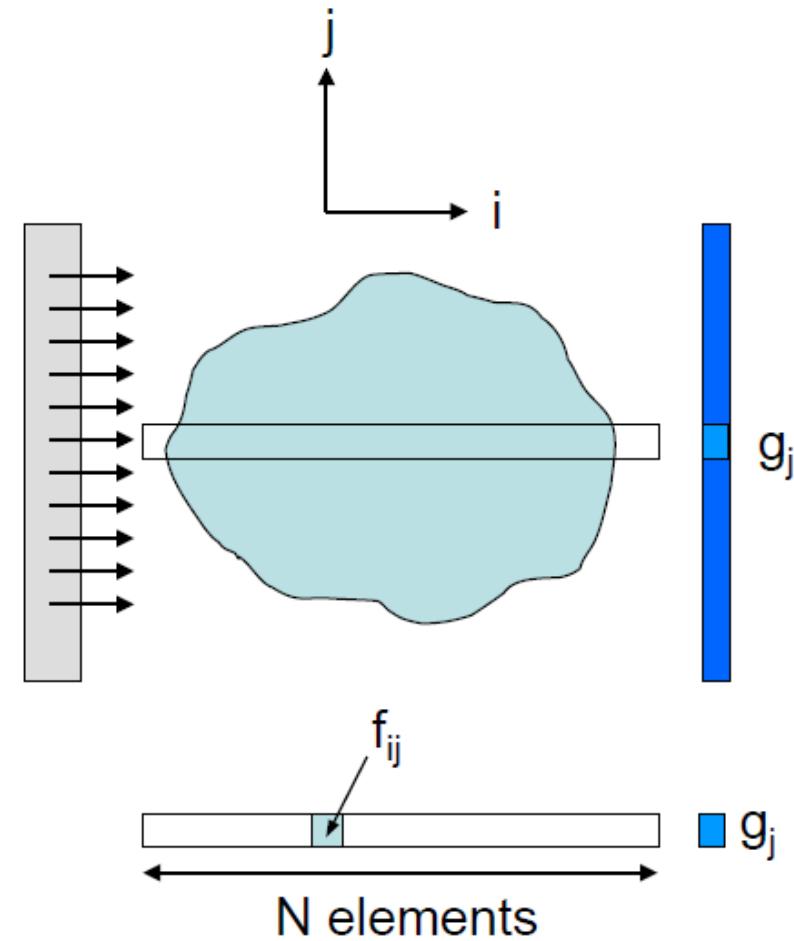
Projection 2

Projection 3

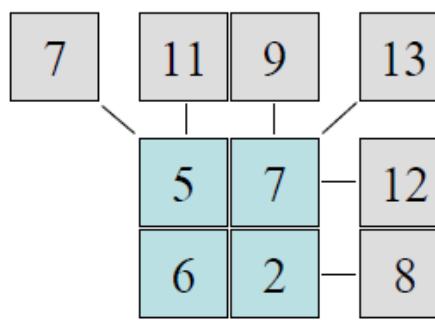
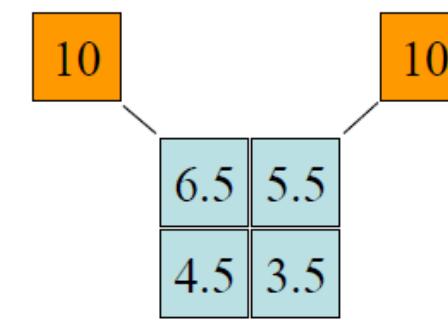
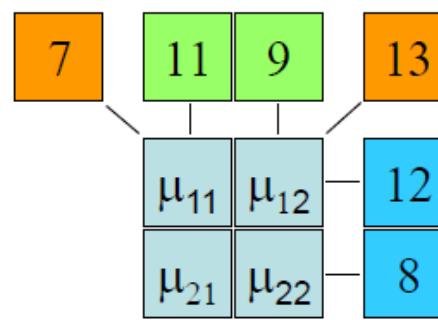
► Algebraic Reconstruction Technique (ART)

- Direct matrix inversion is impractical
- Solve linear equations iteratively

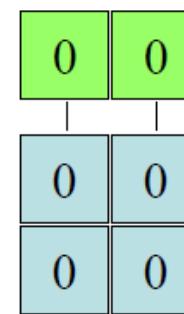
$$g_j - \sum_{i=1}^N f_{ij} = 0$$
$$g_j - \sum_{i=1}^N f_{ij}^k = \frac{1}{\tau} (f_{ij}^{k+1} - f_{ij}^k)$$
$$f_{ij}^{k+1} = f_{ij}^k + \frac{g_j - \sum_{i=1}^N f_{ij}^k}{N}, \quad \tau = \frac{1}{N}$$



▶ Example



$$f_{ij}^{k+1} = f_{ij}^k + \frac{g_j - \sum_{i=1}^N f_{ij}^k}{N}$$



Initial guess

Vertical rays

$$f_{11}^1 = 0 + \frac{11-0}{2} = 5.5$$

$$f_{21}^1 = 0 + \frac{11-0}{2} = 5.5$$

$$f_{12}^1 = 0 + \frac{9-0}{2} = 4.5$$

$$f_{22}^1 = 0 + \frac{9-0}{2} = 4.5$$

Horizontal rays

$$f_{11}^2 = 5.5 + \frac{12-10}{2} = 6.5$$

$$f_{21}^2 = 5.5 + \frac{8-10}{2} = 4.5$$

$$f_{12}^2 = 4.5 + \frac{12-10}{2} = 5.5$$

$$f_{22}^2 = 4.5 + \frac{8-10}{2} = 3.5$$

Diagonal rays

$$f_{11}^3 = 6.5 + \frac{7-10}{2} = 5$$

$$f_{21}^3 = 4.5 + \frac{13-10}{2} = 6$$

$$f_{12}^3 = 5.5 + \frac{13-10}{2} = 7$$

$$f_{22}^3 = 3.5 + \frac{7-10}{2} = 2$$

► Reconstruction Techniques in CT

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- Fourier Reconstruction
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- Further flavors
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► Reminder: 2-D Fourier Reconstruction

- The continuous 2-D Fourier transform $G(u,v)$ of a function $g(x,y)$ is defined as

$$G(u, v) = F[g(x, y)] = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} g(x, y) e^{-i2\pi(ux+vy)} dx dy$$

where u and v are often referred to as spatial frequencies by analogy to 1-D transform

- Physically, each component $G(u,v)$ can be considered as component of the function $g(x,y)$ resulting from a plane wave of wave vector k , and propagating along direction θ , where

$$k = 2\pi\sqrt{u^2 + v^2} \quad \theta = \tan^{-1}(v/u)$$

- The inverse transform is similarly defined as

$$g(x, y) = F^{-1}[G(u, v)] = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} G(u, v) e^{i2\pi(ux+vy)} du dv$$

► Reminder: Fourier Transform Relations

- An impulse in one domain results in a constant level at all coordinates of the transform domain (**delta function response**)

$$F[\delta(x, y)] = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \delta(x, y) e^{-i2\pi(ux+vy)} dx dy = 1$$

- The Fourier Transform operation is linear (**linearity**)

$$F[\alpha \cdot g(x, y) + \beta \cdot h(x, y)] = \alpha \cdot F[g(x, y)] + \beta \cdot F[h(x, y)]$$

- Translation of a function in one space introduces a linear phase shift in the transform domain (**shift**)

$$F[g(x - \alpha, y - \beta)] = G(u, v) e^{-i2\pi(\alpha u + \beta v)}$$

- If $g(x, y) = g_x(x)g_y(y)$ (**separability**), then

$$F[g(x, y)] = F[g_x(x)g_y(y)] = F_x[g_x(x)]F_y[g_y(y)]$$

▶ Projection-Slice Theorem

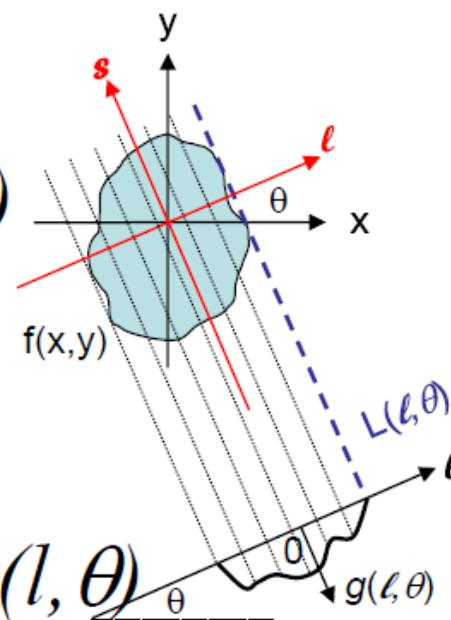
Ref. astra toolbox

▶ Projection-Slice Theorem

- Let start with the projection $g(l, \theta)$

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

- Now let's consider a 1-D FT of $g(l, \theta)$



$$\begin{aligned} G(\rho, \theta) &= \int g(l, \theta) e^{-i2\pi\rho l} dl = \\ &= \int \left[\int \int f(x, y) \delta(x \cos \theta + y \sin \theta - l) dx dy \right] e^{-i2\pi\rho l} dl = \\ &= \int \int f(x, y) \int e^{-i2\pi\rho l} \delta(x \cos \theta + y \sin \theta - l) dl dx dy = \\ &= \int \int f(x, y) e^{-i2\pi\rho(x \cos \theta + y \sin \theta)} dx dy \end{aligned}$$

▶ Projection-Slice Theorem

- From the previous slide

$$\begin{aligned} G(\rho, \theta) &= \int g(l, \theta) e^{-i2\pi\rho l} dl = \\ &= \int \left[\int \int f(x, y) \delta(x \cos \theta + y \sin \theta - l) dx dy \right] e^{-i2\pi\rho l} dl = \\ &= \int \int f(x, y) \int e^{-i2\pi\rho l} \delta(x \cos \theta + y \sin \theta - l) dx dy dl = \\ &= \int \int f(x, y) e^{-i2\pi\rho(x \cos \theta + y \sin \theta)} dx dy \end{aligned}$$

- Which is 2-D Fourier transform

$$F(u, v) = \int \int f(x, y) e^{-i2\pi(ux+vy)} dx dy$$

- In polar coordinates $u = \rho \cos \theta$ and $v = \rho \sin \theta$

▶ Projection-Slice Theorem

- Therefore, the desired function $f(x,y)$ can be reconstructed from $G(\rho, \theta)$

$$\begin{aligned} f(x,y) &= \iint F(u,v) e^{i2\pi(ux+vy)} du dv = \\ &= \int_0^{2\pi} d\theta \int_0^{\infty} G(\rho, \theta) e^{i2\pi\rho(x\cos\theta + y\sin\theta)} \rho d\rho \end{aligned}$$

- where

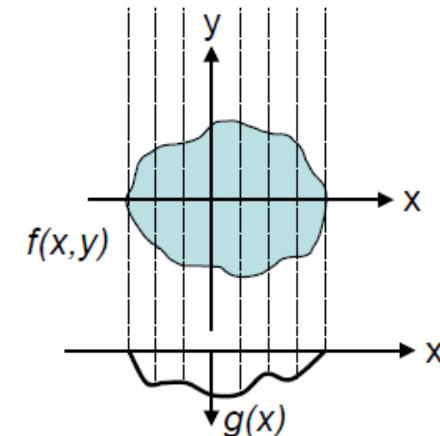
$$dudv = \rho d\rho d\theta$$

$$\begin{vmatrix} \cos\theta - \rho\sin\theta \\ \sin\theta & \rho\cos\theta \end{vmatrix} = \rho(\cos^2\theta + \sin^2\theta) = \rho$$

► Really, what does it all mean...

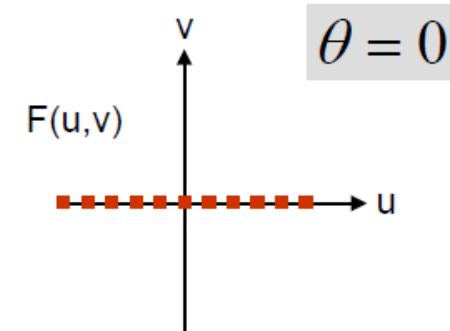
- Single projection of $f(x, y)$

$$g(x) = \int f(x, y) dy$$



- 2-D Fourier transform

$$F(u, v) = \iint f(x, y) e^{-i2\pi(ux+vy)} dx dy$$



- Along $v=0$ line

$$\begin{aligned} F(u, v=0) &= \iint f(x, y) e^{-i2\pi(ux+0y)} dx dy = \\ &= \int \left[\int f(x, y) dy \right] e^{-i2\pi ux} dx = \int g(x) e^{-i2\pi ux} dx = \\ &= F_{1-D}[g(x)] \end{aligned}$$

► Really, what does it all mean...

- Single projection of $f(x,y)$

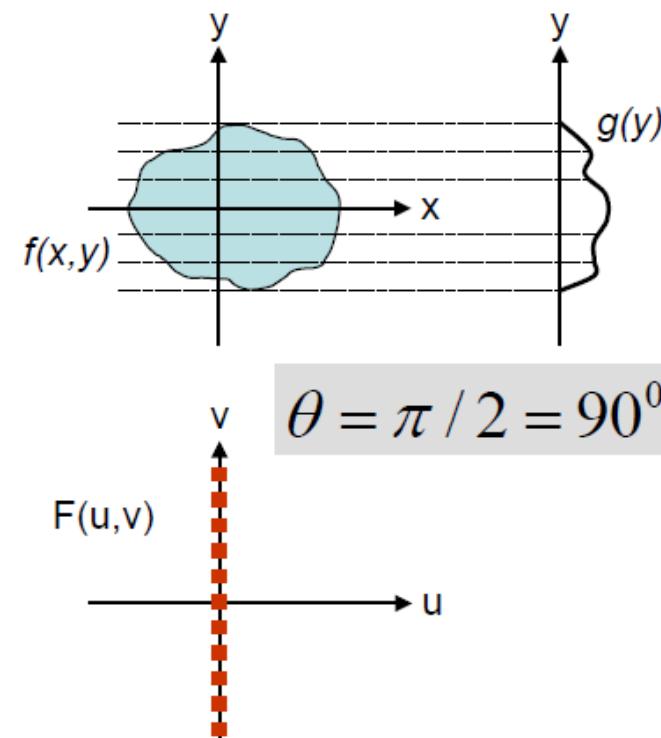
$$g(y) = \int f(x, y) dx$$

- 2-D Fourier transform

$$F(u, v) = \iint f(x, y) e^{-i2\pi(ux+vy)} dx dy$$

- Along $u=0$ line

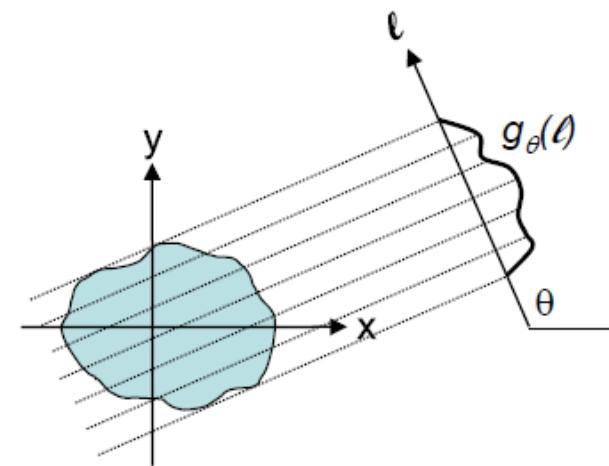
$$\begin{aligned} F(0, v) &= \iint f(x, y) e^{-i2\pi(0x+vy)} dx dy = \\ &= \int \left[\int f(x, y) dx \right] e^{-i2\pi vy} dy = \int g(y) e^{-i2\pi vy} dy = \\ &= F_{1-D}[g(y)] \end{aligned}$$



► Central Selection Theorem

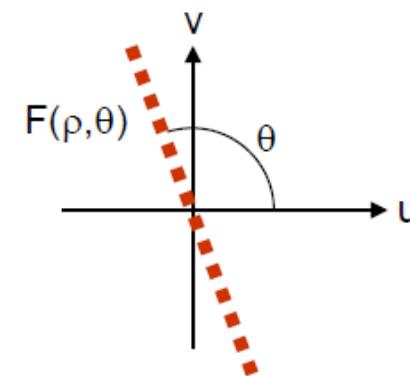
- It can be shown that

$$F(\rho, \theta) = F_{1-D}[g_\theta(l)]$$

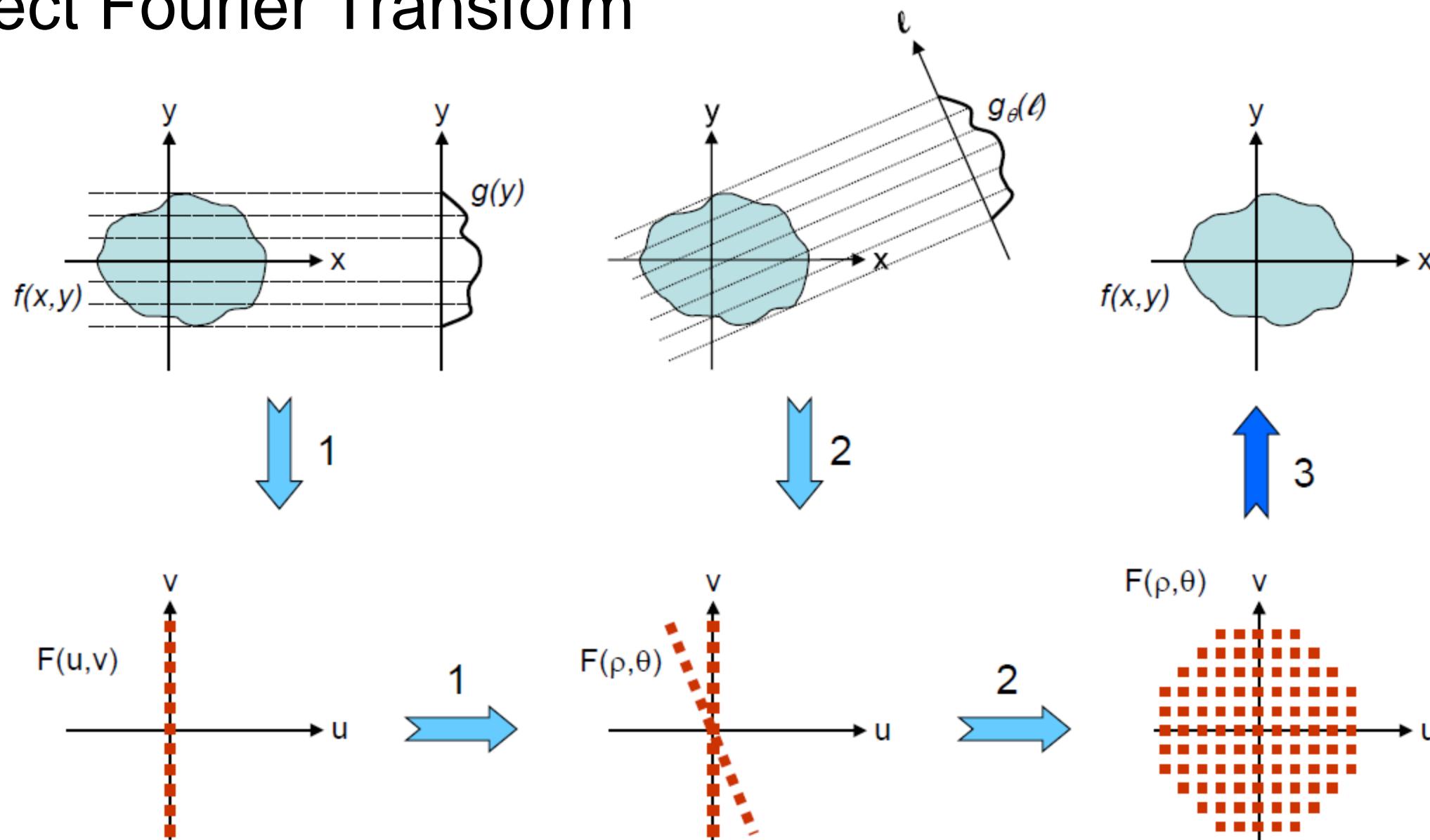


- 2-D Inverse Fourier transform

$$f(x, y) = \int_0^{2\pi} d\theta \int_0^{\infty} F(\rho, \theta) e^{i2\pi\rho(x\cos\theta + y\sin\theta)} \rho d\rho$$

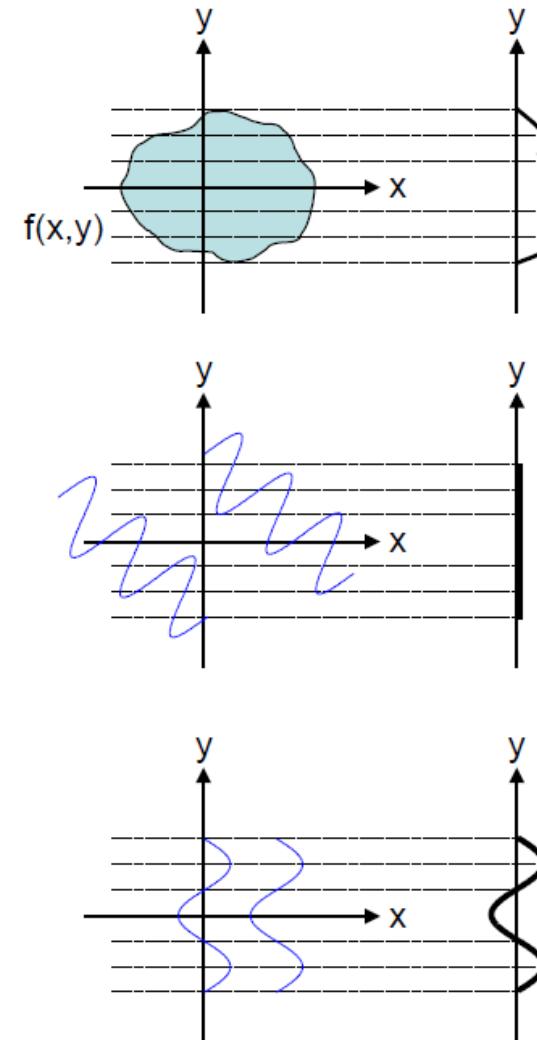


► Direct Fourier Transform

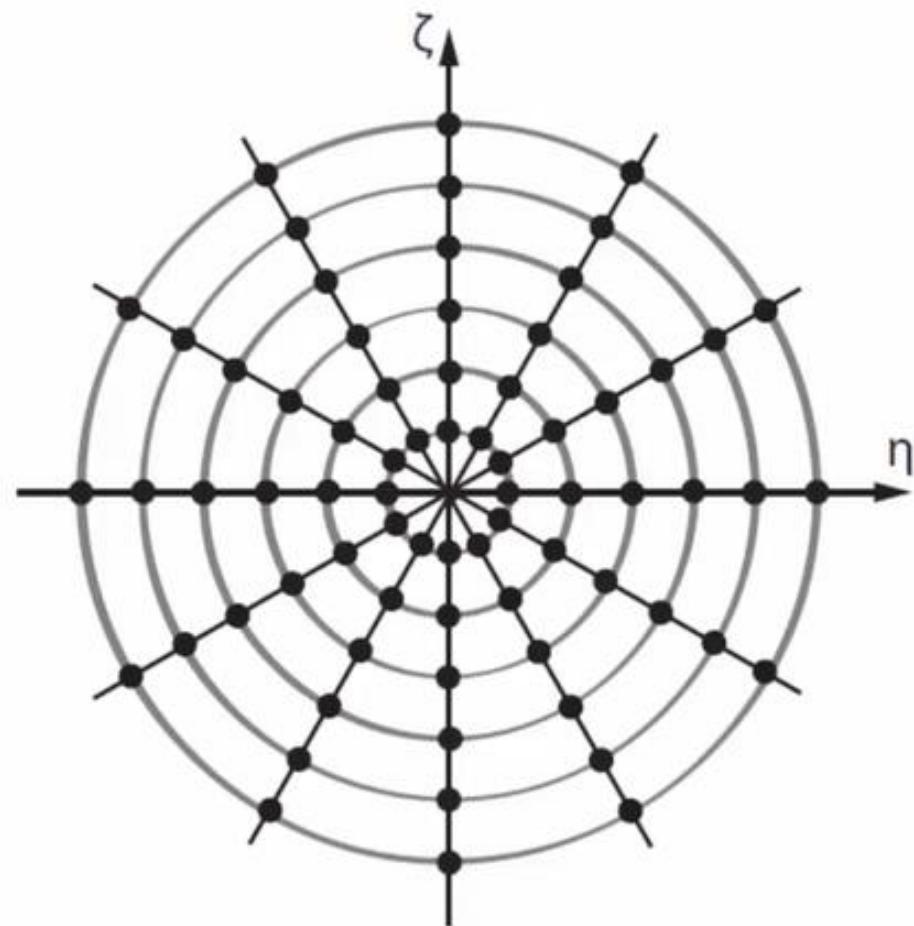


► Physical Insight into Central Selection Theorem

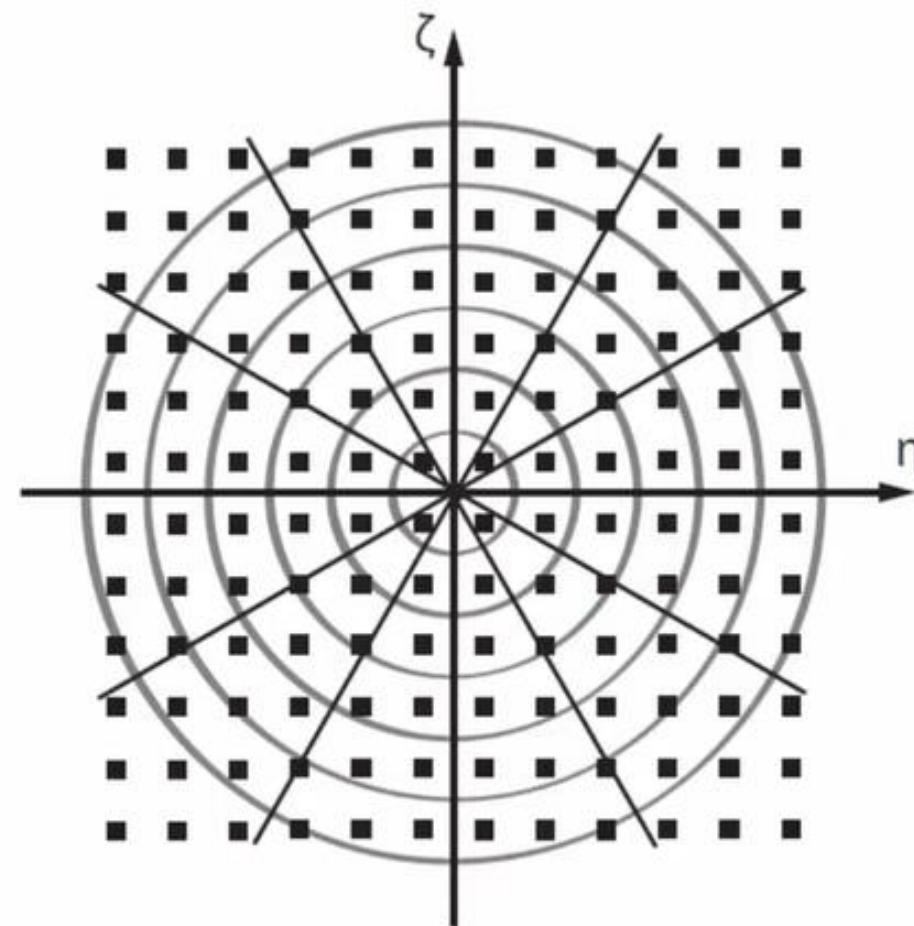
- Single projection of $f(x,y)$
- 2-D Fourier transform
 $f(x,y)$ is decomposed into array of 2-D sinusoids
- Projection of sinusoidal signals



► Different sampling approaches

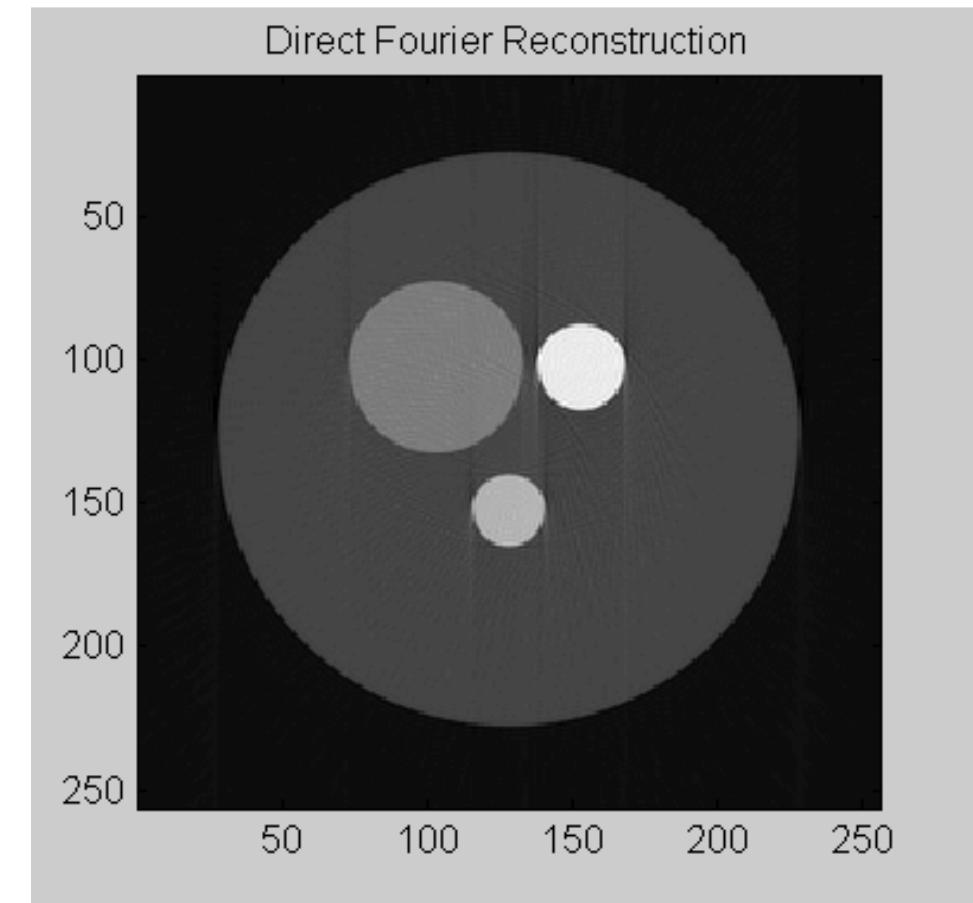
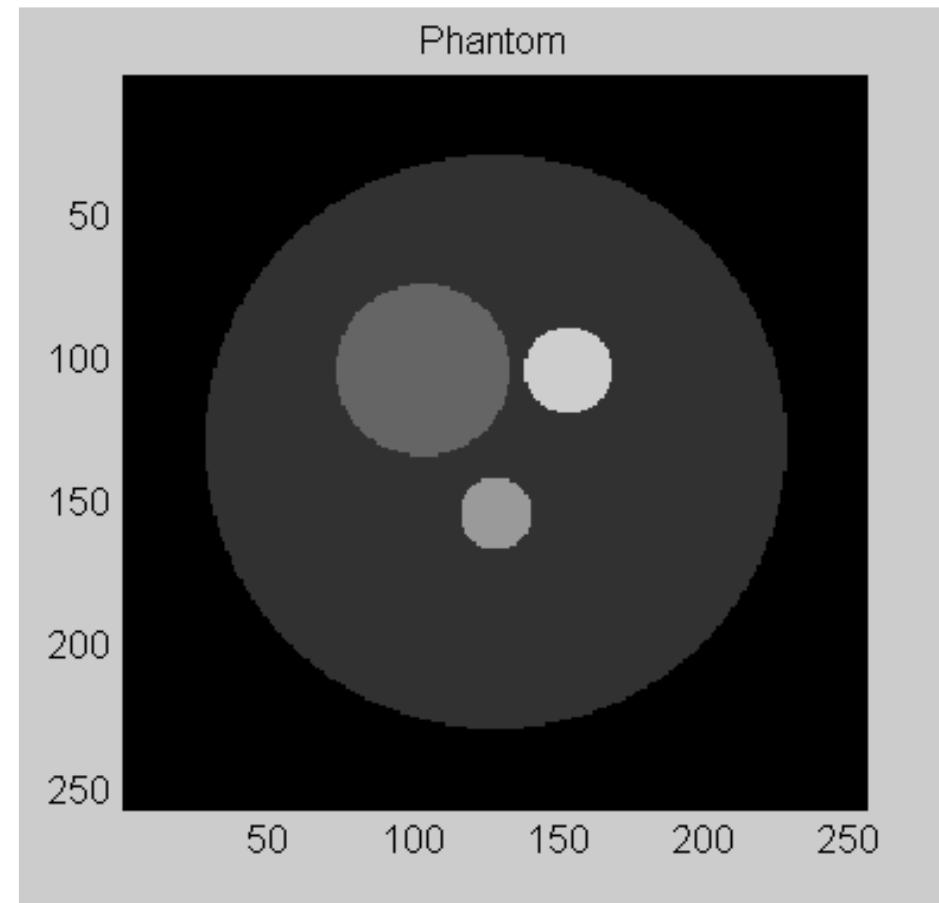


Fourier sampling with
Central selection (Fourier slice) theorem



Sampling required by
Discrete fourier transform

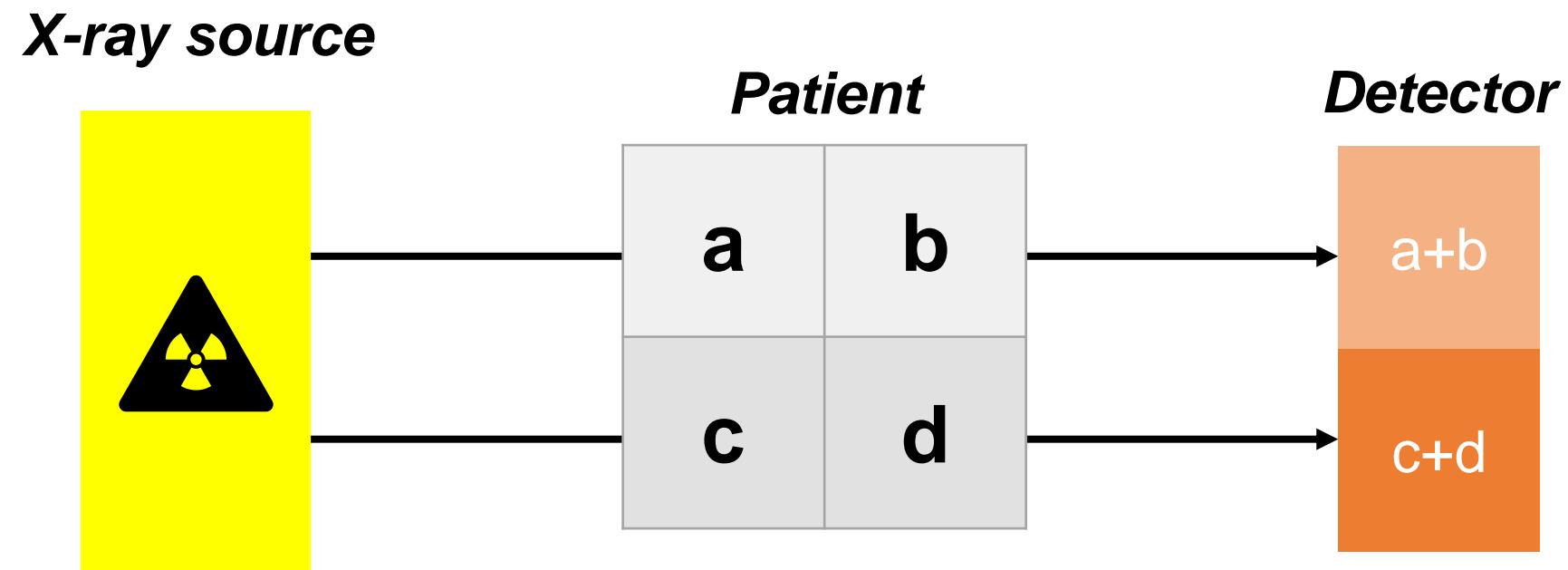
► Fourier Reconstruction: Example



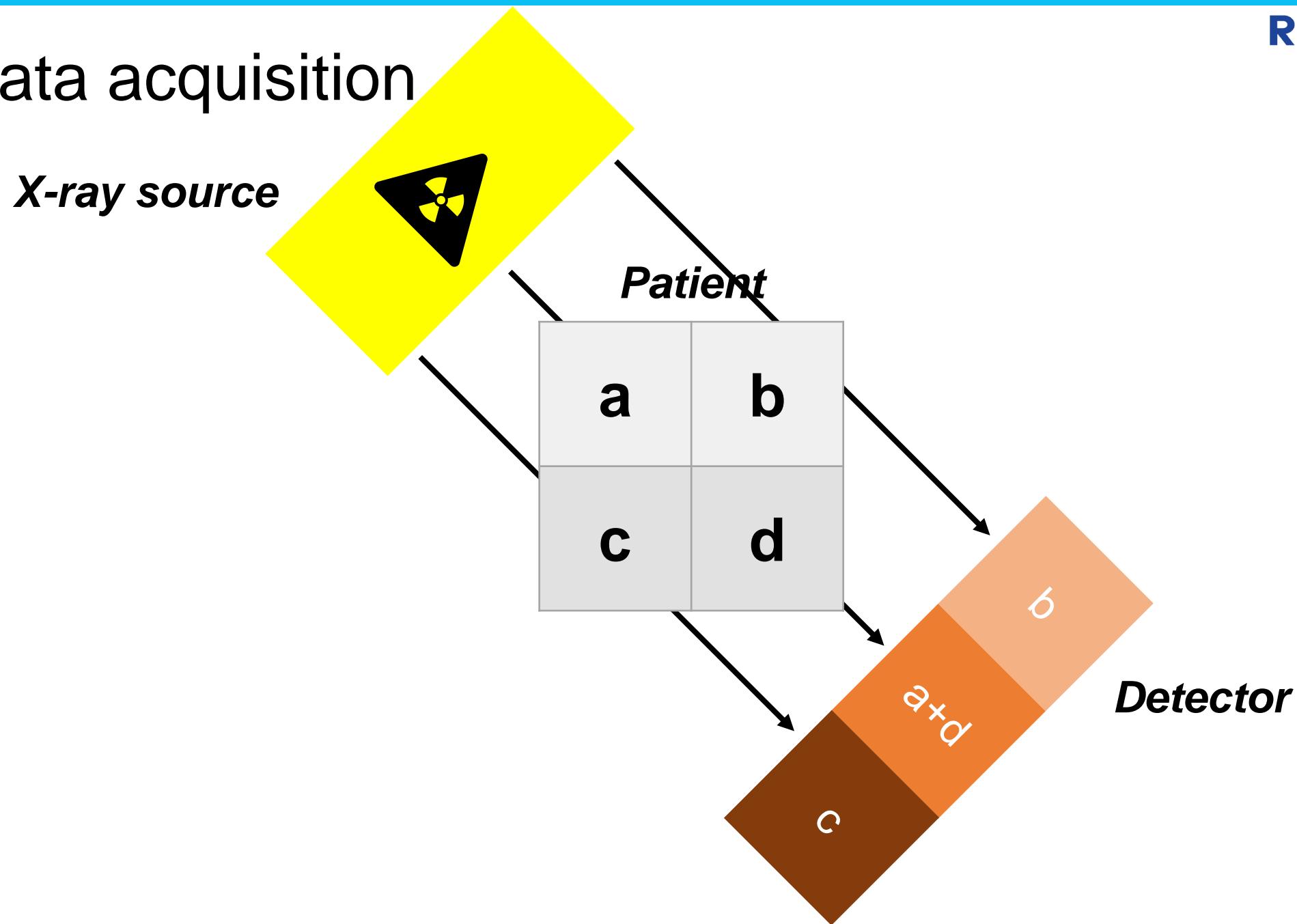
► Reconstruction Techniques in CT

- Algebraic Reconstruction
- Fourier Reconstruction
- Projection Reconstruction
 - Backprojection
 - Filtered backprojection
 - Convolution + backprojection
- Further flavors
 - Parallel beam geometry
 - Fan beam geometry
 - Cone beam geometry
 - Helical CT reconstruction

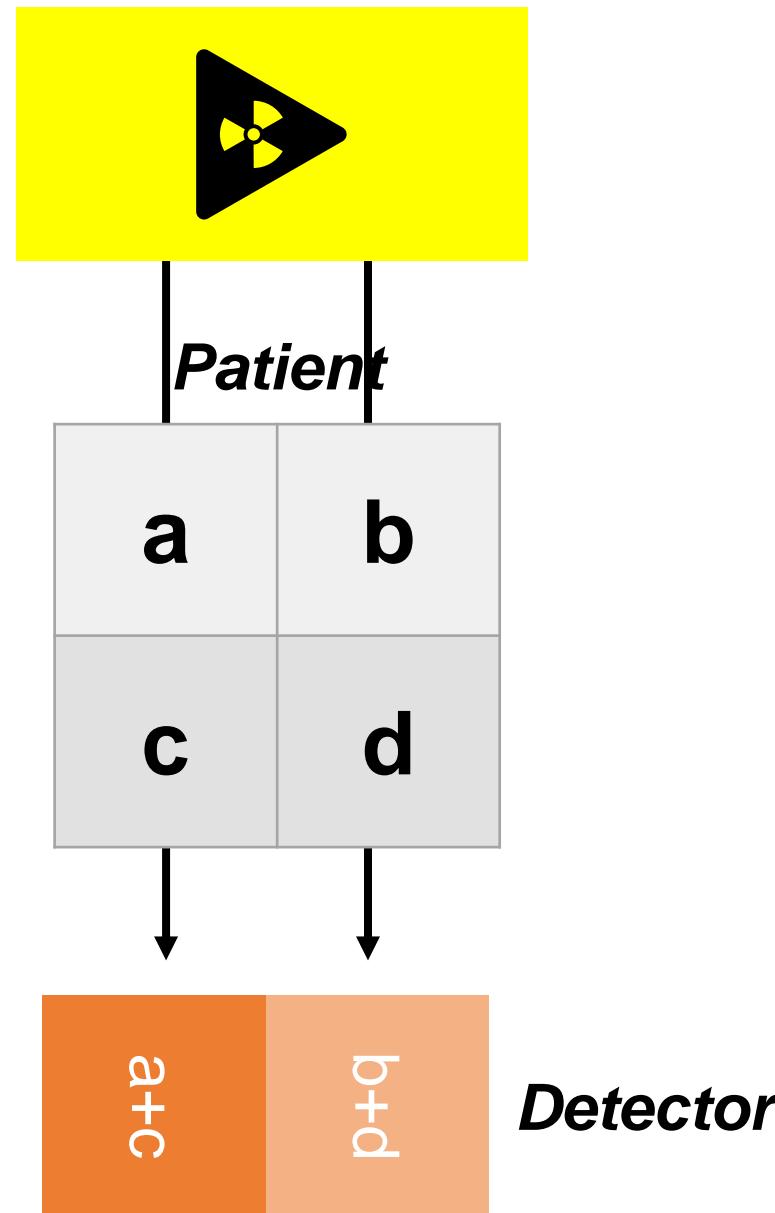
► CT data acquisition



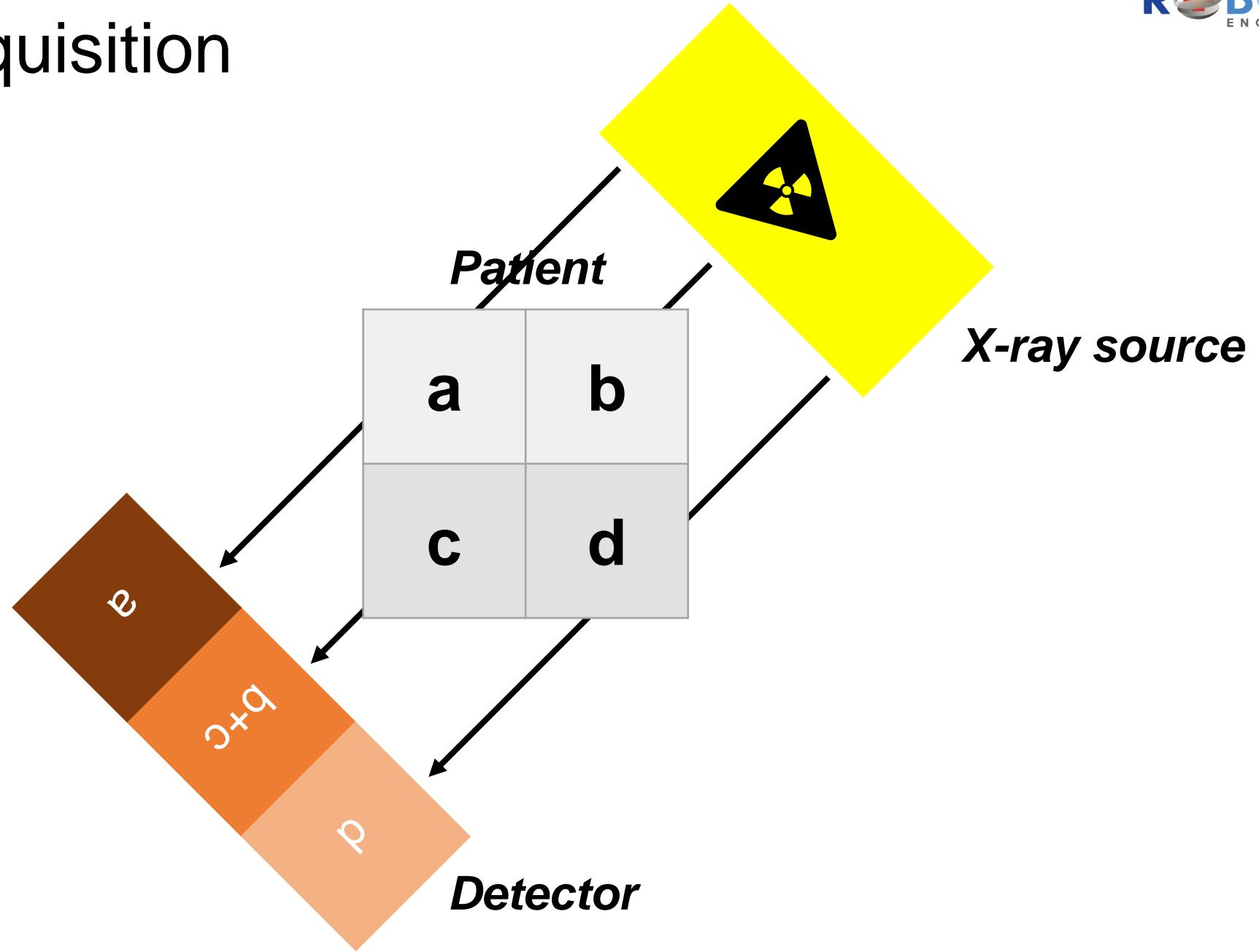
► CT data acquisition



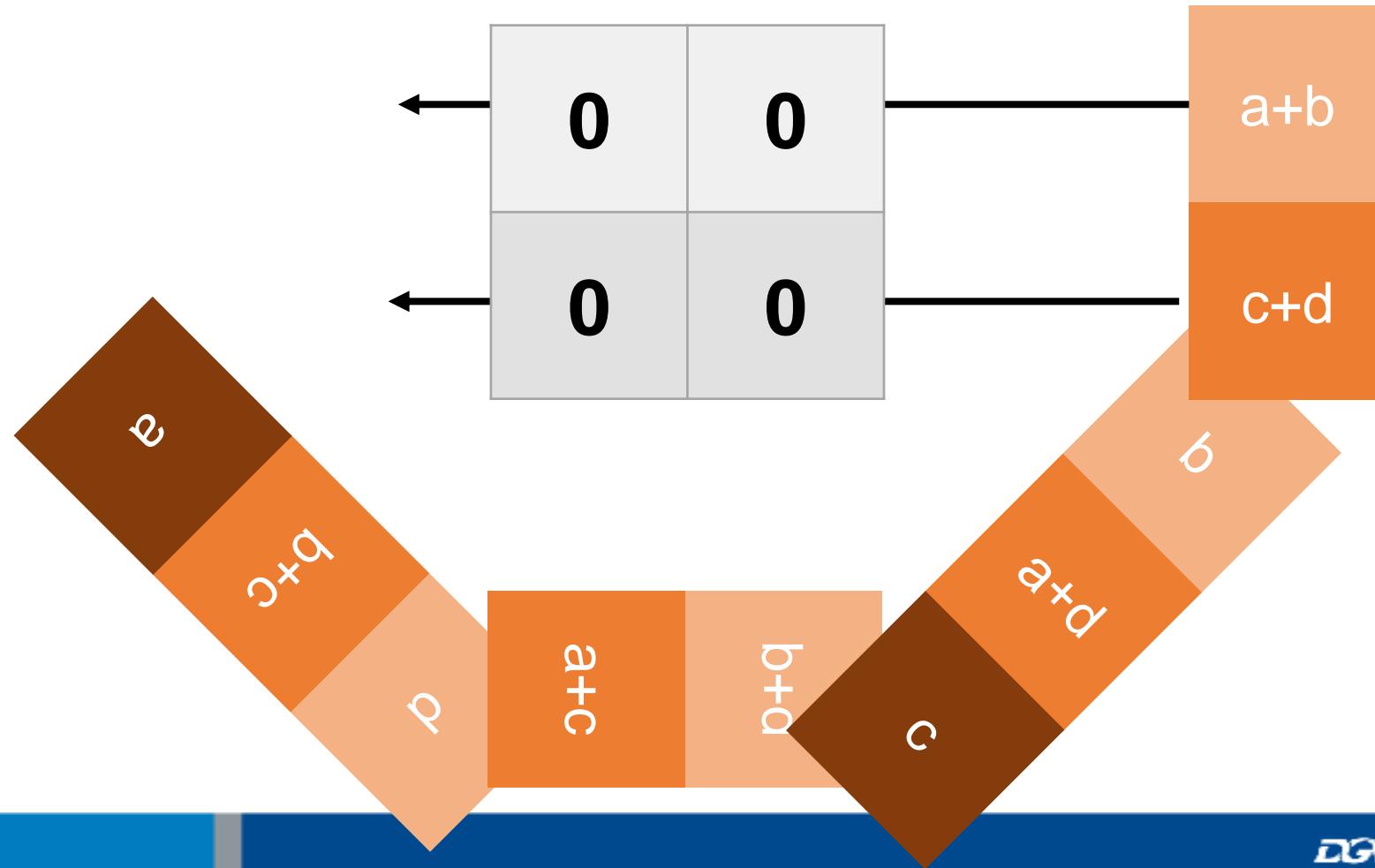
► CT data acquisition
X-ray source



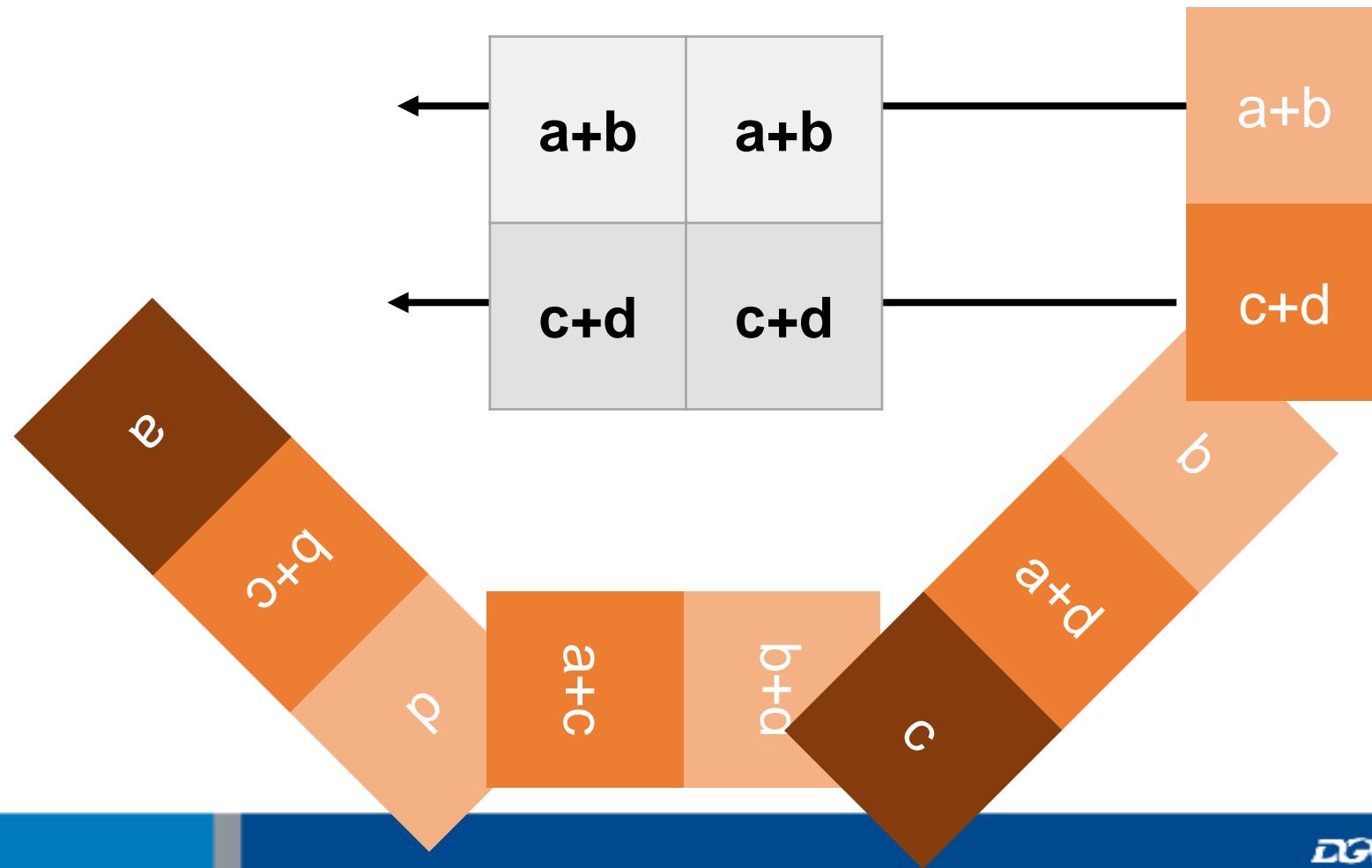
► CT data acquisition



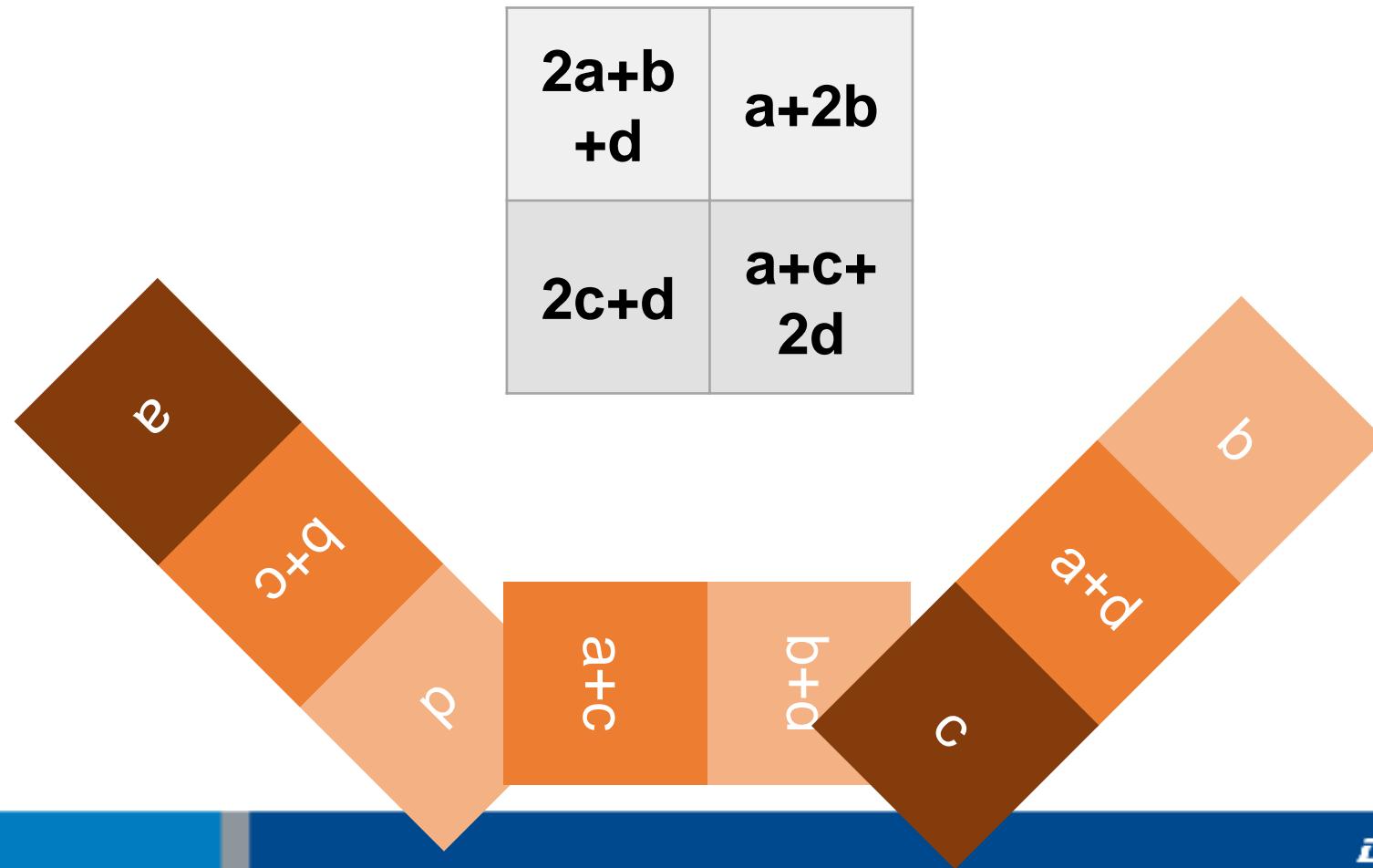
► Reconstruction – Backprojection (Concept, not real)



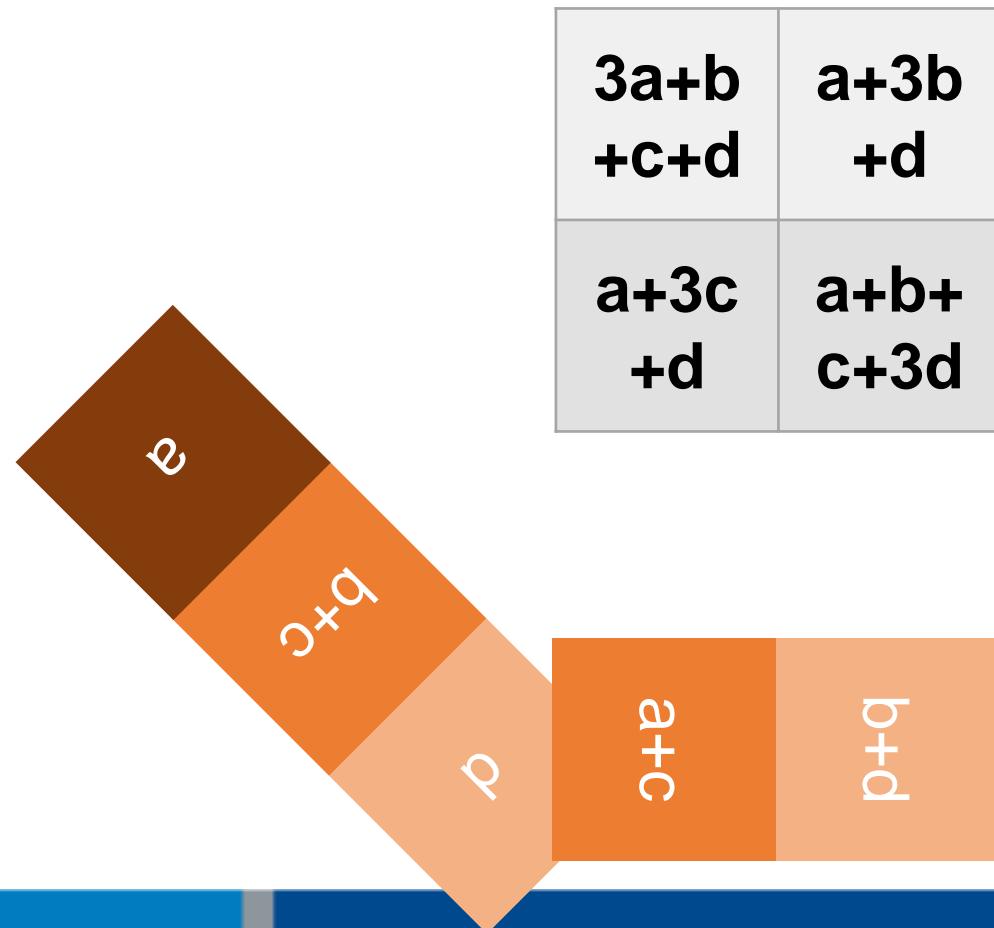
► Reconstruction – Backprojection (Concept, not real)



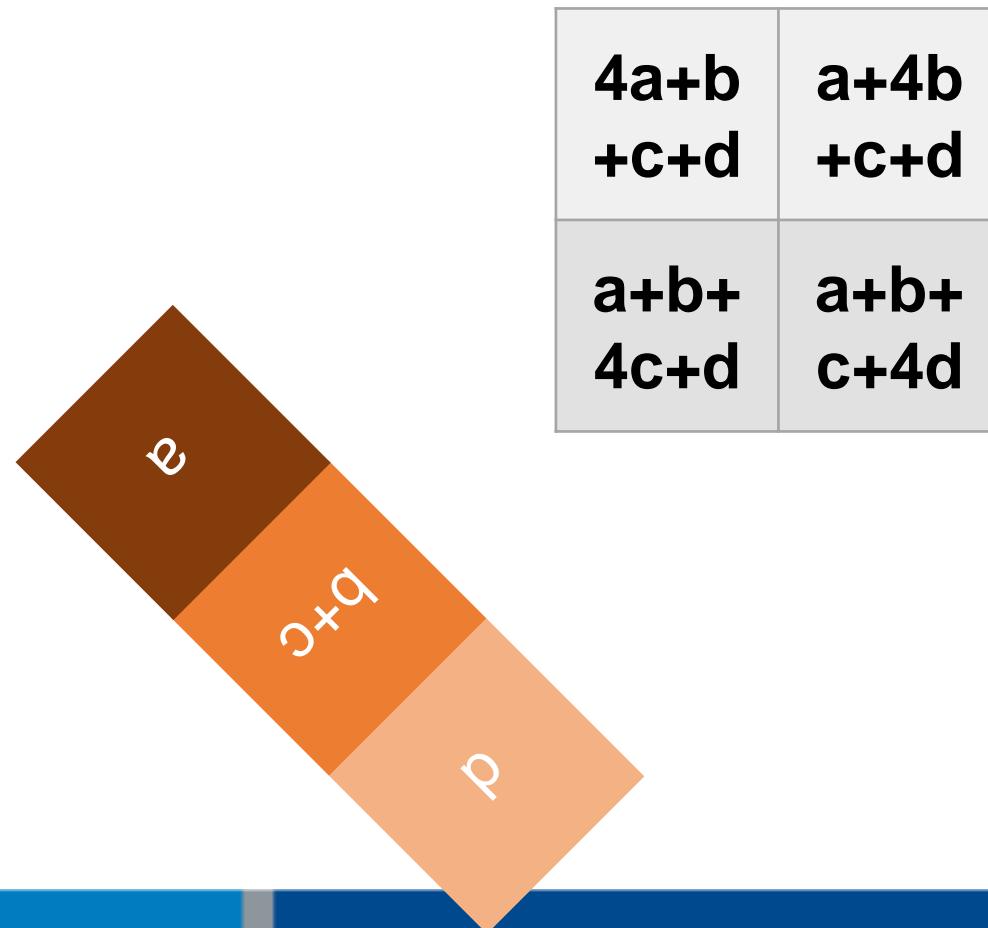
► Reconstruction – Backprojection (Concept, not real)



► Reconstruction – Backprojection (Concept, not real)

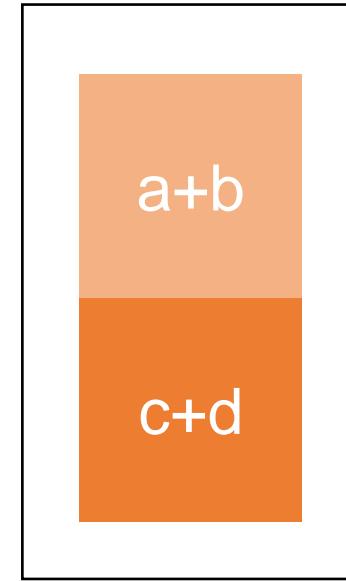


► Reconstruction – Backprojection (Concept, not real)



► Reconstruction – Backprojection (Concept, not real)

$4a+b$ $+c+d$	$a+4b$ $+c+d$
$a+b+$ $4c+d$	$a+b+$ $c+4d$



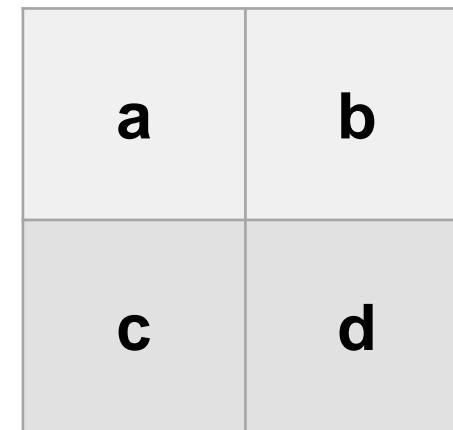
$$= a+b+c+d$$

► Reconstruction – Backprojection (Concept, not real)



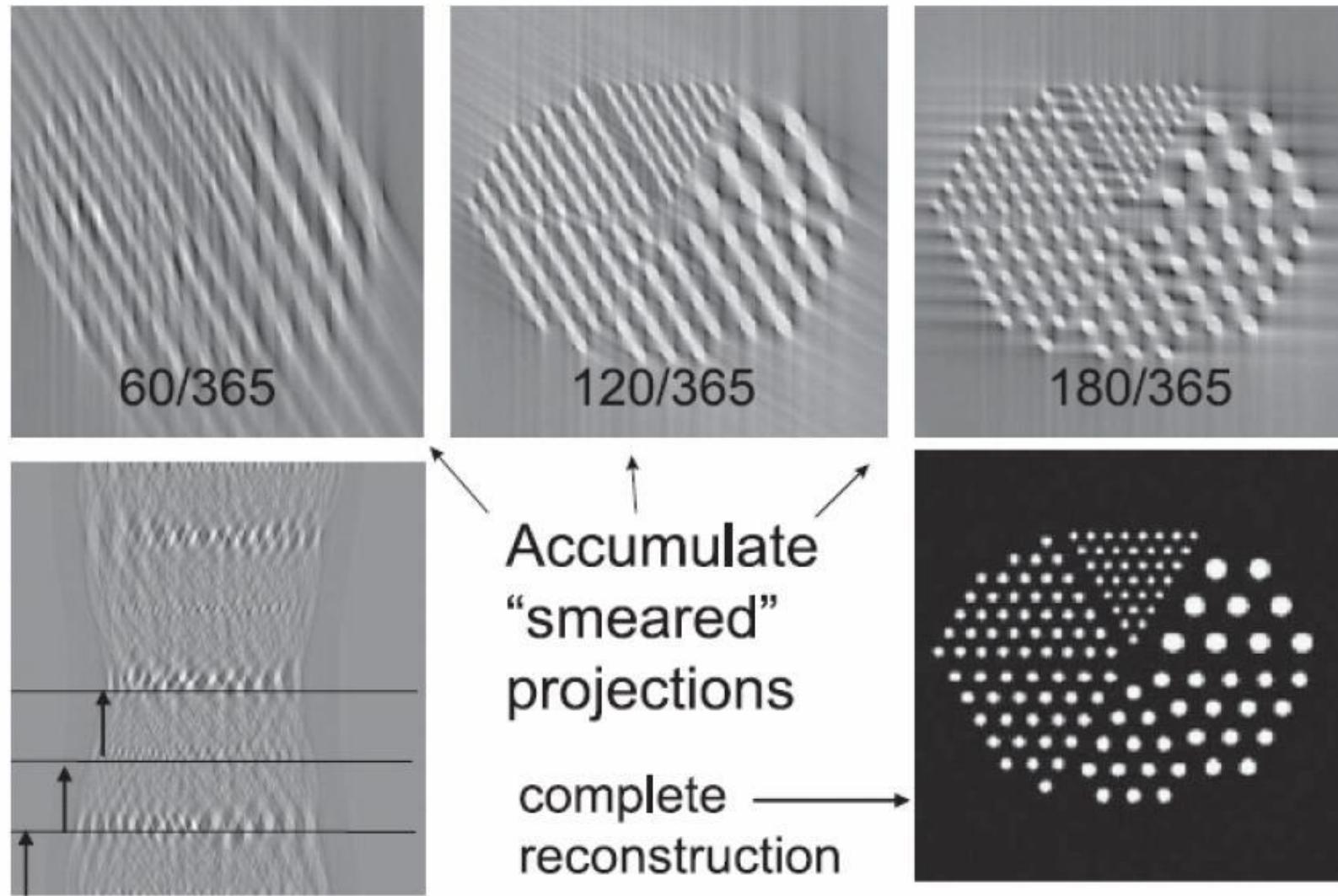
/ 3

► Reconstruction – Backprojection (Concept, not real)

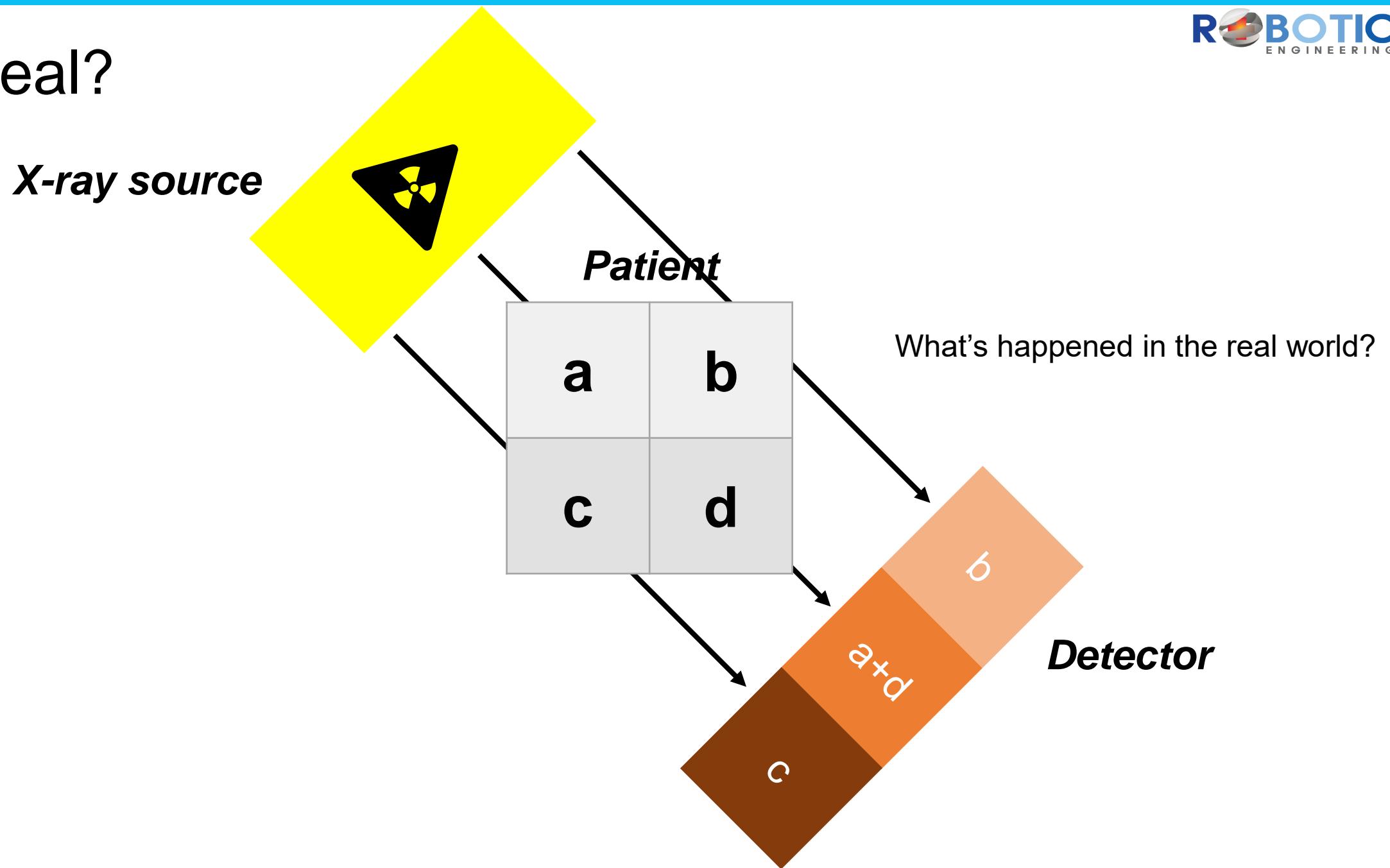


It's perfect in this case, however...

► Reconstruction – Backprojection (Concept, not real)



► Is it real?

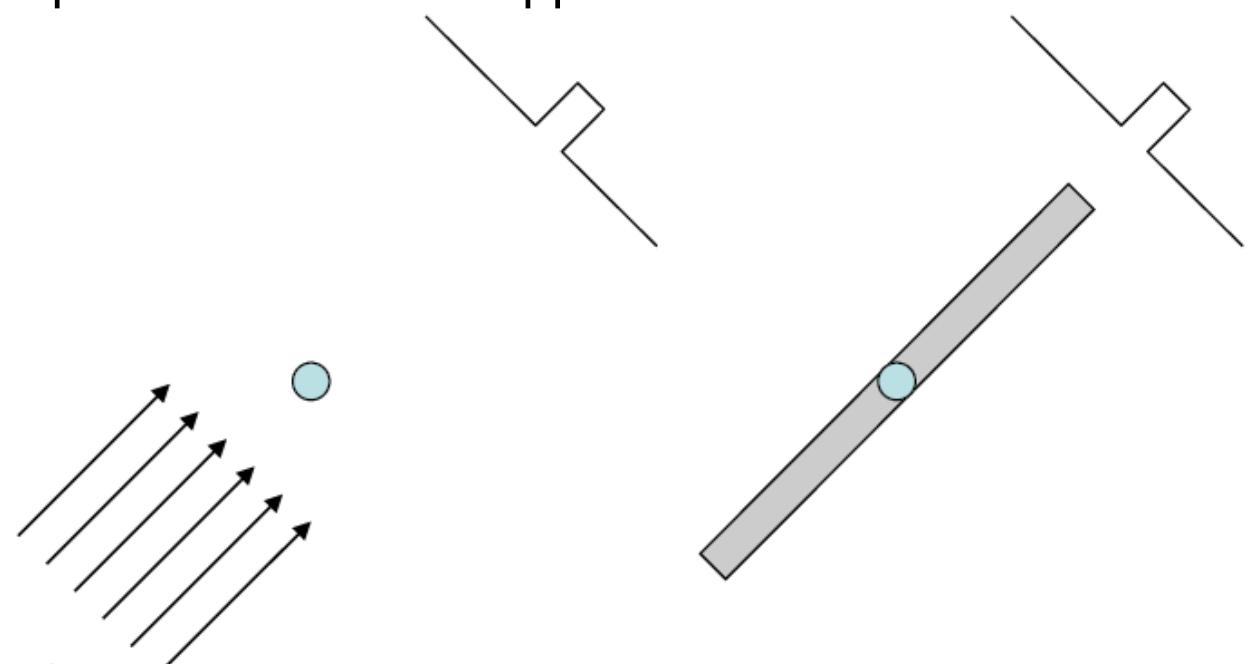


▶ Backprojection

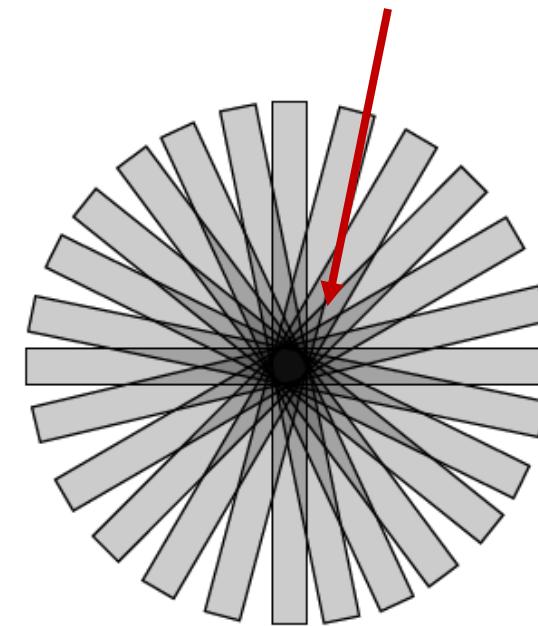
- Alternative direct reconstruction: backprojection
- The measured projections are “**smeared**” across the object: laminogram
- Crude reconstruction
- Nature of distortion: $1/r$

► Backprojection

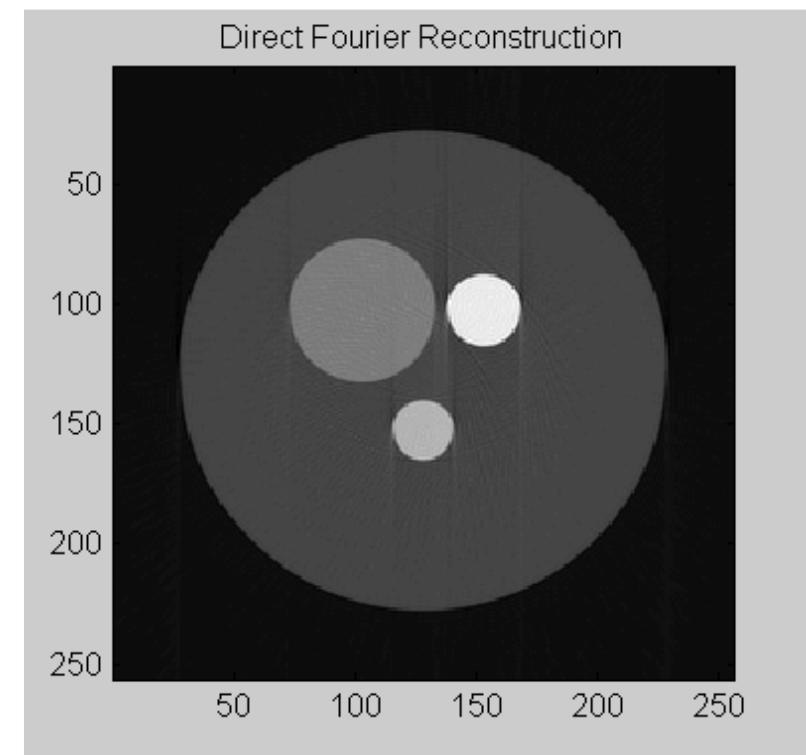
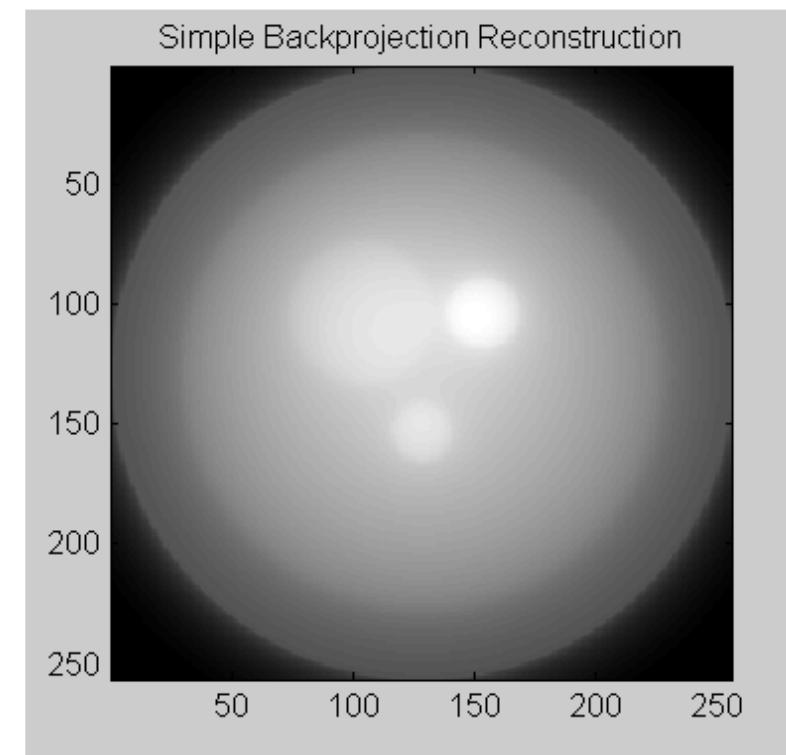
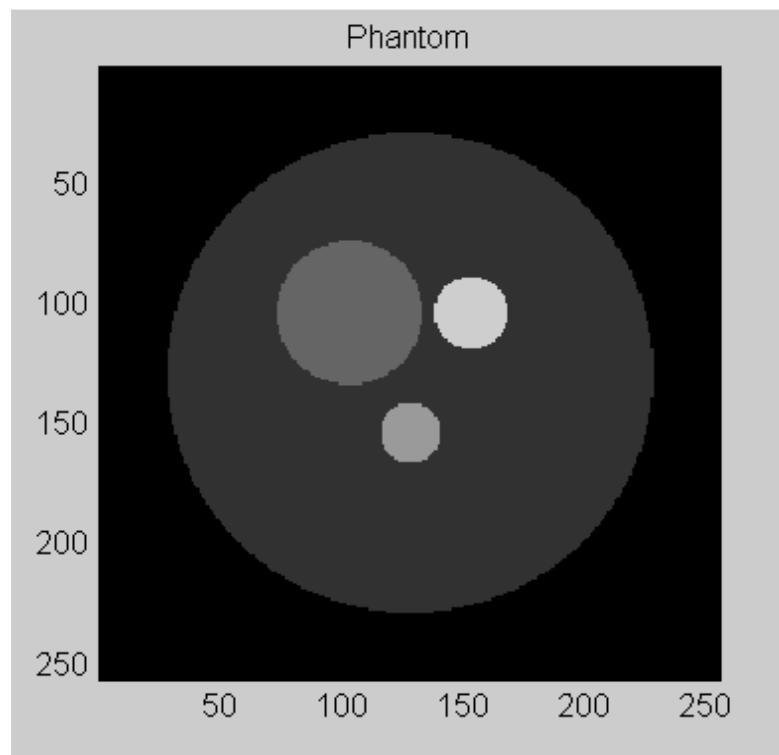
- Consider an example of a simple circular absorber
- The detected profile suggests the presence of the absorber along the x-ray source and detector
- Backproject the profile back to the object, i.e., assign the same value along that line
- Repeat for every projection
- There are problems with this approach



Smeared signal (decay $1/r$)

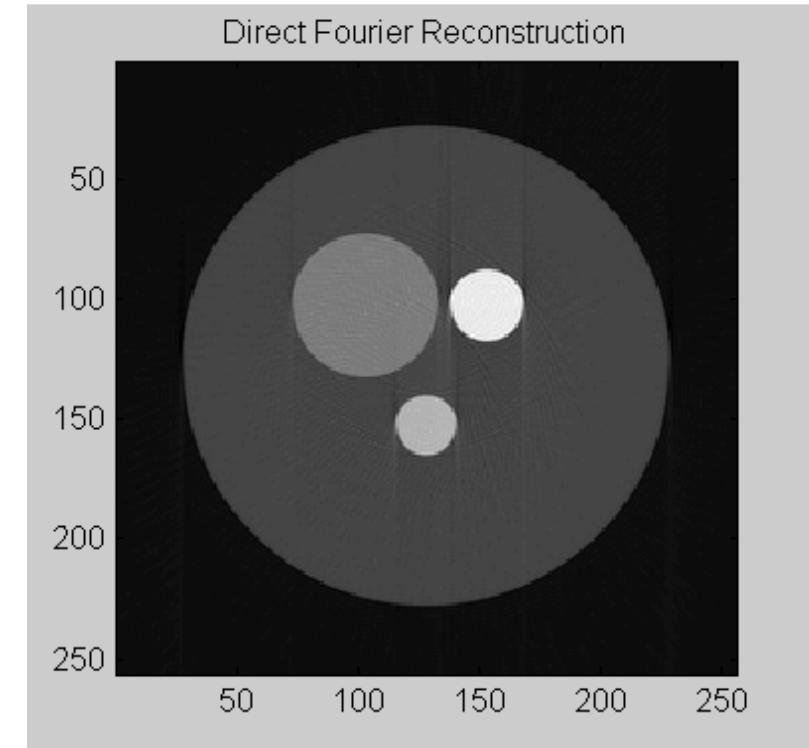
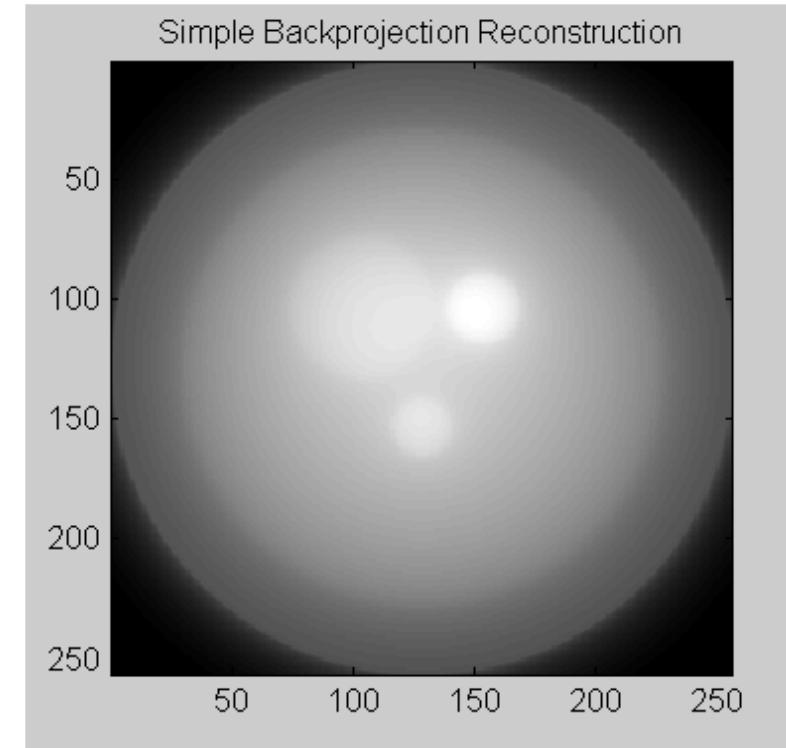
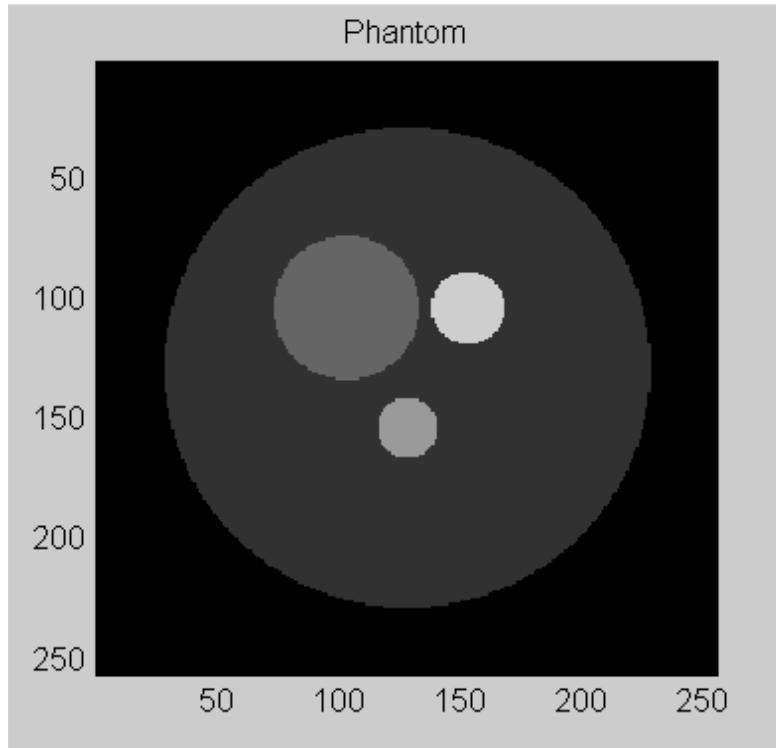


► Backprojection: Example



'Smeared'

► Backprojection vs Fourier Transform



Computationally simple but
inaccurate

Accurate but
computationally intensive

> *Filtered Backprojection: Computationally simple & Accurate*

▶ Filtered(Convolution) Backprojection

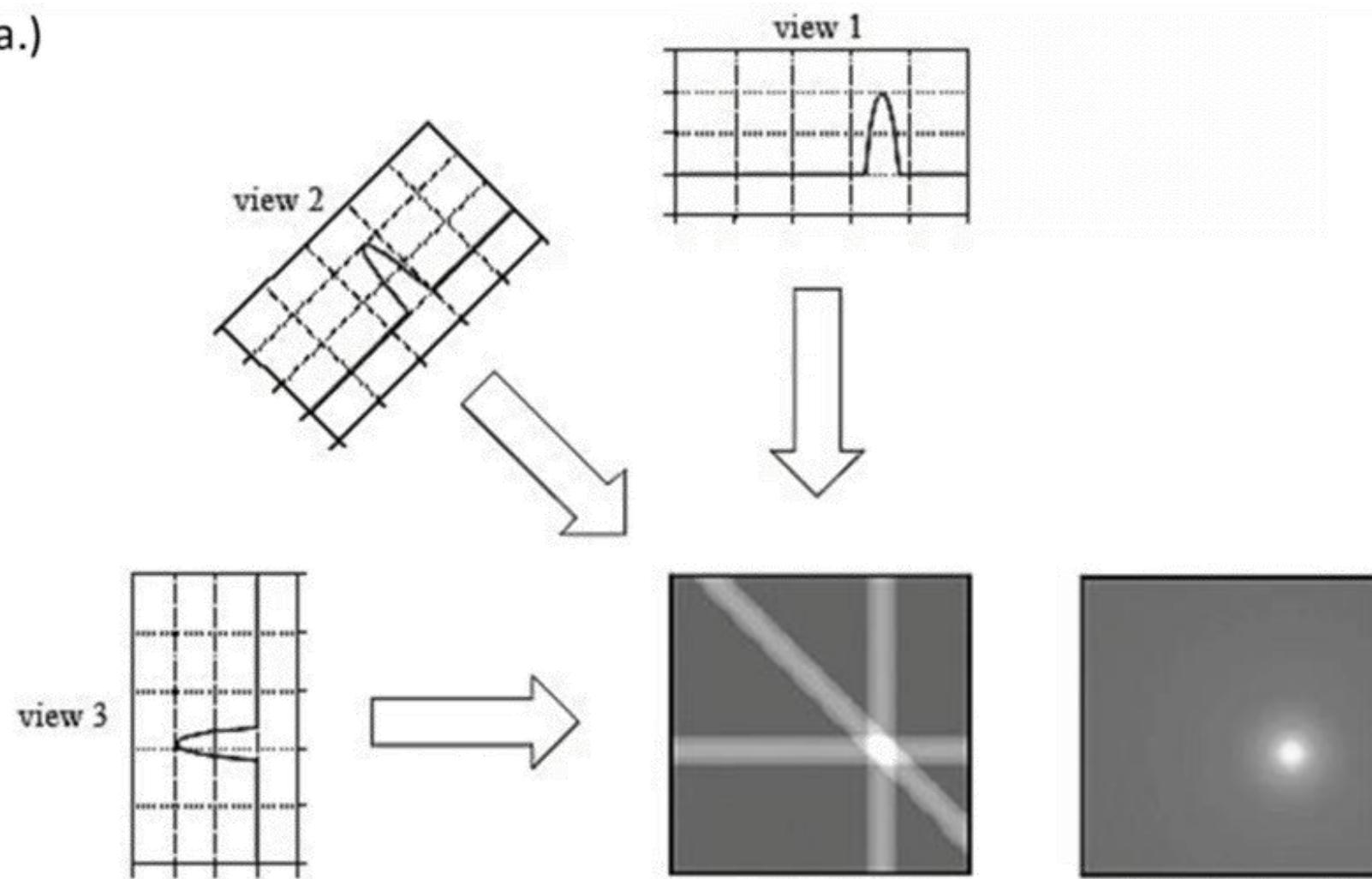
- Filtered Backprojection: corrects for “ $1/r$ blur”
- Convolution Backprojection: Filtered Back projection using convolution theorem of Fourier transform
- Consider $f(x,y)$ as derived previously:

$$\begin{aligned} f(x,y) &= \int_0^{2\pi} d\theta \int_0^{\infty} G(\rho, \theta) e^{i2\pi\rho(x\cos\theta + y\sin\theta)} \rho d\rho = \\ &= \int_0^{\pi} \int_{-\infty}^{\infty} |\rho| G(\rho, \theta) e^{i2\pi\rho(x\cos\theta + y\sin\theta)} d\rho d\theta = \\ &= \int_0^{\pi} \left[\int_{-\infty}^{\infty} |\rho| G(\rho, \theta) e^{i2\pi\rho l} d\rho \right] d\theta \end{aligned}$$

- Here $|\rho|$ is a frequency filter where FT of $g(l, \theta)$ is multiplied by $|\rho|$ and then IFT, and then backprojected

► Concept of Backprojection

a.)

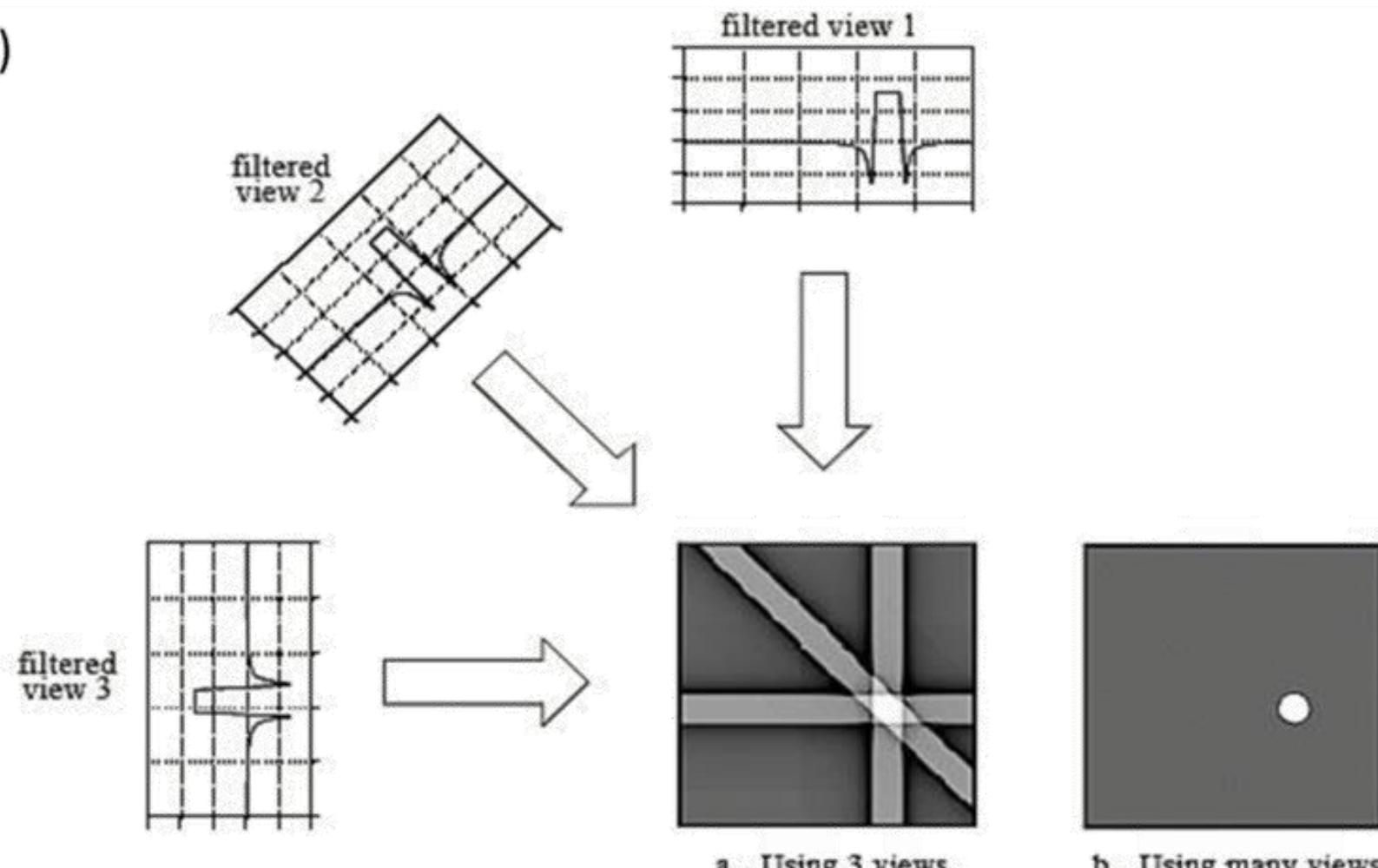


a. Using 3 views

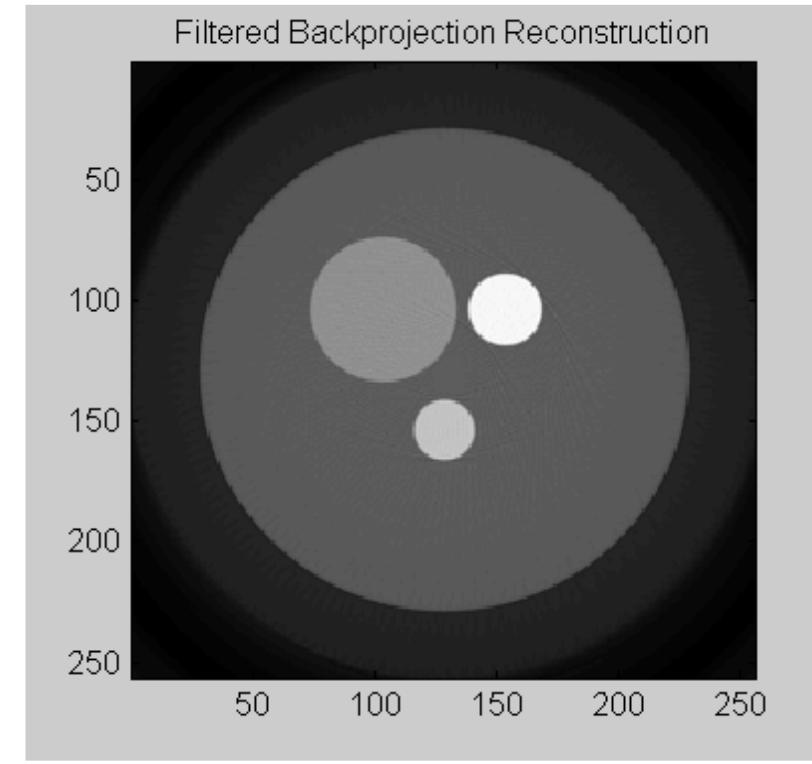
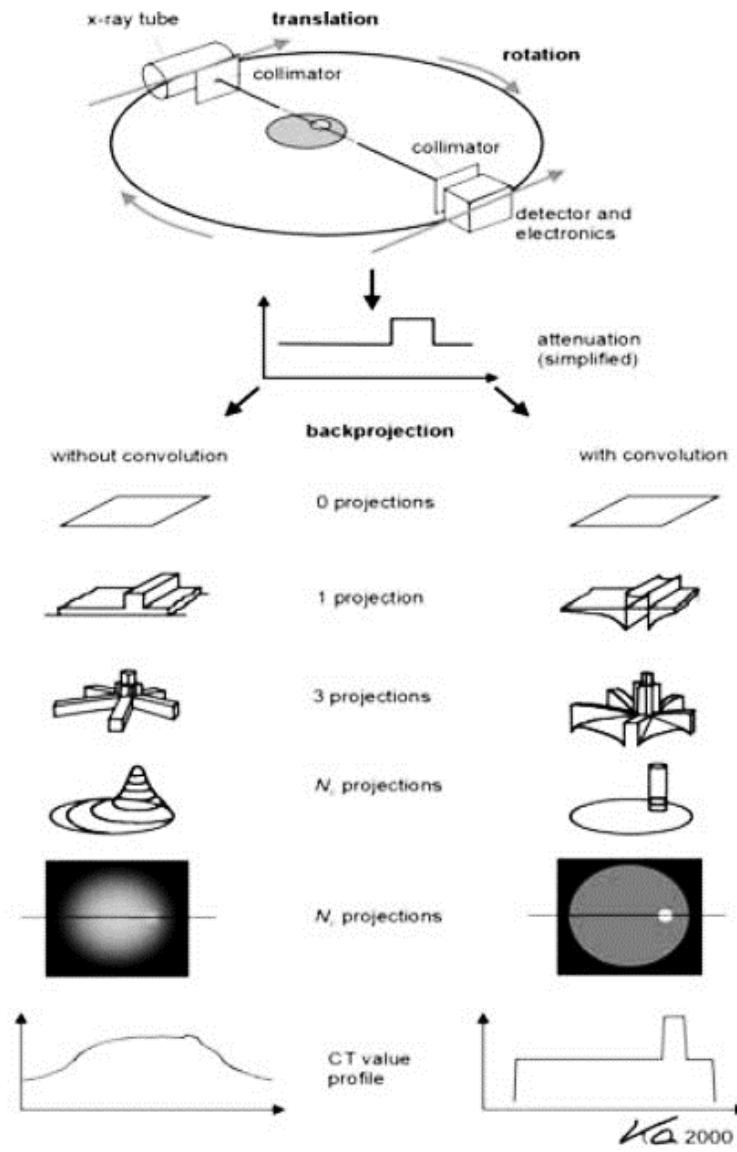
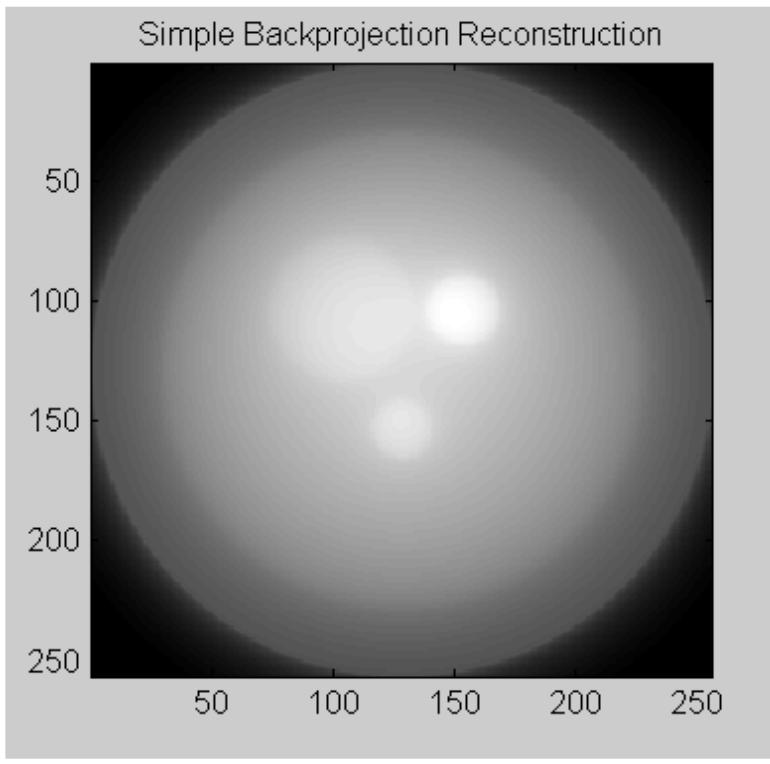
b. Using many views

► Concept of Filtered Backprojection

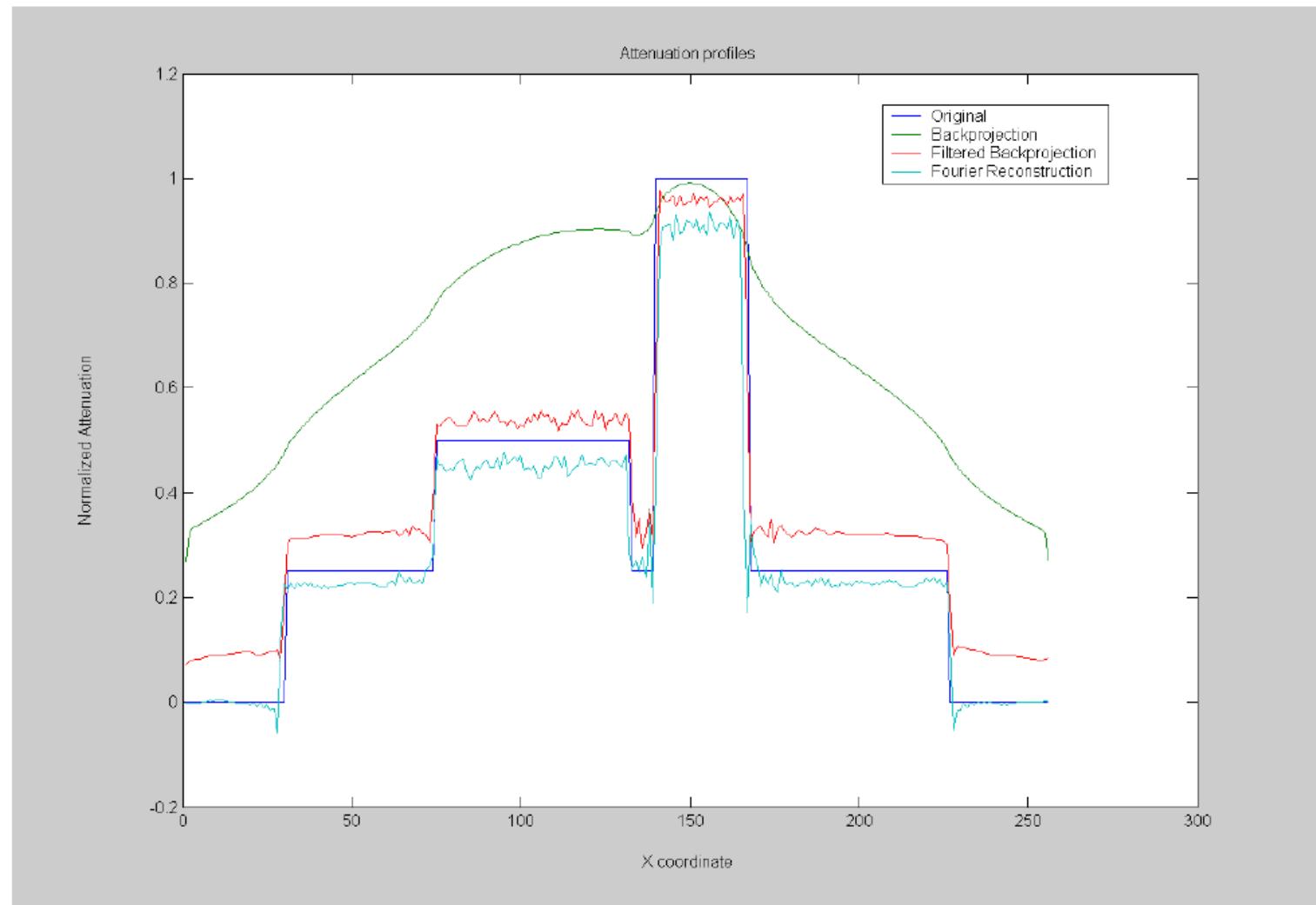
b.)



▶ Comparison



► Comparison



► Reminder: Fourier Transform Relations

- The convolution of two functions in space can be represented by simply multiplying their frequency spectra (**convolution**)

$$F[g(x, y) \otimes h(x, y)] = F \left[\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} g(\xi, \eta) h(x - \xi, y - \eta) d\xi d\eta \right] = G(u, v) H(u, v)$$

NOTE: This is very important relationship for imaging. A 2-D convolution is a common operation in imaging systems (for example, convolve the object with the point spread function of the imaging system). This same operation can be represented as the product of the spectra of the two functions [for example, product of spectra of PSF with object spectra results in image spectra].

▶ Convolution Backprojection

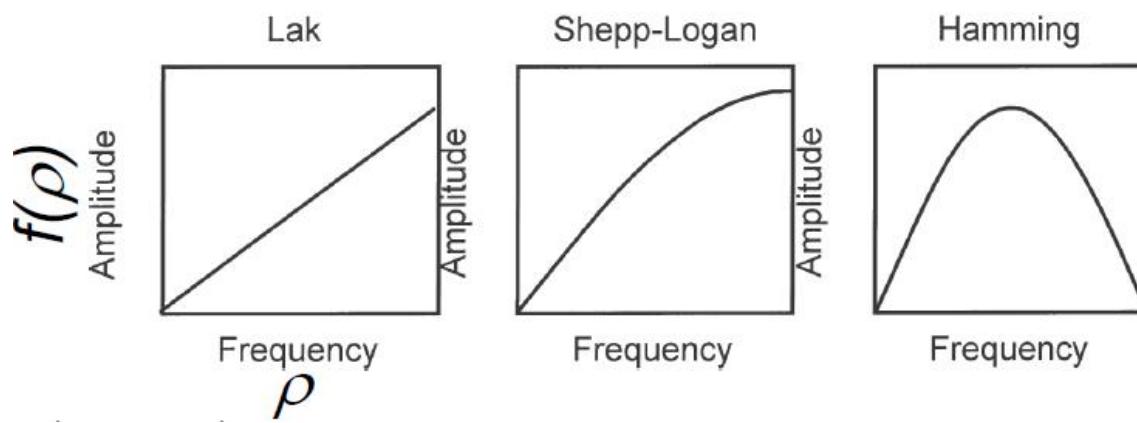
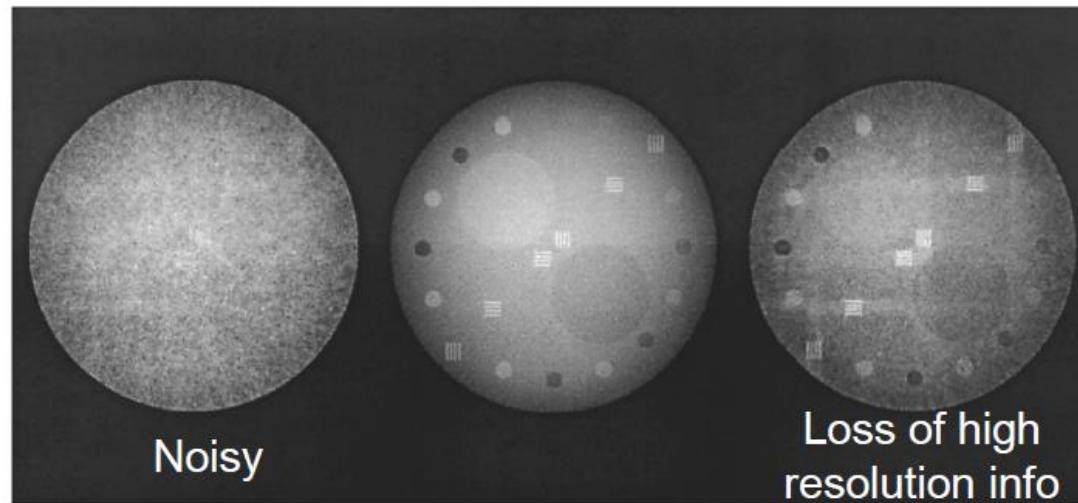
- Consider filtered backprojection equation and look at it from a convolution perspective

$$\begin{aligned}f(x, y) &= \int_0^{\pi} \left[\int_{-\infty}^{\infty} |\rho| G(\rho, \theta) e^{i 2 \pi \rho l} d\rho \right] d\theta = \int_0^{\pi} [F_{1D}^{-1}[|\rho|]^* g(l, \theta)] d\theta = \\&= \int_0^{\pi} [c(l)^* g(l, \theta)]_{l=x \cos \theta + y \sin \theta} d\theta = \\&= \int_0^{\pi} \int_{-\infty}^{\infty} g(l, \theta) c(x \cos \theta + y \sin \theta - l) dl d\theta\end{aligned}$$

- Here $c(\cdot)$ is IFT of $|\rho|$ and the equation above outlines the convolution backprojection

▶ Convolution Backprojection

- Unfortunately, IFT of $f(\rho) = |\rho|$ does not exist, and we often limit/modify $f(\rho)$ to have IFT possible.



Rams-Lak filter (Simple)

$$f(\rho) = \begin{cases} |\rho| & 0 \leq \rho \leq \rho_{\max} \\ 0 & \rho > \rho_{\max} \end{cases}$$

Shepp-Logan filter

Hamming

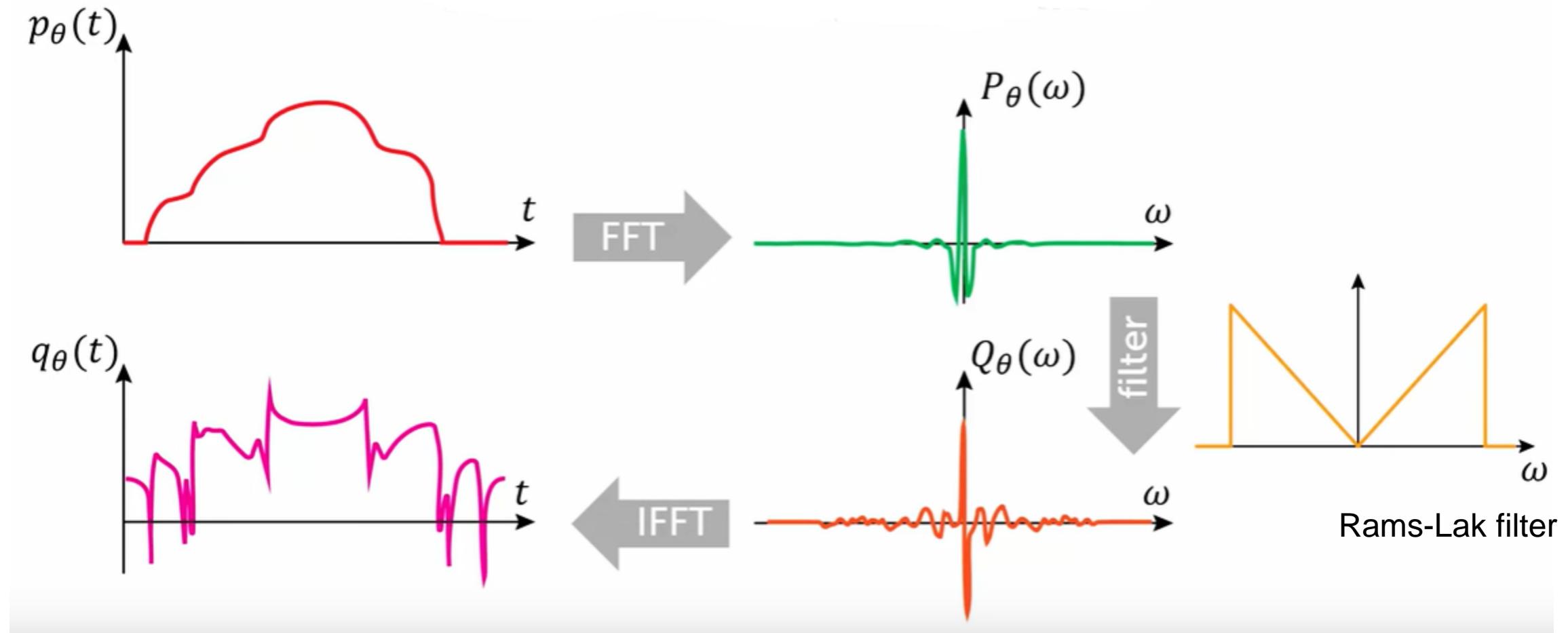
► What does $f(\rho)$ mean? -> Filter

- Consider filtered backprojection equation :

$$\begin{aligned} f(x, y) &= \int_0^{\pi} \left[\int_{-\infty}^{\infty} |\rho| G(\rho, \theta) \cdot e^{i2\pi\rho l} d\rho \right] d\theta = \int_0^{\pi} \left[\int_{-\infty}^{\infty} (f(\rho) \cdot G(\rho, \theta)) e^{i2\pi\rho l} d\rho \right] d\theta = \\ &= \int_0^{\pi} \left[\int_{-\infty}^{\infty} G^{\text{filtered}}(\rho, \theta) \cdot e^{i2\pi\rho l} d\rho \right] d\theta = \int_0^{\pi} g^{\text{filtered}}(l, \theta) d\theta \end{aligned}$$

- In Fourier domain, $G(\rho, \theta)$ is multiplied by $f(\rho)$, i.e., the extent (and content) of $G(\rho, \theta)$ is affected by the functional form of $f(\rho)$. In image domain, i.e., if we consider $g(l, \theta)$, such procedure represents filtering.
- Same applies to convolution-based back projection

▶ Process



► Same image, different display



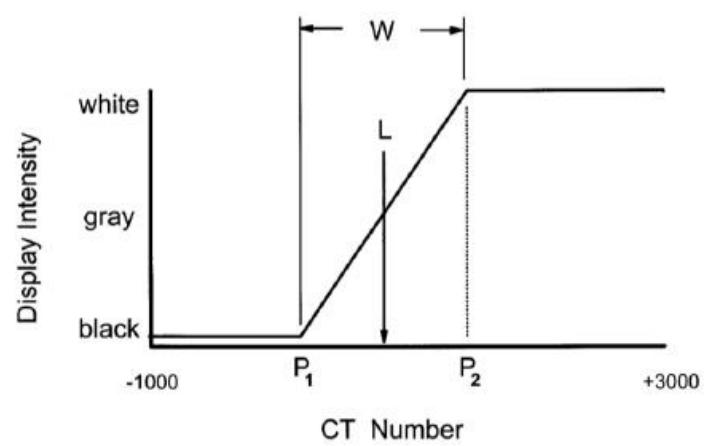
$W = 4095, L = 1048$



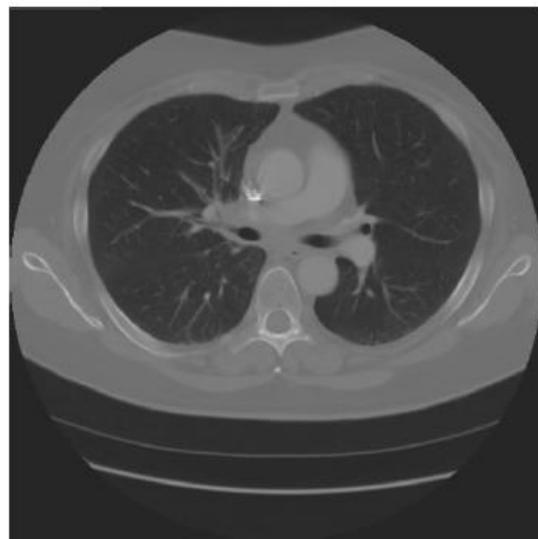
$W = 600, L = -100$



$W = 700, L = -650$



► Same image, different display



$W = 4095, L = 1048$

May be too large DR,
good for bone



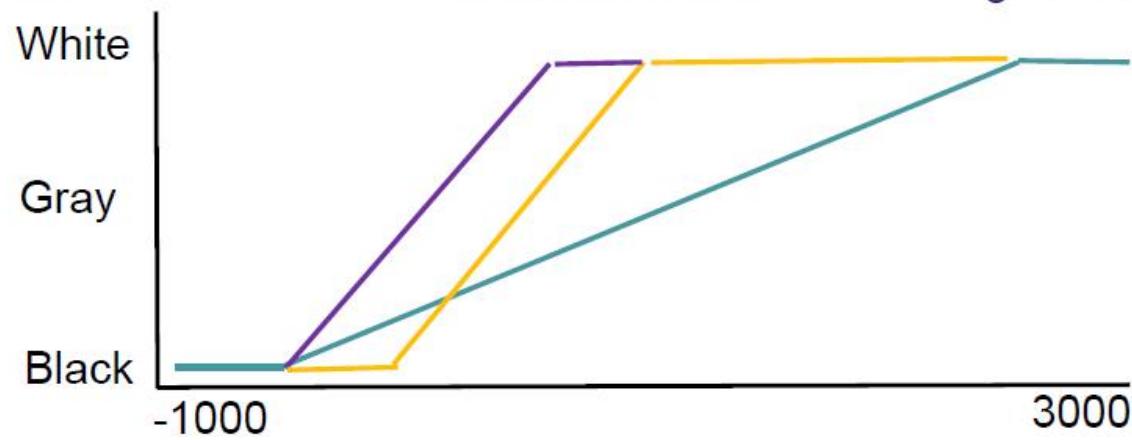
$W = 600, L = -100$

Optimal Dynamic Range
For mediastinum



$W = 700, L = -650$

Saturated for bone,
good for lung structure



► CT number and Display

- 12 bits: 2^{12} combinations = 4096 values
- Contrast scale: Hounsfield unit (HU) scale

$$HU = 1000 \frac{\mu_m - \mu_{water}}{\mu_{water}}$$

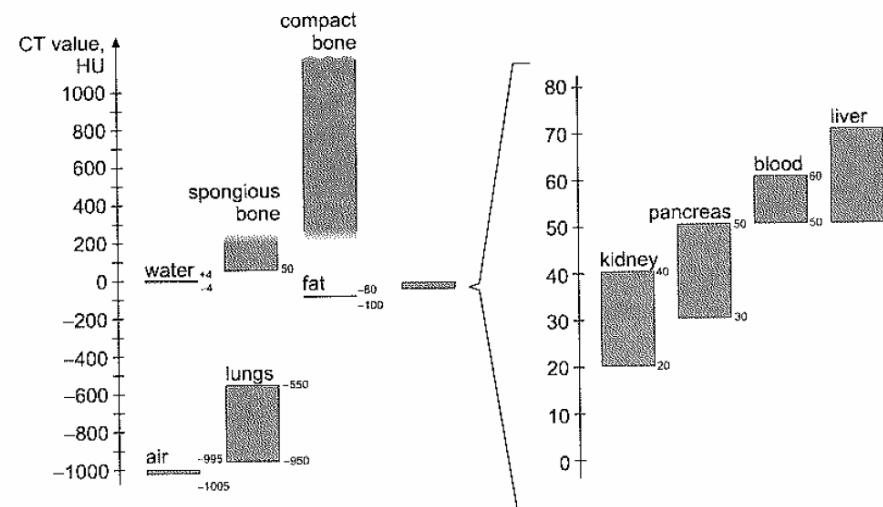
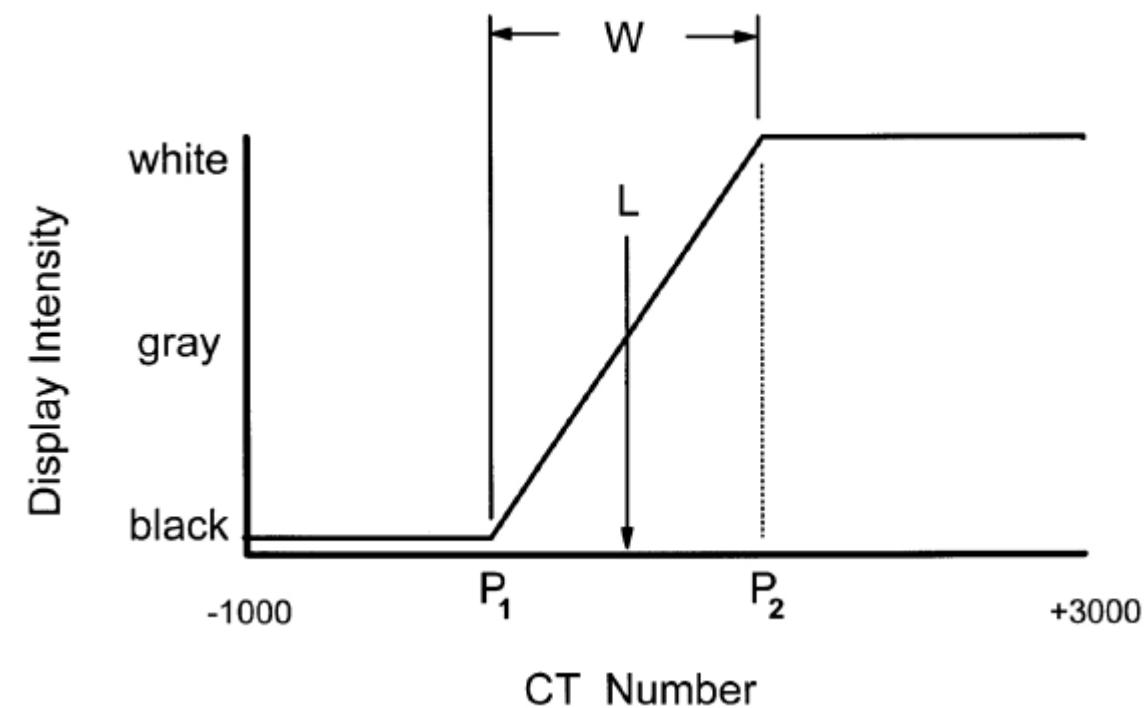


Figure 1.9
The Hounsfield scale. CT values characterize the linear attenuation coefficient of the tissue in each volume element relative to the μ -value of water. The CT values of different tissues are therefore defined to be relatively stable and to a high degree independent of the x-ray spectrum.

$$\Delta N = -\mu \cdot N \cdot \Delta z$$

$$N = N_0 e^{-\mu \cdot z}$$



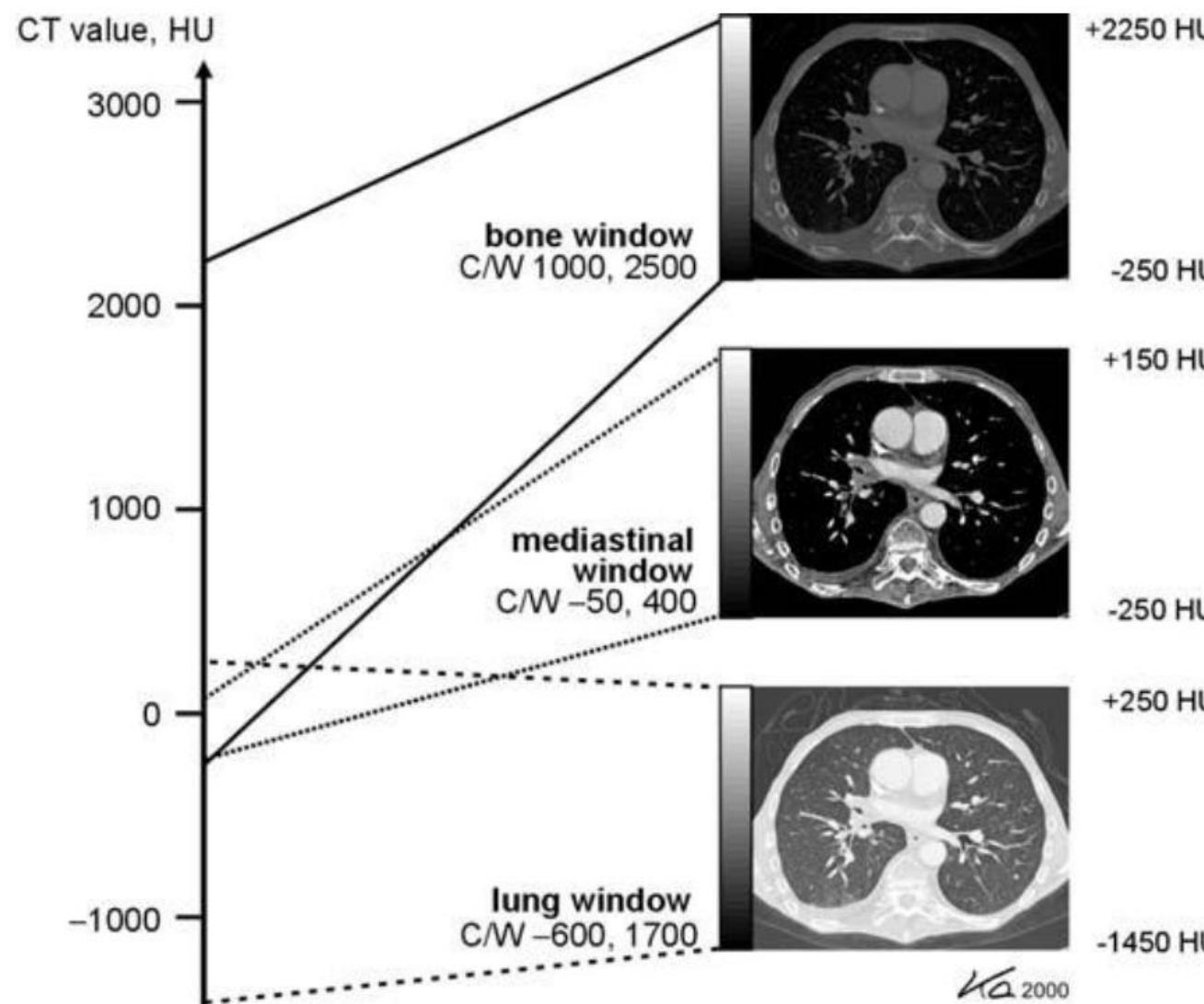
► CT number and Display

Substance		HU
Air		-1000
Fat		-120 to -90 ^[2]
Soft tissue on contrast CT		+100 to +300
Bone	Cancellous	+300 to +400 ^[3]
	Cortical	+1800 to +1900 ^[3]
Subdural hematoma	First hours	+75 to +100 ^[4]
	After 3 days	+65 to +85 ^[4]
	After 10-14 days	+35 to +40 ^[5]
Other blood	Unclootted	+13 ^[6] to +50 ^[7]
	Clotted	+50 ^[8] to +75 ^{[6][8]}
Pleural effusion	Transudate	+2 to +15 ^[9]
	Exudate	+4 to +33 ^[9]
Other fluids	Chyle	-30 ^[10]
	Water	0
	Urine	-5 to +15 ^[2]
	Bile	-5 to +15 ^[2]
	CSF	+15
	Abscess / Pus	0 ^[11] or +20 ^[12] , to +40 ^[12] or +45 ^[11]
	Mucus	0 ^[13] - 130 ^[14] ("high attenuating" at over 70 HU) ^{[15][16]}

Parenchyma	Lung	-700 to -600 ^[17]
	Kidney	+20 to +45 ^[2]
	Liver	60 ± 6 ^[18]
	Lymph nodes	+10 to +20 ^[19]
	Muscle	+35 to +55 ^[2]
Thymus	<ul style="list-style-type: none"> +20 to +40 in children^[20] +20 to +120 in adolescents^[20] 	
	White matter	+20 to +30
	Grey matter	+37 to +45
Gallstone	Cholesterol stone	+30 to +100 ^[21]
	Bilirubin stone	+90 to +120 ^[21]
Foreign body ^[22]	Windowpane glass	+500
	Aluminum, tarmac, car window glass, bottle glass, and other rocks	+2,100 to +2,300
	Limestone	+2,800
	Copper	+14,000
	Silver	+17,000
	Steel	+20,000
	Gold, steel, and brass	+30,000 (upper measurable limit)
Earwax	<0	

From, wikipedia

► CT number and Display



Ref. kalendar

► Parameters for Spatial / Contrast Resolution

- Detector pitch
- Detector aperture
- Number of views
- Number of rays
- Focal spot size
- Object magnification
- Slice thickness
- Slice sensitivity profile
- Helical pitch
- Reconstruction kernel
- Pixel matrix
- Patient motion
- Field of view
- Number of x-ray photons
- Dose
- Pixel size
- Slice thickness
- Reconstruction filter
- Patient size
- Rotation speed

► Artifacts

- Aliasing
- Patient motion
- Beam hardening
- Partial volume averaging

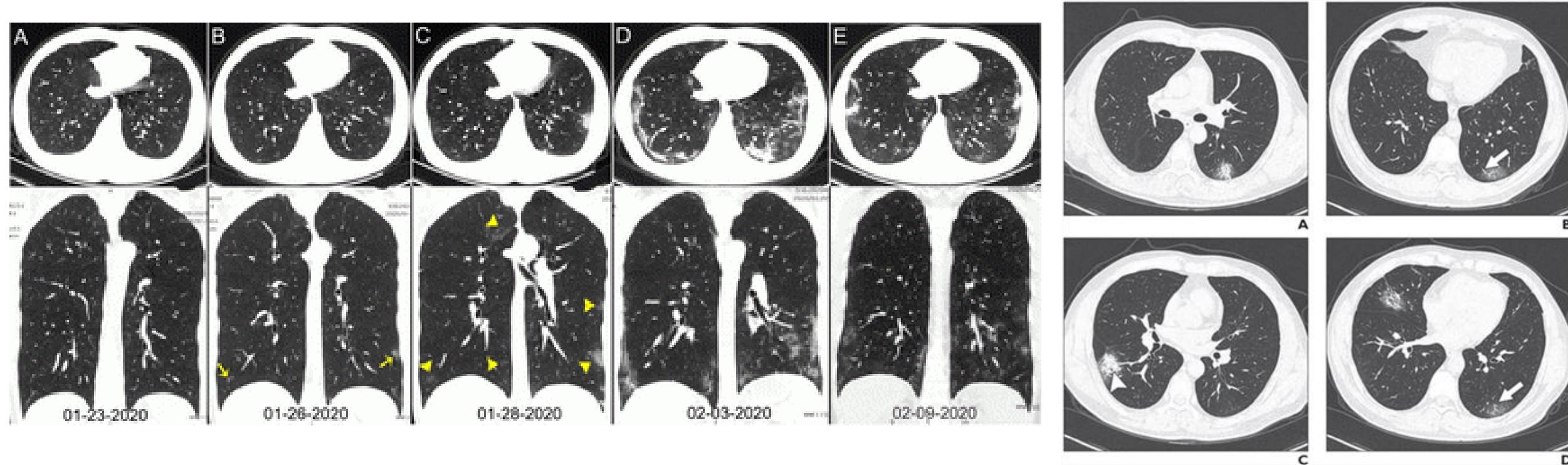
► Applications: Anatomical, Functional, Interventional

- Cardiology
- CT angiography
- Pre-, intra-, and post-operative imaging
- Image guided neurosurgery
- Orthopedic
- Trauma evaluation
- Lung CT imaging
- Pulmonary CT Imaging
- Neuro CT

► Applications: Anatomical, Functional, Interventional

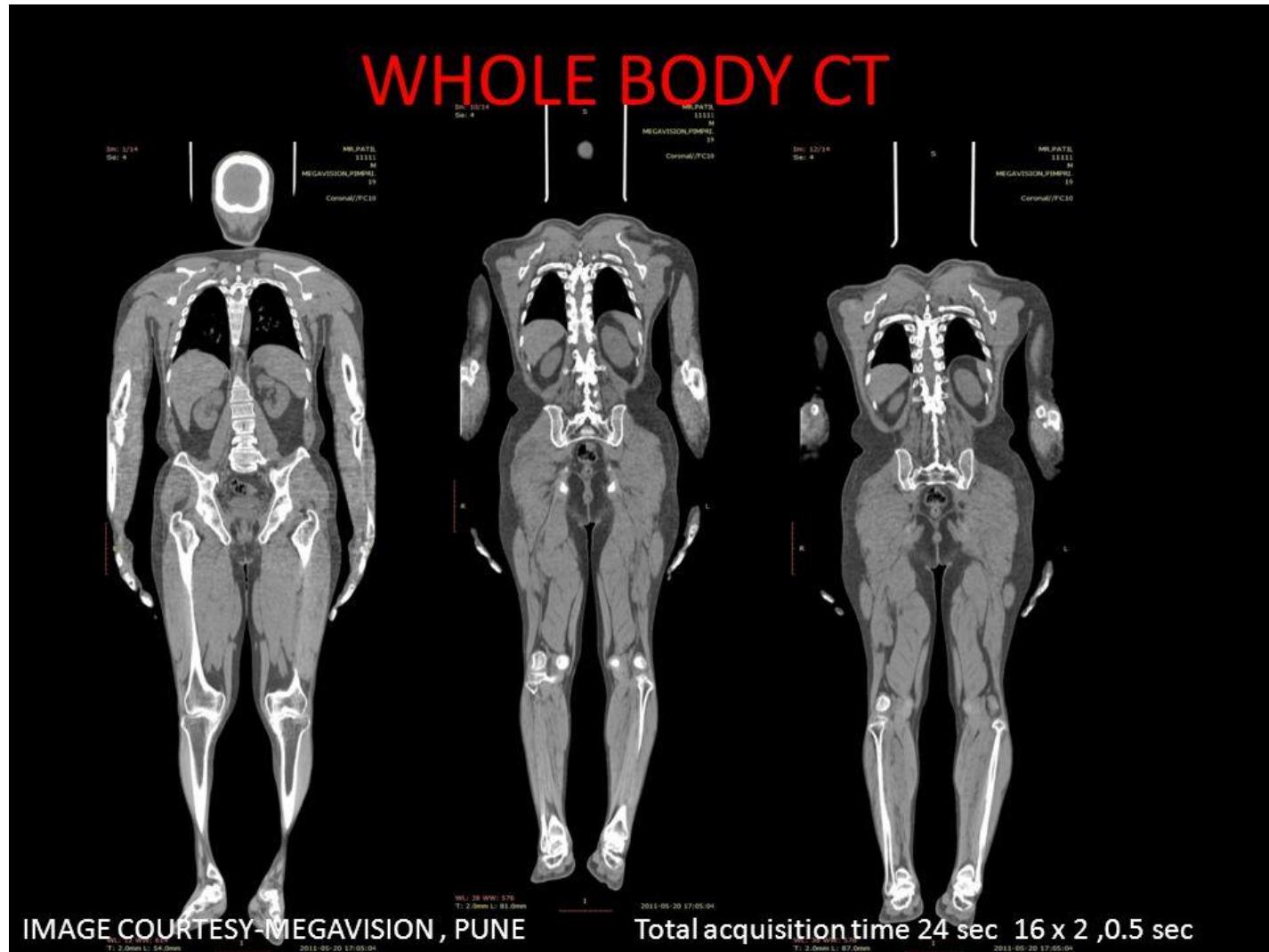
CT Provides Best Diagnosis for Novel Coronavirus (COVID-19)

► Applications: Anatomical, Functional, Interventional



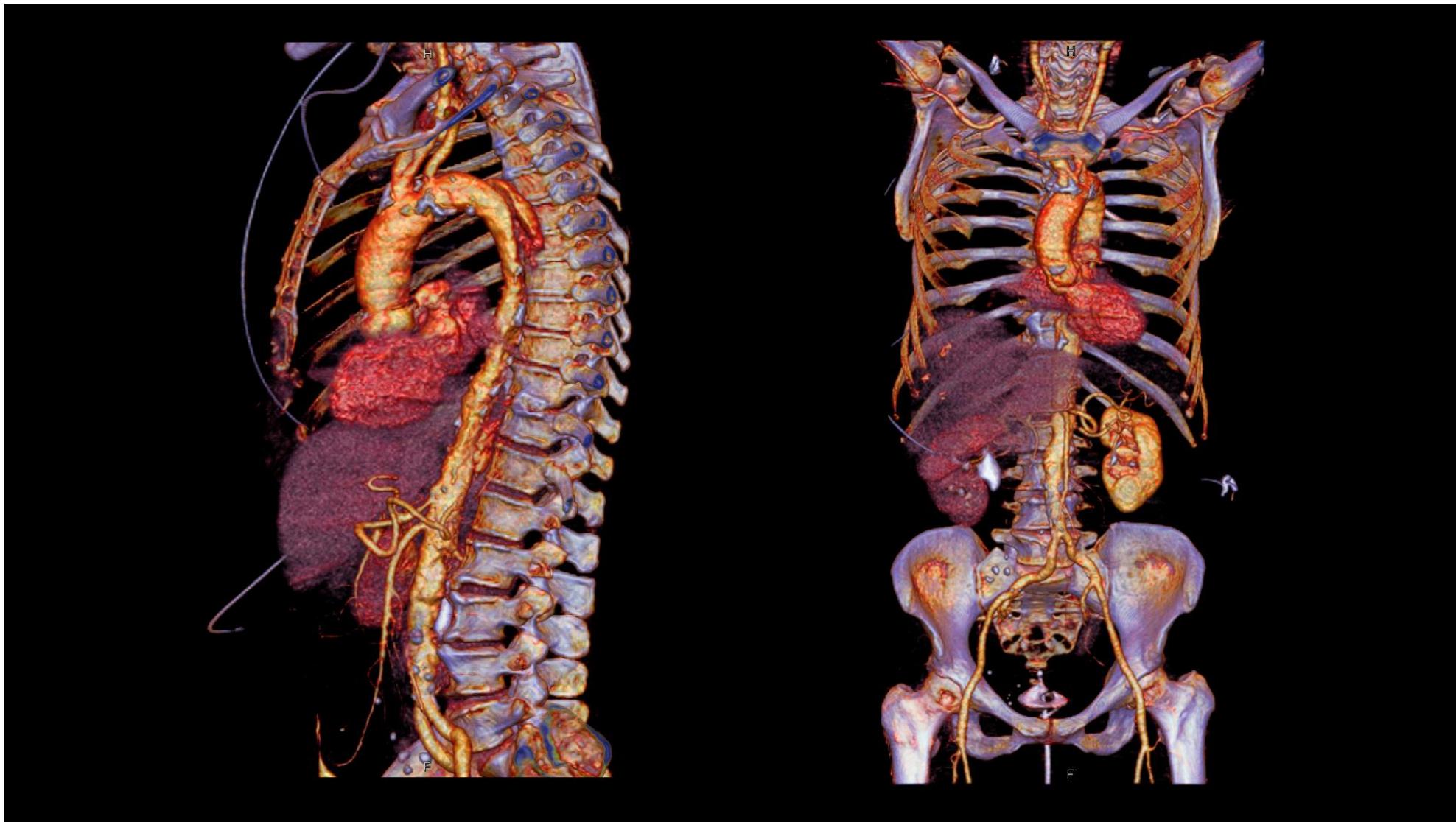
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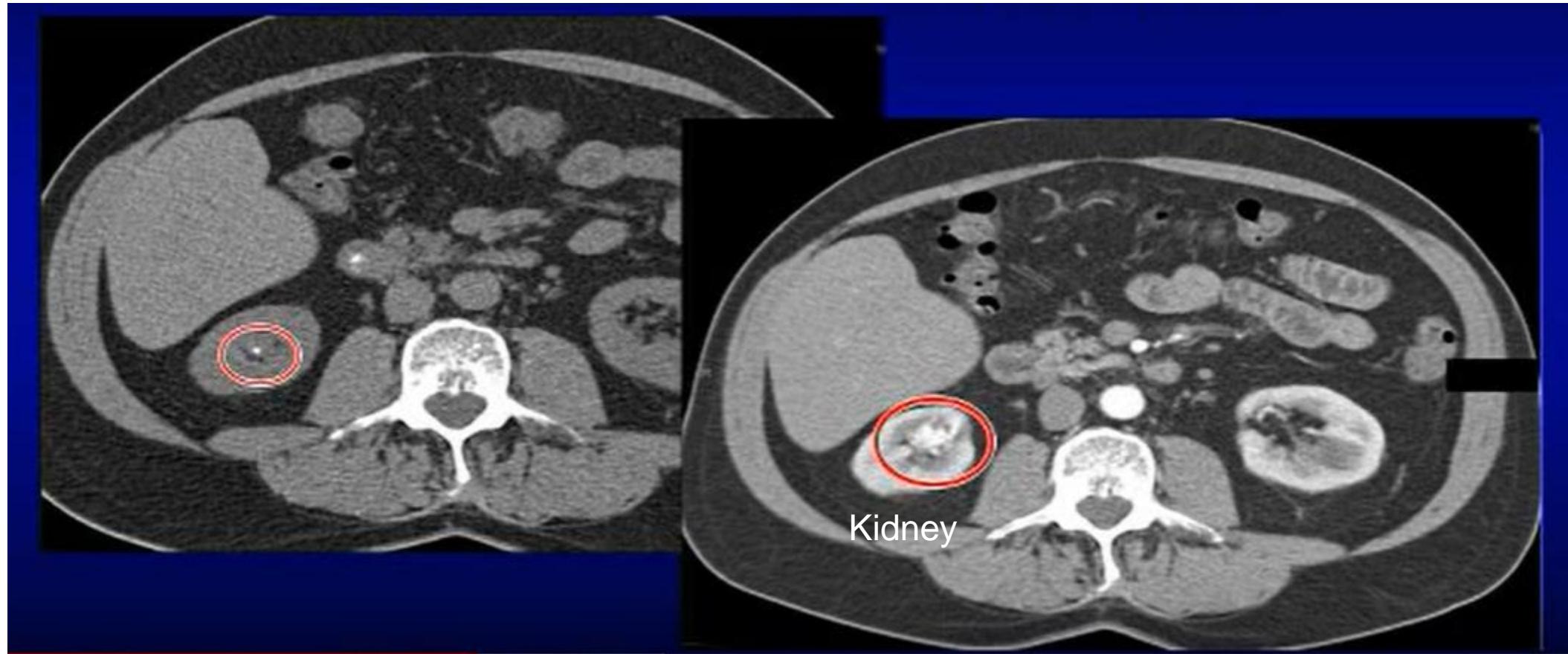


Whole body CT scan

► CT Angiography



► Limitations of Non-contrast CT



► Modern CT scanner, GE healthcare, 2019

S/W revolution in 21c

- Core technology is same
- A.I. powered automation
- A.I. powered diagnosis

► Past and Future of Computed Tomography

- Previous efforts: faster imaging
 - Cardiac imaging
- Current development
 - 64-slice scanners
 - 128 images acquired per second !!!
 - Pulmonary embolism (PE) exam – 5 sec to cover 40 cm, 640 images
 - Spatial resolution
 - Functional Imaging
 - Contrast agent development
 - Combined/Multi-modality

