

Computed Tomography

From X-ray Physics to Applications

Wim van Aarle

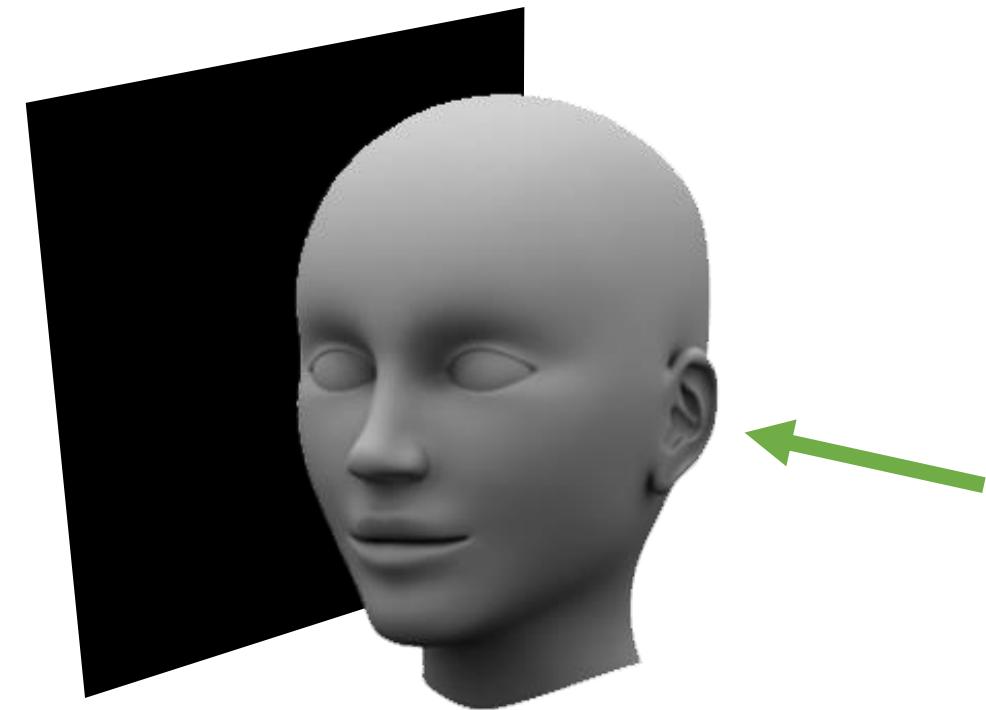
Summer School Szeged 12/07/2016

- **Computed Tomography**
- Some Applications
- A Brief Introduction to X-ray Physics
- A Brief Introduction to Reconstruction Mathematics
- When Physics and Mathematics colide

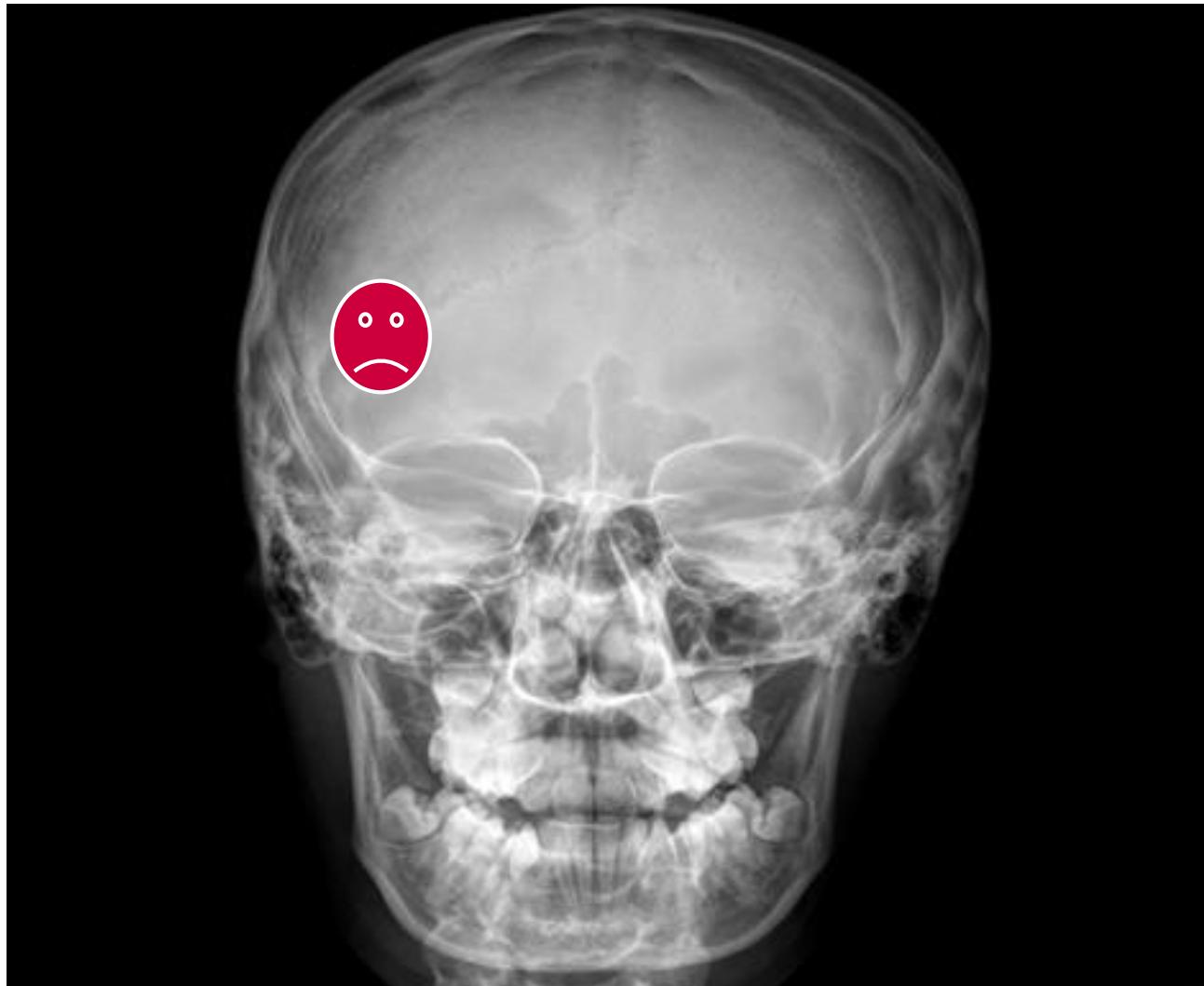
What is Computed Tomography



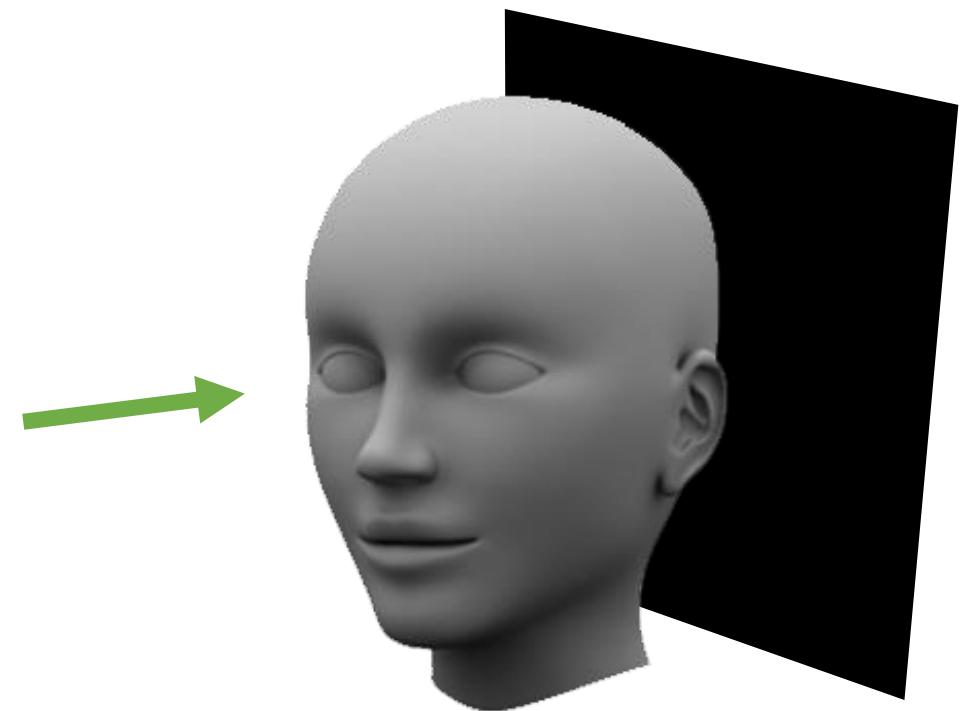
X-ray imaging



What is Computed Tomography



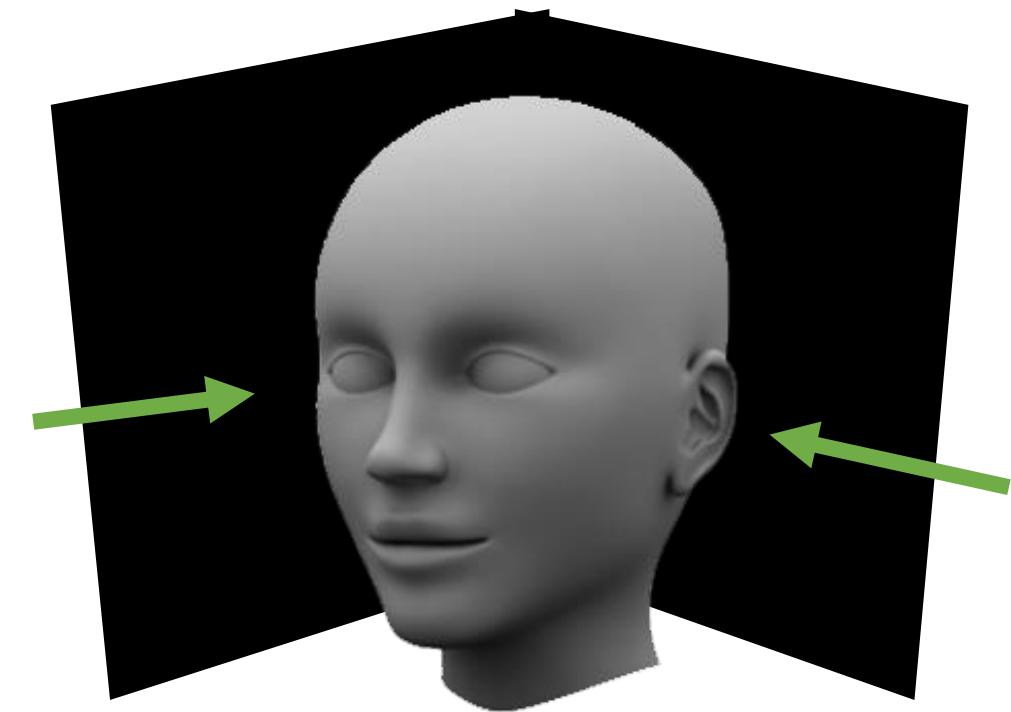
X-ray imaging



What is Computed Tomography



X-ray imaging



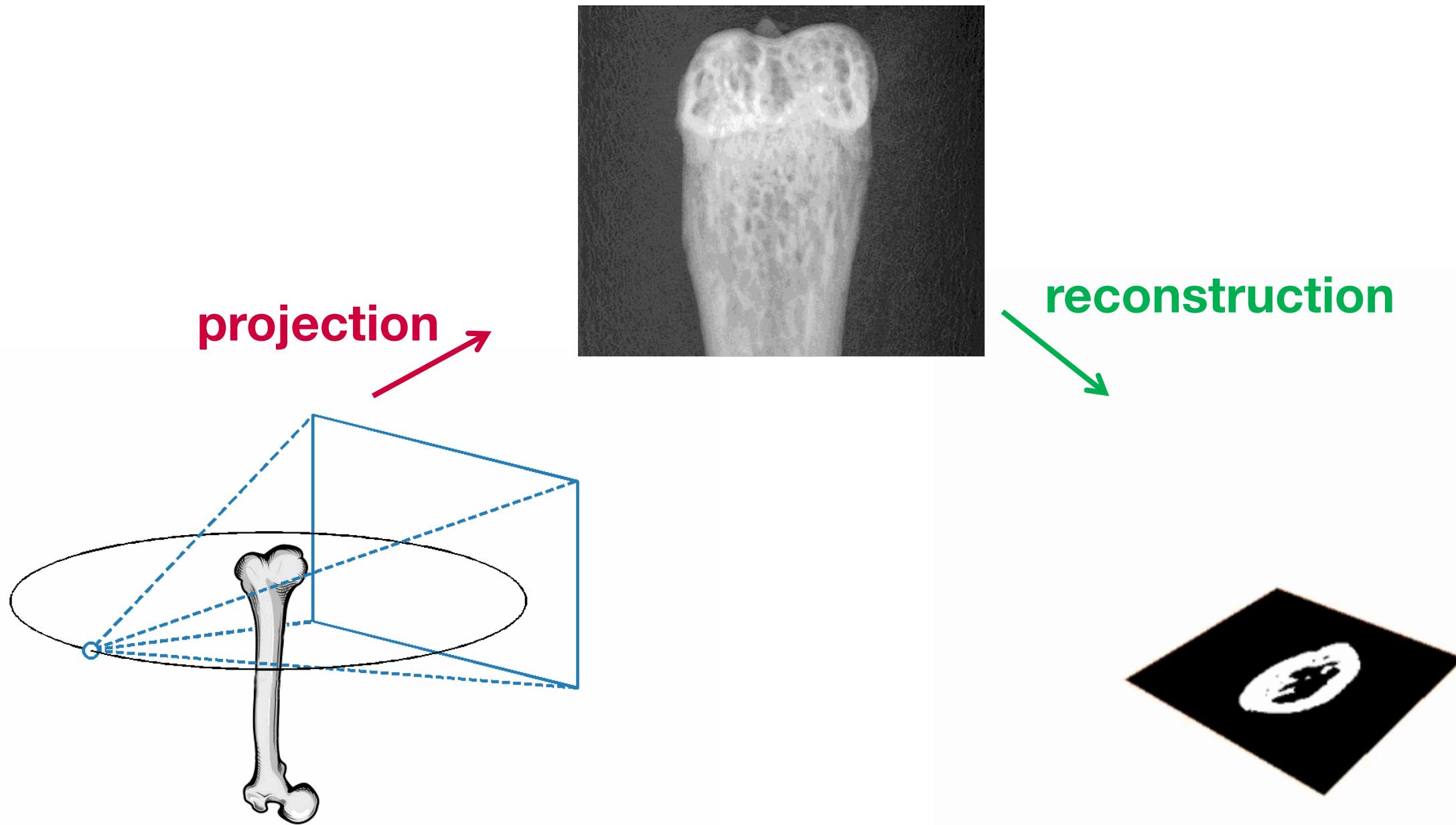
What is Computed Tomography



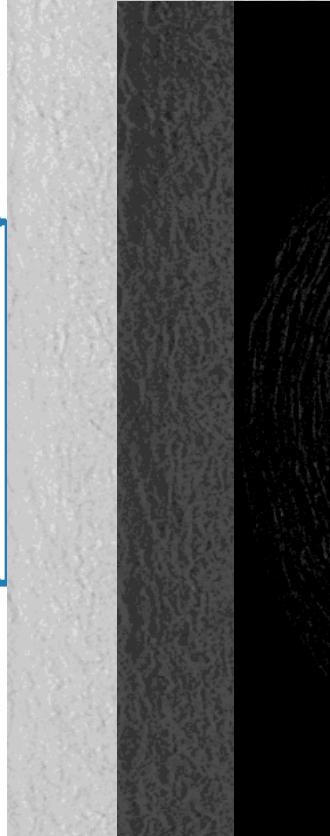
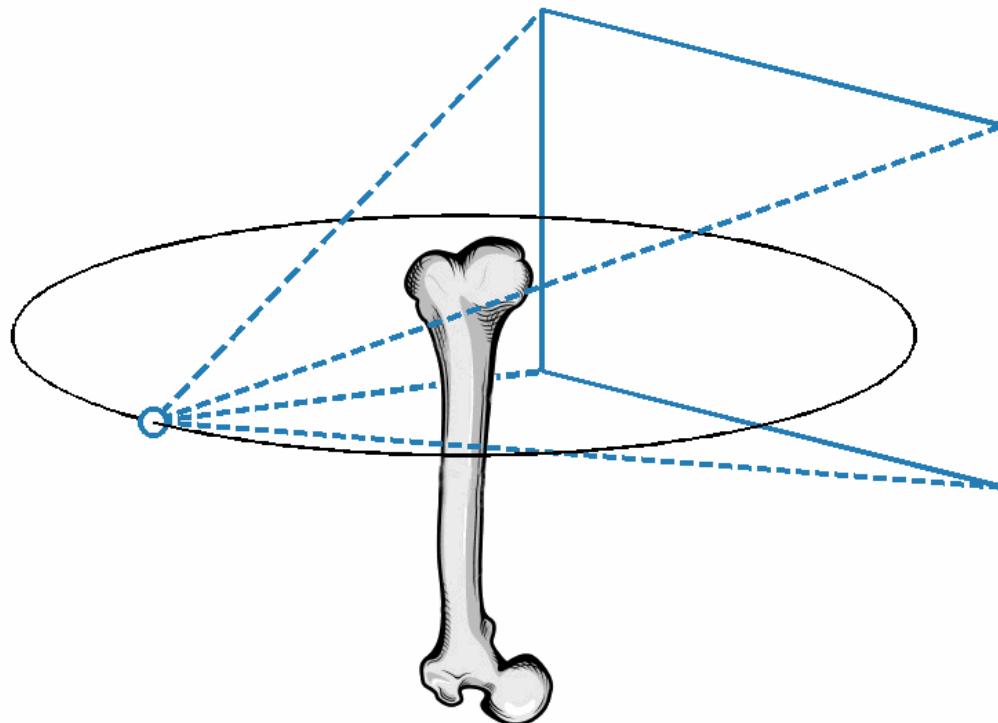
X-ray imaging



What is Computed Tomography



A typical CT workflow



bone parameter	value
BV/TV (%)	6.7
Tb.Th* (μm)	57
Tb.Sp* (μm)	306
Tb.N (1/mm)	1.17
SMI	2.5

acquisition

data
preprocessing

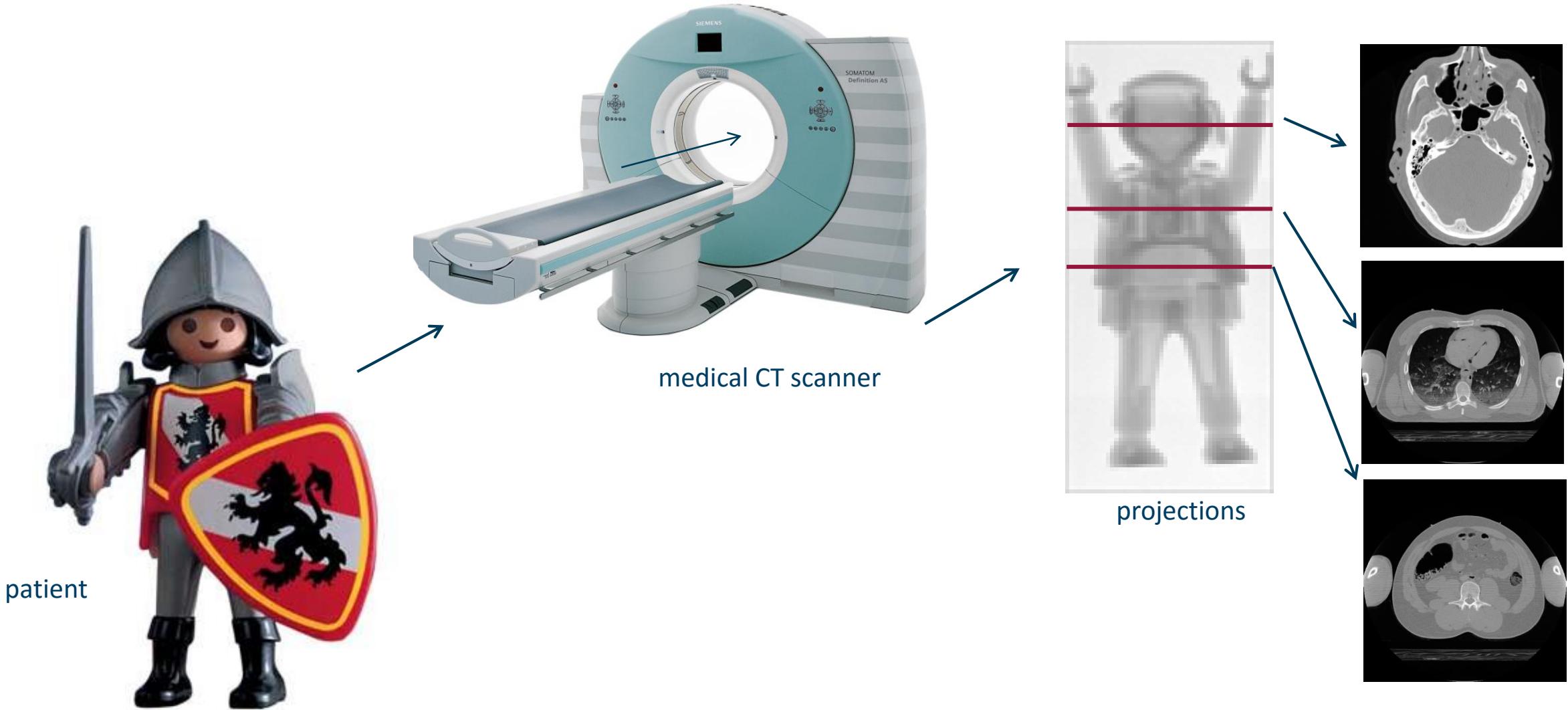
reconstruction

segmentation

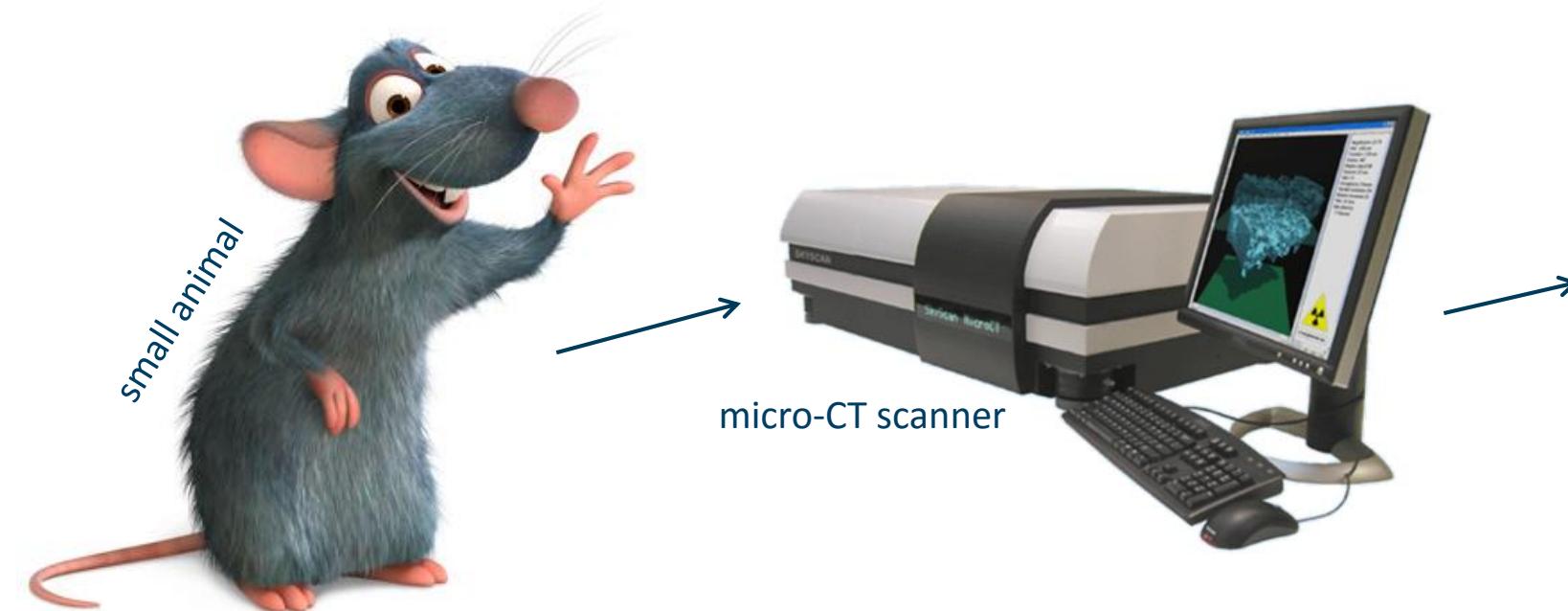
analysis

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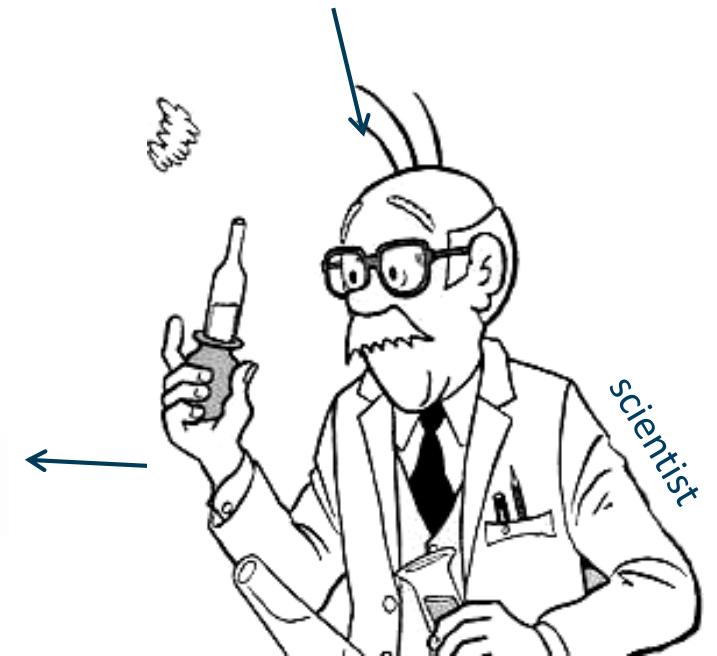
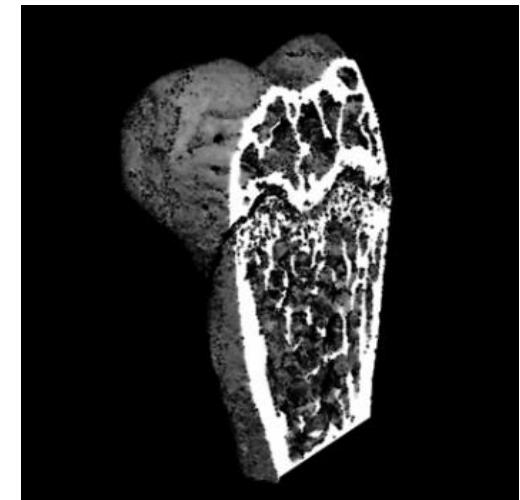
Medical CT



Biomedical CT



reconstruction of femur



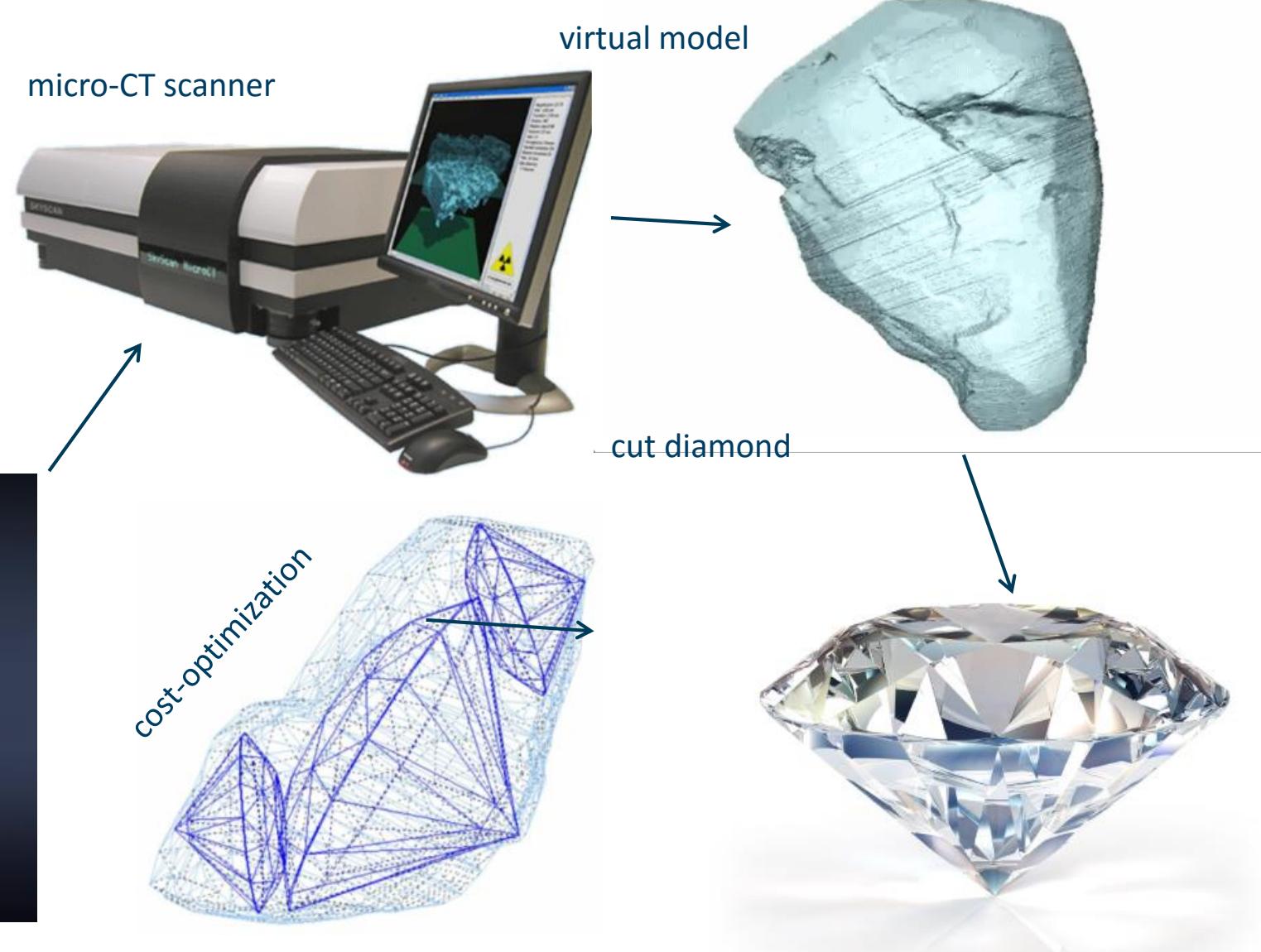
Industrial CT



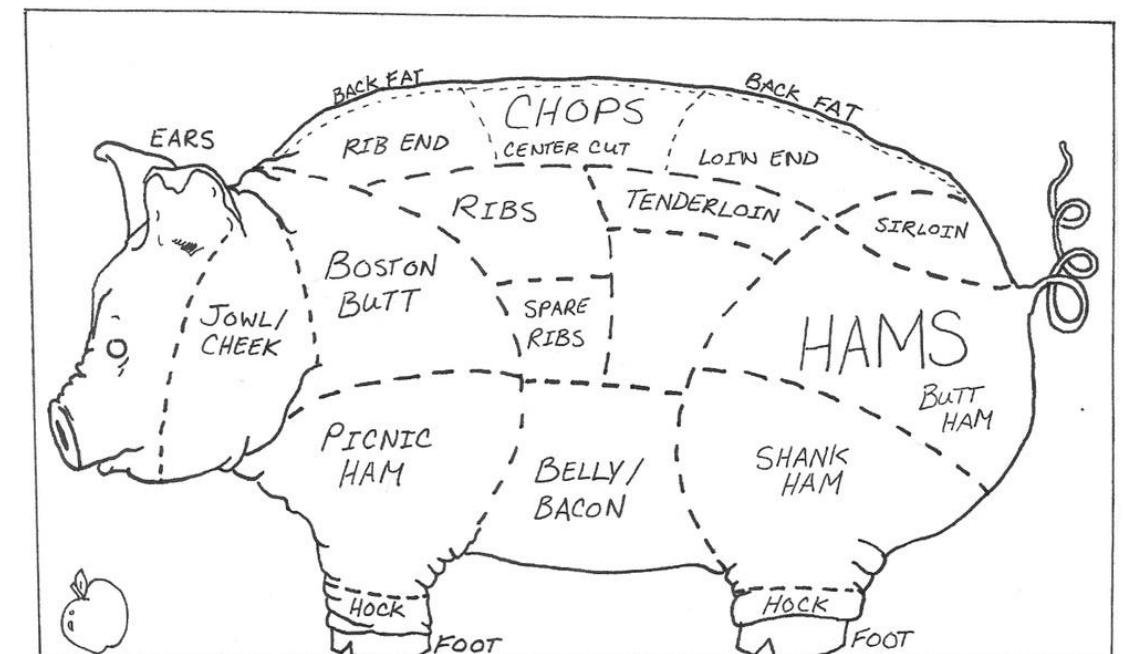
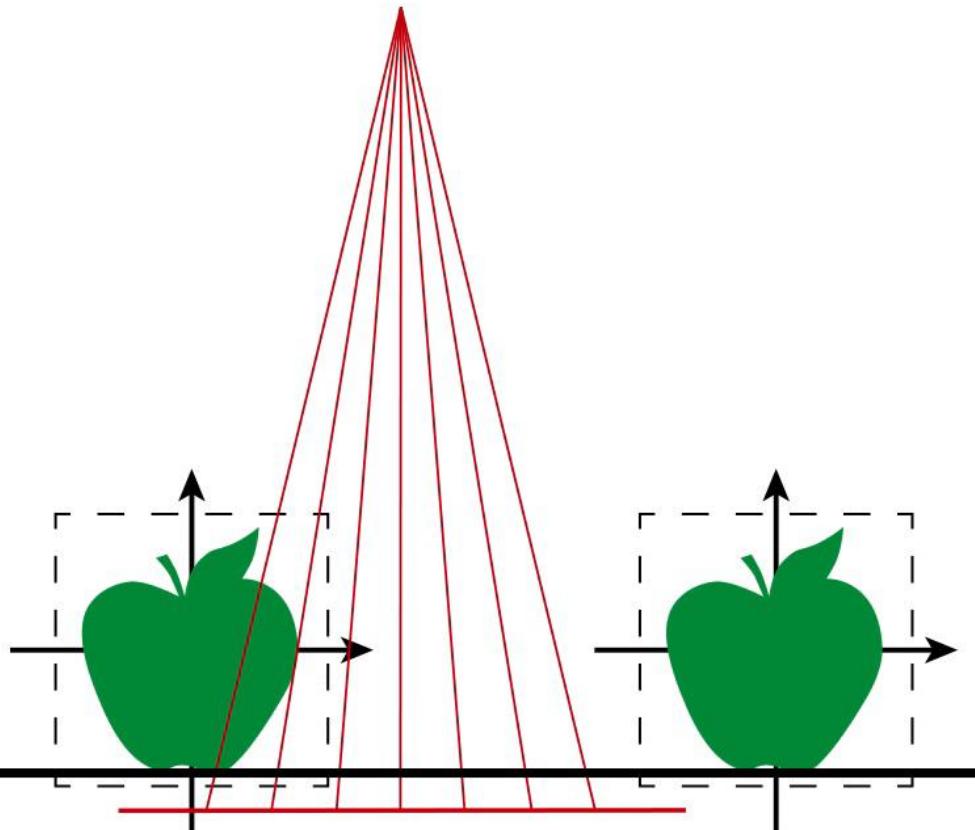
diamond mine



raw diamond



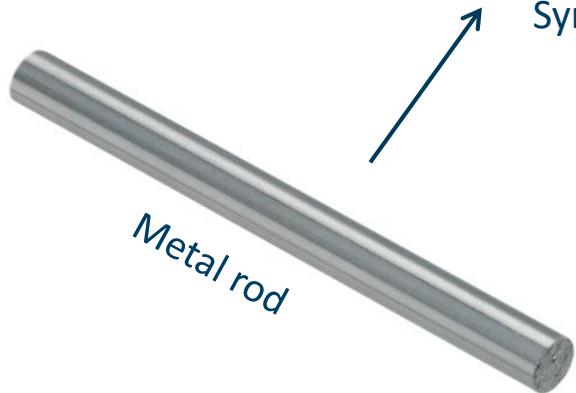
Food Processing



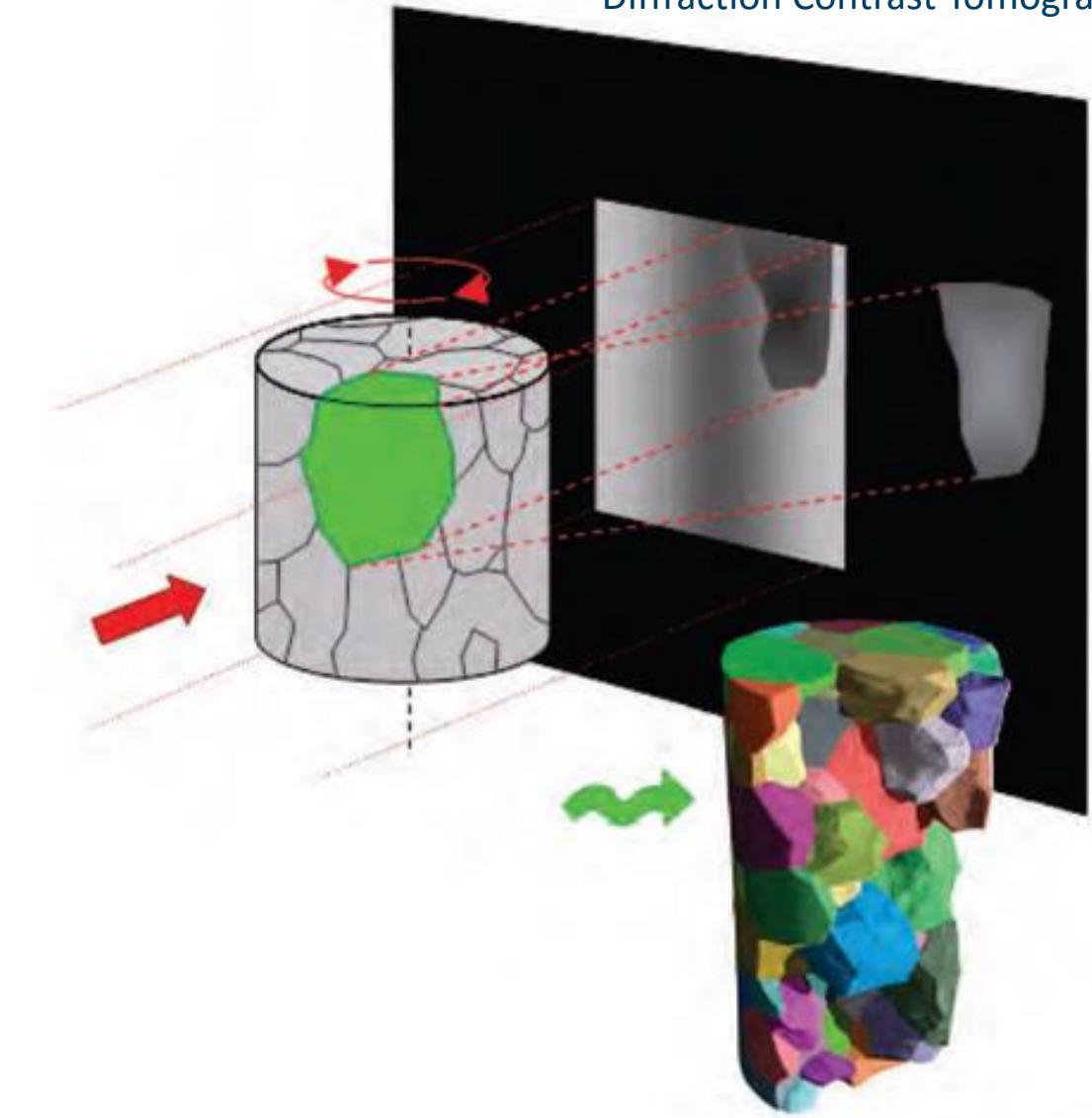
Material Science



Synchrotron facility

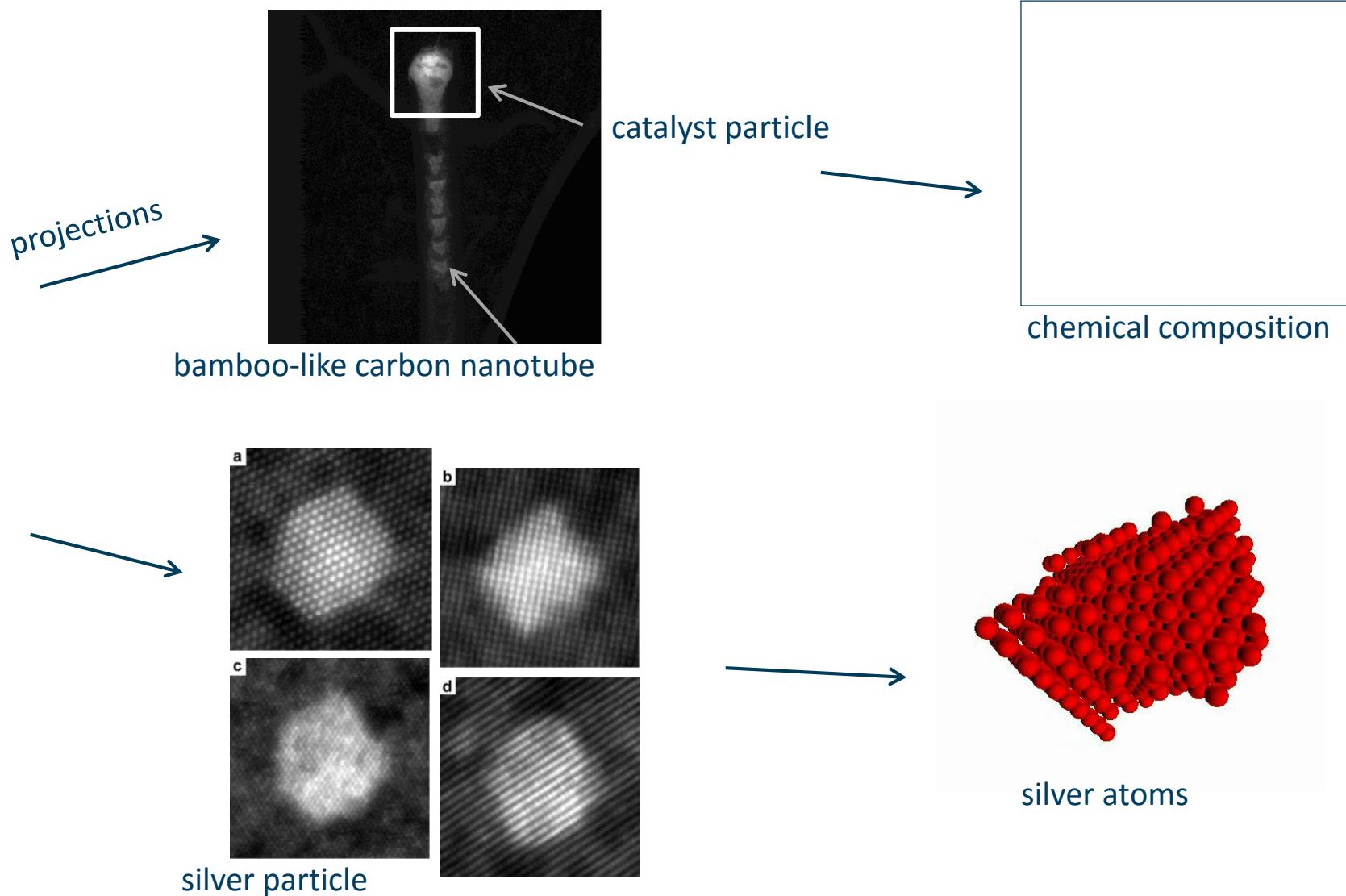


Diffraction Contrast Tomography

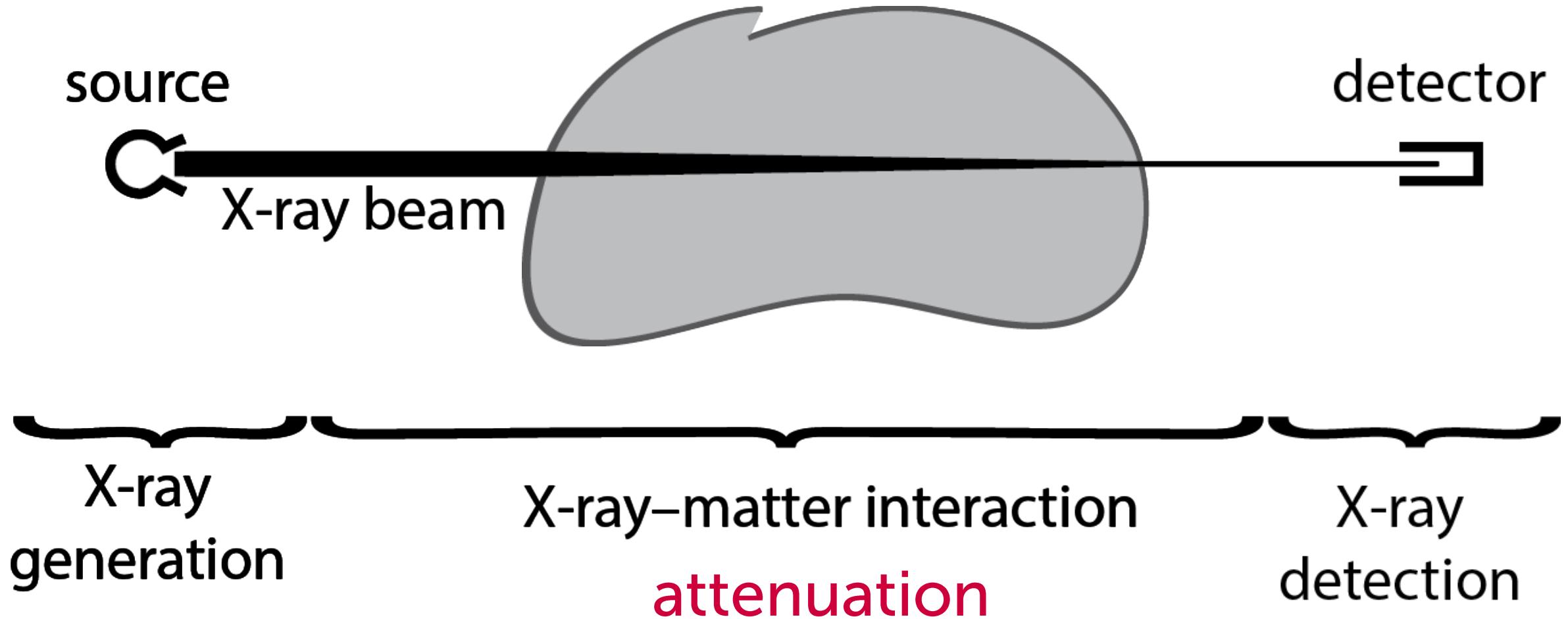




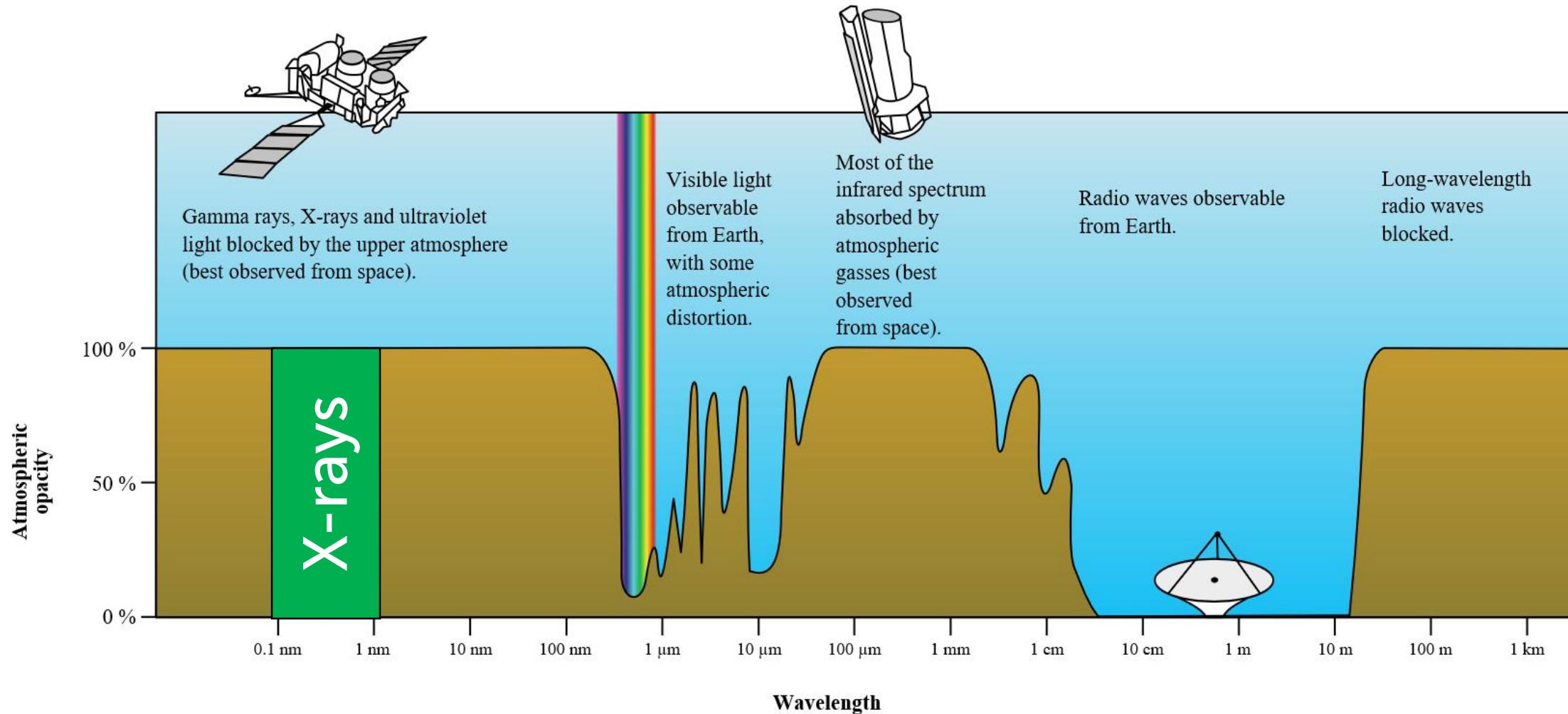
electron microscope



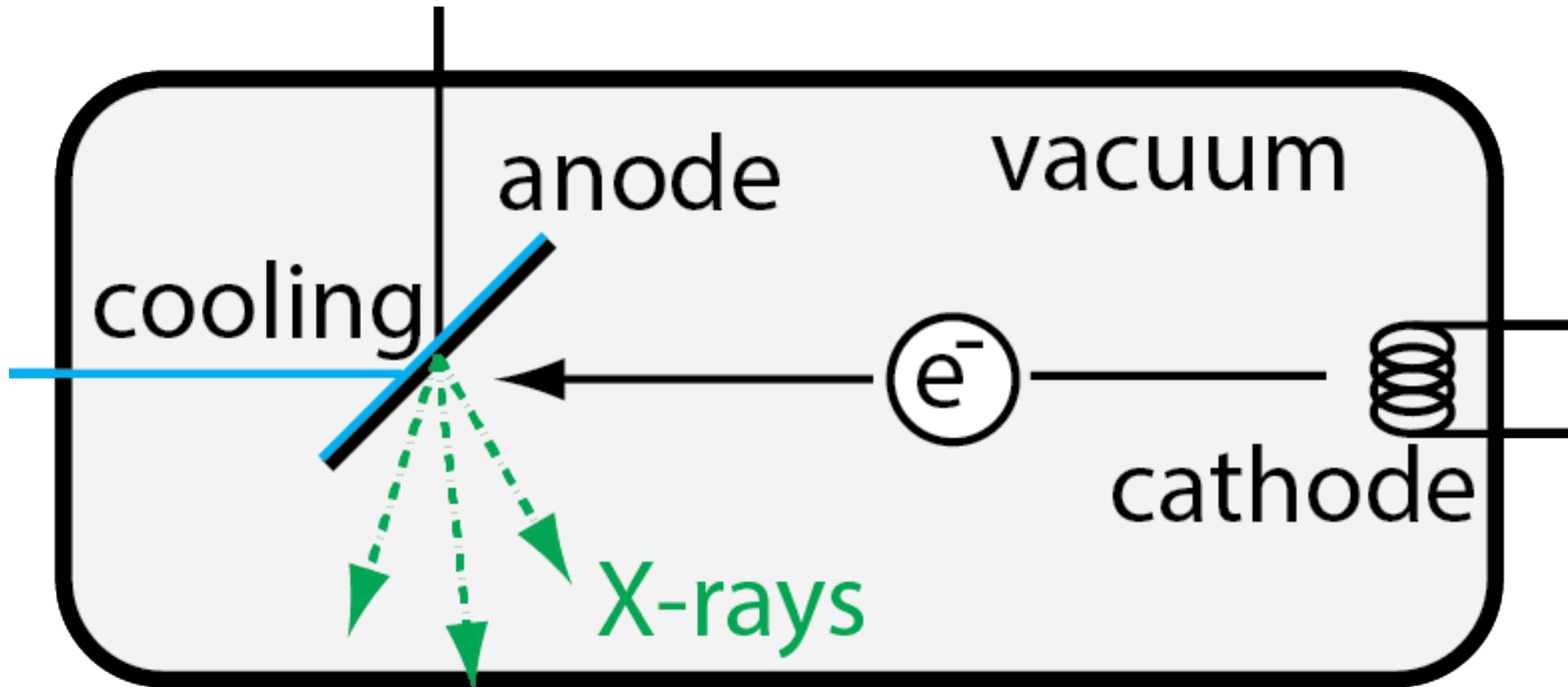
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- A Brief Introduction to Reconstruction Mathematics
- When Physics and Mathematics colide



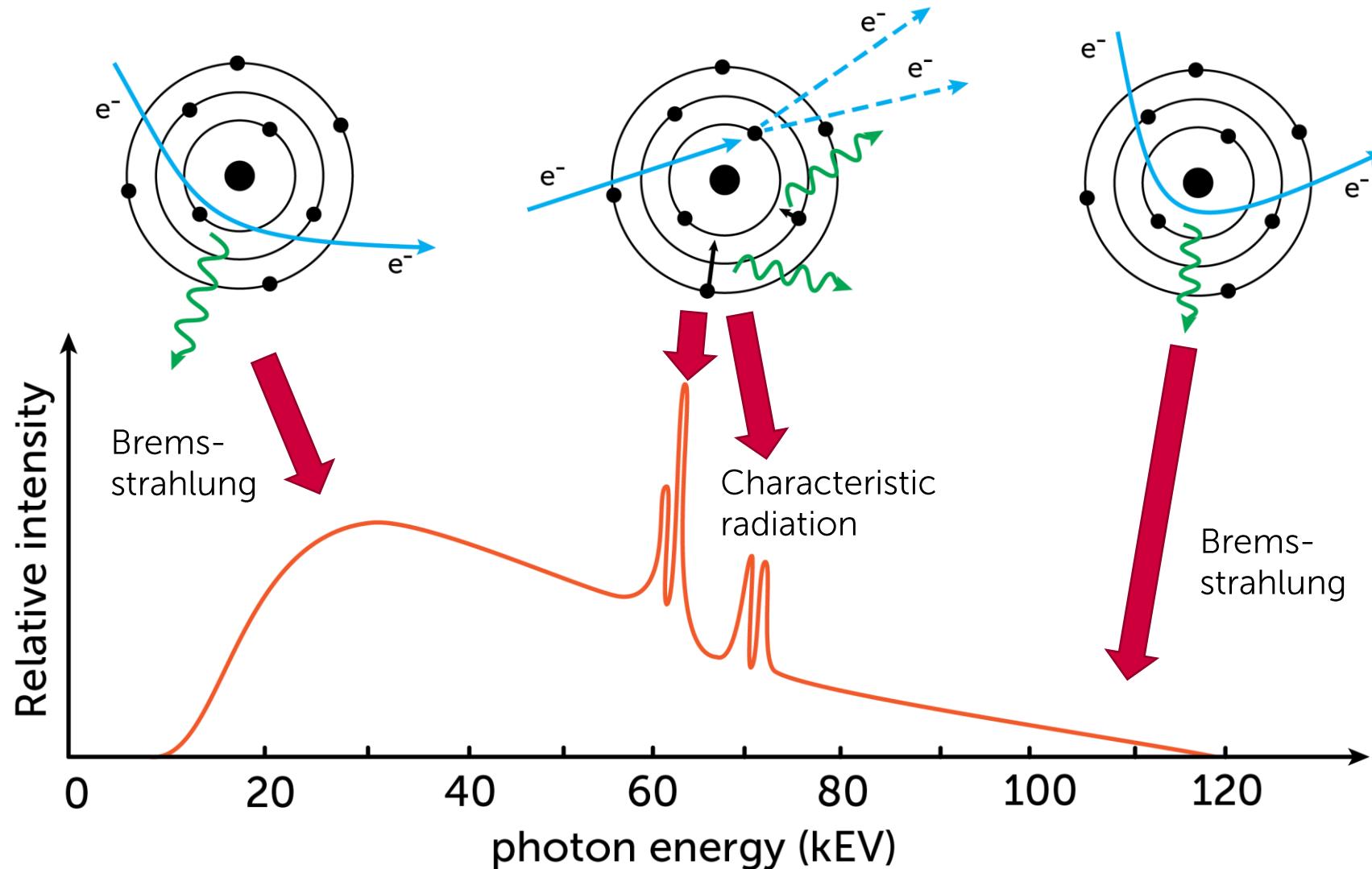
X-ray generation



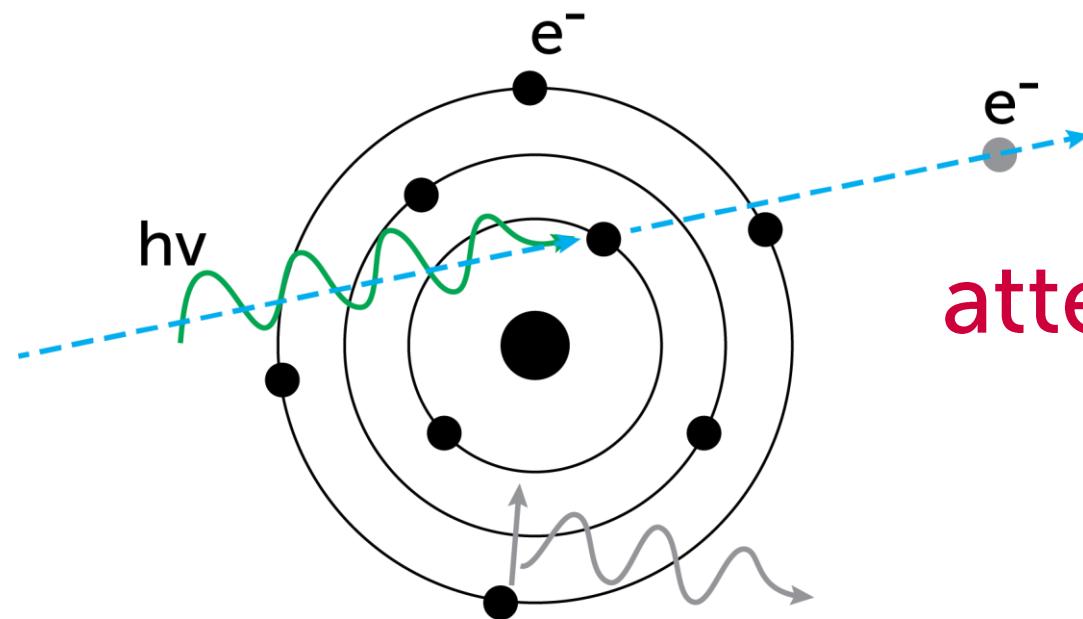
X-ray generation



X-ray generation - spectrum

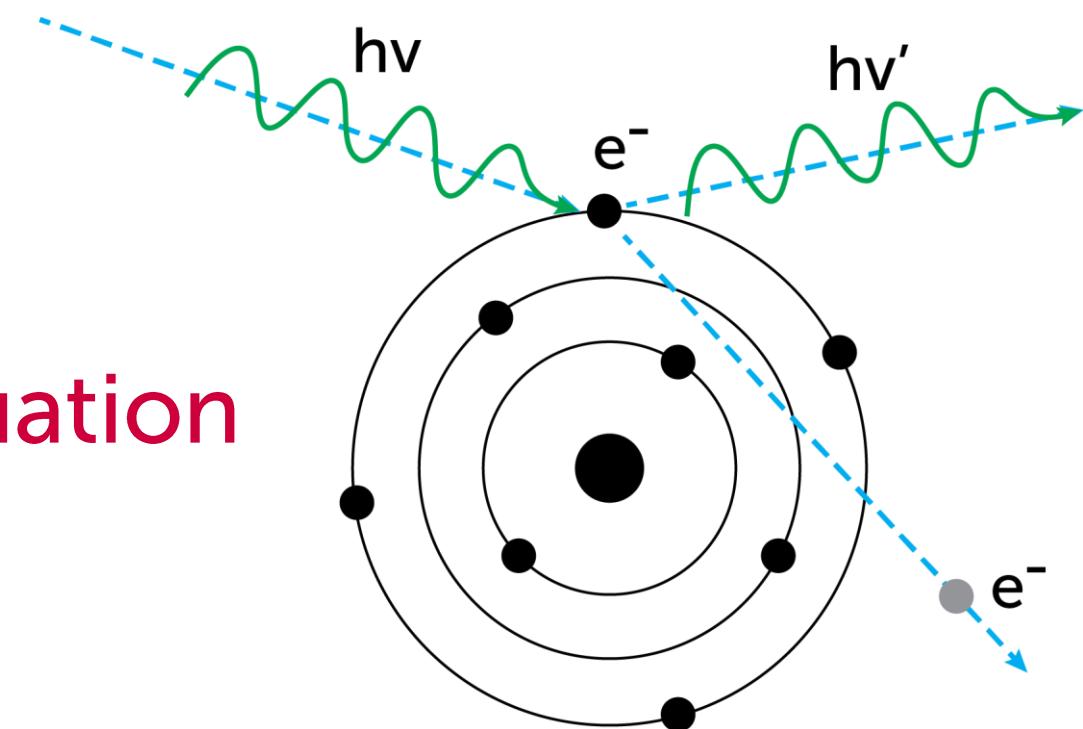


X-ray-matter interaction



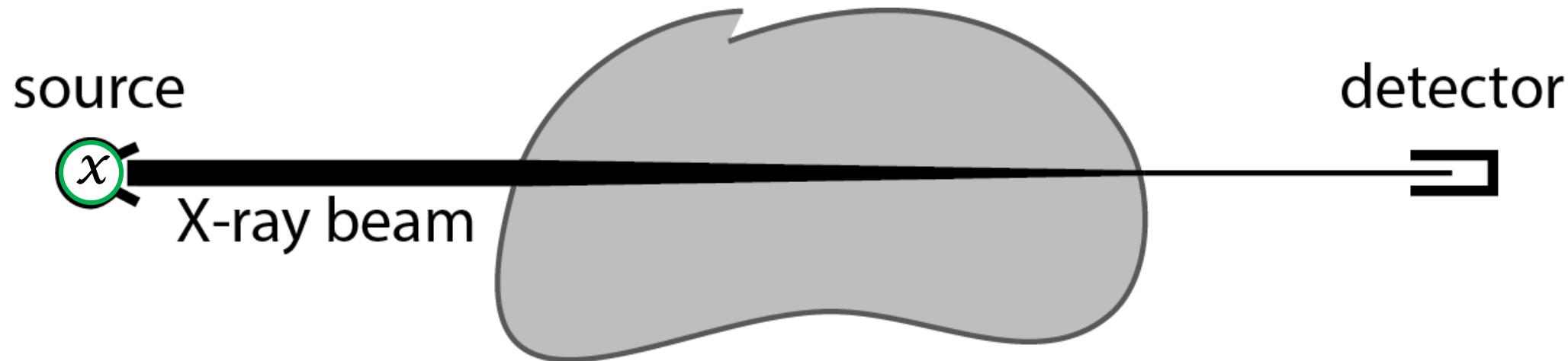
photoelectric effect

attenuation



Compton scattering

X-ray detection



Beam intensity at source

$$I_0(E)$$

Beam intensity at detector

$$I(E) = I_0(E) e^{-\int \mu(\xi, E) d\xi}$$

Beam intensity changes

$$I(\xi + \Delta\xi, E) = I(\xi) - \mu(\xi, E)I(\xi, E)\Delta\xi$$



$$I = I_0 e^{-\int \mu(\xi) d\xi}$$



$$D_0$$

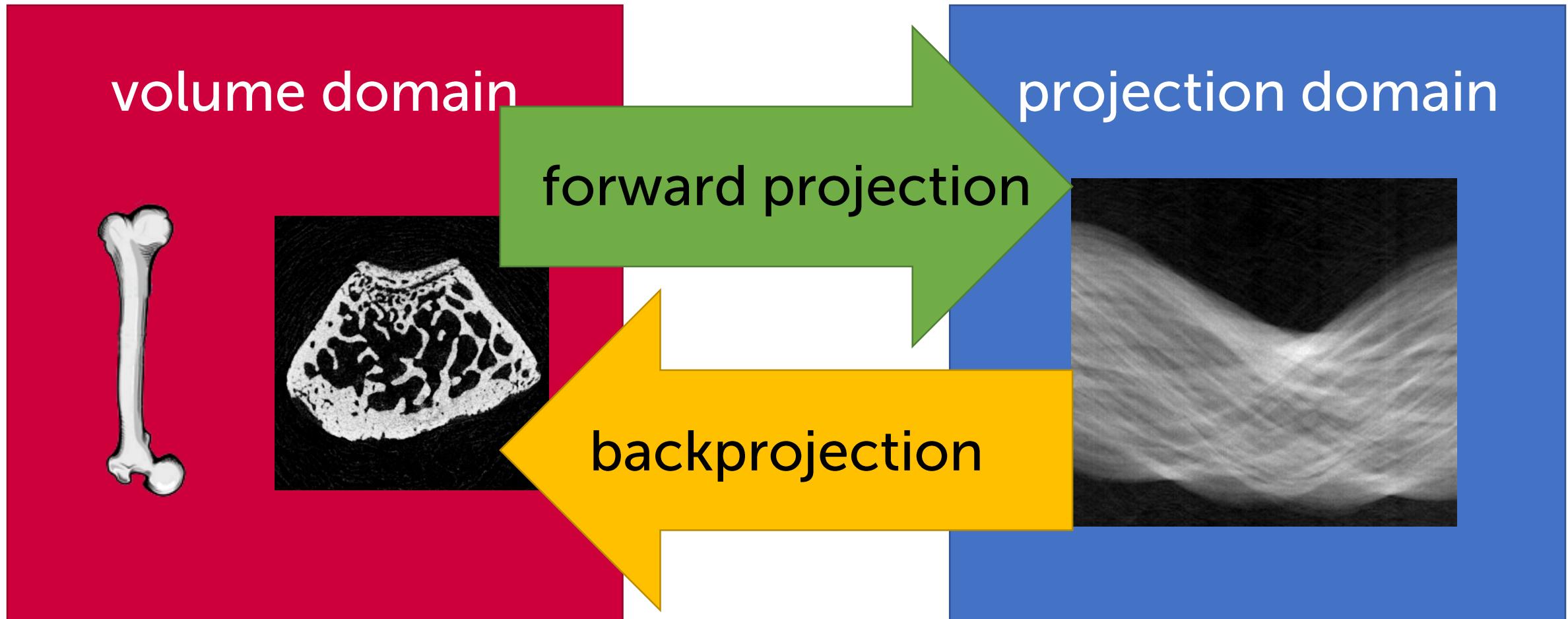


$$I_0$$



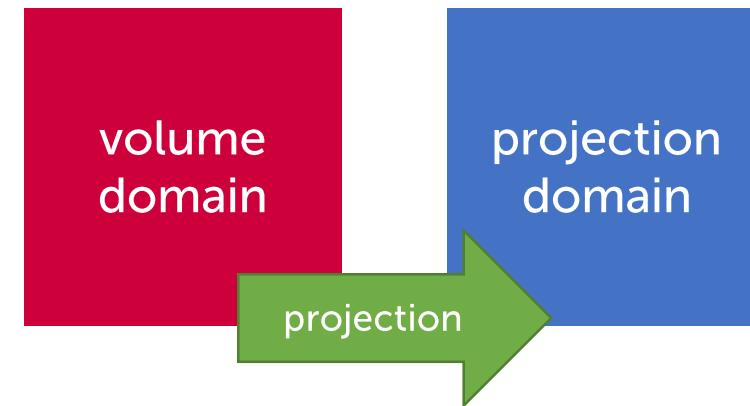
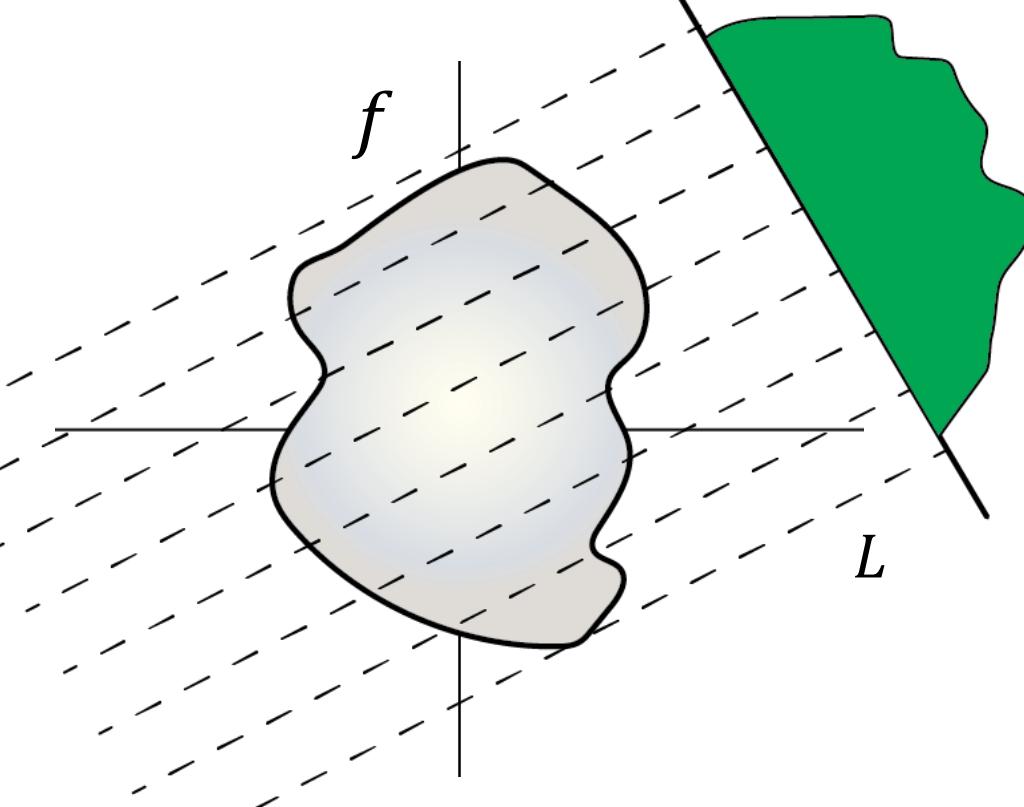
$$\text{Law of Beer-Lambert: } A = -\ln \left(\frac{I-D}{I_0-D} \right)$$

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 - **Analytical approach**
 - **Algebraic approach**
- When Physics and Mathematics colide



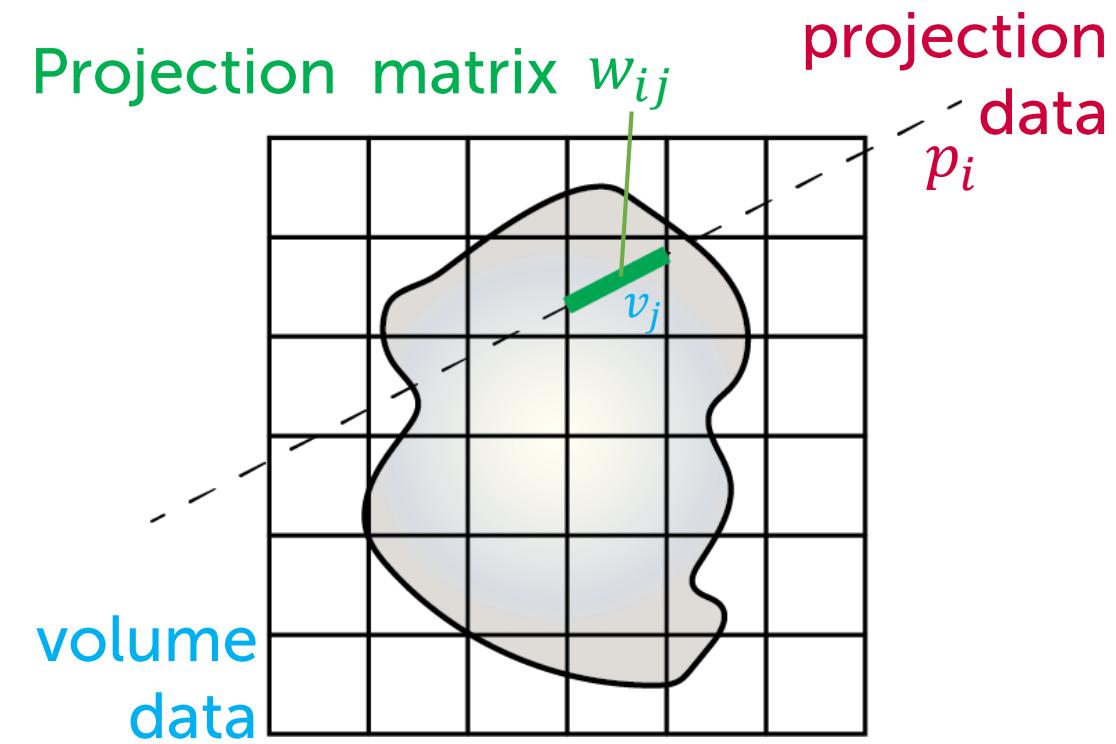
analytically

$$Rf(L) = \int_L f(x)|dx|$$



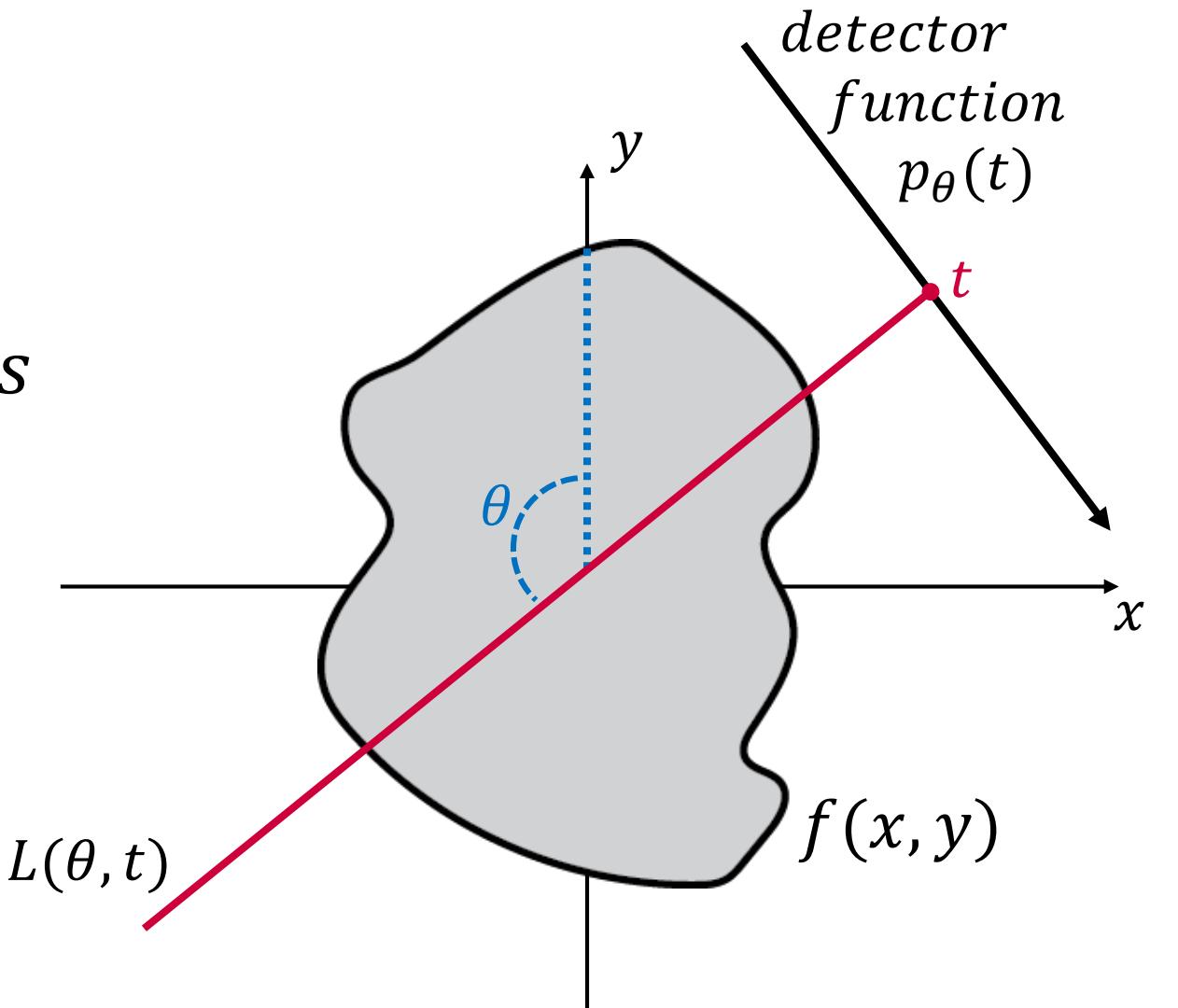
algebraically

$$p = Wv$$



Radon transform

$$p_\theta(t) = \int_{L(\theta,t)} f(x,y) ds$$

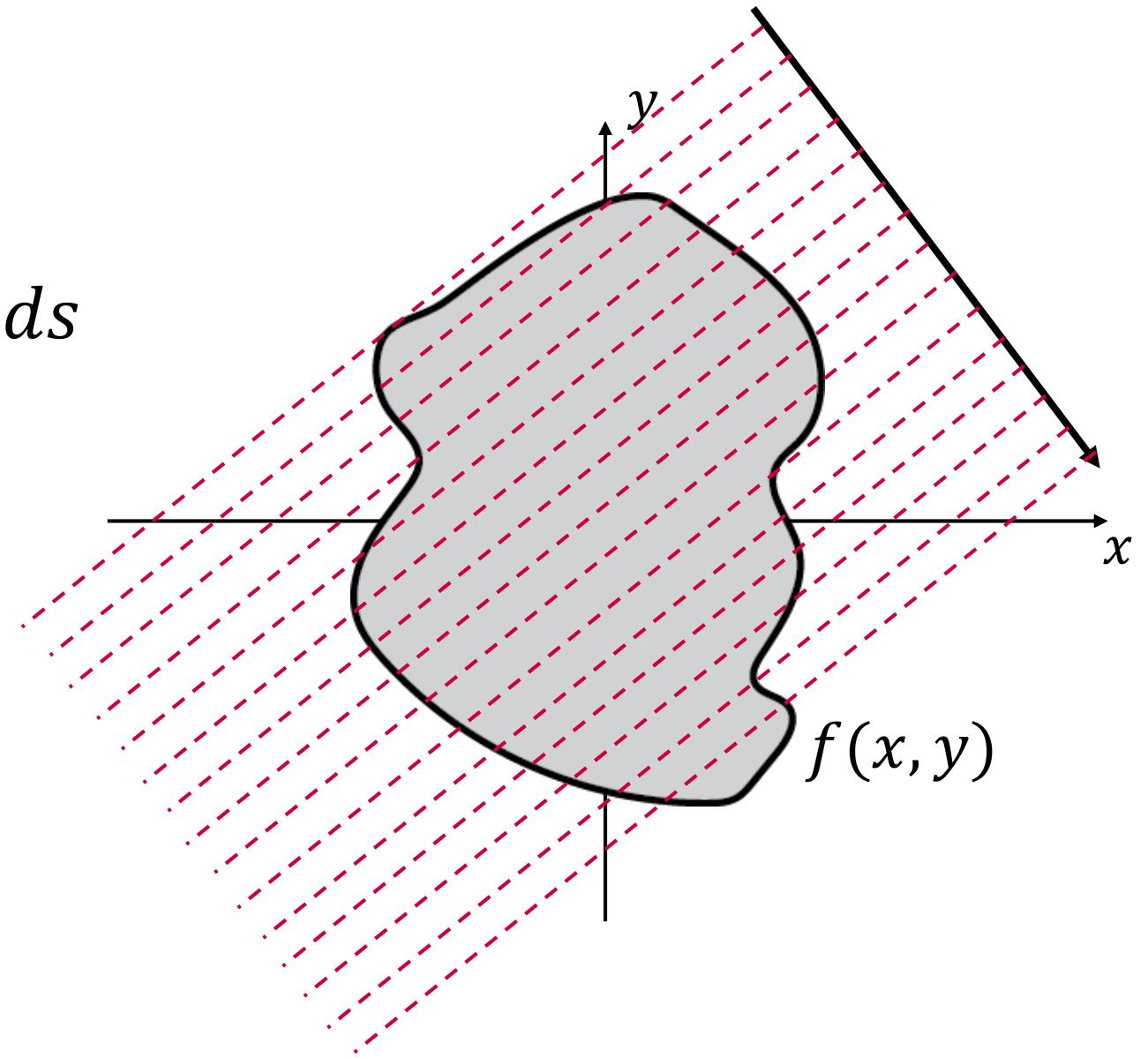
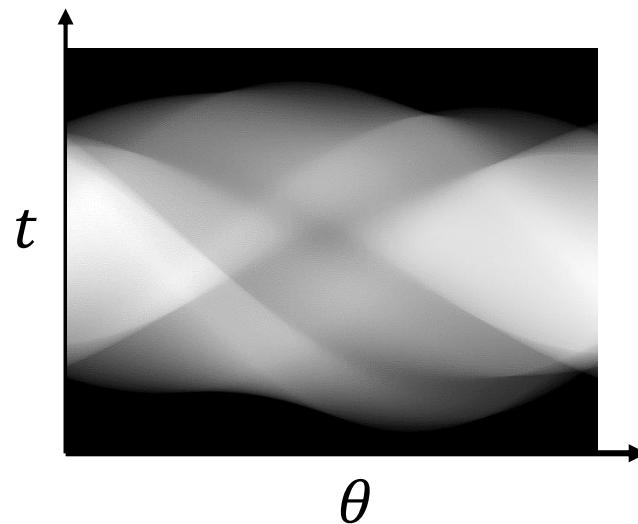


$$L(\theta, t) = \{(x, y) \in \mathbb{R} \times \mathbb{R}: x \cos \theta + y \sin \theta = t\}$$

Radon transform

$$\mathcal{R}f(\theta, t) = \int_{L(\theta, t)} f(x, y) ds$$

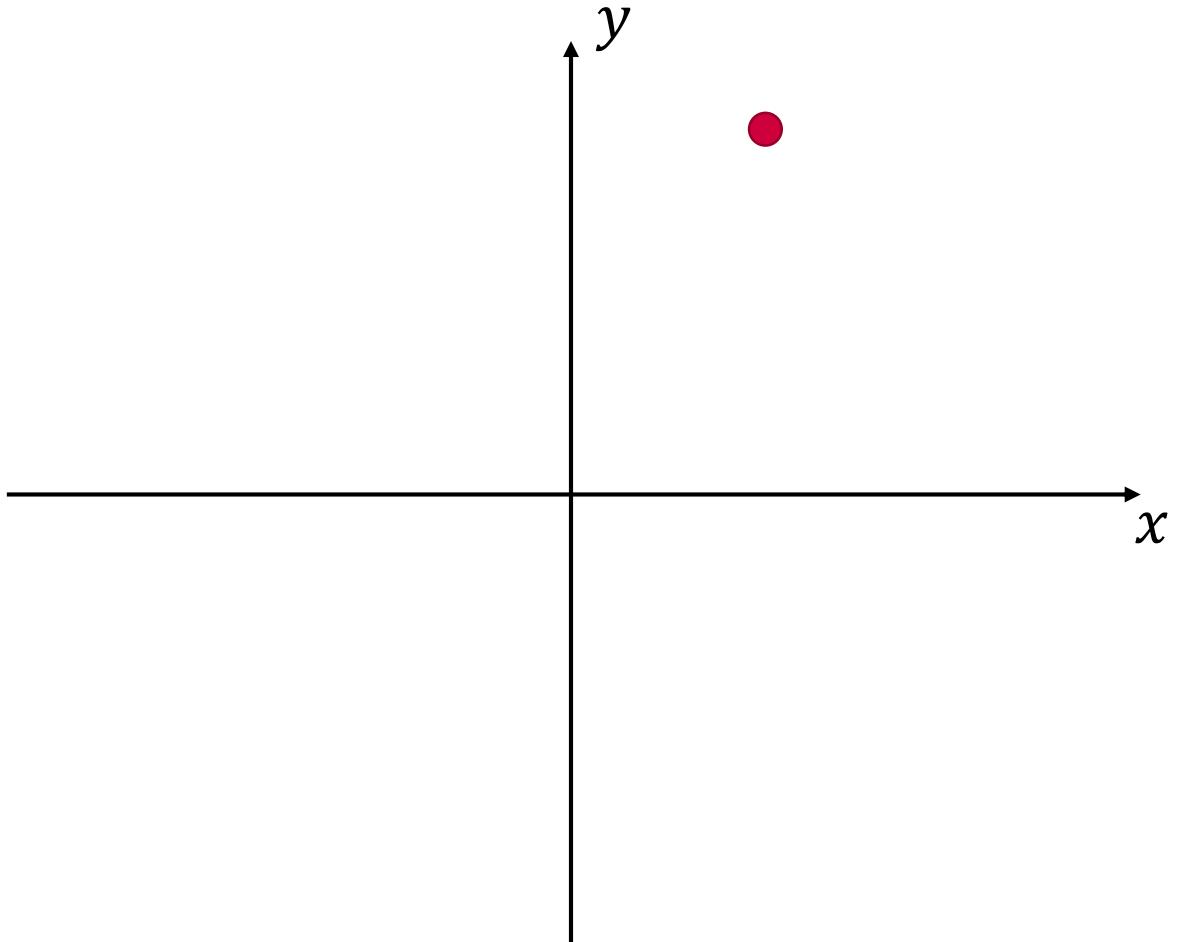
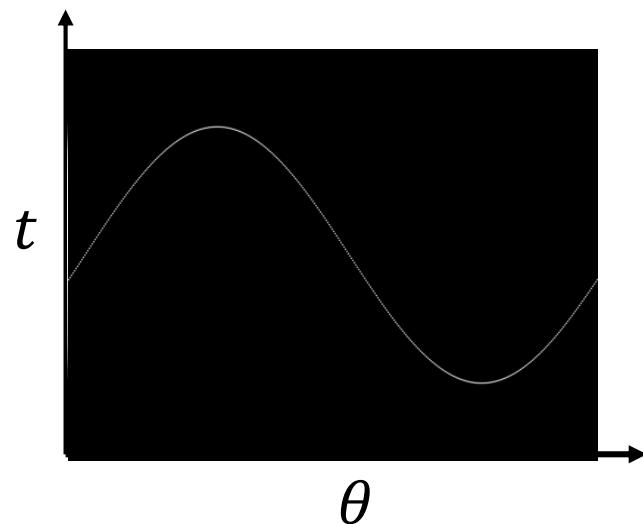
sinogram



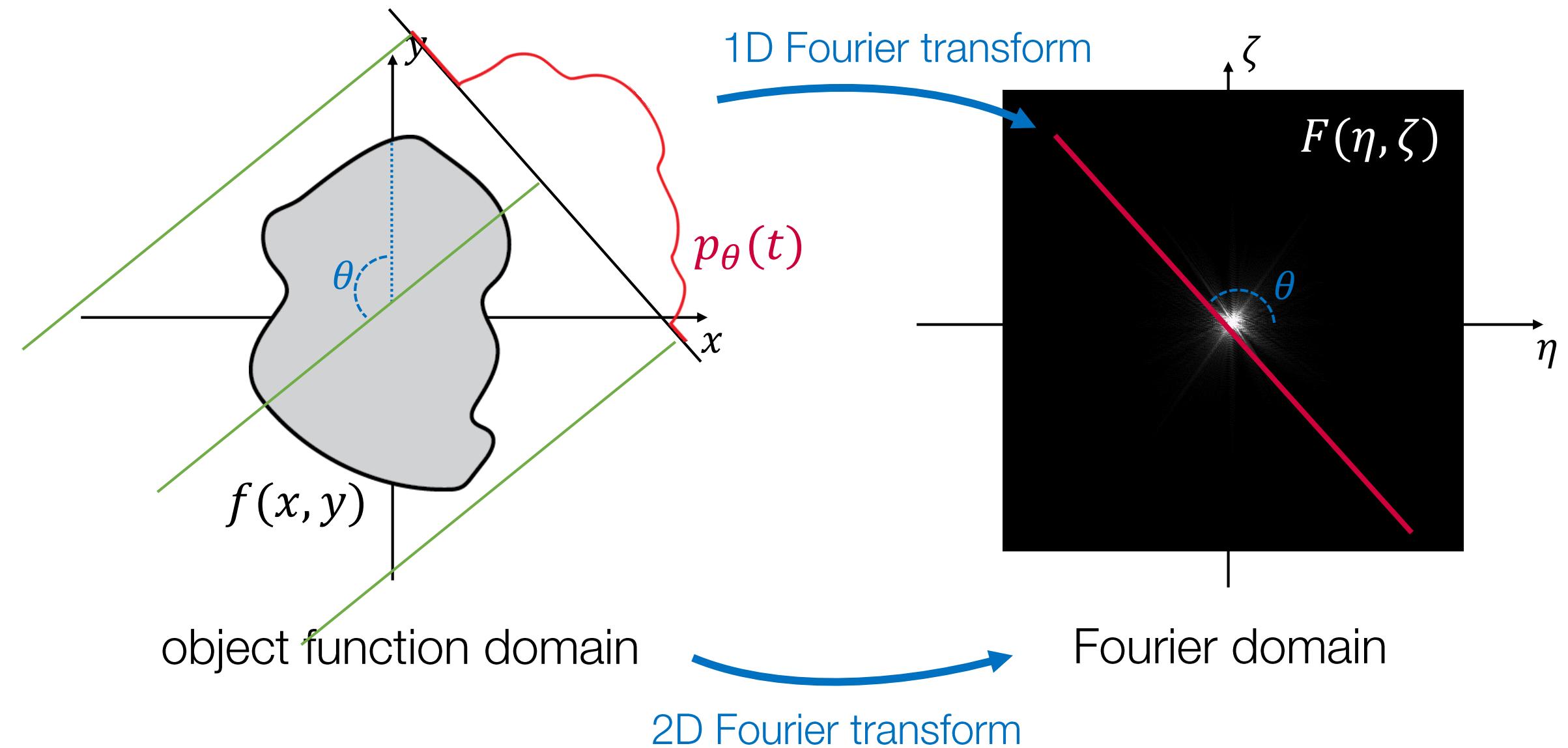
Radon transform

$$\boxed{\mathcal{R}f(\theta, t) = \int_{L(\theta, t)} f(x, y) ds}$$

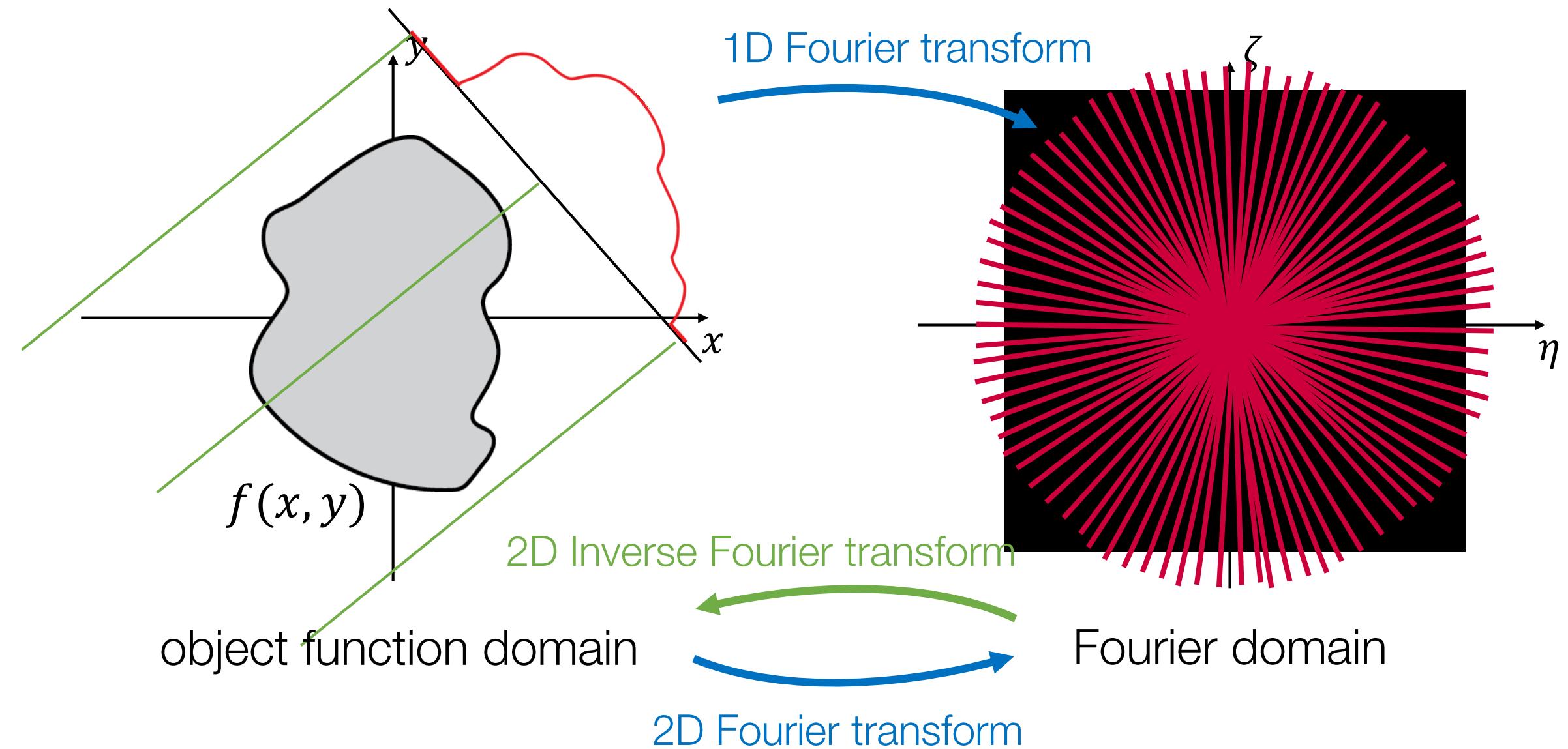
sinogram



Fourier Slice Theorem



Fourier Slice Theorem



Fourier Slice Theorem Proof

The 2D Fourier transform of $f(x, y)$ is:

$$F(\zeta, \eta) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x, y) e^{-i2\pi(\eta x + \zeta y)} dx dy$$

The 1D Fourier transform of $p_\theta(t)$ is:

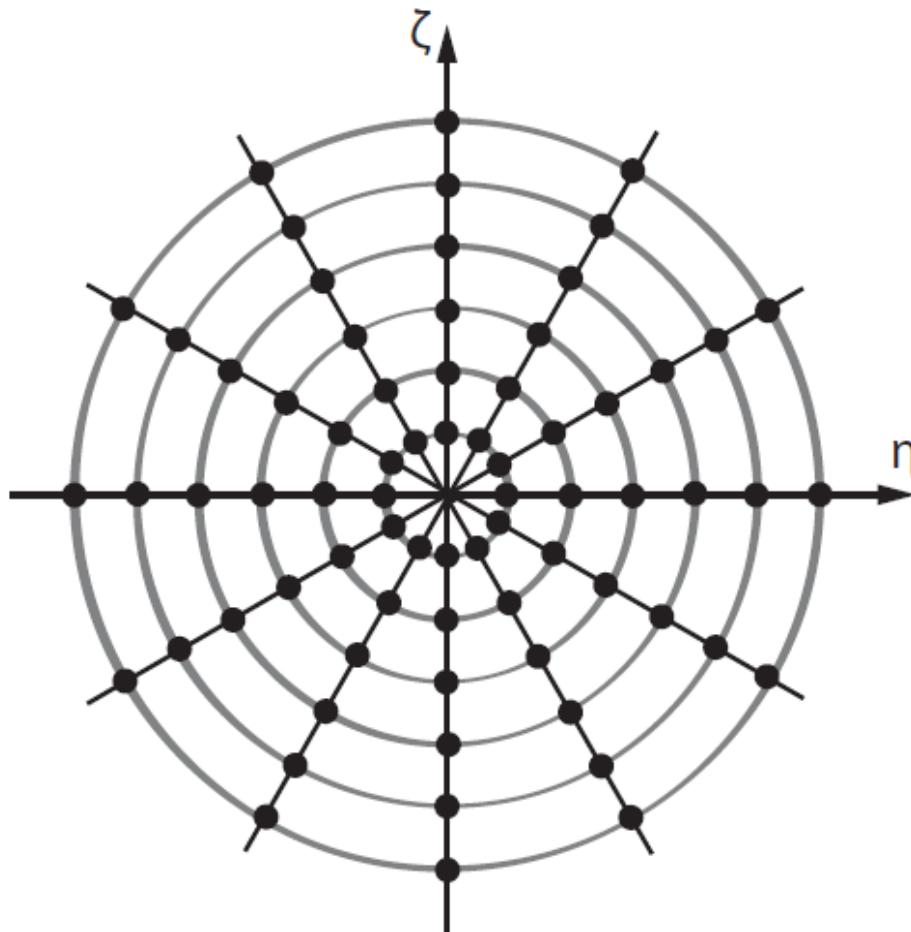
$$P_\theta(\omega) = \int_{-\infty}^{\infty} p_\theta(t) e^{i2\pi\omega t} dt$$

Substituting $\eta = \omega \cos \theta$ and $\zeta = \omega \sin \theta$:
$$p_\theta(t) = \mathcal{R}f(\theta, t) = \int_{L(\theta, t)} f(x, y) ds$$

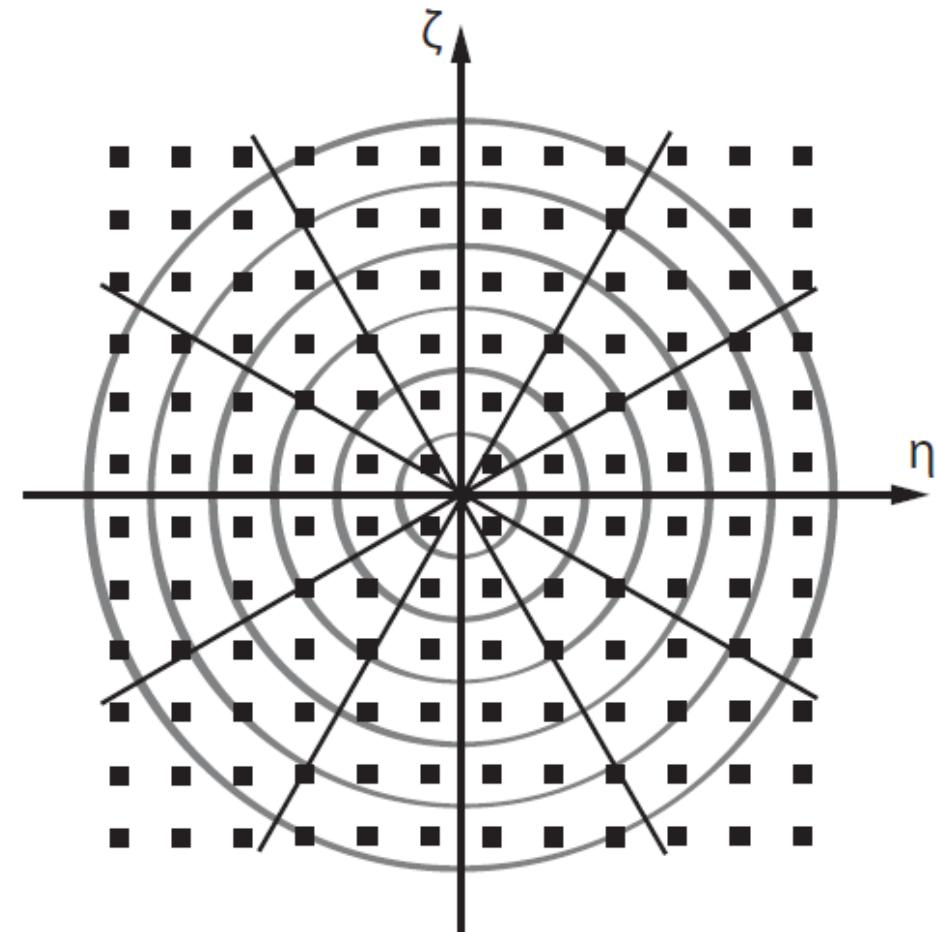
$$P_\theta(\omega) = F(\omega \cos \theta, \omega \sin \theta)$$

$$L(\theta, t) = \{(x, y) \in \mathbb{R} \times \mathbb{R}: x \cos \theta + y \sin \theta = t\}$$

Fourier Slice Theorem



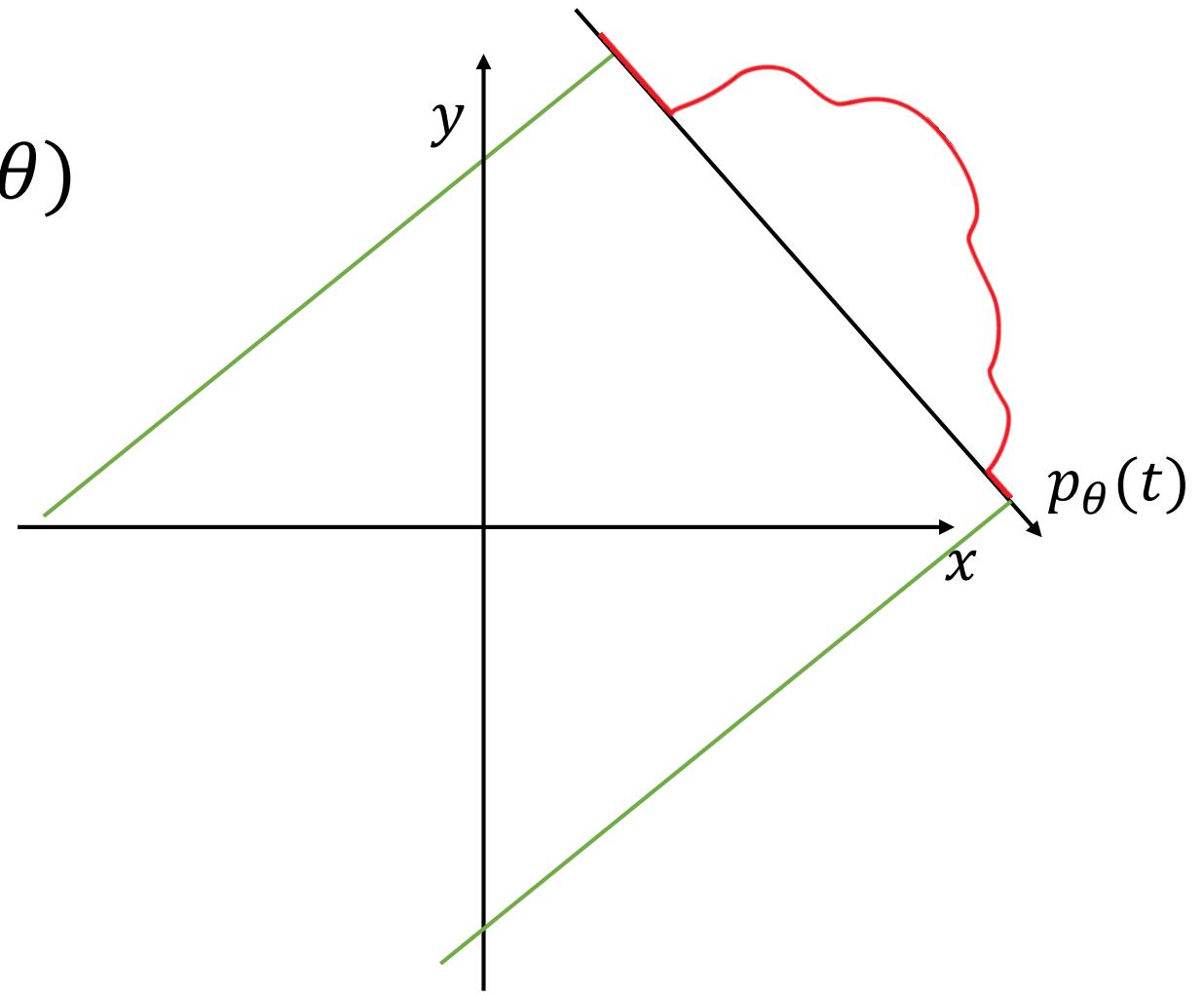
Fourier sampling with Fourier
Slice Theorem



Sampling required by Fast
Fourier Transform (FFT)

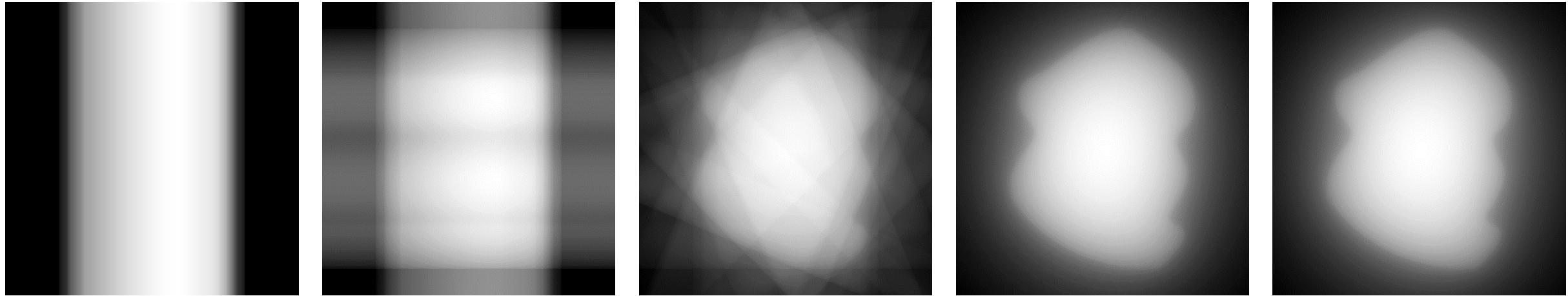
Filtered Backprojection

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



Filtered Backprojection (FBP)

backprojection



1 angle

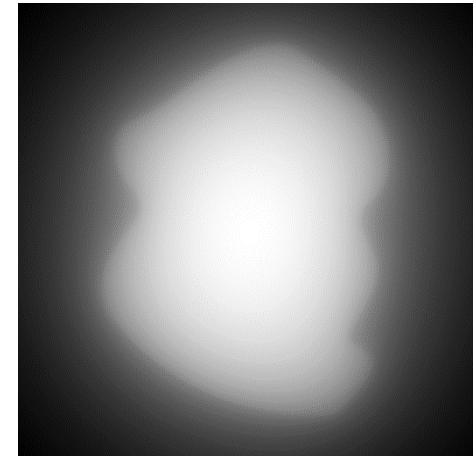
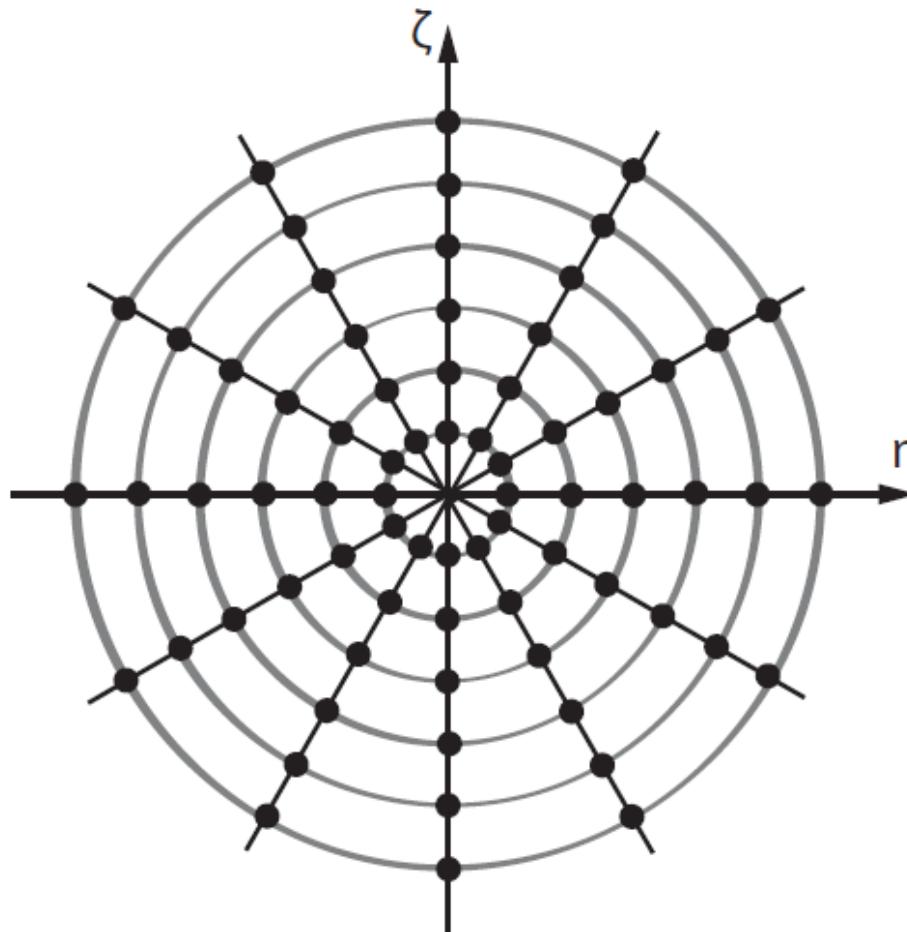
2 angles

8 angles

45 angles

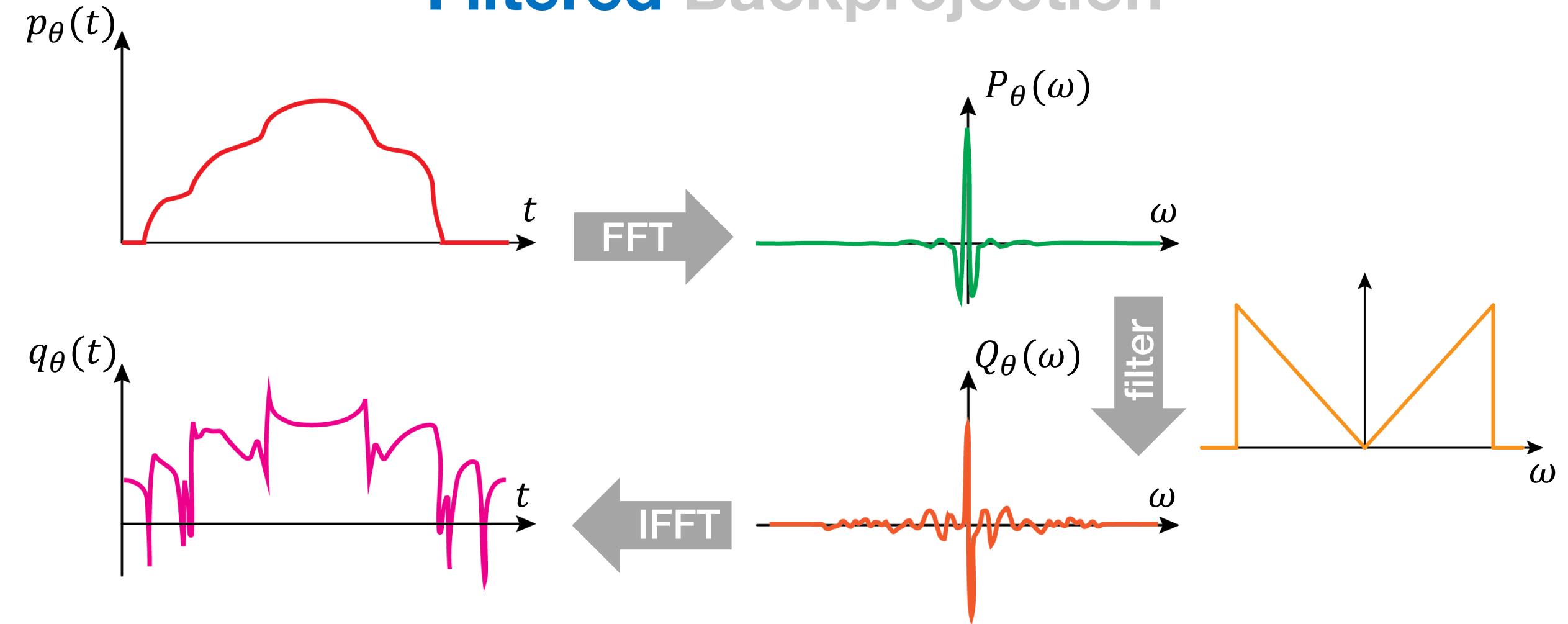
180 angles

Filtered Backprojection (FBP)



Sampling of Fourier domain

Filtered Backprojection



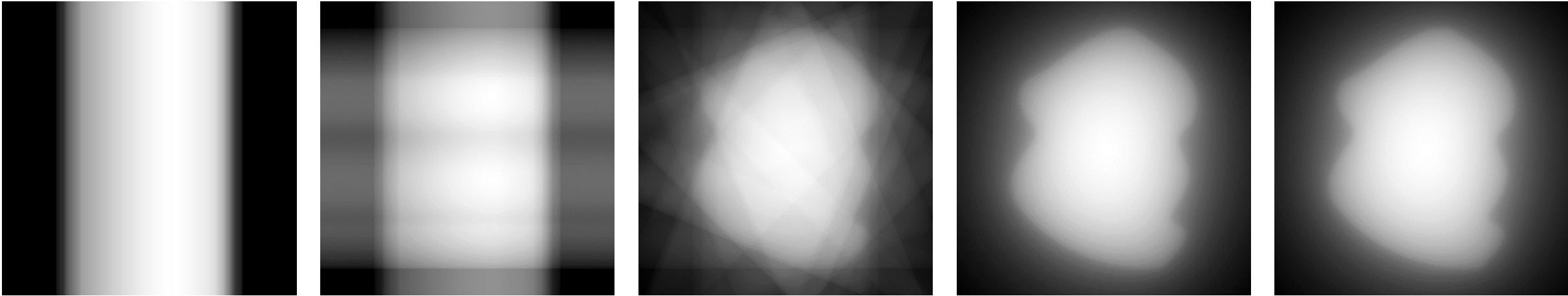
Filtered Backprojection

$$ff_{bp}(x, y) = \int q_\theta(x \cos \theta + y \sin \theta) d\theta$$

with $q_\theta(t) = \int P_\theta(\omega) |\omega| e^{i2\pi\omega t} d\omega$

Filtered Backprojection (FBP)

backprojection



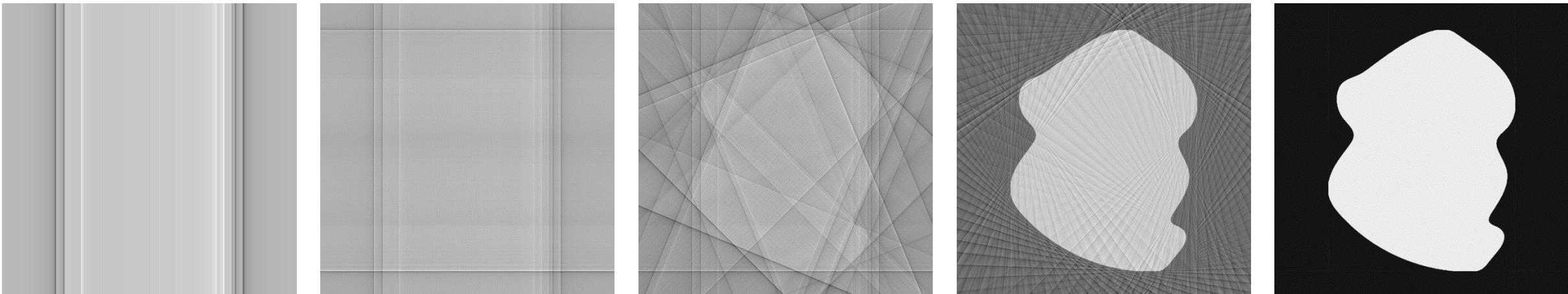
1 angle

2 angles

8 angles

45 angles

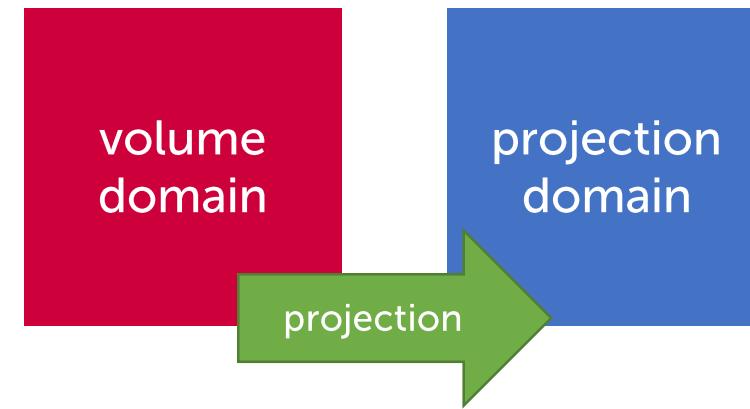
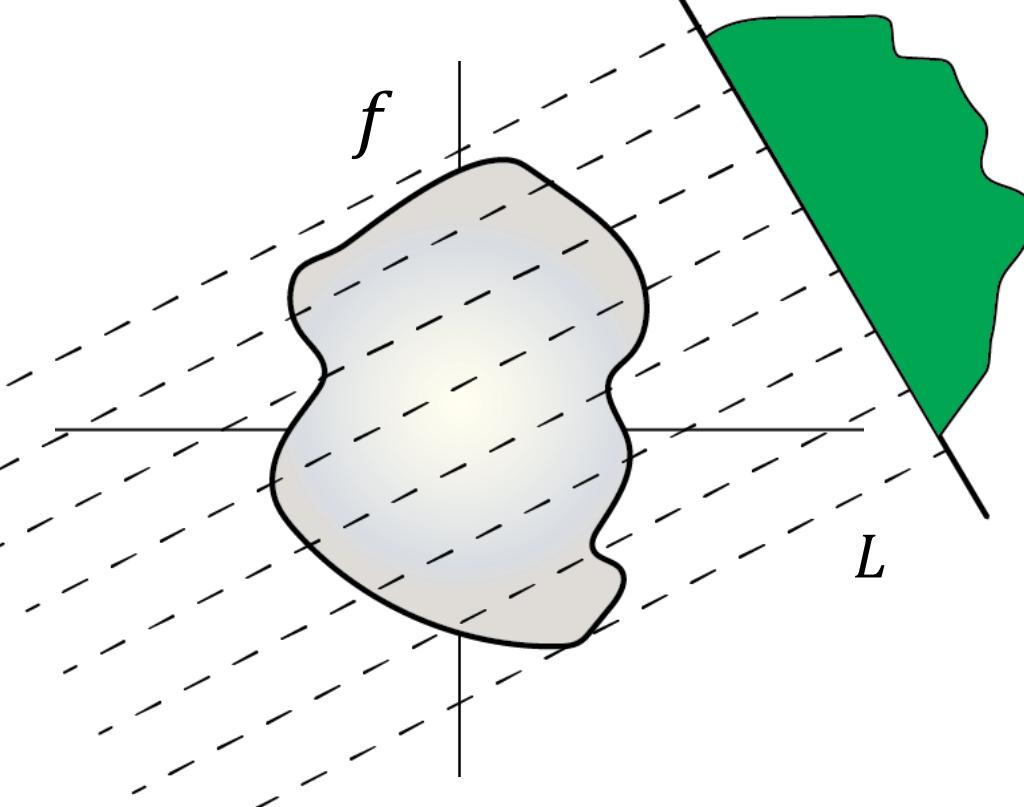
180 angles



filtered backprojection

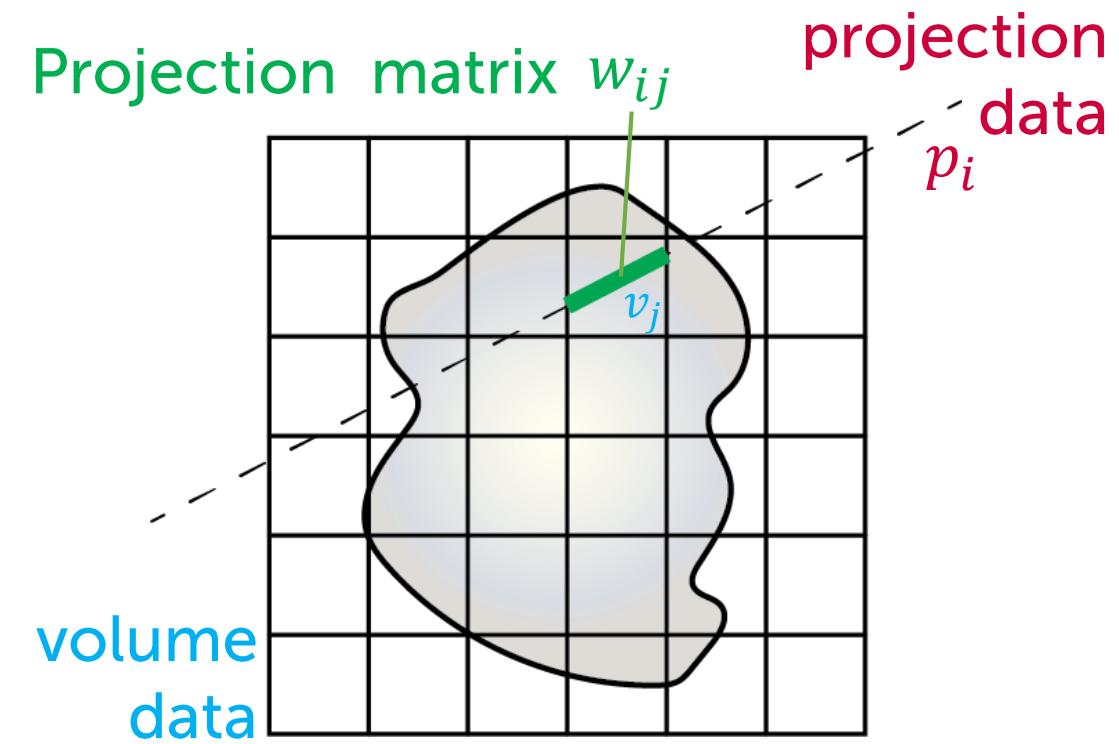
analytically

$$Rf(L) = \int_L f(x)|dx|$$



algebraically

$$p = Wv$$



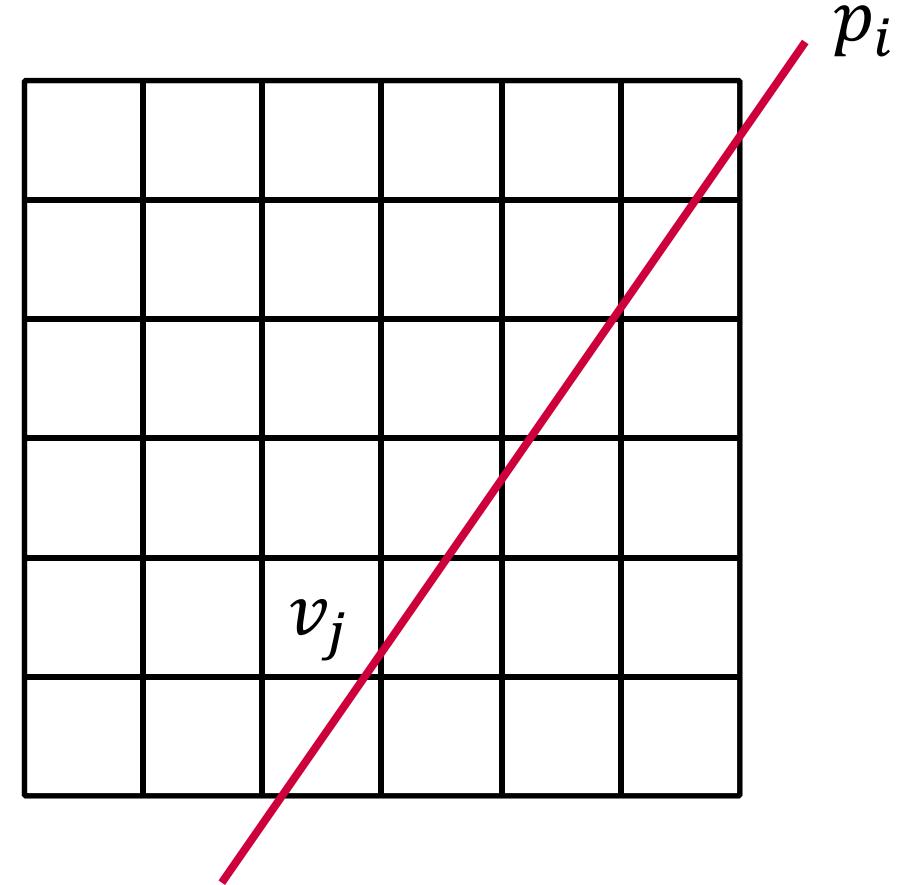
$$Wv = p$$

$v \in \mathbb{R}^n$ Volume Vector

$p \in \mathbb{R}^m$ Projection Data Vector

$W \in \mathbb{R}^{m \times n}$ Projection Matrix

w_{ij} represents how much v_j contributes to p_i

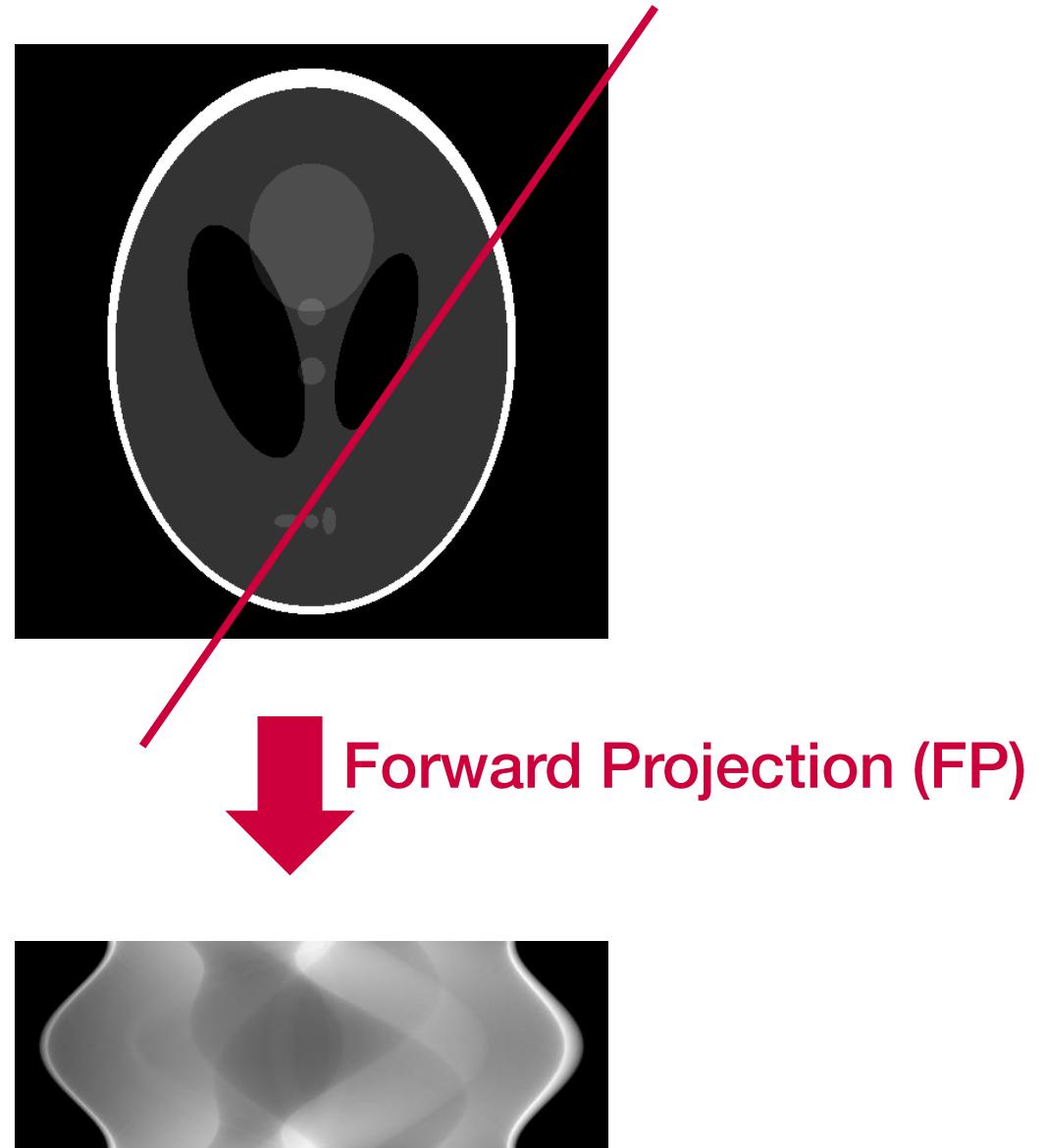


Forward Projection

$$q_i := \mathbf{W}\mathbf{v} \quad \mathbf{W}$$

The diagram illustrates the forward projection process. On the left, a vertical vector \mathbf{v} is shown as a red bar. In the center, a matrix \mathbf{W} is shown as a dark blue square. To the right, a query q_i is shown as a dark blue bar. The equation $q_i := \mathbf{W}\mathbf{v}$ indicates that the query q_i is obtained by multiplying the matrix \mathbf{W} with the vector \mathbf{v} .

$$q_i := \sum_{j=0}^{n-1} w_{ij} v_j$$

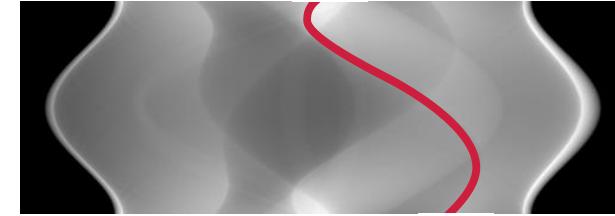


Backprojection

$$u' := \begin{matrix} & \\ & \\ & W^T \\ & \\ & \end{matrix}$$

$$u_j := \sum_{i=0}^{m-1} w_{ij} p_i$$

$$p$$

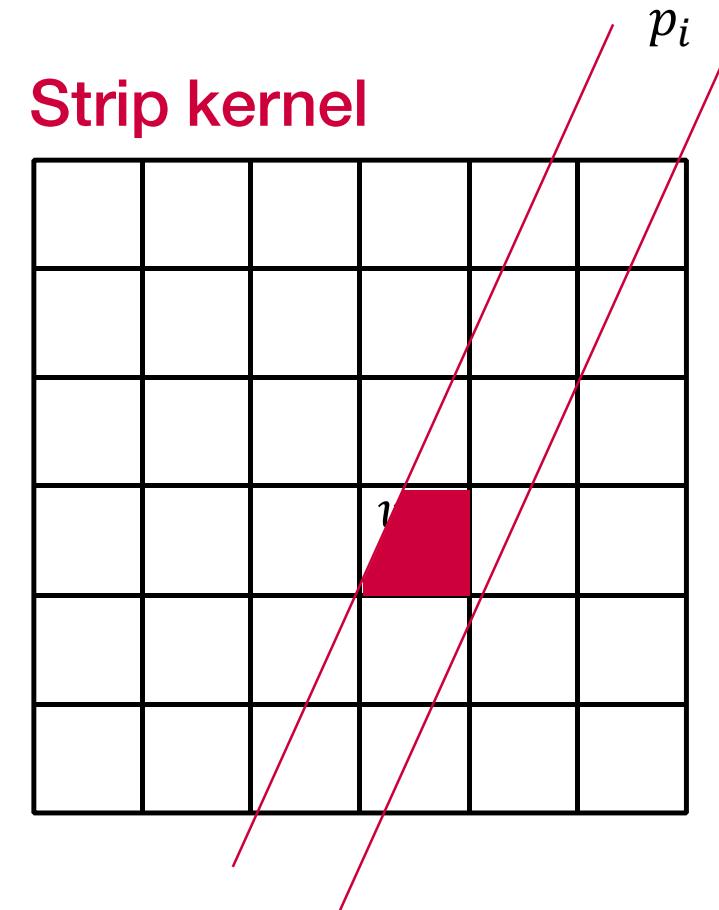
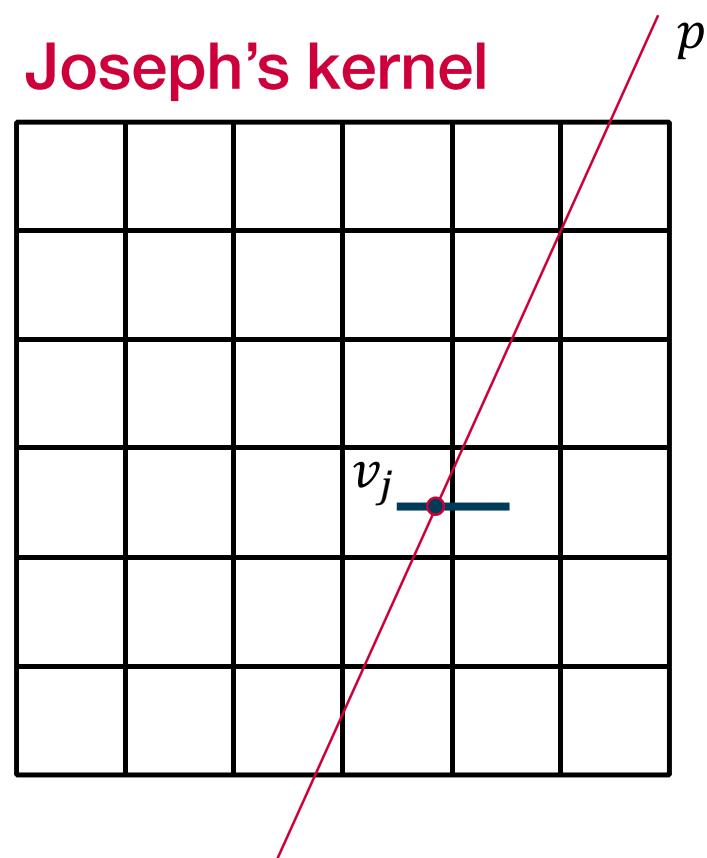
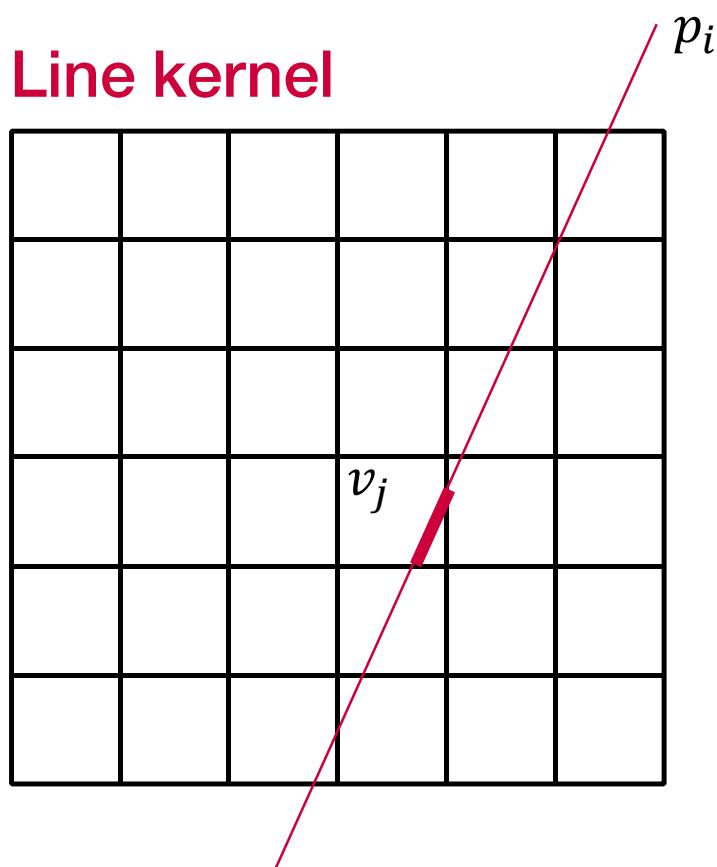


Backprojection (BP)



Projection Matrix

$$\sum_{j=1}^n w_{ij} v_j = p_i$$



Algebraic Reconstruction

Solve $Wv = p$ 



$v^* = \operatorname{argmin}_v \|p - Wv\|$ some norm

/ Projection distance

Solve iteratively: Landweber (ART,SART,SIRT), Krylov Subspace (CGLS), ...

Simultaneous Iterative Reconstruction Technique (SIRT)

Iteration step:

$$\boldsymbol{v}^{(k+1)} = \boldsymbol{v}^{(k)} + \mathbf{C} \mathbf{W}^T \mathbf{R} (\mathbf{p} - \mathbf{W} \boldsymbol{v}^{(k)})$$

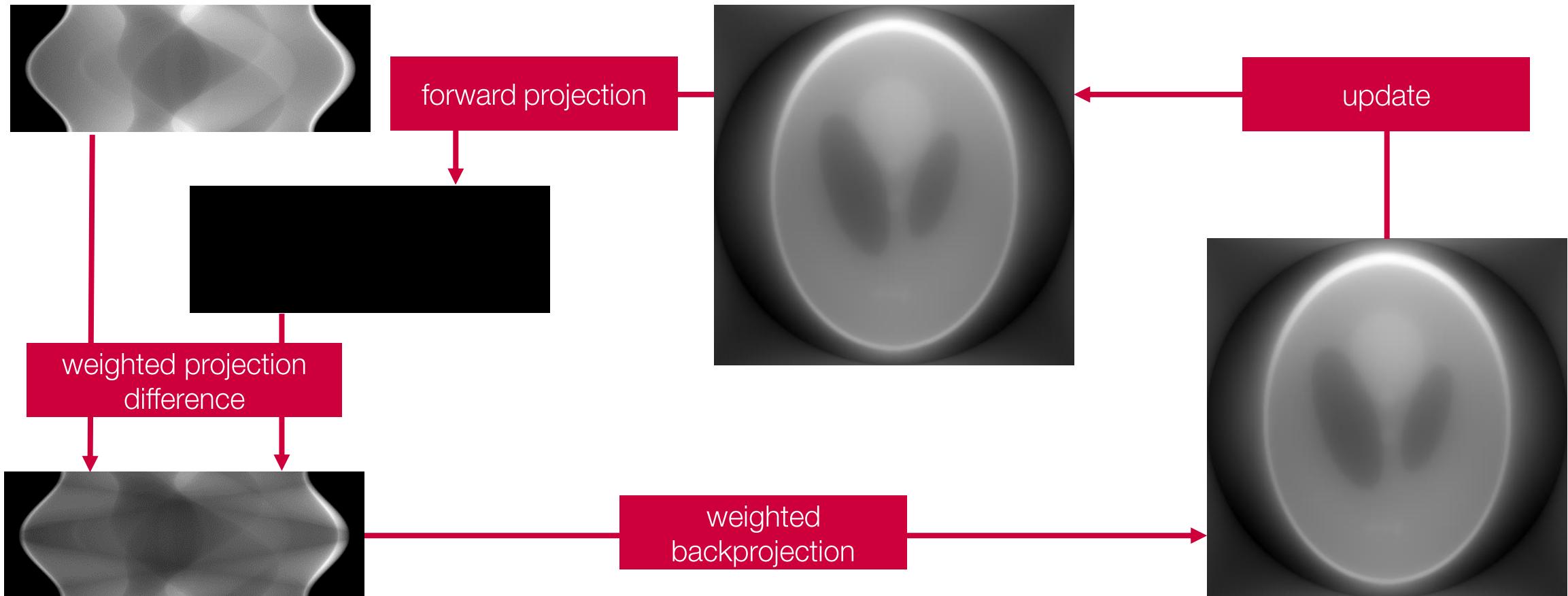
$$\mathbf{R} \in \mathbb{R}^{m \times m} \quad r_{ii} = 1 / \sum_{j=0}^{n-1} w_{ij}$$
$$\mathbf{C} \in \mathbb{R}^{n \times n} \quad c_{jj} = 1 / \sum_{i=0}^{m-1} w_{ij}$$

1. Weighted projection difference
2. Weighted backprojection
3. Update

Solves $\boldsymbol{v}^* = \operatorname{argmin}_{\boldsymbol{v}} \|\mathbf{p} - \mathbf{W} \boldsymbol{v}\|_{\mathbf{R}}$ with $\|\boldsymbol{x}\|_{\mathbf{R}} = \boldsymbol{x}^T \mathbf{R} \boldsymbol{x}$

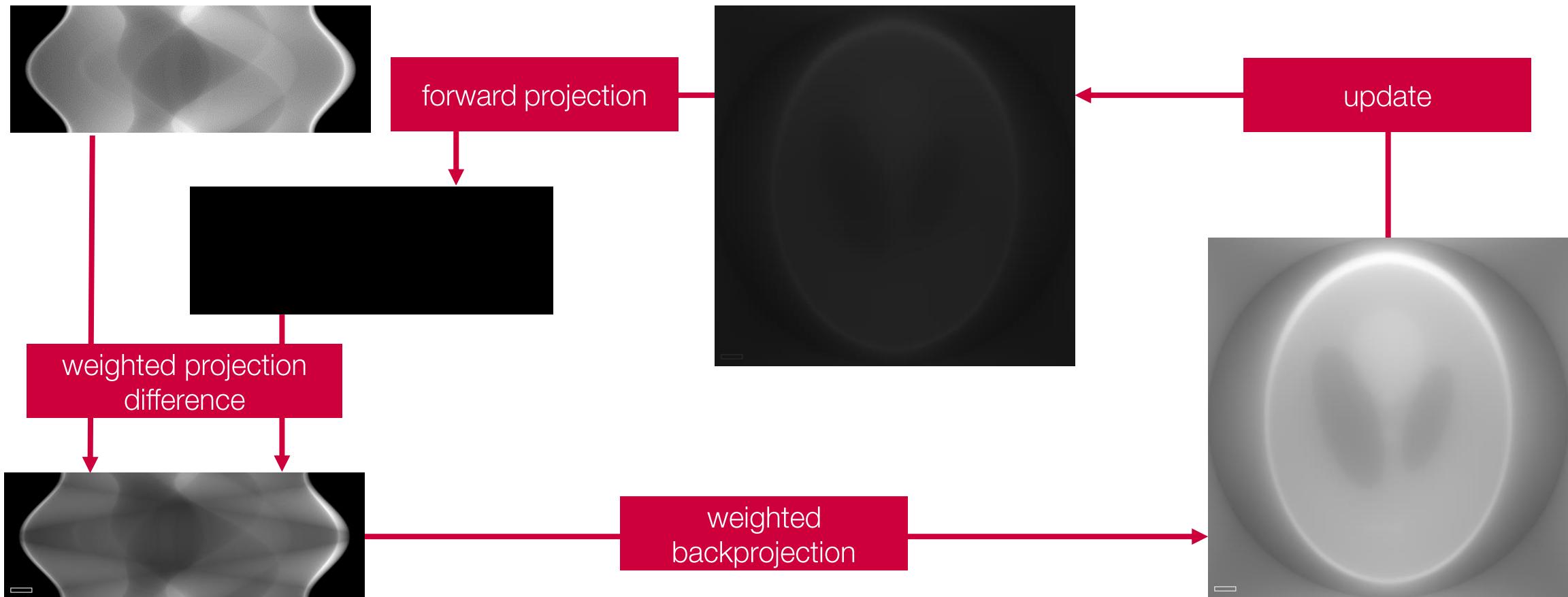
Algebraic Reconstruction

$$\boldsymbol{v}^{(1)} = \boldsymbol{v}^{(0)} + \mathbf{C}\mathbf{W}^T\mathbf{R}(\mathbf{p} - \mathbf{W}\boldsymbol{v}^{(0)})$$



Algebraic Reconstruction

$$\boldsymbol{v}^{(k+1)} = \boldsymbol{v}^{(k)} + \mathbf{C}\mathbf{W}^T\mathbf{R}(\mathbf{p} - \mathbf{W}\boldsymbol{v}^{(k)})$$



Prior knowledge

Sufficient projection data

$$W \quad v = p$$

Insufficient projection data

$$W \quad v = p$$

Insufficient projection data
+ prior knowledge

$$W \quad v = p$$

prior
knowledge

Observation Negative values in the reconstruction make no sense.

Minimum Constraint: Project all negative values to zero.

$$\mathbf{v}^{(k+1)} = \textcolor{red}{\mathbf{v}^{(k)}} + \mathbf{C}^T \mathbf{W}^T \mathbf{R} (\mathbf{p} \mathbf{W} \mathbf{W}^T \mathbf{v}^{(k)})$$

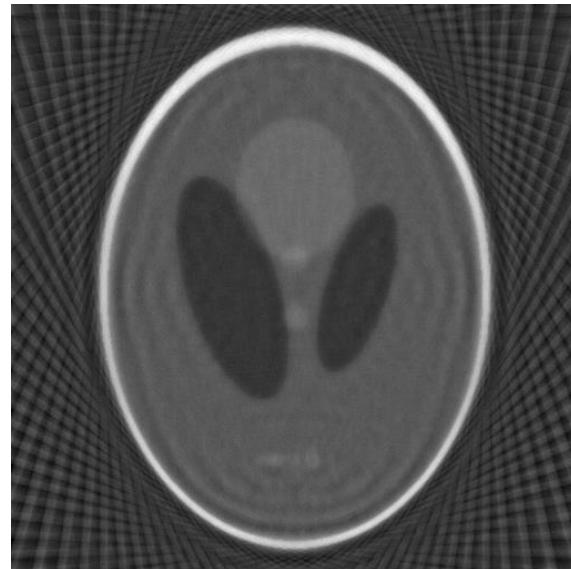
$$\phi(x_i) = \begin{cases} x_i & 0 \leq x_i \\ 0 & \text{otherwise} \end{cases}$$

Breaks mathematical converge theory of SIRT, but improves quality in practice.

Positivity constraint

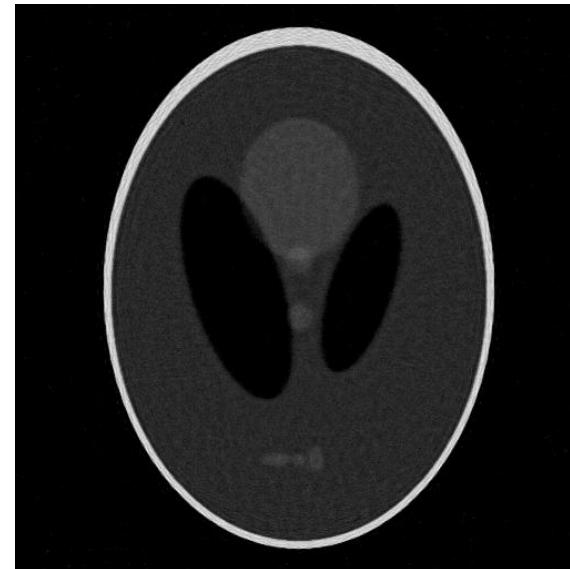
45 projection directions

$$\begin{matrix} \textcolor{red}{\square} & \textcolor{darkblue}{\square} & = & \textcolor{red}{\square} \end{matrix}$$



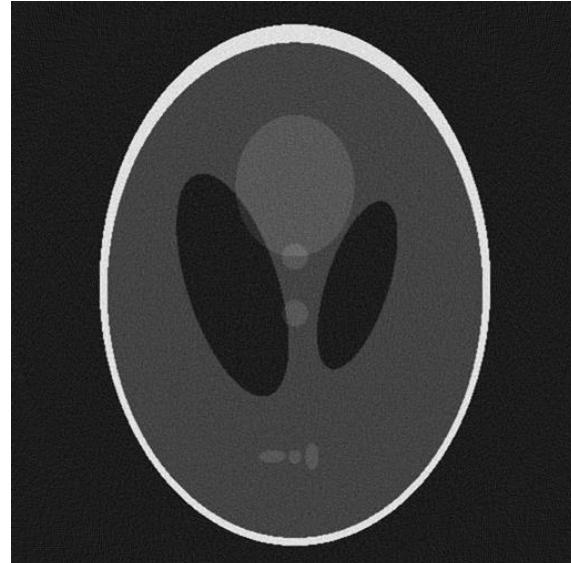
45 projection directions

$$\begin{matrix} \textcolor{red}{\square} & \textcolor{darkblue}{\square} & = & \textcolor{darkgray}{\square} \end{matrix}$$



360 projection directions

$$\begin{matrix} \textcolor{green}{\square} & \textcolor{darkblue}{\square} & = & \textcolor{green}{\square} \end{matrix}$$



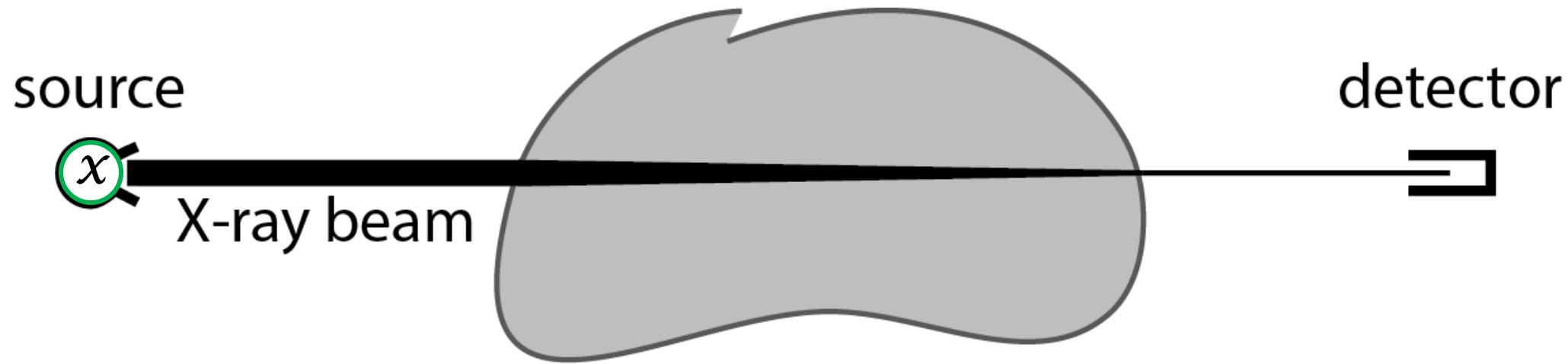
360 projection directions

$$\begin{matrix} \textcolor{green}{\square} & \textcolor{darkblue}{\square} & = & \textcolor{darkgray}{\square} \end{matrix}$$



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- A Brief Introduction to Reconstruction Mathematics
- **When Physics and Mathematics collide**
 - **Polychromatic source and Beam Hardening**
 - **Scattering**

X-ray detection



Beam intensity at source

$$I_0(E)$$

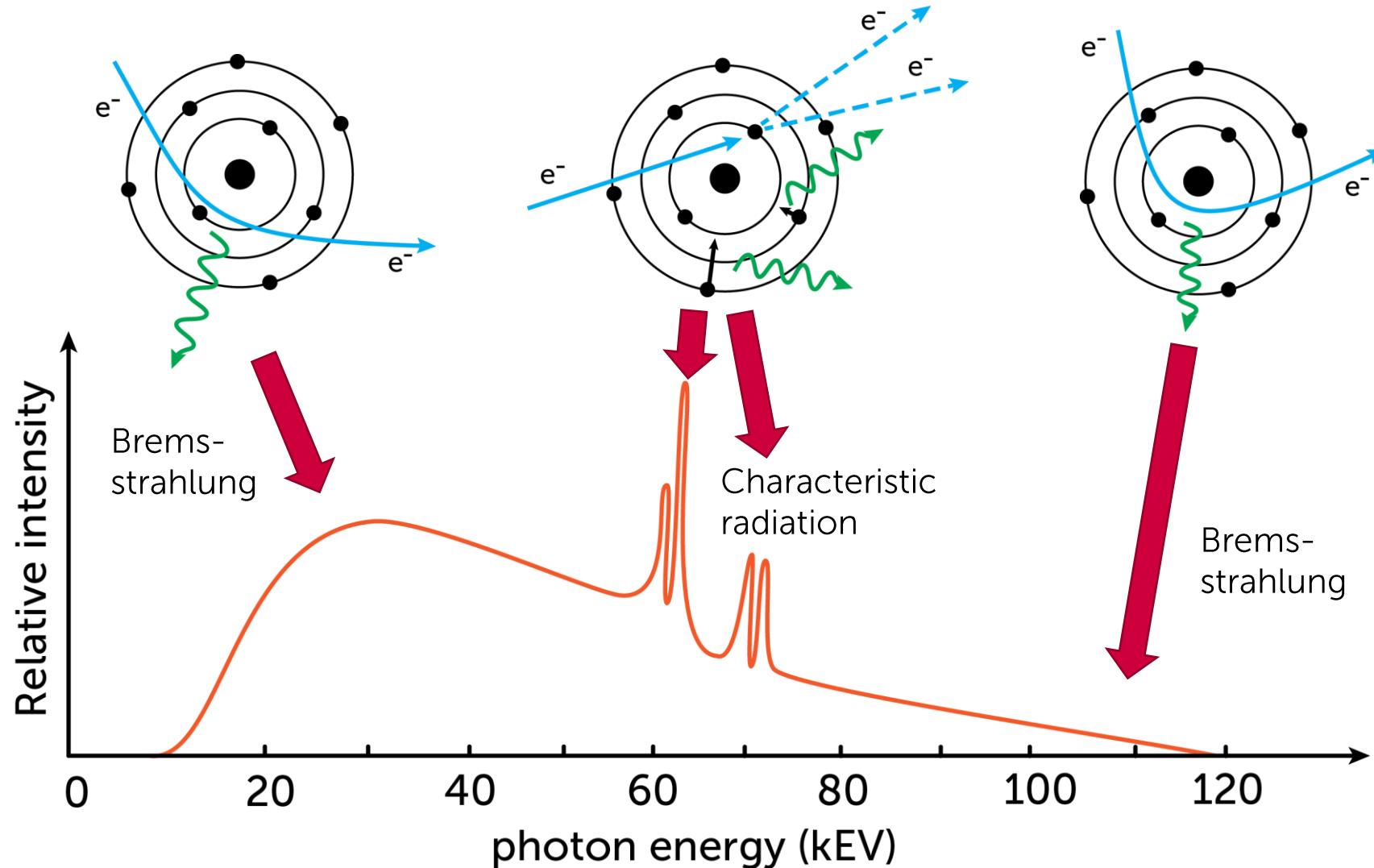
Beam intensity at detector

$$I(E) = I_0(E) e^{-\int \mu(\xi, E) d\xi}$$

Beam intensity changes

$$I(\xi + \Delta\xi, E) = I(\xi) - \mu(\xi, E)I(\xi, E)\Delta\xi$$

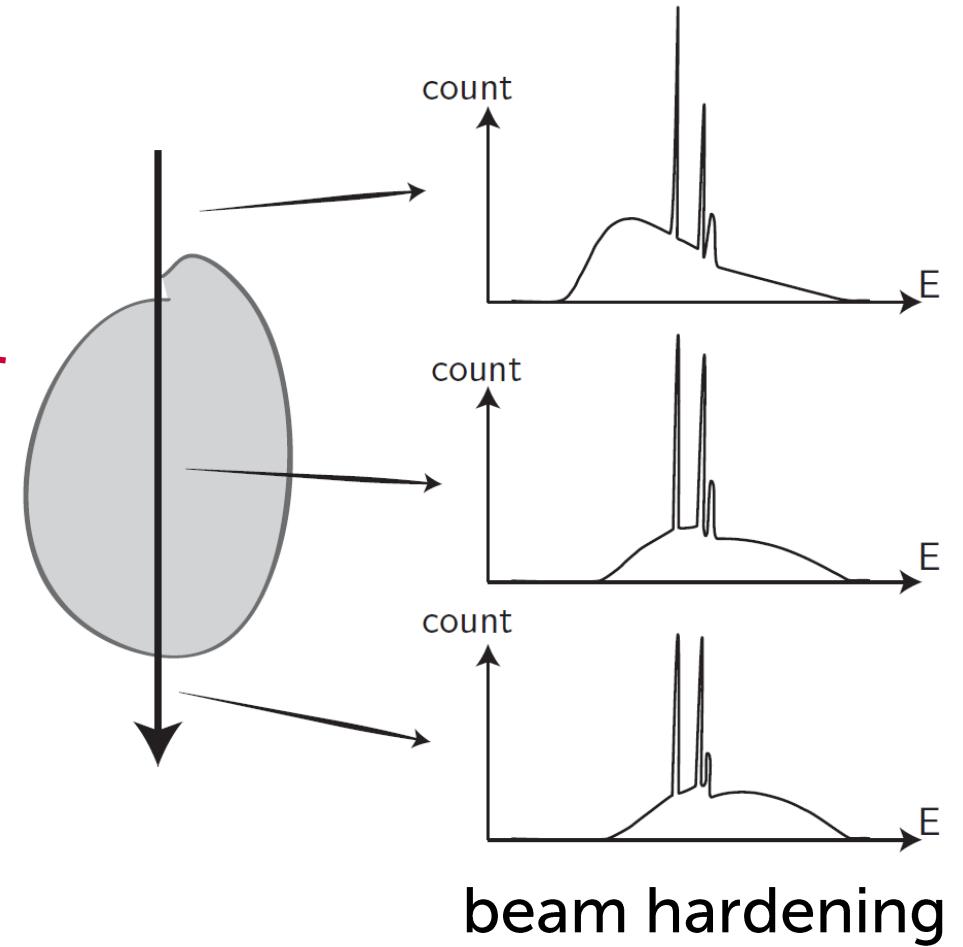
X-ray spectrum



Polychromatic spectrum

$$I_m(u, v) = I_0 \int_E^\infty e^{-\int_{L,E} \mu(x, f_L z) dL} e^{(x, f_L z)} dE$$

~~$$p_m(u, v) = -\ln q \left(\frac{I_m(u, v)}{I_0} \right)$$~~

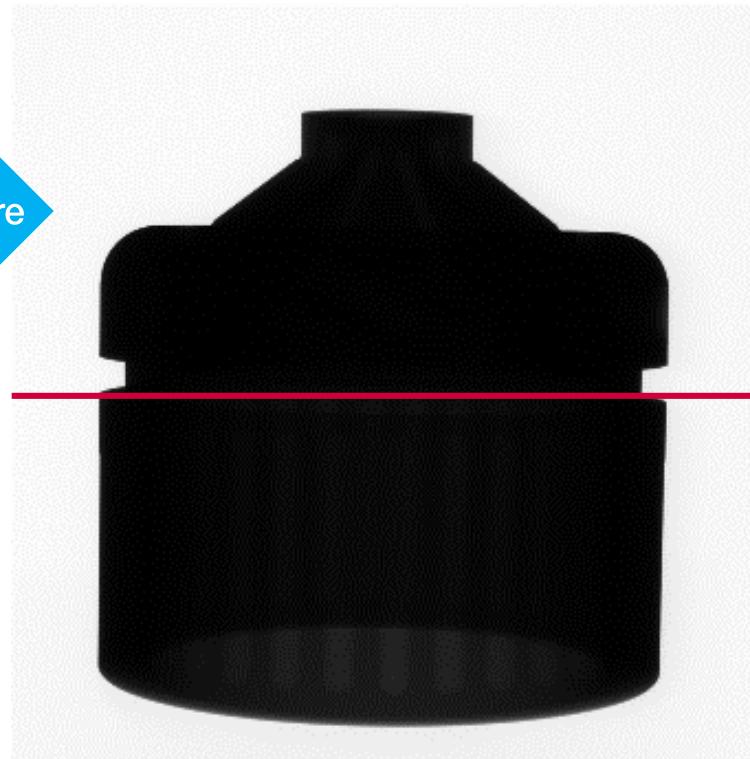


Beam Hardening

injection mould, 3D printed

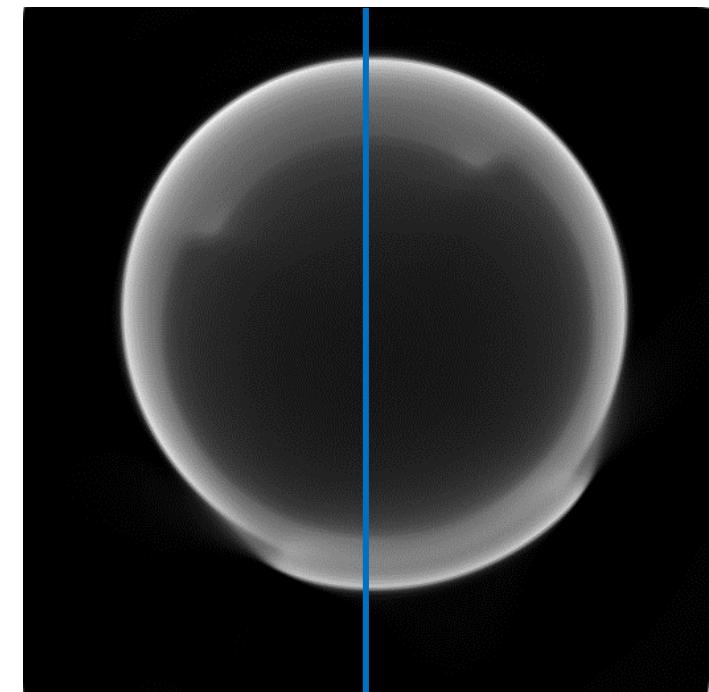


compare

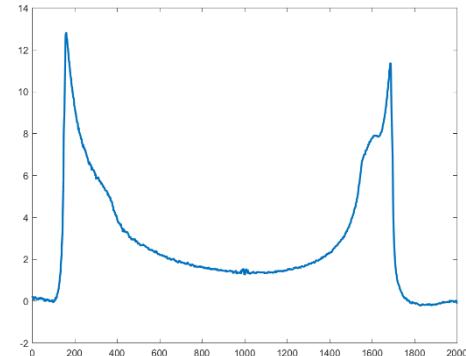


3D CAD model

projection



reconstruction



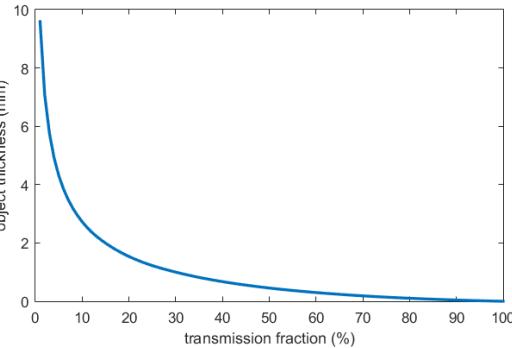
Beam hardening correction strategy



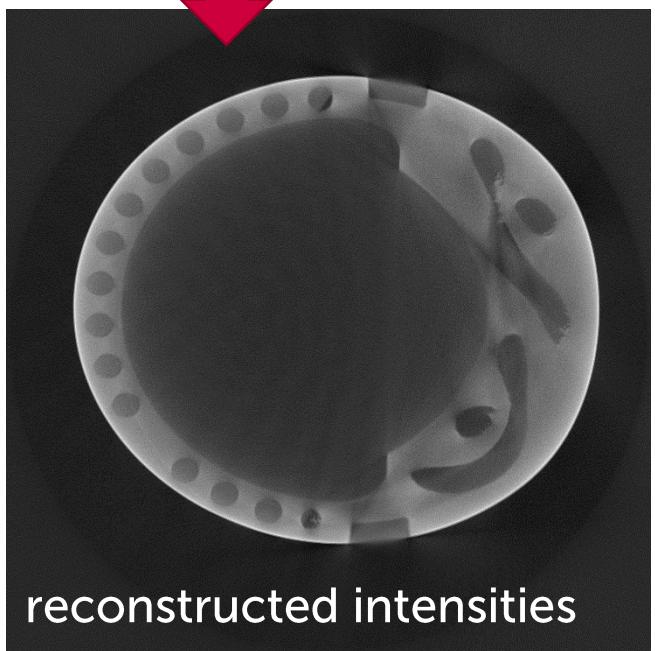
Transmission fraction (%)



Object thickness (mm)

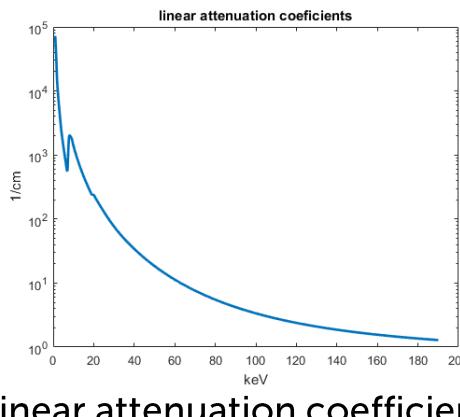
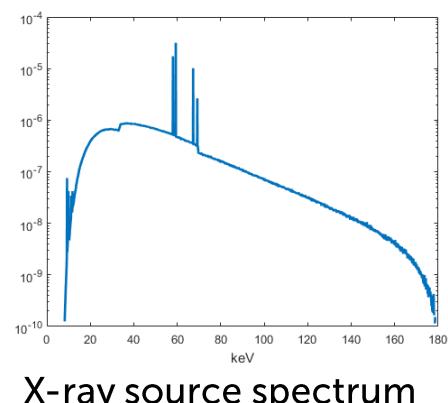


log transform + FBP



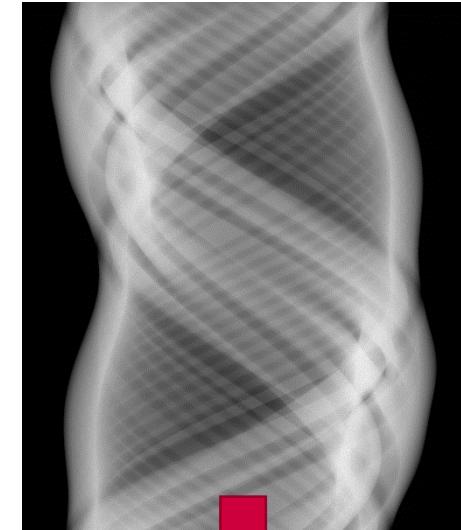
transmission vs thickness

Monte Carlo simulation

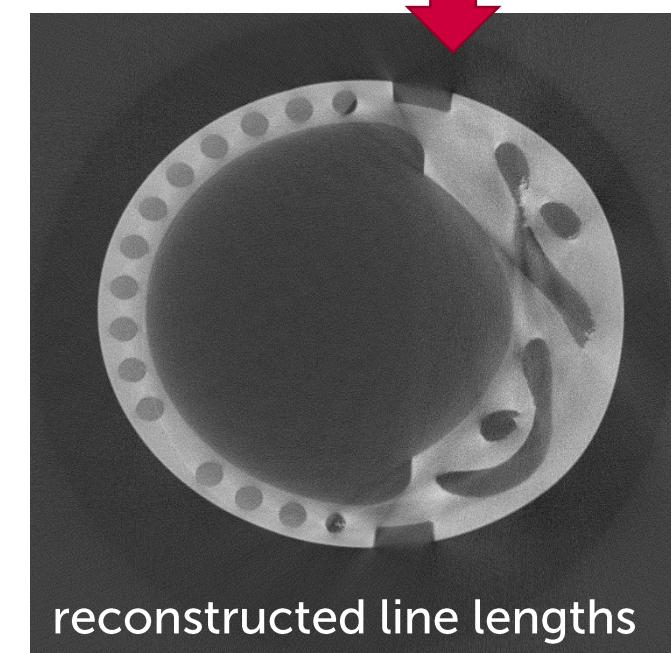


X-ray source spectrum

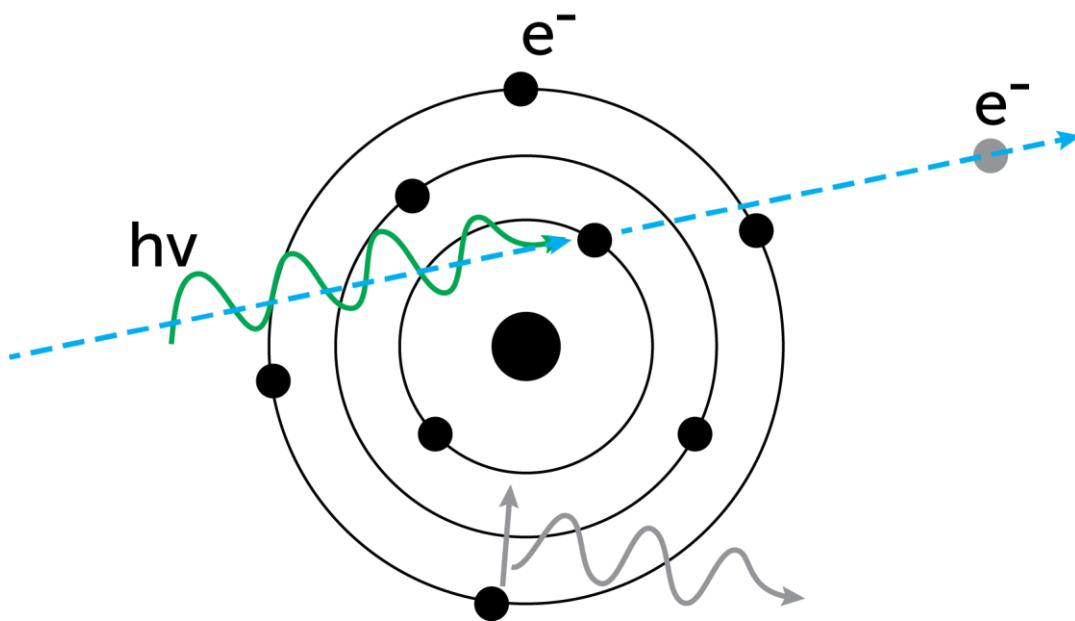
linear attenuation coefficients



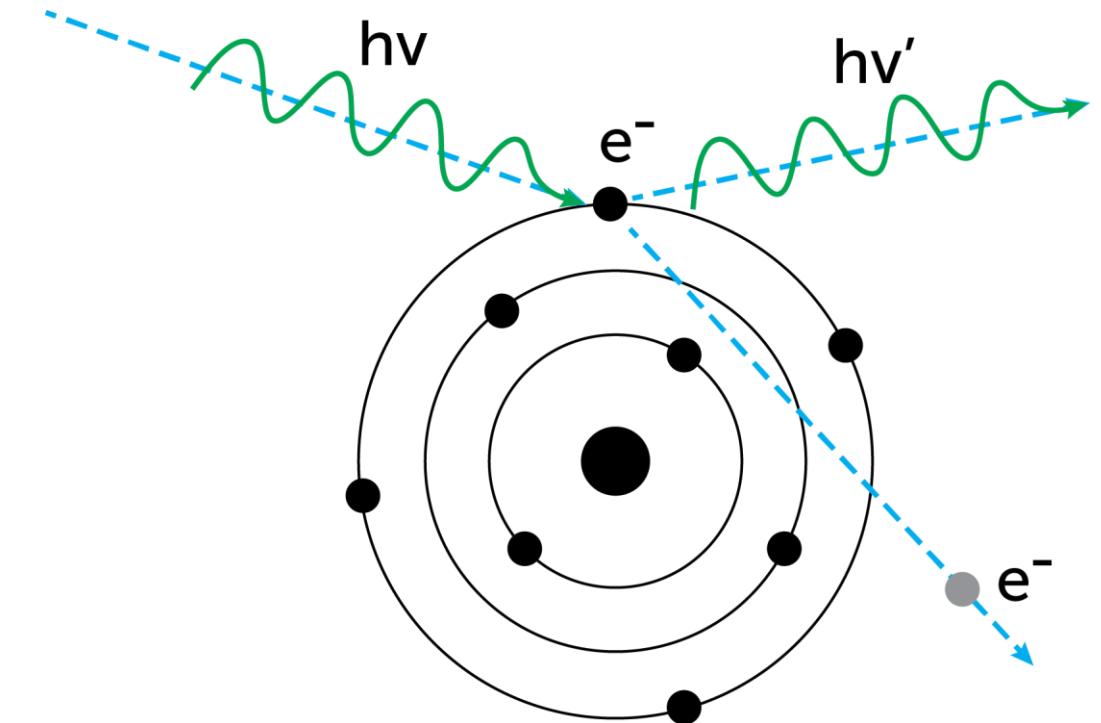
FBP



X-ray Matter Interaction



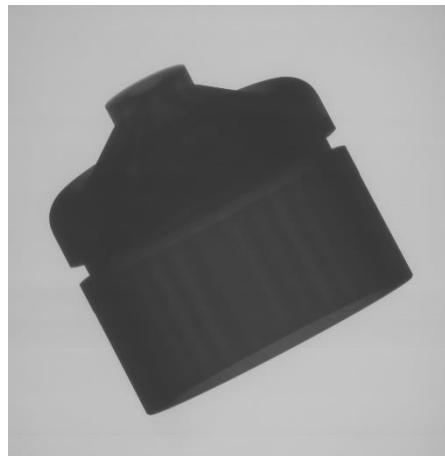
photoelectric effect



Compton scattering

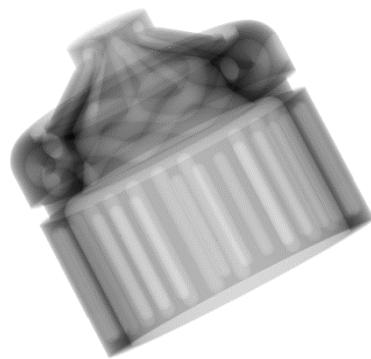
Scattering

$$I_{S,D}(u, v) = I_0 e^{- \int_L \mu(x,y,z) dL} + I_0 e^{- \iint_L s(x,y,z) \mu(x,y,z) dL} dxz$$



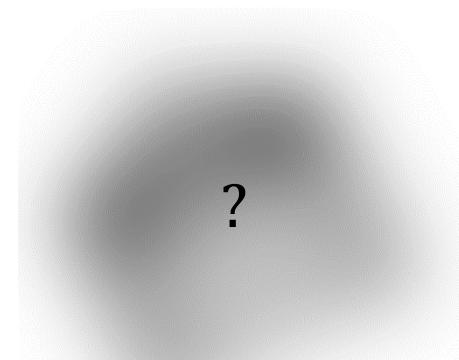
measured data

=



“clean” projection

+



scattering
contribution

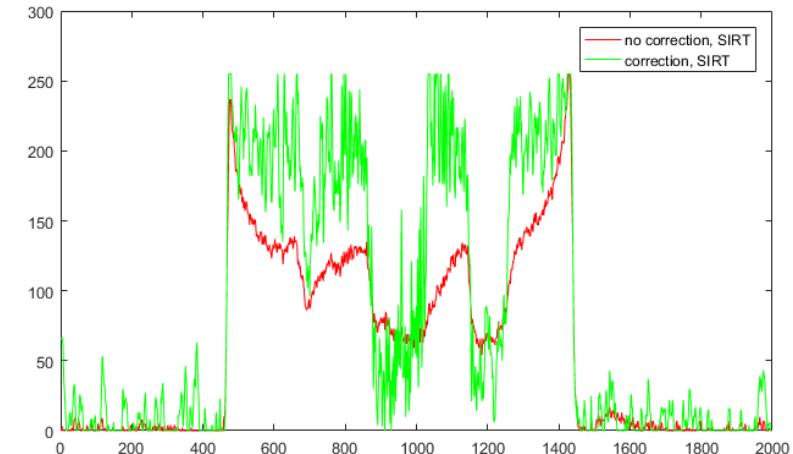
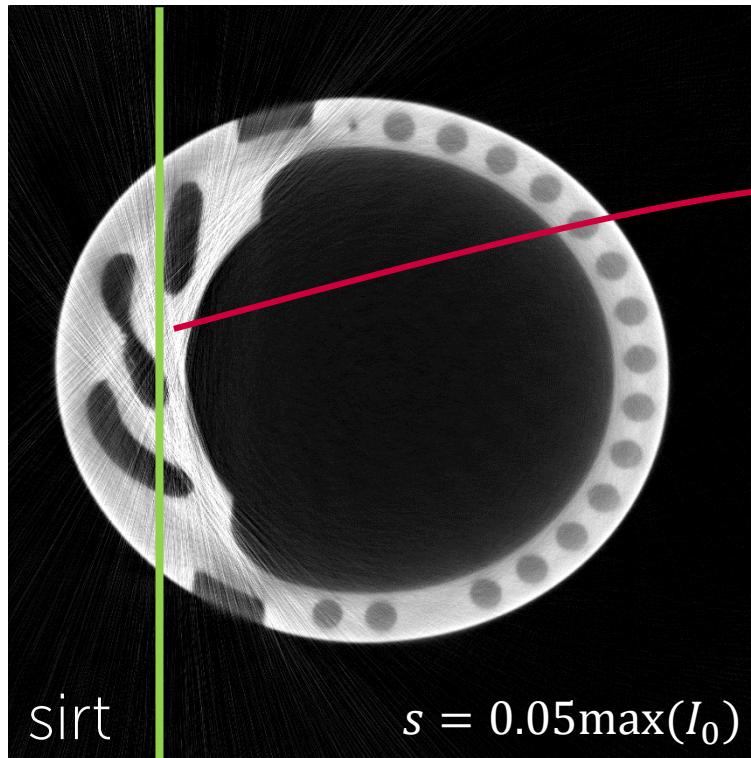
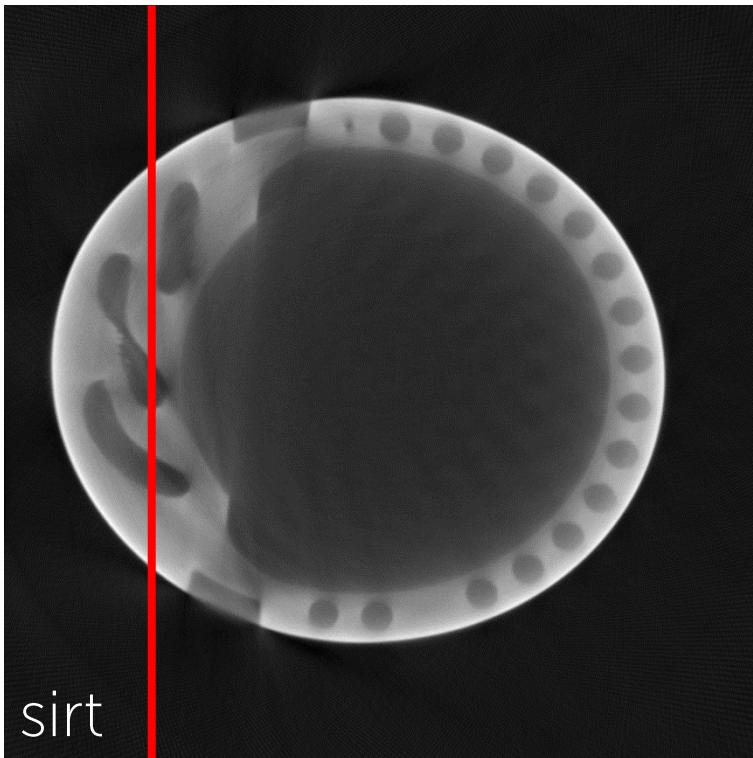
?

Scatter correction strategy

Simple approach: subtract constant value of all projections

$$-\log \left(\frac{I - D}{F - D} \right)$$

$$-\log \left(\frac{I - s - D}{F - D} \right)$$



noise!

Thanks!

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