# **Introduction to Medical Imaging**

X-Ray CT -- Introduction

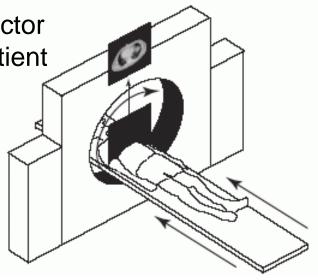
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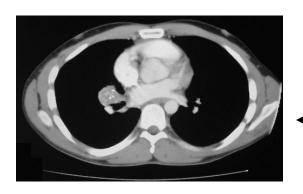
#### **Overview**

Scanning: rotate source-detector pair around the patient

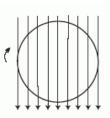




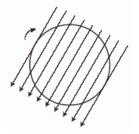
data



reconstructed crosssectional slice

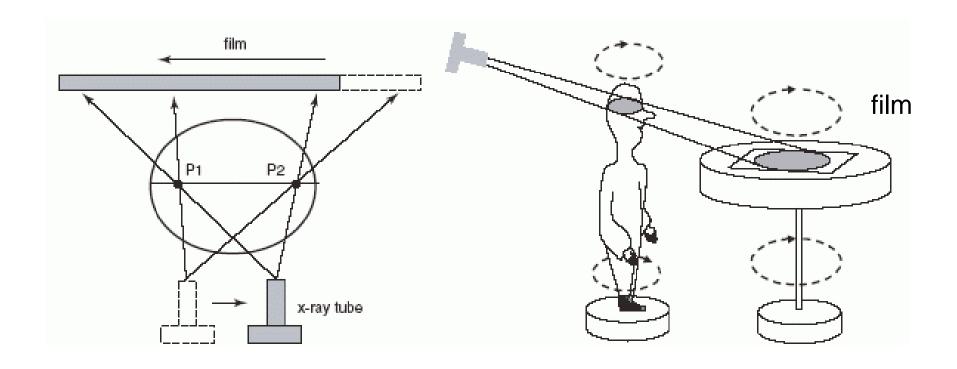


reconstruction routine



sinogram: a line for every angle

## **Early Beginnings**



Linear tomography
only line P1-P2 stays in focus
all others appear blurred

Axial tomography
in principle, simulates the
backprojection procedure used in
current times

### **Current Technology**

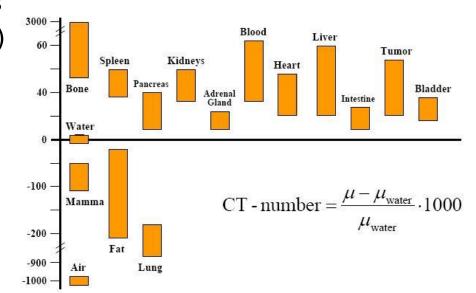
#### Principles derived by Godfrey Hounsfield for EMI

- based on mathematics by A. Cormack
- both received the Nobel Price in medizine/physiology in 1979
- technology is advanced to this day

#### Images:

- size generally 512 x 512 pixels
- values in Hounsfield units (HU) in the range of –1000 to 1000





μ: linear attenuation coefficient

due to large dynamic range, windowing must be used to view an image

#### **CT Detectors**

#### Scintillation crystal with photomultiplier tube (PMT)

(scintillator: material that converts ionizing radiation into pulses of light)

- high QE and response time
- low packing density
- PMT used only in the early CT scanners

#### Gas ionization chambers

- replace PMT
- X-rays cause ionization of gas molecules in chamber
- ionization results in free electrons/ions
- these drift to anode/cathode and yield a measurable electric signal
- lower QE and response time than PMT systems, but higher packing density

## Scintillation crystals with photodiode

- current technology (based on solid state or semiconductors)
- photodiodes convert scintillations into measurable electric current
- QE > 98% and very fast response time

### **Projection Coordinate System**

The parallel-beam geometry at angle  $\theta$  represents a new coordinate system (r,s) in which projection  $I_{\theta}(r)$  is acquired

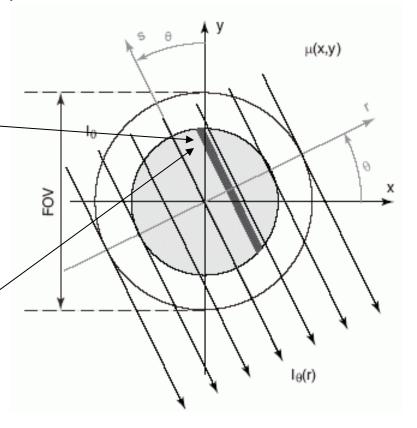
• rotation matrix *R* transforms coordinate system (*x*, *y*) to (*r*, *s*):

that is, all (x,y) points that fulfill  $r = x \cos(\theta) + y \sin(\theta)$  are on the (ray) line  $L_{(r,\theta)}$ 

R<sup>T</sup> is the inverse, mapping (r, s) to (x, y)

$$\begin{pmatrix} x \\ y \end{pmatrix} = R^T \begin{pmatrix} r \\ s \end{pmatrix} = \begin{pmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{pmatrix} \begin{pmatrix} r \\ s \end{pmatrix}$$

s is the parametric variable along the (ray) line  $L_{(r,\theta)}$ 



## **Projection**

Assuming a fixed angle  $\theta$ , the measured intensity at detector position r is the integrated density along  $L_{(r,\theta)}$ :

$$I_{\theta}(r) = I_{0} \cdot e^{-\int_{L_{r,\theta}} \mu(x,y)ds}$$

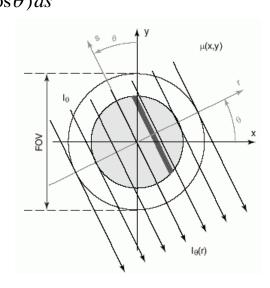
$$= I_{0} \cdot e^{\int_{L_{r,\theta}} \mu(r \cdot \cos \theta - s \cdot \sin \theta, r \cdot \sin \theta + s \cdot \cos \theta)ds}$$

$$= I_{0} \cdot e^{\int_{L_{r,\theta}} \mu(r \cdot \cos \theta - s \cdot \sin \theta, r \cdot \sin \theta + s \cdot \cos \theta)ds}$$

For a continuous energy spectrum:

$$I_{\theta}(r) = \int_{0}^{\infty} I_{0}(E) \cdot e^{L_{r,\theta}} e^{L_{r,\theta}}$$

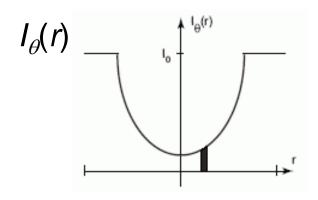
But in practice, is is assumed that the X-rays are monochromatic

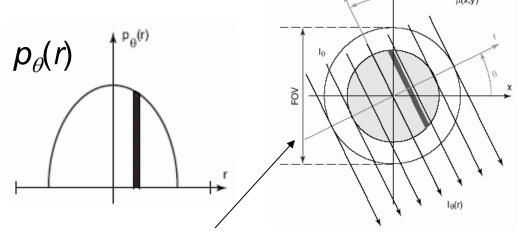


### **Projection Profile**

Each intensity profile  $I_{\theta}(r)$  is transformed to into an attenuation profile  $p_{\theta}(r)$ :

$$p_{\theta}(r) = -\ln \frac{I_{\theta}(r)}{I_{0}} = \int_{L_{r,\theta}} \mu(r \cdot \cos \theta - s \cdot \sin \theta, r \cdot \sin \theta + s \cdot \cos \theta) ds$$





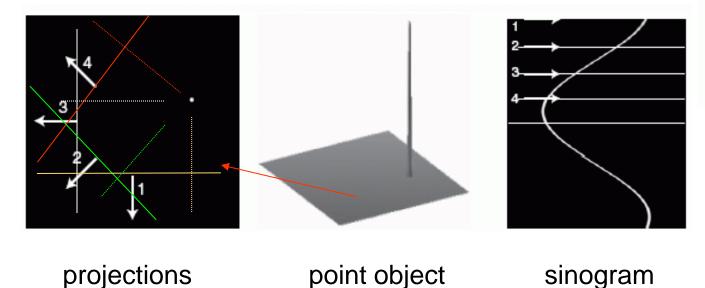
- $p_{\theta}(r)$  is zero for |r| > FOV/2 (FOV = Field of View, detector width)
- $p_{\theta}(r)$  can be measured from  $(0, 2\pi)$
- however, for parallel beam views  $(\pi, 2\pi)$  are redundant, so just need to measure from  $(0, \pi)$

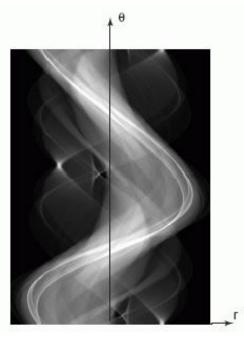
### **Sinogram**

Stacking all projections (line integrals) yields the *sinogram*, a 2D dataset  $p(r, \theta)$ 

To illustrate, imagine an object that is a single point:

• it then describes a sinusoid in  $p(r, \theta)$ :





 note, the sinogram is very similar to the Hough transform where a single point also gives rise to a sinusoid

#### **Radon Transform**

The transformation of any function f(x,y) into  $p(r,\theta)$  is called the 2D Radon Transform

$$p(r,\theta) = R\{f(x,y)\}\$$

$$= \int_{-\infty}^{\infty} f(r \cdot \cos \theta - s \cdot \sin \theta, r \cdot \sin \theta + s \cdot \cos \theta) ds$$

The Radon transform has the following properties:

•  $p(r,\theta)$  is periodic in  $\theta$  with period  $2\pi$ 

$$p(r,\theta) = p(r,\theta + 2\pi)$$

•  $p(r,\theta)$  is symmetric in  $\theta$  with period  $\pi$ 

$$p(r,\theta) = p(-r,\theta \pm \pi)$$

## Sampling (1)

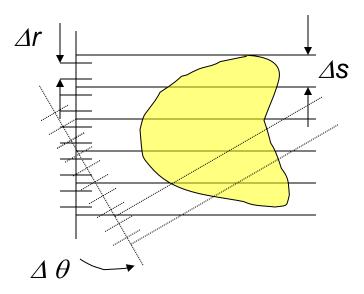
#### In practice, we only have a limited

- number of views, M
- number of detector samples, N
- for example, *M*=1056, *N*=768

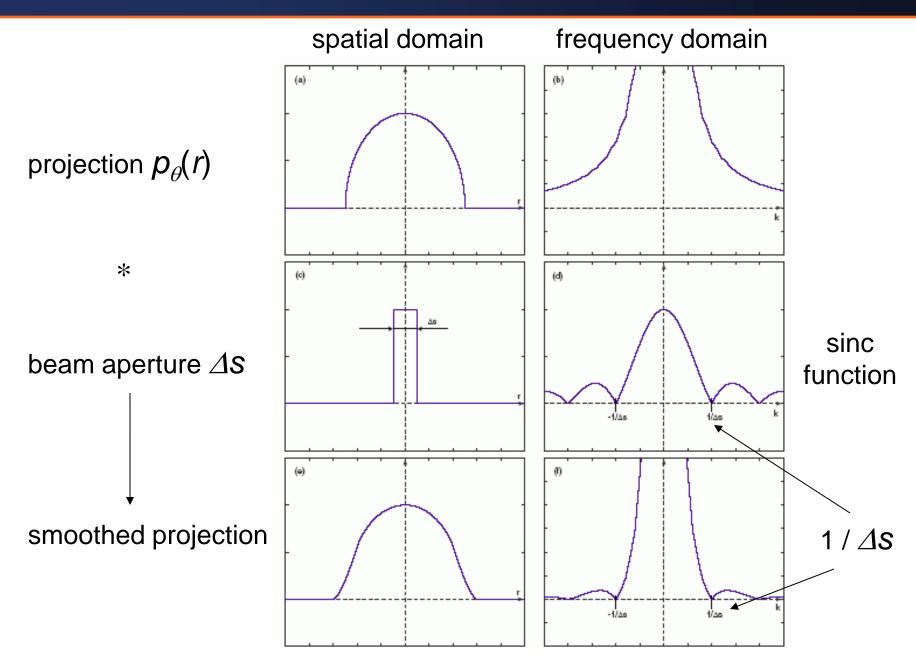
## This gives rise to a discrete sinogram $p(n\Delta r, m\Delta\theta)$

- a matrix with M rows and N columns
   Δr is the detector sampling distance
   Δθ is the rotation interval between subsquent views
- assume also a beam of width ⊿s

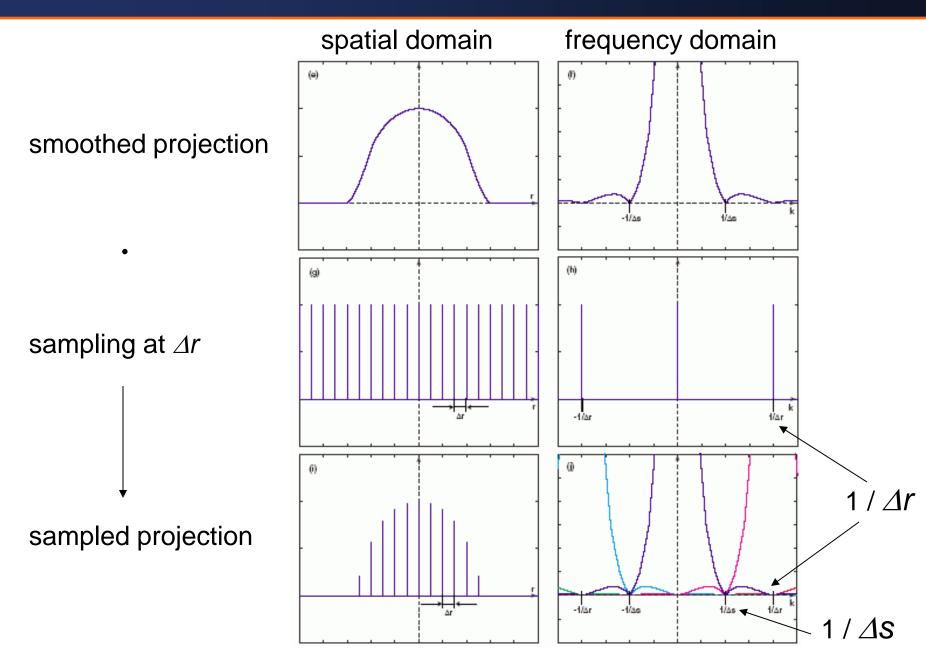
Sampling theory will tell us how to choose these parameters for a given desired object resolution



# Sampling (2)



## Sampling (3)



## **Limiting Aliasing**

### Aliasing within the sinogram lines (projection aliasing):

• to limit aliasing, we must separate the aliases in the frequency domain (at least coinciding the zero-crossings):

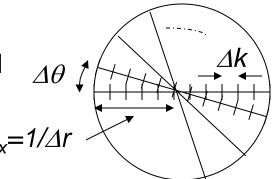
$$\frac{1}{\Delta r} \ge \frac{2}{\Delta s} \quad \to \quad \Delta r \le \frac{\Delta s}{2}$$

• thus, at least 2 samples per beam are required

Aliasing across the sinogram lines (angular aliasing):

$$\Delta\theta = \frac{\pi k_{\rm max}}{M}$$
 distributed around the semi-circle

$$\Delta k = \frac{k_{\text{max}}}{N/2}$$
 N: number of detector samples, give rise to N frequency domain samples for each projection



sinogram in the frequency domain
(2 projections with *N*=12 samples each are shown)

for uniform sampling: 
$$\Delta \theta = \Delta k \rightarrow \frac{\pi k_{\text{max}}}{M} = \frac{k_{\text{max}}}{N/2} \rightarrow M = \pi \frac{N}{2}$$

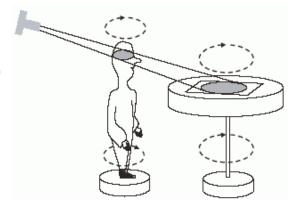
## **Reconstruction: Concept**

Given the sinogram  $p(r,\theta)$  we want to recover the object described in (x,y) coordinates

Recall the early axial tomography method

- basically it worked by subsequently "smearing" the acquired  $p(r, \theta)$  across a film plate
- for a simple point we would get:

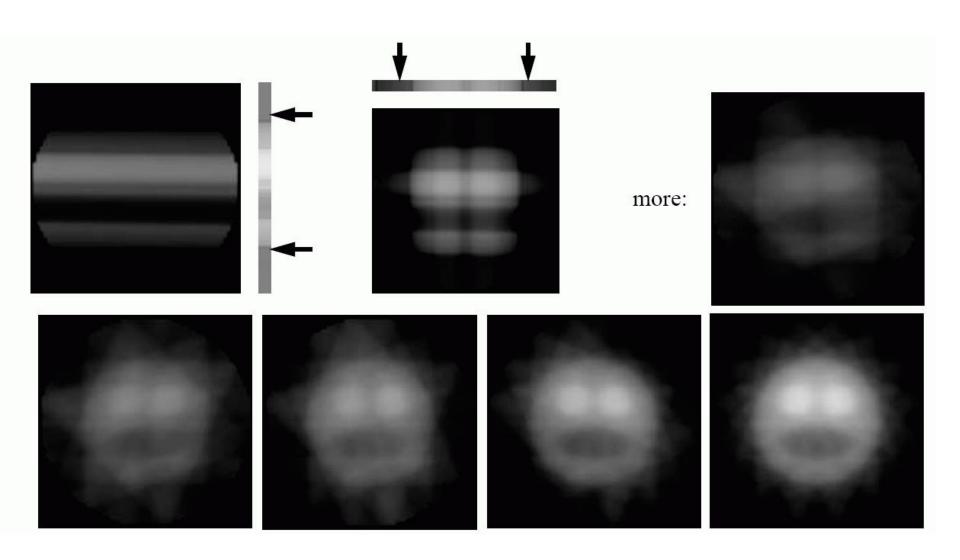




This is called *backprojection*:

$$b(x, y) = B\{p(r, \theta)\} = \int_{0}^{\pi} p(x \cdot \cos \theta + y \cdot \sin \theta, \theta) d\theta$$

# **Backprojection: Illustration**



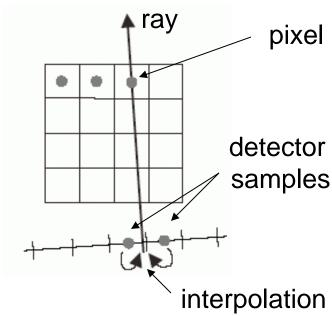
### **Backprojection: Practical Considerations**

#### A few issues remain for practical use of this theory:

we only have a finite set of M projections and a discrete array of N pixels (x<sub>i</sub>, y<sub>j</sub>)

$$b(x_i, y_j) = B\{p(r_n, \theta_m)\} = \sum_{m=1}^{M} p(x_i \cdot \cos \theta_m + y_j \cdot \sin \theta_m, \theta_m)$$

- to reconstruct a pixel  $(x_i, y_j)$  there may not be a ray  $p(r_n, \theta_n)$  (detector sample) in the projection set
  - → this requires interpolation (usually linear interpolation is used)



 the reconstructions obtained with the simple backprojection appear blurred (see previous slides)

#### The Fourier Slice Theorem

To understand the blurring we need more theory → the Fourier

Slice Theorem or Central Slice Theorem

• it states that the Fourier transform  $P(\theta, k)$  of a projection  $p(r, \theta)$  is a line across the origin of the Fourier transform  $F(k_x, k_y)$  of function f(x, y)

polar grid

 $P(\theta,k)$ 



- calculate the 1D FT of all projections  $p(r_m, \theta_m)$ , which gives rise to  $F(k_x, k_v)$  sampled on a polar grid (see figure)
- resample the polar grid into a cartesian grid (using interpolation)
- perform inverse 2D FT to obtain the desired f(x,y) on a cartesian grid

#### However, there are two important observations:

- interpolation in the frequency domain leads to artifacts
- at the FT periphery the spectrum is only sparsely sampled

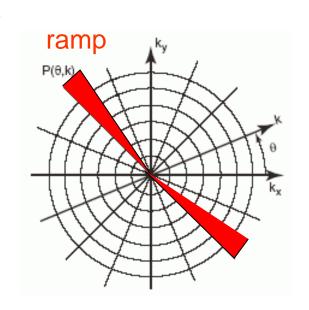
## Filtered Backprojection: Concept

# To account for the implications of these two observations, we modify the reconstruction procedure as follows:

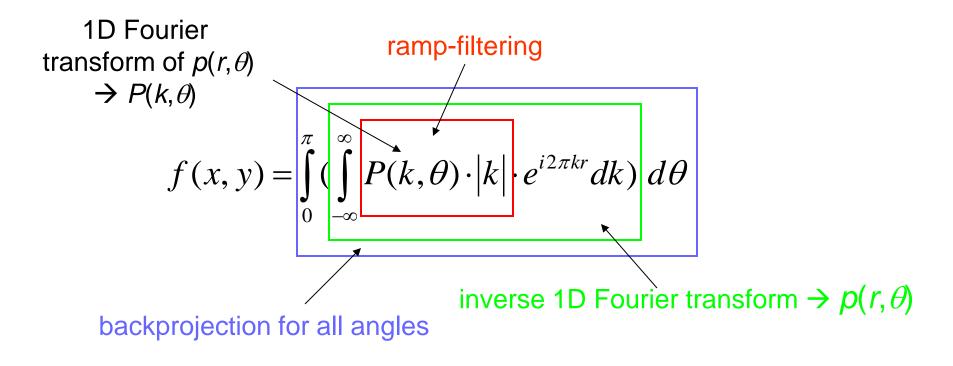
- filter the projections to compensate for the blurring
- perform the interpolation in the spatial domain via backprojection
  - → hence the name Filtered Backprojection

#### Filtering – for now a more practical explanation:

- we need a way to equalize the contributions of all frequencies in the FT's polar grid
- this can be done by multiplying each P(θ,k) by a ramp function → this way the magnitudes of the existing higher-frequency samples in each projection are scaled up to compensate for their lower amount
- the ramp is the appropriate scaling function since the sample density decreases linearly towards the FT's periphery

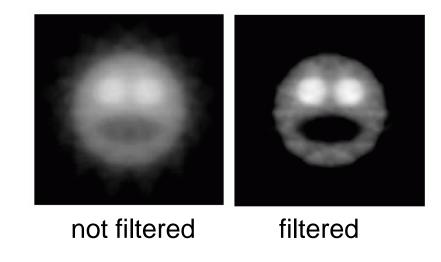


### Filtered Backprojection: Equation and Result



Recall the previous (blurred) backprojection illustration

• now using the filtered projections:



# Filtered Backprojection: Illustration



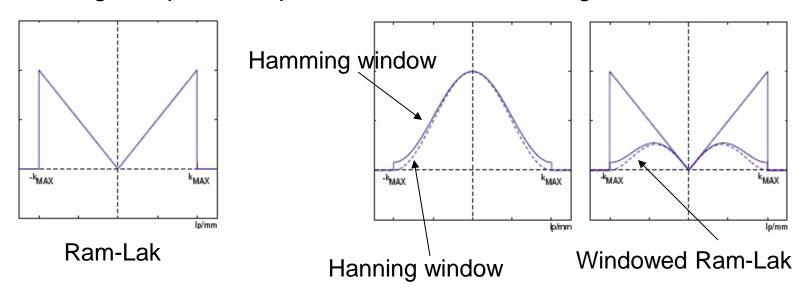
#### **Filters**

#### There are various filters:

- all filters have large spatial extent → convolution would be expensive
- therefore the filtering is usually done in the frequency domain → the required two FT's plus the multiplication by the filter function has lower complexity

#### Popular filters:

- Ram-Lak: original ramp filter limited to interval  $[\pm k_{max}]$
- Ram-Lak with Hanning/Hamming smoothing window: de-emphasizes the higher spatial frequencies to reduce aliasing and noise

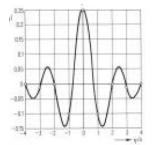


#### **Filters**

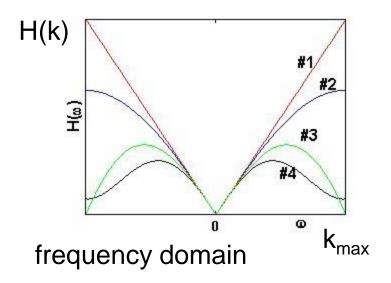
### Frequency domain:

• #1: Ram-Lak

$$H(k) = |k|$$



spatial domain



#2: Shepp-Logan

$$H(k) = |k| \cdot \operatorname{sinc}(\frac{\pi k}{2k_{\max}})$$

- #3: cosine  $H(k) = \cos(k/k_{\text{max}})$
- #4: Hamming ( $\alpha$ =0.54) and Hanning ( $\alpha$ =0.5)

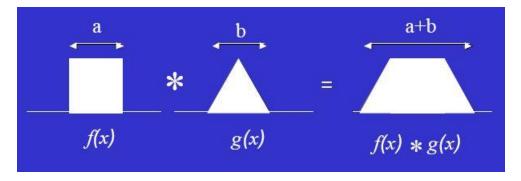
$$H(k) = \alpha + (1 - \alpha)\cos(\frac{\pi k}{k_{\text{max}}})$$

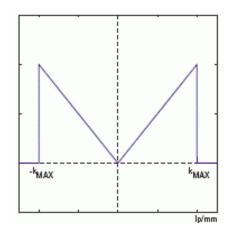
### **Filtering: Details**

#### The filter seems to set the DC term to zero:

- but the reconstructed image has all positive values with a non-zero average
- how can that be explained?

#### Recall convolution theory



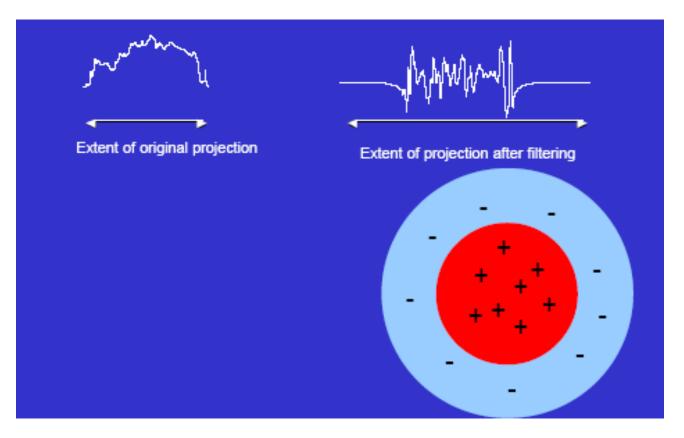


- the length of a convolved signal is the sum of the supports of the individual functions
- since convolution in the spatial domain is equivalent to multiplication in the frequency domain, there is a direct correspondance

## **Filtering: Details**

#### Thus:

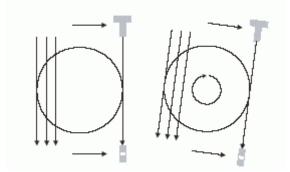
- the extent of the projection doubles with filtering
- and the average value of this image is indeed zero
- but we are only interested in the inner part
- so it all works out!



#### **Beam Geometry**

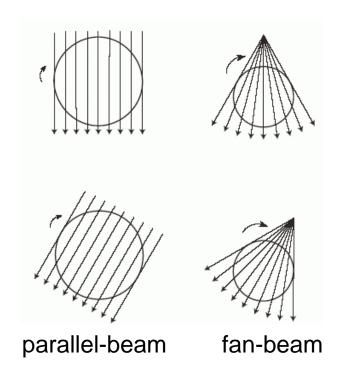
#### The parallel-beam configuration is not practical

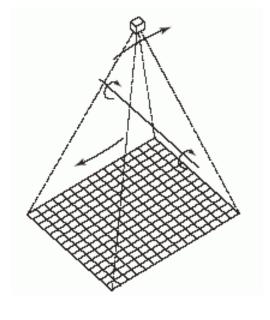
it requires a new source location for each ray



#### We'd rather get an image in "one shot"

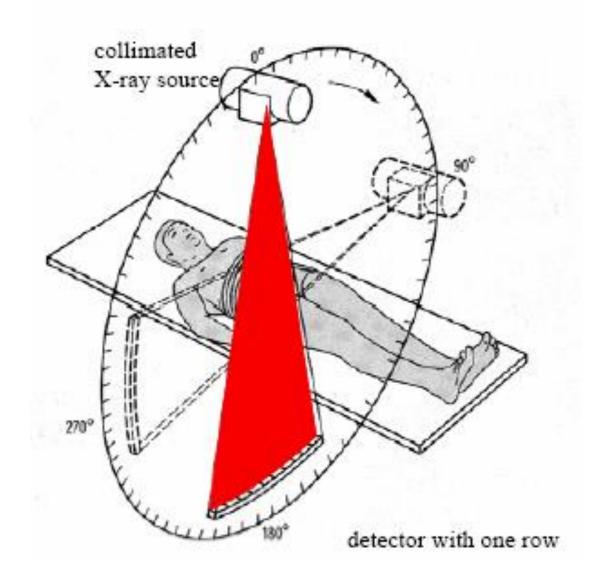
• the requires fan-beam acquisition





cone-beam in 3D

## **Fan Beam Acquisition**



from: Dr. Günter Lauritsch, Siemens

μ(**x**,**y**)

Rewrite the parallel-beam equations into the fan-beam geometry

#### Recall:

• filtering:

$$p*(r,\theta) = \int_{-FOV/2}^{+FOV/2} p(r',\theta)q(r-r')dr'$$

• backprojection:

$$f(x, y) = \int_{0}^{\pi} p^{*}(r, \theta)d\theta$$
 with  $r = x\cos\theta + y\sin\theta$ 

and combine:

$$f(x,y) = \int_{0}^{2\pi} \int_{-FOV/2}^{+FOV/2} p(r',\theta)q(x\cos\theta + y\sin\theta - r')dr'd\theta$$

v(x,y) = distance

from source

$$f(x,y) = \int_{0}^{2\pi + FOV/2} \int_{-FOV/2}^{2\pi + FOV/2} p(r',\theta)q(x\cos\theta + y\sin\theta - r')dr'd\theta$$

with change of variables:

$$\theta = \alpha + \beta$$
  $r = R \sin \alpha$ 

• for a voxel (x,y): v = distance,  $\gamma = \text{angle}$ 

$$v = \sqrt{(x\cos\beta + y\sin\beta)^2 + (x\sin\beta - y\cos\beta + R)^2}$$

$$\gamma = \operatorname{atan}((x\cos\beta + y\sin\beta) / (x\sin\beta - y\cos\beta + R))$$

the projection at  $\boldsymbol{\beta}$ 

$$f(x,y) = \int_{0}^{2\pi} \frac{1}{v^{2}(x,y)} \int_{fan-angle/2}^{fan-angle/2} \frac{R\cos\alpha}{R\cos\alpha} \cdot p(\alpha,\beta) \cdot (\frac{\gamma-\alpha}{2\sin(\gamma-\alpha)}) q(\gamma-\alpha) d\alpha \ d\beta$$
3. weighting during backprojection 1. projection pre-weighting 2. filter

See chapter in Kak-Slaney (posted on the class website) for equations associated with flat detectors

#### So, reconstruction from fan-beam data involves

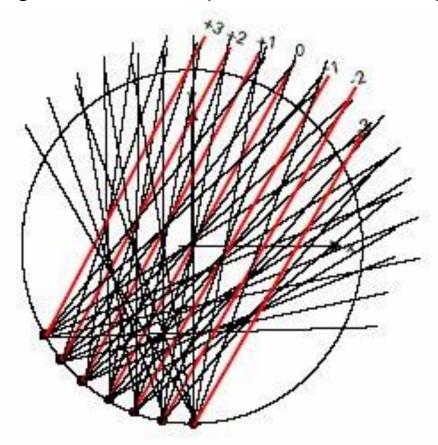
- a pre-weighting of the projection data, depending on  $\alpha$
- a pre-weighting of the filter (here we used the spatial domain filters)
- a backprojection along the fan-beam rays (interpolation as usual)
- a weighting of the contributions at the reconstructed pixels, depending on their distance v(x, y) from the source

#### There are also iterative algorithms

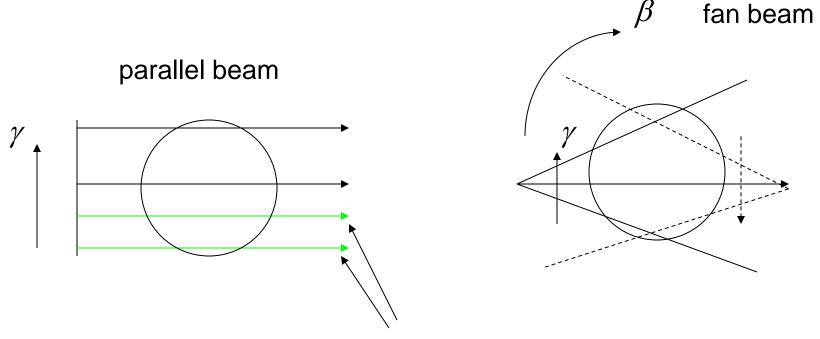
- these pose the reconstruction problem as a system of linear equations
- solution via iterative solvers
- more on this to come in the nuclear medicine lectures

## Alternatively, one could also "rebin" the data into a parallelbeam configuration

 however, this requires an additional interpolation since there is no direct mapping into a uniform parallel-beam configuration

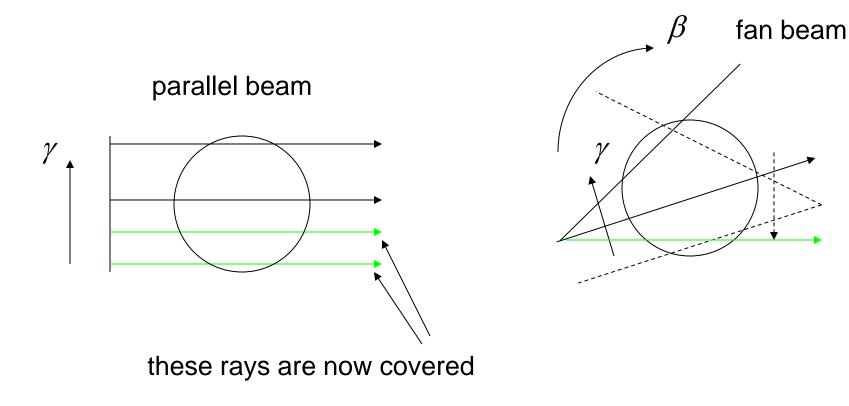


Problem: fan-beam does not fill the sinogram adequately



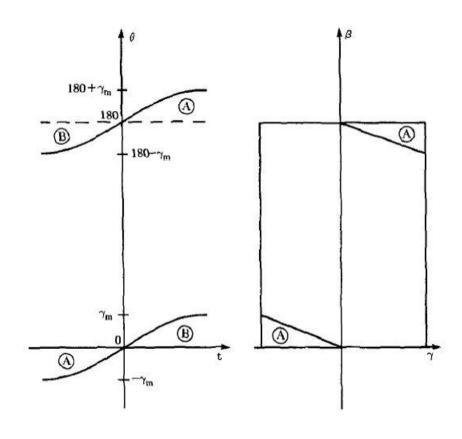
these rays (and others) are not covered by any fan-beam view

Solution: extend the source-detector trajectory by the fan halfangle on both ends



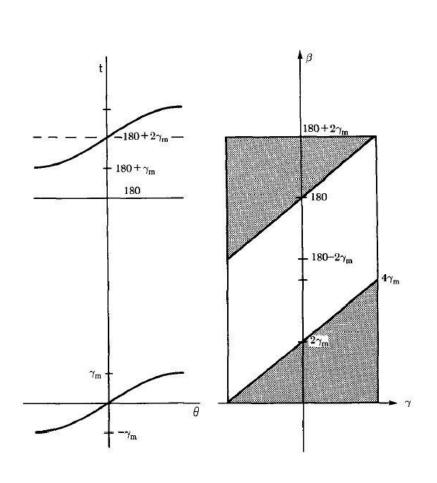
## More formally

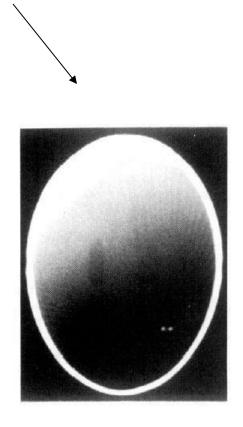
• region A is covered twice, while region B is not covered at all



## Extending the trajectory fills the space

• but some areas are filled twice, which causes problems





# Simply setting these regions to zero will result in heavy streak artifacts

recall the filtering step?

#### Need to use a smoother window

 a smooth window is both continuous and has a continuous derivative at the boundary of single and double-overlap regions

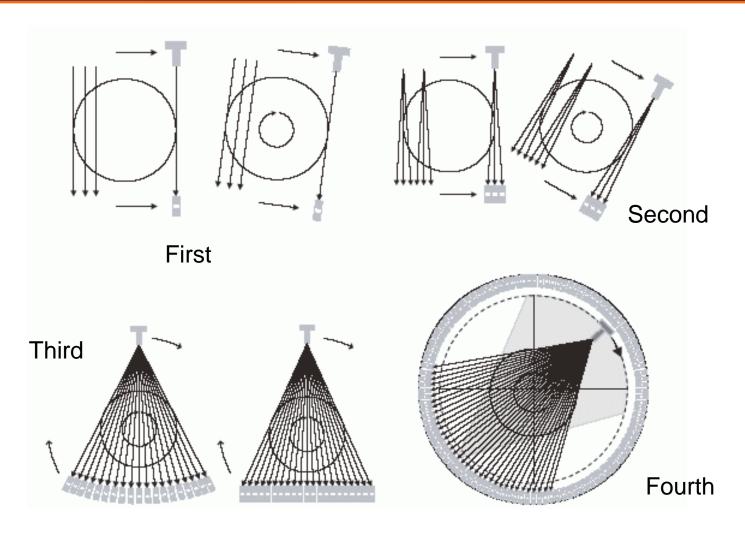
$$\left. \frac{\partial w_{\beta}(\gamma)}{\partial \beta} \right|_{\beta = 2\gamma_m + 2\gamma} = 0$$

$$\frac{\partial w_{\beta}(\gamma)}{\partial \beta}\bigg|_{\beta=180^{\circ}+2\gamma}=0$$

• the window weights for the same rays at opposite sides of the sinogram must be 1.  $w_{\beta_1}(\gamma_1) + w_{\beta_2}(\gamma_2) = 1$ 

• the Parker window fullfills these conditions: 
$$w_{\beta}(\gamma) = \begin{cases} \sin^2\left[\frac{45^{\circ}\beta}{\gamma_m - \gamma}\right], & 0 \leq \beta \leq 2\gamma_m - 2\gamma \\ 1, & 2\gamma_m - 2\gamma \leq \beta \leq 180^{\circ} - 2\gamma \end{cases}$$
$$\sin^2\left[45^{\circ}\frac{180^{\circ} + 2\gamma_m - \beta}{\gamma + \gamma_m}\right], \quad 180^{\circ} - 2\gamma \leq \beta \leq 180^{\circ} + 2\gamma_m.$$

#### **Scanner Generations**

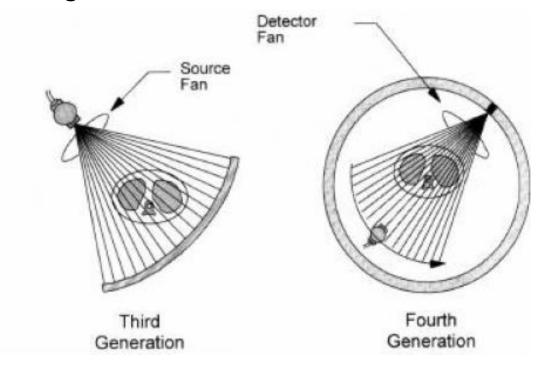


Third generation most popular since detector geometry is simplest

collimation is feasible which eliminates scattering artifacts

#### **Fan Beam Scanners**

The 3<sup>rd</sup> and 4<sup>th</sup> generation scanners:



However, in the 3<sup>rd</sup> generation scanner:

• the detector width (the beam aperture  $\Delta s$ ) = detector spacing  $\Delta r$ 

• recall our earlier discussion on sampling constraints where we found that:  $1 \quad 2 \quad \Delta s$ 

 $\frac{1}{\Delta r} \ge \frac{2}{\Delta s} \quad \to \quad \Delta r \le \frac{\Delta s}{2}$ 

## **Fan Beam Data Acquisition: Practice**

#### So, we should acquire 2 samples per detector width

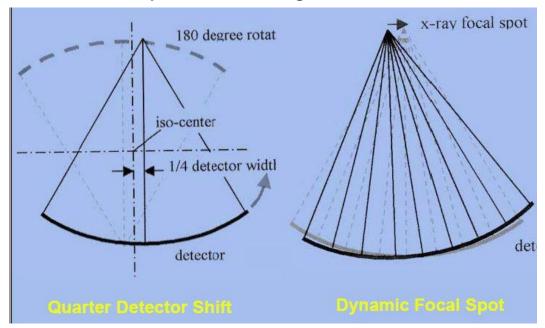
- a symmetrical rotation configuration violates this requirement
- the consequence is ray aliasing:



## Ray Aliasing Remedies

## For 3<sup>rd</sup> generation scanners:

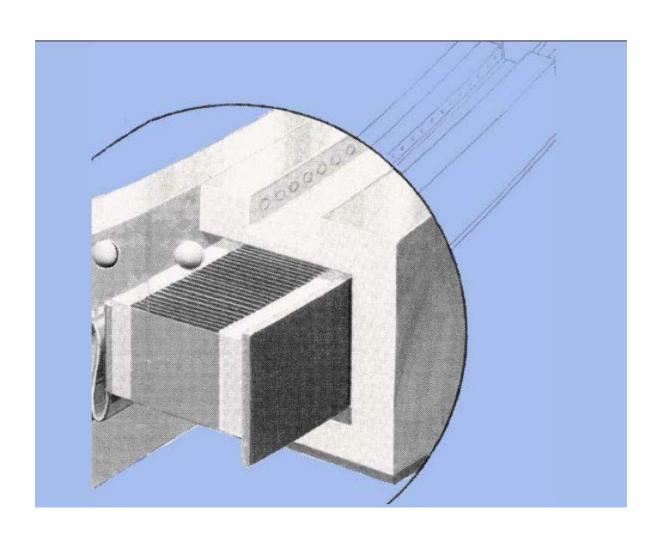
- ¼ detector shift
- dynamic focal spot
- both double the density of the sinogram with little technical overhead



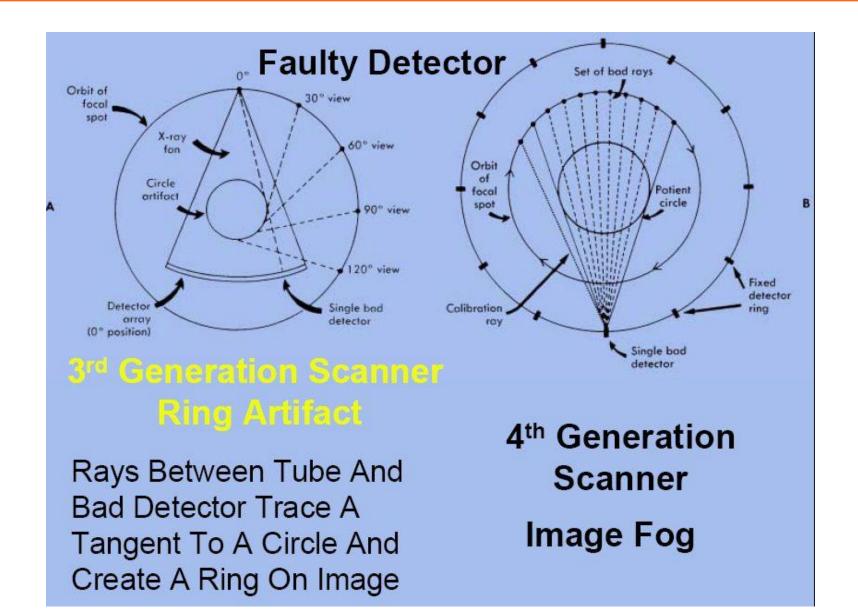
## For 4<sup>rd</sup> generation scanners:

- move the X-ray tube at slower speeds
- this increases the number of ray samples

# **A Detector Element**

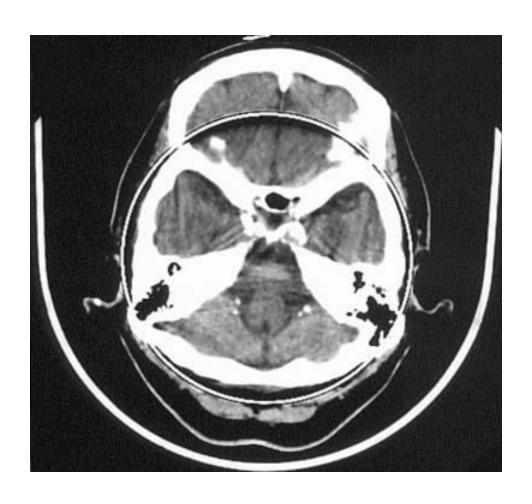


#### **Artifacts Related to Faulty Detectors**



# **Ring Artifacts**

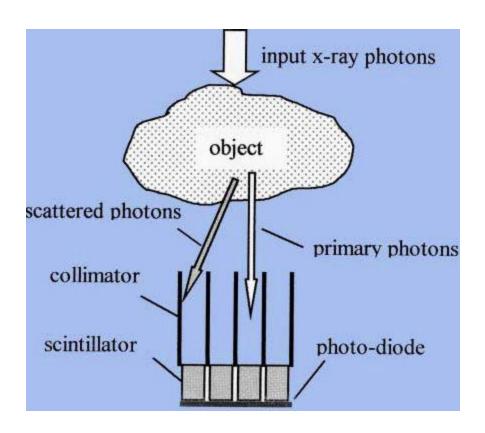
See larger ring just at the edge of the skull



#### **A Note on Collimation**

Collimation ensures that we know the ray direction at each detector bin (perpendicular to the local tangent)

this enables reconstruction theory



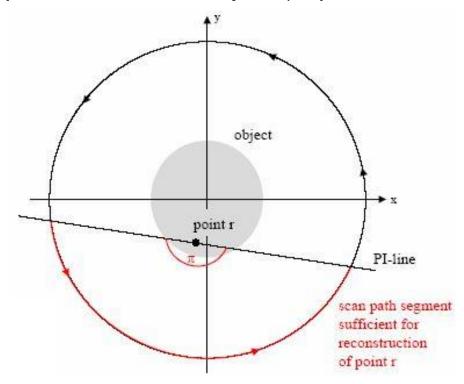
#### **Short-Scan CT**

#### Requirement:

• an object point r can be reconstructed exactly if it sees a scan path segment of an angular range  $\pi$ 

#### Consequence:

• an smaller Region of Interest (ROI) can be reconstructed without acquiring complete data of the object (super short-scan).



#### **Short-Scan CT**

Specific algorithms are needed for reconstruction from a super short-scan

- F. Noo et al., BMP 2002
- H. Kudo et al., IEEE NSS 2002

