
Computed Tomography

BME/EECS 516
CT lecture #2

Announcements

- MRI Project due Tues, Nov 21
- HW #6 (x-ray), Due Tues, Nov 28
- US and MRI Demos – Tues, Nov 21
 - Meet at entrance to Gerstacker Building at 3pm.

Last Lecture

- CT introduction + history + basics
 - Line integral
 - Radon transform
- Central Section Theorem

Computed Tomography

- **Computed tomography (CT):** digital projective geometry processing is used to generate a 3D image of an object from a large series of x-ray 2D projection taken around a single axis of rotation.

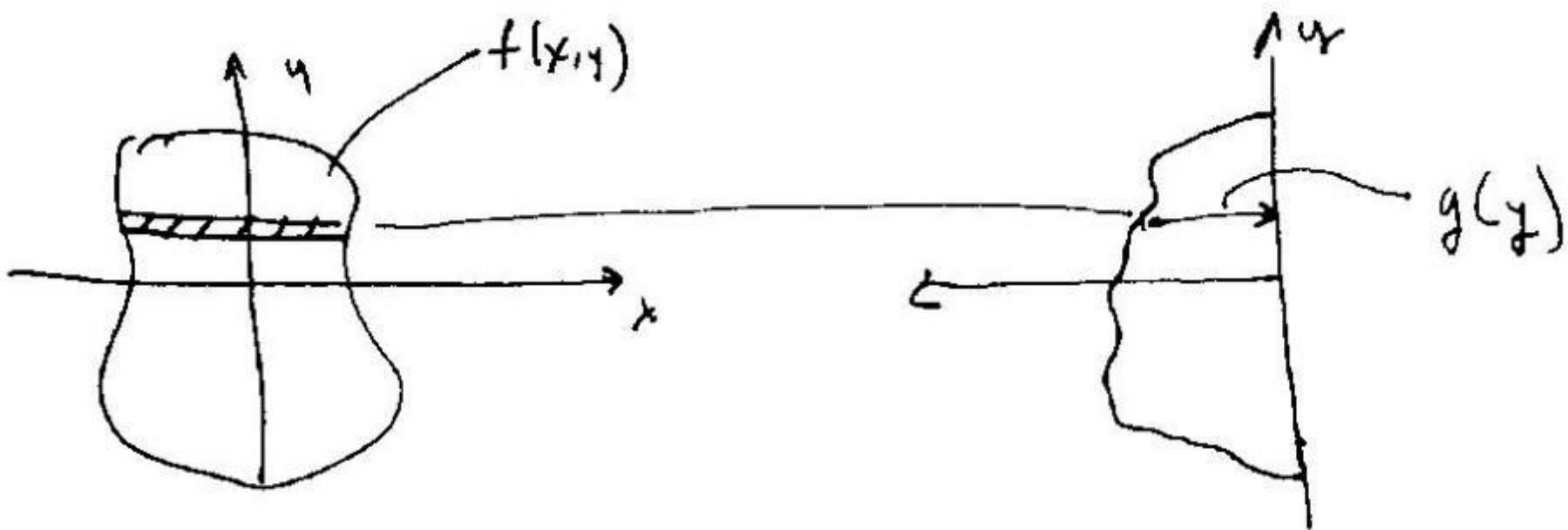
Computed Tomography

- Define the “line integral” through f in the x direction

$$g(y) = \ln \frac{I_0}{I_d(y)} = \int \mu(x, y) dx = \int f(x, y) dx$$

- Goal: solve $f(x,y)$ from $g(y)$ contain spatial attenuation coefficient info

Computed Tomography



- Sampling at $y = R$

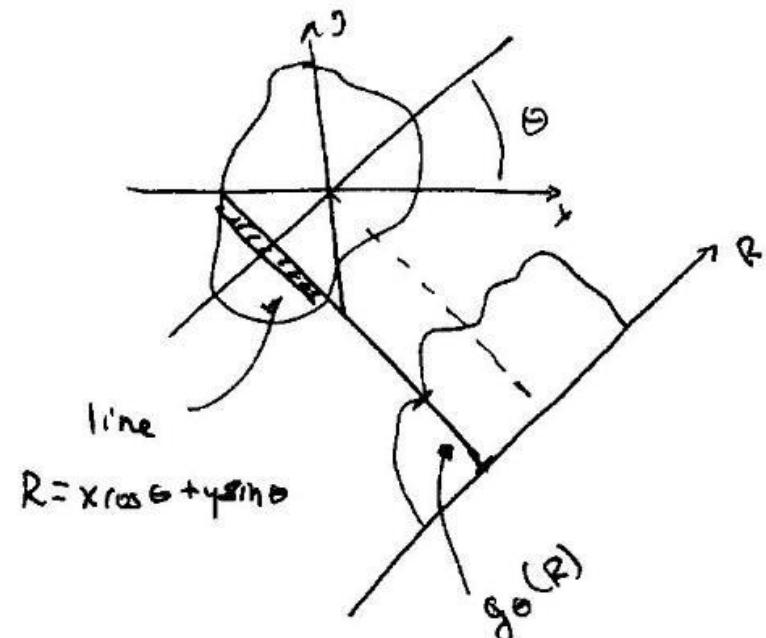
$$g(R) = \iint f(x, y) \delta(y - R) dx dy = \int f(x, R) dx$$

Computed Tomography

Line integral at an arbitrary angle, θ

When $R = x \cos \theta + y \sin \theta$:

$$g_\theta(R) = \iint f(x, y) \delta(x \cos \theta + y \sin \theta - R) dx dy$$



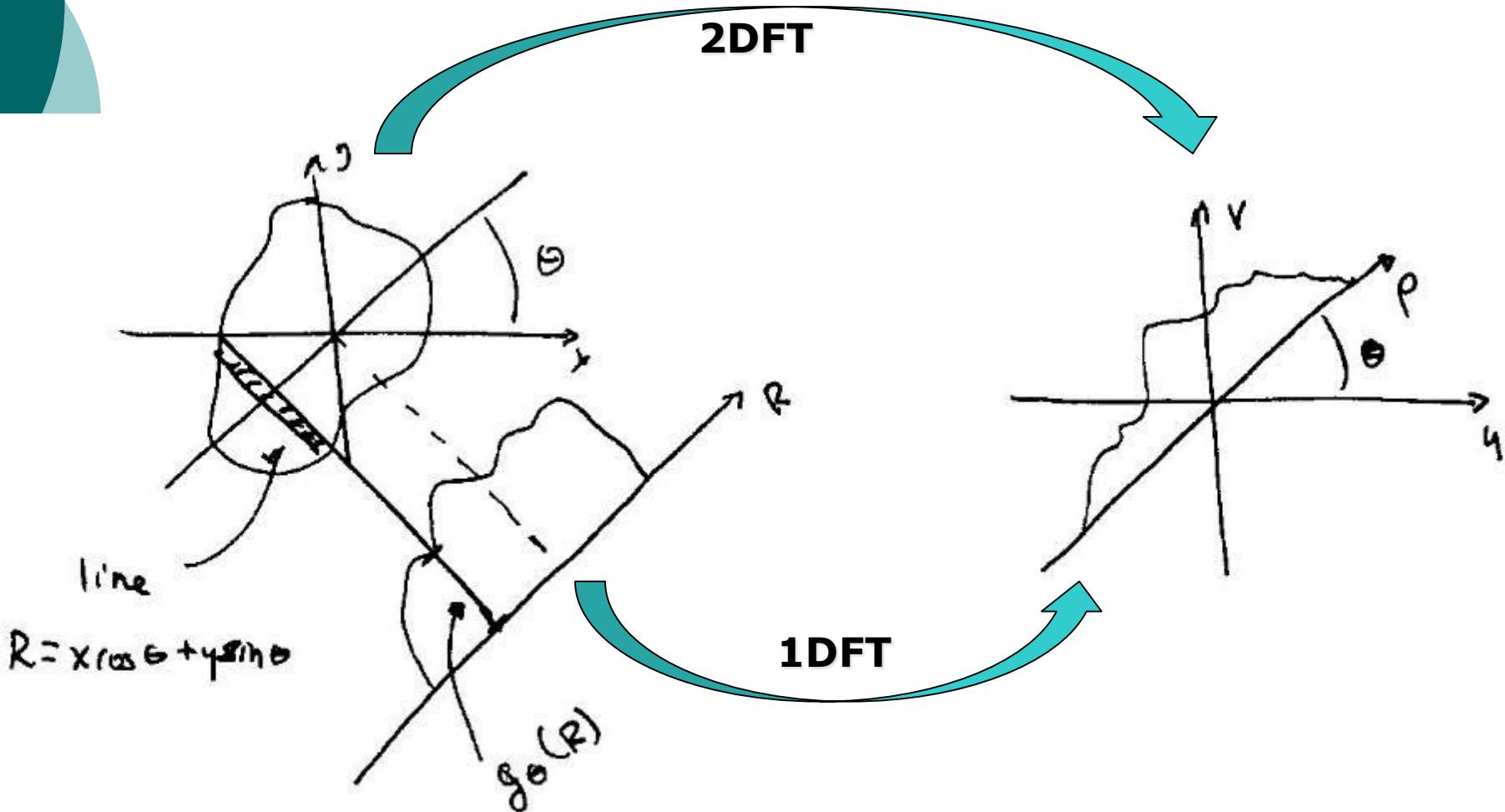
$g_\theta(R)$ is known as the **Radon transform** of $f(x, y)$.

Central Section Theorem

The central section theorem:

- The 1D FT of a projection $g_\theta(R)$ is the 2D FT of $f(x,y)$ evaluated at angle θ .

Central Section Theorem



$$G_\theta(\rho) = F_{1D,R}\{g_\theta(R)\} = F_{2D}\{f(x, y)\}_{u=\rho \cos \theta, v=\rho \sin \theta}$$

Central Section Theorem

- 2D FT of $f(x,y)$

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

- 1D FT of $g_\theta(R)$

$$G_\theta(\rho) = F_{1D(R)} \{ g_\theta(R) \}$$

$$= \iint f(x,y) \exp(-i2\pi(\rho \cos \theta x + \rho \sin \theta y)) dx dy$$

- Compare the two - same when $u = \rho \cos \theta, v = \rho \sin \theta$ (polar coords.)

Central Section Theorem

- 2D FT in Polar Coordinates.

- $u = \rho \cos \theta, v = \rho \sin \theta$
- A line in ρ at angle θ

- Thus:

$$G_\theta(\rho) = F_{1D(R)}\{g_\theta(R)\}$$

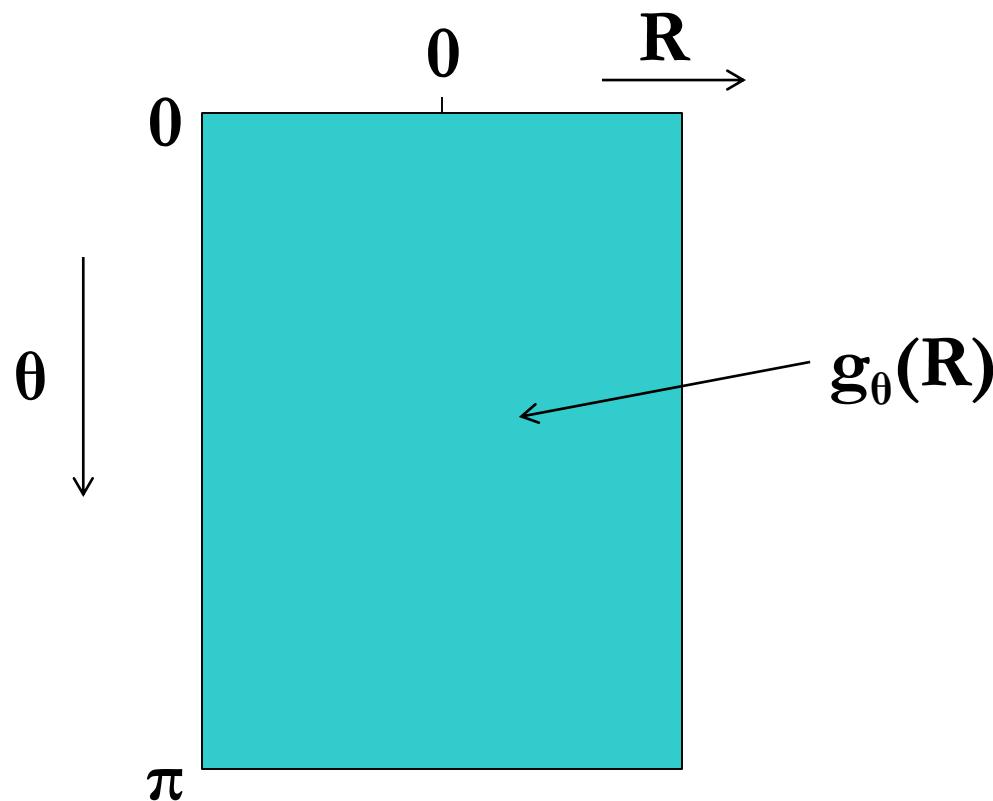
$$G_\theta(\rho) = F(u, v) \Big|_{u=\rho \cos \theta, v=\rho \sin \theta} = F(\rho, \theta)$$

Computed Tomography

- Projections → Images
- Image reconstruction: Inverse problems
- We will present several approaches

Sinograms

- Sinograms - $g_\theta(R)$ data matrix for many different angles θ



Sinograms

- Example – point projection at (x_0, y_0)

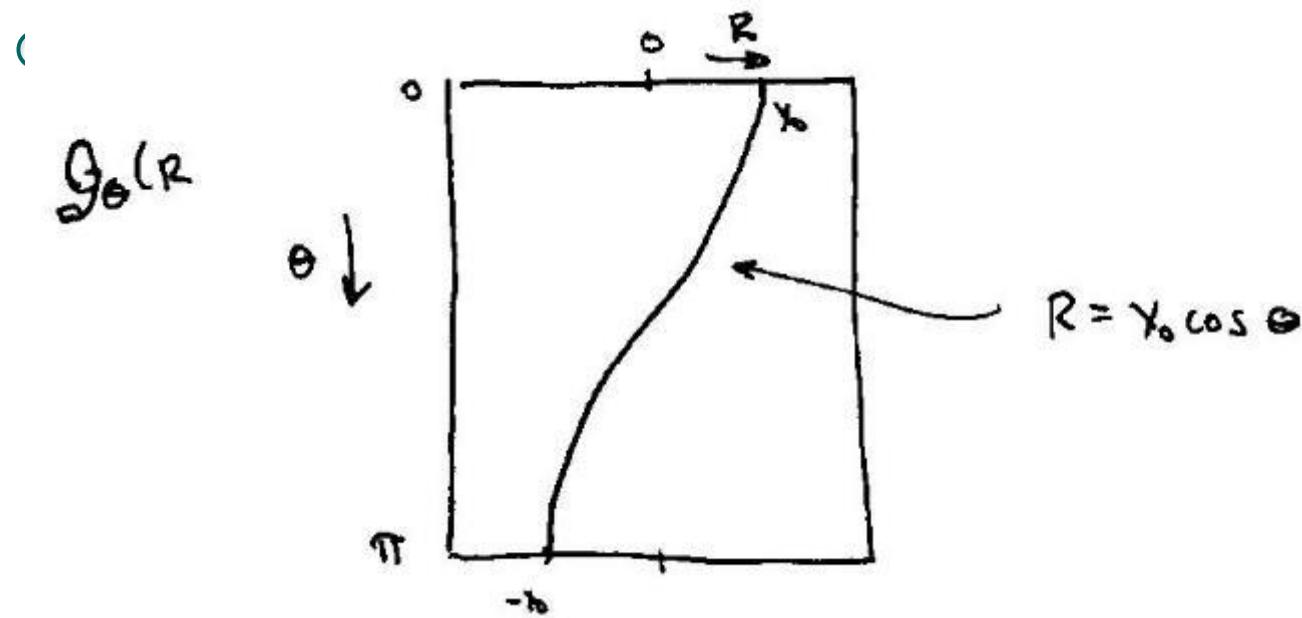
$$\begin{aligned}g_\theta(R) &= \iint \delta(x - x_0)\delta(y - y_0)\delta(x\cos\theta + y\sin\theta - R)dxdy \\&= \delta(x_0\cos\theta + y_0\sin\theta - R)\end{aligned}$$

- Letting $y_0 = 0$, then:

$$g_\theta(R) = \delta(x_0\cos\theta - R)$$

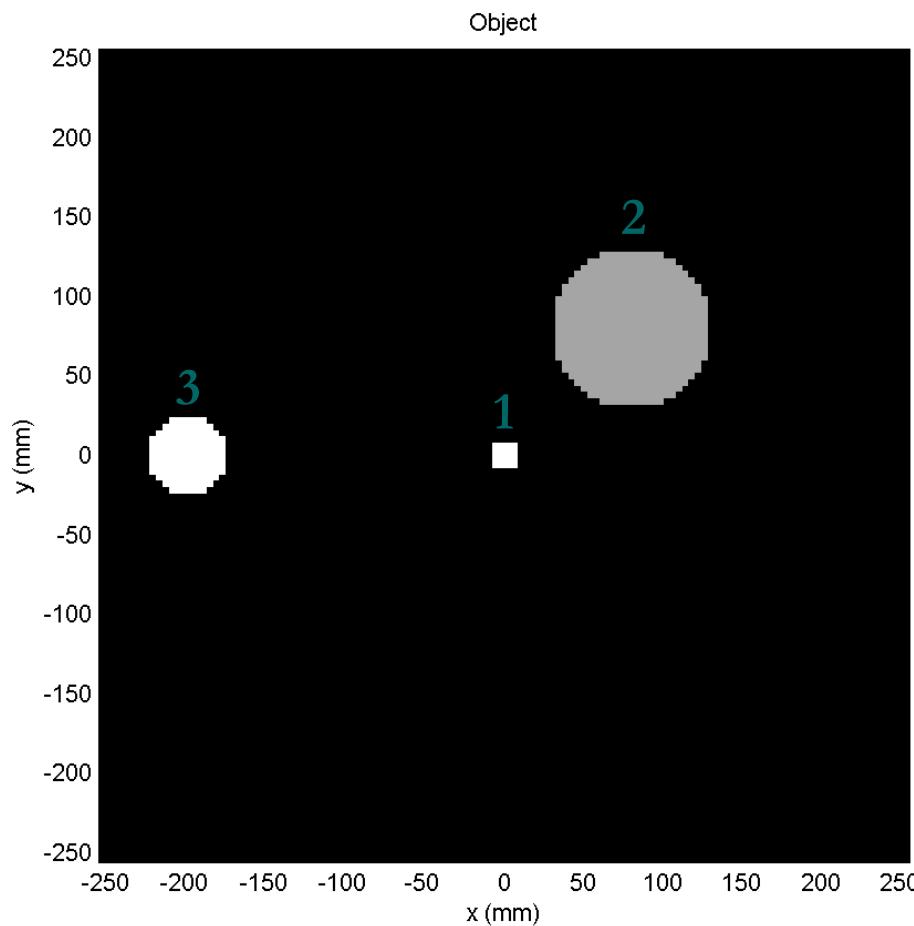
Sinograms

- A point traces out a sinusoid in the $R-\theta$ space --thus the name sinogram.



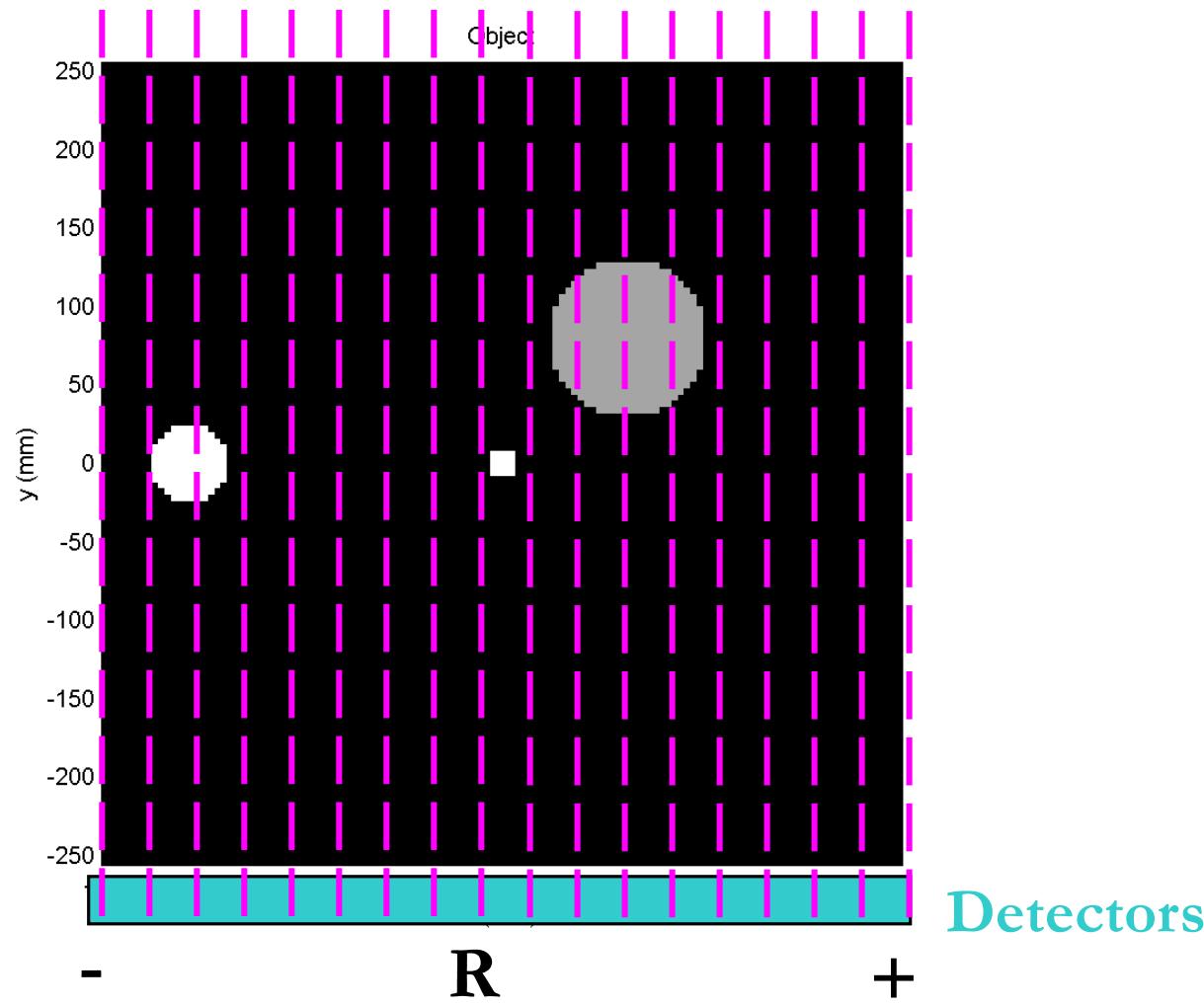
Sinograms

○ Objects



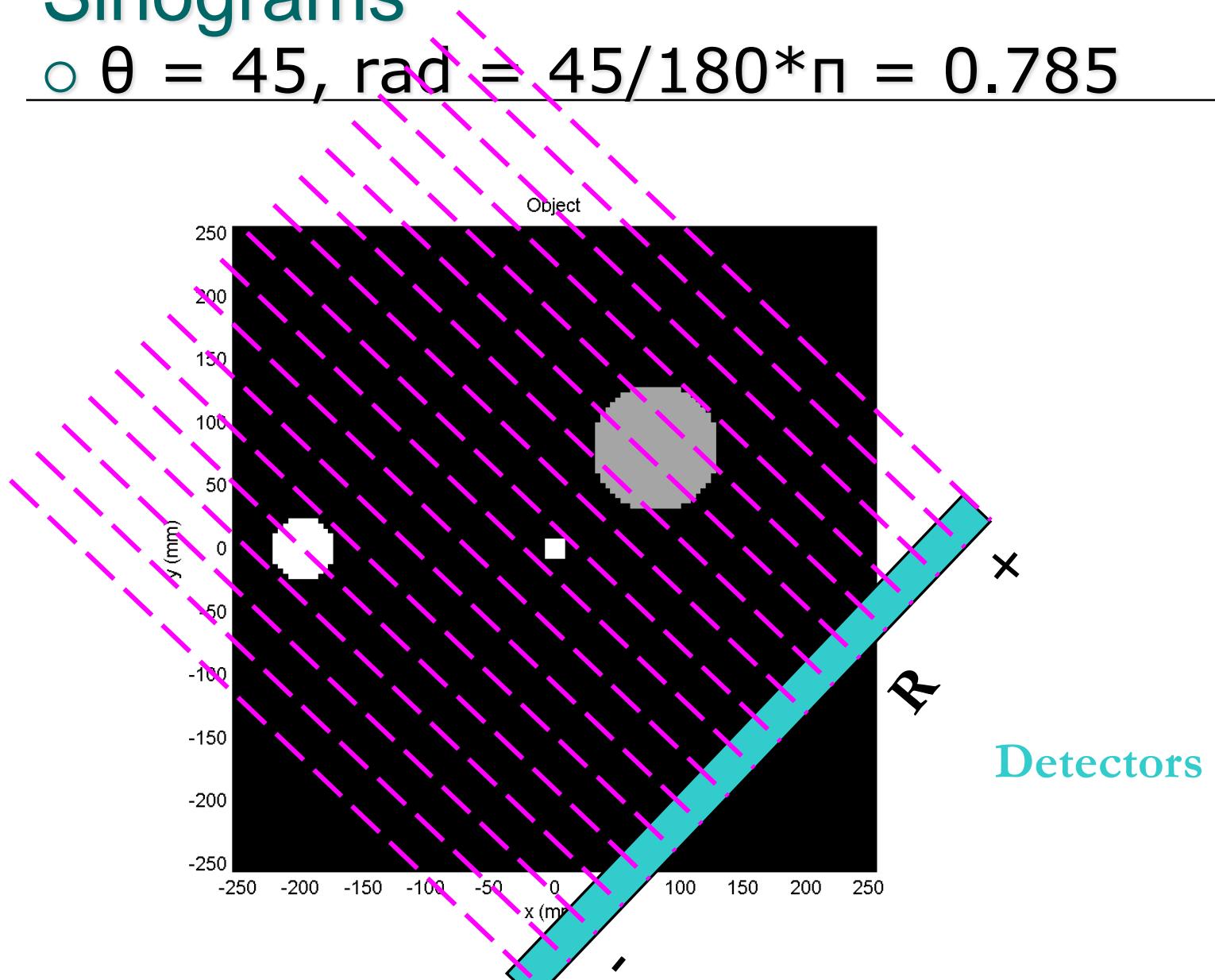
Sinograms

- $\theta = 0$, get project

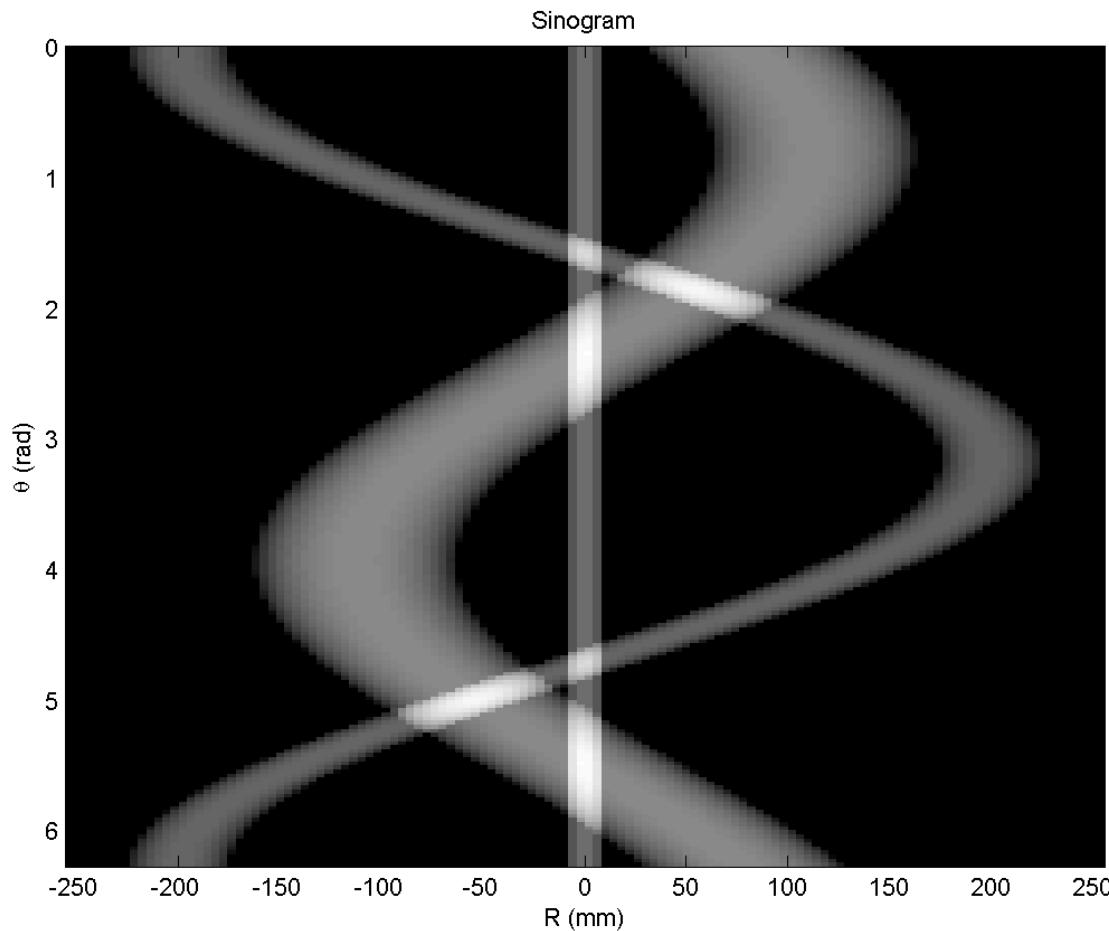


Sinograms

- $\theta = 45, \text{rad} = 45/180\pi = 0.785$



Sinograms



Sinograms

- The maximum deviation describes an object's distance from the origin
- The point of peak deviation describes the angular location of object.
- Three objects above, from smallest to largest are located at
 - $(r, \theta) = (0,0)$
 - $(r, \theta) = (113, \pi/4)$
 - $(r, \theta) = (-200,0) \text{ or } (200, \pi)$.
- Symmetry: $\rho(r, \theta) = \rho(-r, \theta + \pi)$



Question?

Sinograms

Reconstruction Methods

1. Direct Fourier Interpolation Method
2. Backprojection-Filtering Method
3. Direct Fourier Superposition and Filtering Method
4. Filtered Backprojection Method
5. Algebraic Reconstruction Technique (ART)

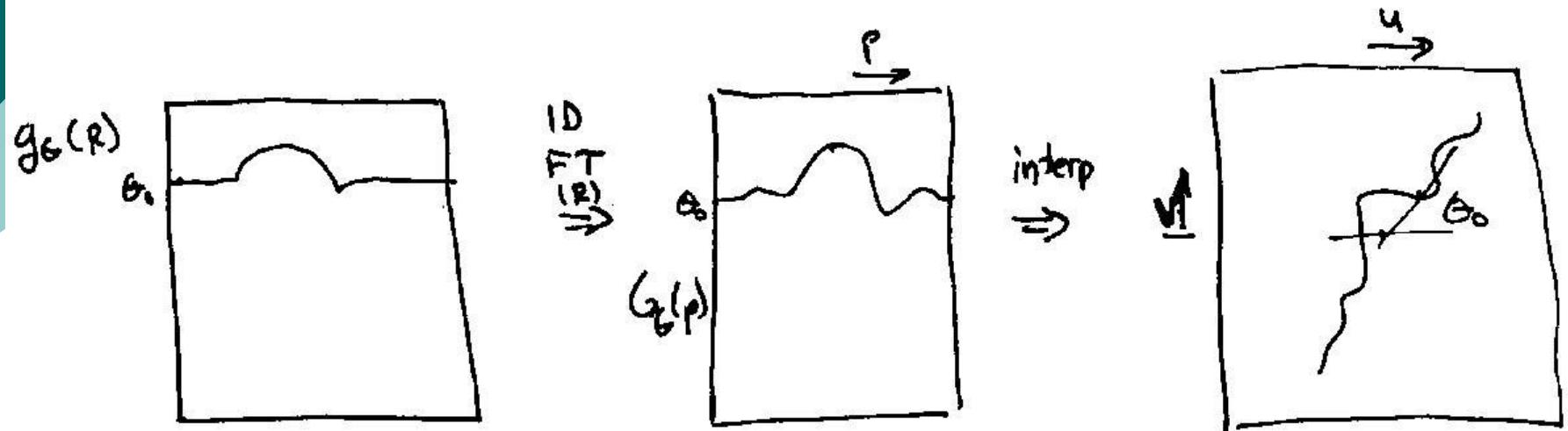
Reconstruction Methods

- 1. Direct Fourier Interpolation Method – CT project**
- 2. Backprojection-Filtering Method – Introduce backprojection concept**
- 3. Direct Fourier Superposition and Filtering Method**
- 4. Filtered Backprojection Method – CT project**
- 5. Algebraic Reconstruction Technique (ART) – conventional method**

1. Direct Fourier Interpolation Method

- Direct use of the Central Section Theorem. Steps:
 1. 1D FT each of the projections:
$$F_{1D}\{g_\theta(R)\} = G_\theta(\rho) = F(\rho, \theta)$$
 2. Interpolate $F(\rho, \theta)$ to $\hat{F}(u, v)$ (polar to rectangular coordinates – e.g. Matlab functions interp2 or griddata)
 3. Inverse 2D FT

1. Direct Fourier Interpolation Method



Can use interp2 or griddata in Matlab.

1. Direct Fourier Interpolation Method

- Simple concepts but have drawbacks:
 - Two-dimensional inverse FT is required for each projection line.
 - Involves interpolations and coordinate transformations.
 - Errors associated with interpolation in Fourier space
 - Complex implementation for practical systems

Question?

Direct Fourier Interpolation Method



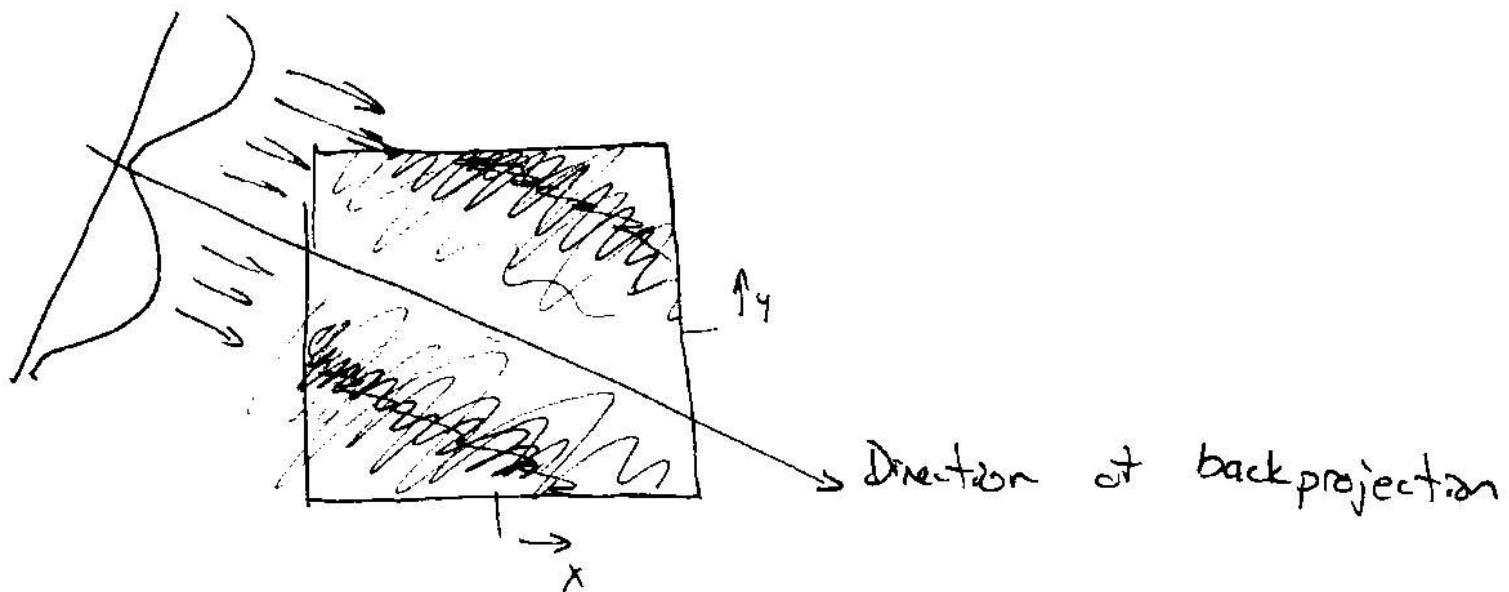
Backprojection

Backprojection

- Alternative reconstruction method
- Based on the same general principles but having distinct computational advantages
- Introduce backprojection

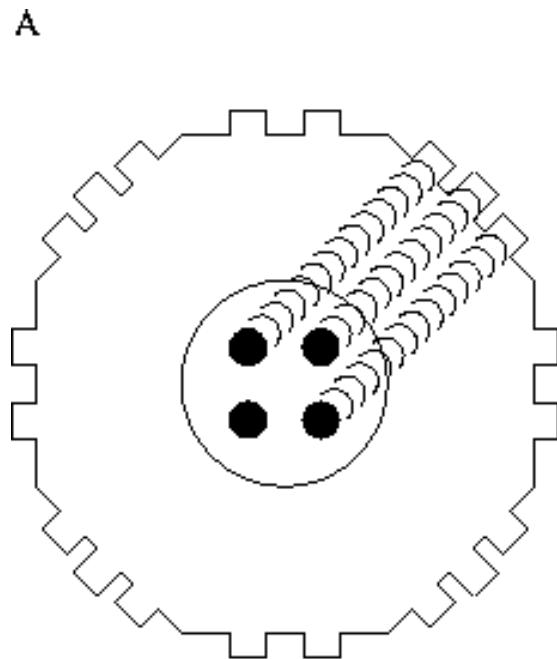
Backprojection

- Backprojection – the measurements obtained at each projection are projected back along the same line (hence “backproject) - “Smear” the projection data back across the object space

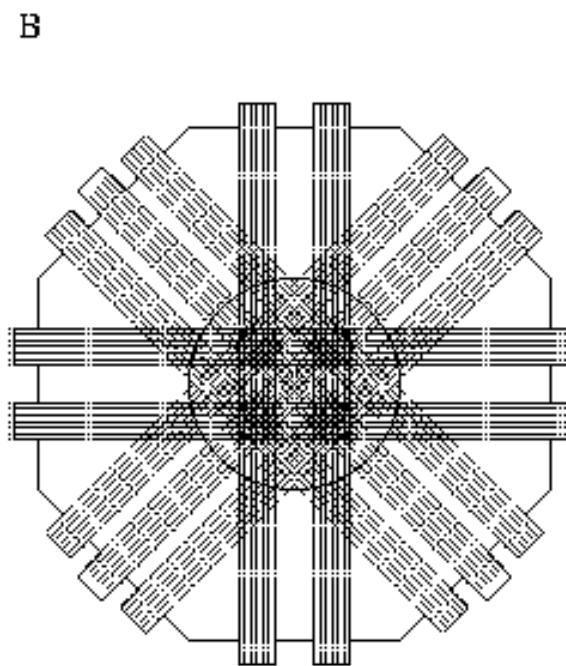


Backprojection

Illustration of back projection



Projection



Backprojection

Backprojection

- Backprojection of a single measured projection at angle θ

$$b_\theta(x, y) = \int_{-\infty}^{\infty} g_\theta(R) \delta(x \cos \theta + y \sin \theta - R) dR$$

- The total backprojected image is the integral (sum) of this over all angles:

$$f_b(x, y) = \int_0^{\pi} b_\theta(x, y) d\theta$$

$$= \int_0^{\pi} \int_{-\infty}^{\infty} g_\theta(R) \delta(x \cos \theta + y \sin \theta - R) dR d\theta$$

Backprojection

- Now consider:

$$g_\theta(R) = F_{1D,\rho}^{-1}\{F(\rho, \theta)\} = \int_{-\infty}^{\infty} F(\rho, \theta) \exp(i2\pi\rho R) d\rho$$

- The backprojected image can be written as:

$$\begin{aligned} f_b(x, y) &= \int_{-\infty}^{\infty} \int_0^{\pi} \int_{-\infty}^{\infty} F(\rho, \theta) \delta(x \cos \theta + y \sin \theta - R) \exp(i2\pi\rho R) dR d\theta d\rho \\ &= \int_0^{\pi} \int_{-\infty}^{\infty} F(\rho, \theta) \exp(i2\pi\rho(x \cos \theta + y \sin \theta)) d\theta d\rho \end{aligned}$$

Backprojection

$$f_b(x, y) = \int_0^{\pi} \int_{-\infty}^{\infty} F(\rho, \theta) \exp(i2\pi\rho(x \cos \theta + y \sin \theta)) d\theta d\rho$$

- Comparing to inverse 2D FT of $F(\rho, \theta)$

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

- Differences:
 - limits of integration
 - $d\theta d\rho$ vs. $\rho d\theta d\rho$

Backprojection

- To address difference #1: limit of integration

$$F(\rho, \theta) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x, y) \exp(-i2\pi(\rho \cos \theta x + \rho \sin \theta y)) dx dy$$

$$\begin{aligned} F(-\rho, \theta) &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x, y) \exp(-i2\pi(-\rho \cos \theta x - \rho \sin \theta y)) dx dy \\ &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x, y) \exp(-i2\pi(\rho \cos(\theta + \pi)x + \rho \sin(\theta + \pi)y)) dx dy \\ &= F(\rho, \theta + \pi) \end{aligned}$$

Backprojection

$$f_b(x, y) = \int_0^{\pi} \int_{-\infty}^{\infty} F(\rho, \theta) \exp(i2\pi\rho(x \cos \theta + y \sin \theta)) d\rho d\theta$$

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

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

$$= F_{2D}^{-1} \left\{ \frac{F(\rho, \theta)}{\rho} \right\}$$

$$= f(x, y) \ast \ast F_{2D}^{-1} \left\{ \frac{1}{\rho} \right\}$$

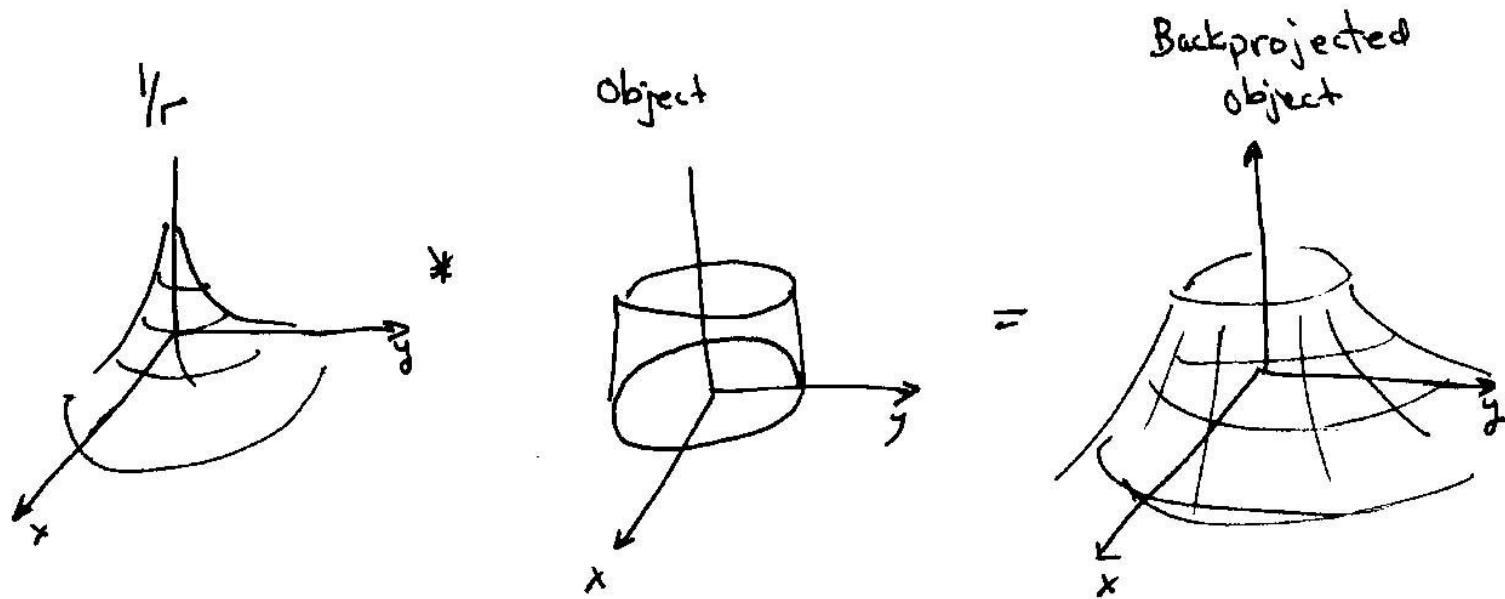
$$= f(x, y) \ast \ast \frac{1}{r}$$

Address
difference #1

Address
difference #2

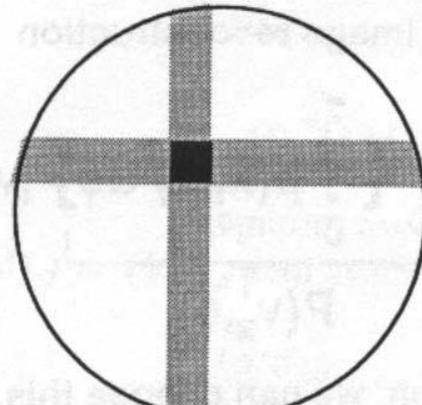
Backprojection

- Backprojected image is equal to the desired image convolved with a $1/r$ blurring function

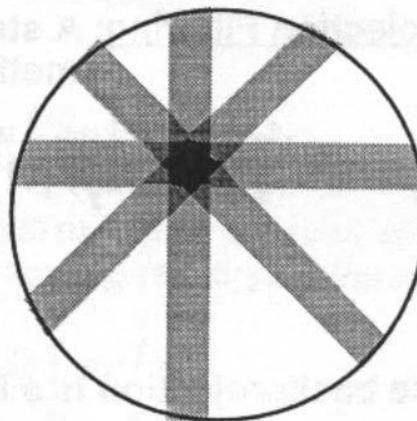


Unfiltered Backprojection

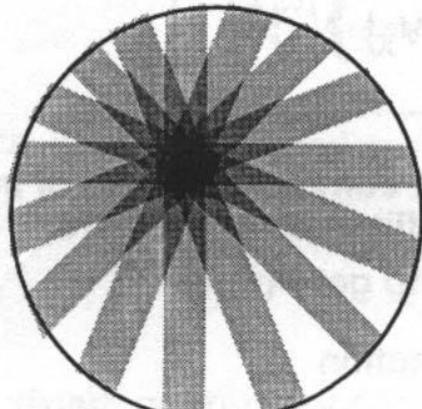
Unfiltered Backprojection Example



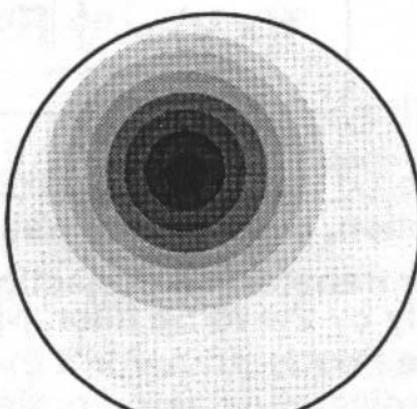
2 Backprojections



4 Backprojections



8 Backprojections



Many Backprojections

Backprojection Summary

1. Backprojection of a single measured projection at angle θ

$$b_\theta(x, y) = \int_{-\infty}^{\infty} g_\theta(R) \delta(x \cos \theta + y \sin \theta - R) dR$$

2. Total backprojected image - integral (sum) of this over all angles:

$$f_b(x, y) = \int_0^\pi b_\theta(x, y) d\theta$$

3. Backprojected image is equal to the desired image convolved with $1/r$

$$F_{2D}\{f_b(x, y)\} = \frac{F(r, q)}{r}, \quad f_b(x, y) = f(x, y) \ast \ast \frac{1}{r}$$

Backprojection

- Image reconstruction – remove blurring function to recover $f(x, y)$ from $f_b(x, y)$ (or $b(x, y)$)
- Methods
 - Backprojection-filtered Method
 - Filtered-backprojection Method



Questions?

Backprojection



Backprojection Filtering Method

2. Backprojection-Filtering Method

- Steps:
 1. Backproject all projections, $g_\theta(R)$ to get $f_b(x,y)$
 2. Forward 2D FT to get $\frac{F(\rho,\theta)}{\rho}$
 3. Filter with ρ (or $|\rho|$) to get $F(\rho,\theta)$.
 4. Inverse 2D FT to get $\hat{f}(x,y)$

2. Backprojection-Filtering Method

- Interpolation happens in image domain (better for artifacts)
- The big problem is that the blur $\left(\frac{1}{r}\right)$ extends over a long distance and you need accurately backproject over an extended region
- Presented for comparison (doesn't really get used)



Questions?

Backprojection-Filtering Method

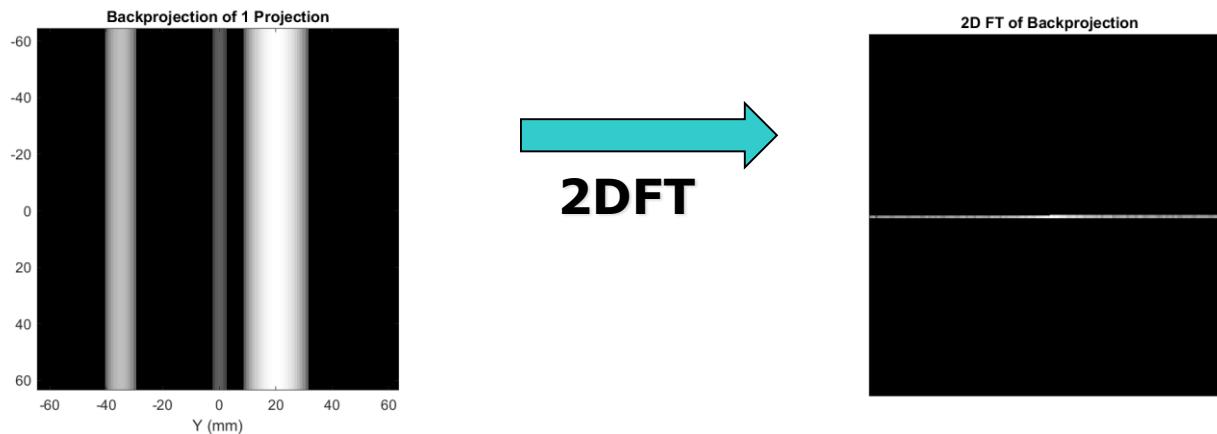


Fourier Superposition and Filtering

3. Fourier Superposition-Filtering

- Steps:
 1. 1D FT all projections, $g_\theta(r)$, to get $G_\theta(\rho)$.
 2. Add $G_\theta(\rho)$ to the 2D FT space to $\hat{F}(\rho, \theta)$ along the line at θ for all θ .
(Note: that this is equivalent to backprojection at all θ)
 3. Filter with $|\rho|$ to get $F(\rho, \theta)$.
 4. Inverse 2D FT to get $\hat{f}(x, y)$.

3. Fourier Superposition-Filtering



- Interpolation happens in Fourier image domain (artifacts)
- Can filter before or after superposition (before is usually better)
- This is equivalent to a method called convolution gridding



Questions?

Fourier Superposition - Filtering
Method



Filtered Backprojection Method

4. Filtered Backprojection Method

Solution to eliminate 2D transforms:

- Still use backprojection
- Undo the $1/r$ blur without:
 - 2D transforms and
 - Backprojection over extended space
- **Idea: Reverse the order of backprojection and filtering**

4. Filtered Backprojection Method

Steps:

1. Filter the backprojection with a $|\rho|$ filter (ramp filter)

- Fourier method:

$$F_{1D}\{g'_q(R)\} = |r| F_{2D}(r, q)$$

$$g'_q(R) = F_{1D}^{-1}\{|r| F_{1D}\{g_q(R)\}\} = F_{1D}^{-1}\{|r| F_{2D}(r, q)\}$$

- Convolution method:

$$g'_\theta(R) = g_\theta(R) * c(R), \text{ where } c(R) \approx F_{1D}^{-1}\{|\rho|\}$$

2. Backproject for all angles to get $\hat{f}(x, y)$

4. Filtered Backprojection Method

- Putting it all together:

$$\hat{f}(x, y) = \int_0^{\pi} \int_{-\infty}^{\infty} F_{1D}^{-1}\left\{\rho \middle| F_{1D}\left\{g_{\theta}(R)\right\}\right\} \delta(x \cos \theta + y \sin \theta - R) dR d\theta$$

$$= \int_0^{\pi} \int_{-\infty}^{\infty} F_{1D}^{-1}\left\{\rho \middle| F(\rho, \theta)\right\} \delta(x \cos \theta + y \sin \theta - R) dR d\theta$$

$$= \int_{-\infty}^{\infty} \int_0^{\pi} \int_{-\infty}^{\infty} \left| \rho \middle| F(\rho, \theta) \exp(i2\pi\rho R) \delta(x \cos \theta + y \sin \theta - R) \right| d\rho dR d\theta$$

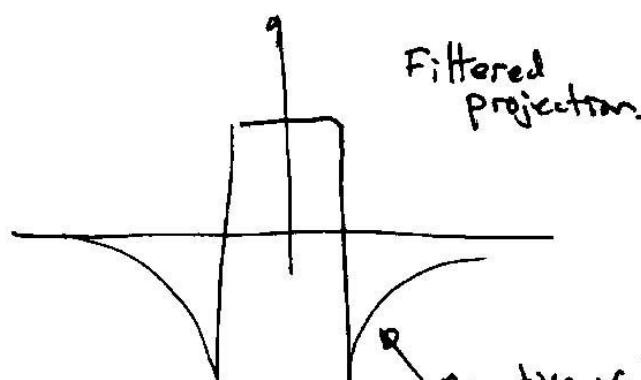
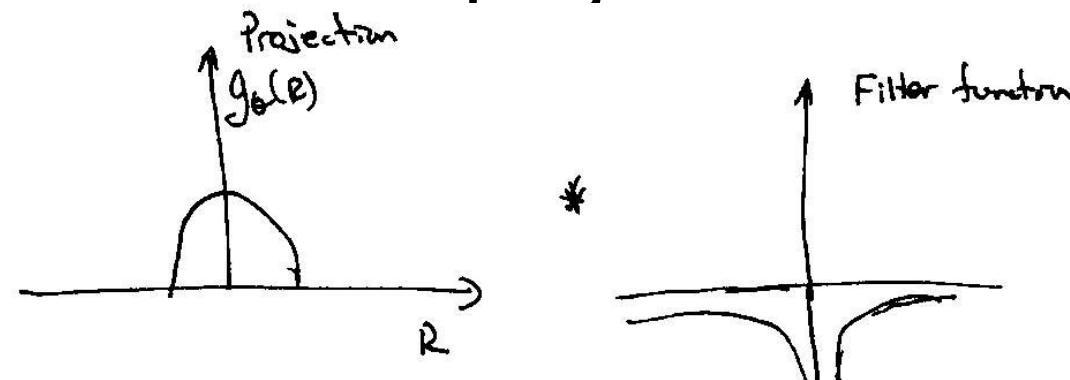
$$= \int_0^{\pi} \int_{-\infty}^{\infty} F(\rho, \theta) \exp(i2\pi\rho(x \cos \theta + y \sin \theta)) \left| \rho \right| d\rho d\theta$$

4. Filtered Backprojection Method

- Changing the limits of integration to $(0, 2\pi)$ and $(0, \infty)$,

$$\begin{aligned}\hat{f}(x, y) &= \int_0^{2\pi} \int_0^{\infty} F(\rho, \theta) \exp(i2\pi\rho(x \cos \theta + y \sin \theta)) \rho d\rho d\theta \\ &= F_{2D}^{-1}\{F(\rho, \theta)\} = f(x, y)\end{aligned}$$

4. Filtered Backprojection Method

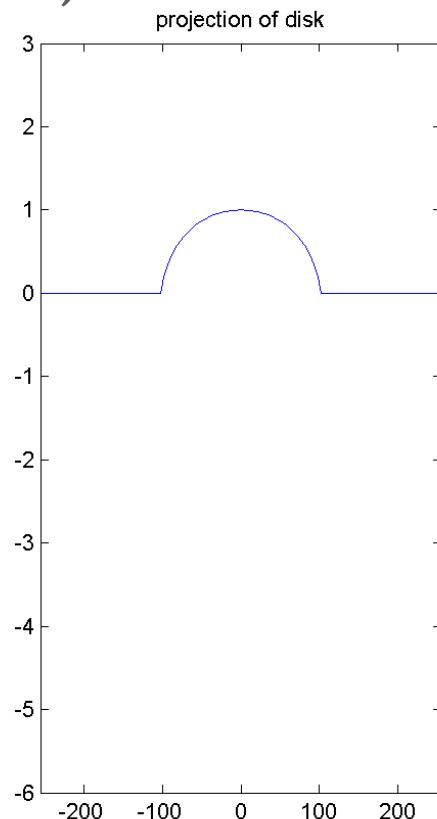


Negative values are necessary
to correct for projections from
other ~~complexe~~ angles

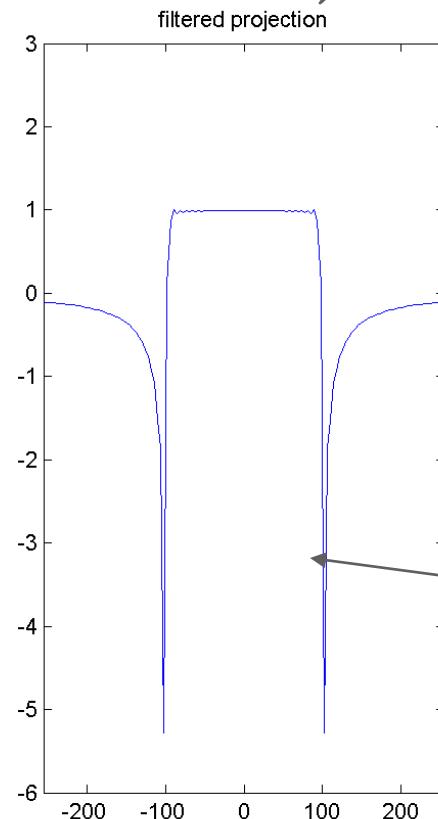
4. Filtered Backprojection Method

One Example – projection of a disk

Projection of disk



Filtered Projection

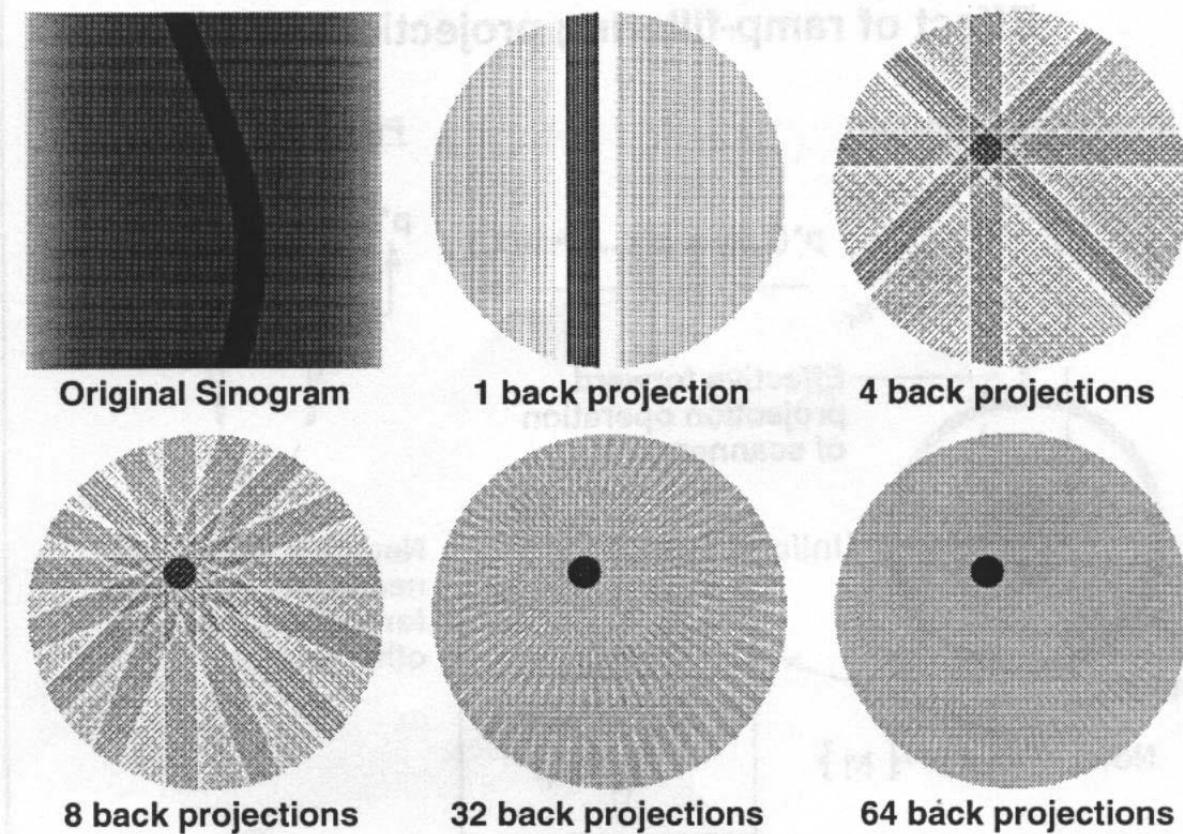


Negative values
necessary to correct
projections from
other angles

4. Filtered Backprojection Method

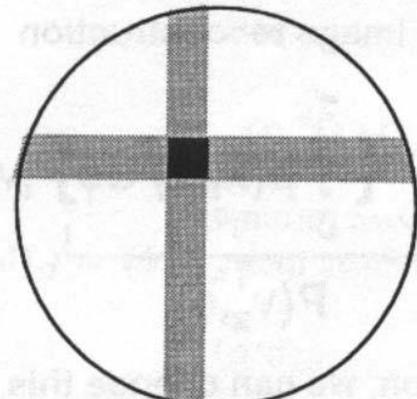
- Example cont. – backprojecting to reconstruct

Filtered Backprojection Example

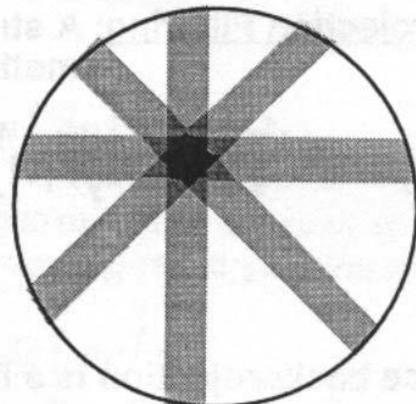


Recall Unfiltered Backprojection

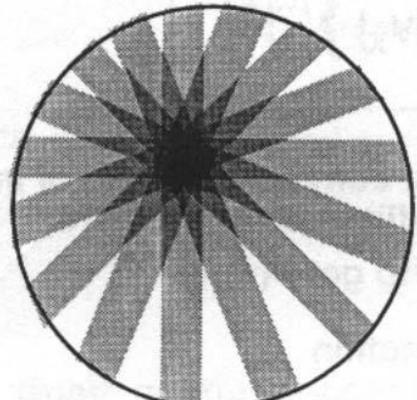
Unfiltered Backprojection Example



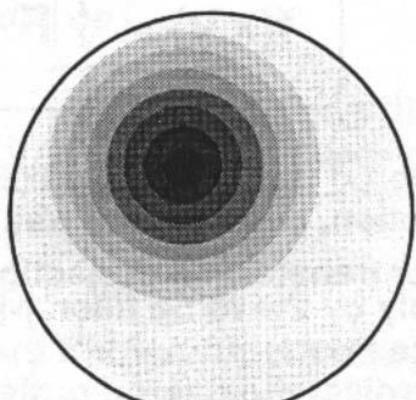
2 Backprojections



4 Backprojections



8 Backprojections



Many Backprojections

4. Filtered Backprojection Method

Steps:

1. Filter the backprojection with a $|\rho|$ filter (ramp filter)

- Fourier method:

$$g'_\theta(R) = F_{1D}^{-1}\{\rho|F_{1D}\{g_\theta(R)\}\} = F_{1D}^{-1}\{\rho|F_{2D}(\rho, \theta)\}$$

- Convolution method:

$$g'_\theta(R) = g_\theta(R) * c(R), \text{ where } c(R) \approx F_{1D}^{-1}\{\rho\}$$

2. Backproject for all angles to get $\hat{f}(x, y)$

4. Filtered Backprojection Method

- Filter Methods:
 - Fourier method – needs two 1D transforms (1D FT and 1D inverse FT) -
 - Convolution Method – one convolution – simpler computation, needs filter $c(R)$

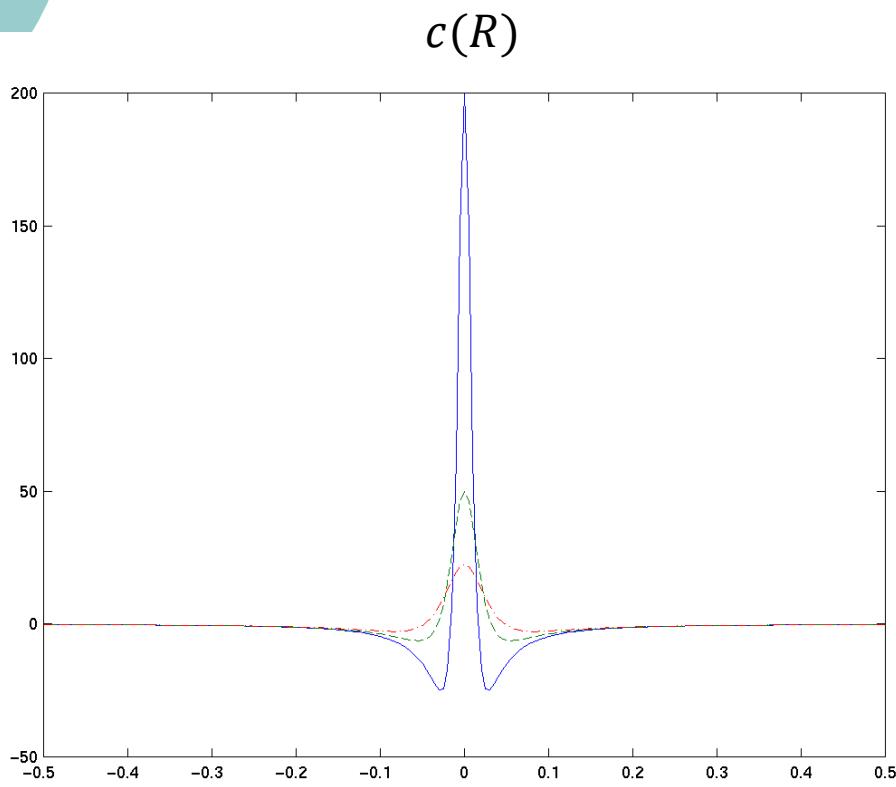
4. Filtered Backprojection Method

- Convolution Filtering Method:
- Convolution filter function $-c(R) = F^{-1}\{\|\rho\|\}$
- Problem: $c(R) = F^{-1}\{\|\rho\|\} = \int_{-\infty}^{\infty} |\rho| \exp(i2\pi R\rho) d\rho$
does not exist (not integrable), goes to ∞

4. Filtered Backprojection Method

- Find FT of a variety of functions that approach $|\rho|$ in the limit
- Filter 1:
$$c(R) = \lim_{\varepsilon \rightarrow 0} F^{-1}\{\rho|\exp(-\varepsilon|\rho|)\}$$
$$= \lim_{\varepsilon \rightarrow 0} \frac{2(\varepsilon^2 - 4\pi^2 R^2)}{(\varepsilon^2 + 4\pi^2 R^2)^2}$$
 - for small R , $c(R)$ will approach $2/\varepsilon^2$
 - for large R , $c(R)$ will approach $-1/2\pi^2 R^2$

4. Filtered Backprojection Method

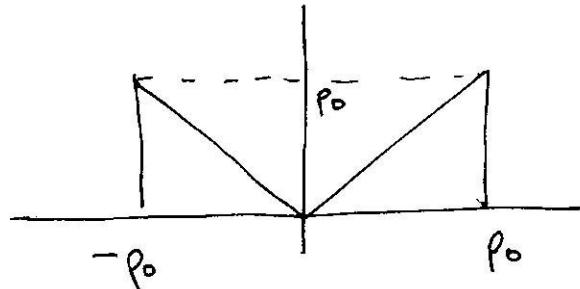


- The function $\exp(-\varepsilon|\rho|)$ cuts off the high frequency parts of $|\rho|$ - cutoff frequency of ρ_0

4. Filtered Backprojection Method

- Filter 2: rect function

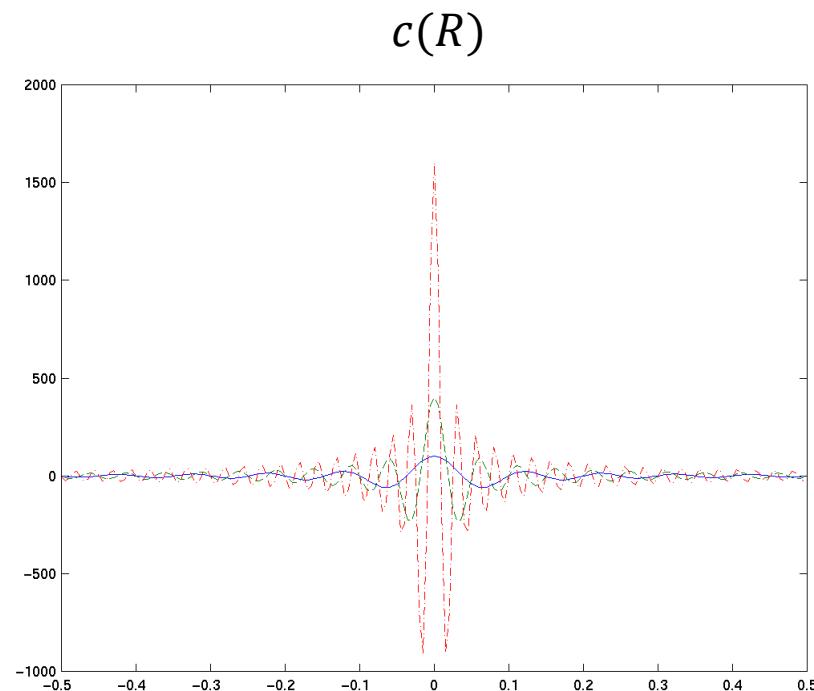
$$\begin{aligned} C(\rho) &= |\rho| \text{rect}\left(\frac{\rho}{2\rho_0}\right) \\ &= \rho_0 \left[\text{rect}\left(\frac{\rho}{2\rho_0}\right) - \text{tri}\left(\frac{\rho}{\rho_0}\right) \right] \end{aligned}$$



$$c(R) = \rho_0^2 \left(2 \text{sinc}(2\rho_0 R) - \text{sinc}^2(\rho_0 R) \right)$$

4. Filtered Backprojection Method

- Has substantial ringing effect

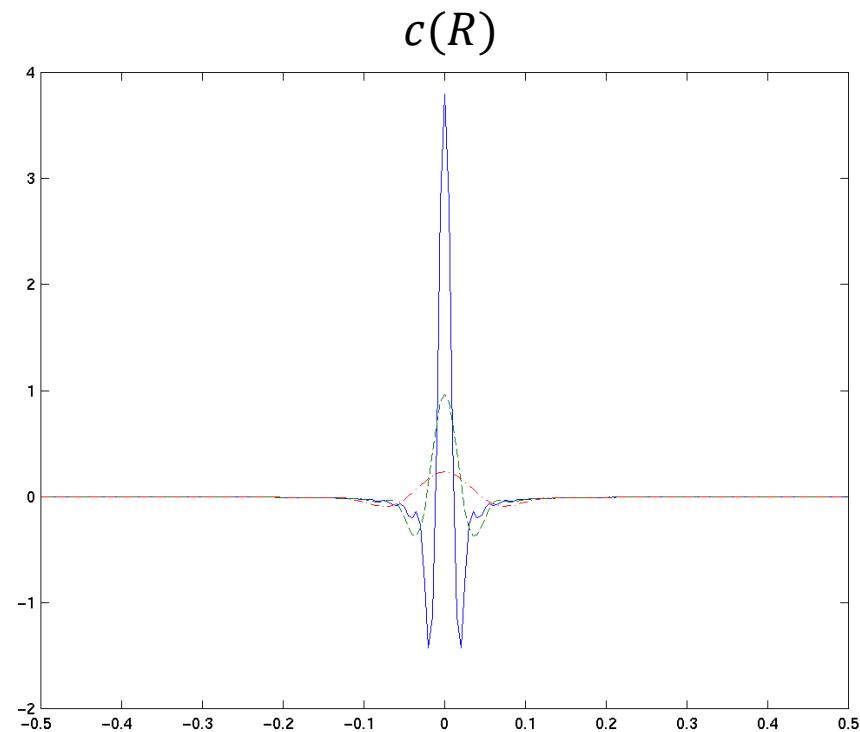


4. Filtered Backprojection Method

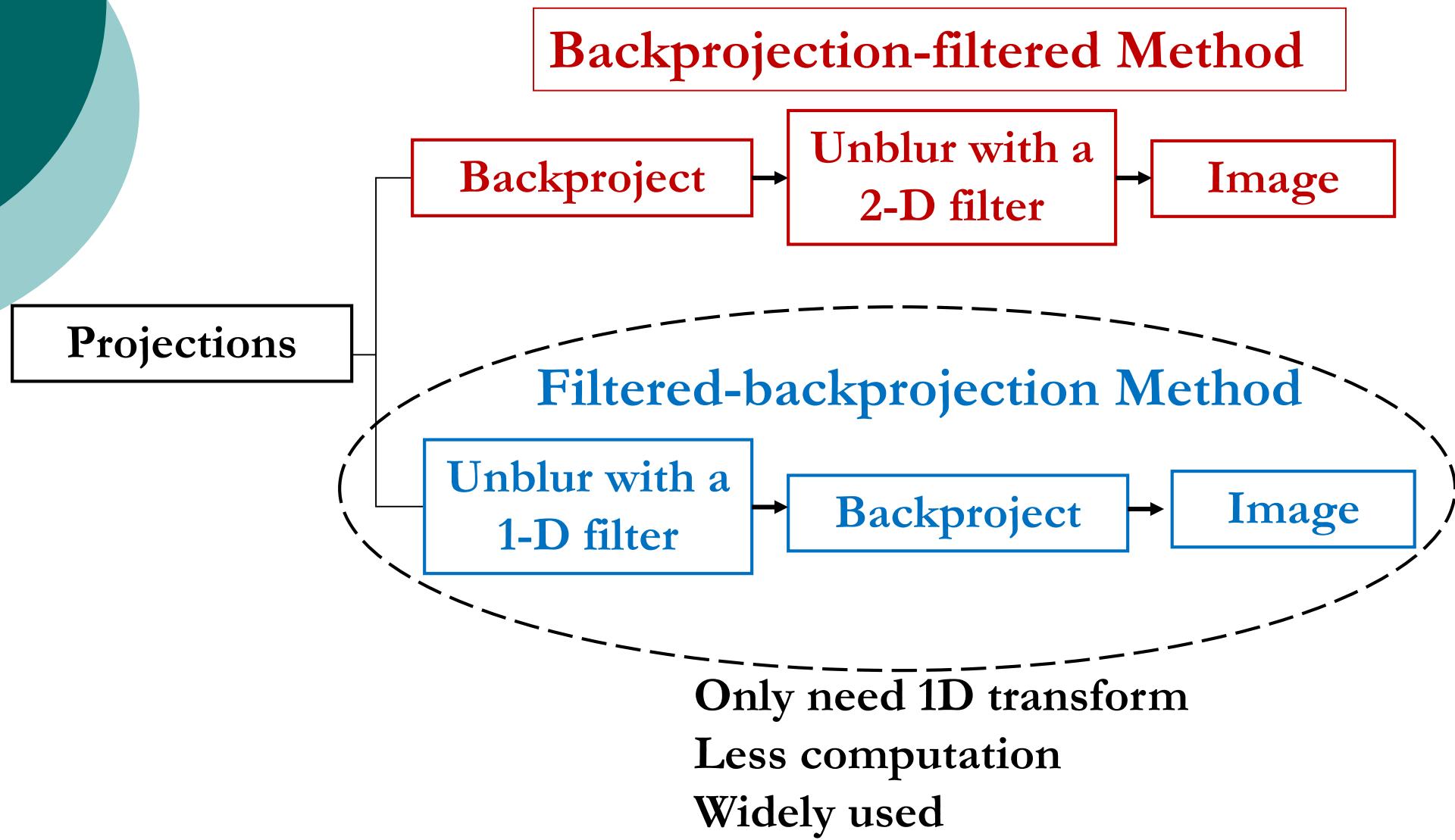
- Filter 3: Gaussian or Hanning filter (suppress ringing)

$$C(r) = |r| \exp \left(-\rho \frac{r^2}{r_0^2} \right)$$

$$C(r) = |r| \text{rect} \left(\frac{r}{2r_0} \right)^{\frac{1}{2}} + \frac{1}{2} \cos \left(\rho \frac{r}{r_0} \right)$$



Comparison between Two Backprojection Methods





Questions?

Filtered-backprojection Method

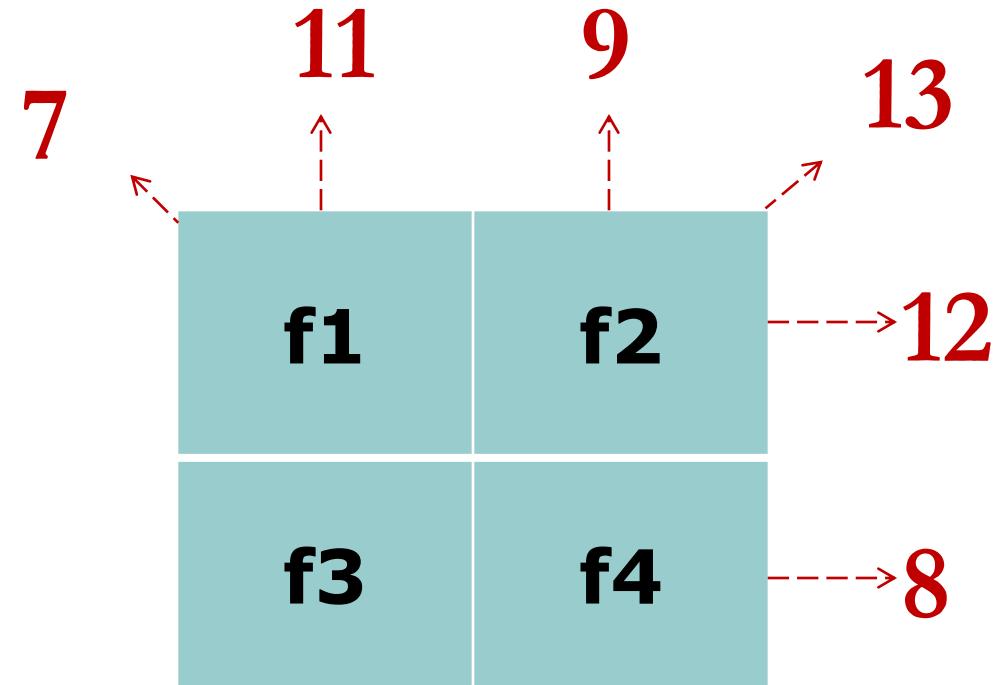
5. Algebraic Reconstruction Technique (ART)

- Like regular backprojection, but uses iterative corrections
- Straightforward, although computationally inefficient solution involves linear algebra
- Used initially, relatively old and now seldomly used now

5. Algebraic Reconstruction Technique (ART)

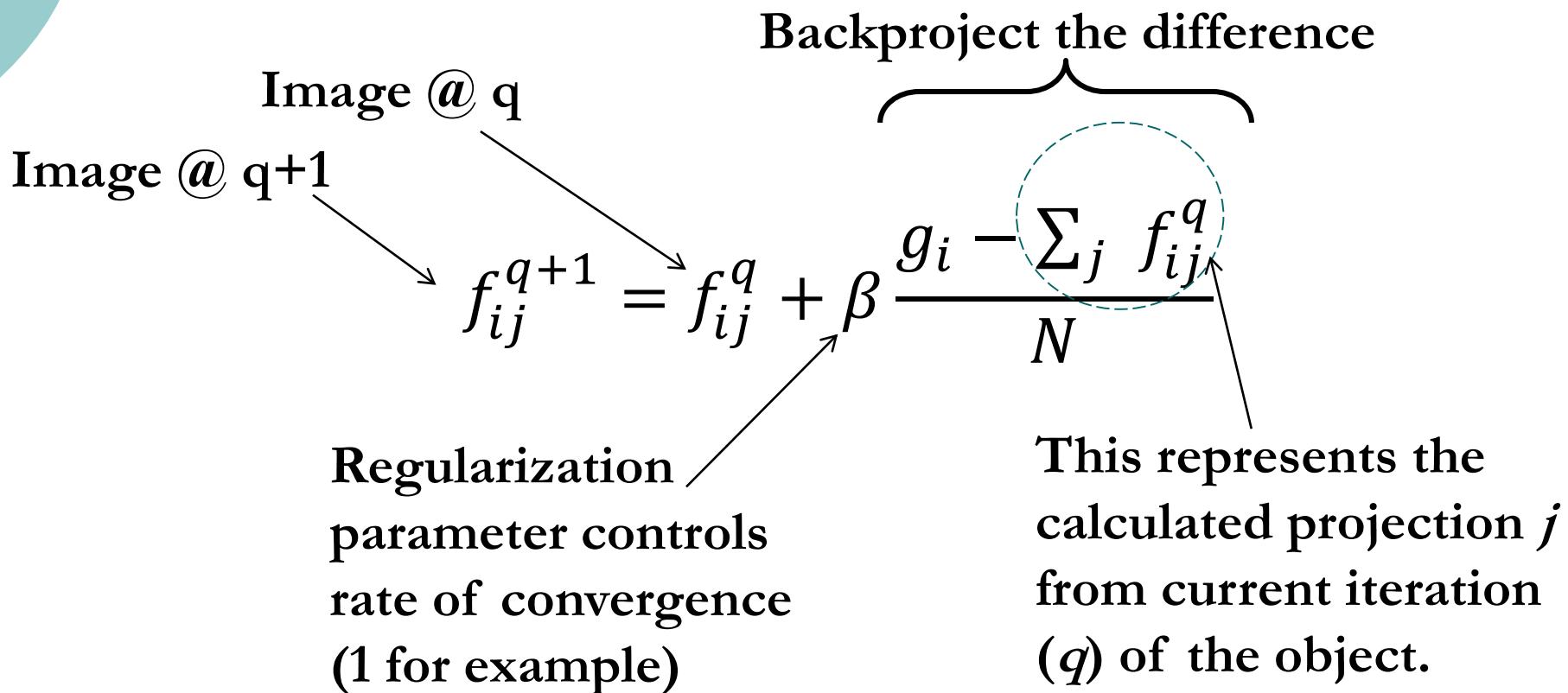
- Example

Object and projection



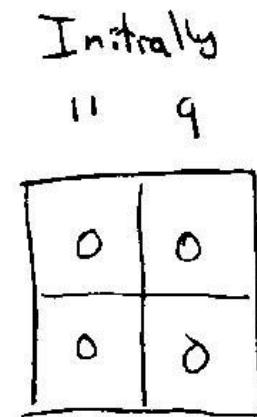
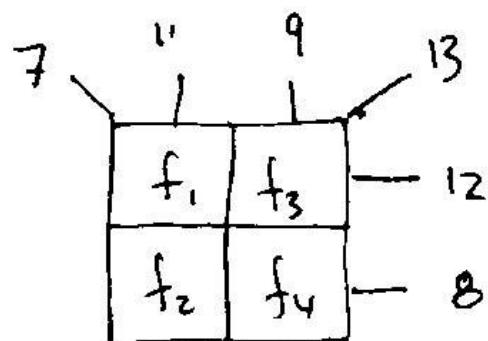
5. Algebraic Reconstruction Technique (ART)

- g_i be the measured projections and f_{ij}^q be the image at iteration q



5. Algebraic Reconstruction Technique (ART)

- $q = 1$, Looking at the projections from top to bottom



$$f_1^1 = 0 + \frac{11 - 0}{2} = 5.5; \quad f_3^1 = 0 + \frac{9 - 0}{2} = 4.5$$

$$f_2^1 = 0 + \frac{11 - 0}{2} = 5.5; \quad f_4^1 = 0 + \frac{9 - 0}{2} = 4.5$$

5. Algebraic Reconstruction Technique (ART)

- $q = 2$, looking at the left-right projections:

5.5	4.5	- 12
5.5	4.5	- 8

$$f_1^2 = 5.5 + \frac{12 - (5.5 + 4.5)}{2} = 6.5;$$

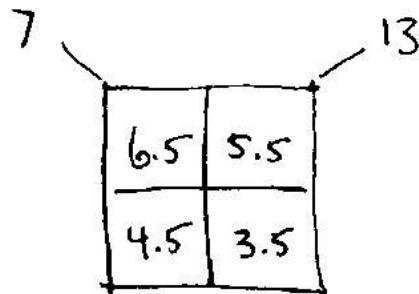
$$f_2^2 = 5.5 + \frac{8 - (5.5 + 4.5)}{2} = 4.5;$$

$$f_3^2 = 4.5 + \frac{12 - (5.5 + 4.5)}{2} = 5.5$$

$$f_4^2 = 4.5 + \frac{8 - (5.5 + 4.5)}{2} = 3.5$$

5. Algebraic Reconstruction Technique (ART)

- $q = 3$, looking at the diagonal projections



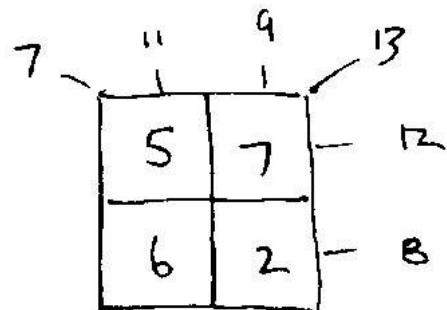
$$f_1^3 = 6.5 + \frac{7 - (6.5 + 3.5)}{2} = 5; \quad f_3^3 = 5.5 + \frac{13 - (5.5 + 4.5)}{2} = 7$$

$$f_2^3 = 4.5 + \frac{13 - (5.5 + 4.5)}{2} = 6; \quad f_4^3 = 3.5 + \frac{7 - (6.5 + 3.5)}{2} = 2$$

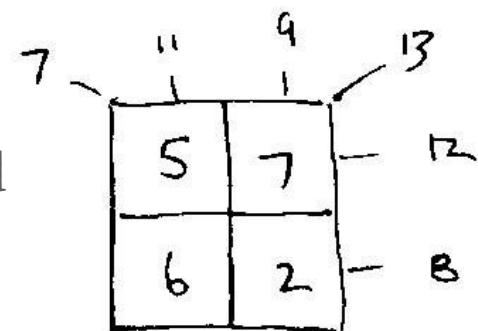
5. Algebraic Reconstruction Technique (ART)

○ Final Solution

Final
Sol.



Object and
projection



5. Algebraic Reconstruction Technique (ART)

- Drawbacks
- In the presence of noise (that is, the projections aren't exactly consistent), this method converges very slowly, and in some cases, not at all.
- Usually, one goes through the entire set of projections multiple times.



Questions?

Algebraic Reconstruction Technique
(ART)

Reconstruction Methods

- 1. Direct Fourier Interpolation Method – CT project**
- 2. Backprojection-Filtering Method – Introduce backprojection concept**
- 3. Direct Fourier Superposition and Filtering Method**
- 4. Filtered Backprojection Method – CT project**
- 5. Algebraic Reconstruction Technique (ART) – conventional method**

Image Recon

- Matlab Demos



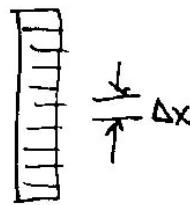
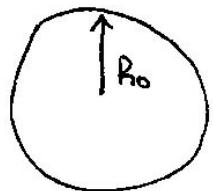
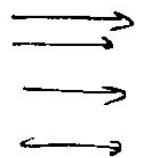
Practical considerations

Sampling

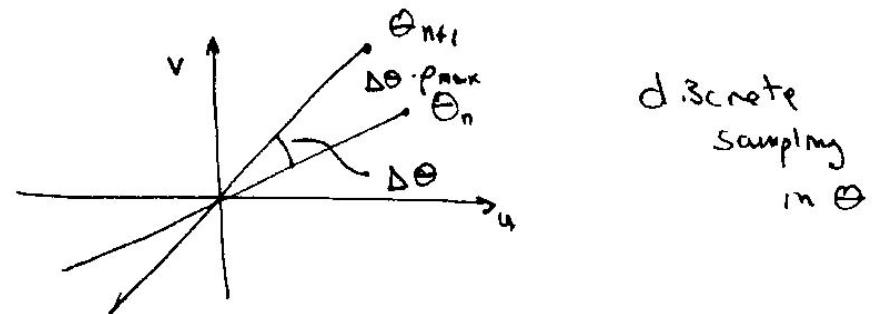
Fan-beam geometric

Circular convolution

Sampling in R and θ



discrete sampling
in R



- Usually we think about sampling θ finely enough for the information content in each projection

Sampling in R and θ

- Recall the minimum required sampling distance is given by FOV:

$$\Delta k_{\max} \leq \frac{1}{FOV} = \frac{1}{2R_0}$$

- And the maximum extent in k-space is given by the spatial resolution:

$$\rho_{\max} = \frac{1}{2\Delta x}$$

Sampling in R and θ

- Recognizing the arc length in k -space between radial lines is approx:

$$\Delta k_{\max} = \Delta\theta \cdot \rho_{\max}$$

- Then we can set the critical angular sampling as:

$$\Delta\theta \leq \frac{\Delta x}{R_0}$$

Number of angles

- Number of projection angles:

$$N_{proj} = \frac{\theta_{\max} - \theta_{\min}}{\Delta\theta}$$

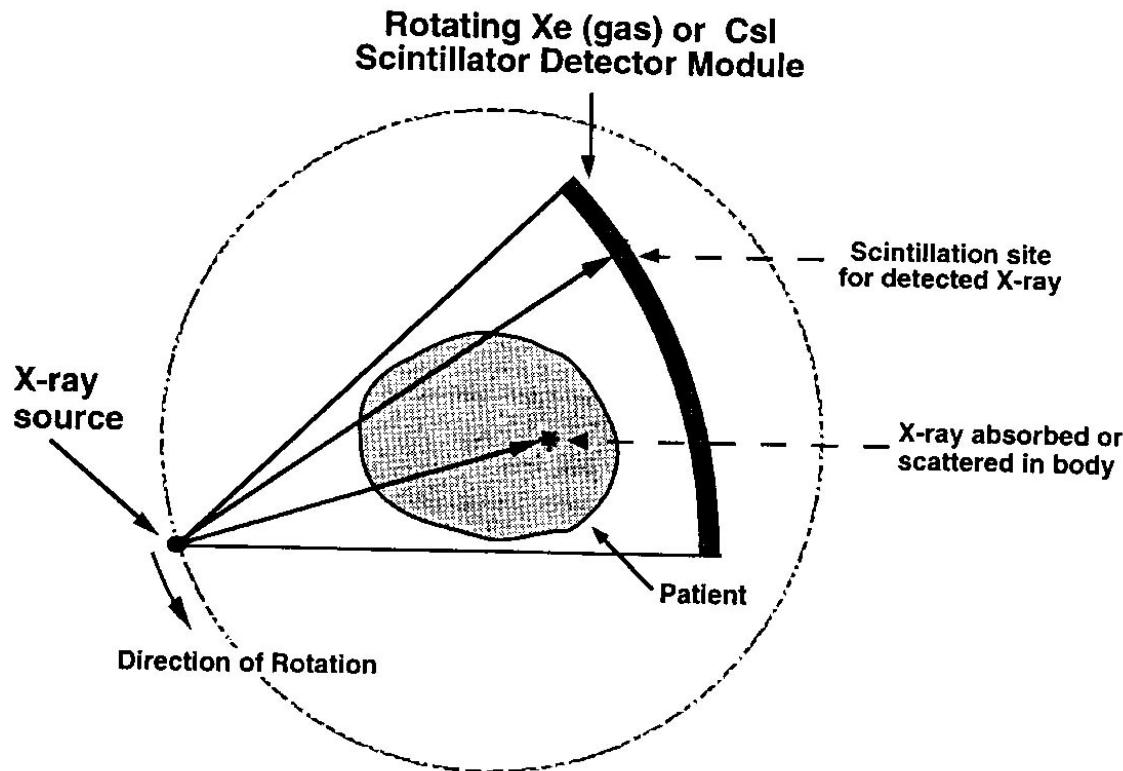
- Which for 512 samples on each projection over π radians is:

$$\frac{2R_0}{\Delta x} = 512 \quad \Delta\theta \leq \frac{1}{256} \quad \text{and then}$$

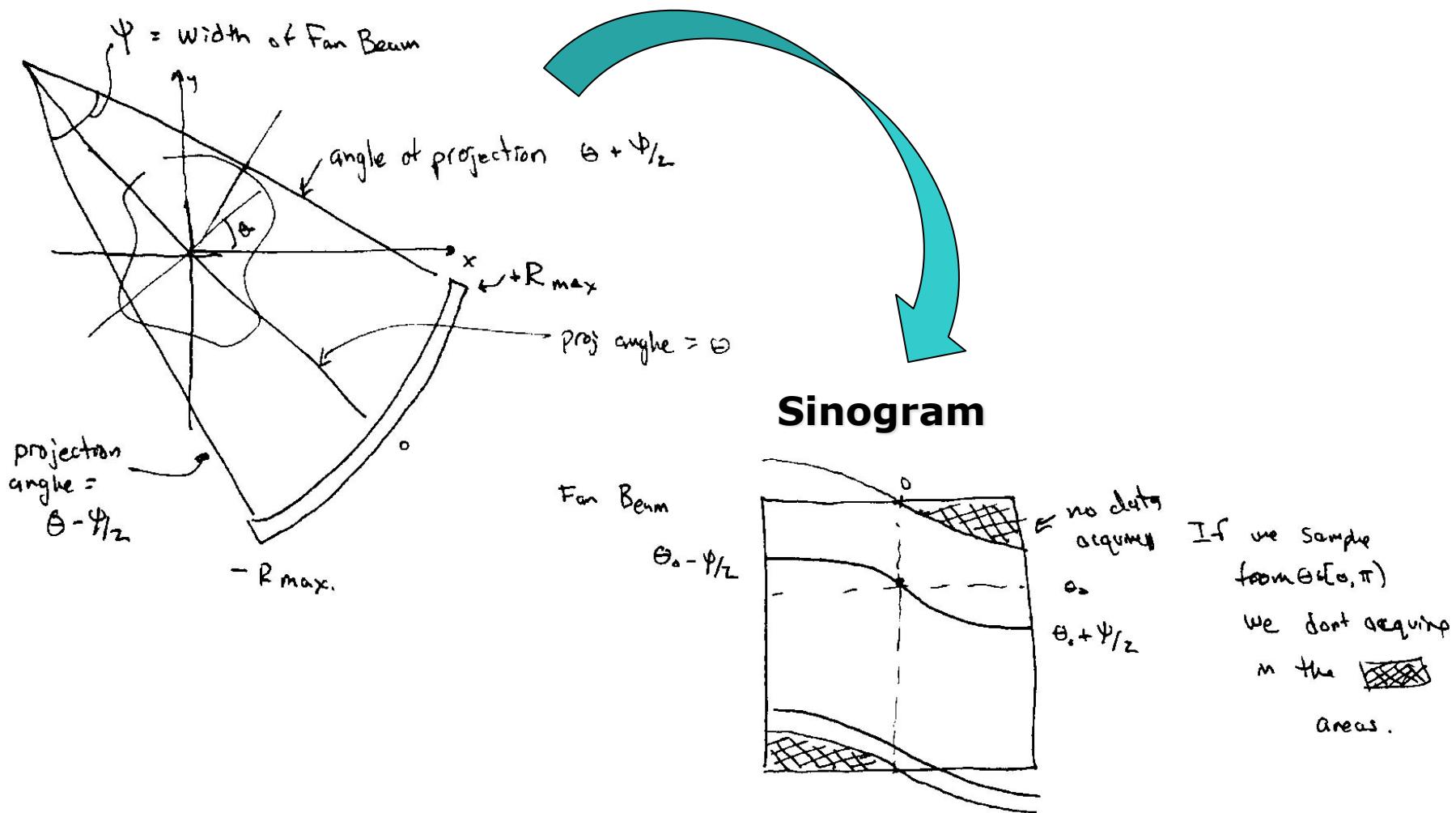
$$N_{proj} = \frac{\theta_{\max} - \theta_{\min}}{\Delta\theta} = 256\pi \approx 804$$

Fan Beam Geometry

- Beams are no longer parallel

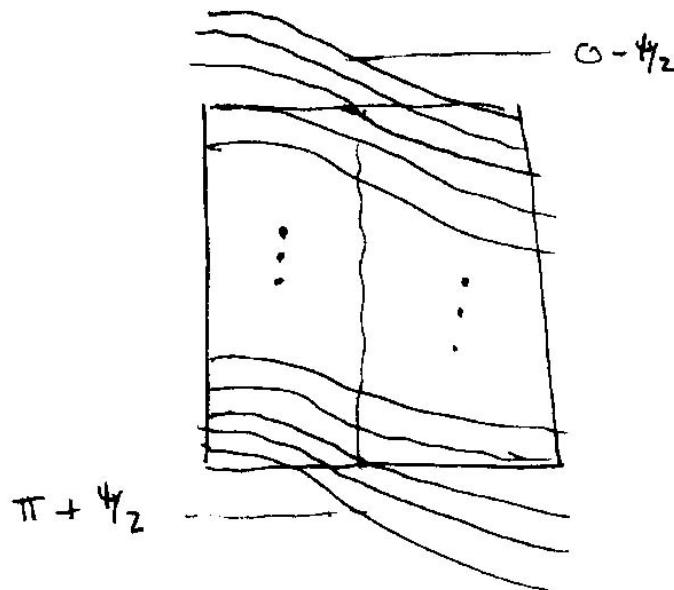


Fan Beam Geometry



Fan Beam Geometry

- Need to acquire slightly more than π radians to fill sinogram space
- Need $\theta \in \left[-\frac{\psi}{2}, \pi + \frac{\psi}{2}\right]$ or $\pi + \psi$ radians



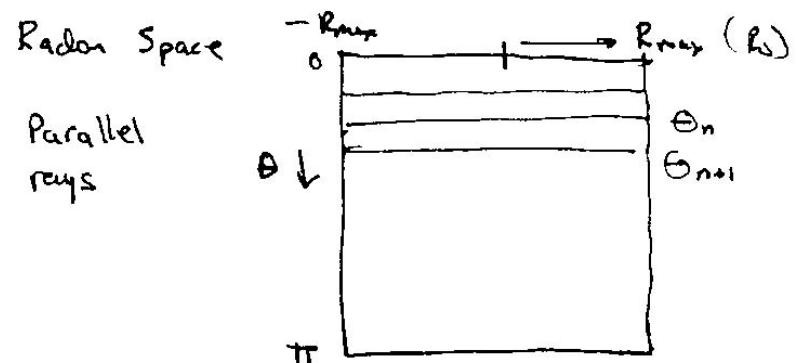
For full Data

we need

$$\theta \in \left[-\frac{\psi}{2}, \pi + \frac{\psi}{2}\right]$$

Reconstruction of Fan Beam Data

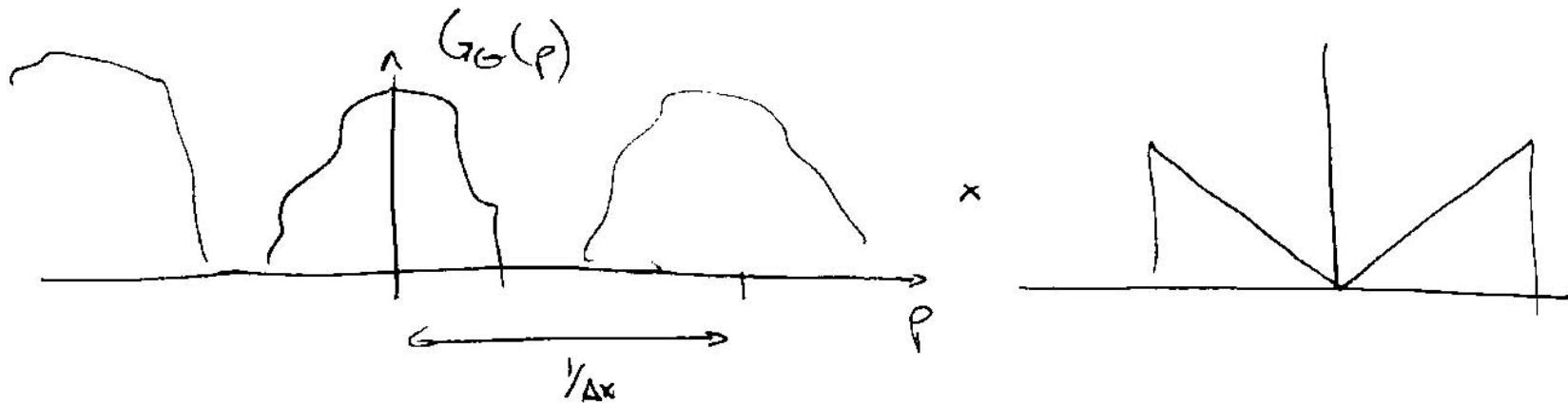
- Interpolate to parallel ray sinogram and then use any method including Fourier interpolation



- Other methods
 - Fan beam filtered backprojection
 - Fan beam ART

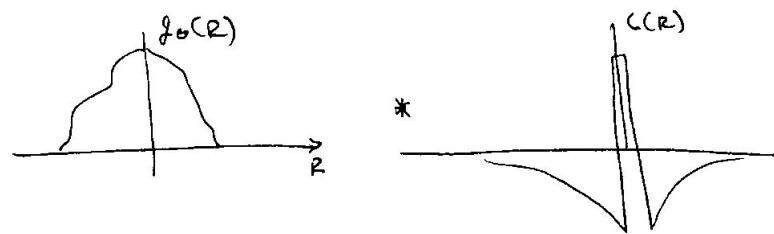
Circular convolution

- When filtering in the Fourier domain for filtered backprojection, one needs to worry about circular convolution

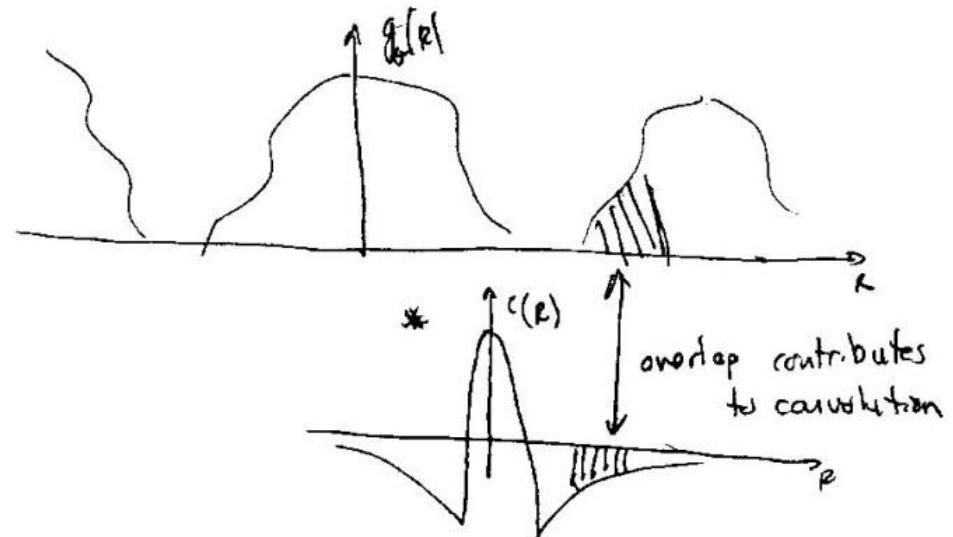


Circular convolution

- In the image domain instead of this:

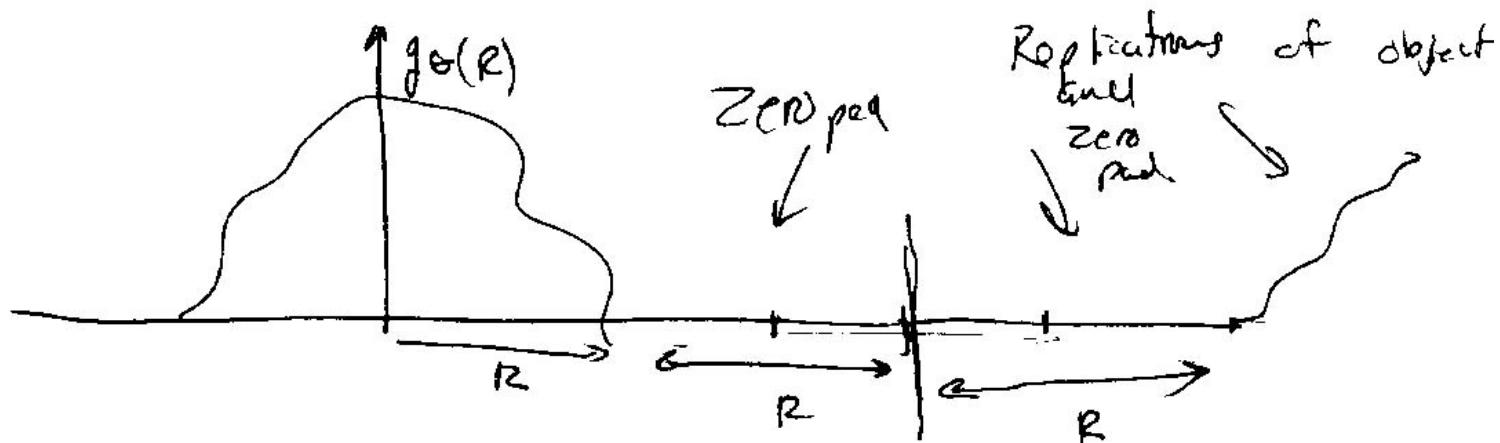


- We get this:



Circular convolution

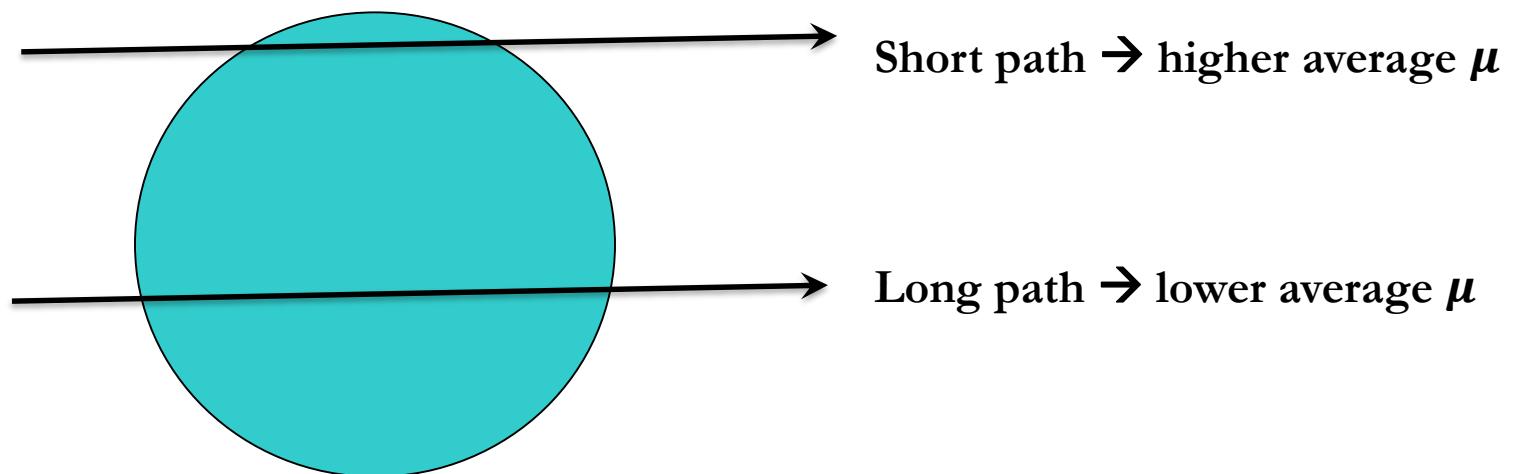
- Solution: zero pad before taking 1D FT:



which pushes the replicates out farther, minimizing these effects

Beam Hardening

- We talked about beam hardening making deeper structure appear to have a lower μ than expected.



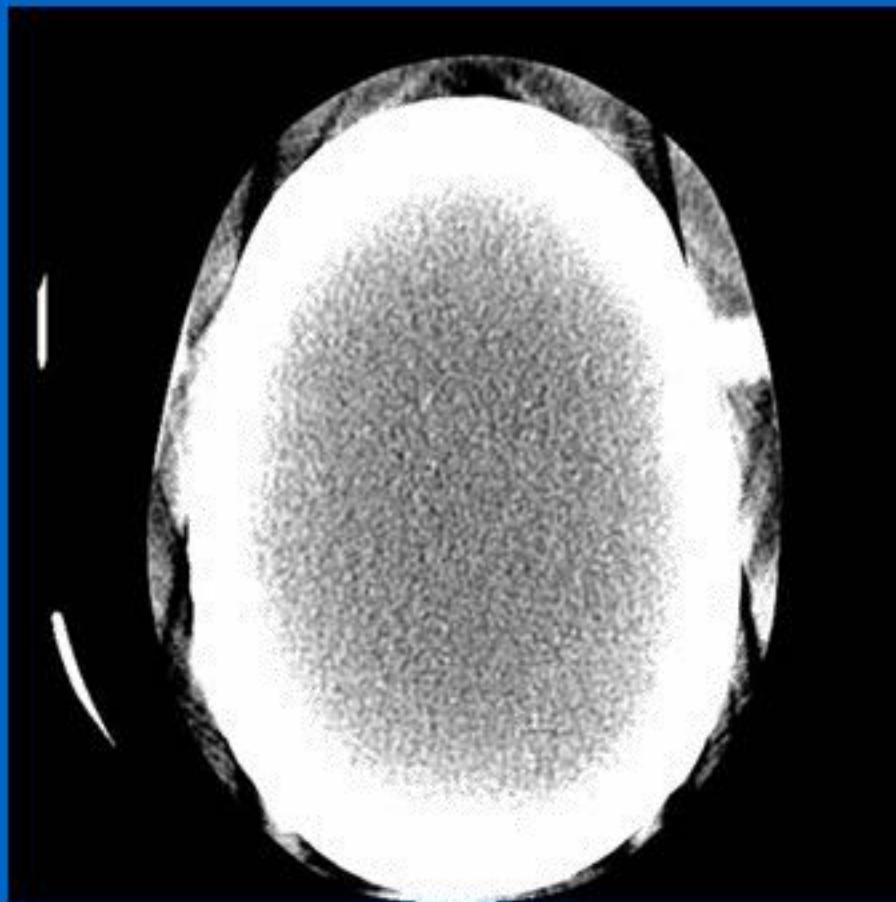
Beam Hardening

- Results in apparent lower μ in center of objects (cupping)

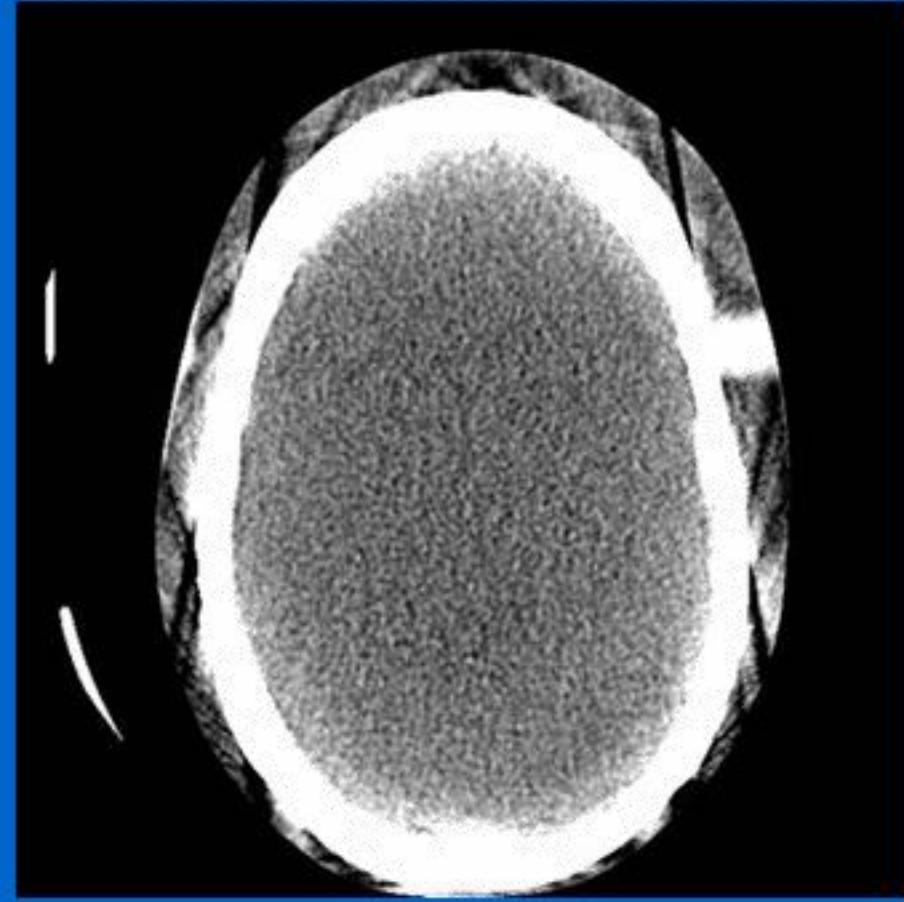


- Also shadows around implants

Beam hardening correction software



Raw image



After beam hardening
correction

Beam Hardening Artifact

