

ROI Reconstruction in CT

From Tomo reconstruction in the 21st century, IEEE Sig. Proc. Magazine (R.Clackdoyle M.Deprise)

L. Desbat

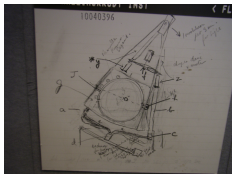
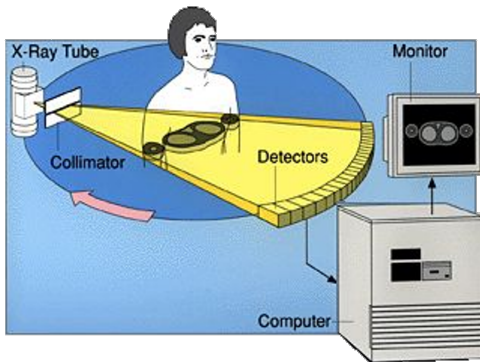
TIMC-IMAG

September 10, 2013

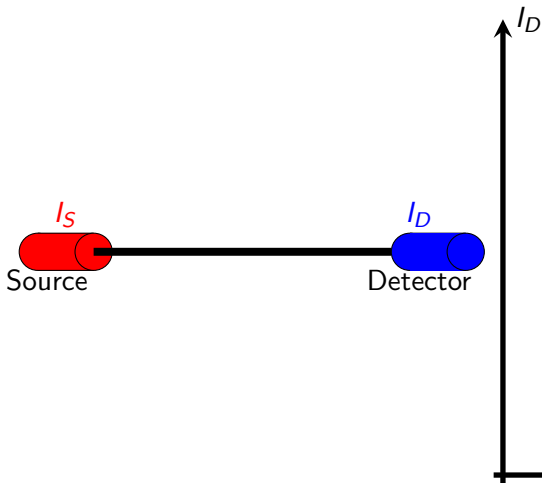
Outline

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 - CT
 - Radiology
- 2 Radon transform and its inversion
 - Radon Transform
 - Radon and Fourier
 - Non Local Inversion
 - Fan Beam
- 3 ROI reconstruction
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 - parallel Fan Beam Hilbert Projection Equality
 - DBP-H Inversion

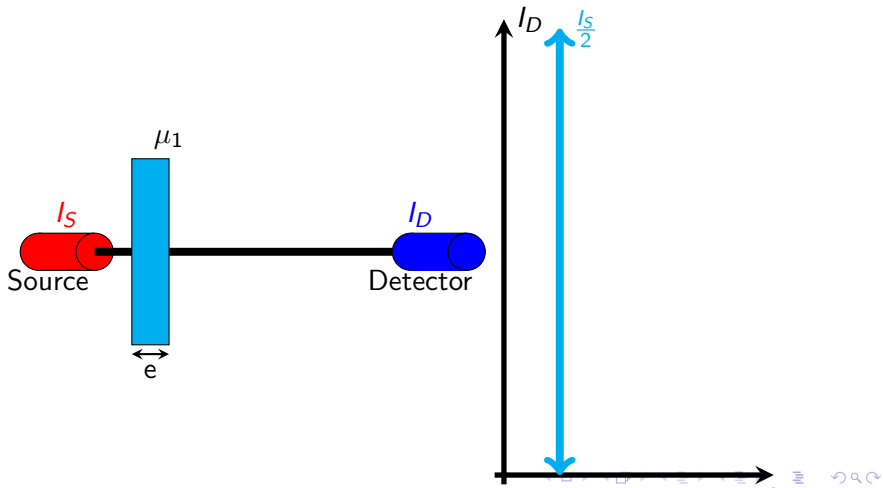
CT scanner: principle



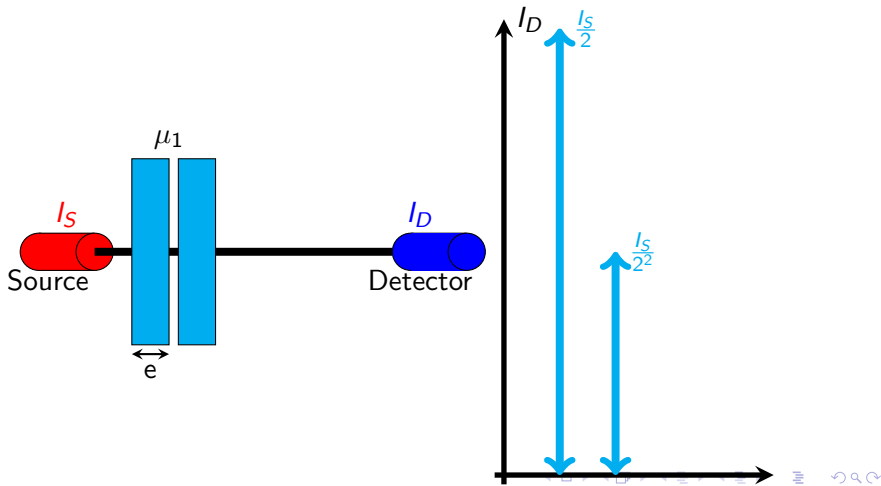
X-ray attenuation



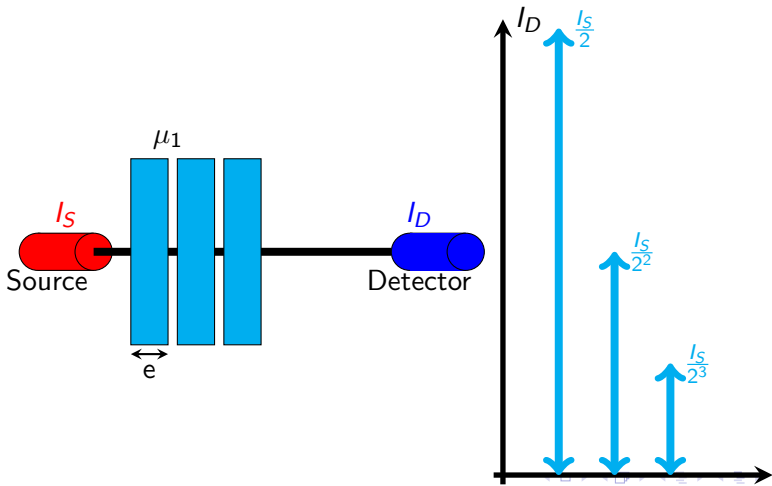
X-ray attenuation



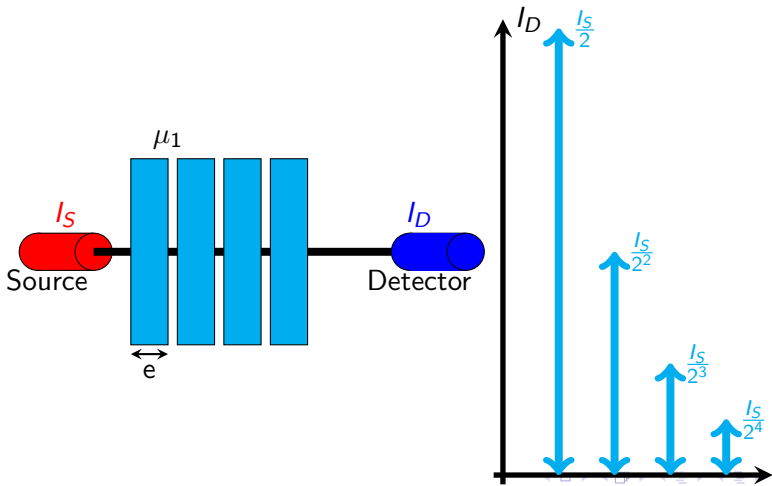
X-ray attenuation



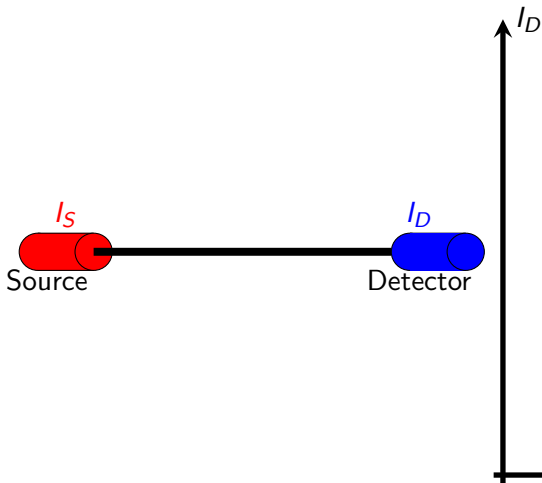
X-ray attenuation



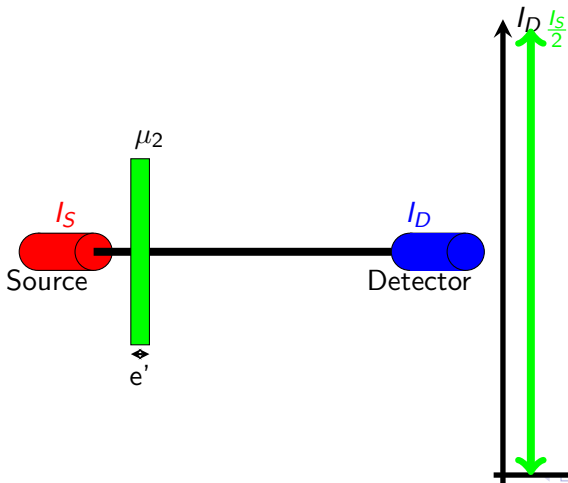
X-ray attenuation



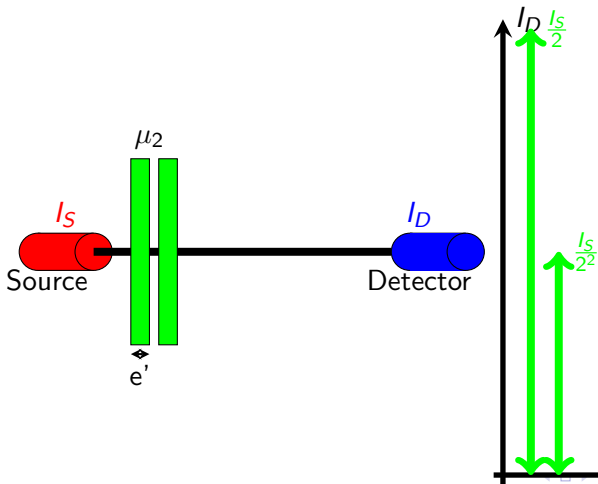
X-ray attenuation



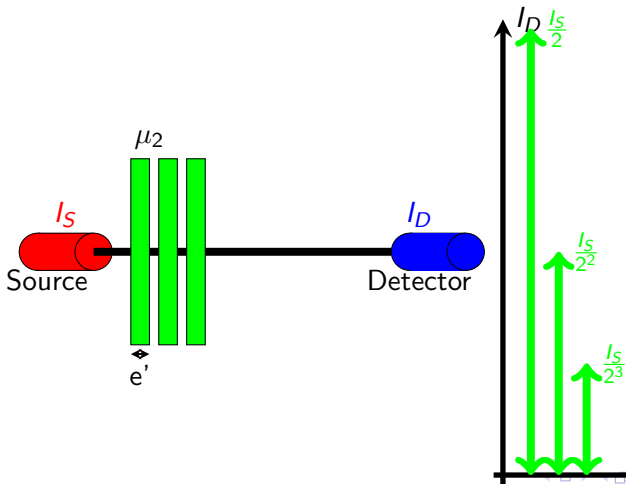
X-ray attenuation



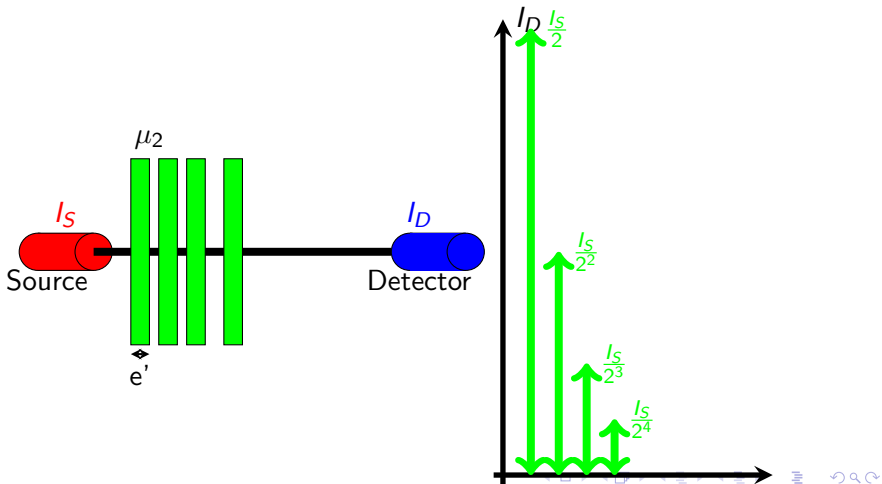
X-ray attenuation



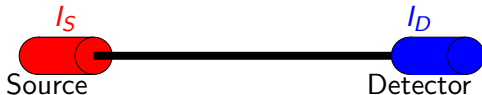
X-ray attenuation



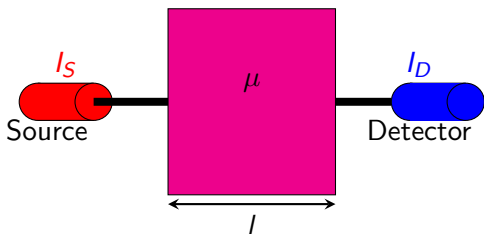
X-ray attenuation



X-ray attenuation



X-ray attenuation



Lambert-Beer's Law

$$I_D(x) = I_S e^{-\mu l}$$

I_S : intensité initiale

μ : coefficient d'atténuation

l : épaisseur du tissu

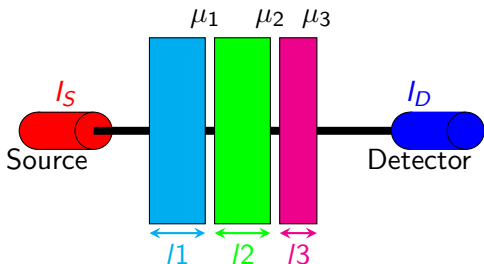
X-ray attenuation

Lambert-Beer's Law

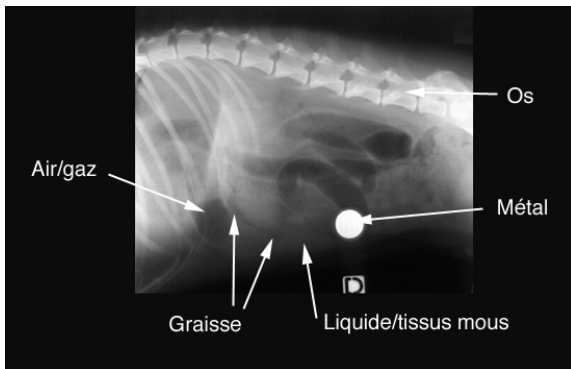
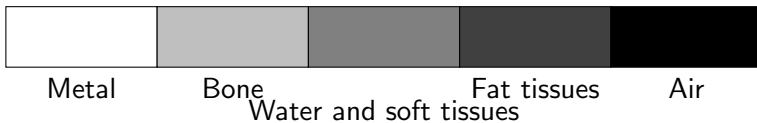
$$I_D = I_S e^{-(\mu_1 l_1 + \mu_2 l_2 + \mu_3 l_3)}$$

$$I_D = I_S e^{-\sum_i \mu_i d l_i}$$

$$I_D = I_S e^{-\int_S^D \mu(l) dl}$$



x-ray image formation



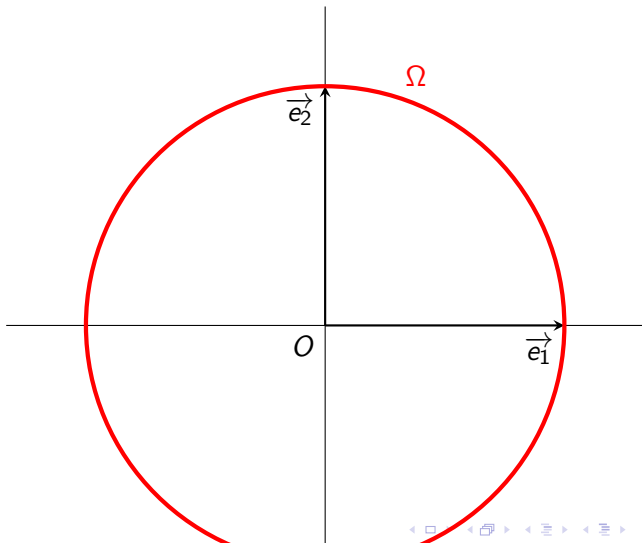
From Radiology to Tomography

- Radon Transform

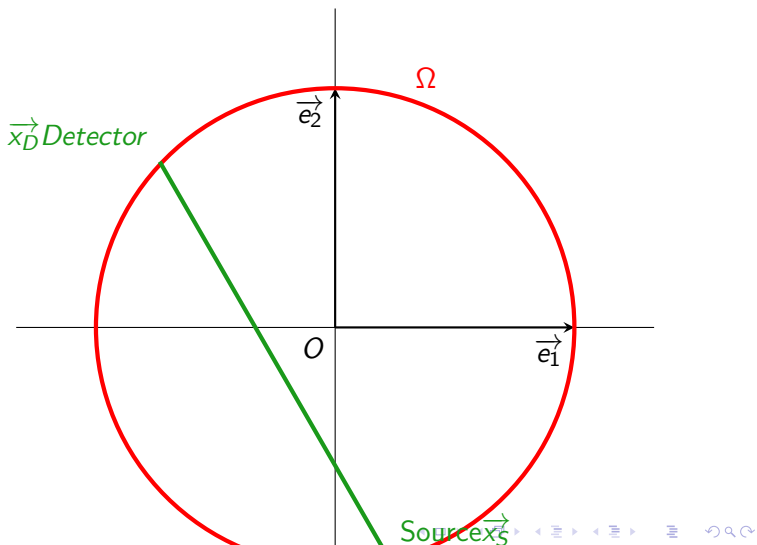
$$I_D = I_S e^{-\int_S^D \mu(l) dl}$$
$$-\ln\left(\frac{I_D}{I_S}\right) = \int_S^D \mu(l) dl$$

- With a CT we measure the integral of μ over lines

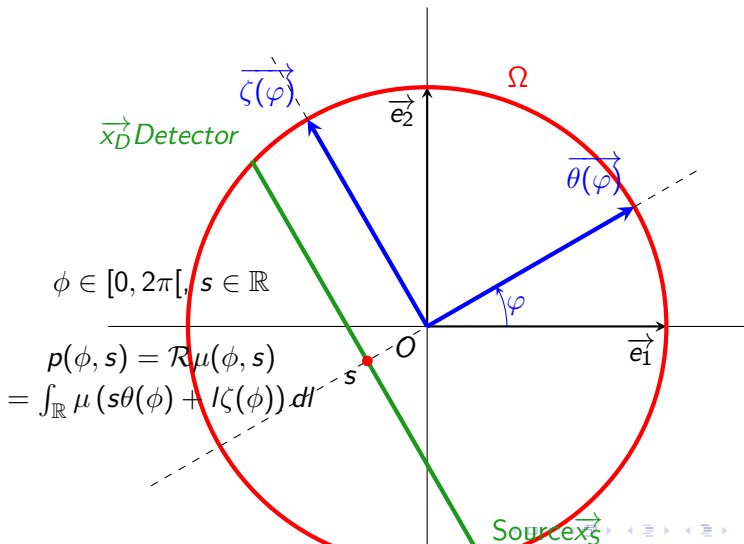
Radon Transform parametrization



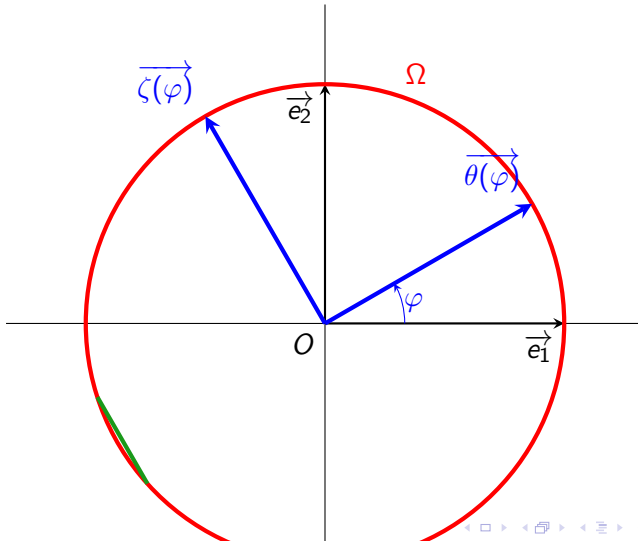
Radon Transform parametrization



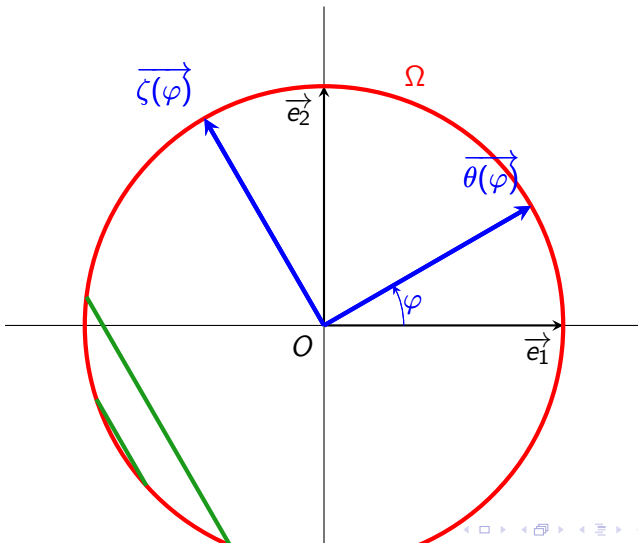
Radon Transform parametrization



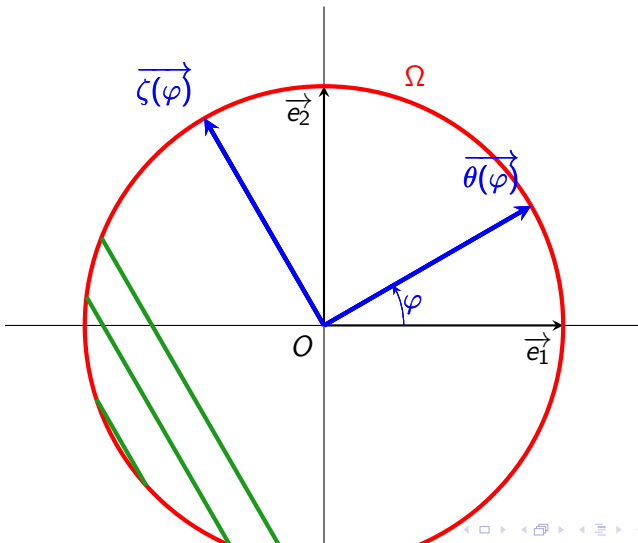
Radon Transform



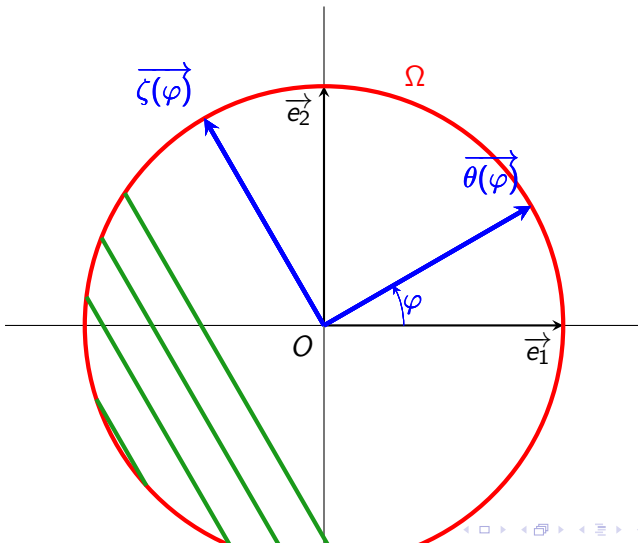
Radon Transform



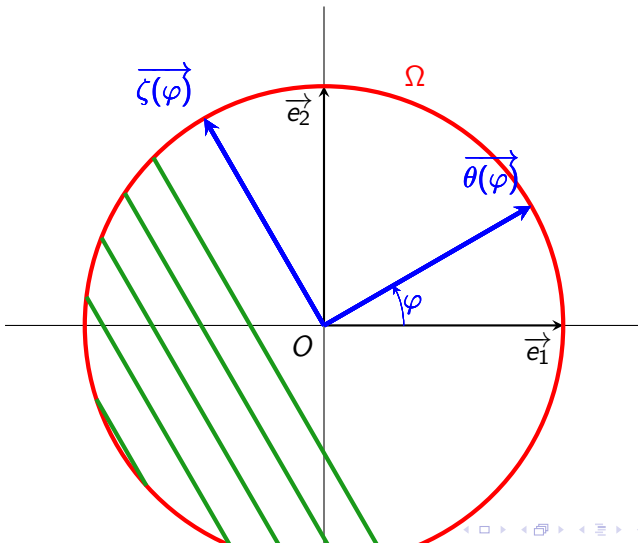
Radon Transform



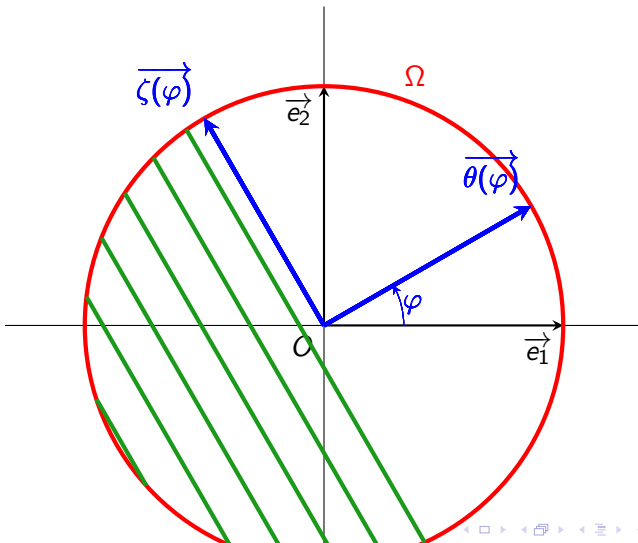
Radon Transform



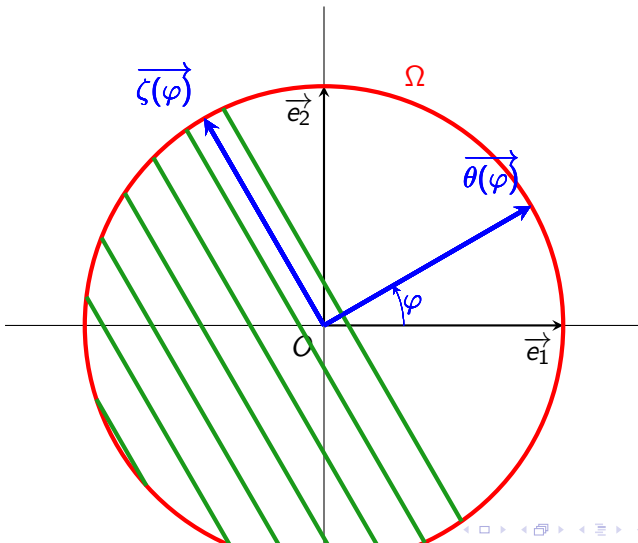
Radon Transform



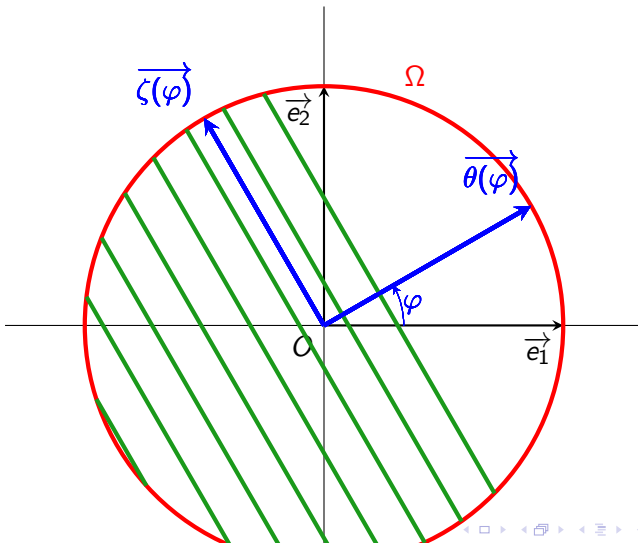
Radon Transform



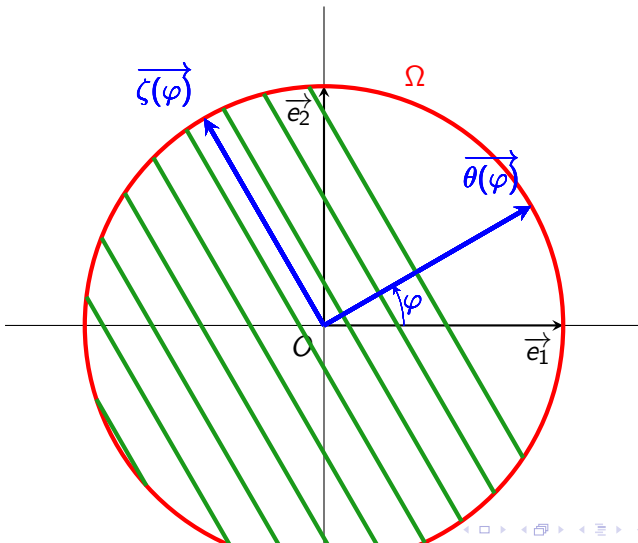
Radon Transform



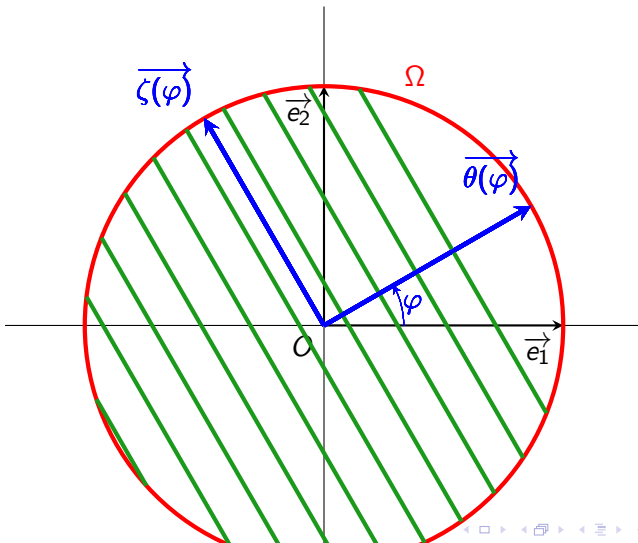
Radon Transform



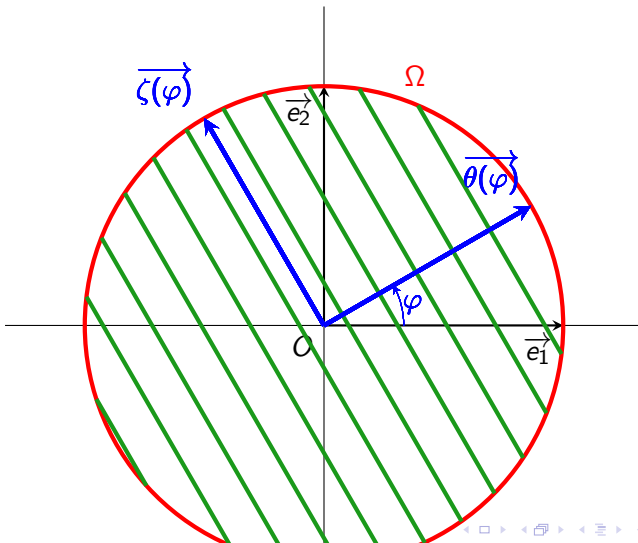
Radon Transform



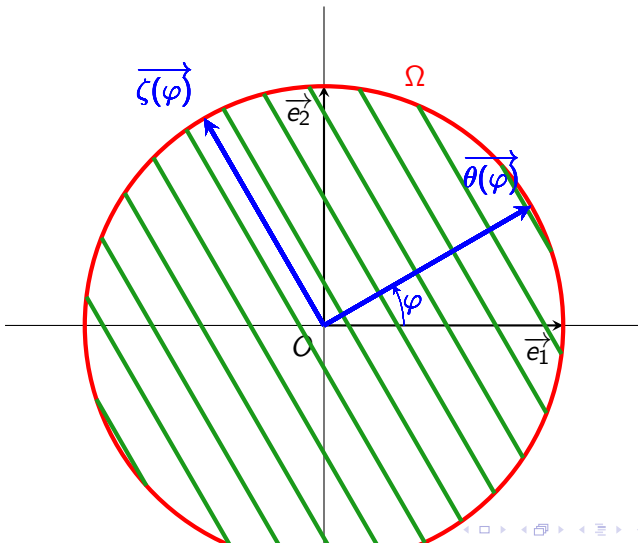
Radon Transform



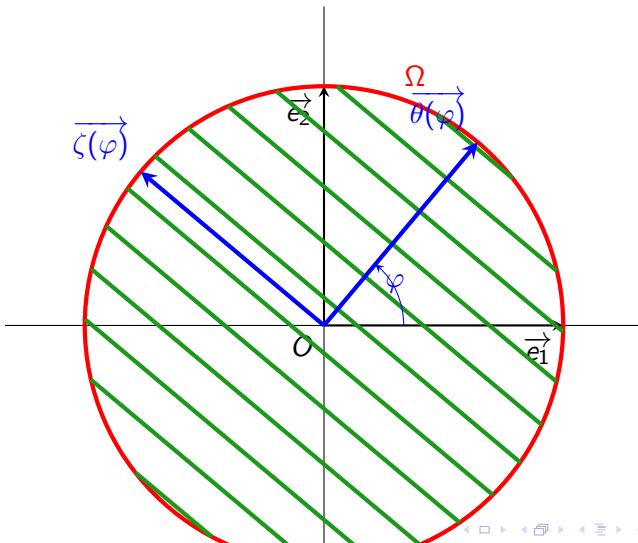
Radon Transform



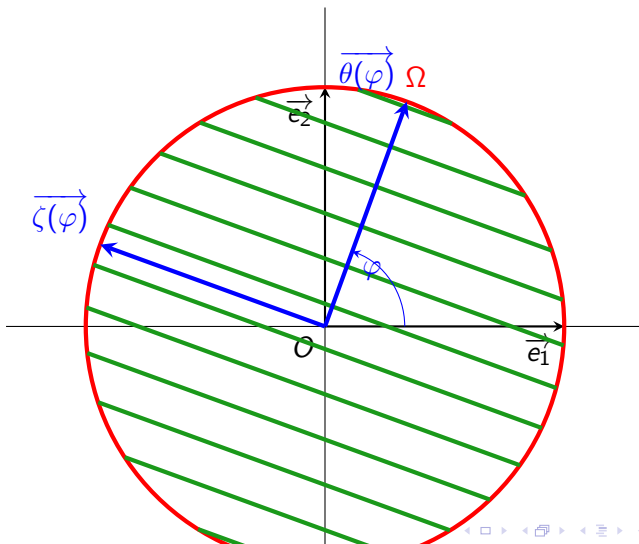
Radon Transform



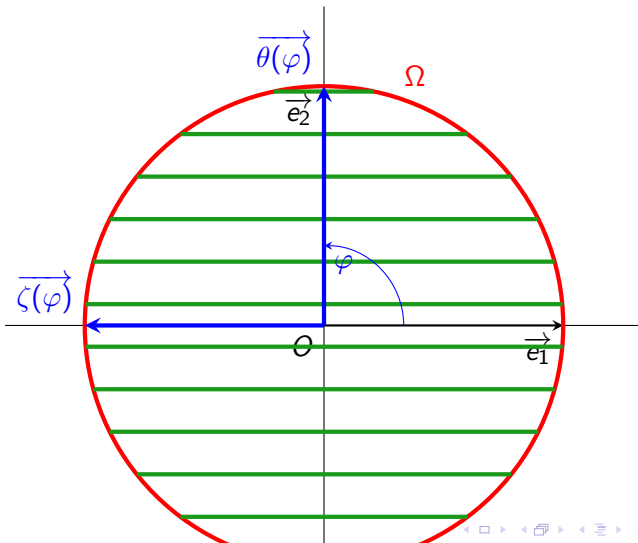
Radon Transform



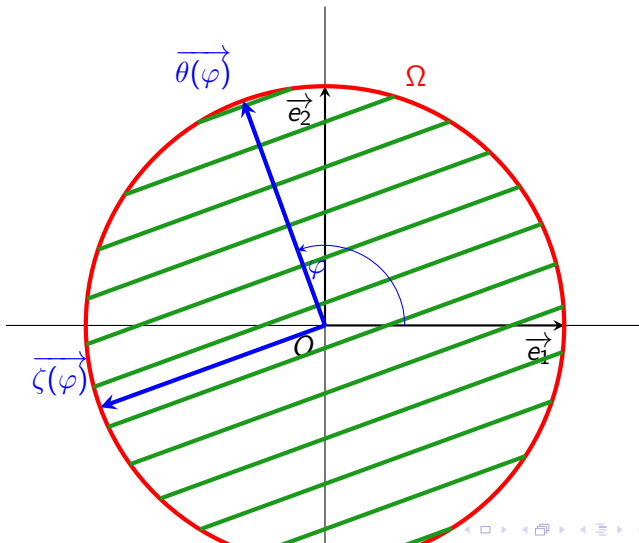
Radon Transform



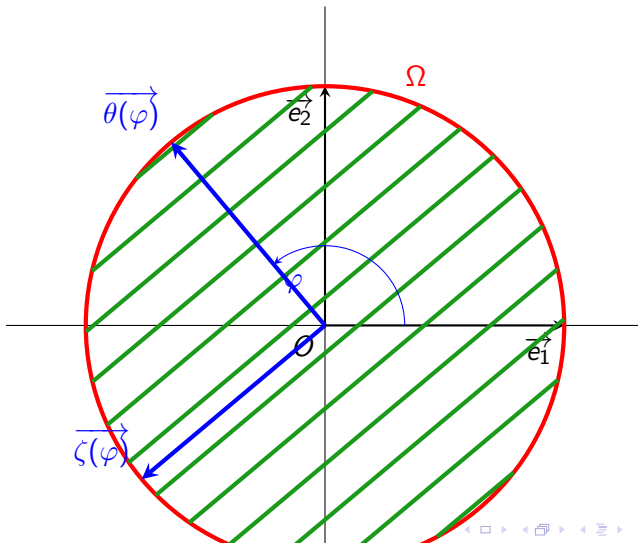
Radon Transform



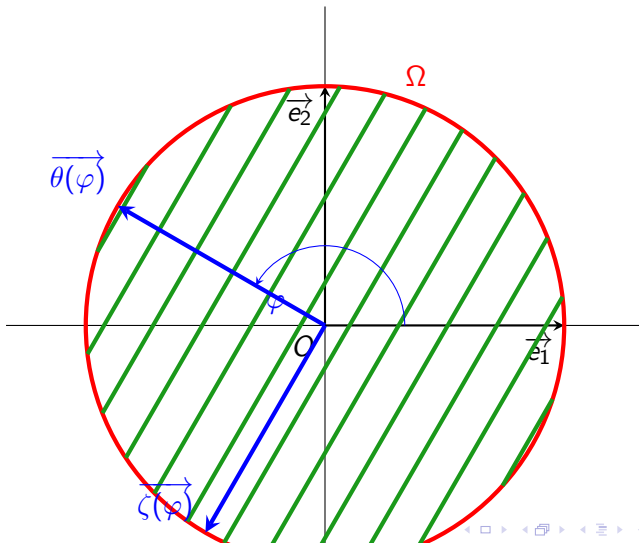
Radon Transform



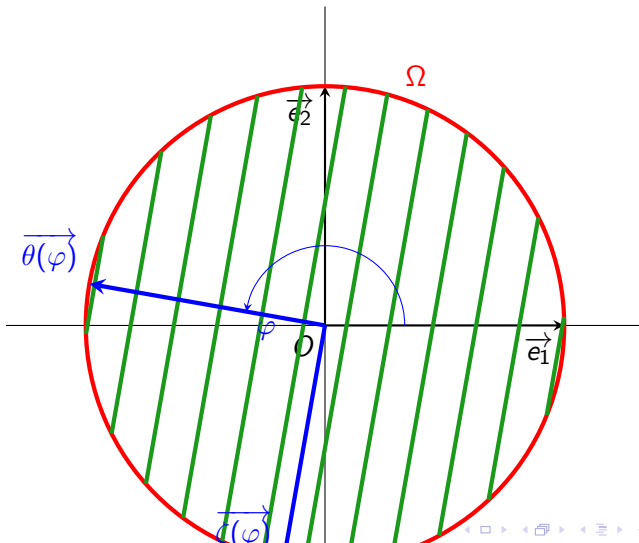
Radon Transform



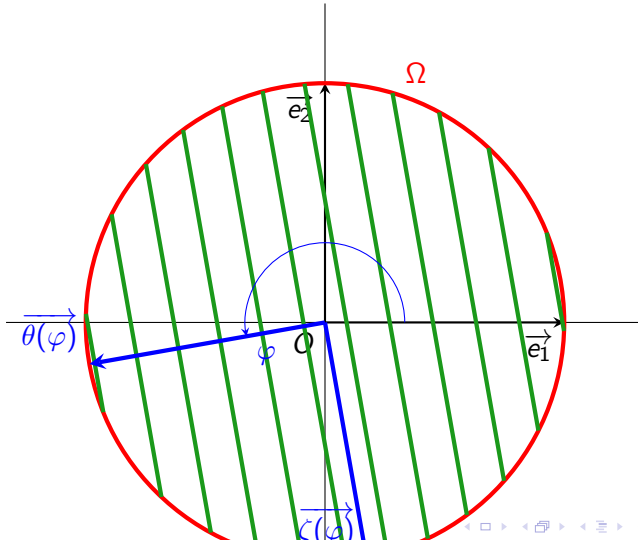
Radon Transform



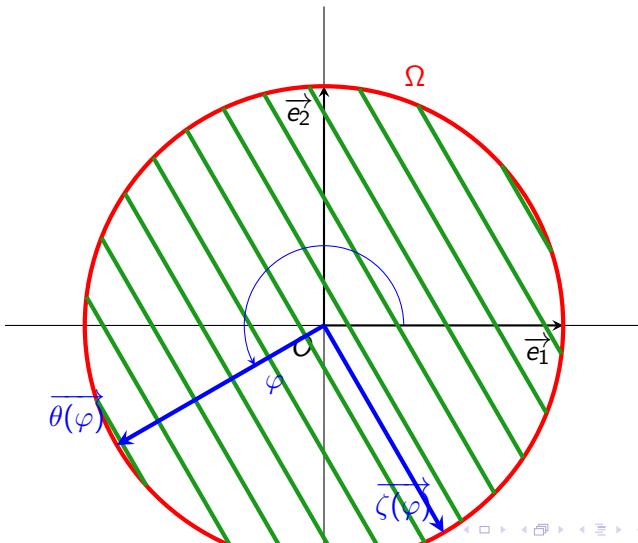
Radon Transform



Radon Transform



Radon Transform



Center Slice Theorem

Definition

$$\mathcal{R}_{\theta}\mu f(s) \stackrel{\text{def}}{=} \mathcal{R}\mu f(\theta, s) \quad (1)$$

Theorem

Let $\mu \in \mathbb{L}^1(\mathbb{R}^2)$ then

$$\widehat{\mathcal{R}_{\theta}\mu}(\sigma) = \hat{\mu}(\sigma\theta)$$

Proof of Center Slice Theorem

Proof.

$$\begin{aligned}\widehat{\mathcal{R}_\theta \mu}(\sigma) &= \int_{\mathbb{R}} \mathcal{R}_\theta \mu(s) e^{-2i\pi\sigma s} ds \\ &= \int_{\mathbb{R}} \int_{\mathbb{R}} \mu(s\theta + l\zeta) dl e^{-2i\pi\sigma s} ds \\ &= \int_{\mathbb{R}^2} \mu(x) e^{-2i\pi\sigma\theta \cdot x} dx \\ &= \hat{\mu}(\sigma\theta)\end{aligned}$$

where we made the change of variables (s, l) to (x_1, x_2) i.e. $x = s\theta + l\zeta$. This is a rotation thus $dl ds = dx_1 dx_2$ □

Filtered Back Projection

Theorem

Let $\mu \in \mathbb{L}^1(\mathbb{R}^2)$ sufficiently smooth then

$$\mu(x) = \int_0^\pi \int_{\mathbb{R}} \widehat{\mathcal{R}_\theta \mu}(\sigma) |\sigma| e^{2i\pi \sigma x \cdot \theta} d\sigma d\phi$$

Filtered Back Projection, proof

Proof.

$$\begin{aligned}\mu(x) &= \int_{\mathbb{R}^2} \hat{\mu}(\xi) e^{2i\pi\xi \cdot x} d\xi \\ &= \int_0^\pi \int_{\mathbb{R}} \hat{\mu}(\sigma\theta) e^{2i\pi\sigma\theta \cdot x} |\sigma| d\sigma d\phi \\ &= \int_0^\pi \int_{\mathbb{R}} \widehat{\mathcal{R}_\theta \mu}(\sigma) |\sigma| e^{2i\pi\sigma\theta \cdot x} d\sigma d\phi\end{aligned}$$

where we have made the polar change of variable $\xi = \sigma\theta(\phi)$, $(\phi, \sigma) \in [0, \pi[\times \mathbb{R}$, thus $d\xi_1 d\xi_2 = |\sigma| d\sigma d\phi$, and we have used theorem 2. □

Non Local Filter

We define p_H the Hilbert transform of the parallel projection p

$$p_H(\phi, s) = \int_{-\infty}^{+\infty} p(\phi, t) h(s - t) dt \text{ where } h(u) = \frac{1}{\pi u}$$

and $\hat{h}(\sigma) = -i \text{sgn}(\sigma)$ (distribution). The ramp filtering p_R of p is

$$p_R(\phi, s) = \frac{1}{2\pi} \frac{\partial}{\partial s} p_H(\phi, s)$$

The filter $p_\phi(s) \xrightarrow{\mathcal{F}} \widehat{p}_\phi(\sigma) \xrightarrow{\text{filter}} \widehat{p}_\phi(\sigma) |\sigma| \xrightarrow{\mathcal{F}^{-1}} p_{\phi \text{ filtered}}(s)$ is non local
 !!! $|\sigma| = \frac{1}{2\pi} (2i\pi\sigma) (-i \text{sgn}(\sigma))$ is the Hilbert filtering composed by the derivation.

Fan Beam geometry

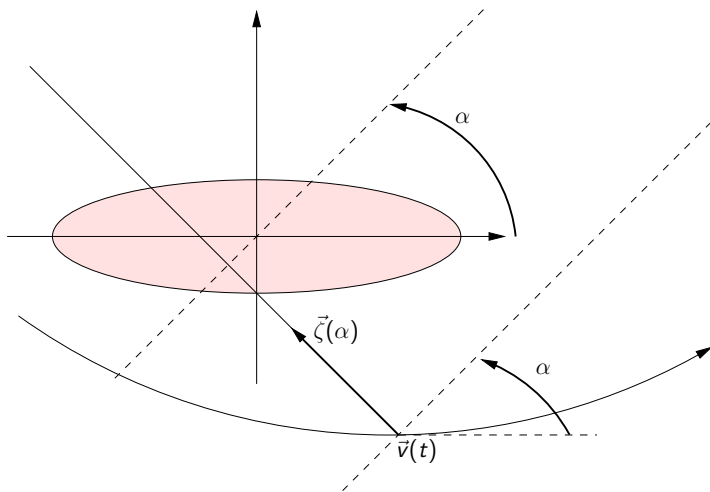


Figure: The Fan Beam variables (t, α)

Fan Beam geometry

We first define the source trajectory along a curve

$$\begin{aligned}v &: \mathcal{C} \longrightarrow \mathbb{R}^2 \\ t &\longrightarrow v(t)\end{aligned}$$

The fan-beam data are then defined by

$$g(v_t, \alpha) = \int_0^{+\infty} \mu(v_t + l\zeta(\alpha)) dl \quad (2)$$

We remark that

$$p(\phi, s) = g(v_t, \phi) + g(v_t, \phi + \pi) \text{ where } s = v_t \cdot \theta(\phi)$$

Fan Beam Inversion

We consider the circular trajectory, $v_t = (-R_v \cos t, -R_v \sin t)$,

Theorem

Let $\mu \in \mathbb{L}^1(\mathbb{R}^2)$ sufficiently smooth then

$$\mu(x) = \frac{1}{2} \int_0^{2\pi} \frac{1}{\|x - v_t\|^2} g_{WF}(v_t, \arg(x - v_t)) dt$$

where

$$g_{WF}(v_t, \phi) = \int_{t-\pi/2}^{t+\pi/2} R_v \cos(\psi - t) g(v_t, \psi) r(\sin(\phi - \psi)) d\psi$$

where r is the ramp filter ($\hat{r}(\sigma) = |\sigma|$).

Proof ... change of variables

Short Scan

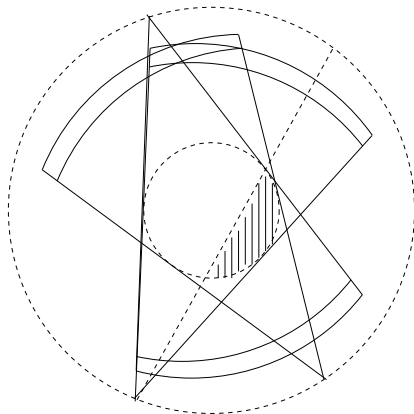


Figure: Short scan with (Parker) weight is possible....

Trajectories, small detectors

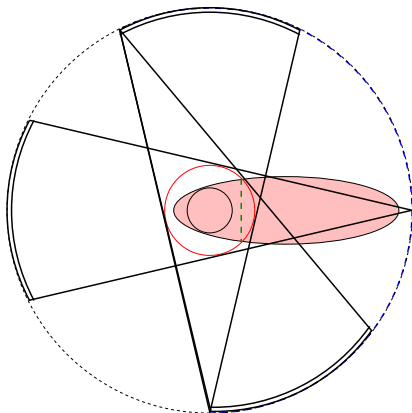


Figure: Small detector yields truncated data.

Hilbert and Fan Beam

We define g_H the Hilbert transform of the fan beam projection g

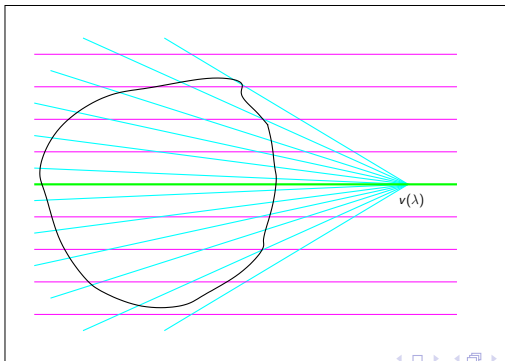
$$g_H(v_t, \phi) = \int_0^{2\pi} g_H(v_t, \psi) h(\sin(\phi - \psi)) d\psi \text{ where } h(u) = \frac{1}{\pi u}$$

The parallel Fan Beam Hilbert Projection Equality

Theorem

$$p_H(\phi, v_t \cdot \theta) = g_H(v_t, \phi) \quad (3)$$

(idea compute $p_H(\phi, s)$ from $g_H(v_t, \phi)$ with $v_t \cdot \theta = s$)



New Reconstruction Conditions

Recall

$$p_R(\phi, s) = \frac{1}{2\pi} \frac{\partial}{\partial s} p_H(\phi, s)$$

with

$$p_H(\phi, s) = g_H(v_t, \phi)$$

Theorem

The point x can be reconstructed from FB non truncated projections provided a fan beam vertex can be found on each line passing through x .

New Reconstruction Formula

$$\mu(x) = \frac{1}{2} \int_{\mathcal{C}} \frac{1}{\|x - v_t\|} w(v_t, \arg(x - v_t)) g_F(v_t, \arg(x - v_t)) dt$$

where

$$g_F(v_t, \phi) = \frac{1}{2\pi} \int_{t-\pi/2}^{t+\pi/2} h(\sin(\psi - \phi)) \frac{\partial g}{\partial t}(v_t, \psi) d\psi$$

where h is the hilbert filter.

Very Short Scan

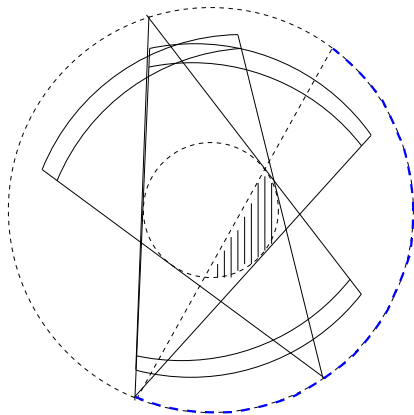


Figure: Very Short Scan....

Virtual Fan Beam

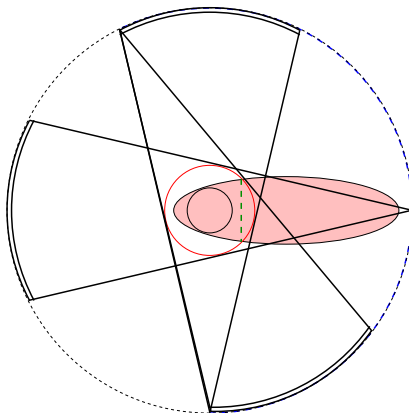


Figure: Data truncation: spine....

Idea

The idea of the Differentiated Backprojection is to compute the Hilbert transform of μ along a direction α from the back projection of the deviation of the projection. μ is then reconstructed from the inversion of the Hilbert transform.

Truncated projections

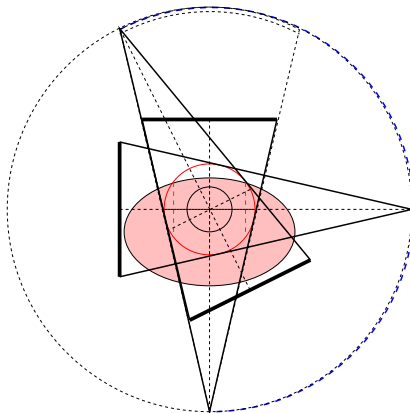


Figure: Data truncation: spine....