

Efficient Solution Methods for Inverse Problems with Application to Tomography

Practical Tomography

Alfred K. Louis

Institut für Angewandte Mathematik
Universität des Saarlandes
66041 Saarbrücken
<http://www.num.uni-sb.de>
louis@num.uni-sb.de
and
<http://www.isca-louis.com>
louis@isca-louis.com

Novosibirsk NSU, October, 2011

Content

- 1 Imaging Systems and Mathematical Models
- 2 Fan Beam Geometry
- 3 3D X rays

Content

- 1 Imaging Systems and Mathematical Models
- 2 Fan Beam Geometry
- 3 3D X rays

Transmission Tomography

- X-Ray CT (Nobel Prize 1979)
- Phase Contrast Tomography
- Magnetic Resonance Imaging (Nuclear Magnetic Resonance; Nobel Prize 2003)
- Ultrasound CT
- Electromagnetic Waves
- Impedance
- Light
- Electron Paramagnetic Resonance Imaging
- Transmission Electron Microscopy

Emission Tomography

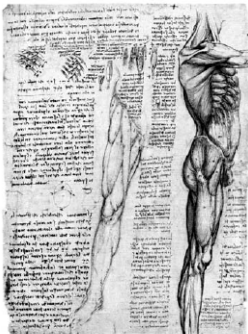
- PET
- SPECT
- EEG / MEG
- Bioluminescence Imaging

Nondestructive Testing

- Nondestructive Testing
 - X-Ray CT
 - Ultrasound
 - Microwaves
 - Backscattering X-Ray CT
 - Synchrotron Rays
 - Phase Contrast Tomography
 - Transmission Electron Microscopy

Historical Image

Leonardo da Vinci, 1500



Leonardo da Vinci. Anatomische Zeichnung. Um 1500

Historical Image

Hand of Dean of Röntgen, A.v. Koelliker, 23.01.1896

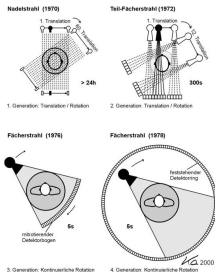


X-Ray CT



Content

- 1 Imaging Systems and Mathematical Models
- 2 Fan Beam Geometry**
- 3 3D X rays



Fan Beam Geometry

$$\mathbf{D}f(a, \theta) = \int_0^\infty f(a + t\theta) dt$$

Relation to Radon transform

$$\mathbf{D}f = URf$$

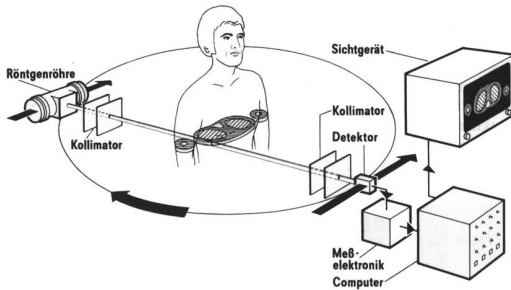
where U is a unitary transform.

Transformation of the parallel geometry inversion to fan beam inversion.

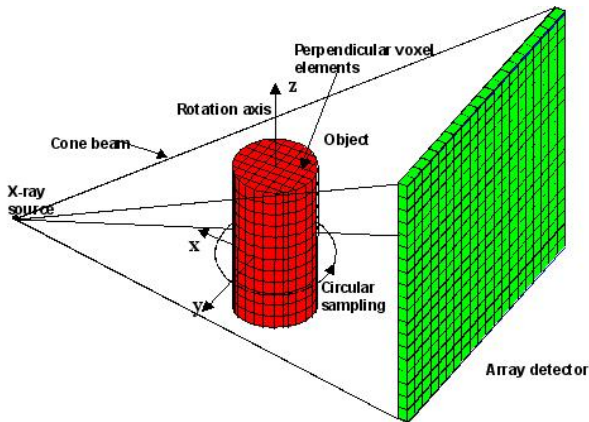
Content

- 1 Imaging Systems and Mathematical Models
- 2 Fan Beam Geometry
- 3 3D X rays**

Basics



Typical Scanning Geometry in NDT



Some 3D Transforms

- 3D Radon Transform: $\mathbf{R}f(\theta, \mathbf{s}) = \int_{\theta^\perp} f(\mathbf{s}\theta + \mathbf{y})d\mathbf{y}$

Some 3D Transforms

- 3D Radon Transform: $\mathbf{R}f(\theta, \mathbf{s}) = \int_{\theta^\perp} f(\mathbf{s}\theta + \mathbf{y})d\mathbf{y}$
Inversion Formula

$$f(\mathbf{x}) = -\frac{1}{8\pi^2} \int_{S^2} (\mathbf{R}f)''(\theta, \mathbf{x}^\top \theta) d\theta$$

Some 3D Transforms

- 3D Radon Transform: $\mathbf{R}f(\theta, \mathbf{s}) = \int_{\theta^\perp} f(\mathbf{s}\theta + \mathbf{y})d\mathbf{y}$
Inversion Formula

$$f(\mathbf{x}) = -\frac{1}{8\pi^2} \int_{S^2} (\mathbf{R}f)''(\theta, \mathbf{x}^\top \theta) d\theta$$

- 3D Parallel X - Ray Transform $\mathbf{P}f(\theta, \mathbf{y}) = \int_{-\infty}^{\infty} f(\mathbf{y} + t\theta)dt$

Some 3D Transforms

- 3D Radon Transform: $\mathbf{R}f(\theta, s) = \int_{\theta^\perp} f(s\theta + y)dy$
Inversion Formula

$$f(x) = -\frac{1}{8\pi^2} \int_{S^2} (\mathbf{R}f)''(\theta, x^\top \theta) d\theta$$

- 3D Parallel X - Ray Transform $\mathbf{P}f(\theta, y) = \int_{-\infty}^{\infty} f(y + t\theta)dt$
Inversion Formula

$$f = c\mathbf{P}^*\mathbf{I}^{-1}\mathbf{P}f$$

where

$$\mathbf{I}^{-1}g(\xi) = |\xi|\hat{g}(\xi)$$

and

$$\mathbf{P}^*g(x) = \int_{S^2} g(\theta, E_\theta x) d\theta$$

Some 3D Transforms

- 3D Radon Transform: $\mathbf{R}f(\theta, s) = \int_{\theta^\perp} f(s\theta + y)dy$
Inversion Formula

$$f(x) = -\frac{1}{8\pi^2} \int_{S^2} (\mathbf{R}f)''(\theta, x^\top \theta) d\theta$$

- 3D Parallel X - Ray Transform $\mathbf{P}f(\theta, y) = \int_{-\infty}^{\infty} f(y + t\theta)dt$
Inversion Formula

$$f = c\mathbf{P}^*\mathbf{I}^{-1}\mathbf{P}f$$

where

$$\mathbf{I}^{-1}g(\xi) = |\xi|\hat{g}(\xi)$$

and

$$\mathbf{P}^*g(x) = \int_{S^2} g(\theta, E_\theta x) d\theta$$

Too many data!

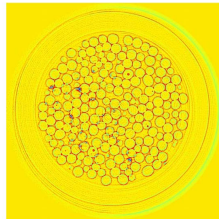
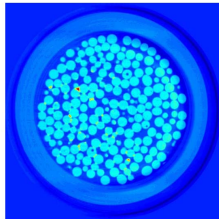
Synchrotron Measurements

Parallel X - Ray Transform for $\theta \in \mathcal{S}^1 \times \{0\}$.
Then Transform

$$\mathbf{P} = \mathbf{P}_2 \otimes \mathbf{I}$$

Adapt 2D-Inversion (B. Hahn)

3D Synchrotron Data



Better results using 3D algorithms: **Bernadette HAHN**

Big Challenge

Application: Determination of fluid flow in porous media

Big Challenge

Application: Determination of fluid flow in porous media

Object time dependent

Big Challenge

Application: Determination of fluid flow in porous media

Object time dependent

Data inconsistent for classical transform

3D Cone Beam : Mathematical Model

Source position: $a \in \Gamma$

Direction: $\theta \in S^2$

3D Cone Beam : Mathematical Model

Source position: $a \in \Gamma$

Direction: $\theta \in S^2$

$$Df(a, \theta) = \int_0^\infty f(a + t\theta) dt$$

3D Cone Beam : Mathematical Model

Source position: $\mathbf{a} \in \Gamma$

Direction: $\theta \in S^2$

$$Df(\mathbf{a}, \theta) = \int_0^\infty f(\mathbf{a} + t\theta) dt$$

For flat detectors use other geometry

3D Cone Beam : Mathematical Model

Source position: $a \in \Gamma$

Direction: $\theta \in S^2$

$$\mathbf{D}f(a, \theta) = \int_0^\infty f(a + t\theta) dt$$

For flat detectors use other geometry

Compare with Radon transform in \mathbb{R}^3

$$\mathbf{R}f(\omega, s) = \int f(x) \delta(s - x^\top \omega) dx$$

Dual Transform

$$\mathbf{D} : L_2(\mathbb{R}^3) \rightarrow L_2(\Gamma \times S^2)$$
$$\mathbf{D}^* g(x) = \int_{\Gamma} |x - a|^{-2} g\left(a, \frac{x - a}{|x - a|}\right) da$$

Relation between Radon and Cone Beam Transform

Grangeat:

$$\frac{\partial}{\partial \mathbf{s}} \mathbf{R}f(\omega, \mathbf{a}^\top \omega) = - \int \mathbf{D}f(\mathbf{a}, \theta) \delta'(\omega^\top \theta) d\theta$$

Proof:

$$\begin{aligned} \int \mathbf{R}f(\omega, \mathbf{s}) \psi(\mathbf{s}) d\mathbf{s} &= \int f(\mathbf{x}) \psi(\mathbf{x}^\top \omega) d\mathbf{x} \\ \int \mathbf{D}f(\mathbf{a}, \theta) h(\theta) d\theta &= \int f(\mathbf{x}) h\left(\frac{\mathbf{x} - \mathbf{a}}{|\mathbf{x} - \mathbf{a}|}\right) |\mathbf{x} - \mathbf{a}|^{-2} d\mathbf{x} \end{aligned}$$

Put $\psi(\mathbf{s}) = \delta'(\mathbf{s} - \mathbf{a}^\top \omega)$ and $h(\theta) = \delta'(\theta^\top \omega)$ and use δ' homog. of degree -2 in \mathbb{R}^3 .

References

- Hamaker, Smith, Solmon, Wagner, 1980
- Tuy, 1984
- Grangeat 1986
- Dietz 1999
- AKL, 2000
- Katsevich, 2000
- Zhao, Jiang, Zhuang, Wang, 2006
- ...

Inversion Formula

Theorem (AKL 2004)

Let the condition of Tuy-Kirillov be fulfilled. Then the Inversion formula can be given as

Inversion Formula

Theorem (AKL 2004)

Let the condition of Tuy-Kirillov be fulfilled. Then the Inversion formula can be given as

$$f = -\frac{1}{8\pi^2} D^* T M_{\Gamma, a} T D f$$

where

$$D^* g(x) = \int_{\Gamma} |x - a|^{-2} g(a, \frac{x - a}{|x - a|}) da$$

$$Tg(\omega) = \int_{S^2} g(\theta) \delta'(\theta^T \omega) d\theta$$

$$M_{\Gamma, a} h(\omega) = |a'^T \omega| m(\omega, a^T \omega) h(\omega)$$

Crofton Symbol

$$m = 1/n$$

where n is the Crofton symbol, counting how often a plane through a point cuts the source path Γ

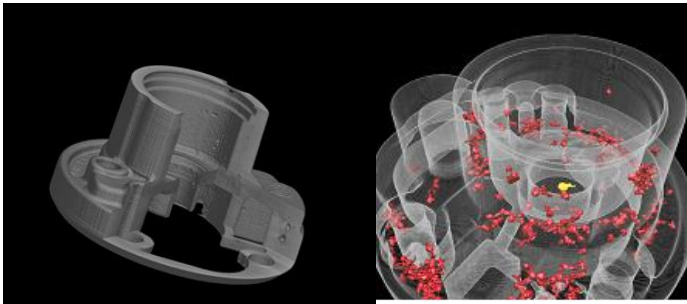
Hence $n \in \mathbb{N}_0$ and therefore

m is not differentiable!

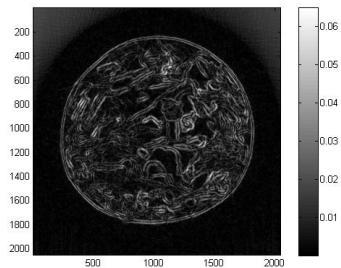
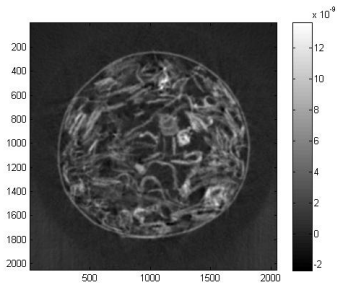
Challenge

Inversion formula contains detailed information about curve Γ .
But in practice measurements are taken only at discrete points.
How to include in algorithms?

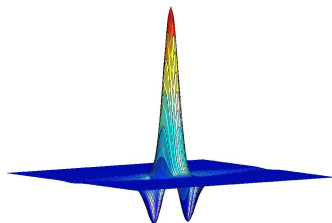
Images



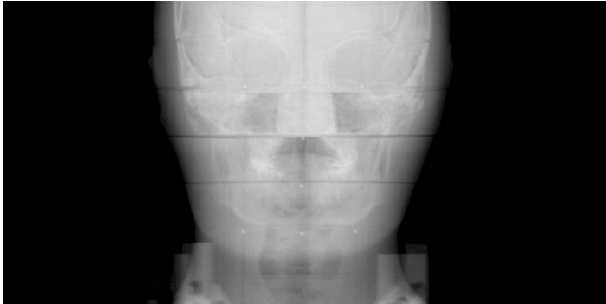
Images



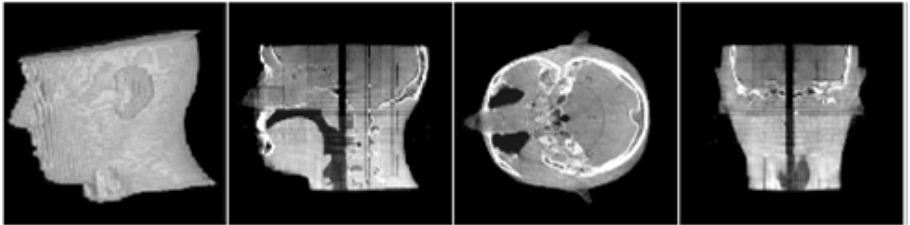
Reconstruction Kernel



Data from DKFZ

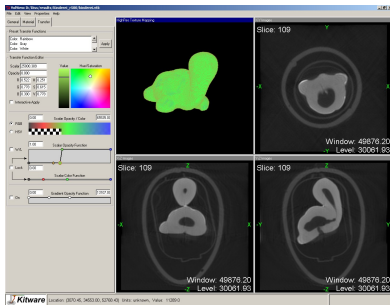


Reconstruction



Surprise Egg

Data from IzfP



Comparison with Method of Katsevich

Comparison with Method of Katsevich

- Katsevich uses π - lines

Comparison with Method of Katsevich

- Katsevich uses π - lines
- hence integration over parts of Γ where n is constant

Comparison with Method of Katsevich

- Katsevich uses π - lines
- hence integration over parts of Γ where n is constant
- Backprojection depends on reconstruction point x

Comparison with Method of Katsevich

- Katsevich uses π - lines
- hence integration over parts of Γ where n is constant
- Backprojection depends on reconstruction point x
- Jump of n at the end introduces δ - distributions, hence point evaluation of data

Laminography

Application for example: printed circuit board

- Data given in very small range

Laminography

Application for example: printed circuit board

- Data given in very small range
- Classical Algorithms fail

Laminography

Application for example: printed circuit board

- Data given in very small range
- Classical Algorithms fail
- So far: Iterative algorithms realized on graphical processors (GPU)

Laminography

Application for example: printed circuit board

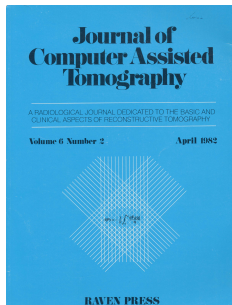
- Data given in very small range
- Classical Algorithms fail
- So far: Iterative algorithms realized on graphical processors (GPU)
- Optimizing order of used equations

Laminography

Application for example: printed circuit board

- Data given in very small range
- Classical Algorithms fail
- So far: Iterative algorithms realized on graphical processors (GPU)
- Optimizing order of used equations
- Use a priori information (see Lavrentiev)

Importance of Mathematics



$$f(\mathbf{P}) = -\frac{1}{\pi} \int_0^{\infty} \frac{d\bar{F}_p(\mathbf{q})}{q}$$