Introduction to Medical Imaging

Cone-Beam CT

Klaus Mueller

Computer Science Department Stony Brook University

Introduction

Available cone-beam reconstruction methods:

- exact
- approximate
- algebraic

Our discussion:

- exact (now)
- approximate (next)
- algebraic (nuclear imaging)

The Radon transform and its inverse are important mechanisms to understand cone-beam CT

Cone-Beam Transform

$$D\mu\left(\vec{a}(t),\vec{\beta}\right) = \int_{0}^{\infty} \mu\left(\vec{a}(t) + s\vec{\beta}\right) ds, \quad \left(\vec{a},\vec{\beta}\right) \in \Gamma \times S^{2}$$

 $\vec{a}(t)$ is the source position along trajectory Γ $\vec{\beta}$ the unit vector pointing along a particular x-ray beam

The cone-beam transform reflects the data acquisition process of measuring line integrals of the attenuation coefficient μ .



2D Radon Transform

The analytical approach of reconstruction by projections has to be done in the context of the Radon transform 9२

$$\begin{split} \Re \mu(\rho, \vec{\theta}) = &\int d^2 r \, \delta(\vec{r} \cdot \vec{\theta} - \rho) \cdot \mu(\vec{r}) = \\ & \int_{-\infty}^{+\infty} dl \, \mu(\rho \cdot \vec{\theta} + l \cdot \vec{\theta}_{\perp}) \end{split}$$

Thus in the 2D case the Radon transform $\Re \mu$ is identical to the measured cone beam transform $D \mu$

$$D\mu(\vec{a},\vec{\theta}_{\perp})\Big|_{\vec{a}\cdot\vec{\theta}=\rho} = \Re\mu(\rho,\vec{\theta})$$

with projection angle θ .



3D Radon Transform



which is a severe complication compared to the 2D case. As we will see the link to the measured cone beam transform $D\mu$ is not trivial.



Fourier-Slice Theorem in 2D

$$F_{\rho} \Re \mu(\rho, \vec{\theta}) = (F_2 \mu)(\omega_{\rho} \cdot \vec{\theta})$$

The radial 1D Fourier transform F_{ρ} of the Radon transform $\Re \mu$ along $\vec{\theta}$ is equal to the 2D Fourier transform F_2 of the object μ along $\vec{\theta}$ perpendicular to the direction of the projection.



Fourier-Slice Theorem in 3D

$$F_{\rho}\mathfrak{R}\mu(\rho,\bar{\theta}) = (F_{3}\mu)(\omega_{\rho}\cdot\bar{\theta})$$

The radial 1D Fourier transform F_{ρ} of the Radon transform $\Re \mu$ along $\vec{\theta}$ is equal to the 3D Fourier transform F_3 of the object μ along $\vec{\theta}$ perpendicular to the direction of the projection.



Exact Reconstruction in 2D and 3D

In 2D:

- use 2D inversion formula: the filtered backprojection procedure
- we have seen a spatial technique, only performing filtering in the frequency domain (in a polar grid)
- but may also interpolate the polar grid in the frequency domain and invert the resulting cartesian lattice
- employ linogram techniques for the latter (see later)

In 3D:

- use 3D inversion formula: not nearly as straightforward than 2D inversion
- full frequency-space methods also exist
- more details next (on all)

Exact Inversion Formula

The basic 3D inversion filtered backprojection formula, due to Natterer (1986):

$$f(\boldsymbol{x}) = \frac{-1}{8\pi^2} \int_{S^2} \frac{\partial^2}{\partial \rho^2} \Re f(|\rho|\theta) \,\mathrm{d}\theta.$$

- θ is the angle, a unit vector on a unit sphere
- x, ρ are object and Radon space coordinates, resp.: $|\rho| = x \cdot \theta$
- involves a 2nd derivative of the 3D Radon transform
- the second derivative operator can be treated as a convolution kernel

Some manipulations can reduce the second derivative to a first derivative, along with convolution operators

$$f(x) = \frac{1}{2} \int_{S^2} \frac{-1}{4\pi^2} \frac{\partial^2}{\partial \rho^2} \Re f(|\rho|\theta) \, \mathrm{d}\theta = \frac{1}{2} \int_{S^2} \frac{-1}{2\pi^2 \rho^2} * \frac{\partial}{\partial \rho} \left[\frac{1}{2\pi^2 \rho} * \Re f(|\rho|\theta) \right] \mathrm{d}\theta$$

- many different variants have been proposed
 - for example: Kudo/Saito (1990), Smith (1985)

Grangeat's Algorithm

Phase 1:

• from cone-beam data to derivatives of Radon data

Phase 2:

• from derivatives of Radon data to reconstructed 3D object

There are many ways to achieve Phase 2

- direct, O(N⁵)
- a two-step procedure, O(N⁴) [Marr et al, 1981]
- a Fourier method, O(N³ log N), [Axelsson/Danielsson, 1994]
- a divide-and-conquer strategy, O(N³ log N) [Basu/Bresler, 2002]
- we shall discuss the first three here

But first let us see how Radon data are generated from conebeam data

Transforming Cone-Beam to Radon Data



from Axelsson/Danielsson

Transforming Cone-Beam to Radon Data

$$\frac{\mathrm{d}}{\mathrm{d}\rho}[\Re f(\rho)] = \int_{-\pi/2}^{\pi/2} \int_0^\infty \frac{\mathrm{d}}{\mathrm{d}\rho} f(\rho, r, \gamma) r \,\mathrm{d}r \,\mathrm{d}\gamma = \frac{\mathrm{d}}{\mathrm{d}\kappa} \int_{-\pi/2}^{\pi/2} X f(\rho, \gamma) \frac{1}{\cos\gamma} \,\mathrm{d}\gamma$$
$$= \frac{SC}{\cos^2 \beta} \frac{\mathrm{d}}{\mathrm{d}s} \int_{-\infty}^\infty \frac{1}{SA} X f(\rho, t) \,\mathrm{d}t.$$

Strategy:

- weigh detector data with a factor 1/SA
- integrate along all intersections (lines) between the detector plane and the required Radon planes
 - there are N² such lines (N lines and N rotations)
- take the derivative in the s-direction (in the detector plane perpendicular to t)
- weight the 2D data set resulting from a single source position by the factor SC / $\cos^2\beta$

The order of these operations can be switched since they are all linear (Grangeat swapped the order of operation 2 and 3)

Radon Data to Object: Direct Method

There are O(N³) data points in Radon (derivative) space Each is due to a plane integral



The direct method simply inserts the plane data into the object space, one by one

- this is basically the expansion of a point into a plane, defined by (θ, ρ)
- this gives rise to an O(N⁵) algorithm

Radon Data to Object: Two-Step Method



Radon Data to Object: Two-Step Method

- Each vertical plane holds all Radon points due to plane integrals of perpendicularly intersecting planes
 - filtered backprojection reduces the plane integrals to line integrals, confined to horizontal planes
- The horizontal planes are then reconstructed with another filtered backprojection
- Each such operation is $O(N^3)$ and there are O(N) of them, resulting in a complexity of $O(N^4)$

Radon Data to Object: Fourier Space Approach



from Axelsson/Danielsson

Radon Data to Object: Fourier Space Approach

- Takes advantage of the O(N log N) complexity of the FFT at various steps
- It also uses linograms [Edholm/Herman, 1987] to reduce 2D interpolation to 1D interpolation

The complexity is then $O(N^3 \log N)$

Long Object Problem



Reconstruction of an ROI should be feasible from projection data restricted to the ROI and some surrounding.

The basic 3D Radon inversion formula does not fulfill this request.

from: Dr. Günter Lauritsch, Siemens

Tuy's Sufficiency Condition



Concept of PI-Lines



For a point x on a PI line any plane containing x has at least one intersection point with the PI segment associated with the PI line.

The PI segment is that portion of the source trajectory needed for reconstructing the point x.

Examples of Complete Trajectories



Circular Source Path

A prominent example of an incomplete trajectory



- Due to incomplete data sampling cone artifacts show up at sharp z-edges of objects with high contrast.
- Almost horizontal rays (or integration planes) are missing to distinguish between the members of the object stack.

Thorax simulation study. Coronal slice. C=0, W=200

from: Dr. Günter Lauritsch, Siemens

3D Radon Data Acquired by a Circular Trajectory



By a circular source trajectory a donut shaped region is acquired in 3D Radon space. At the z-axis a cone-like region is missing.

from: Dr. Günter Lauritsch, Siemens

Challenges in Cone-Beam Reconstruction

The naive application of the 3D Radon inversion formula is prohibitive due to

- long object problem
- enormous computational expense

Simplifications have to found to end up in an efficient and numerically stable reconstruction algorithm preferably in a shift-invariant 1D-filtered backprojection algorithm

Utilization of redundant data is obscure. Ideally redundancy in collected Radon planes has to be considered. However, this approach is suboptimal because:

- it is quite complicated
- underestimates the redundancy of data
- typically in cone beam, the data are highly redundant in approximation

The Feldkamp-Davis-Kress Algorithm

Approximate cone-beam algorithm

Works well for smaller cone-beam angles

Widely in use



The Feldkamp Algorithm: Details

for each projection ^Xd projection Уd weight pixels by a/b voxel v_i ramp-filter each column (along y_d direction for each grid voxel v_i Х project v_i onto image along cone-ra interpolate voxel update dv_i weight dv_i by depth factor c_i : $dv_i = dv_i \cdot c_i$ add result to grid voxel: $v_i = v_i + dv_i$

$$c_{j} = \frac{a^{2}}{(a + \sqrt{v_{jy}^{2} + v_{jz}^{2}}\cos(\varphi - \varphi_{k}))^{2}}$$