

3D Forward and Back-Projection for X-Ray CT Using Separable Footprints with Trapezoid Functions

Yong Long and Jeffrey A. Fessler

EECS Dept.
University of Michigan

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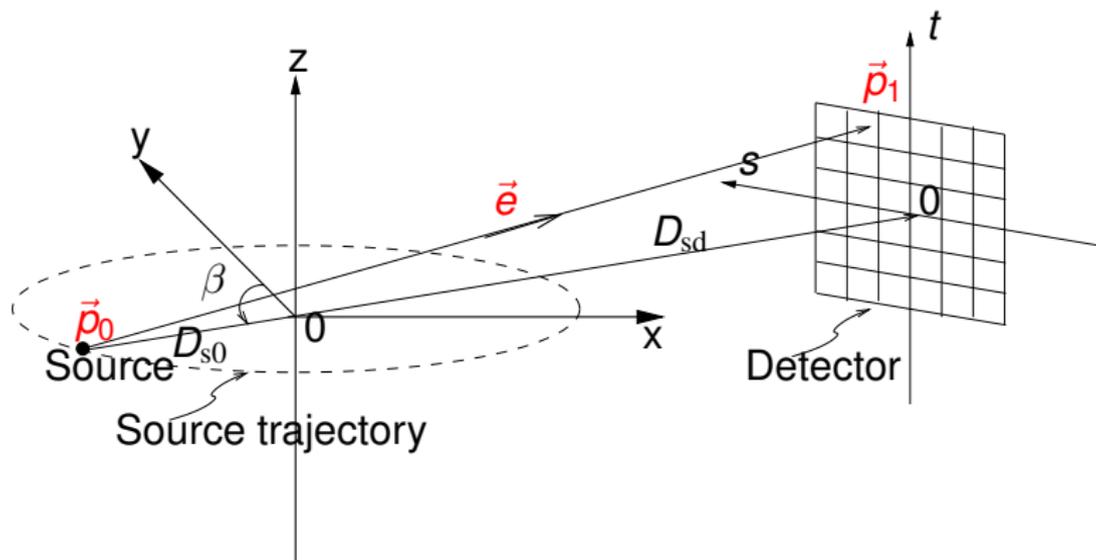
Introduction

- **Iterative methods** for 3D image reconstruction have the potential to improve image quality over conventional filtered back projection (FBP) based methods in X-ray CT.
- The primary computation burden of iterative methods is **3D cone-beam forward-** and **back-projection** using system matrix (\mathbf{A}').
- Infeasible to store \mathbf{A} due to its huge size.
- Forward projection goal: compute \mathbf{Ax} **rapidly** but **accurately** on the fly.
- Back projection goal: compute $\mathbf{A}'\mathbf{y}$ using exact transpose, **rapidly** but **accurately** on the fly.
- Many methods exist, all of which **compromise** between **computational complexity** and **accuracy**.

Separable Footprint (SF) Method

- The **distance-driven (DD)** is a current state-of-the-art method (De Man and Basu, PMB 2004).
- We proposed the **SF-TR** (trapezoid/rectangle) method (Long *et al.* Fully3D 2009, IEEE T-MI to appear).
- **SF-TT** (trapezoid/trapezoid) is the new extension reported here. (Inspired by reviewer inquiry.)
- Summary:
 - Quantitative accuracy: SF-TT > SF-TR > DD
 - Qualitative results: SF-TT \approx SF-TR > DD
 - Computation time: SF-TT > SF-TR \approx DD

Axial Cone-Beam Geometry



β denotes the angle of the source point counter-clockwise from the y axis.
 s and t denote detector plane coordinates

Cone-Beam Projection

- Ideal **cone-beam projection** of a 3D object $f(\vec{x})$:

$$p(\mathbf{s}, t; \beta) = \int_{\mathcal{L}(\mathbf{s}, t, \beta)} f(x, y, z) d\ell, \quad (1)$$

where ray is

$$\mathcal{L}(\mathbf{s}, t, \beta) = \left\{ \vec{p}_0 + \alpha \vec{e} : \alpha \in \left[0, \sqrt{D_{\text{sd}}^2 + \mathbf{s}^2 + t^2} \right] \right\}, \quad (2)$$

and $\vec{e} = \vec{e}(\mathbf{s}, t, \beta)$ is the direction vector from \vec{p}_0 to \vec{p}_1 .

Detector blur model

- **Shift-invariant detector blur** accounts for finite detector cell size:

$$h(\mathbf{s}, t) = \frac{1}{r_s r_t} \text{rect}\left(\frac{\mathbf{s}}{r_s}\right) \text{rect}\left(\frac{t}{r_t}\right), \quad (3)$$

where r_s and r_t denote the width along s and t respectively.

- Ideal (noiseless) sampled projection views:

$$\begin{aligned} \bar{y}_\beta[\mathbf{s}_k, t_l] &= \iint h(\mathbf{s}_k - \mathbf{s}, t_l - t) \rho(\mathbf{s}, t; \beta) \, d\mathbf{s} \, dt \\ &= \iint h(\mathbf{s}_k - \mathbf{s}, t_l - t) \left(\int_{\mathcal{L}(\mathbf{s}, t, \beta)} f(\mathbf{x}, y, z) \, d\ell \right) \, d\mathbf{s} \, dt, \end{aligned} \quad (4)$$

where $k = 0, \dots, N_s - 1$ and $l = 0, \dots, N_t - 1$.

- Discretized object based on a common basis function $\beta_0(\vec{\mathbf{x}})$:

$$f(\vec{\mathbf{x}}) = \sum_{\vec{\mathbf{n}}} f[\vec{\mathbf{n}}] \beta_0\left((\vec{\mathbf{x}} - \vec{\mathbf{c}}[\vec{\mathbf{n}}]) \oslash \vec{\Delta}\right), \quad (5)$$

where the grid spacing is $\vec{\Delta} = (\Delta_1, \Delta_2, \Delta_3)$, and $\Delta_1 = \pm \Delta_2$ hereafter.

Cone-Beam 3D System Model (cont.)

- Substitute object model (5) into measurement model (4) \Rightarrow

$$\bar{y}_\beta[\mathbf{s}_k, t_l] = \sum_{\vec{n} \in \mathcal{S}} \mathbf{a}_\beta[\mathbf{s}_k, t_l; \vec{n}] f[\vec{n}]. \quad (6)$$

- **System matrix \mathbf{A} elements** (“blurred footprint”)

$$\begin{aligned} \mathbf{a}_\beta[\mathbf{s}_k, t_l; \vec{n}] &= F(\mathbf{s}_k, t_l; \beta; \vec{n}) \\ &= \iint h(\mathbf{s}_k - \mathbf{s}, t_l - t) \mathbf{q}(\mathbf{s}, t; \beta; \vec{n}) \, d\mathbf{s} \, dt. \end{aligned} \quad (7)$$

- **Footprint** of the basis function

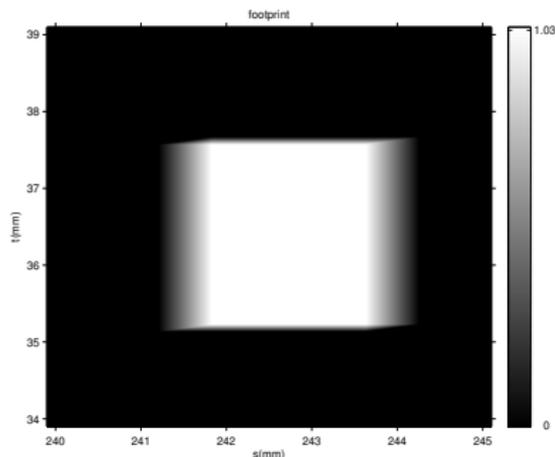
$$\mathbf{q}(\mathbf{s}, t; \beta; \vec{n}) = \int_{\mathcal{L}(\mathbf{s}, t, \beta)} \beta_0 \left((\vec{\mathbf{x}} - \vec{\mathbf{c}}[\vec{n}]) \oslash \vec{\Delta} \right) \, d\ell. \quad (8)$$

- Forward projection methods compute (8) and (7) approximately.

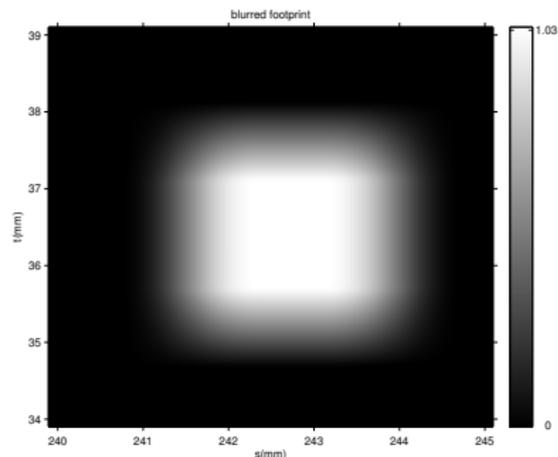
Exact footprint and blurred footprint, for rectangular cuboid (voxel)

$$F(s_k, t_l; \beta; \vec{n}) = \iint h(s_k - s, t_l - t) q(s, t; \beta; \vec{n}) ds dt$$

$$h(s, t) = \frac{1}{r_s r_t} \text{rect}\left(\frac{s}{r_s}\right) \text{rect}\left(\frac{t}{r_t}\right)$$

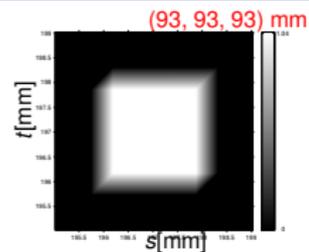
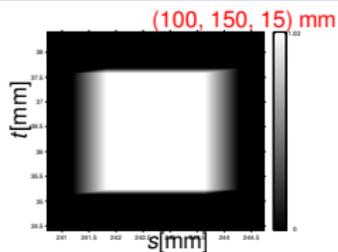
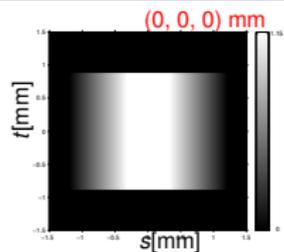


$$q(s, t; \beta; \vec{n})$$

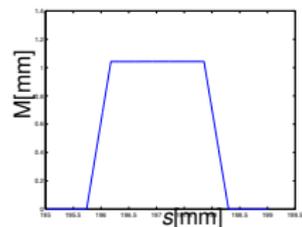
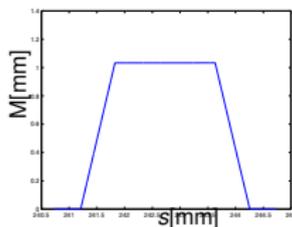
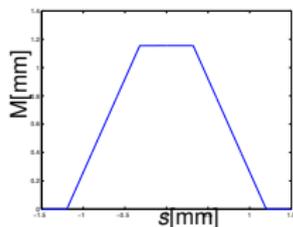


$$F(s, t; \beta; \vec{n})$$

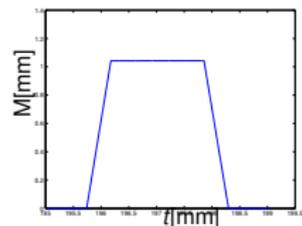
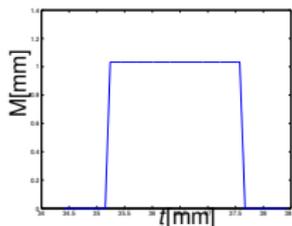
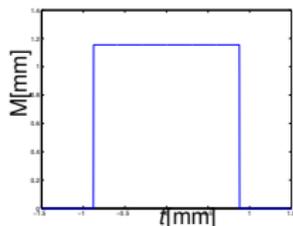
Cone-Beam Projections of Voxel Basis Functions (*Footprints*)



True footprints.



Profiles in s .



Profiles in t

Separable Footprint (SF) Approximation

- Why separable footprint approximation?
 - Inspired by shapes of true footprint functions.
 - 2D convolutions simplify to 1D convolutions.

- $$q(\mathbf{s}, t; \beta; \vec{n}) \approx q_{\text{ap}}(\mathbf{s}, t; \beta; \vec{n}) \triangleq l(\mathbf{s}, t; \beta; \vec{n}) q_{\text{sf}}(\mathbf{s}, t; \beta; \vec{n}).$$

- 2D separable function with unit maximum amplitude

$$q_{\text{sf}}(\mathbf{s}, t; \beta; \vec{n}) \triangleq q_1(\mathbf{s}; \beta; \vec{n}) q_2(t; \beta; \vec{n}).$$

- $l(\mathbf{s}, t; \beta; \vec{n})$ denotes the “amplitude”.

Separable Blurred Footprint

- Assume separable and shift-invariant detector blur:

$$h(s, t) = h_1(s)h_2(t).$$

- (Almost) **separable blurred footprint**:

$$F_{\text{sf}}(\mathbf{s}_k, t_l; \beta; \vec{n}) = \underbrace{I(\mathbf{s}_k, t_l; \beta; \vec{n})}_{\text{amplitude}} \underbrace{F_1(\mathbf{s}_k; \beta; \vec{n})}_{\text{transaxial}} \underbrace{F_2(t_l; \beta; \vec{n})}_{\text{axial}},$$

where

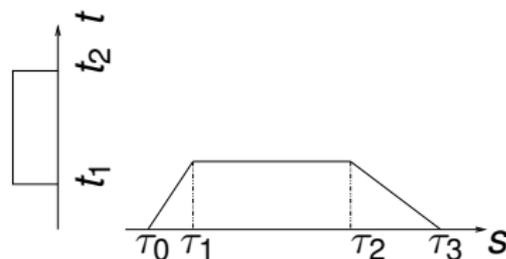
$$F_1(\mathbf{s}_k; \beta; \vec{n}) \triangleq \int h_1(\mathbf{s}_k - \mathbf{s})q_1(\mathbf{s}; \beta; \vec{n}) d\mathbf{s}$$

$$F_2(t_l; \beta; \vec{n}) \triangleq \int h_2(t_l - t)q_2(t; \beta; \vec{n}) dt.$$

SF-TR and ST-TT Projector

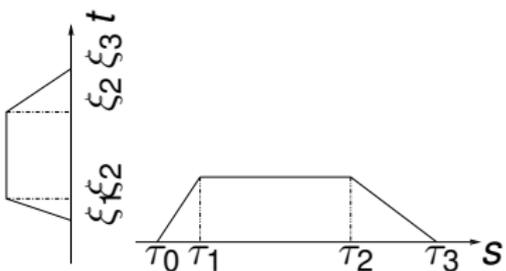
- SF-TR: Trapezoid/Rectangle

- Trapezoid function in s : $q_1(s; \beta; \vec{n}) \triangleq \text{trap}(s; \tau_0, \tau_1, \tau_2, \tau_3)$
- **Rectangular** function in t : $q_2(t; \beta; \vec{n}) \triangleq \mathbb{1}_{\{t_1 \leq t \leq t_2\}}$



- SF-TT: Trapezoid/Trapezoid

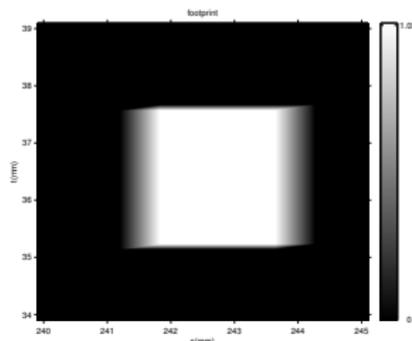
- Trapezoid function in s : $q_1(s; \beta; \vec{n}) \triangleq \text{trap}(s; \tau_0, \tau_1, \tau_2, \tau_3)$
- **Trapezoid** function in t : $q_2(t; \beta; \vec{n}) \triangleq \text{trap}(t; \xi_0, \xi_1, \xi_2, \xi_3)$



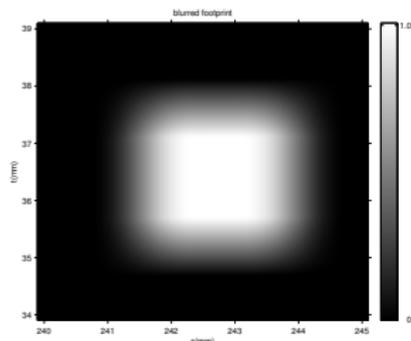
SF-TR and ST-TT Projector (cont.)

- Choose **vertices** of the approximating footprints to **match** exactly the **projections of the voxel boundaries**.
 - Approximation **adapts** to relative positions of source, voxels and detector, **as true footprints do**.
 - Models **depth-dependent magnification** accurately.
- SF-TR vs. SF-TT
 - **SF-TR** is accurate for **small** cone angle ($< 2^\circ$) geometries, *e.g.*, multi-slice detector geometries.
 - **SF-TT** is more accurate for **large** cone angle ($> 10^\circ$) geometries, *e.g.*, flat-panel detector geometries
 - SF-TR is **faster** than SF-TT.
 - Could be **combined** to balance computation and accuracy.

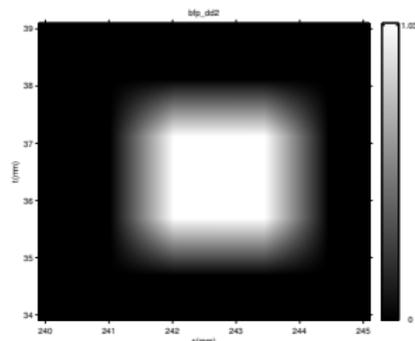
Blurred footprint approximations: near $z = 0$



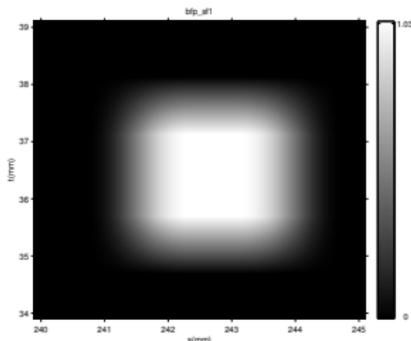
$$q(\mathbf{s}, t; \beta; \vec{n})$$



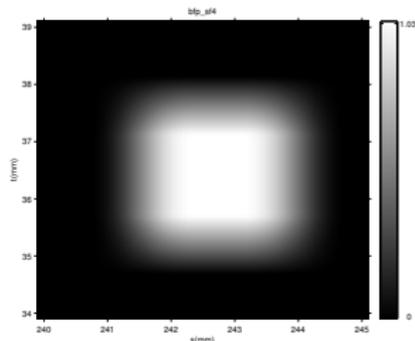
$$F(\mathbf{s}, t; \beta; \vec{n})$$



$$F_{DD}(\mathbf{s}, t; \beta; \vec{n})$$



$$F_{SF-TR}(\mathbf{s}, t; \beta; \vec{n})$$

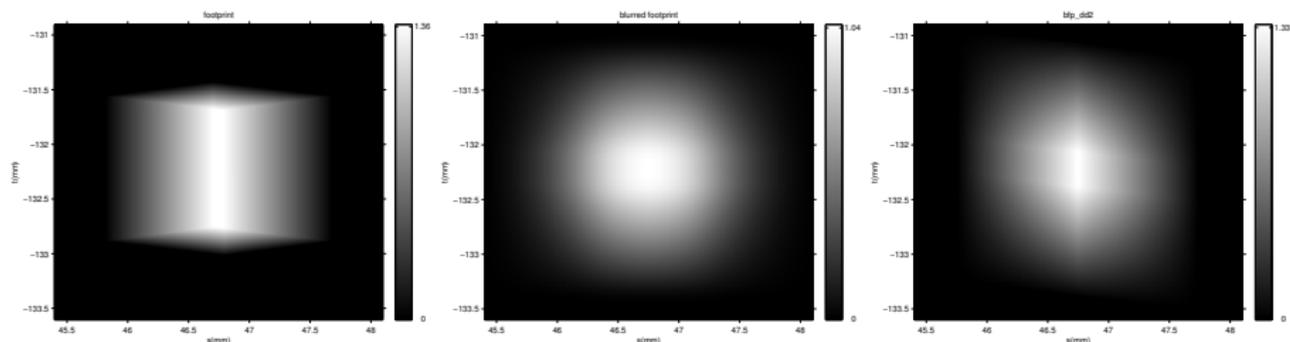


$$F_{SF-TT}(\mathbf{s}, t; \beta; \vec{n})$$

$\vec{n} = (100, 150, 15), \beta = 0$

Azimuthal angle through voxel center: 14.3° . Polar angle: 2.1° .

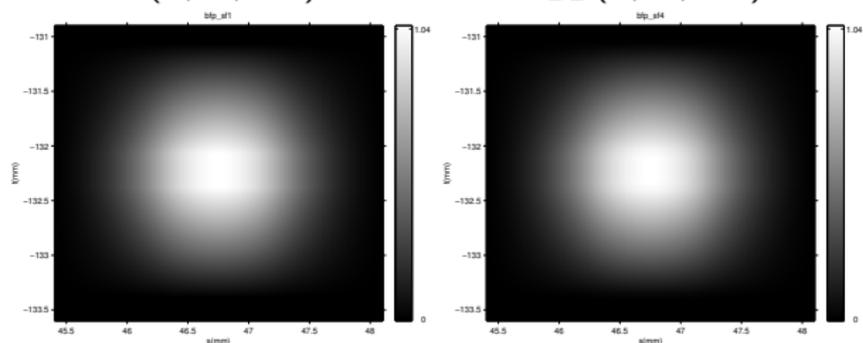
Blurred footprint approximations: off center



$$q(s, t; \beta; \vec{n})$$

$$F(s, t; \beta; \vec{n})$$

$$F_{DD}(s, t; \beta; \vec{n})$$



$$F_{SF-TR}(s, t; \beta; \vec{n})$$

$$F_{SF-TT}(s, t; \beta; \vec{n})$$

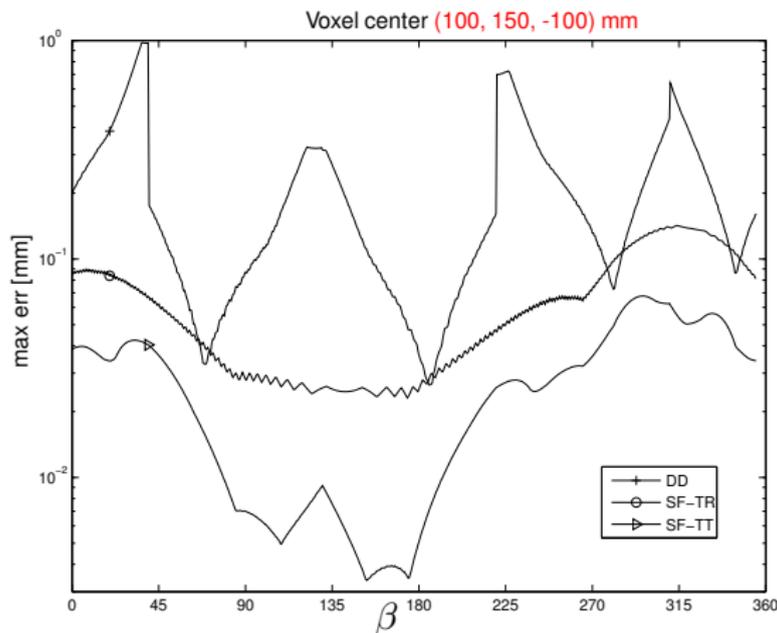
$\vec{n} = (100, 150, -100)$, $\beta = 135^\circ$

Azimuthal angle through voxel center: 138° . Polar angle: 7.8° .

Results: Projector maximum error (single voxel)

Error between exact blurred footprint and an approximation:

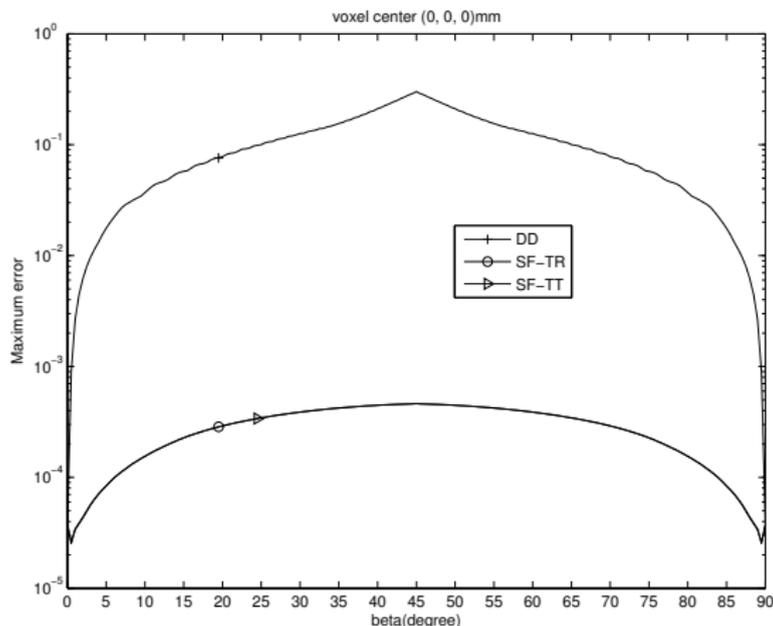
$$e(\beta; \vec{n}) \triangleq \max_{s, t \in \mathbb{R}} |F(s, t; \beta; \vec{n}) - F_{\text{approximation}}(s, t; \beta; \vec{n})|$$



Maximum errors on a **logarithmic scale** for a 1mm^3 size voxel.

Results: Projector maximum error (single voxel)

For a voxel at origin:



Maximum errors on a **logarithmic scale** for a 1mm³ size voxel.

Results: Projector computation time

Projectors	DD	SF-TR	SF-TT
Forward (s)	46	35	91
Back (s)	49	44	92

Forward project a $512 \times 512 \times 128$ image with $\Delta_x = 0.5$ mm.

Back-project a $512 \times 512 \times 984$ sinogram with detector cell spacing of 1 mm.

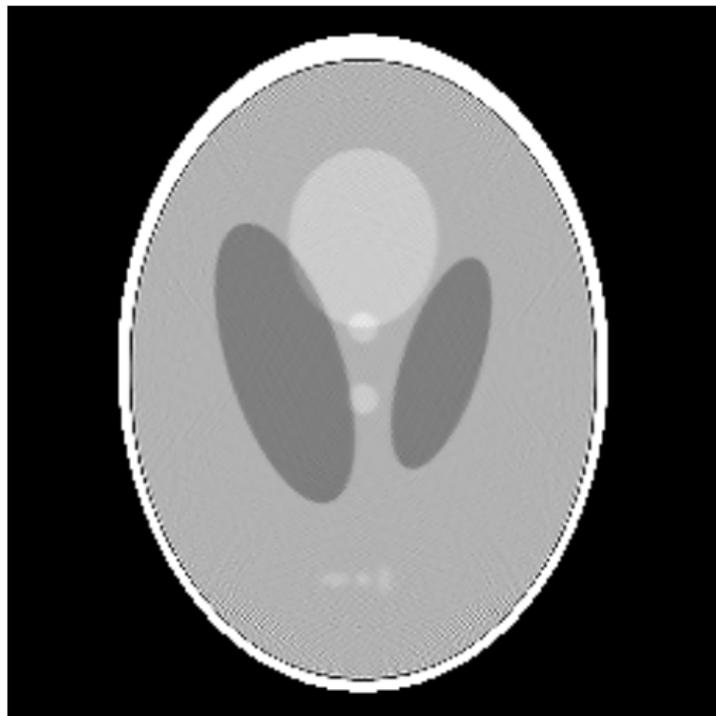
Elapsed time for 16 POSIX threads, averaged over 5 projector runs.

Iterative full FOV reconstruction: SF-TR vs. DD



Axial views of reconstructions by PWLS-CG iterative method.
Image size: $256 \times 256 \times 64$ with $\bar{\Delta} = 0.9766 \times 0.9766 \times 0.6250 \text{ mm}^3$.
SF-TT visually indistinguishable from SF-TR for this object and geometry,
as well as for several other geometries investigated, including both axial and helical
with flat-panel cone angle $> 10^\circ$.

DD reconstruction (1 mm voxels)



SF-TR reconstruction (1 mm voxels)



- Proposed the **SF-TT** projector.

Projector	speed	accuracy	cone angle
SF-TR	Fast	High	$< 2^\circ$
SF-TT	Slow	Higher	$> 10^\circ$

- Could **combine them** to balance computation and accuracy.
- SF-TR adequate for reconstructions under typical CT geometries (no obvious visual differences between SF-TR and SF-TT).