

Image Reconstruction from Sparse-view Data Acquired with Portable X-ray Devices

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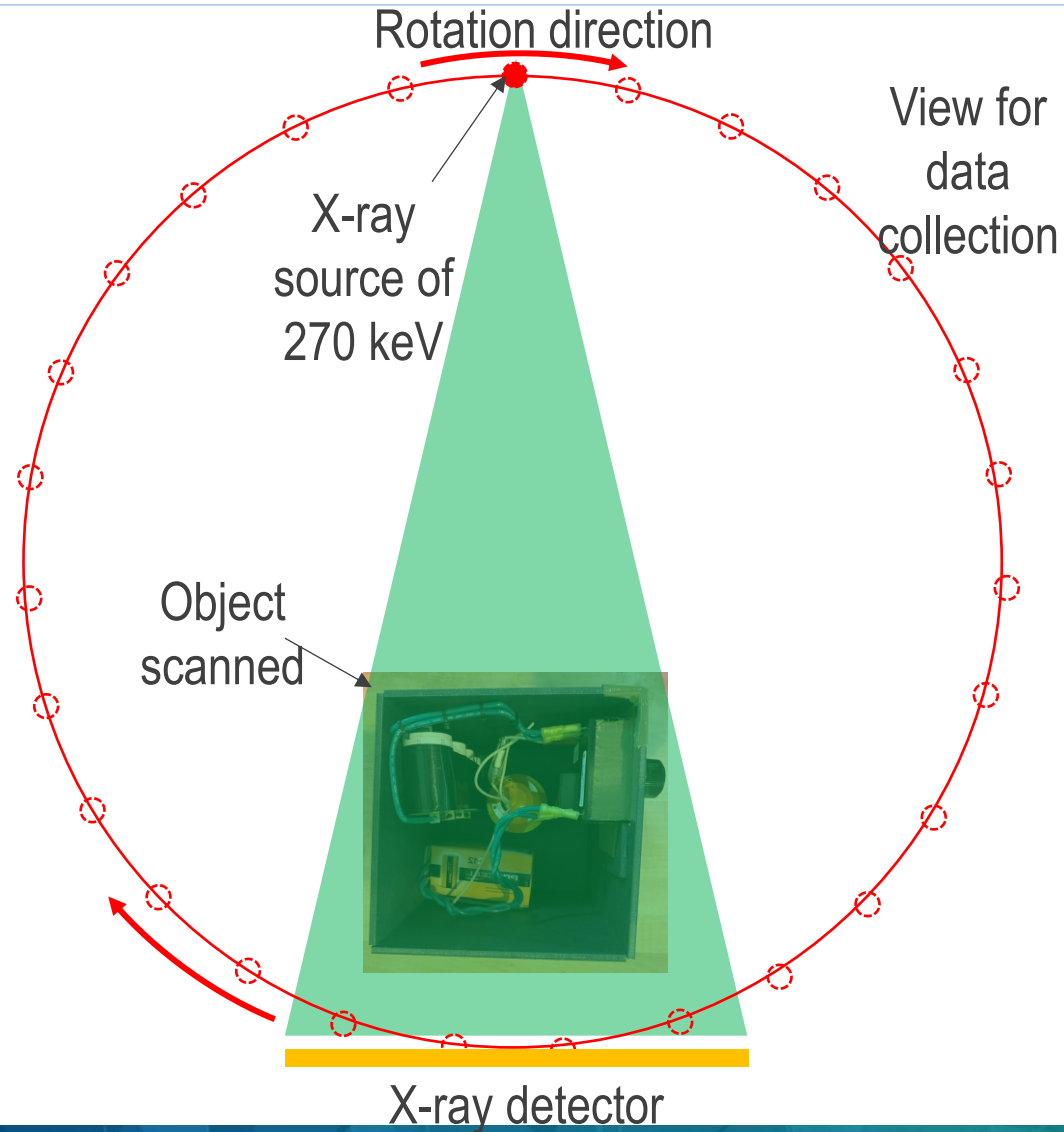
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Introduction

- Portable X-ray systems enable on-site 3D imaging for non-invasive inspection of suspicious packages and explosives.
- Existing reconstruction algorithms (e.g., FDK or Feldkamp, Davis and Kress) require hundreds of projections over 360 degrees.
- Sparse-view scan reduces scanning time and setup effort, making it ideal for field use in time-critical scenarios.
- Existing reconstruction algorithms introduce severe artifacts when applied to sparse-view data.
- We developed a total variation (TV)-based optimization algorithm for yielding 3D images from sparse-view data collected with our portable X-ray imaging system.



Cone-beam-CT Sparse-view Scan



- Angular ranges of 360°
- Sparse-view scans: 45 and 12 views
- Non-uniformly distributed views
- Non-circular trajectories
- Cone-beam illumination at each view

Image Model and Reconstruction

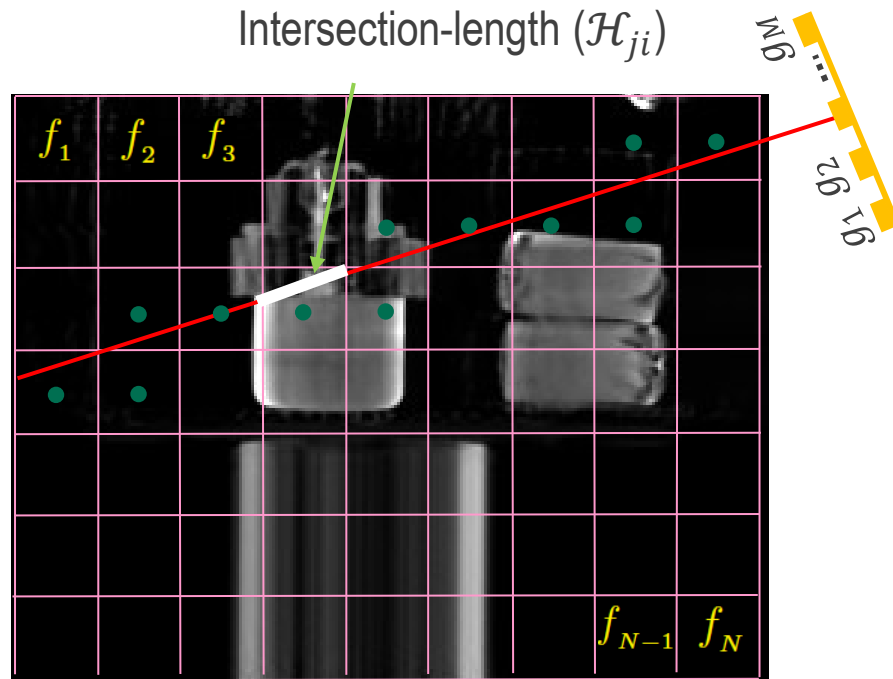


Image model – discrete X-ray transform

$$g = \mathcal{H}f$$

g : discrete model data

f : discrete image

\mathcal{H} : system matrix, each element \mathcal{H}_{ji} denotes the intersection-length of ray j and voxel i

Observation: Image reconstruction is equivalent to inverting the linear equation set.

Matrix \mathcal{H} encodes the system geometry and can accommodate arbitrary angular sampling schemes, including full 4π and partial 4π configurations, with either uniform or non-uniform angular distributions.



Optimization-based TV Reconstruction

Image reconstruction is formulated as a constrained optimization program:

$$f^* = \underset{f}{\operatorname{argmin}} \{ \|\nabla f\|_1 \} ,$$

$$s. t. \quad \|(\mathcal{H}f - g^{[\mathcal{M}]})\|_2 \leq \varepsilon, \text{ and } f_i \geq 0$$

➤ Objective function

- convex total-variation fidelity, in which $\|\nabla f\|_1$ denotes ℓ_1 -norm of the gradient magnitude image.

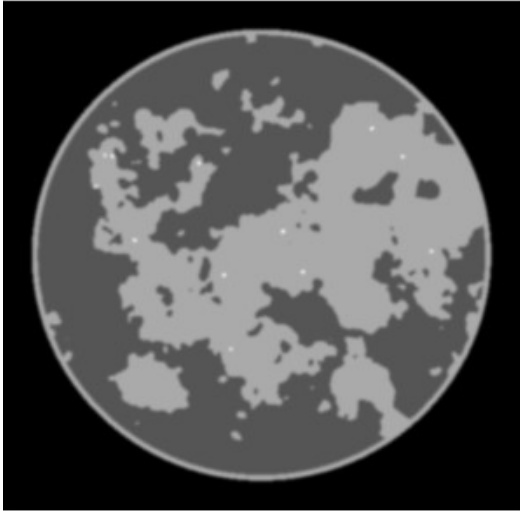
➤ Constraints:

- data- ℓ_2 constraint, and non-negative parameter, ε , is used to control the inconsistency between measured data (i.e., $g^{[\mathcal{M}]}$) and modeled data (i.e., $\mathcal{H}f$).
- non-negativity constraint is on image



Rationale for TV Minimization

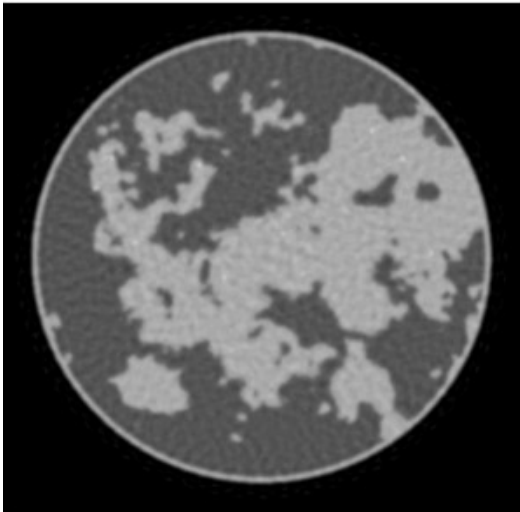
Phantom



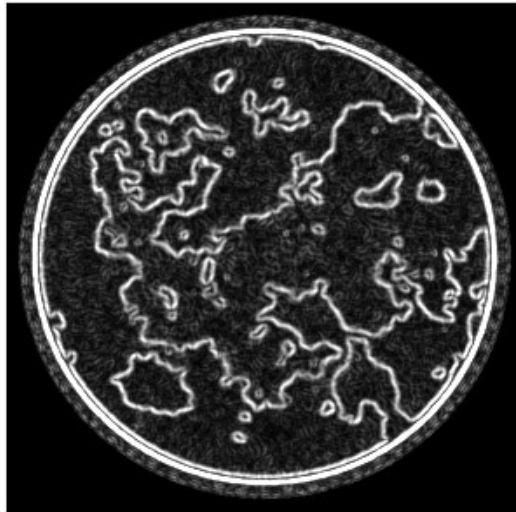
Gradient Magnitude



Least-squares



TV/TV₀ = 1.357



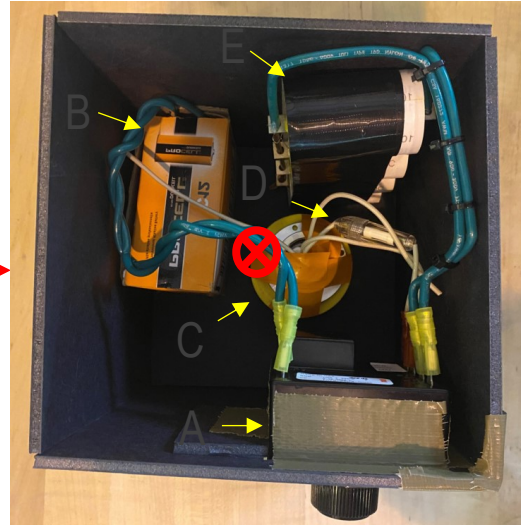
- Sparse view configuration leads to underdetermined imaging model.
- Both phantom and least-squares (LSQ) image solve $g = \mathcal{H}f$ exactly.
- Phantom is sparse in the gradient magnitude image (GMI), while LSQ image is not sparse in the GMI.
- Total variation (TV) measure is the sum of GMI pixel values.
- TV of LSQ image is greater than TV of phantom.
- Minimization of the TV rules out the LSQ image (and hopefully all other images that solve $g = \mathcal{H}f$).



Scanned Object and Scanning Configurations

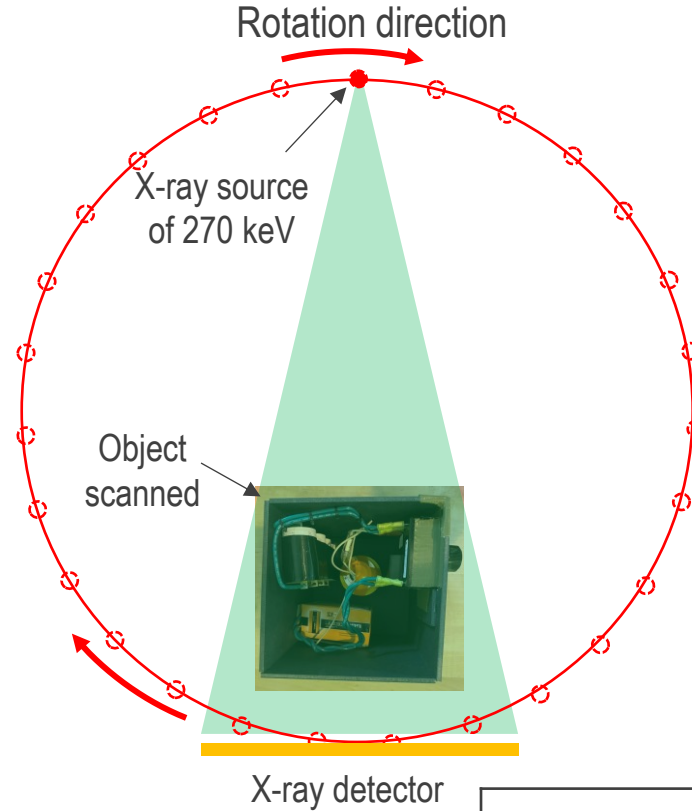
Photo of scanned object

Side view



Front view

- A: Timer
- B: Battery pack
- C: D₂ lamp in Ta cup
- D: Inline fuse
- E: Capacitor bank



Scanning configuration:

Source-to-rotation-axis distance: 130 cm

Source-to-detector distance: 152.5 cm

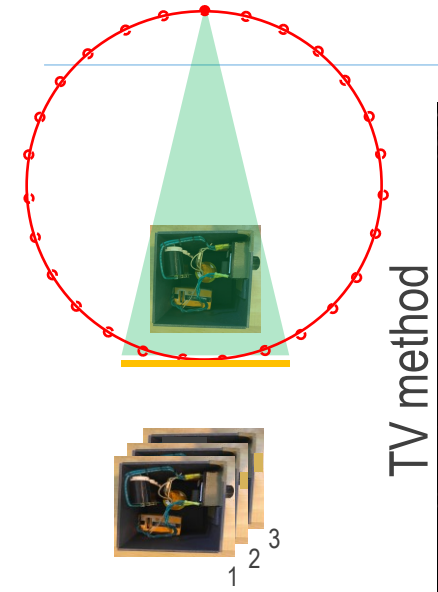
Detector size: 1152×1600 bins of size 0.254 mm

	Full data set	Data set 1	Data set2
Number of views	90	45	12
Angular coverage	360°	360°	360°
Angular interval	4°	8°	30°

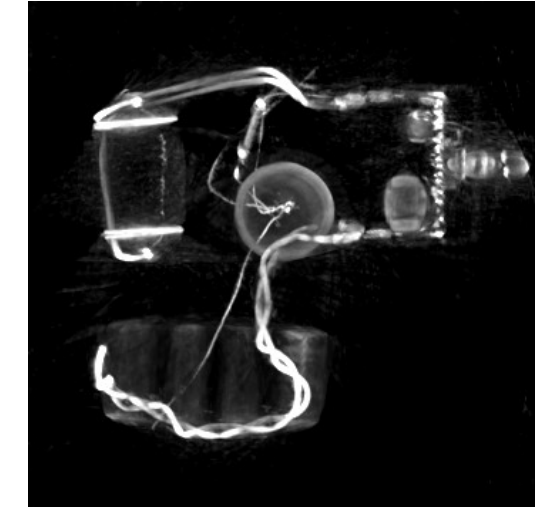
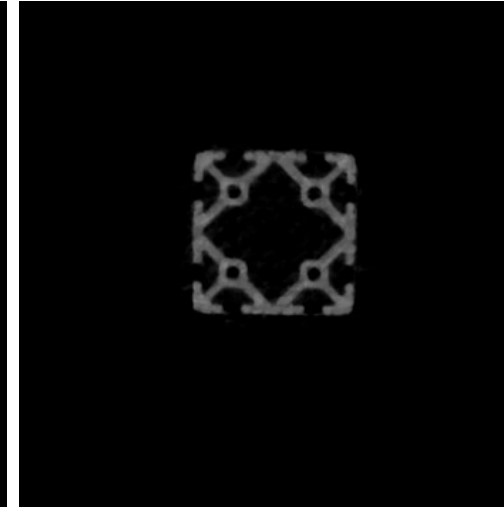
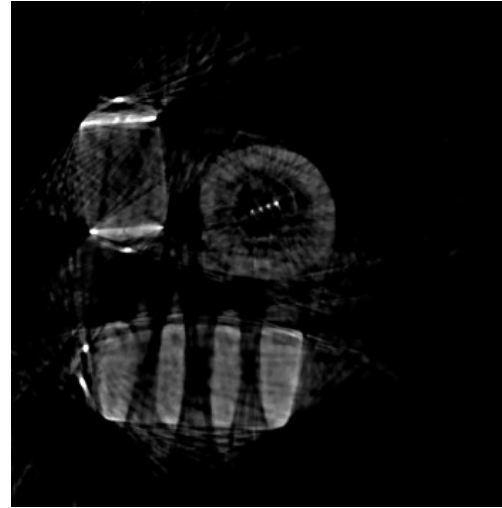
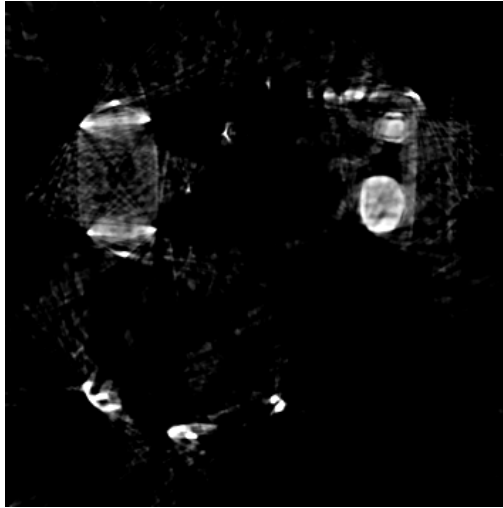


Transverse Slices and Maximum Intensity Projection (MIP) Images

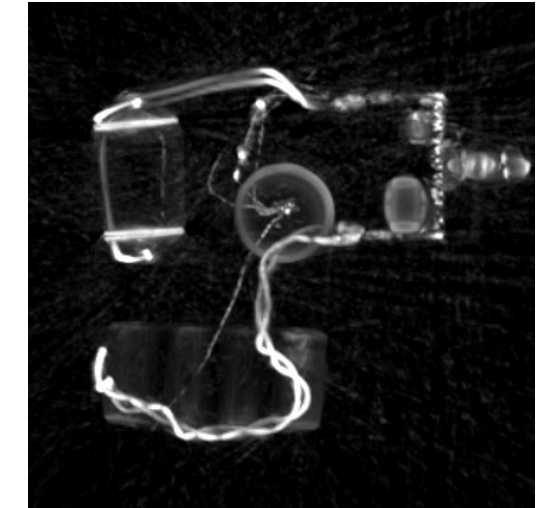
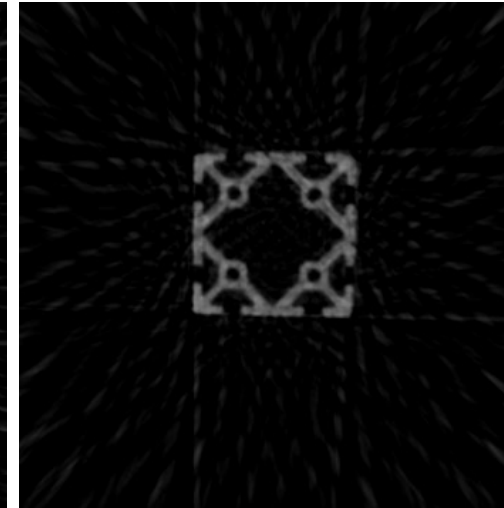
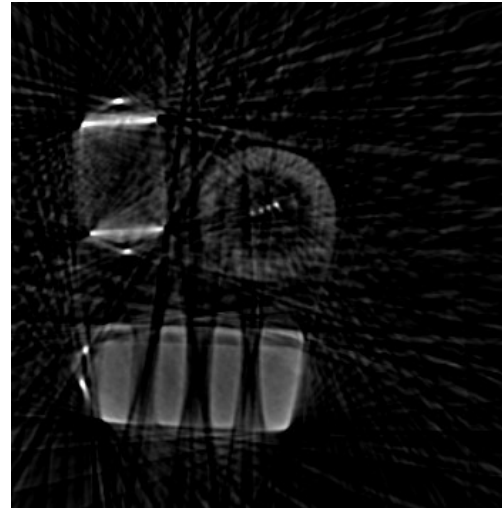
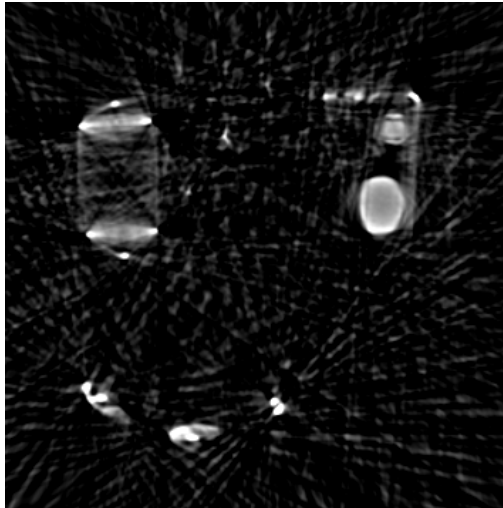
(45 views over 360° with an angular interval of 8°)



TV method



Existing FDK method



Slice 1

Slice 2

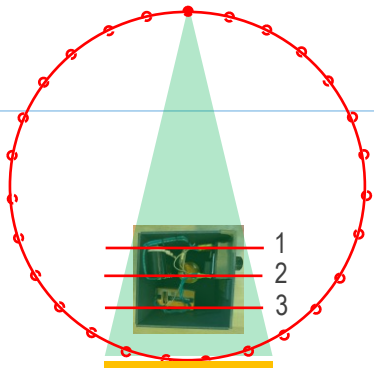
Slice 3

MIP

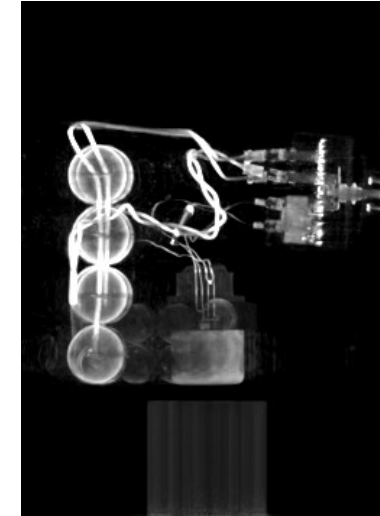
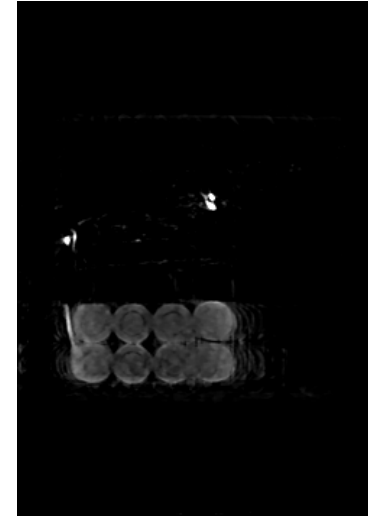
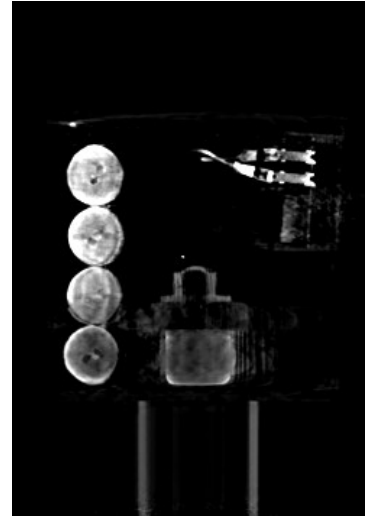


Coronal Slices and MIP Images

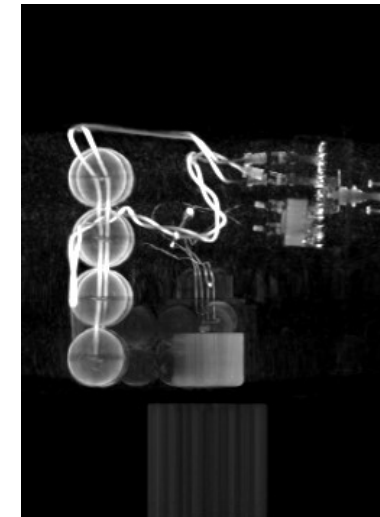
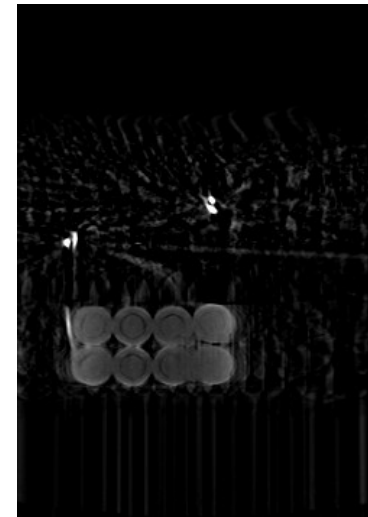
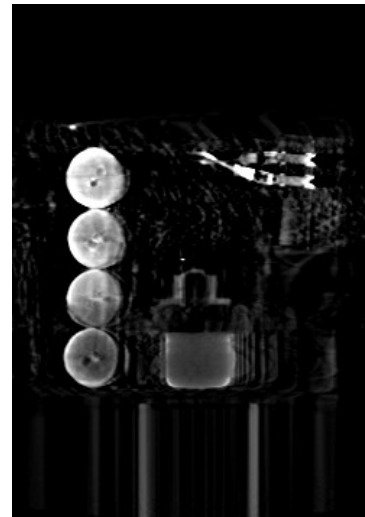
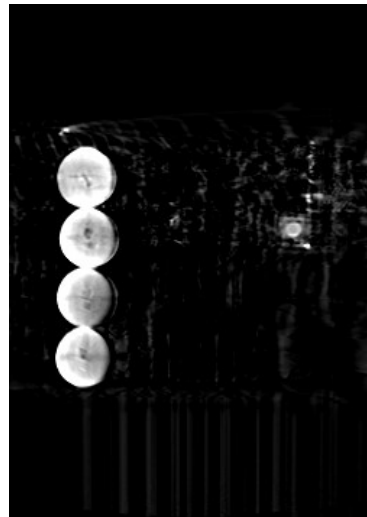
(45 views over 360° with an angular interval of 8°)



TV method



Existing FDK method



Slice 1

Slice 2

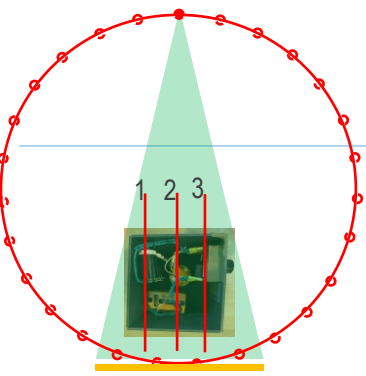
Slice 3

MIP

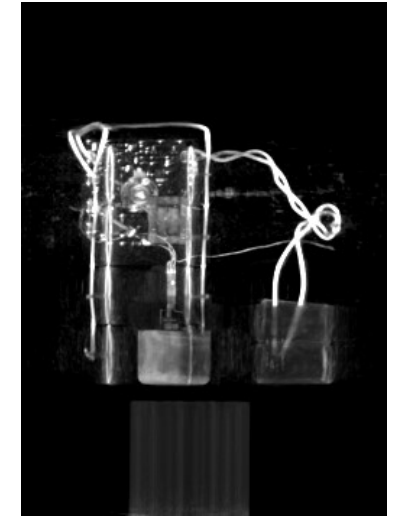
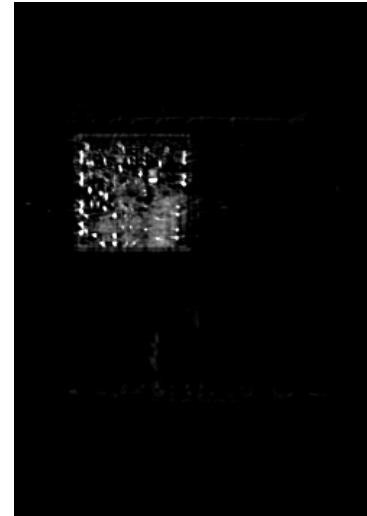


Sagittal Slices and MIP Images

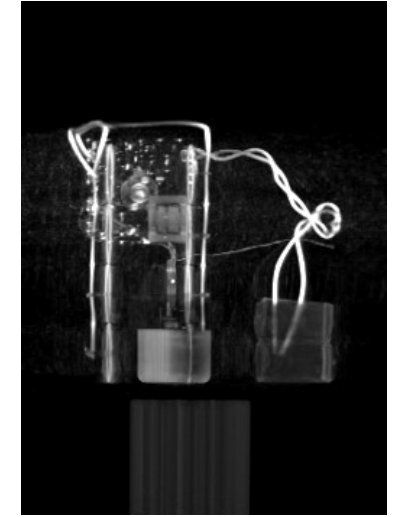
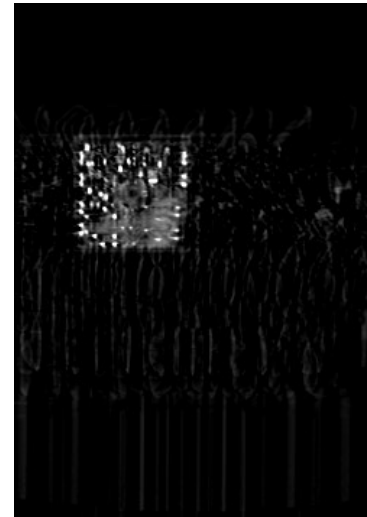
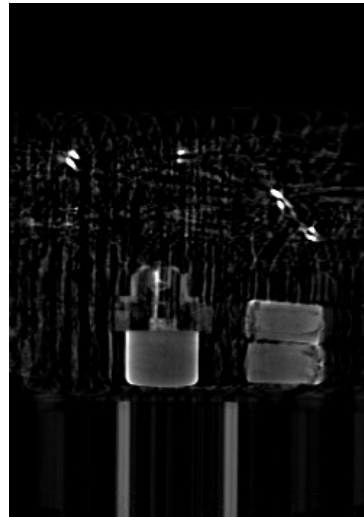
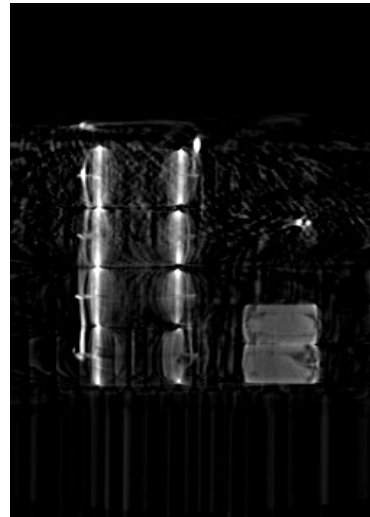
(45 views over 360° with an angular interval of 8°)



TV method



Existing FDK method



Slice 1

Slice 2

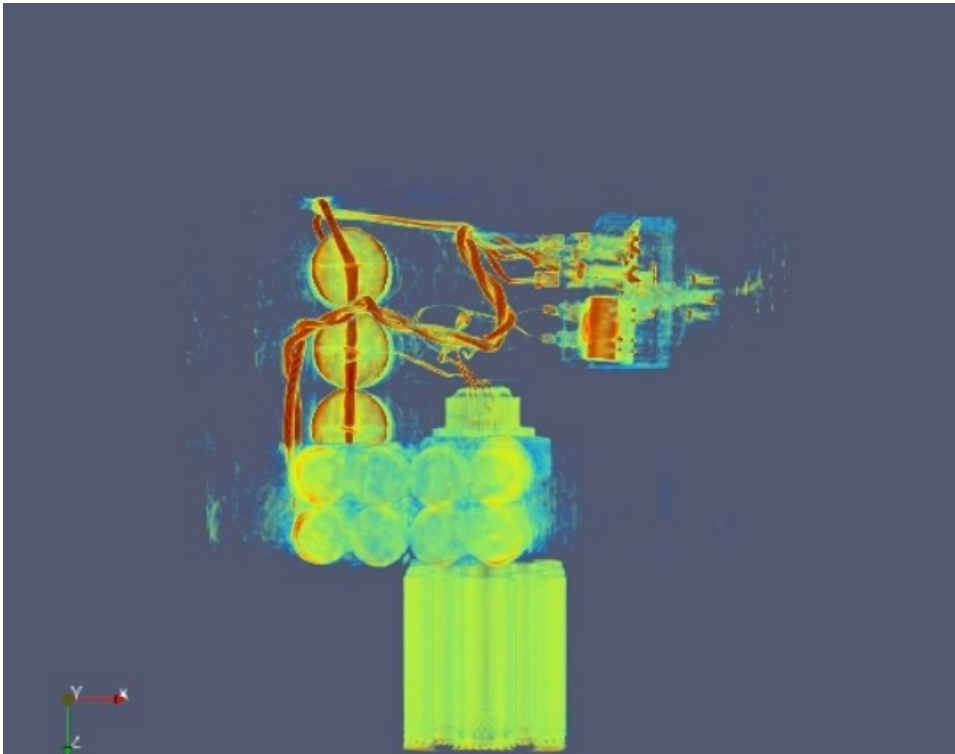
Slice 3

MIP

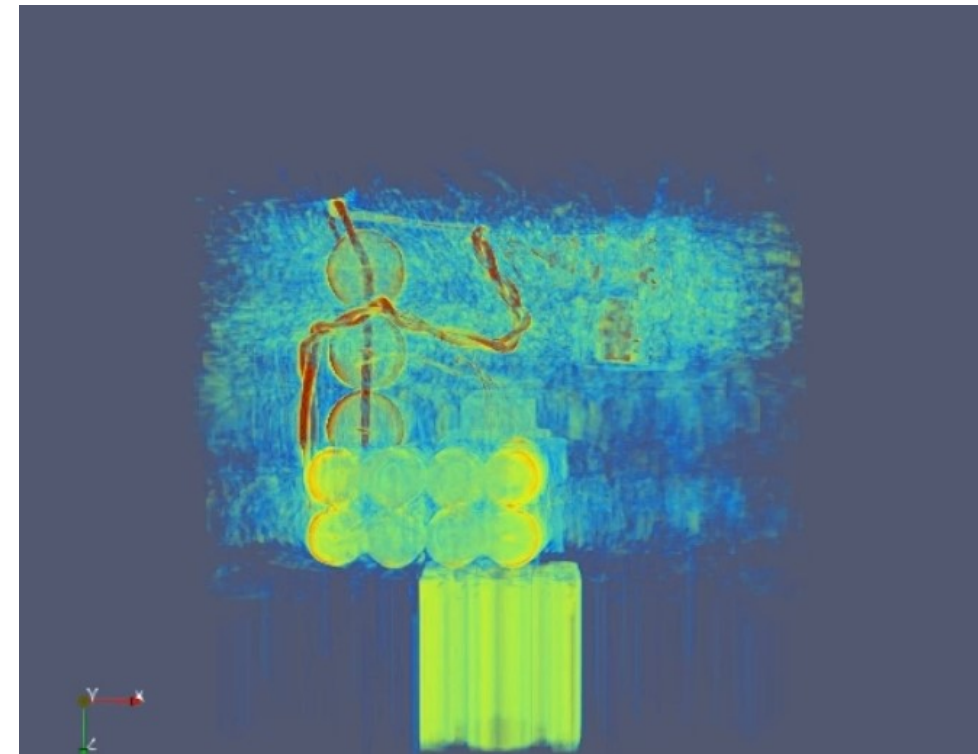


Rendered CT Images

(45 views over 360° with an angular interval of 8°)



TV method

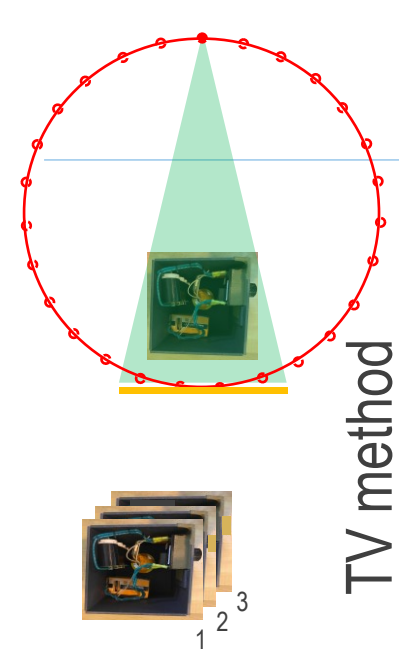


Existing FDK method

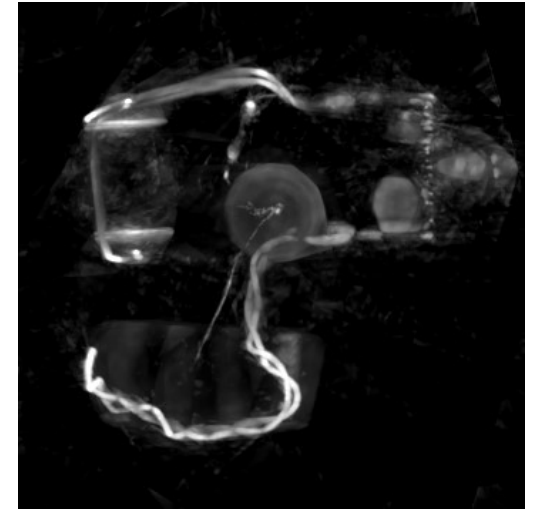
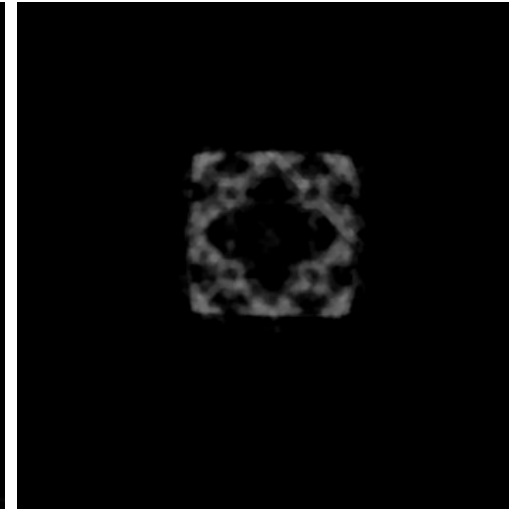
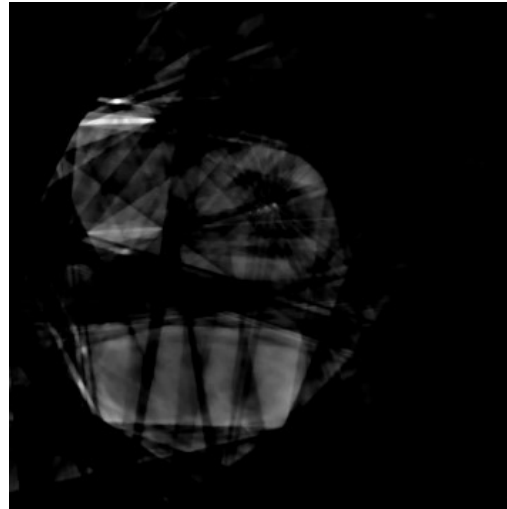
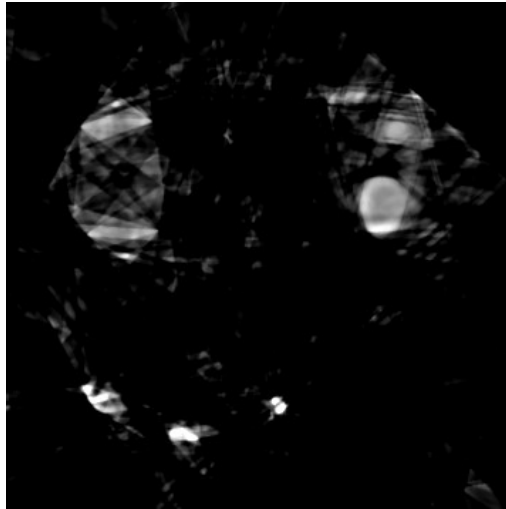


Transverse Slices and MIP Images

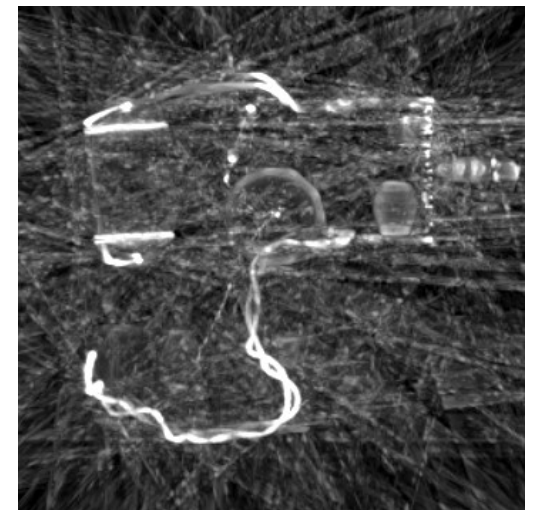
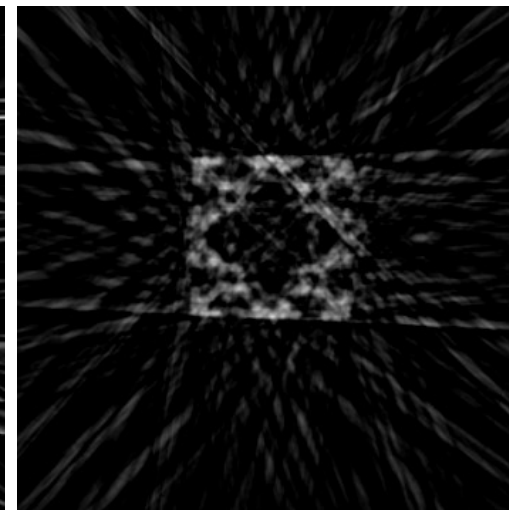
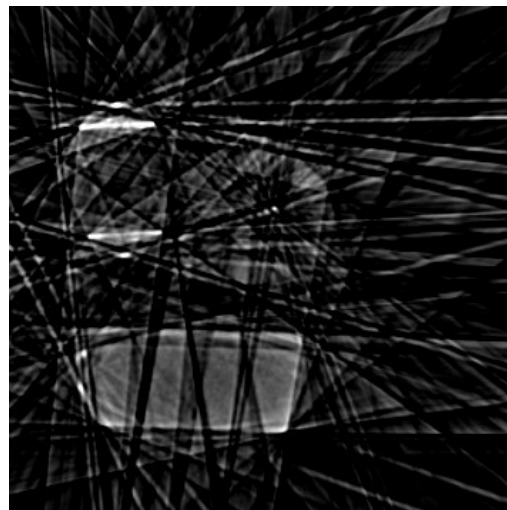
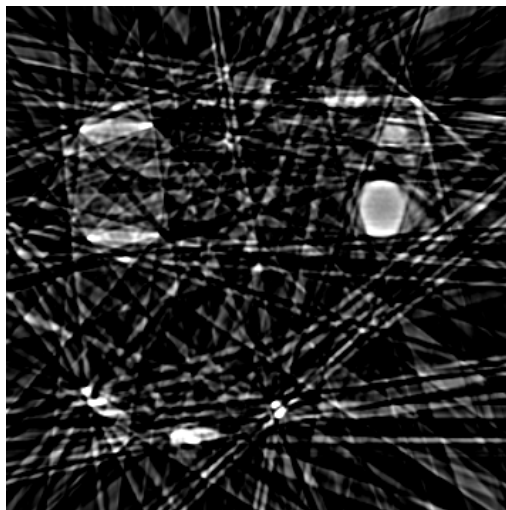
(12 views over 360° with an angular interval of 30°)



TV method



Existing FDK method



Slice 1

Slice 2

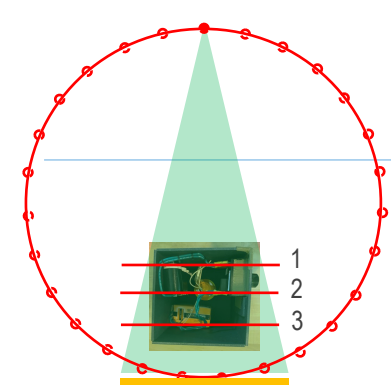
Slice 3

MIP

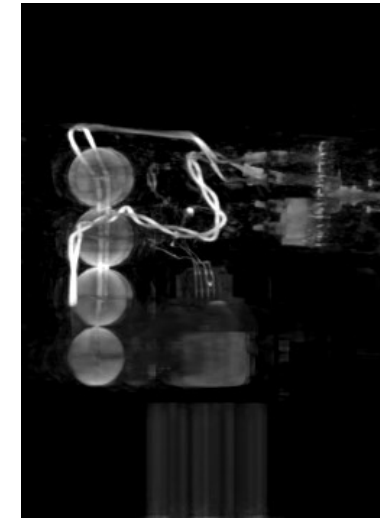
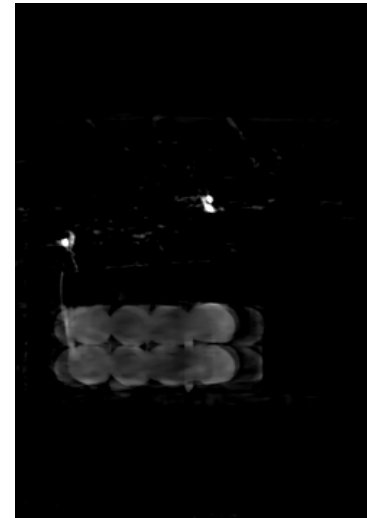
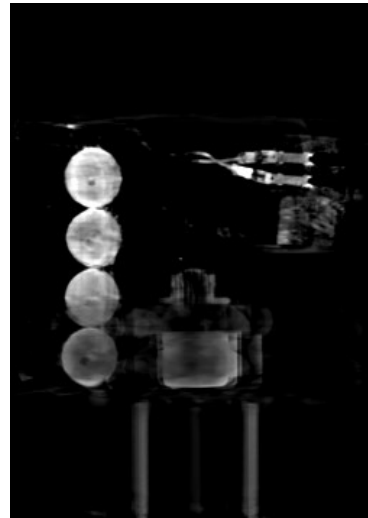
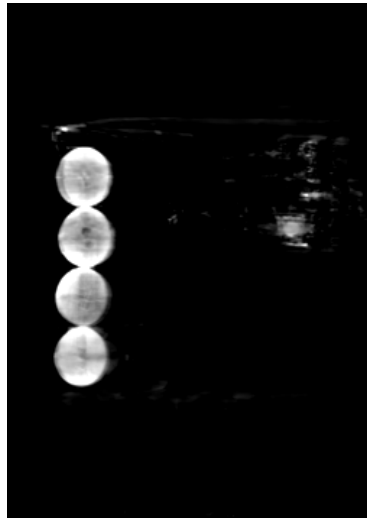


Coronal Slices and MIP Images

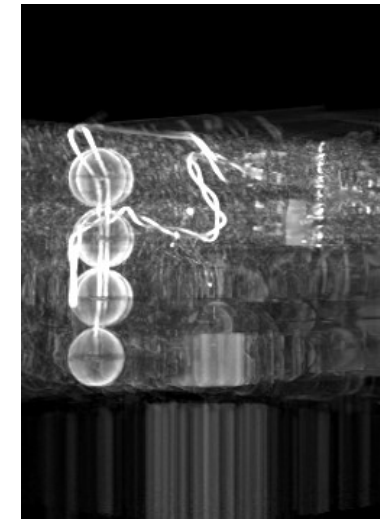
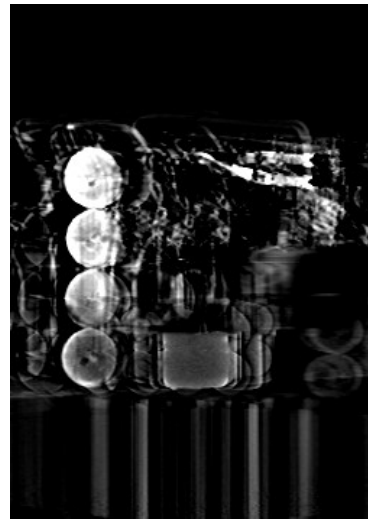
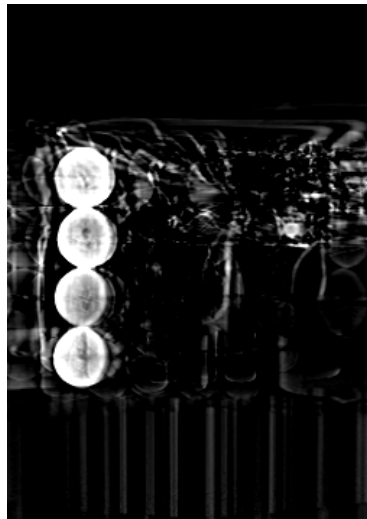
(12 views over 360° with an angular interval of 30°)



TV method



Existing FDK method



Slice 1

Slice 2

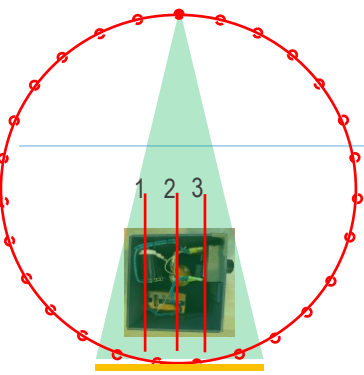
Slice 3

MIP

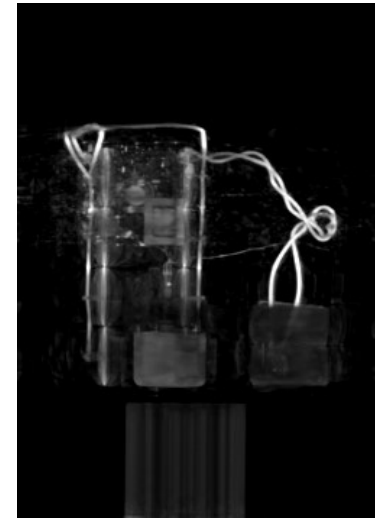
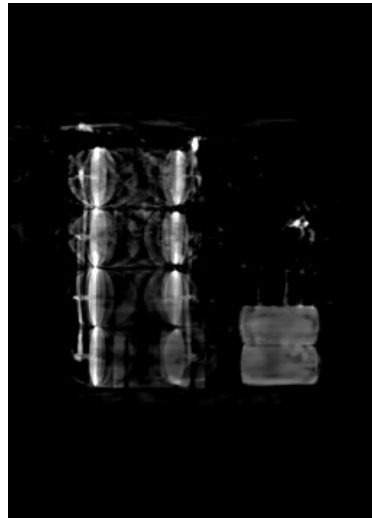


Sagittal Slices and MIP Images

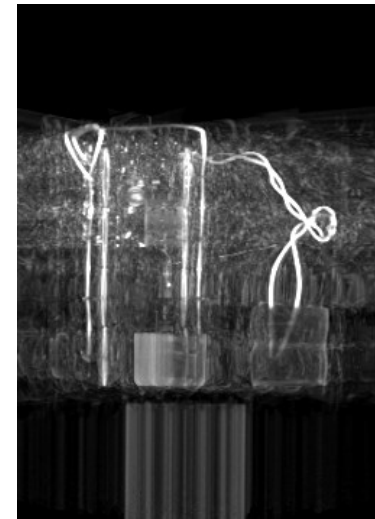
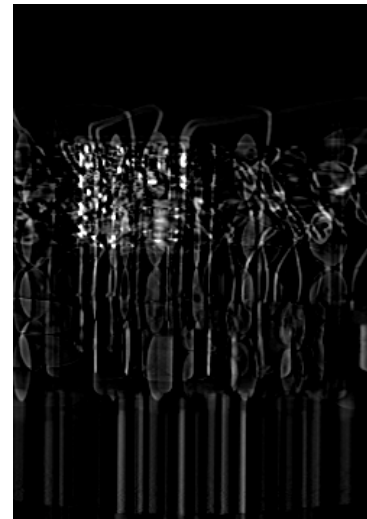
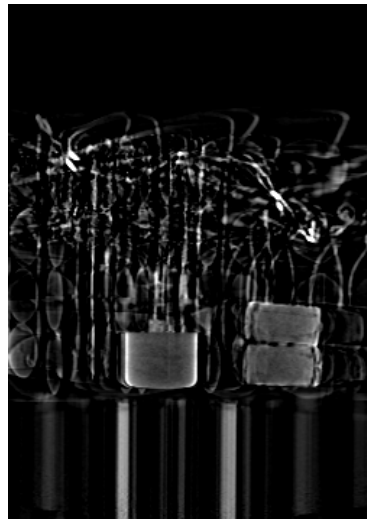
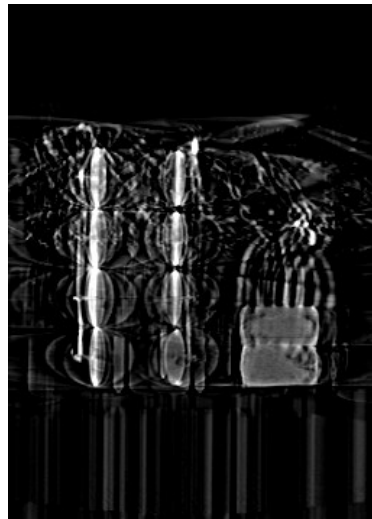
(12 views over 360° with an angular interval of 30°)



TV method



Existing FDK method



Slice 1

Slice 2

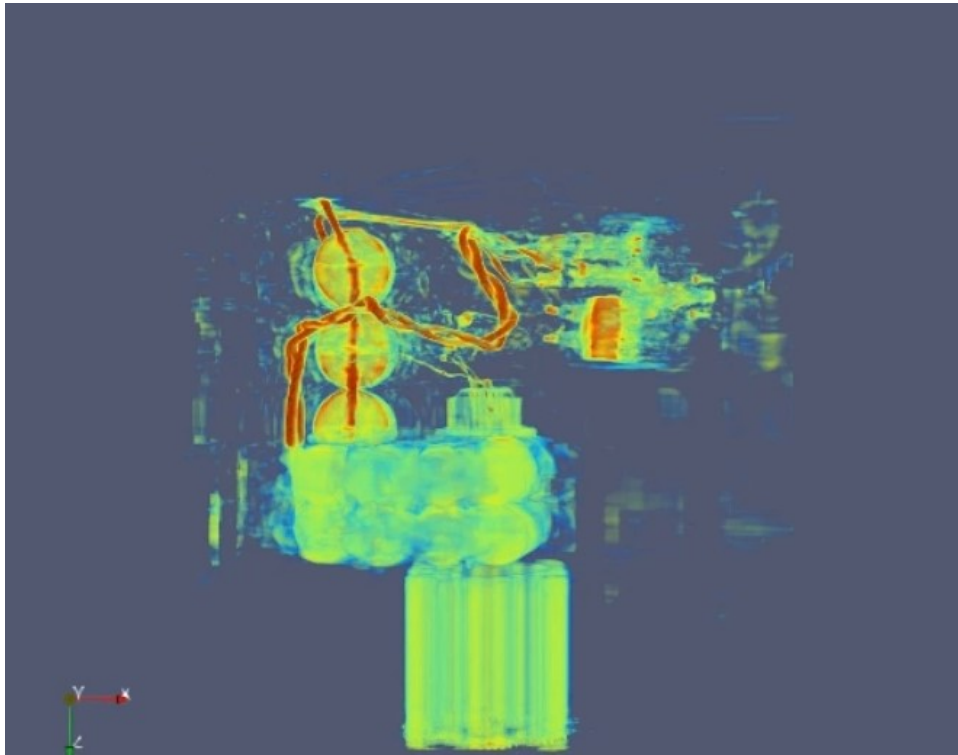
Slice 3

MIP

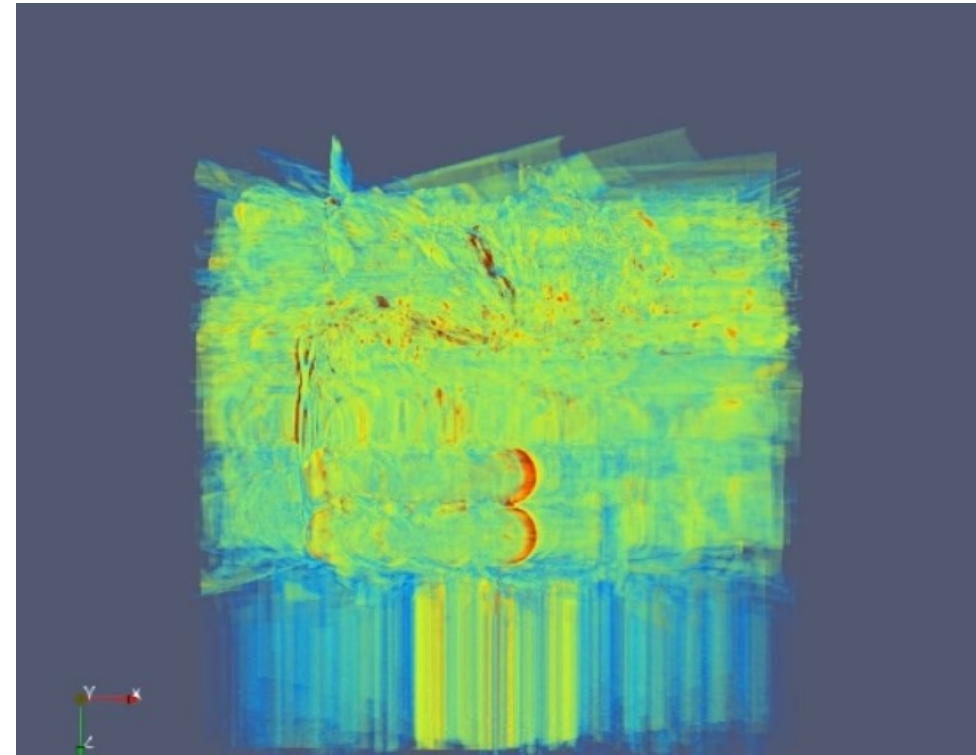


Rendered CT Images

(12 view over 360° with an angular interval of 30°)



TV method



Existing FDK method



Conclusions

- Projection data were collected using a portable X-ray imaging system, with a hand-made phantom of practical relevance
- Developed and tailored a TV algorithm to reconstruct 3D CT images from sparse-view data collected with a portable X-ray imaging system
- Demonstrated that the TV algorithm can yield images with significantly reduced artifacts compared to those obtained with existing algorithms
- Enabled faster data acquisition and supported the use of portable CT imaging in field operations, security screening, law enforcement, and emergency response scenarios



Thank You!

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