

Application of iterative reconstruction algorithms to limited-angle tomography



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MOTIVATION

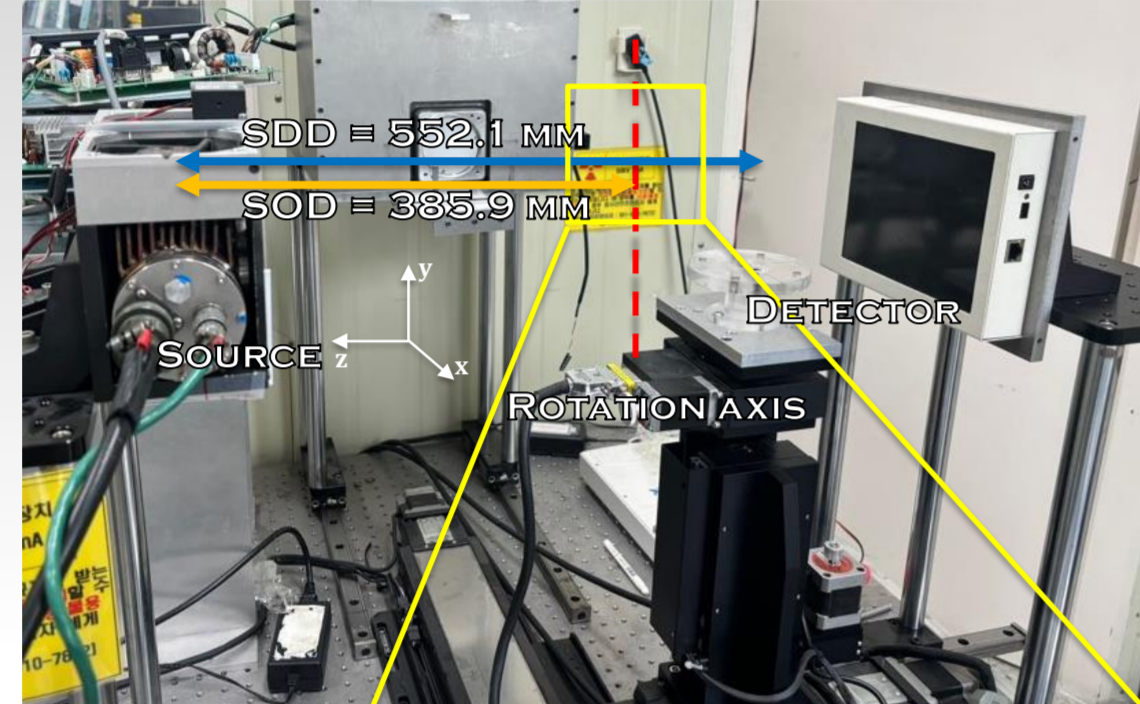
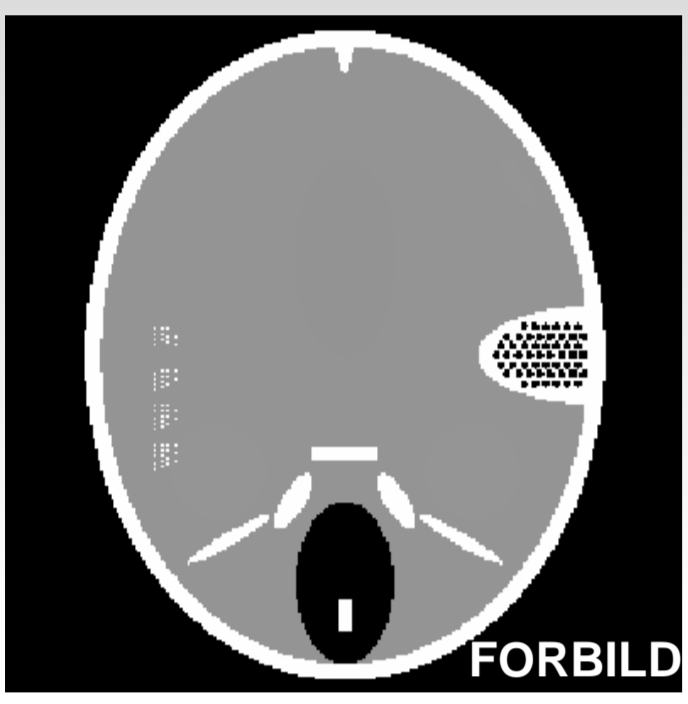
- Limited-angle tomography (LAT) is a solution for inspecting the objects when the full 360° scanning is not feasible
- However, the incomplete data acquired from less than 180° scans introduces out-of-plane and streak artifacts in the reconstructed images, which are intensified in conventional methods such as the filtered backprojection (FBP)
- Iterative reconstruction (IR) methods can be an alternative to reduce out-of-plane artifacts compared to FBP, but they may still be insufficient, leading this study to investigate IR algorithms with prior images

OBJECTIVES

- To apply various IR methods to LAT, given as the system of linear equations (with and without regularization) and statistical maximum likelihood
- To compare the performance of IR methods using ideal detectability, artifact-spread function (ASF), and structural similarity index measure (SSIM)
- To compare the reconstruction results of IR and prior-image constrained IR for simulation data

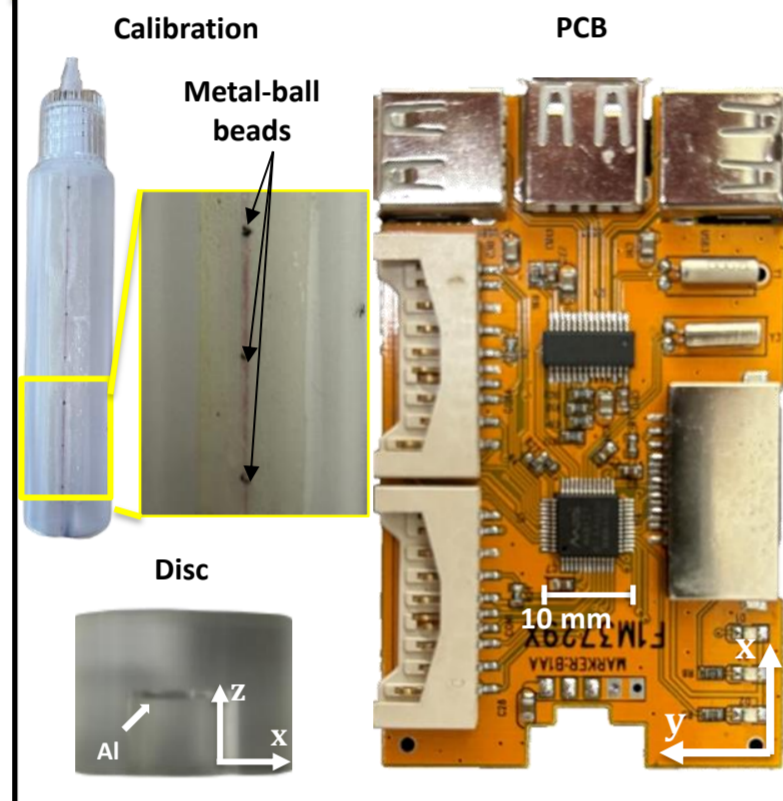
MATERIALS AND METHODS

Simulation phantom Experimental setup



CT system specifications

Category	Details
X-ray	45 kVp, 0.1593 mGy/s
Detector	Shad-o-Box HS, CMOS x-ray sensor Pixel format: 1548 × 1032 Pixel pitch: 99 μm Readout time: 200 ms
Scan	Scan range (α): 30°, 60°, 90°, 120° Step angle (β): 1°, 2°, 3°, 5°



Algorithms used for reconstruction

- Feldkamp, Davis, Kress (FDK), Hann filter**
$$x = \int \frac{1}{\lambda^2} \int_{-\infty}^{\infty} \frac{LSD}{\sqrt{L^2_D + \xi^2 + \zeta^2}} b_{\theta}(\xi, \zeta) * h(\xi' - \xi) d\zeta d\theta$$
- Simultaneous algebraic reconstruction technique (SART)**
$$x^{k+1} = x^k + \lambda \frac{A^T(b - Ax^k)}{A^T A 1}$$
- Conjugate gradient least squares (CGLS)**
$$r_0 = b - Ax^0, p_0 = A^T r_0$$

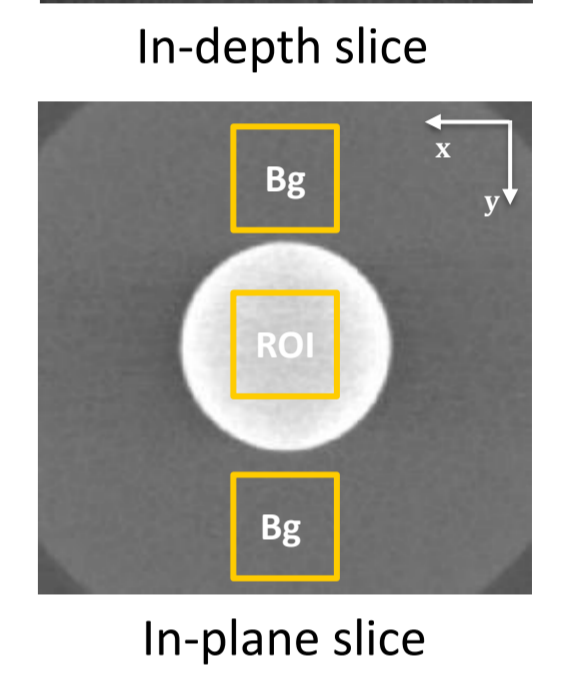
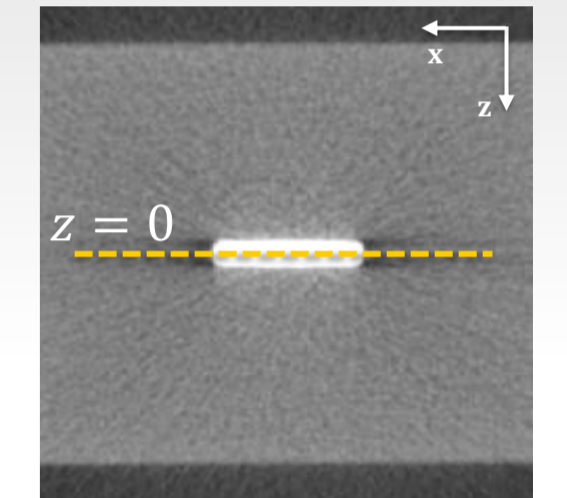
$$x^{k+1} = x^k + \frac{\|p_k\|_2}{\|Ax^k\|_2} p_k$$

$$r_{k+1} = r_k - \frac{\|p_k\|_2}{\|Ax^k\|_2} A p_k$$
- Maximum likelihood-expectation maximization (MLEM)**
$$p_{k+1} = A^T r_{k+1} + \frac{\|A^T r_{k+1}\|_2^2}{\|A^T r_k\|_2^2} p_k$$
- SART with total variation regularization (SART-TV)**
$$x^* = \arg \min \|x\|_{TV}, \text{ subject to } Ax = b$$
- Prior image constrained compressed sensing (PICCS)**
$$x^* = \arg \min ((1 - \alpha) \|x - x_p\|_{TV} + \alpha \|x\|_{TV})$$
- Stopping criteria**
when $\|Ax^{k+1} - b\|_2 > \|Ax^k - b\|_2$

Evaluation method

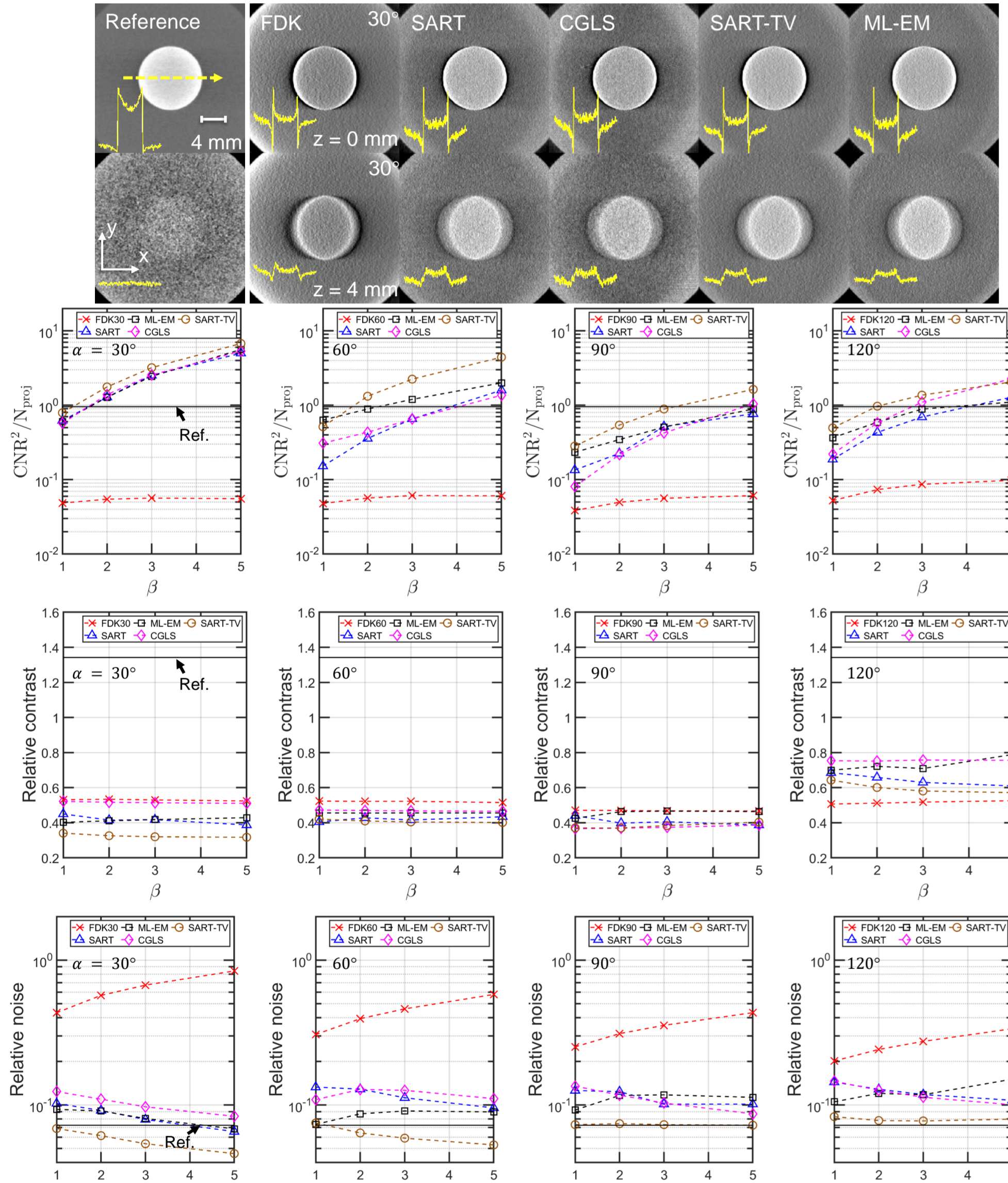
- Contrast-to-noise ratio (CNR)**
 - Calculate the CNR of a 0.5 mm thick Al disc phantom
 - Normalize the squared CNR by the number of projection views used for reconstruction, which is equivalent to the **ideal detectability**
 - $$CNR = \frac{\bar{\mu}_{ROI}(0) - \bar{\mu}_{BG}(0)}{\sigma}, \text{ where } \sigma = \sqrt{\frac{\sigma_{ROI}^2 + \sigma_{BG}^2}{2}}$$
- Relative contrast**
 - Calculate the relative contrast in an in-plane (x-y) image of the Al disc phantom
 - $$RC = \frac{\bar{\mu}_{ROI} - \bar{\mu}_{BG}}{\bar{\mu}_{ROI} + \bar{\mu}_{BG}}$$
- Relative noise**
 - Calculate the relative noise in an in-plane (x-y) image of the Al disc phantom
 - $$RN = \frac{\sqrt{2} \sqrt{\frac{\sigma_{ROI}^2 + \sigma_{BG}^2}{2}}}{\bar{\mu}_{ROI} + \bar{\mu}_{BG}}$$
- Artifact-spread function (ASF)**
 - Estimate the streak artifact from the ASF obtained for an Al disc phantom
 - $$ASF(z) = \frac{\bar{\mu}_{ROI}(z) - \bar{\mu}_{BG}(z)}{\bar{\mu}_{ROI}(0) - \bar{\mu}_{BG}(0)}$$
- Structural similarity index measure (SSIM)**
 - $$SSIM(x, y) = \frac{(2\mu_x \mu_y + c_1)(\sigma_{xy} + c_2)}{(\mu_x^2 + \mu_y^2 + c_1)(\sigma_x^2 + \sigma_y^2 + c_2)}$$
 - μ : the pixel sample mean of x and y (kernel)
 - σ : the variance of x and y
 - σ_{xy} : the covariance of x and y

Disc phantom



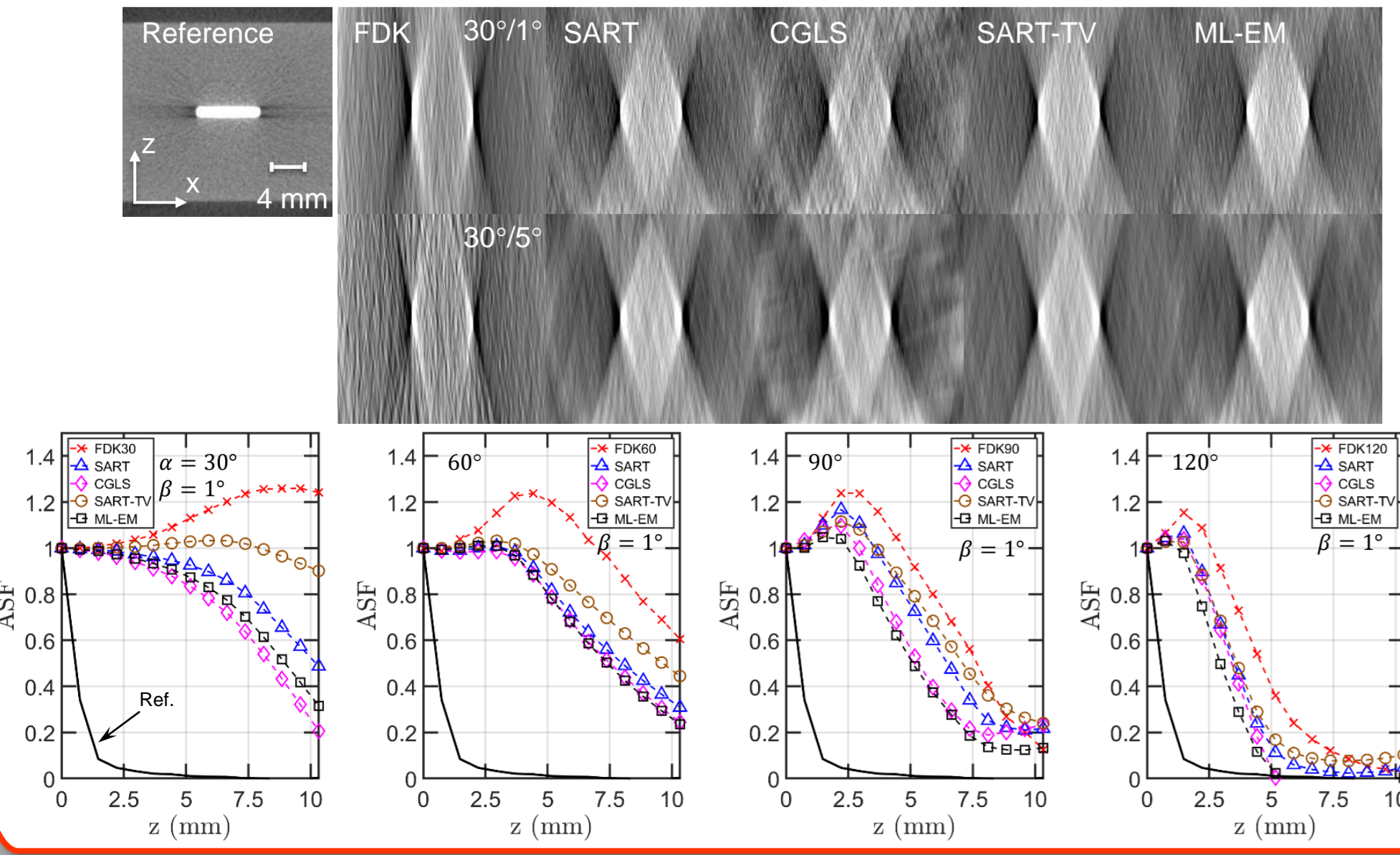
RESULTS

Ideal detectability



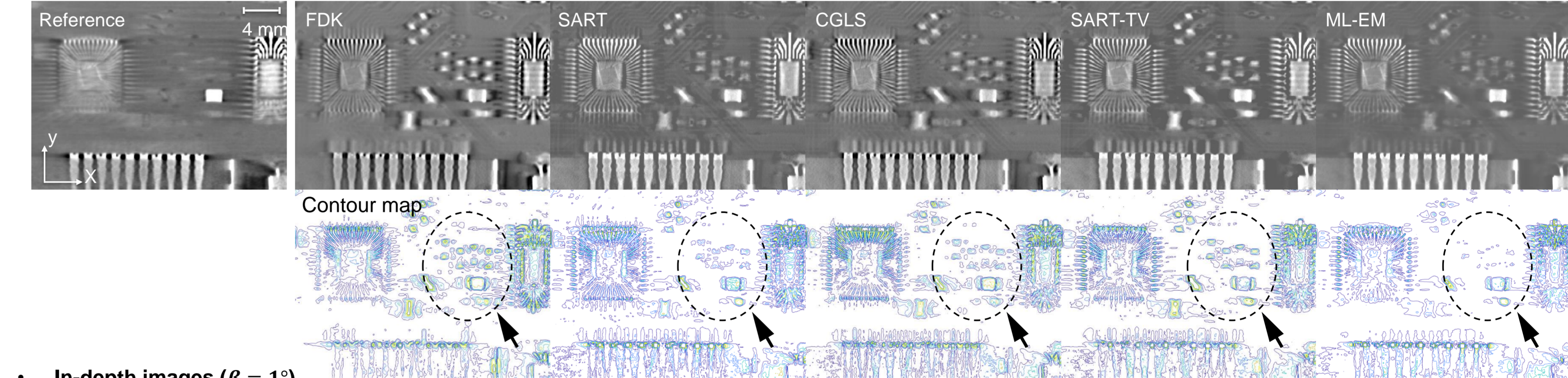
- The ideal detectability decreases as the α increases and β decreases
- IR methods demonstrate higher ideal detectability than FDK
- In terms of relative contrast, there is no significant difference between algorithms except for α = 120°
- The IR methods show better noise performance than FDK, with SART-TV being the best

Artifact-spread function

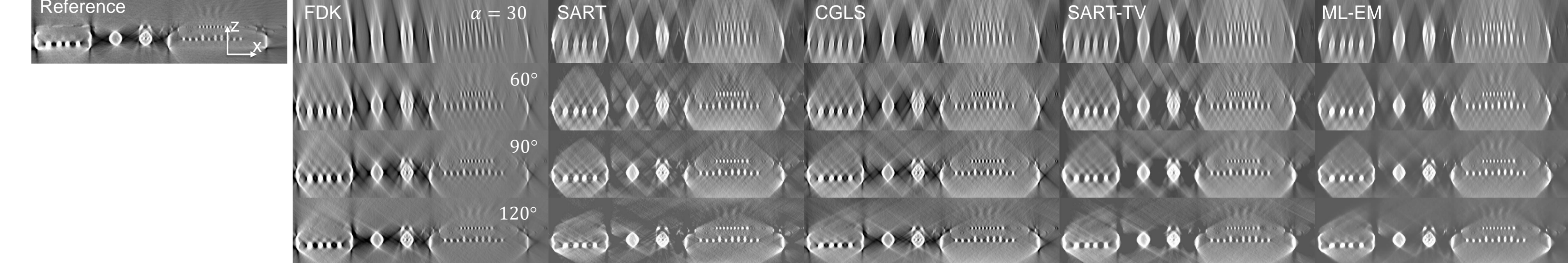


PCB reconstruction result

In-plane images (α = 60°, β = 1°)



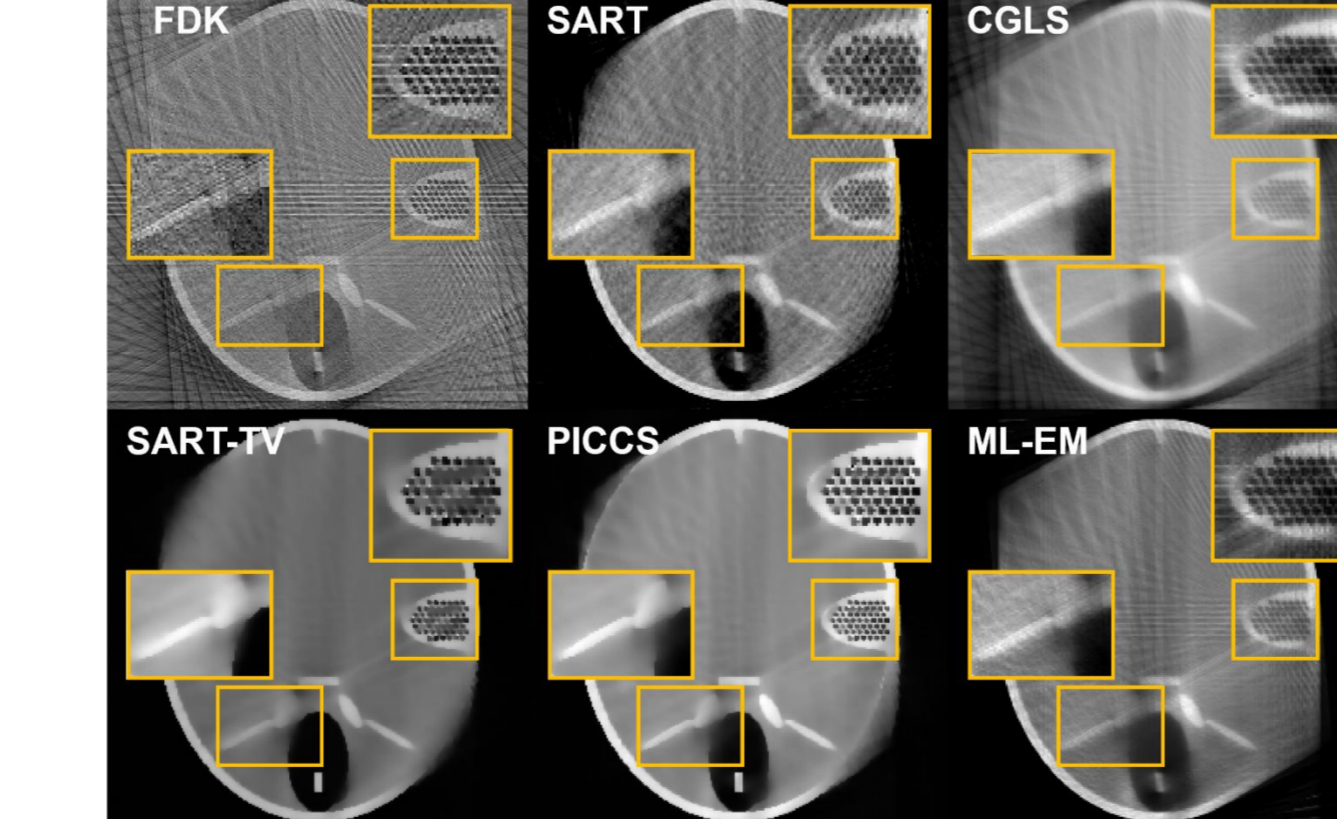
In-depth images (β = 1°)



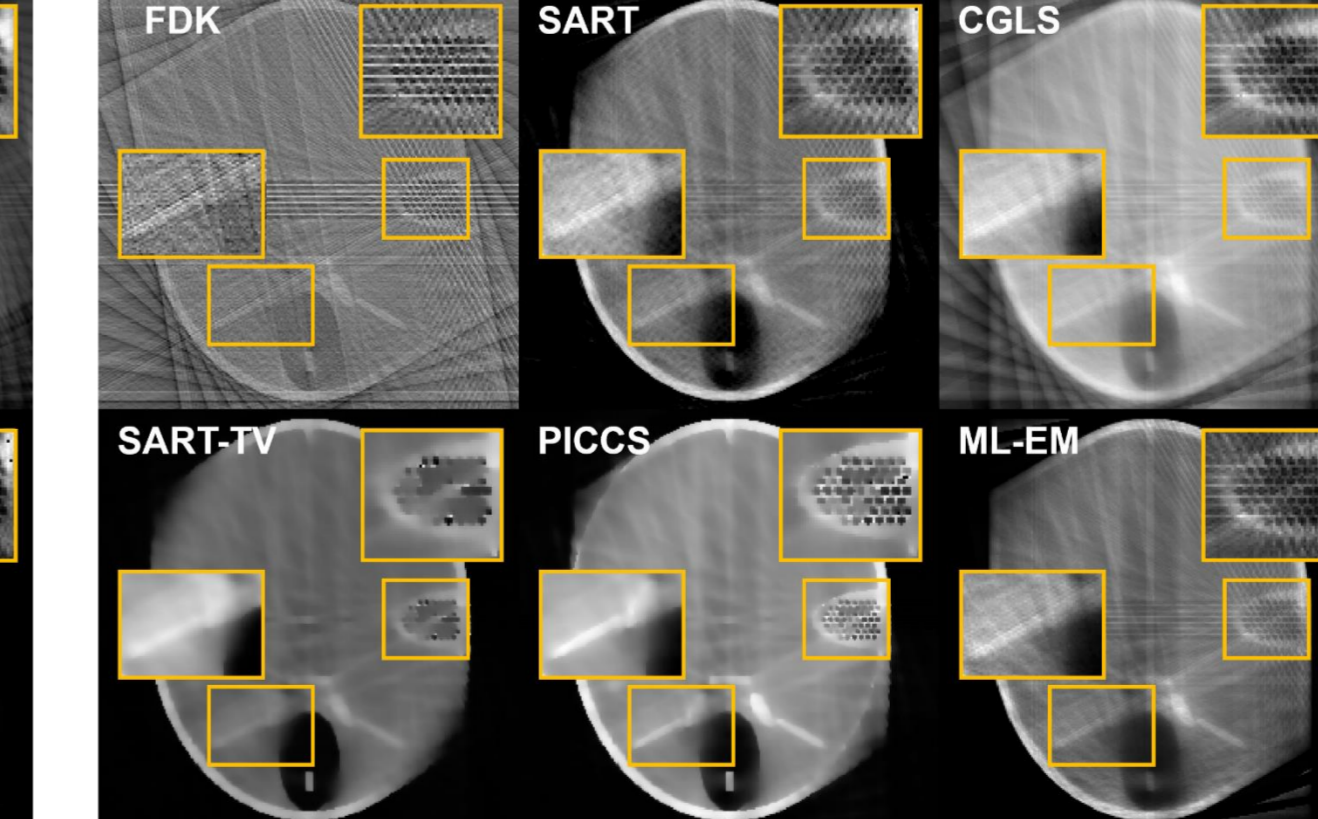
- IR methods shows less out-of-plane artifacts, and among IR methods the ML-EM is the best; however, the difference is marginal
- The out-of-plane artifacts can be well seen in in-depth view, and the artifacts decrease as the α increases

FORBILD reconstruction result

α = 120°, β = 5°



α = 120°, β = 10°



β = 5°/10°	FDK	SART	CGLS	SART-TV	PICCS	ML-EM
SSIM	0.077 / 0.052	0.471 / 0.469	0.252 / 0.209	0.747 / 0.631	0.762 / 0.664	0.524 / 0.504
MSE	0.693 / 1.477	0.038 / 0.047	0.092 / 0.103	0.029 / 0.041	0.028 / 0.038	0.055 / 0.060

- The IR methods are superior to the FDK but still suffer from the streak artifacts
- The prior image used algorithm (PICCS) shows the well-preserving detail of the phantom and shows the best performance among all methods

DISCUSSION

- The ideal detectability increases as the α increases, which is one of the advantages of LAT
- Unlike our expectation that all IR algorithms would outperform the FDK in every aspect, the streak artifacts in particular, the overall performance of SART, CGLS, and SART-TV are comparable to that of FDK in experiments
- However, in terms of noise, IR methods generally offer superior performance
- PICCS, which utilizes the prior image, reduces artifacts and demonstrates a significant advantage in preserving structural details