

Measurement-Based Spectral Deep Scatter Estimation for Photon-Counting CT

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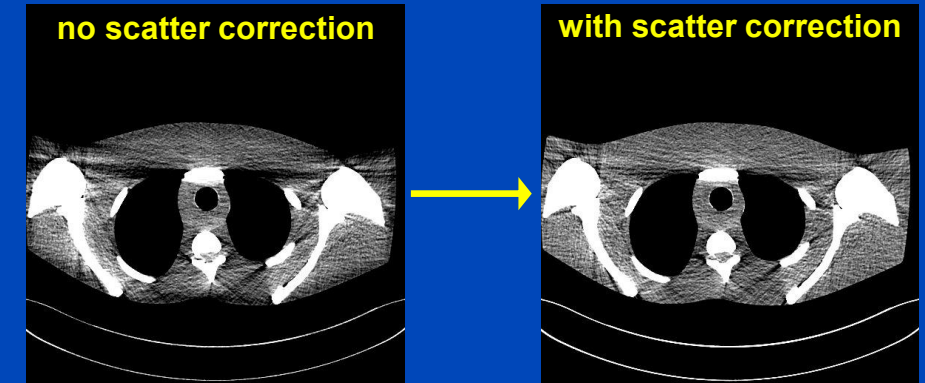
³Ruprecht-Karls-Universität, Heidelberg, Germany

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Motivation

- Scatter degrades image quality: loss of contrast, cupping artifacts, streaks and inaccurate quantitative values.
- Accurate scatter estimation is crucial for spectral CT and quantitative imaging tasks.
- Software-based scatter correction methods:

Method	Speed	Accuracy
Analytical models ¹	+ Fast	○ Moderate
Monte Carlo simulations	– Slow	+ High
Deep learning-based ^{2,3}	+ Fast	+ High (depends on data)



Kernel Qr40f, C = 40 HU, W = 200 HU

**Deep learning approaches rely on large, realistic datasets.
However, Monte Carlo simulations may not fully reflect system-specific physics.
Using real measured data are alternative and may be the more practical choice for training.**

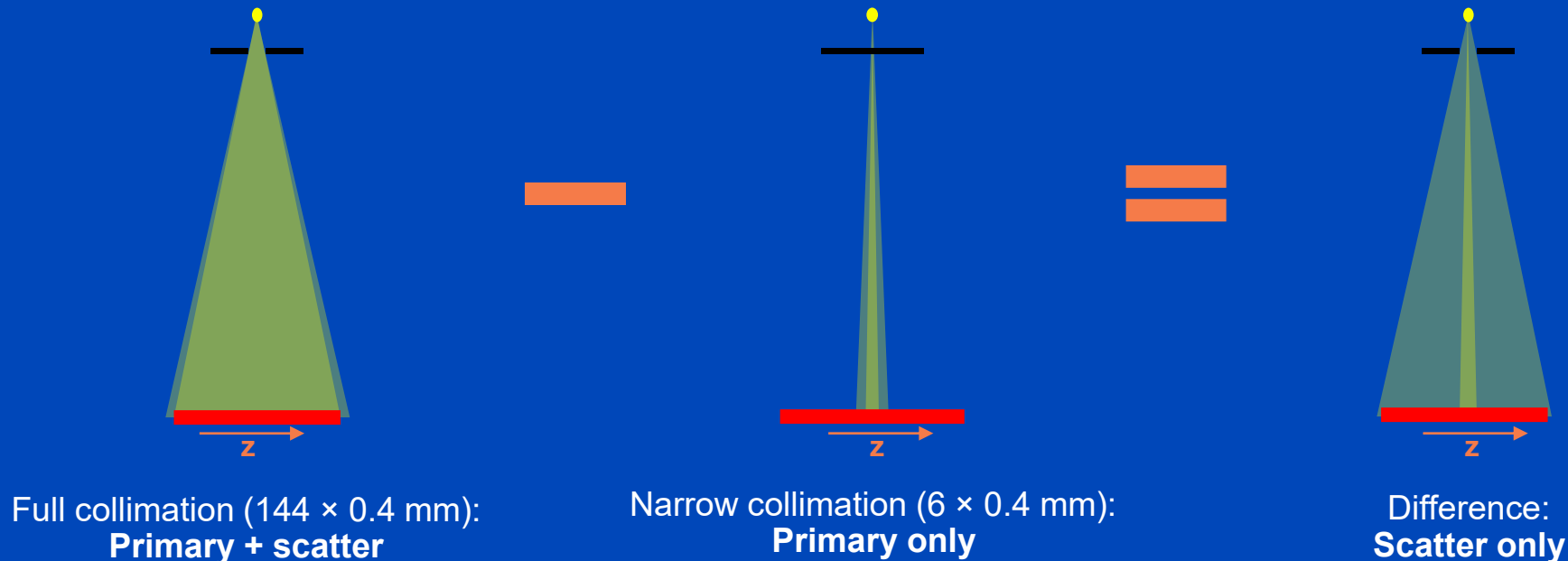
¹B. Ohnesorge, K. Klingenberg-Regn et al. "Efficient object scatter correction algorithm for third and fourth generation CT scanners" European Radiology 9, 563–569, March 1999.

²J. Maier, M. Kachelrieß et al. "Deep Scatter Estimation (DSE)", SPIE 2018 and J. of Nondestruct. Eval. 37:57, July 2018.

³J. Maier, M. Kachelrieß et al. "Robustness of DSE", Med. Phys. 46(1):238-249, January 2019.

Methodology

- Performing two scans:
 - Full collimation, contains scatter and primary intensity
 - Narrow collimation, contains primary intensity as the scatter should be negligible
- Scatter intensity:
 - Difference between the intensities of full and narrow collimated data is used as label
 - Only the data within the primary fan of the narrow collimated scan are used



Measurement-Based Training Data Generation

- Naeotom Alpha.Peak (Siemens Healthineers, Forchheim, Germany)
- Several phantoms (water phantoms, multi-energy QA phantoms, and anthropomorphic patient-like phantoms) are used in the data set.
- Automated script that moves the patient table in predefined steps in z-direction
- The table height is chosen randomly around the isocenter.
- Two scans at each position
 - $6 \times 0.4 \text{ mm} = 2.4 \text{ mm}$
 - $144 \times 0.4 \text{ mm} = 57.6 \text{ mm}$
- Default bowtie filter
- Energy thresholds at 20 keV and 65 keV
- 140 kV tube voltage and 100 mAs / rotation

Total data set: 5760 projections

- 36 projections per z-position
- 160 different z-positions



Measurement-Based Spectral DSE (mbsDSE)

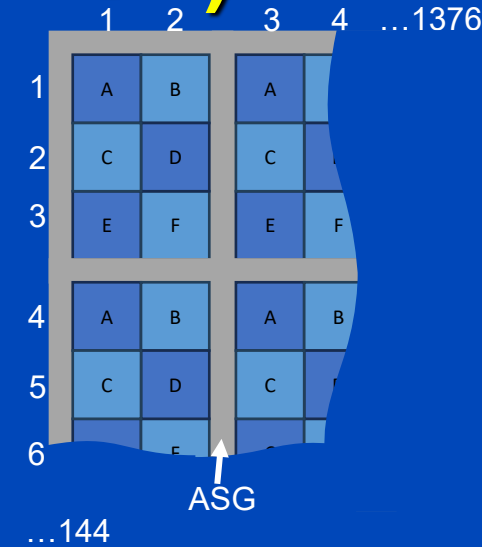
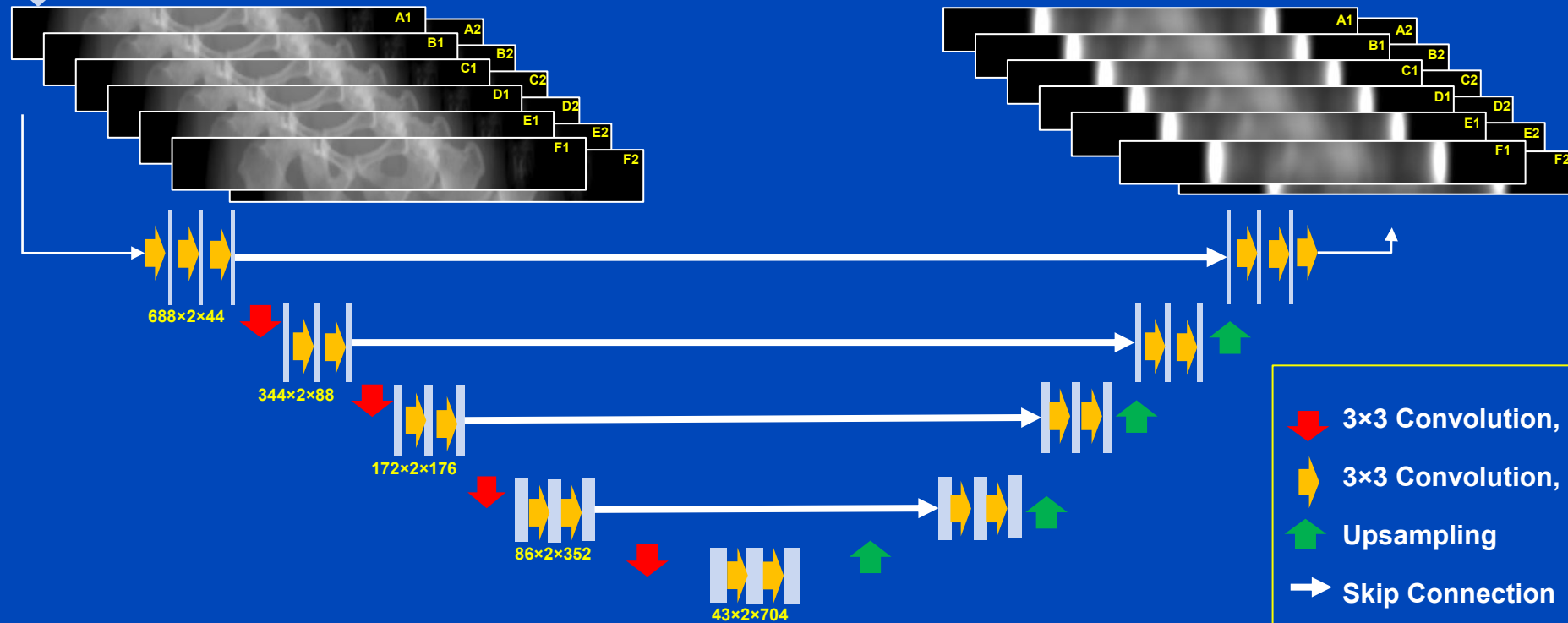
Detector dimension 1376×6

Each channel corresponds to a different pixel position between the lamellae of the ASG

merging 6 pixel positions

Input: 12 channels
(2 energy threshold \times 6 pixel positions)
Dimension: 688×2

Output: 12 channels



- 3x3 Convolution, Stride 2
- 3x3 Convolution, Stride 1
- Upsampling
- Skip Connection

[1] J. Maier, M. Kachelrieß et al. "Deep Scatter Estimation (DSE)", SPIE 2018 and J. of Nondestruct. Eval. 37:57, July 2018.

[2] J. Maier, M. Kachelrieß et al. "Robustness of DSE", Med. Phys. 46(1):238-249, January 2019.

[3] J. Erath, M. Kachelrieß et al. "Monte-Carlo-Free Deep Scatter Estimation (DSE) for X-Ray CT and CBCT", RSNA 2019.

[4] J. Erath, T. Vöth, J. Maier, E. Fournié, M. Petersilka, K. Stierstorfer, and M. Kachelrieß, "Deep Scatter Correction in DSCT", CT Meeting August 2020.

[5] J. Erath, T. Vöth, J. Maier, E. Fournié, M. Petersilka, K. Stierstorfer, and M. Kachelrieß, "Deep Learning-Based Forward and Cross-Scatter Correction in DS CT" Med. Phys. 2021.

RESULTS

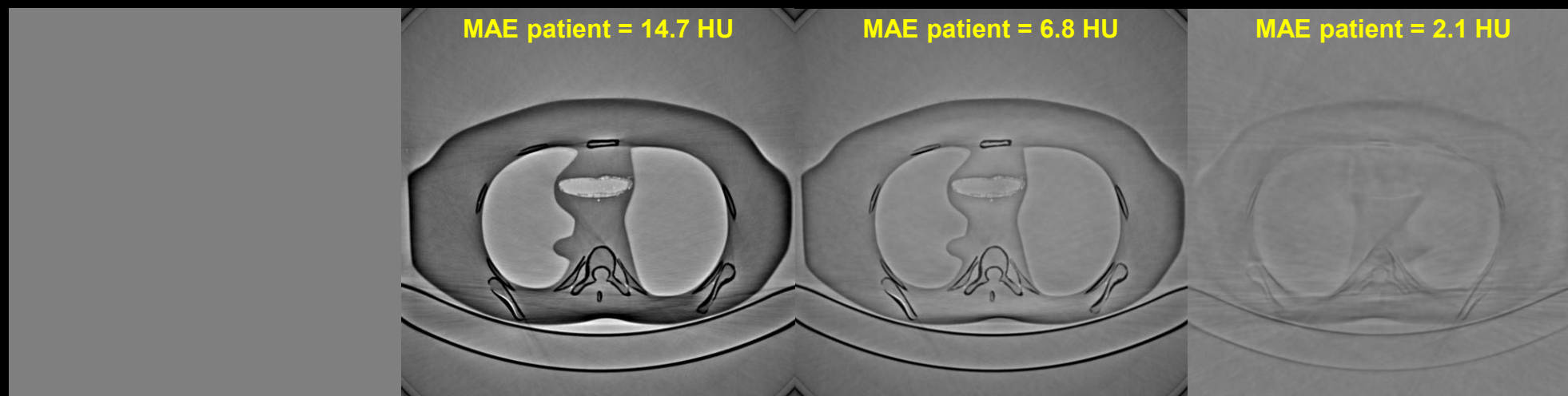
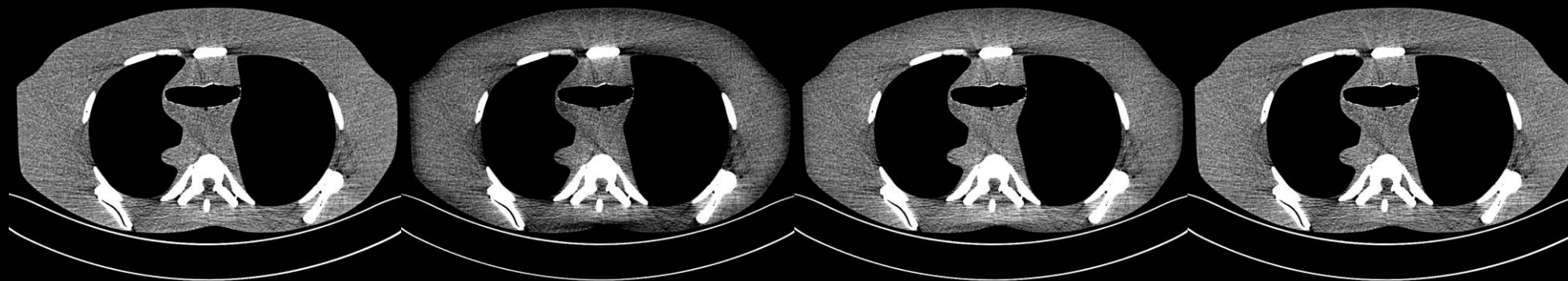
Low Energy Threshold Image

ground truth

uncorrected

analytical
model

mbsDSE



DSE was only applied to the innermost 6 detector rows and the images shown are reconstructions from these rows

Kernel Qr40f, reconstructed images $C = 60$ HU, $W = 120$ HU

difference images $C = 0$ HU, $W = 100$ HU

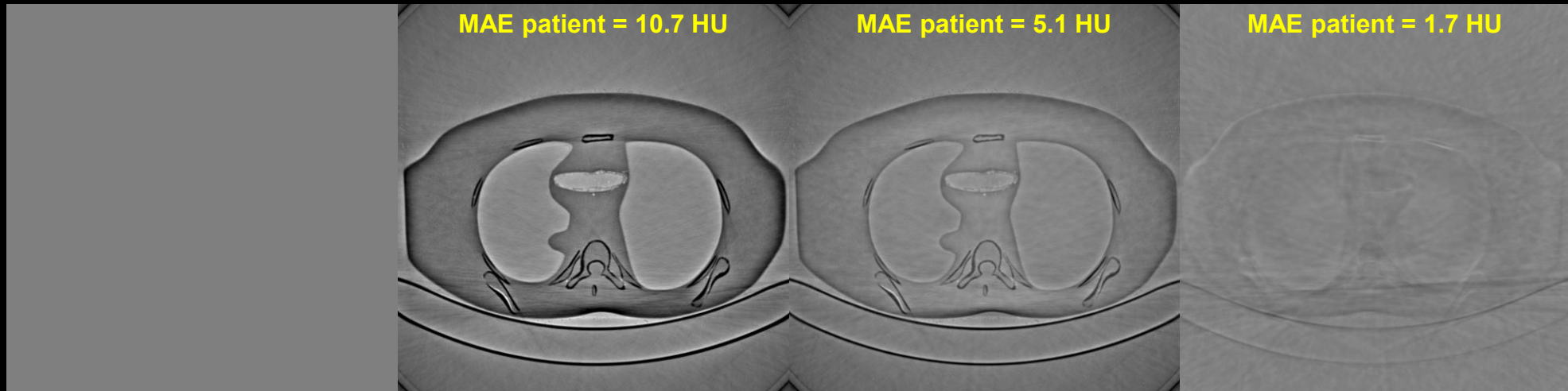
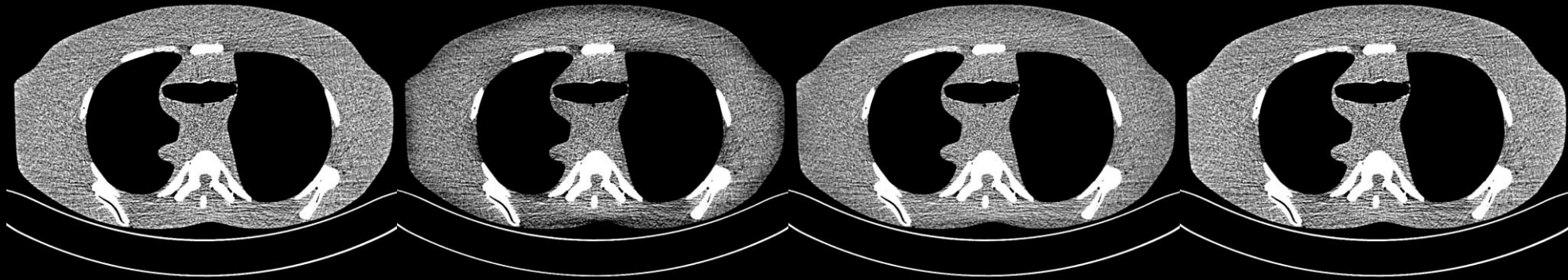
High Energy Threshold Image

ground truth

uncorrected

analytical
model

mbsDSE



DSE was only applied to the innermost 6 detector rows and the images shown are reconstructions from these rows

Kernel Qr40f, reconstructed images $C = 60$ HU, $W = 120$ HU

difference images $C = 0$ HU, $W = 100$ HU

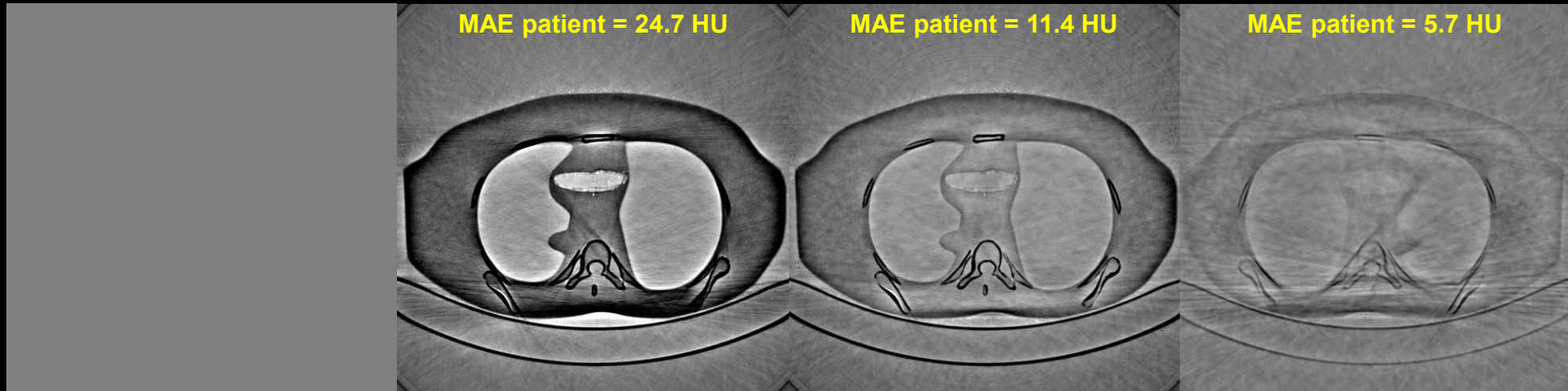
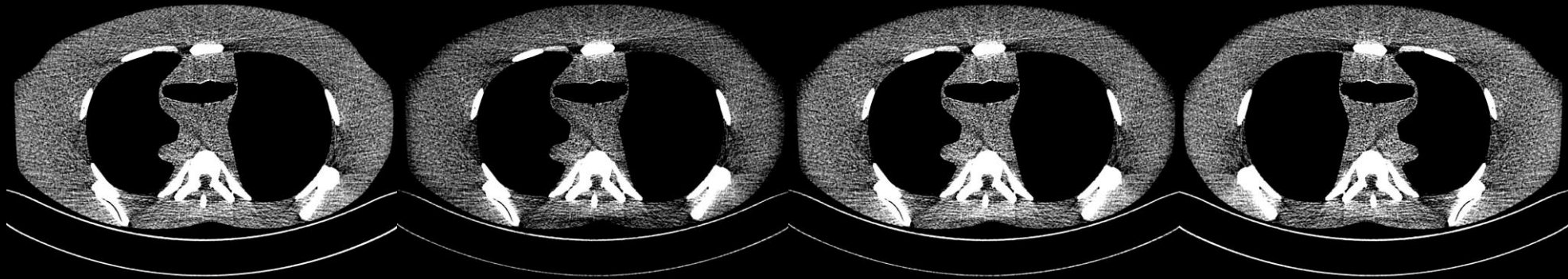
VMI 50 keV Image

ground truth

uncorrected

analytical
model

mbsDSE



DSE was only applied to the innermost 6 detector rows and the images shown are reconstructions from these rows

Kernel Qr40f, reconstructed images $C = 40$ HU, $W = 200$ HU

difference images $C = 0$ HU, $W = 100$ HU

Conclusions & Outlook

- **Conclusions**

- mbsDSE achieves substantially stronger MAE reductions (77 to 86%) compared to the analytical model (52 to 54 %) across low energy, high energy, and spectral data for the test phantom.
- The largest gains appear in low energy threshold data (86% reduction).

- **Outlook**

- mbDSE is not limited to PCCT but can also be used for EICT.
- A measurement-based trained cross-scatter DSE should also be possible.
- In the future, multiple scans should be used to scan the entire detector and thus obtain a label for the complete detector size.
- The influence of iodine on mbsDSE performance has not yet been investigated, but will be addressed in future work.

Thank You!



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