Deep Scatter Estimation for Coarse Anti-Scatter Grids as used in Photon-Counting CT

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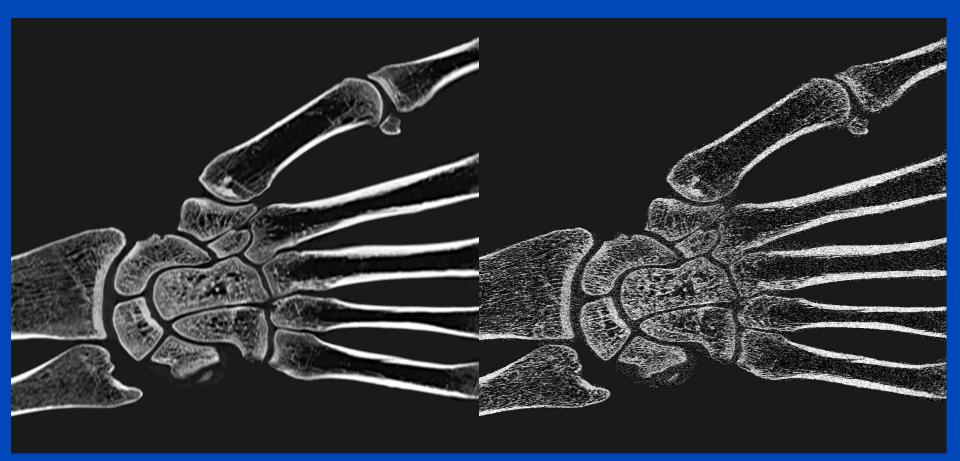
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Photon-Counting enables Visualization of Small Details



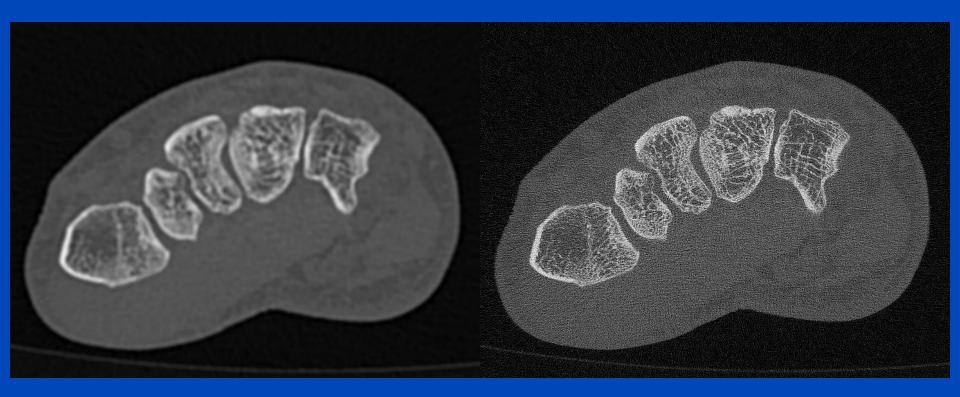
Reconstruction to mimick conventional energy-integrating CT, e.g. Somatom Flash



Reconstruction of data scanned at photon-counting CT Naoetom Alpha



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SIEMENS Healthineers Reconstruction of data scanned at photon-counting CT Naoetom Alpha



Properties of Photon-Counting Detectors

Smaller detector pixels

- can deliver ultra high resolution imaging
- less dose due for conventional spatial resolution¹ ("small pixel effect")
- No electronic noise
 - advanced image quality in obese patients and low-dose scans
- No downweighting of lower energy quanta
 - improved image contrast
 - less dose due to increased iodine CNR² ("iodine effect")
- Intrinsic spectral sensitivy
 - established dual energy applications available in any scan

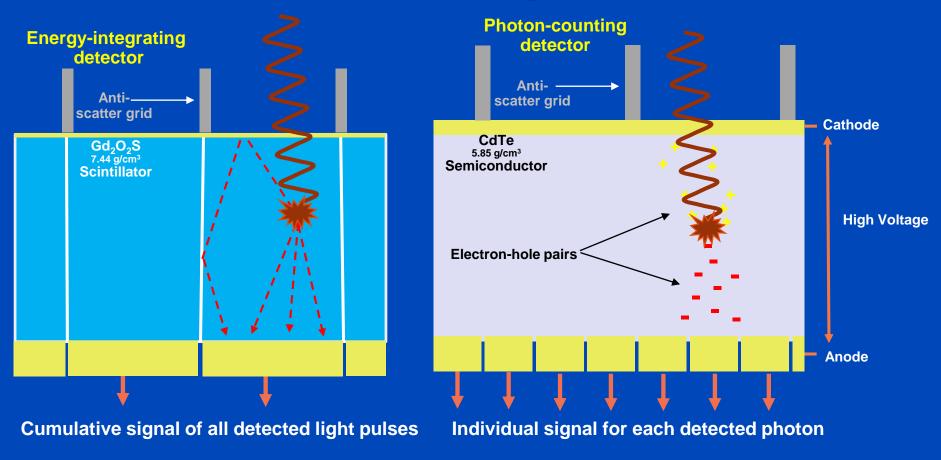
[1] Klein, Kachelrieß, Sawall et al. "Effects of Detector Sampling on Noise Reduction in Clinical Photon-Counting Whole-Body Computed Tomography." *Investigative Radiology* vol. 55(2): 111-119, 2020.

[2] Sawall, Kachelrieß et al. "Iodine Contrast-to-Noise Ratio Improvement at Unit Dose and Contrast Media Volume Reduction in Whole-Body Photoncounting CT." *European journal of radiology* vol. 126: 108909, 2020.





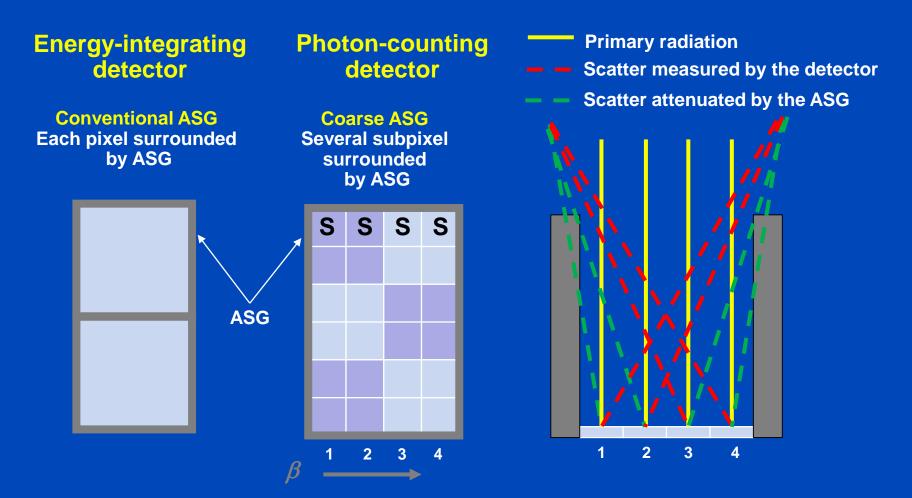
Energy-Integrating vs. Photon-Counting Detectors



Photon-counting detector has several subpixels between lamellae of ASG





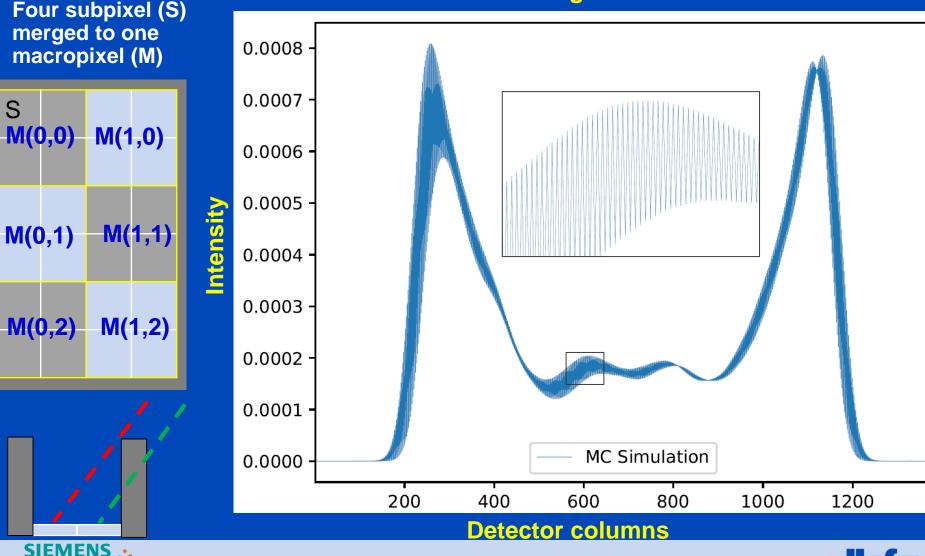


Coarse ASG leads to changing scatter intensity between neighboring pixel depending on the incident angle of the photon

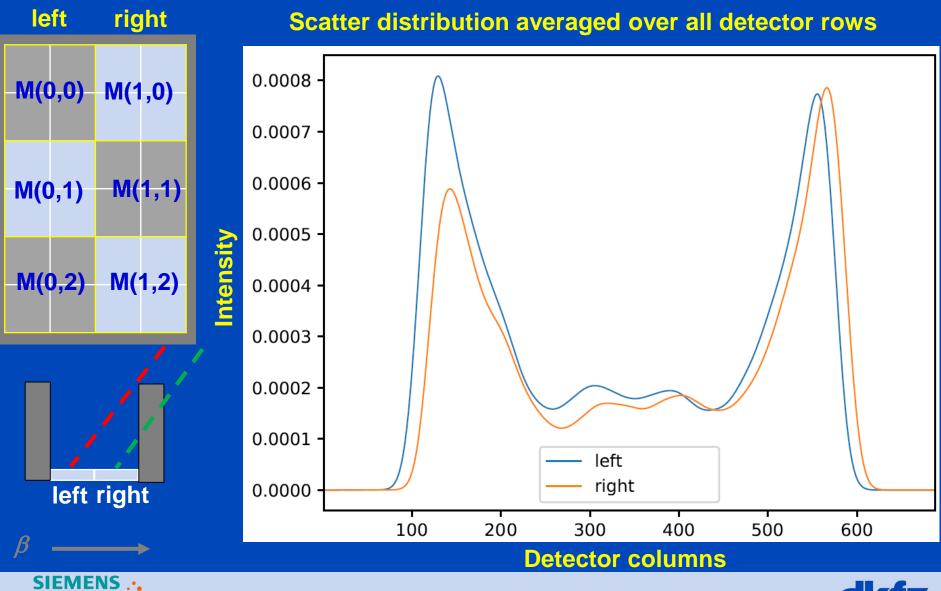




Scatter distribution averaged over all detector rows

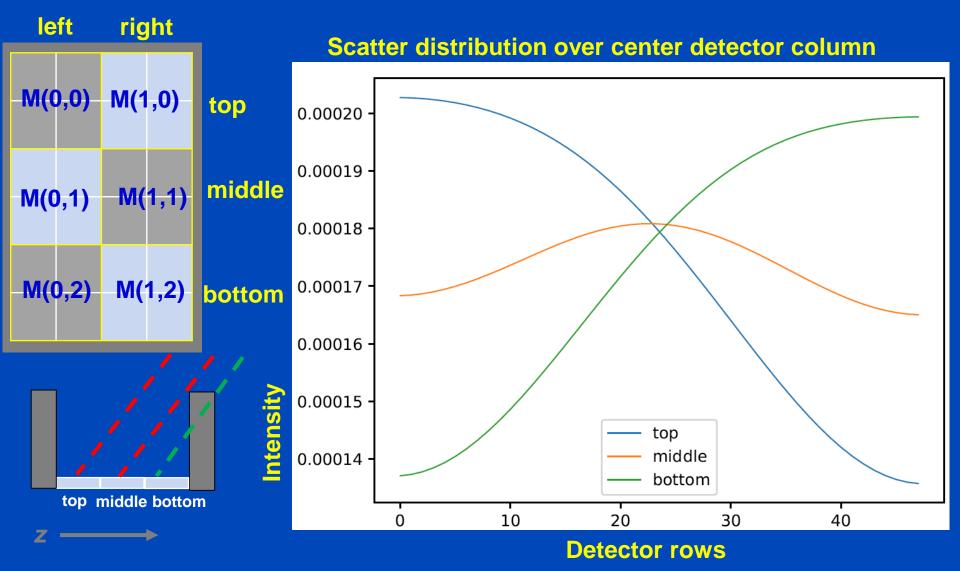


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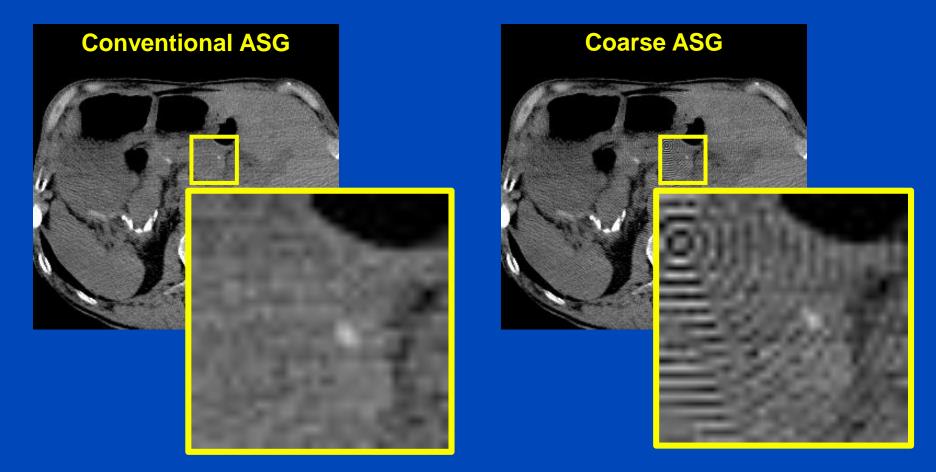








Scatter Artifacts of Coarse ASG



Coarse ASG can lead to scatter-induced moiré artifacts.

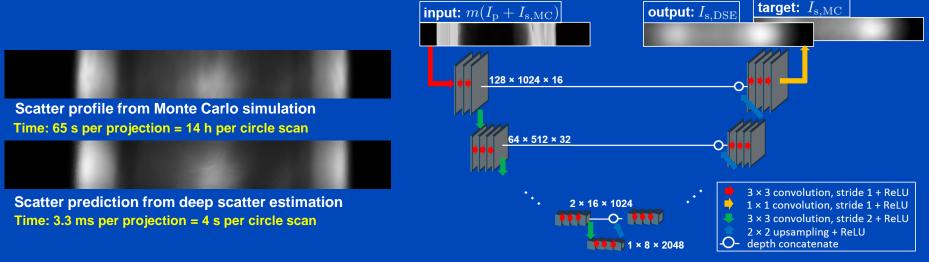


Reconstruction: C = 40 HU, W = 300 HU



Deep Scatter Estimation (DSE)

- Use a deep convolutional neural network to estimate scatter using the • acquired projection data as input.^{1,2}
- Train the network to predict Monte Carlo scatter estimates based on the ٠ acquired projection data.^{1,2}
- DSE outperforms other scatter estimation techniques.^{1,2,4,5} •
- DSE is much faster than the Monte Carlo simulation.^{1,2,5} •
- DSE can also be trained with measured scatter data.³ •
- DSE shows great potential to correct for cross-scatter in dual source CT.4,5 •

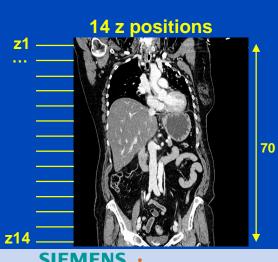


- [1] J. Maier, M. Kachelrieß et al. "Deep Scatter Estimation (DSE)", SPIE 2017 and J. of Nondest. Eval. 37:57, July 2018.
- [2] J. Maier, M. Kachelrieß et al. "Robustness of DSE", Med. Phys. 46(1):238-249, January 2019.
- [3] J. Erath, M. Kachelrieß et al "Monte-Carlo-Free Deep Scatter Estimation (DSE) 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



Training and Validation Data

- Monte Carlo simulation with the geometry of the photon counting CT scanner NAEOTOM Alpha (Siemens Healthineers)
- 12 patients for training and 4 for validation
- 14 z-positions with 36 projections each simulated for each patient
- 8064 paired scatter and primary data pairs
- Simulation of coarse ASG with macro pixel with detector dimension of 1376 × 144 pixels
- 6 different macro pixels locations
- Smooth only across same macro-pixel locations



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Training and validation patients with high variety and different clinical situations, important to consider scatter-to-primary ratio

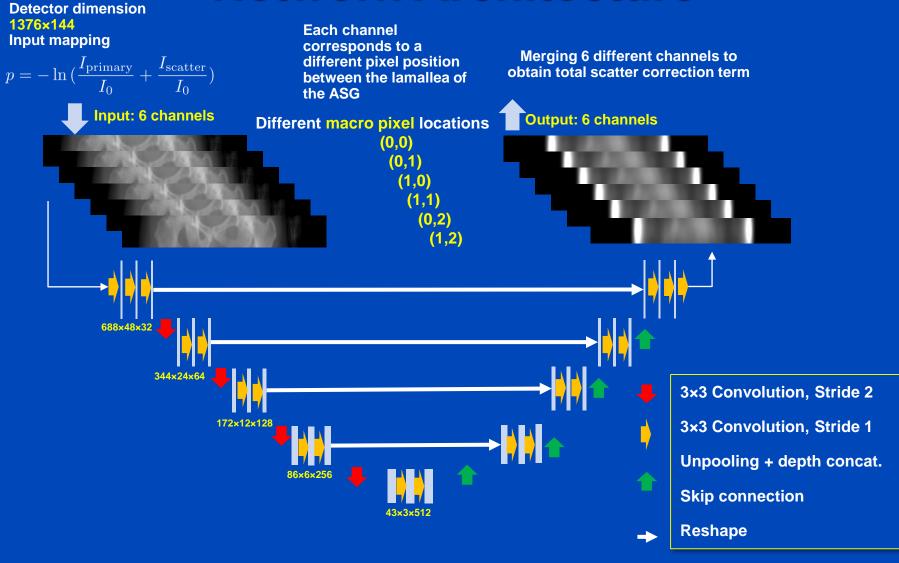
Example of validation data set:



M(0,0)M(1,0)M(0,1)M(1,1)M(0,2)M(1,2)



Network Architecture

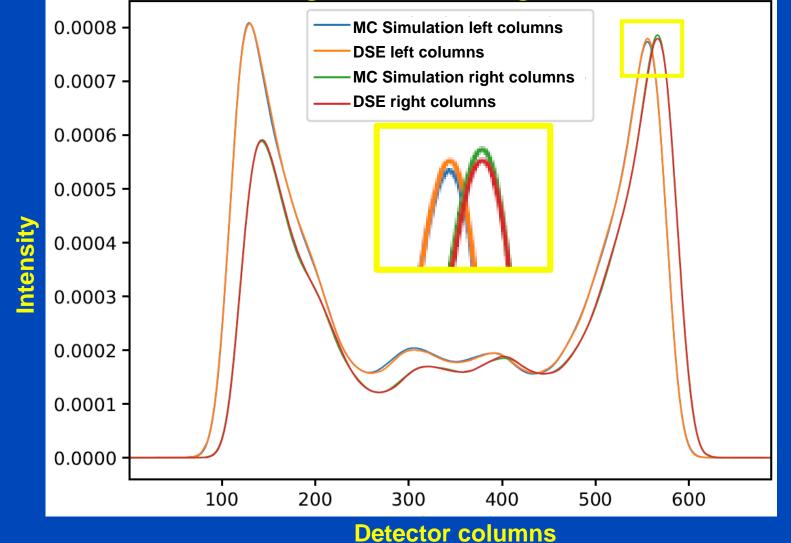






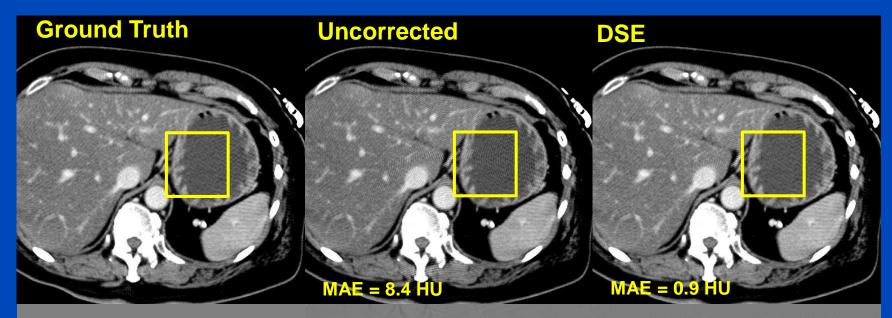
Predicton of Scatter Intensity

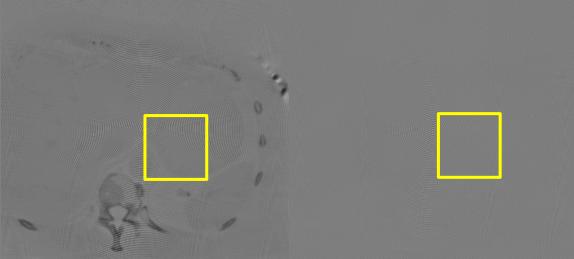
Scatter distribution of left and right columns averaged over all detector rows





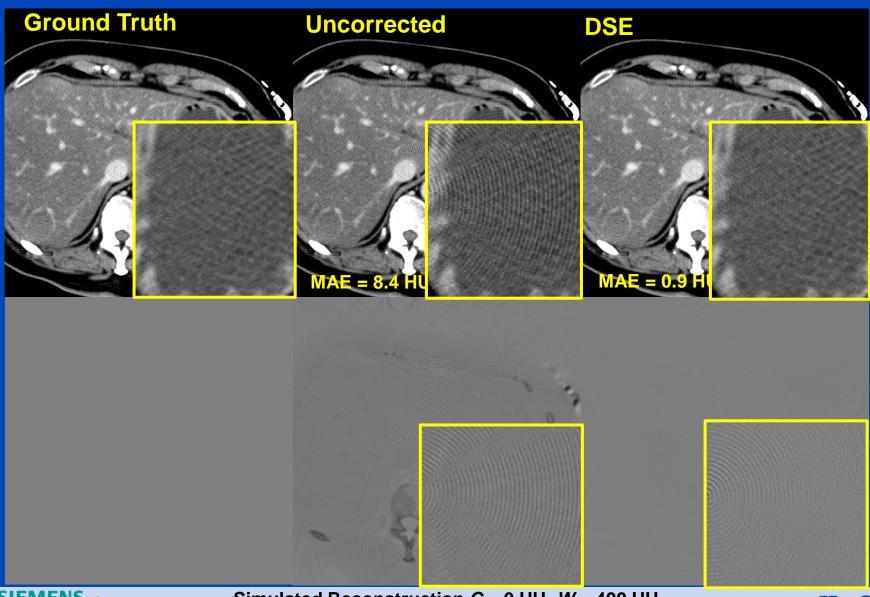






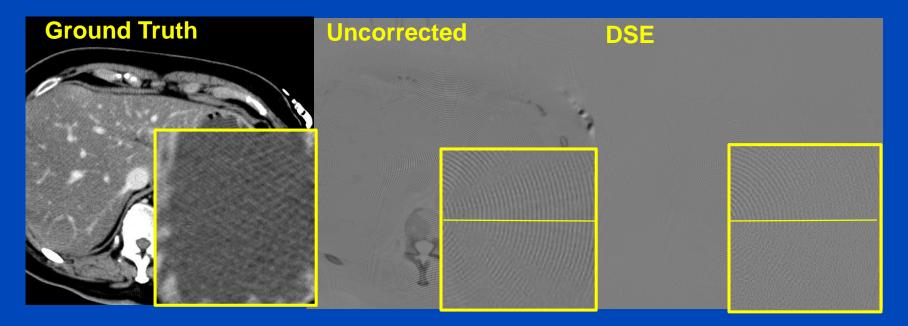


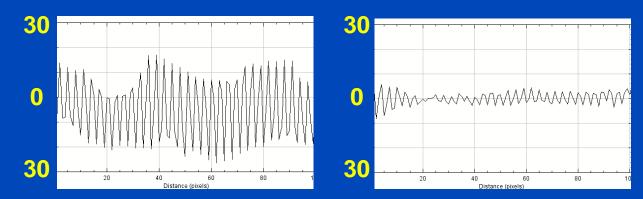








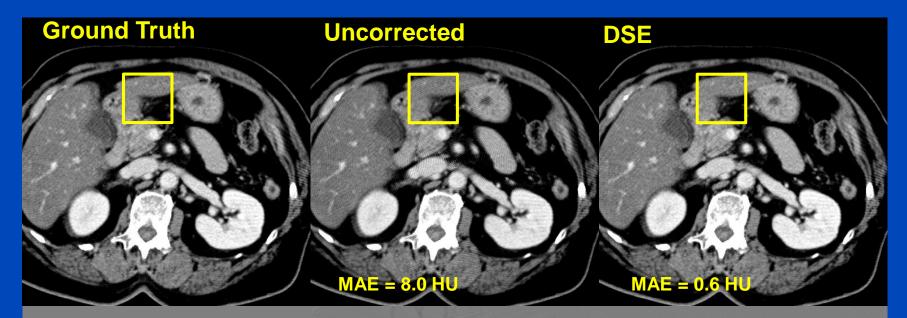


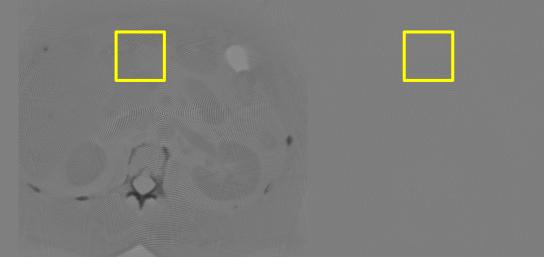


Amplitude of the moiré effect gets corrected from 30 HU to less than 5 HU.



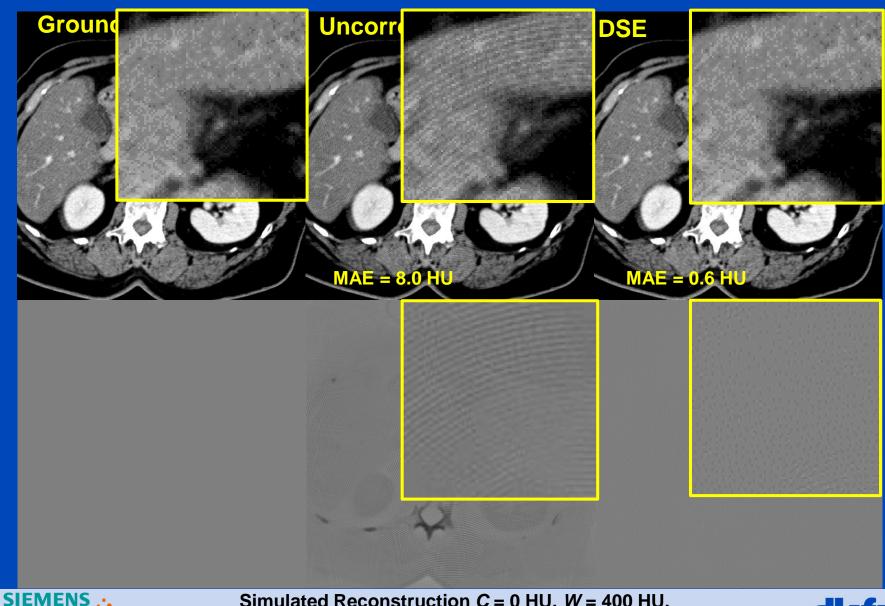












Simulated Reconstruction C = 0 HU, W = 400 HU, Difference to GT C = 0 HU, W = 50 HU

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Conclusions

- Smaller detector pixel and coarse anti-scatter grid can lead to moiré artifacts.
- Scatter induced moiré effect can be clearly corrected with deep learning-based scatter correction.
- With the proposed algorithm the mean absolute error (MAE) could be reduced from uncorrected about 9 HU to under 1 HU.
- The amplitude of the scatter induced moiré effect can be corrected from 30 HU to less than 5 HU.
- Next step: apply deep-learning based correction for measurements.







This presentation will soon be available at www.dkfz.de/ct. Job opportunities through DKFZ's international PhD or Postdoctoral Fellowship programs (marc.kachelriess@dkfz.de).



