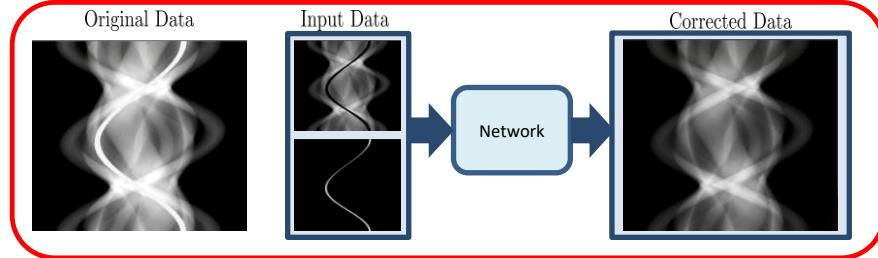


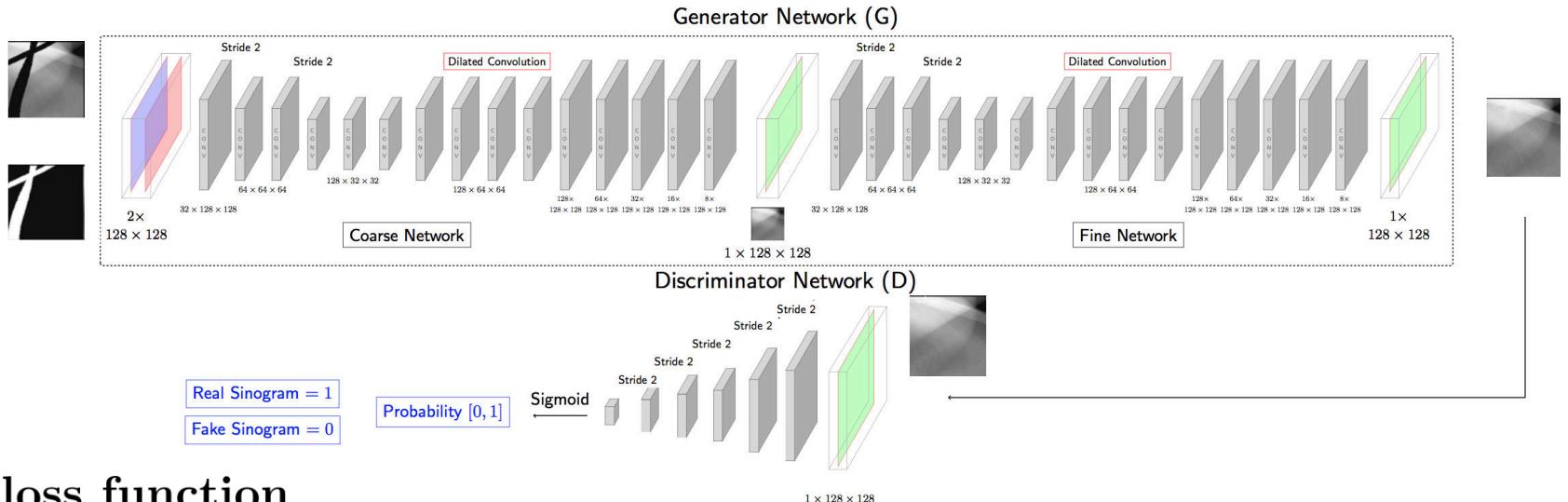
Motivation

- Design a network, which can be applied directly to the measured projection data
- Integration in modern CT image reconstruction to get the best possible result

Trained network



cGAN



loss function

Adversarial loss + l1 loss (local and global)

$$G^* = \arg \min_G \max_D L_{GAN}(G, D) + \lambda \mathbb{E}_{x, x_{ref}} [\|(x_{ref} - G(x))_{\text{local}}\|_1 + \|(x_{ref} - G(x))_{\text{global}}\|_1]$$

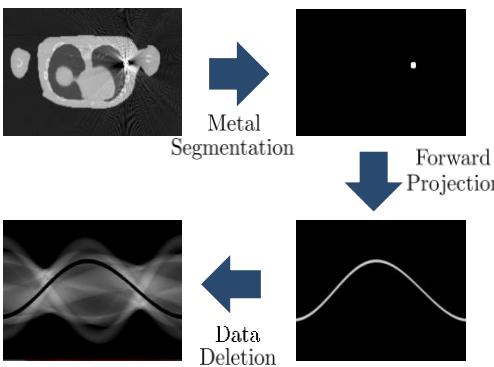
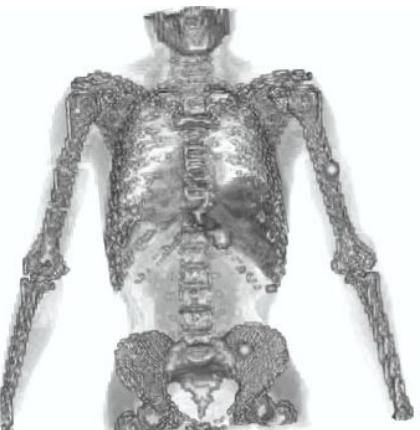
with

$$L_{GAN}(G, D) = \mathbb{E}_{x, x_{ref}} [\log(D(x, x_{ref}))] + \mathbb{E}_x [1 - D(x_{ref}, G(x))]$$

Data

120 simulated sinograms of the XCAT Phantom

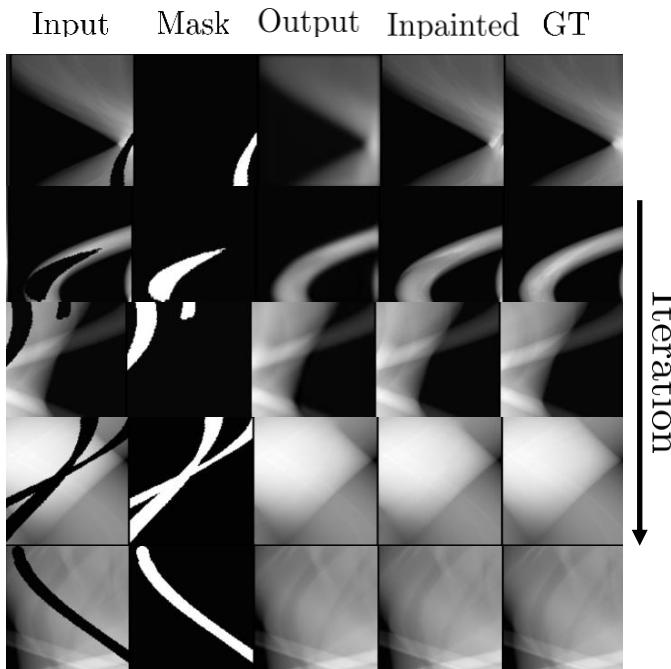
Divided in training/validation/test data (90/15/15)



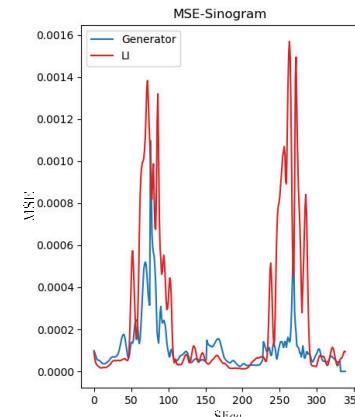
- G Generator network
- D Discriminator network
- x Training data: missing metal trace
- x_{ref} Training data: ground truth
- \mathbb{E} Expectation value over the training data
- λ Weighting parameter

Training

Parameter	Value
number training data	183 000
number validation data	31 000
batch size	32
epochs	100
learning rate	$1 \cdot 10^{-5}$

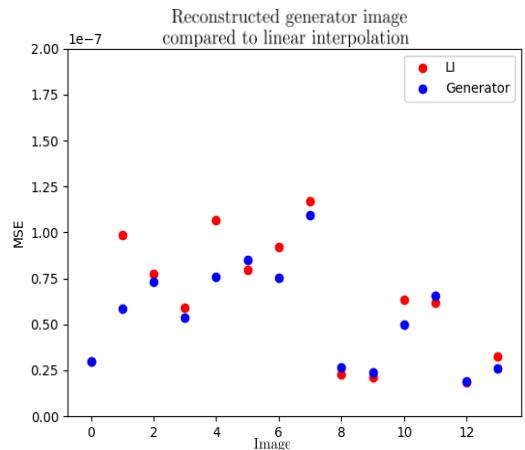


Projection domain



Results

Image domain



Input

Linear
InterpolationProposed
Method