

Oct, 2017

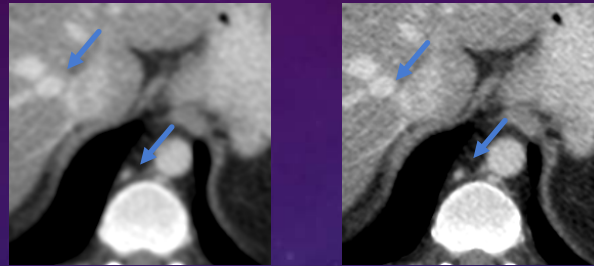
# Deep reconstruction of security data & AAPM Grand Challenge

Quanzheng Li

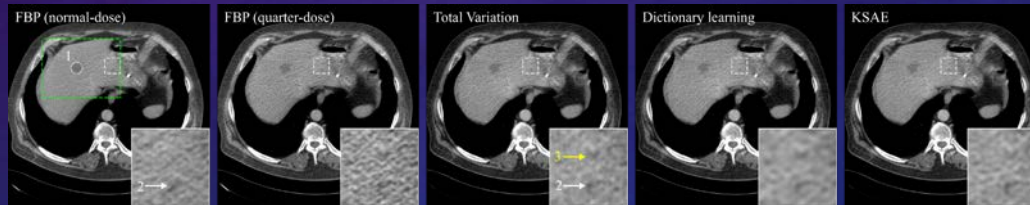
Core Faculty, Center for Clinical Data Science, Harvard Medical School  
Director, Computational Imaging and Artificial Intelligence lab, Gordon Center,  
Mass General Hospital

# Summary

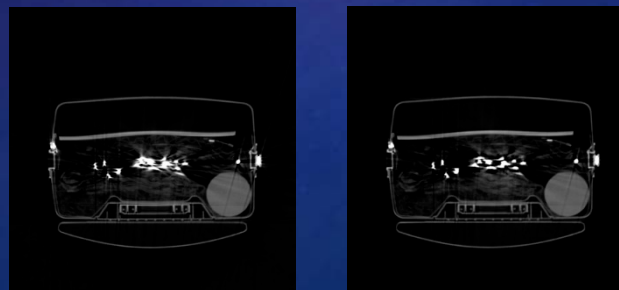
- An iterative algorithm with Spatially Encoded Non-Local Penalty is developed, and won **AAPM Low Dose CT Grant Challenge**



- An iterative recon **with deep learning based** prior is developed and published on TMI special issue on low dose CT



- Similar iterative recon with **deep learning based** prior is applied to security CT





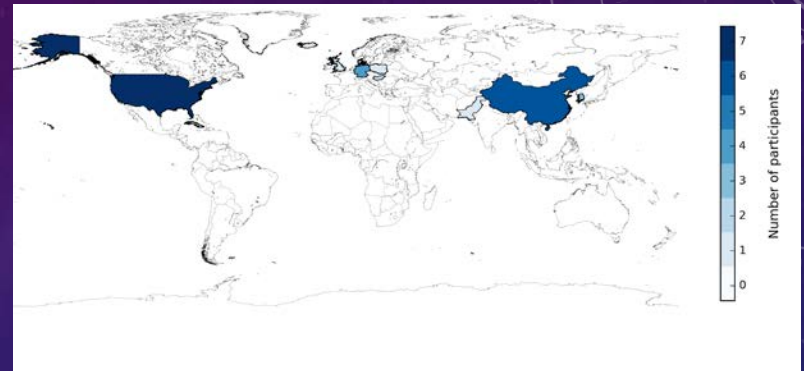
# Low Dose CT Grand Challenge

- First CT Grand Challenge
- Public Available Data and Parameters
- An Open Test Bed for CT Algorithms

- World Wide Participants

**Low Dose CT Grand Challenge**

NIH National Institute of Biomedical Imaging and Bioengineering  
AMERICAN ASSOCIATION OF PHYSICISTS IN MEDICINE  
MAYO CLINIC  
CT Clinical Innovation Center



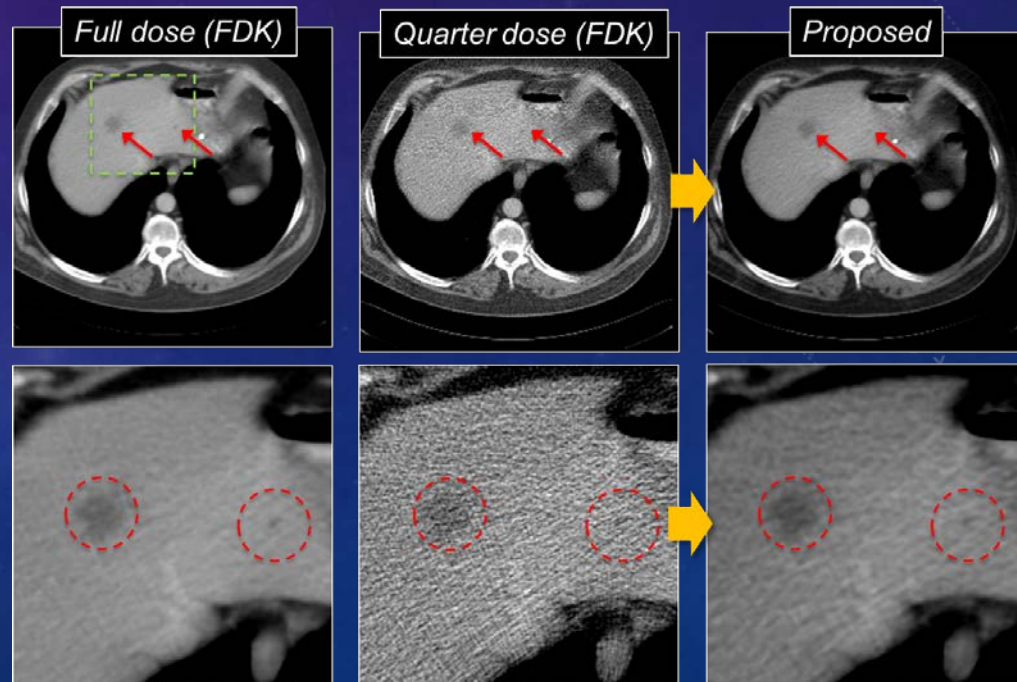
First Place!

## Spatially Encoded Non-Local Penalty



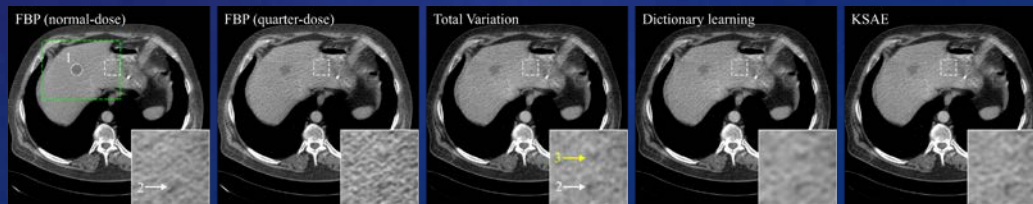
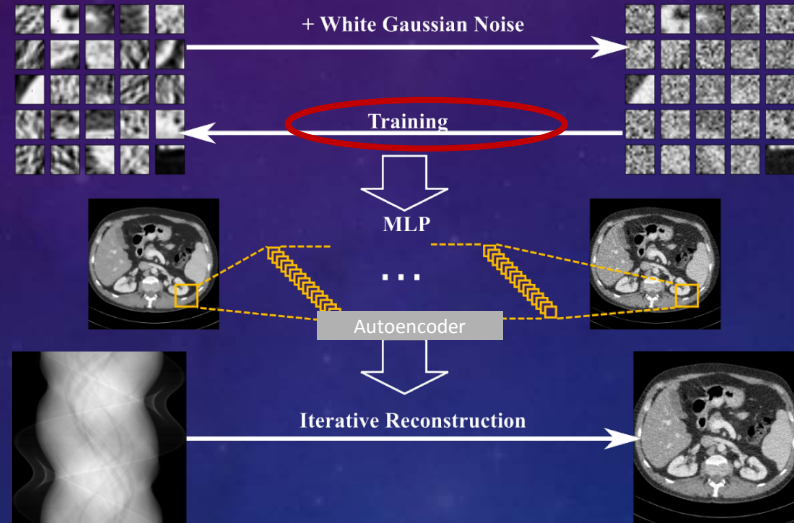
Traditional non-local mean

New non-local mean

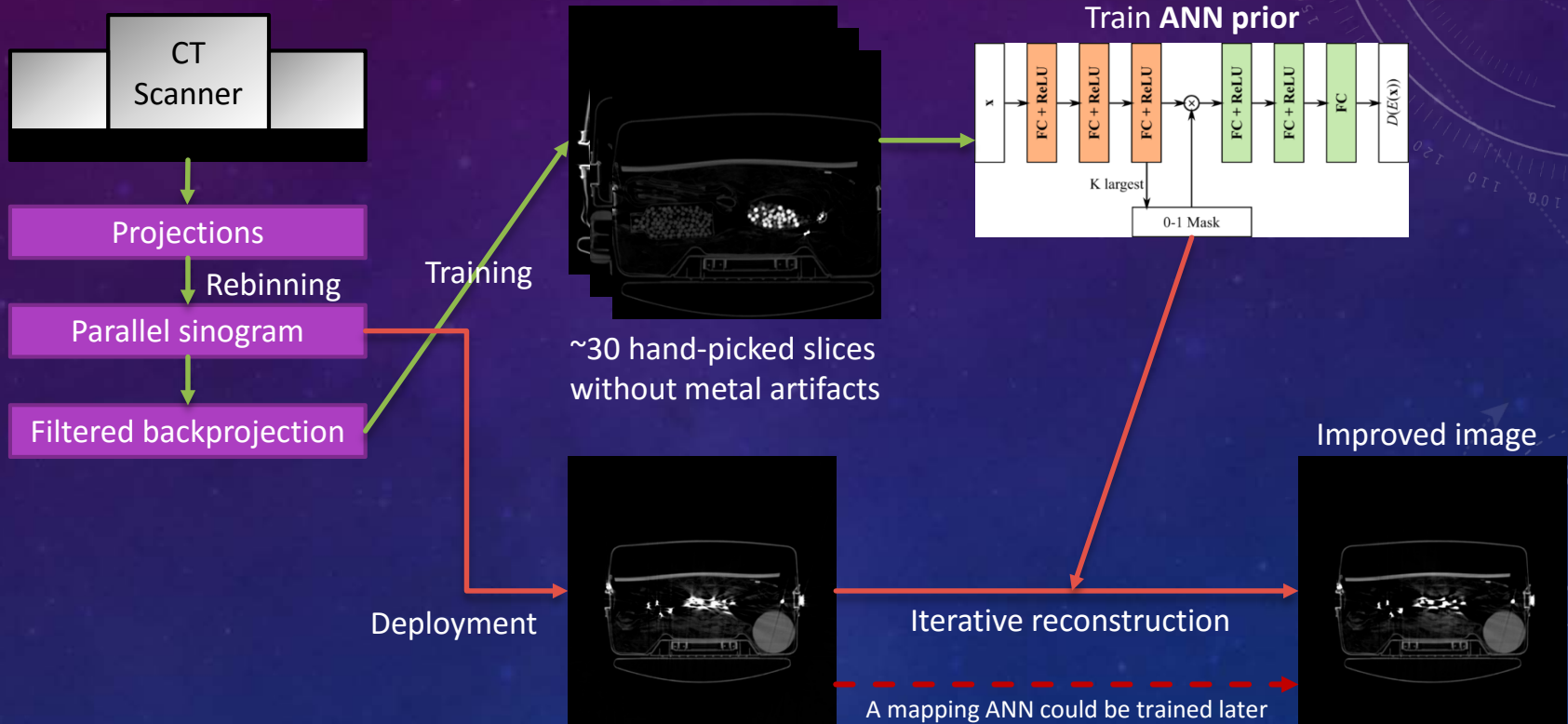


# DL based Recon: Clinical Data

- Because noises in  $x$  changes during the iterations, it has to be learned in an **unsupervised** way;
- A solution with denoising autoencoders:



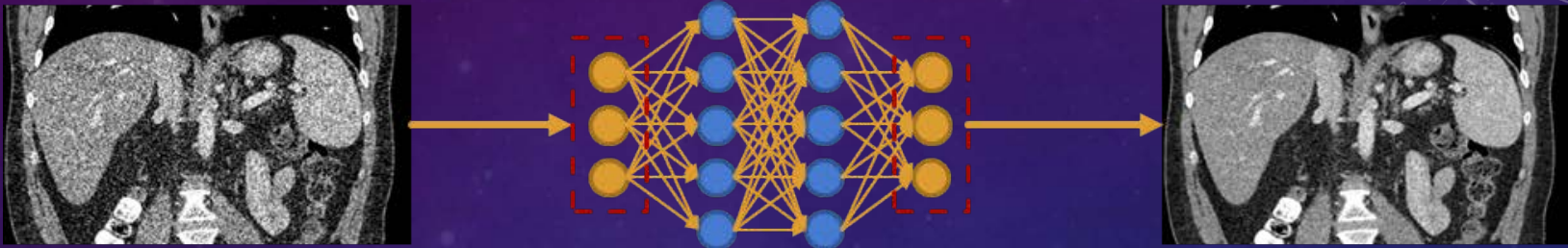
# DL BASE RECON: TSA DATA



ML Knowledge can be transferred to security, and significantly improve image quality.

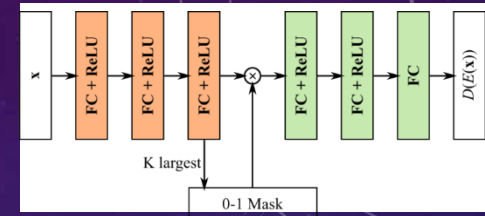
# MACHINE LEARNING BASED RECONSTRUCTION

- Artificial neural networks (ANN) have great nonlinear mapping ability

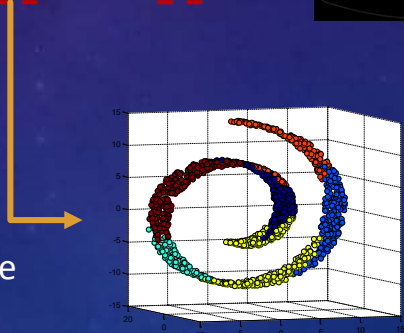
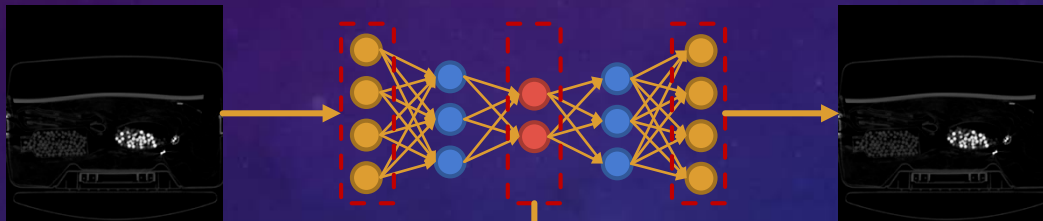


- Problem
  - Require “ground truth” data: unavailable for security CT
- Solution: unsupervised learning
  - Train ANN on clean data
  - Apply ANN to “dirty” data during reconstruction
  - Train another “mapping” ANN for real-time application

# ANN PRIOR: AUTOENCODER



- An ANN that maps clean images back to themselves



\*The ANN was trained on patches in practice

The features mapped to images to some “computer understandable” manifold:  
It told the computer where clean data was.

# ITERATIVE RECONSTRUCTION

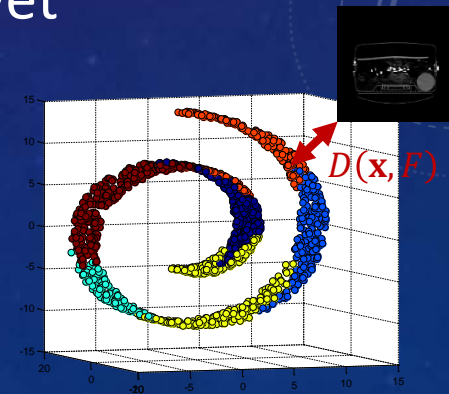
- Penalized reconstruction

$$\mathbf{x} = \arg \min \left[ \|\mathbf{Ax} - \mathbf{b}\|_{\mathbf{w}} + \beta D(\mathbf{x}, F) \right]$$

Noise weighted data loss

Distance between  $\mathbf{x}$  and trained space

- No explicit metal artifacts reduction applied yet





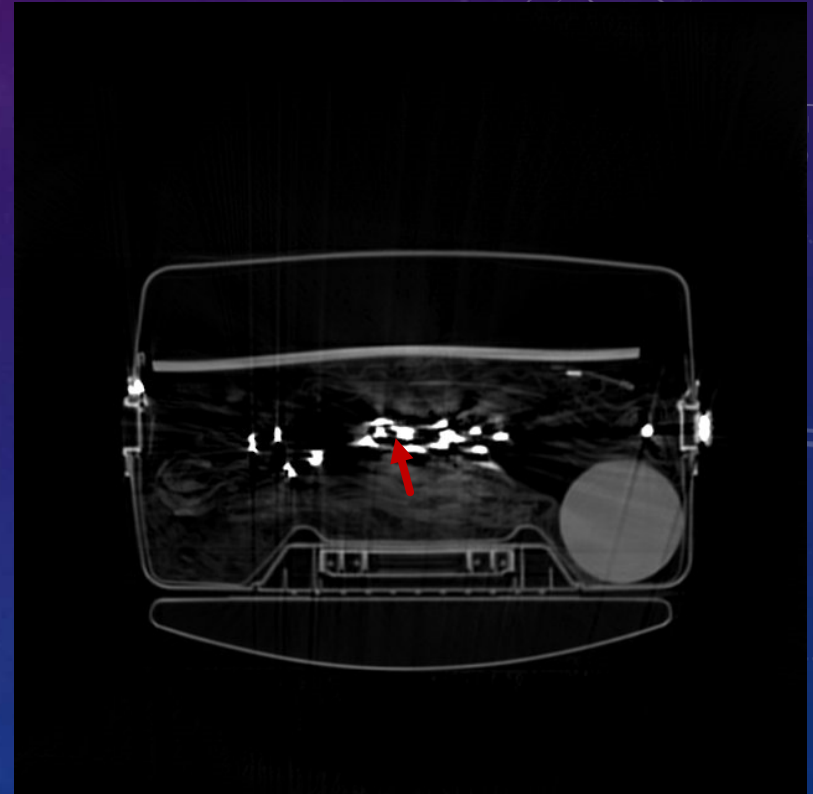
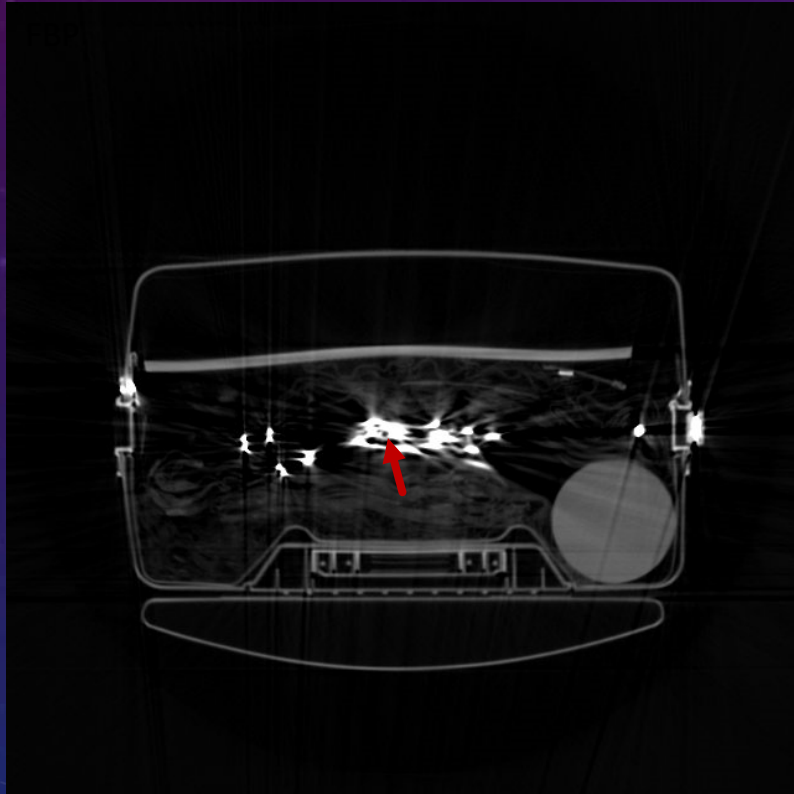


# PRELIMINARY RESULTS

- Data
  - Provided by Boston University
  - Imatron C300 electronic beam tomography (EBT) medical CT scanner
  - Reconstructed at MGH



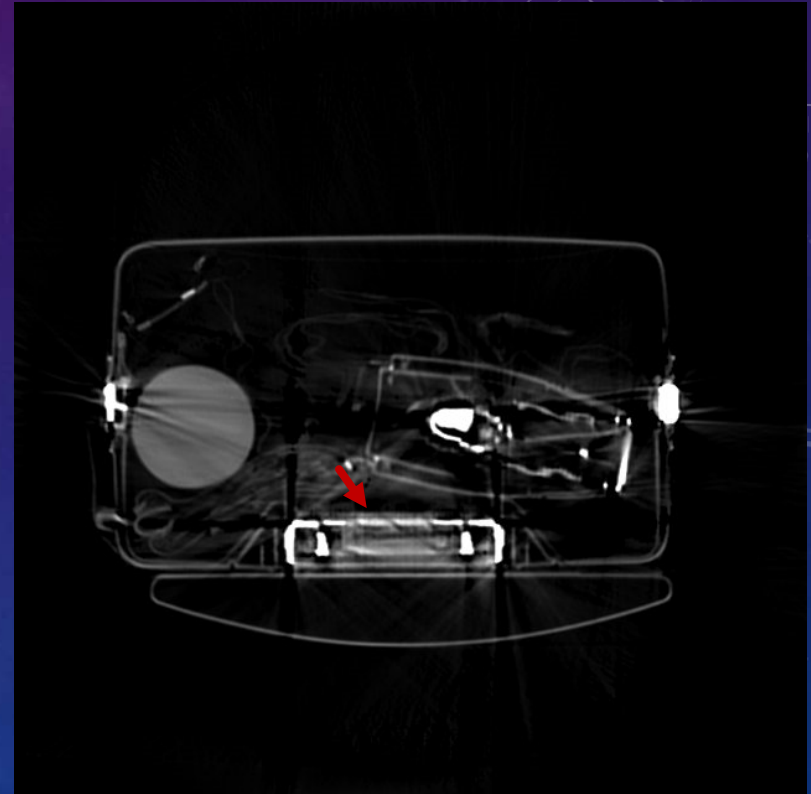
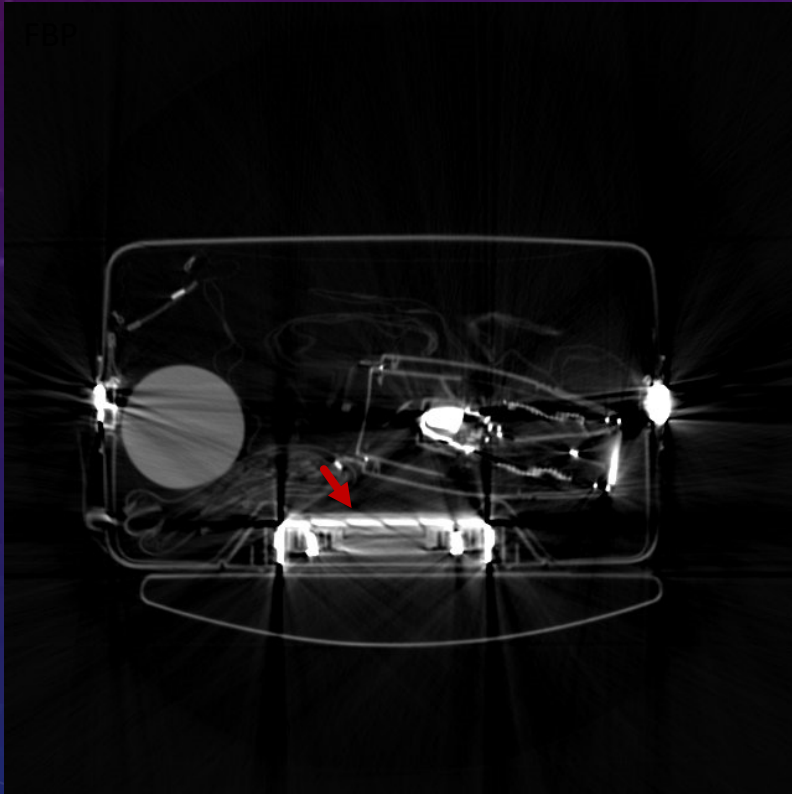
# PRELIMINARY RESULTS



# PRELIMINARY RESULTS



# PRELIMINARY RESULTS





# CONCLUSION AND FUTURE WORKS

- With unsupervised ANN approach
  - We are able to commence training without ground truth
  - Promising artifacts reduction potential was shown
- Future improvements
  - Incorporate explicit metal artifacts reduction technique
  - Train the final “mapping” ANN for real-time application



# CIAI Lab

## Image Recon and Analysis

### Image Recon:

- PET
- CT
  - Low Dose CT
  - Spectrum CT/Material Decomposition
  - Phase Contrast CT
  - Static CT / Nano CT
- MRI/Optical
- Microscope – EM
- Hybrid: PET/CT, PET/MRI

### Image Analysis:

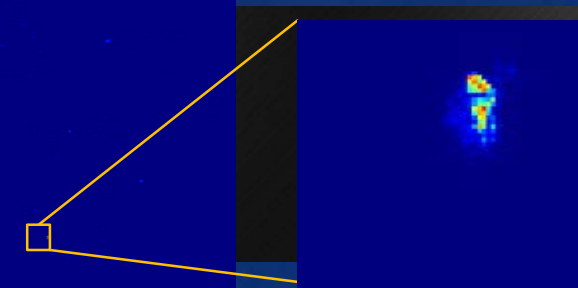
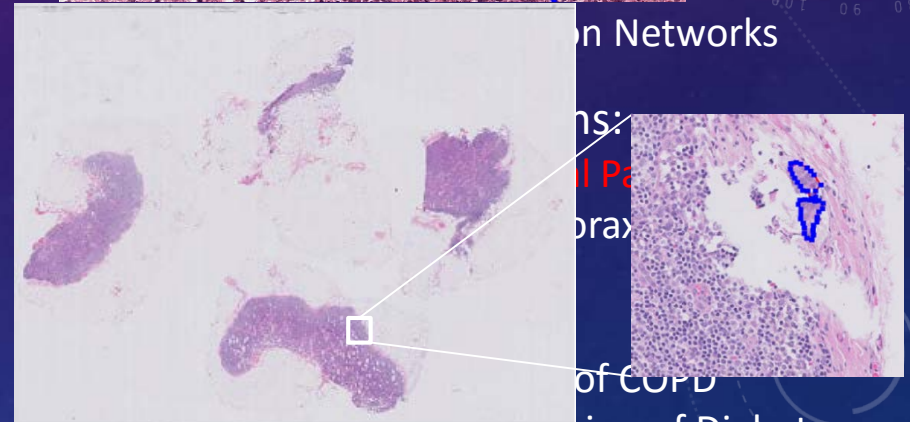
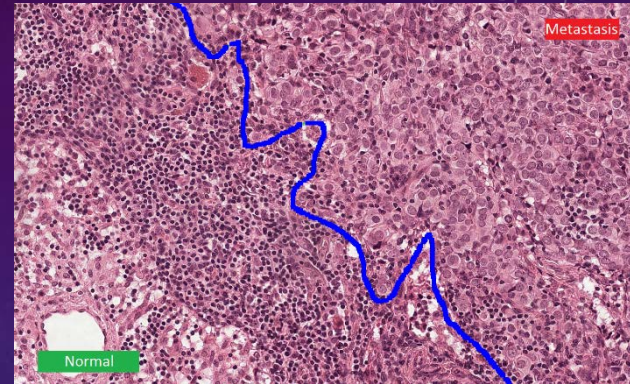
- Image Denoising and Restoration
- Segmentation and Registration
- Novel Image Biomarkers
- Radiomics/Radiogenomics
- Diagnosis/Prognosis

**Low Dose CT  
Grand Challenge**



**First Place!**

## Medicine





Thanks for your attention !

[Li.Quanzheng@mgh.harvard.edu](mailto:Li.Quanzheng@mgh.harvard.edu)