### Speed Talk: Learned Reconstruction in Medical Imaging



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Most obvious place for machine learning is in post-processing













(initially) less obvious place: image reconstruction



### Why bother with AI/ML in recon?

#### In Magnetic Resonance Imaging (MRI)...

- Physical limits to how fast data can be acquired
- Acquiring "fully sampled" data time-consuming
- Taking fewer measurements gives noise/artifacts

#### In X-ray Computed Tomography (CT)...

- Uses ionizing radiation potentially harmful to patient
- Can reduce dose but at the expense of noise/artifacts

#### accelerated knee MRI

#### "fully-sampled" knee MRI



standard dose breast CT scan

reduced dose breast CT scan











### Example: Learned recon of accelerated MRI acquisitions

• Ex: Train neural network to de-artifact accelerated MRI acquisitions





## Issue: Hallucination of clinically relevant details

ground truth



4-fold undersampled MRI

2306

figure from:

#### neural network recon

![](_page_7_Picture_9.jpeg)

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![](_page_7_Picture_11.jpeg)

#### Results of the 2020 fastMRI Challenge for Machine Learning MR Image Reconstruction

## Issue: Erasure of clinically relevant details

#### breast CT phantom

![](_page_8_Picture_2.jpeg)

# breast CT simulation study

#### low-dose recon

#### neural network recon

#### O., Sidky, Resier, & Pan, SPIE Med Img 2021

![](_page_9_Picture_0.jpeg)

# My focus:

- 1. NN architecture design
- 2. Loss function design
- 3. Deep learning theory

![](_page_9_Picture_6.jpeg)

Key Problem: How do we ensure learning approaches to image reconstruction yield faithful, reliable results, with task-specific improvements?

# Thanks! Questions?

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