

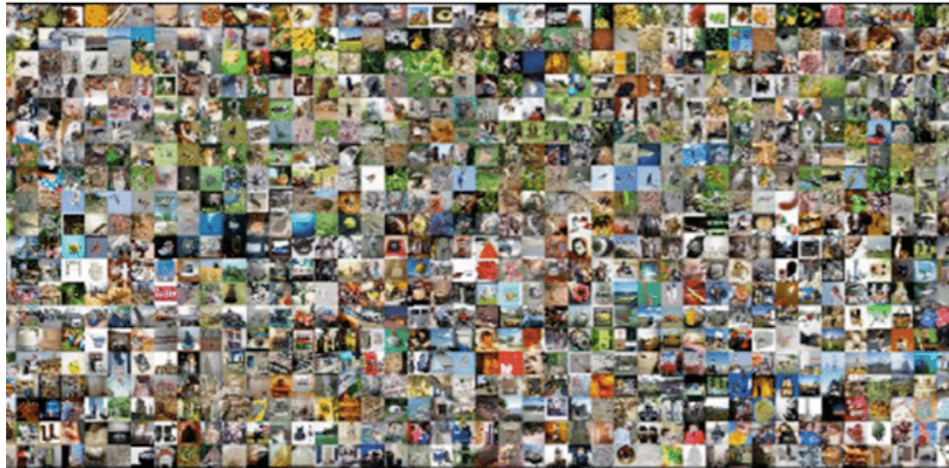


# Overcoming the data bottleneck in AI for the sciences

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MHI Finalist Talk 2021

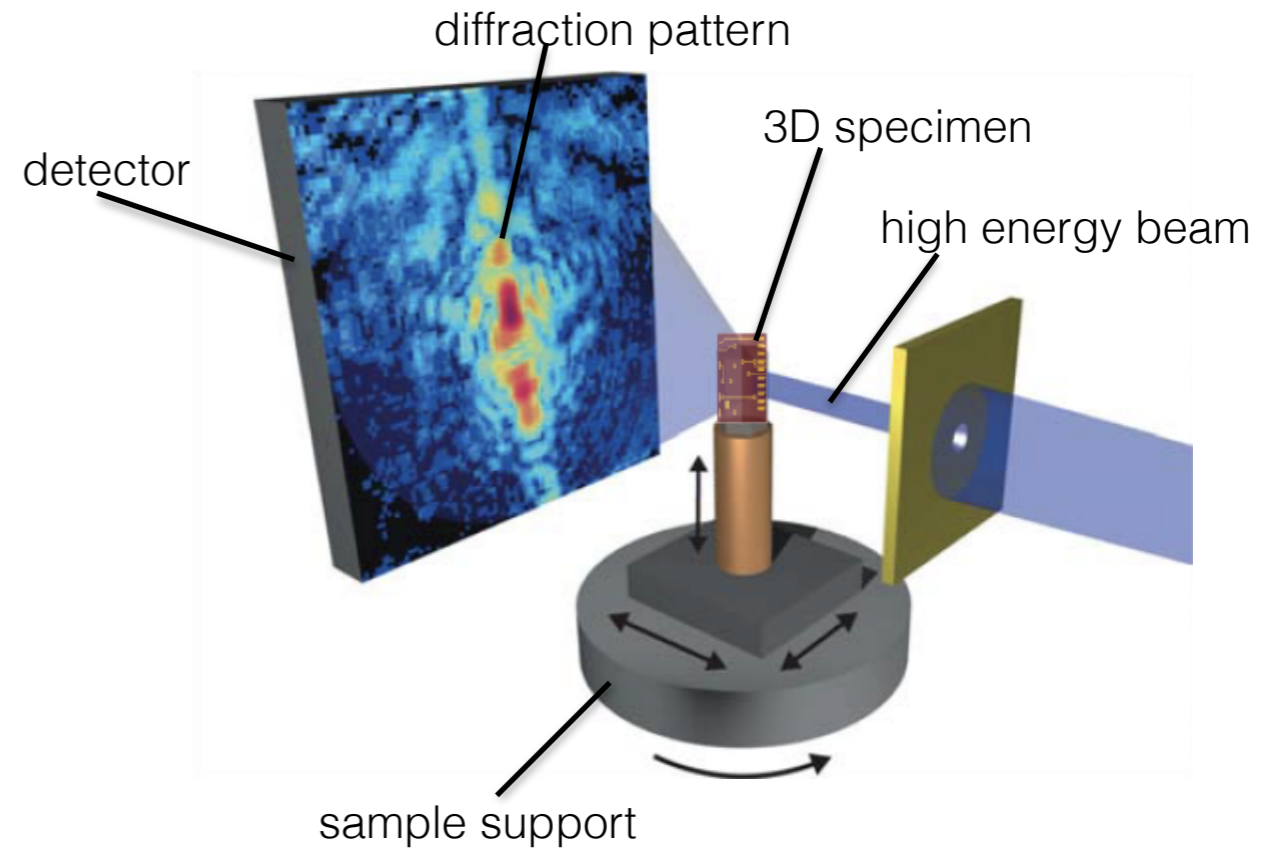
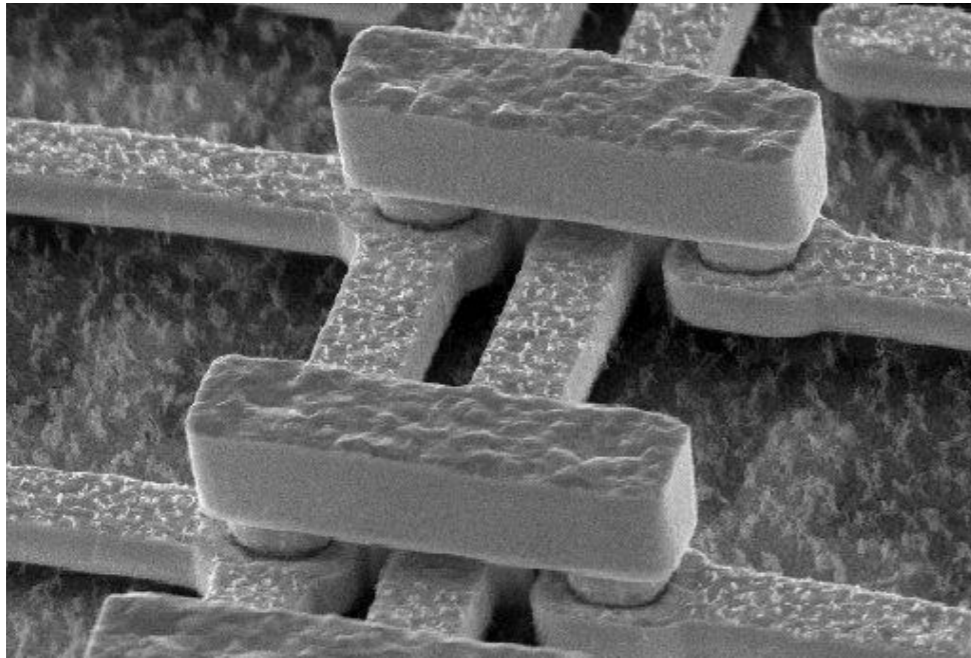
# Deep learning is data-hungry



		'Internet data'
Data acquisition		readily available
Cost		very low
Amount of data		huge

**How about scientific data?**

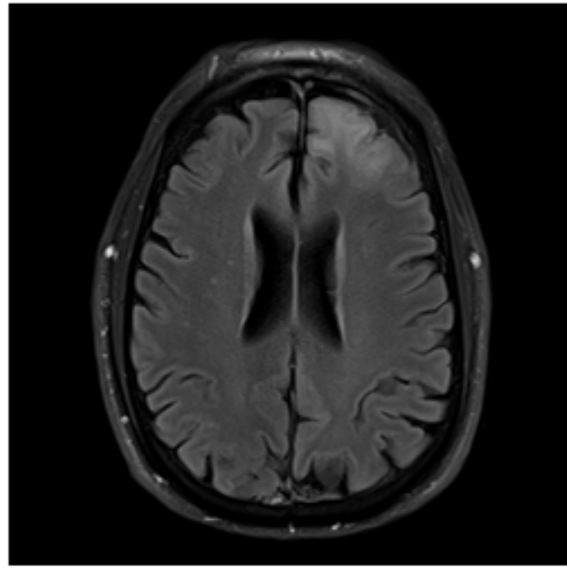
# Case study: nano-scale imaging



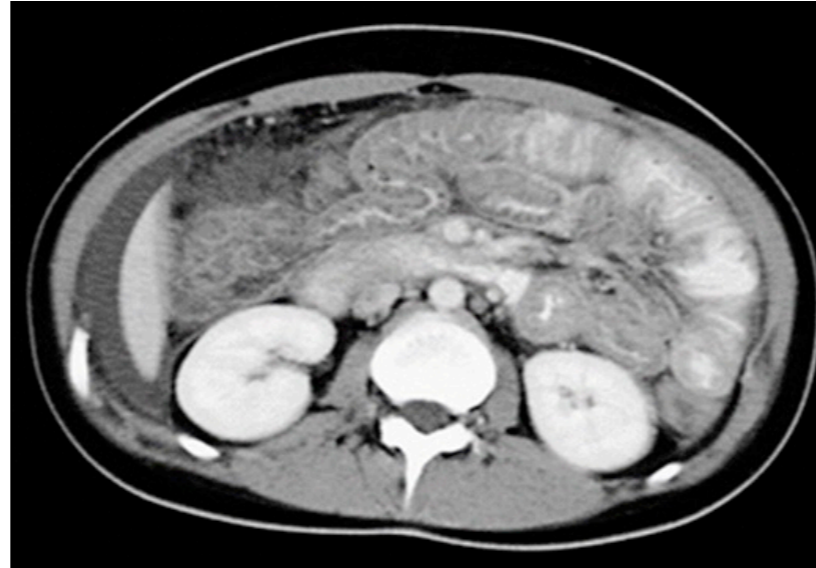
## 3D CDI data

<b>Data acquisition</b>	A circular icon containing symbols for a lightning bolt, a thermometer, a computer monitor, and a power button.	1cm <sup>3</sup> volume $\approx$ 2500 days
<b>Cost</b>	A green circular icon containing a white dollar sign.	very high
<b>Amount of data</b>	A blue circular icon containing a white database symbol (cylinder with horizontal lines).	some simulated data

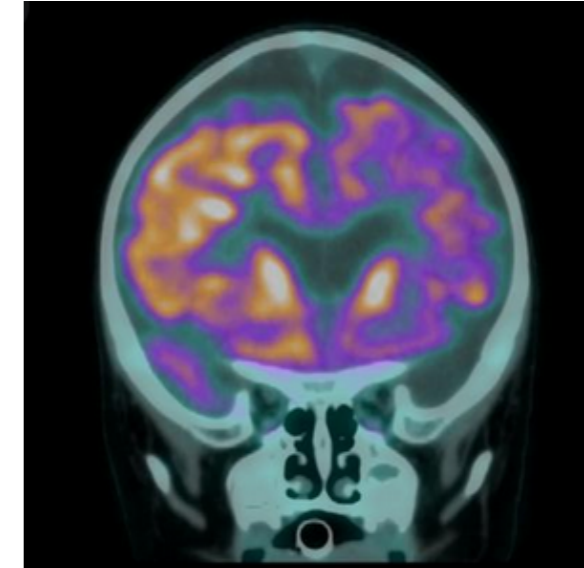
# Case study: medical imaging






MRI



CT



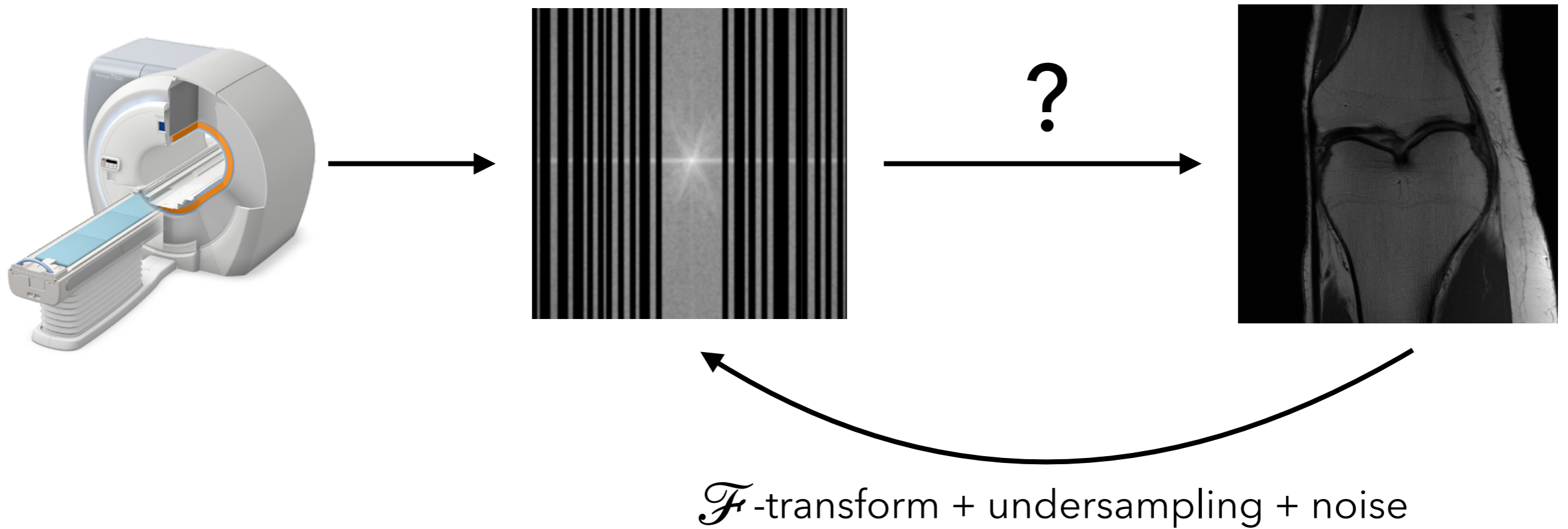
PET

		Medical imaging data
<b>Data acquisition</b>		slow, harmful radiation
<b>Cost</b>		high
<b>Amount of data</b>		small public datasets

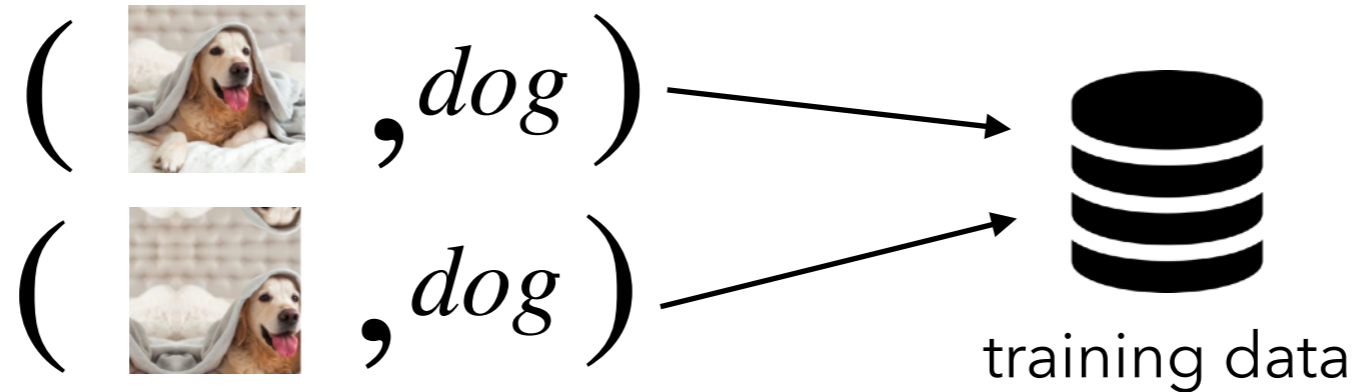
**How to address the data bottleneck?**

# Reducing training data for MRI reconstruction

# MRI reconstruction

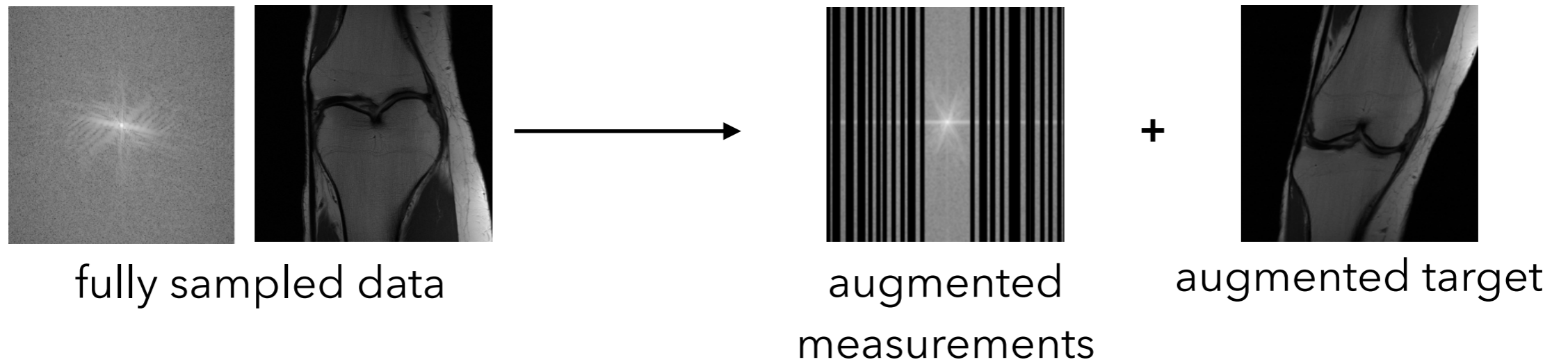


# Data augmentation in classification: straightforward



# Data augmentation in regression: non-trivial

1. Output is **not** invariant to transformations

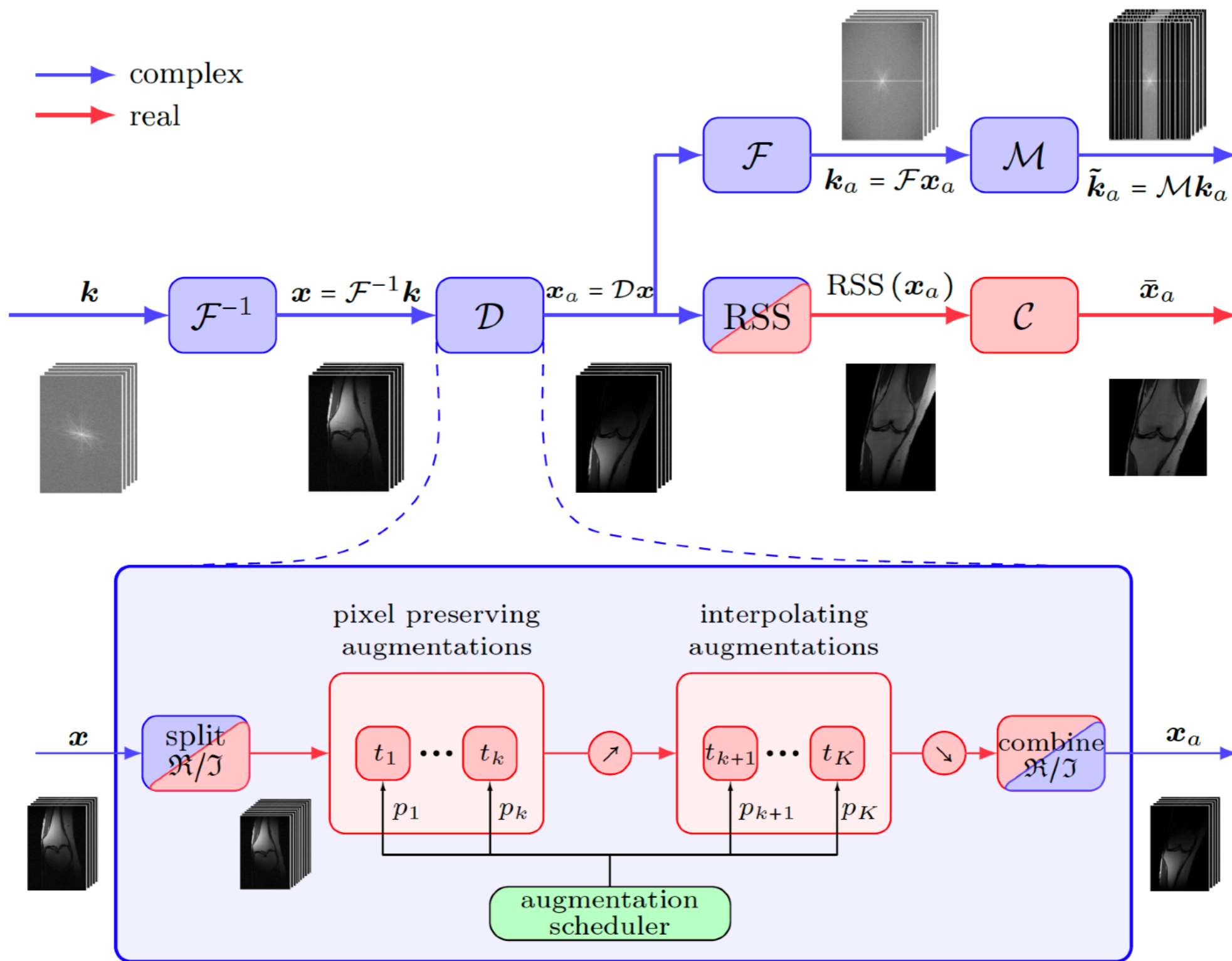


2. Distribution shift due to noise

$$x = x^* + n \longrightarrow x_{aug.} = \mathcal{D}x^* + \mathcal{D}n$$

augmented signal                      augmented noise!

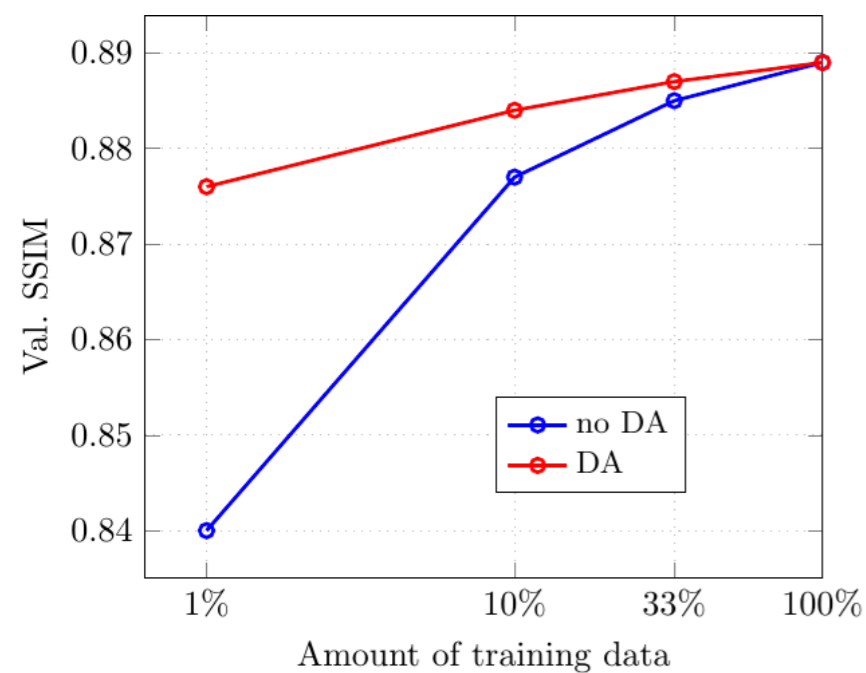
# MRAugment pipeline



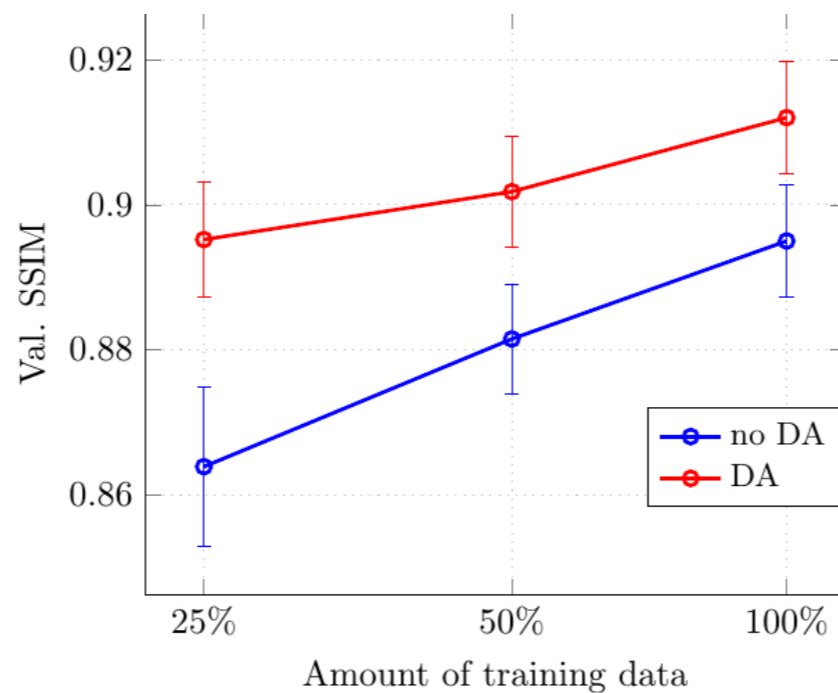


# Results on various datasets

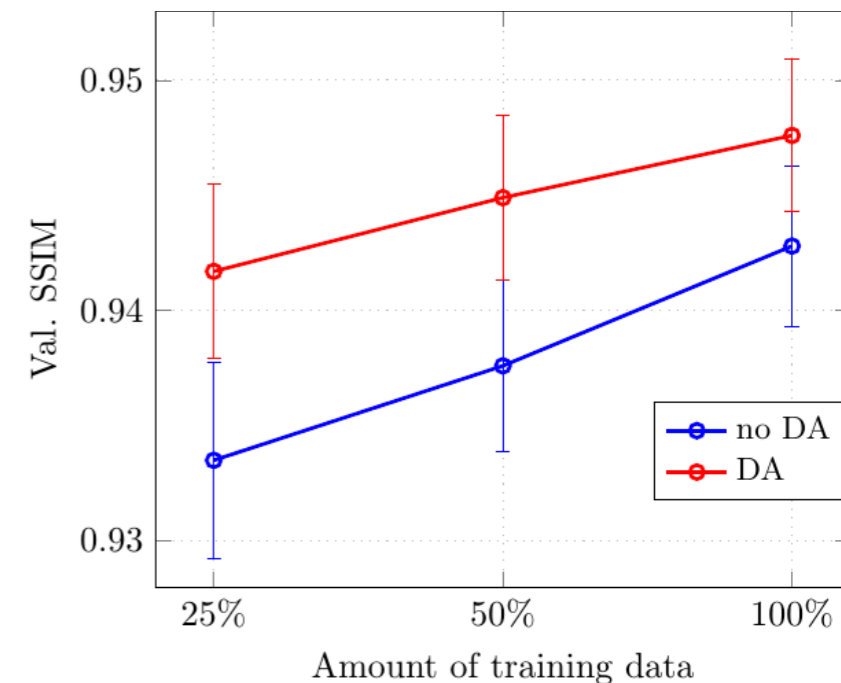
fastMRI knees



Stanford 2D FSE



Stanford 3D FSE knees



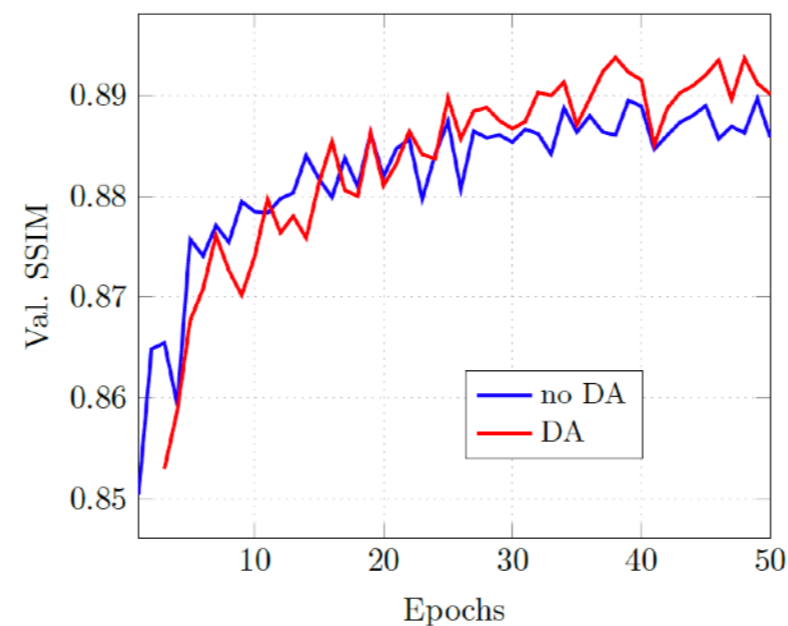
# Robustness experiments

## Unseen scanners

2% train	no DA	DA
$3T \rightarrow 3T$	0.8646	<b>0.9049</b>
$3T \rightarrow 1.5T$	0.8241	<b>0.8551</b>
$1.5T \rightarrow 3T$	0.8174	<b>0.8913</b>

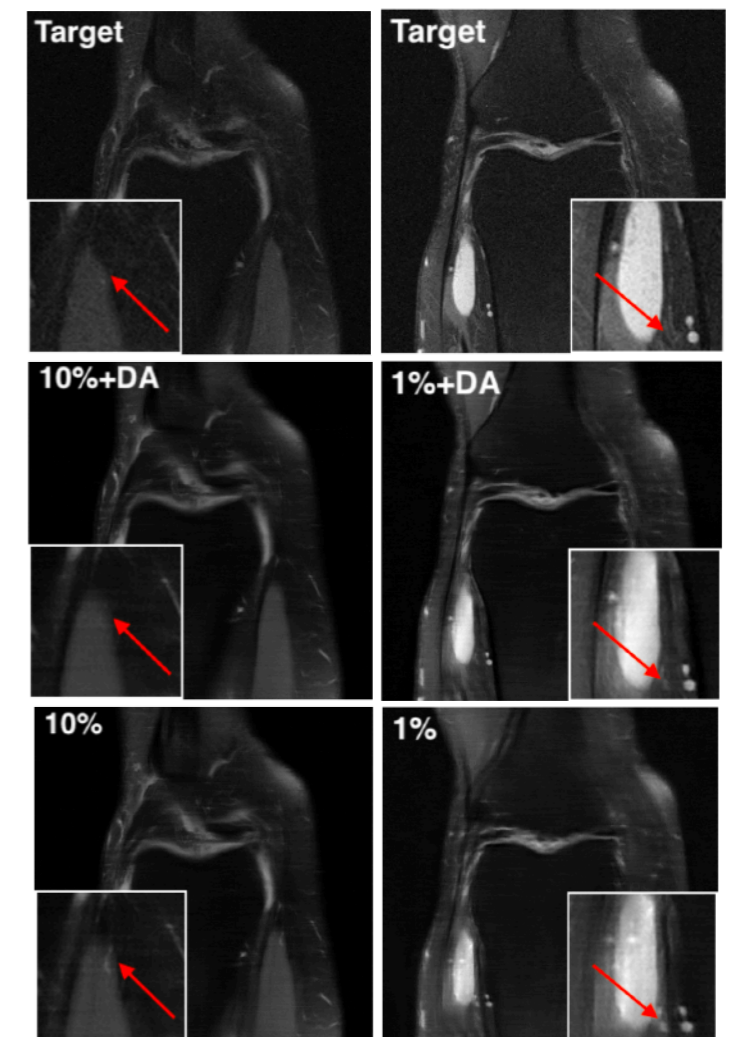
100% train	no DA	DA
$3T \rightarrow 3T$	0.9177	<b>0.9185</b>
$3T \rightarrow 1.5T$	0.8686	<b>0.8690</b>
$1.5T \rightarrow 3T$	0.9043	<b>0.9062</b>

## Unseen anatomies



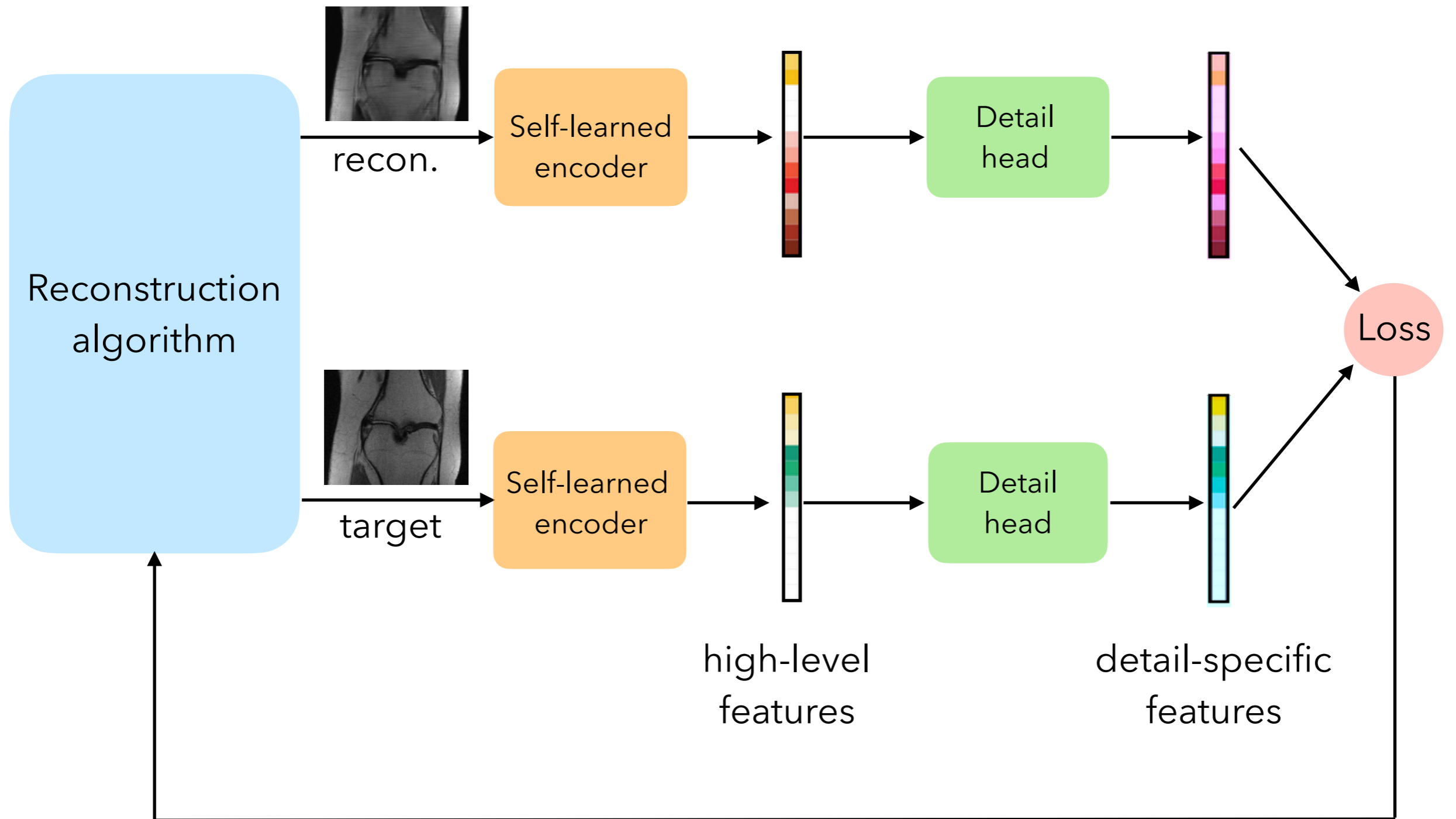
knee → brain

## Hallucinations



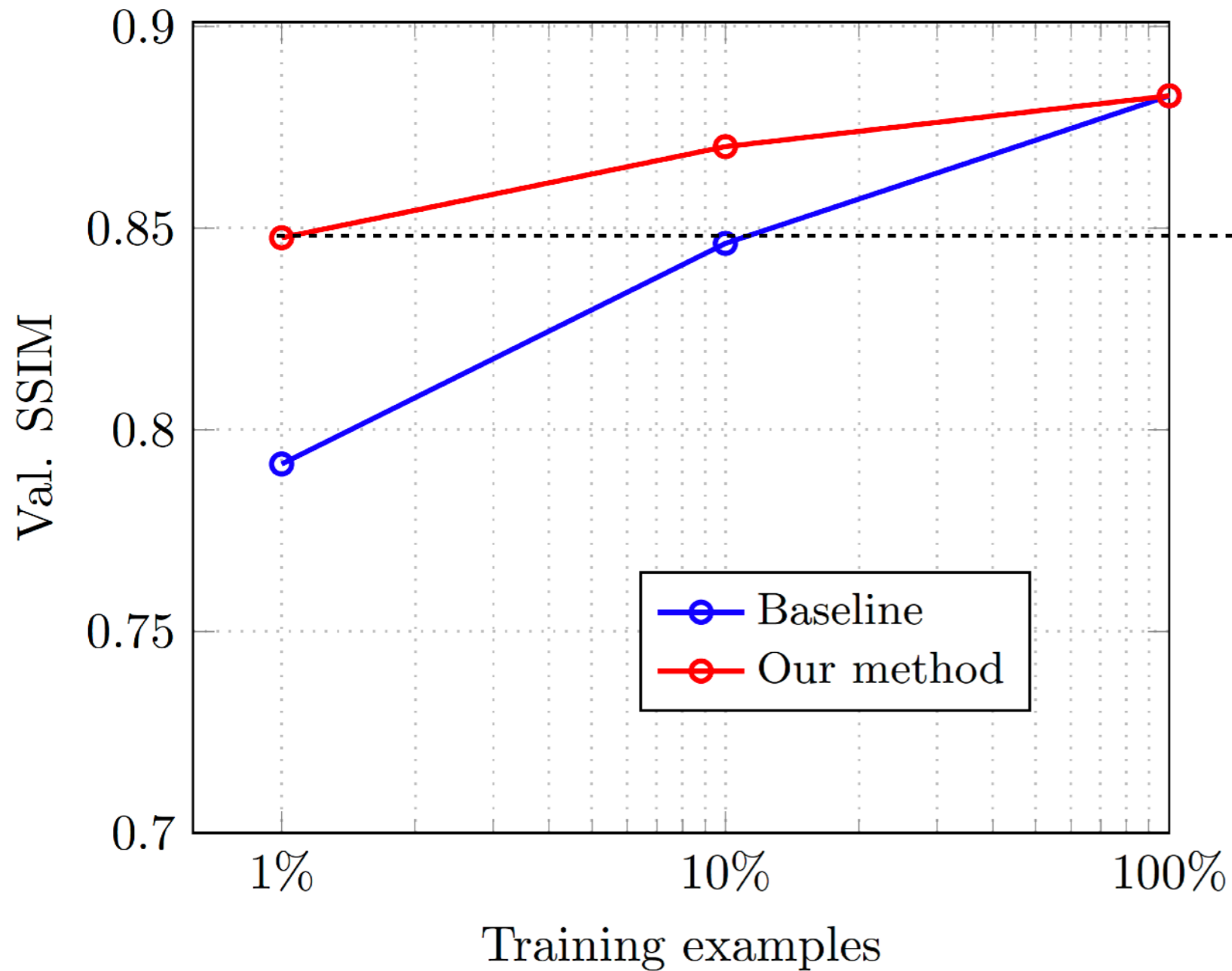
# Leveraging self-learned models for data reduction

# Detail encoding



# Detail encoding results

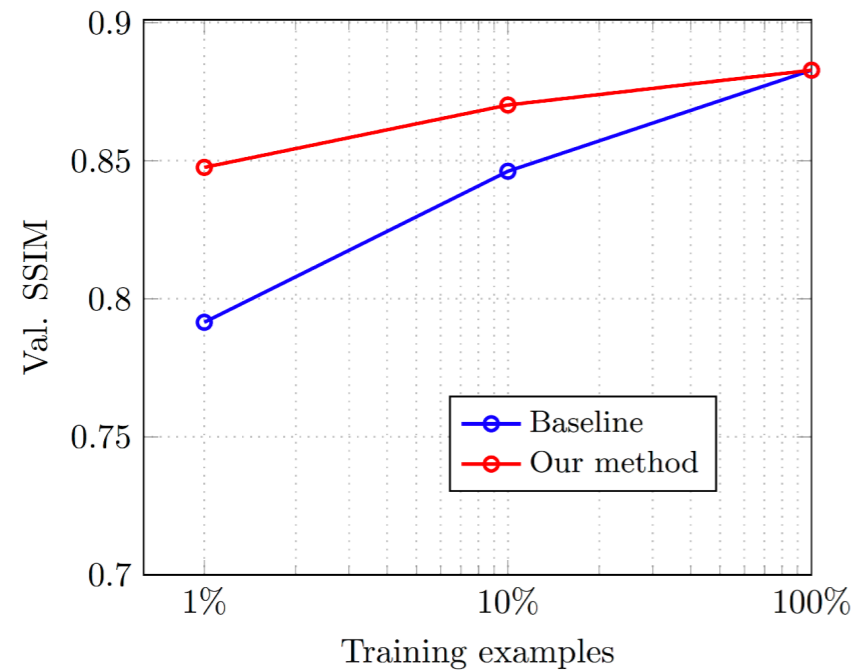
CelebA-HQ dataset



**10x data reduction!**

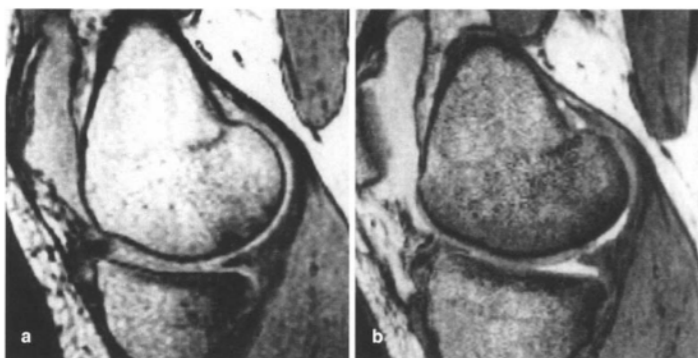
# Future work

- Closing the gap



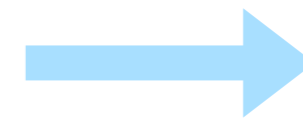
- 10x reduction in low-data regime ✓
- achieve 100% performance with 1-10% data

- Low-field MRI



0.2T

1.5T



Thank you for your attention!