

Self-Supervised Learning for MeerKAT Images

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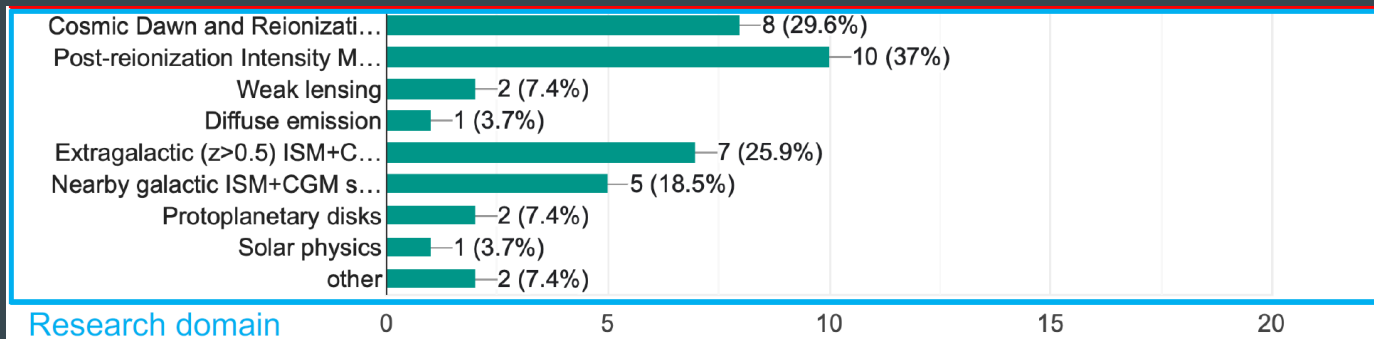
E. Lastufka

O. Bait, O. Taran, V. Kinakh, M. Audard, M. Dessauges-Zavadsky, T. Holotyak,
D. Schaerer, S. Voloshynovskiy

Motivation

Deliver products useful to the SKA community

- Adaptable to diverse range of science topics
- Scalable

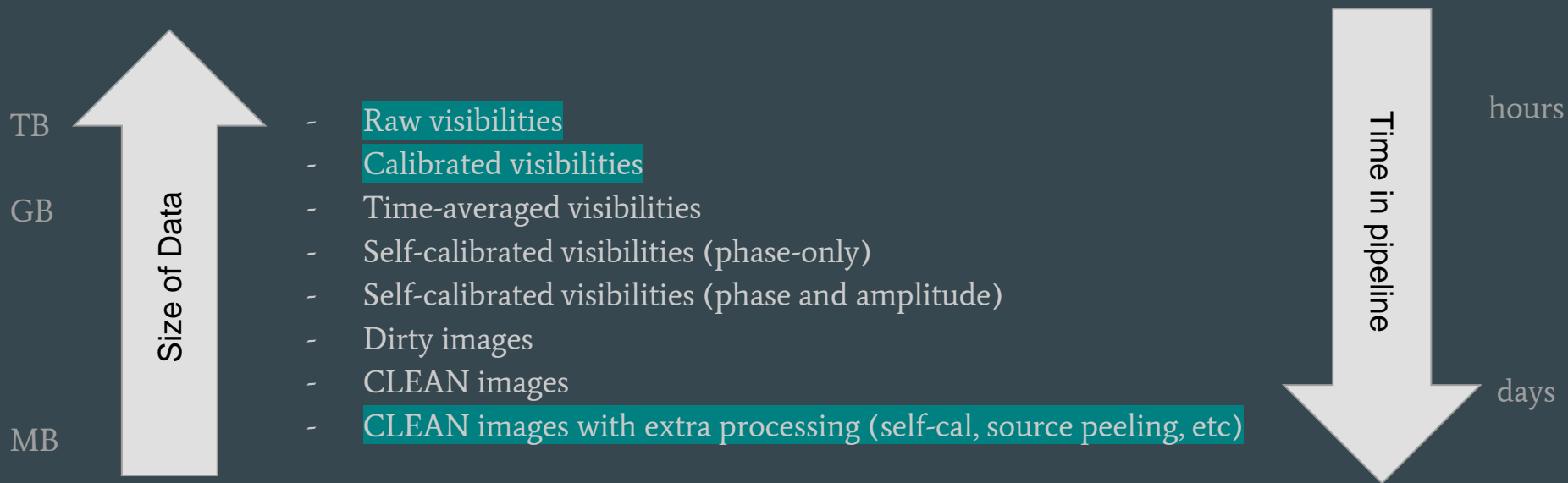


MeerKAT Datasets

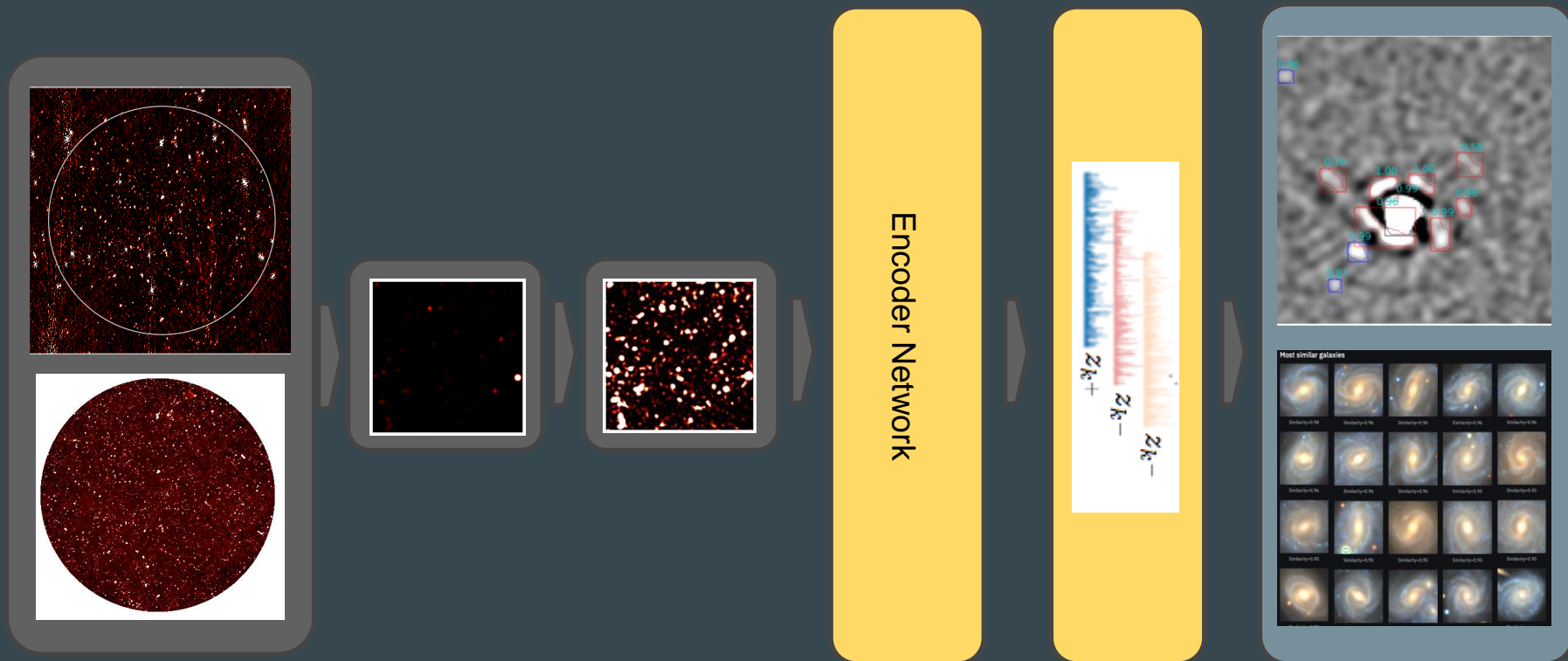
- L Band (900 MHz - 1.6 GHz)
- Continuum sensitivity of 1-2 $\mu\text{Jy}/\text{beam}$
- 6 - 15 hours time-on-source
- CLEAN images, source catalogs publicly available



Data Formats



Goal: Learn useful representations of radio data



Downstream Tasks

Welcome to Galaxy Finder
Created by [@George.Stein](#)
Interested in learning how this works?

Query galaxy
RA, Dec = (199.3324, 20.6382)

Most similar
Similarity=0.98
Similarity=0.96
Similarity=0.95

Galaxy Finder (Stein 2021)

ANOMALY SCORING VISUALISATION

Metadata
filename: /home/m
Class6.1: 0.916
score: 4.91
human_label: 5

Features
Residual_90: 2.58
Residual_80: 23.2
Residual_70: 12.9
Residual_60: 8.1

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HOW INTERESTING IS THIS OBJECT?

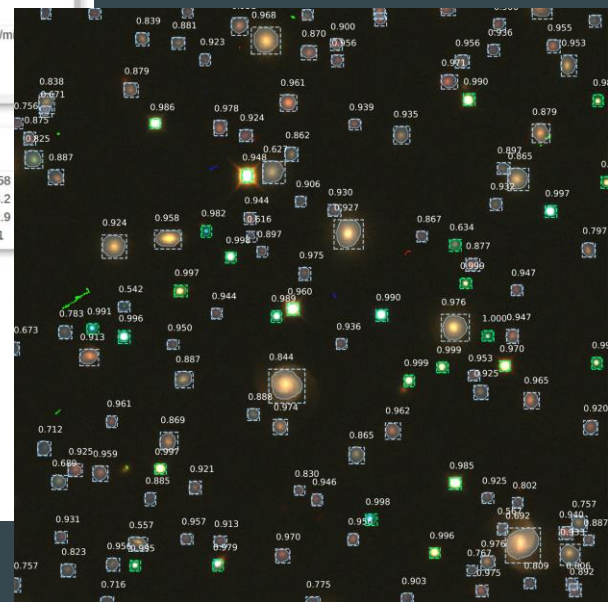
0 1 2 3 4 5

Raw anomaly score

Scoring method to sort by

RETRAIN

Astronomy (Lochner 2021)

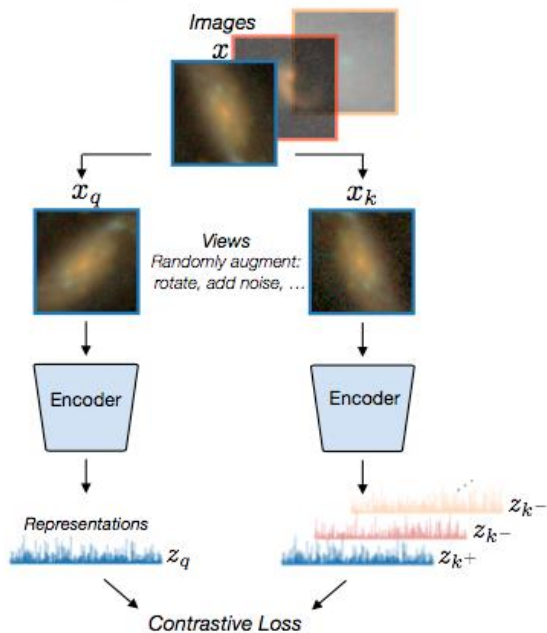


Astro-rcnn (Burke 2019)

What is possible?

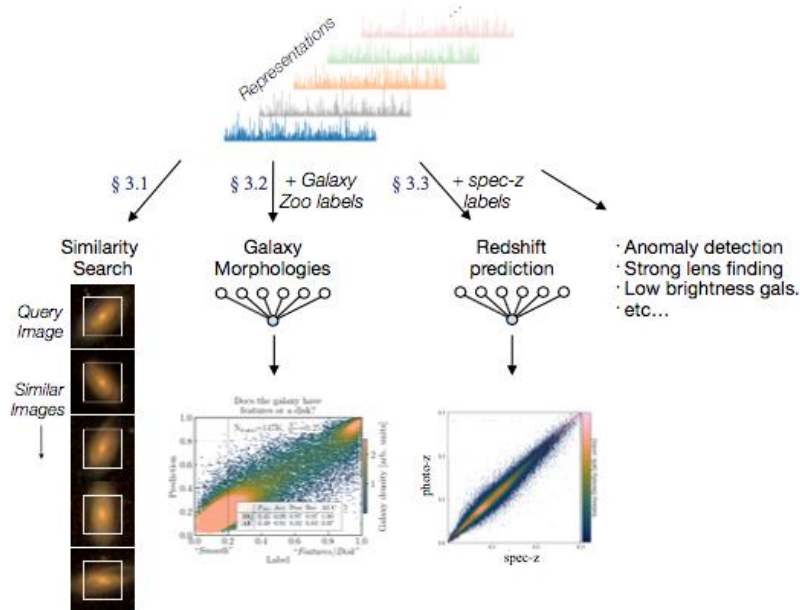
1. Self-supervised contrastive representation learning

Learn representations in an unsupervised manner



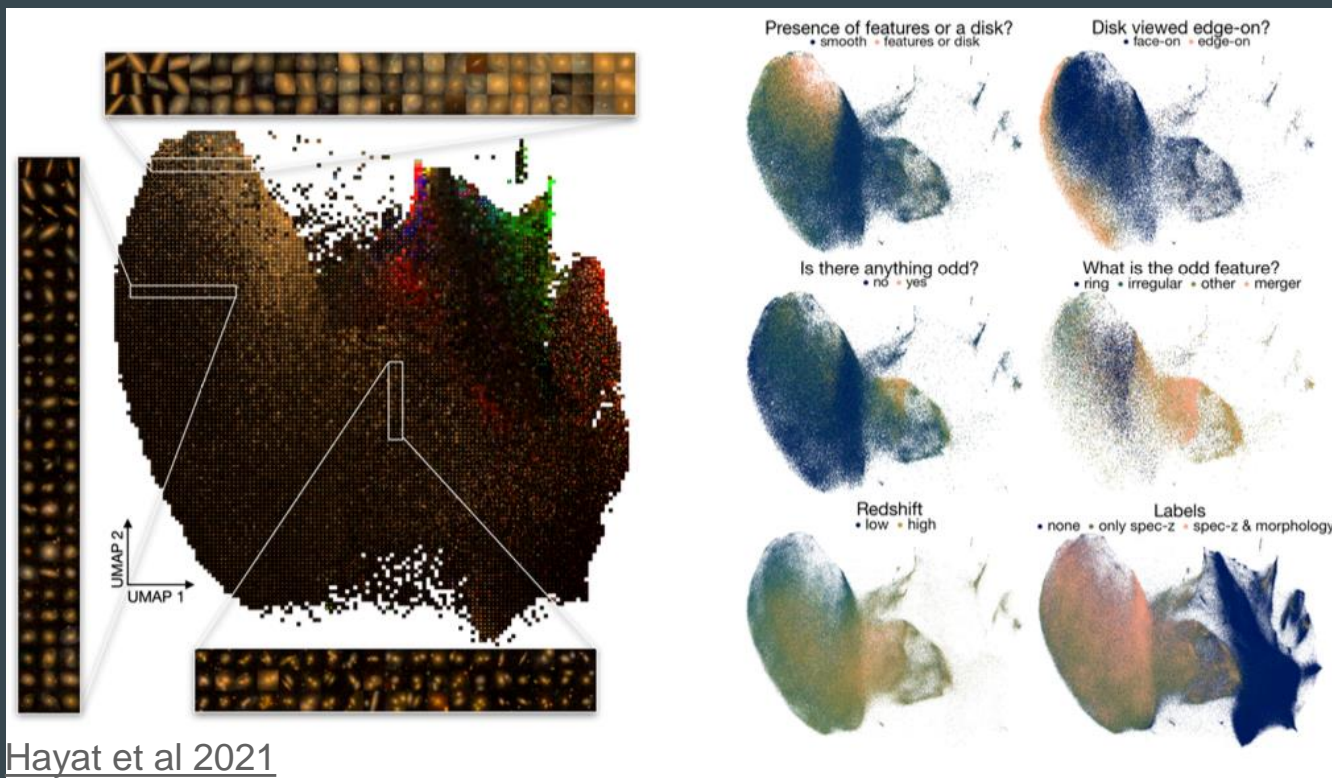
2. Downstream tasks

Use representations for a variety of applications



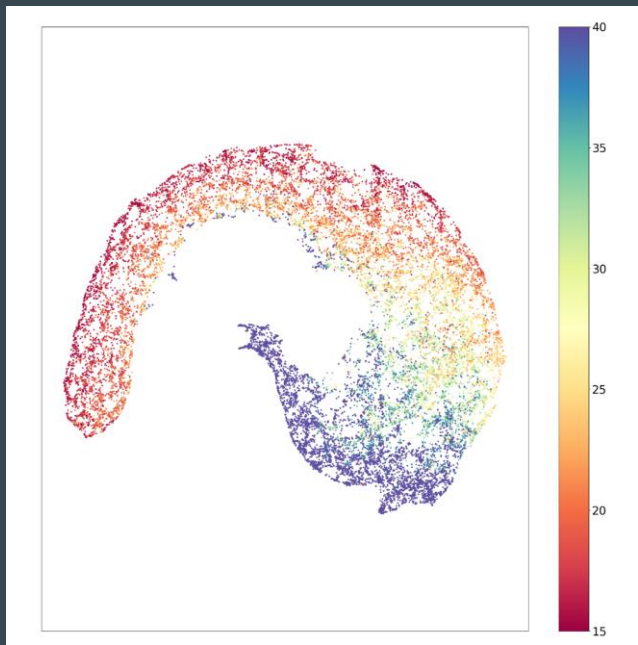
Hayat et al 2021

What is possible?



Hayat et al 2021

What is possible?



I. Slijepcevic et al, 2022 [🔗](#)

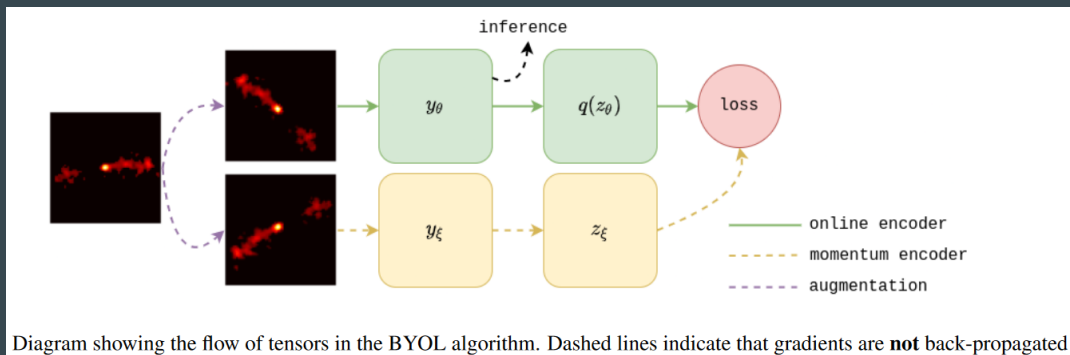


Diagram showing the flow of tensors in the BYOL algorithm. Dashed lines indicate that gradients are **not** back-propagated

Hayat et al 2021

Challenges

- Difficult to generalize from one survey to another
- Astrophysical images can be very large and contain many sources
- High dynamic range
- Need task-agnostic image augmentations

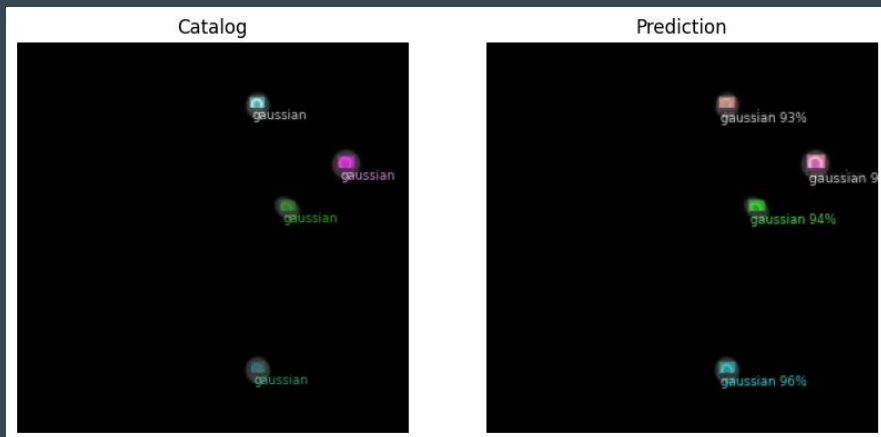
Radio-specific

- Work up until now has depended on CLEAN images
 - Slow to produce, often reconstructed to suit the science use case rather than in a standardized way
- Choice of channels is huge
- More variation possible than in optical

Solutions

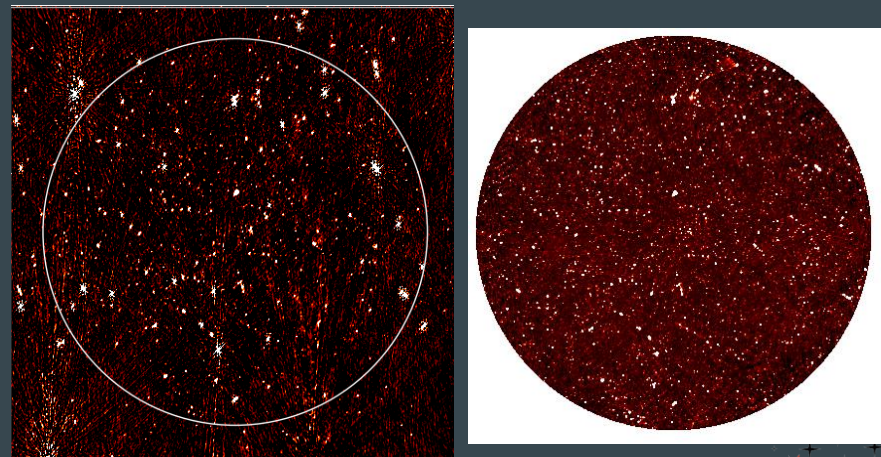
Use simulations in parallel

- Ability to change parameter space
- Reliable method of verification
- Early indicator of success

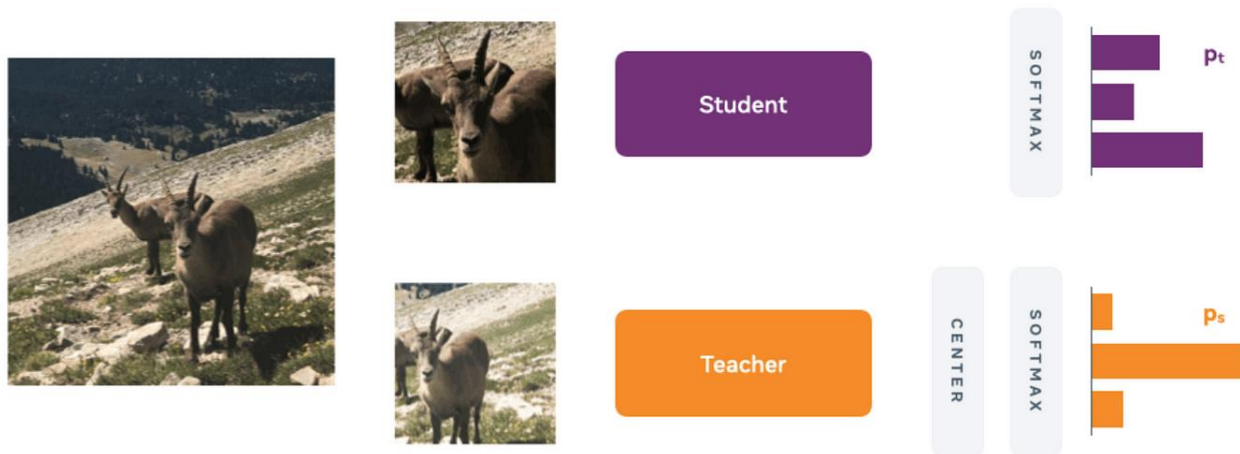


Dirty Images

- Contains same information as CLEAN
- Lower processing level
- Potential to generalize better

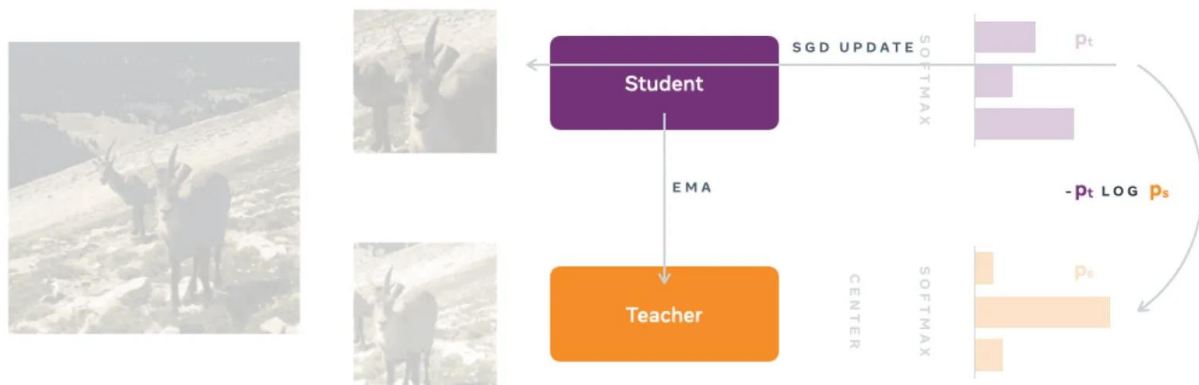


Self-Supervised Learning with DINO



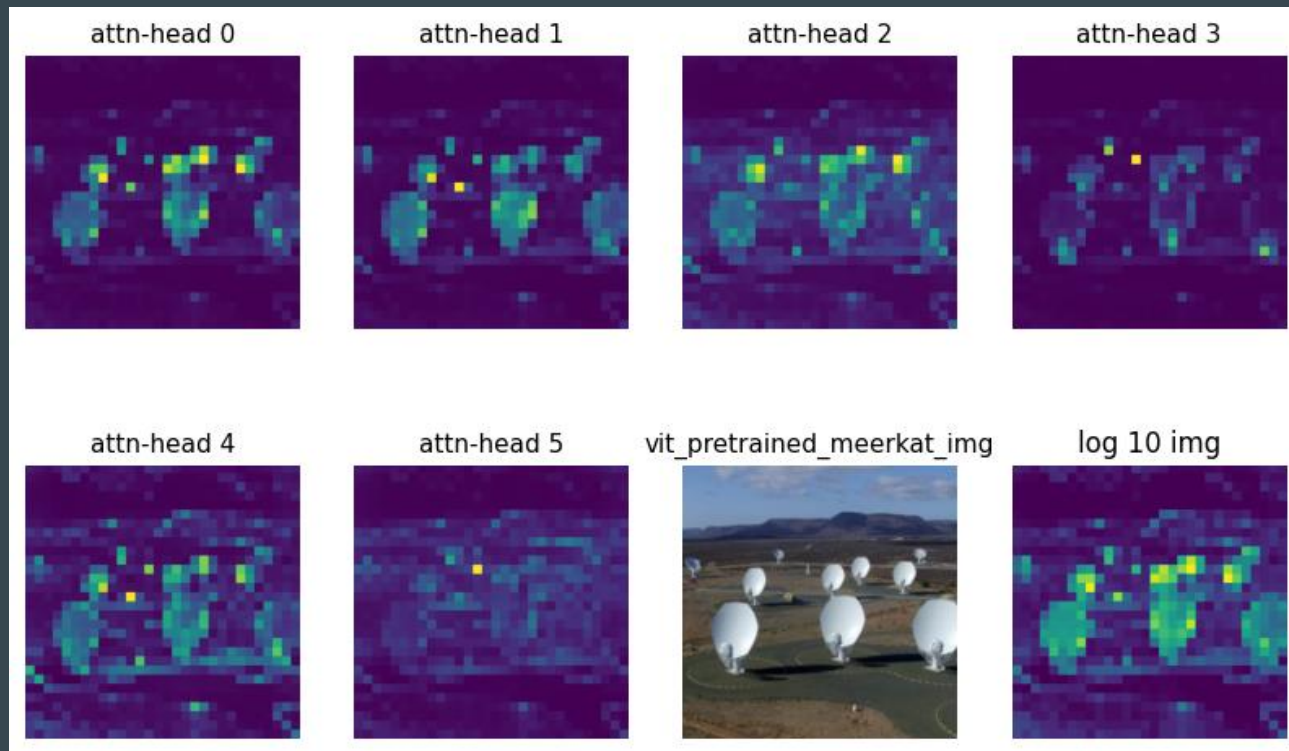
<https://github.com/facebookresearch/dino>

Self-Supervised Learning with DINO

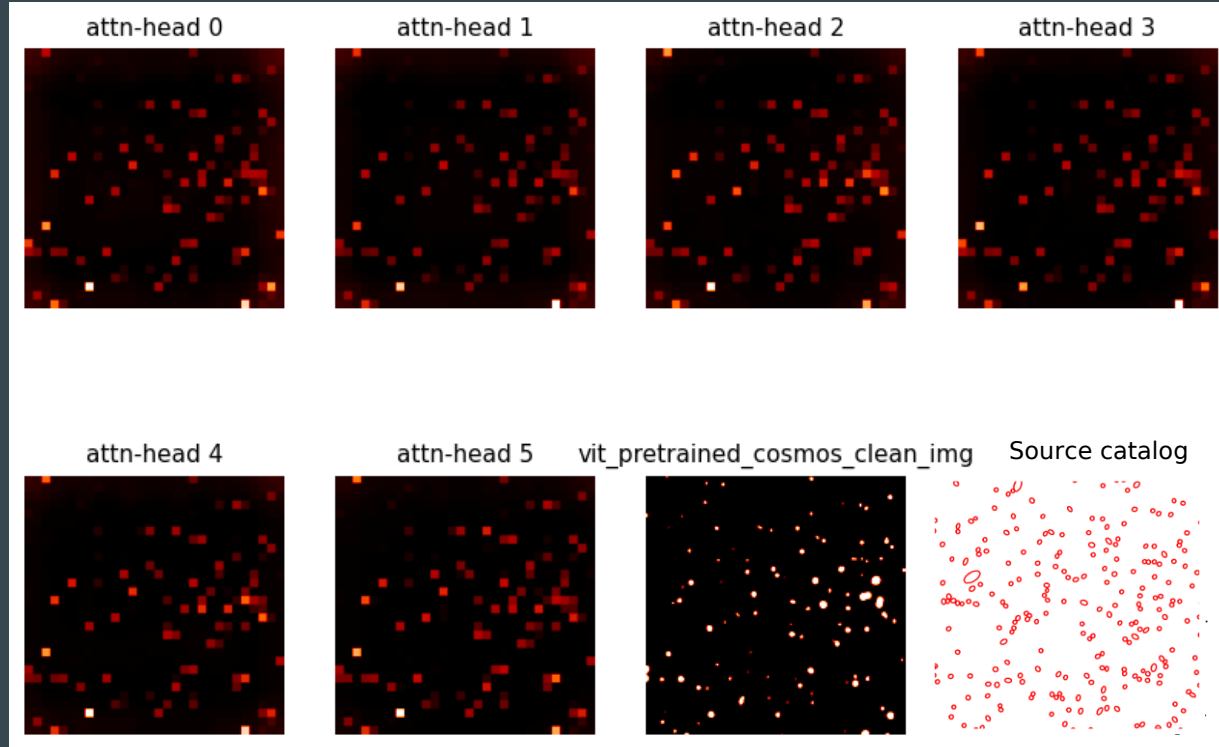


<https://github.com/facebookresearch/dino>

DINO is a promising approach!



DINO is a promising approach!



Motivational poster of a colorful stegosaurus with the words “Thank You”

