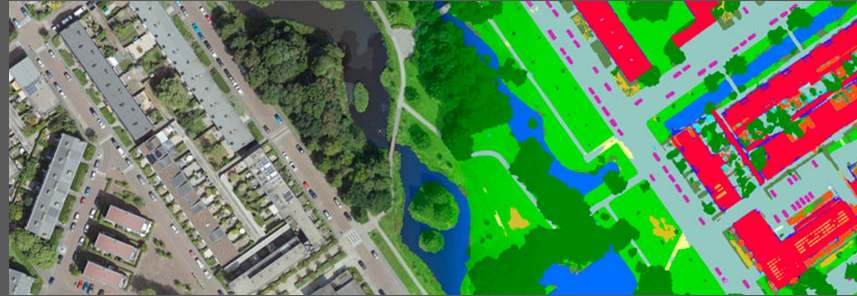


The Alan Turing Institute

Remote Sensing reading group

Conference round-up
18 July 2024



Outline

- Thijs van der Plas | CVPR (Earthvision)
- Anna Zanchetta & Barbara Metzler | ML4EO

GeoSynth

Stable diffusion (tuned). Generate HR sat given
OSM + gen captions + location. $p(s \mid m, c, l)$

Encoders for im, text, loc. Diffusion in latent space.



Control

Seattle,
Washington

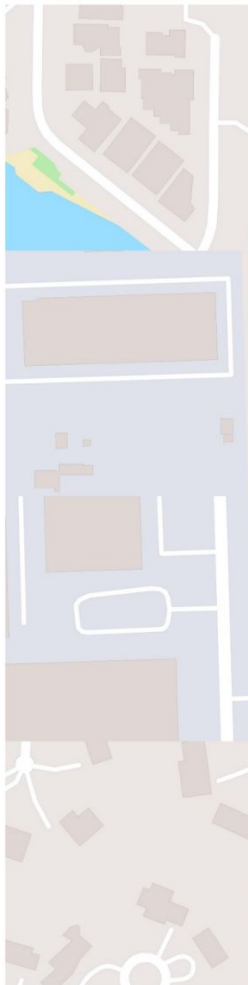
San Diego,
California

Des Moines,
Iowa

New York City,
New York

Boulder,
Colorado

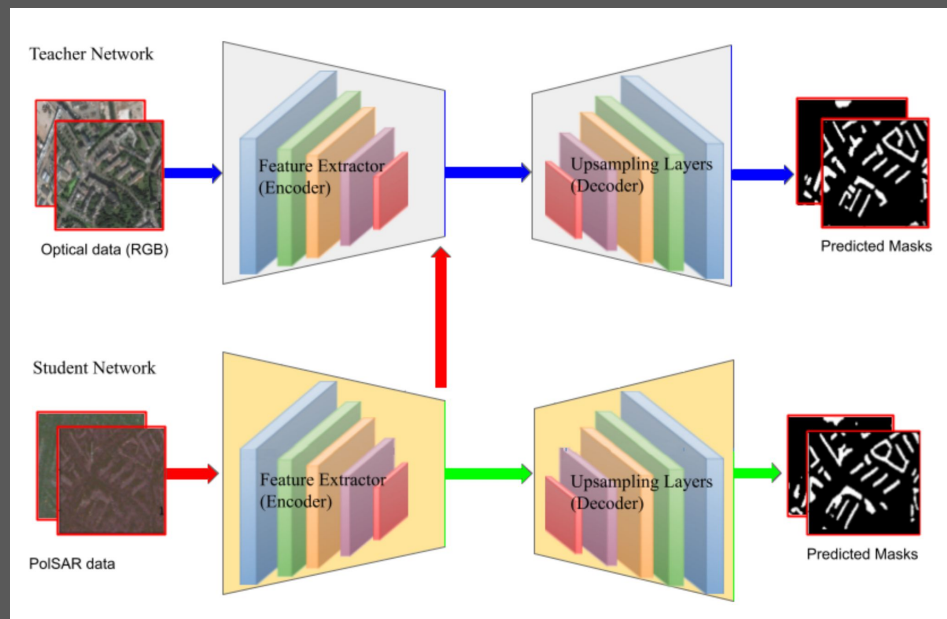
Dallas,
Texas



Cross model Knowledge Distillation

1. Train optical supervised
2. KD SAR (student) and optical (teacher)
3. Fine tune SAR supervised

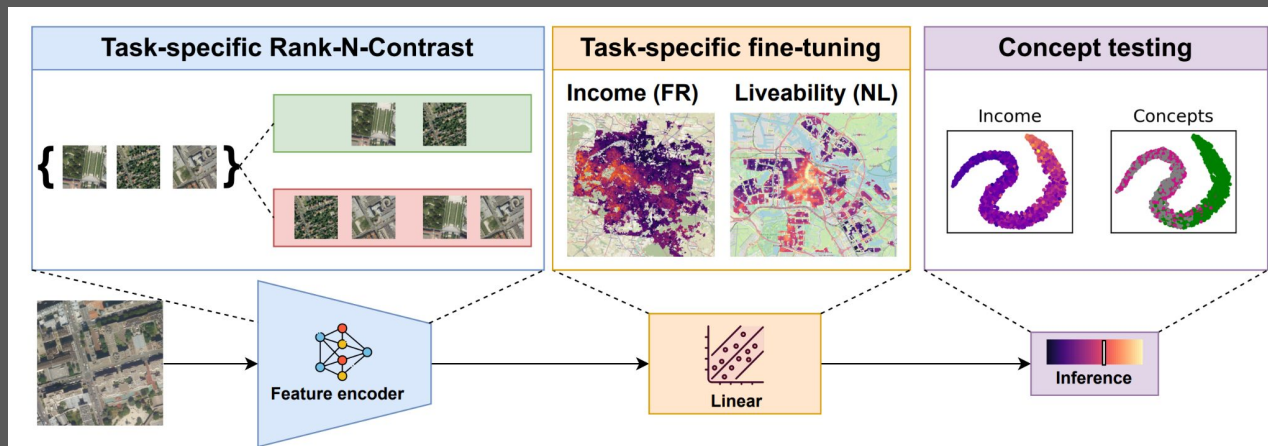
Improves SAR model (but does not exceed optical model)



Contrastive pretraining for explainability

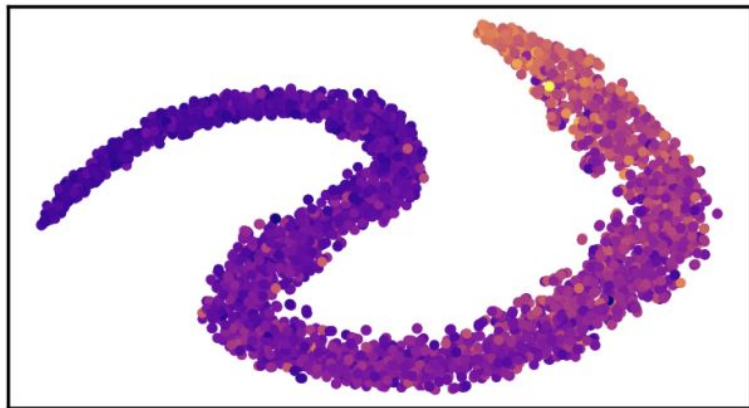
Rank-N-contrast (Neurips 2023)

1. Train Resnet with CL.
2. Freeze and train final layer on task.



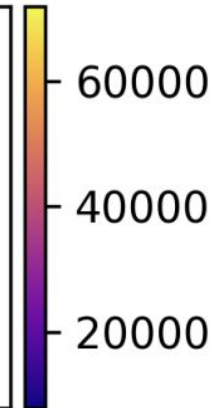
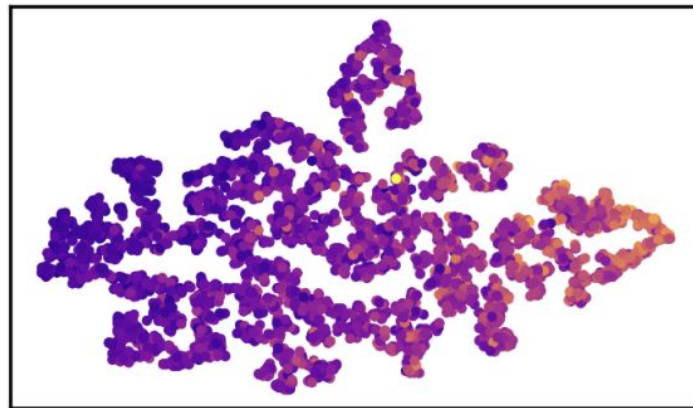
Rank-N-Contrast loss

Income

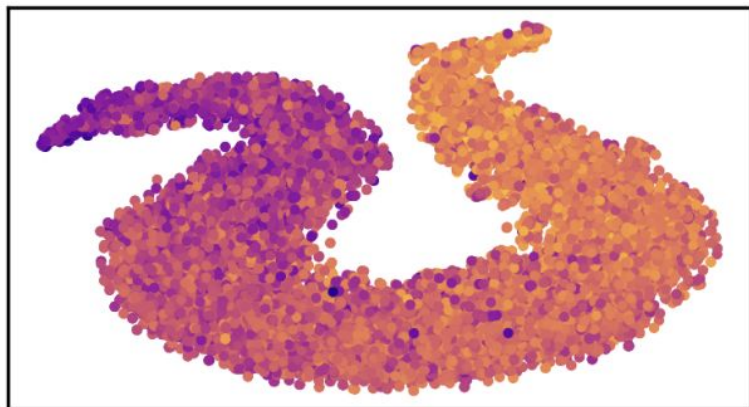


L_1 loss

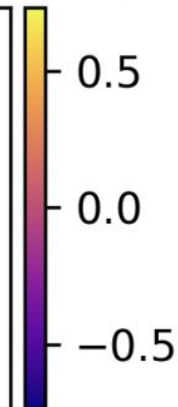
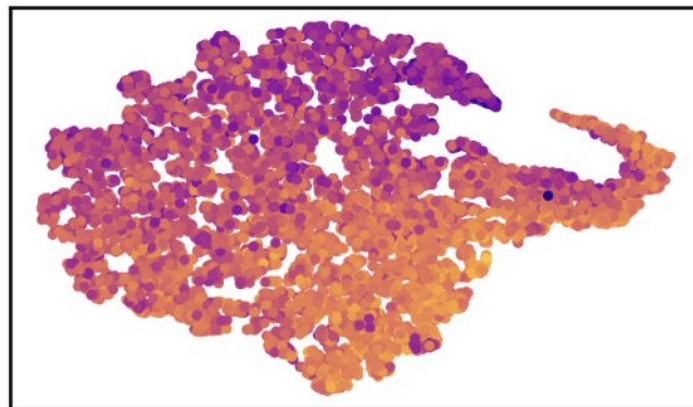
Income



Liveability



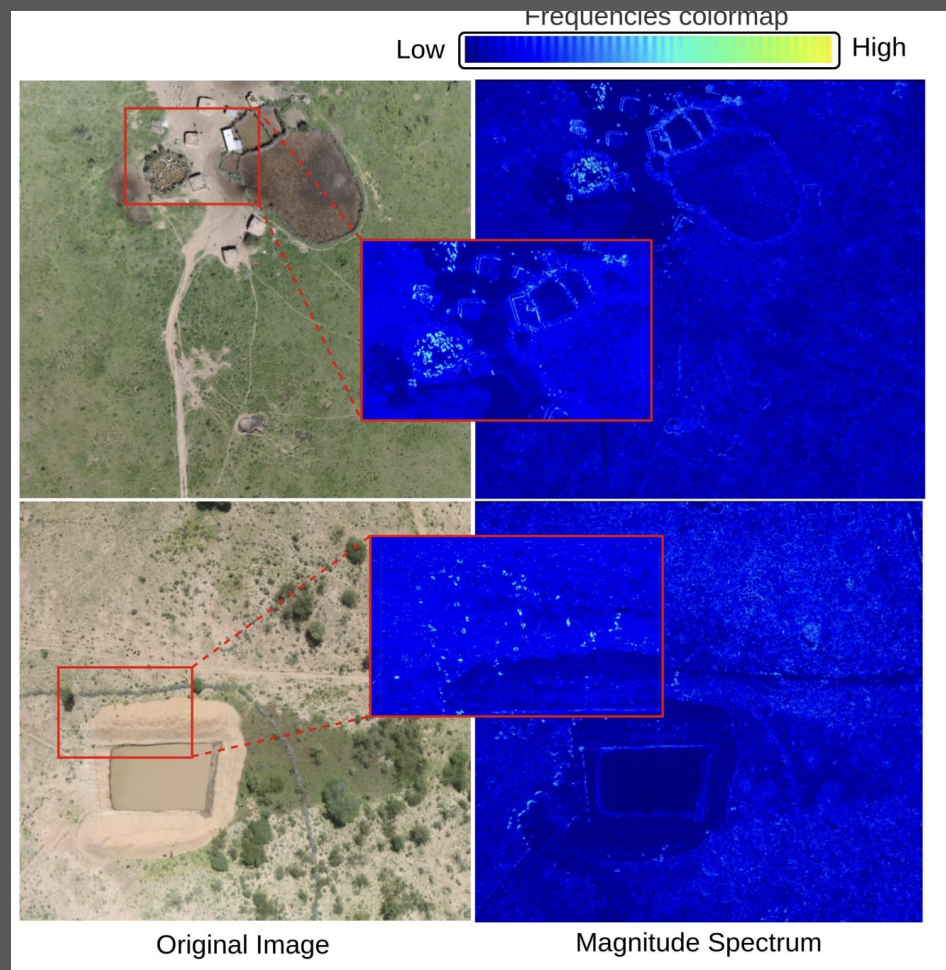
Liveability

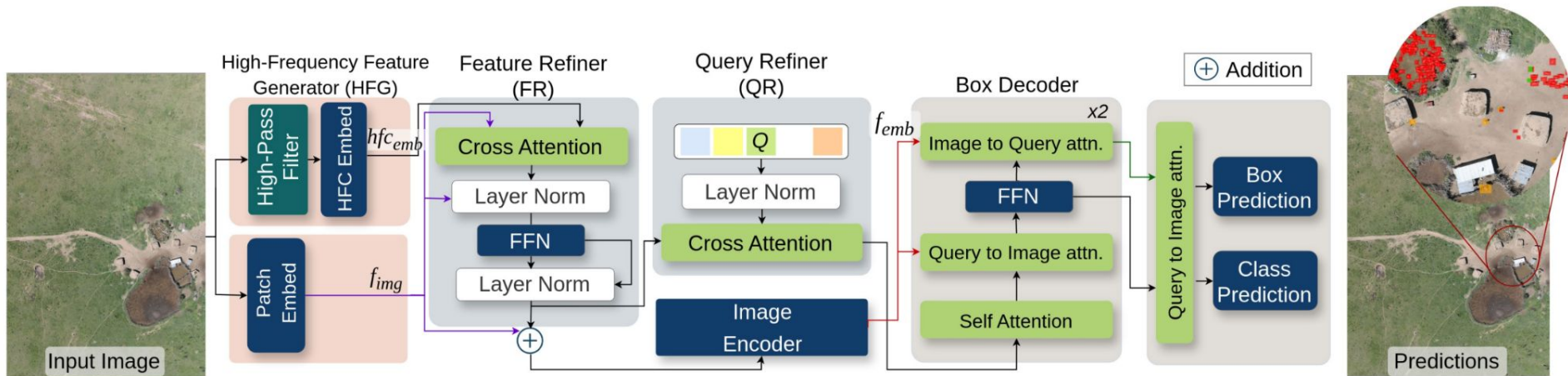


WildlifeMapper

Object detection (large mammals) -> very small

Generate high freq map
via FT, use with high
pass filter







Foundation models

Benchmarking GPT4

Scenario

Example

Results

Location Recognition [2.1]

- 👤: Make an educated guess about the name of the landmark shown in the image.
- 🤖: Based on the style, the dome, and layout of the grounds, my choice is: A. Nebraska State Capitol

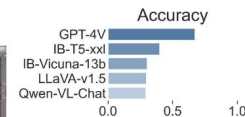
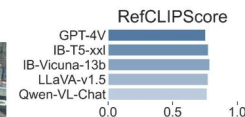


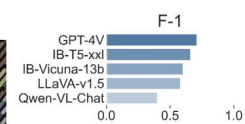
Image Captioning [2.2]

- 👤: Generate a caption for the image in one sentence.
- 🤖: Aerial view of an airport terminal with adjacent aircraft, taxiways, and parking areas.



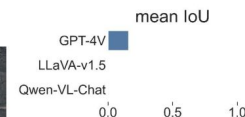
Land Use & Land Cover Classification [2.3]

- 👤: Classify the image into one of the following options.
- 🤖: The best option to describe the given image is: 28. Shipping Yard



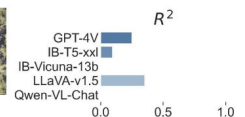
Object Localization [3.1]

- 👤: Identify the extent of the object in the description below. Description: The gray windmill in the middle
- 🤖: [233, 383, 376, 542]



Object Counting [3.2]

- 👤: Count the number of trees in the given image to the best of your ability.
- 🤖: 134



Change Detection [4]

- 👤: Count the number of buildings in each damage category in JSON format.
- 🤖:

```
```json
{"count_before": 75,
 "no_damage": 2,
 "minor_damage": 73,
 "major_damage": 0,
 "destroyed": 0}
```
```

