

Seminar: Hot topics in Computer Vision

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DALL-E: "Generate a high-resolution photorealistic satellite image from one of the Sentinel-1 satellites with false colours."



Motivation

urban?



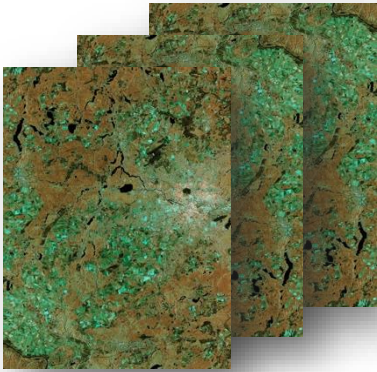
agriculture



water?



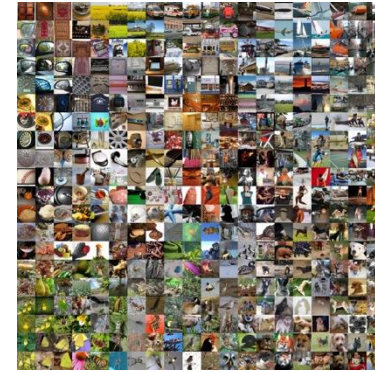
Motivation



Sentinel I/II



Landsat



ImageNet

...

Trained using **contrastive learning** or **reconstruction learning**.

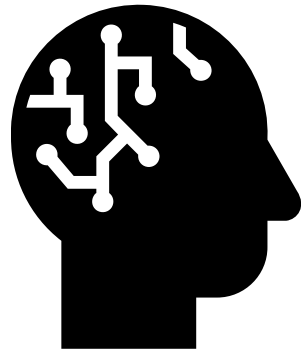


Foundational Model



Goal

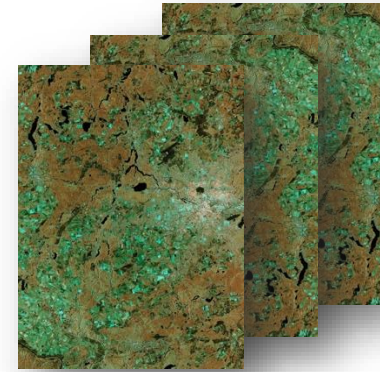
“Neural Plasticity-Inspired Foundation Model for Observing the Earth Crossing Modalities” - Zhitong Xiong et al. - 2024



DOFA Model



Test & Benchmark

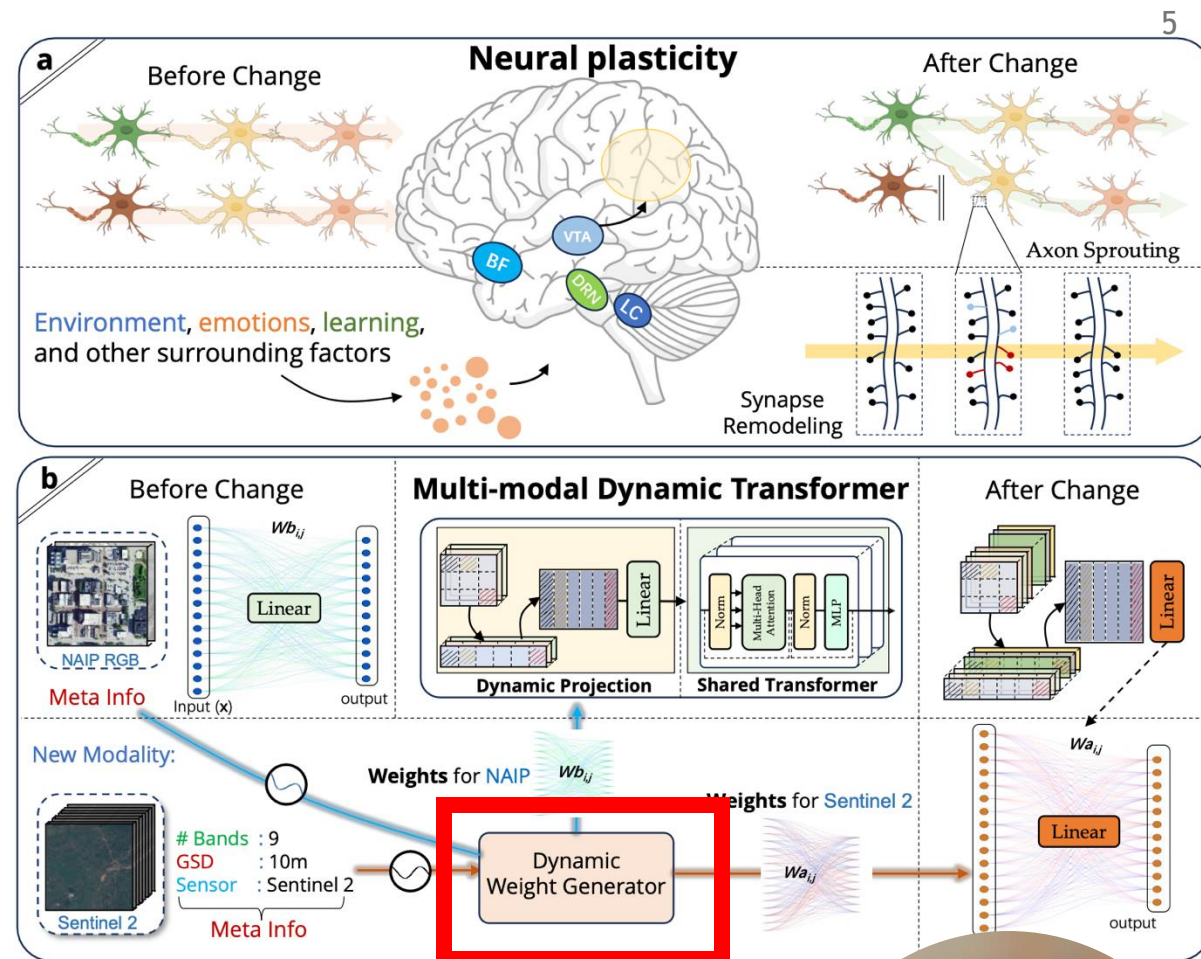
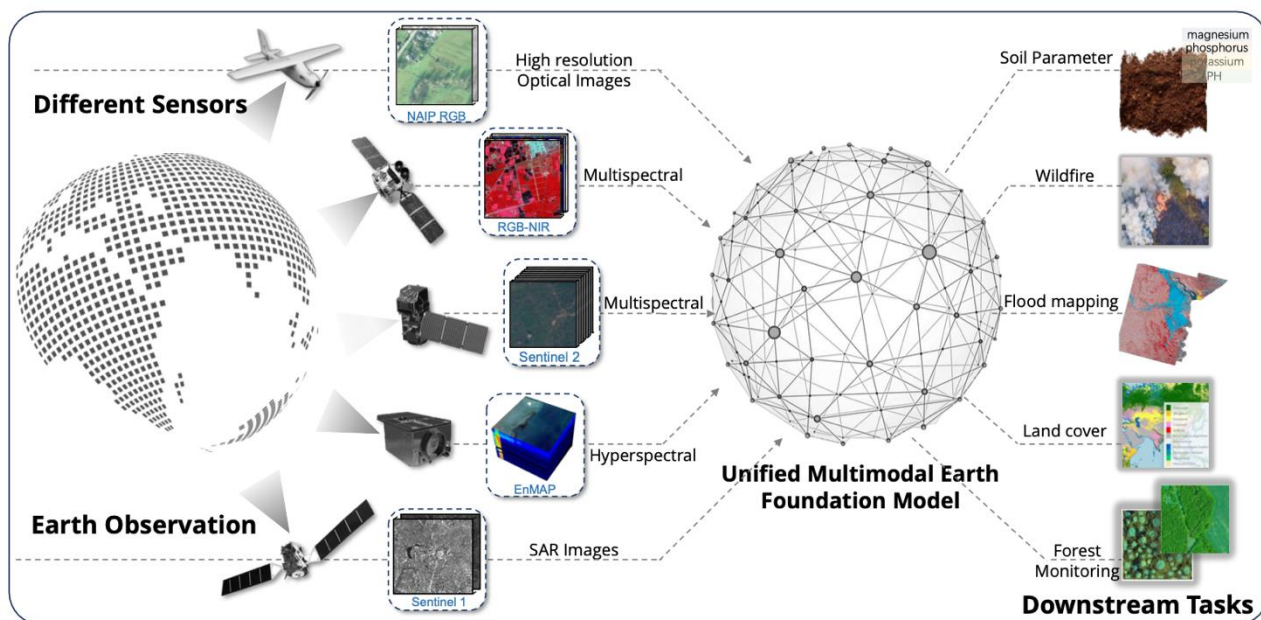


BigEarthNet

- Foundational model for remote sensing
- Multimodal



DOFA stands for "Dynamic-One-For-All"



1st base concept:
Hypernetwork

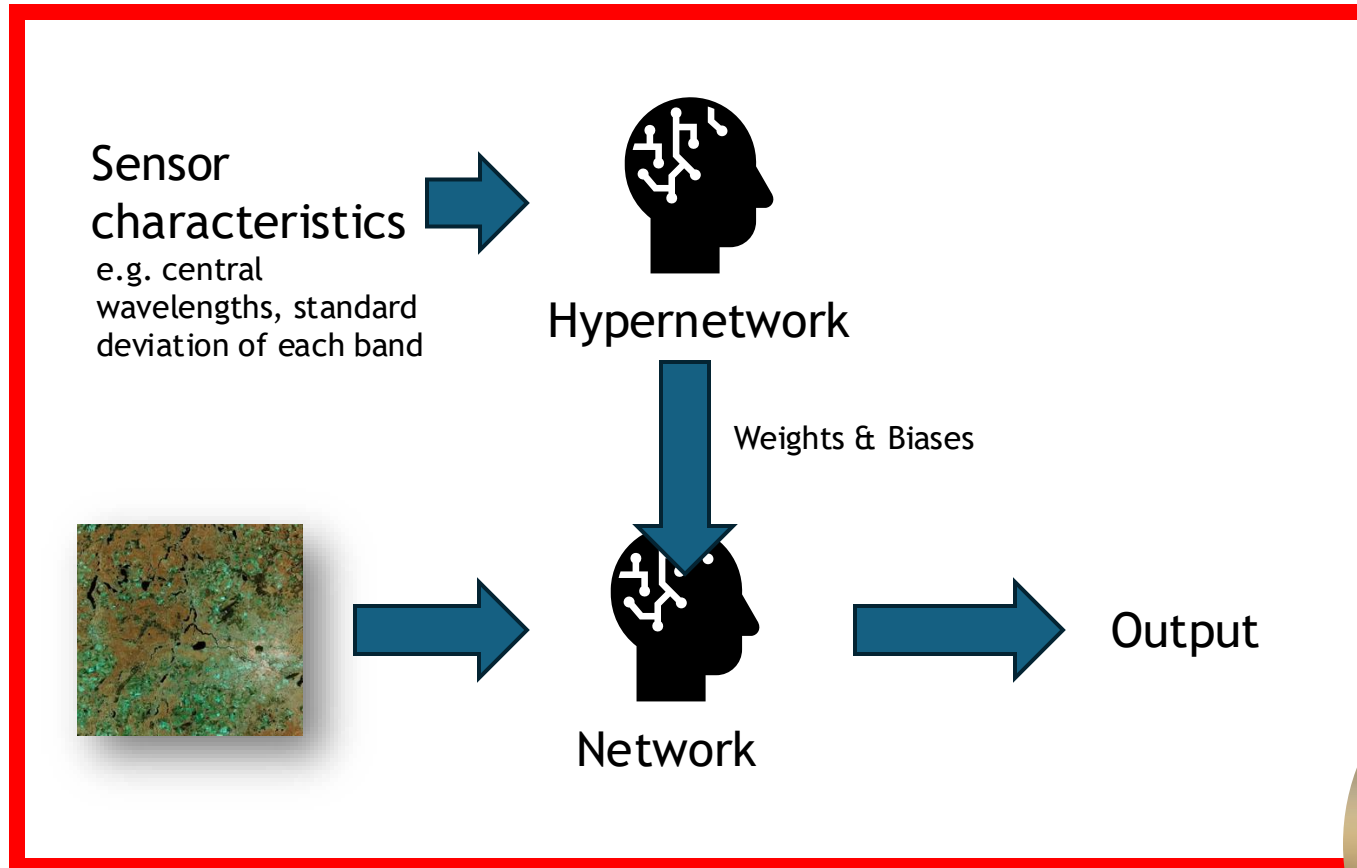


DOFA stands for "Dynamic-One-For-All"

1st base concept:
Hypernetwork

Goal:

- "reduce computational overhead" and complexity
- Multimodality (also on unseen data)



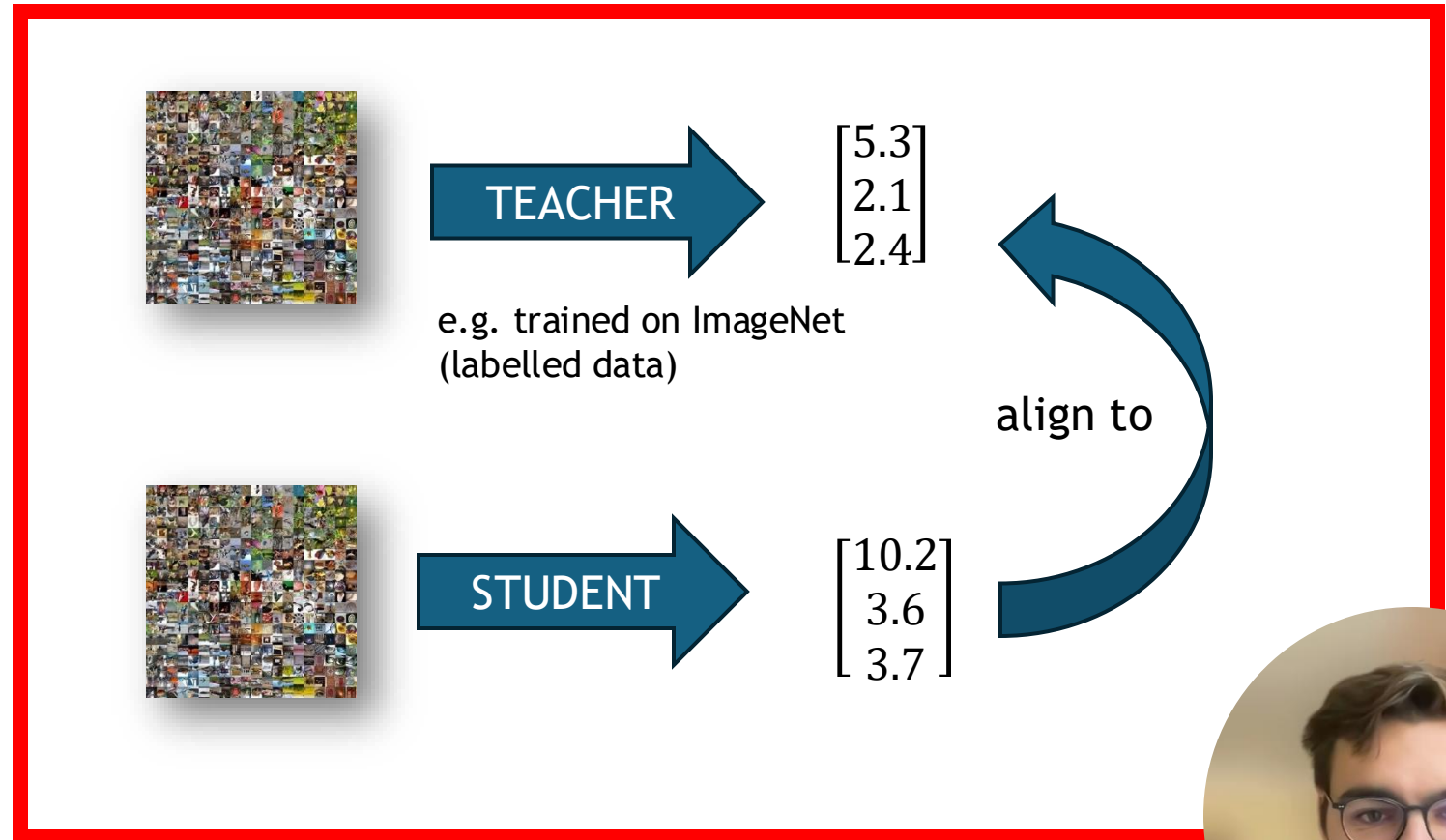
DOFA stands for "Dynamic-One-For-All"

2nd base concept: Distillation loss

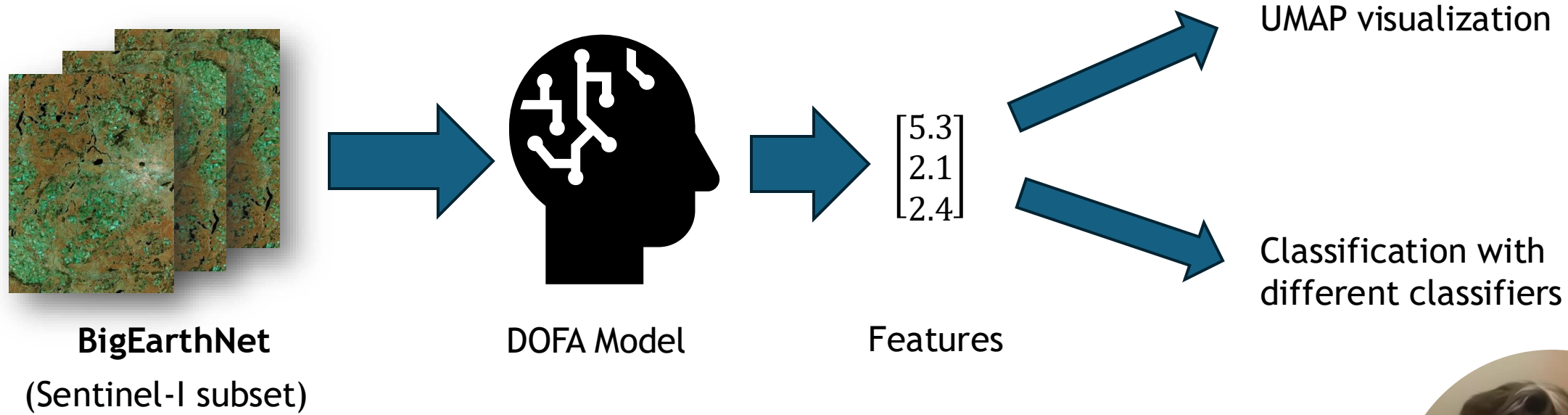
Goal:

- Accelerate training convergence
- Enhance overall performance

Combined with a reconstruction loss
→ Reproduce input data correctly +
produce informed representations

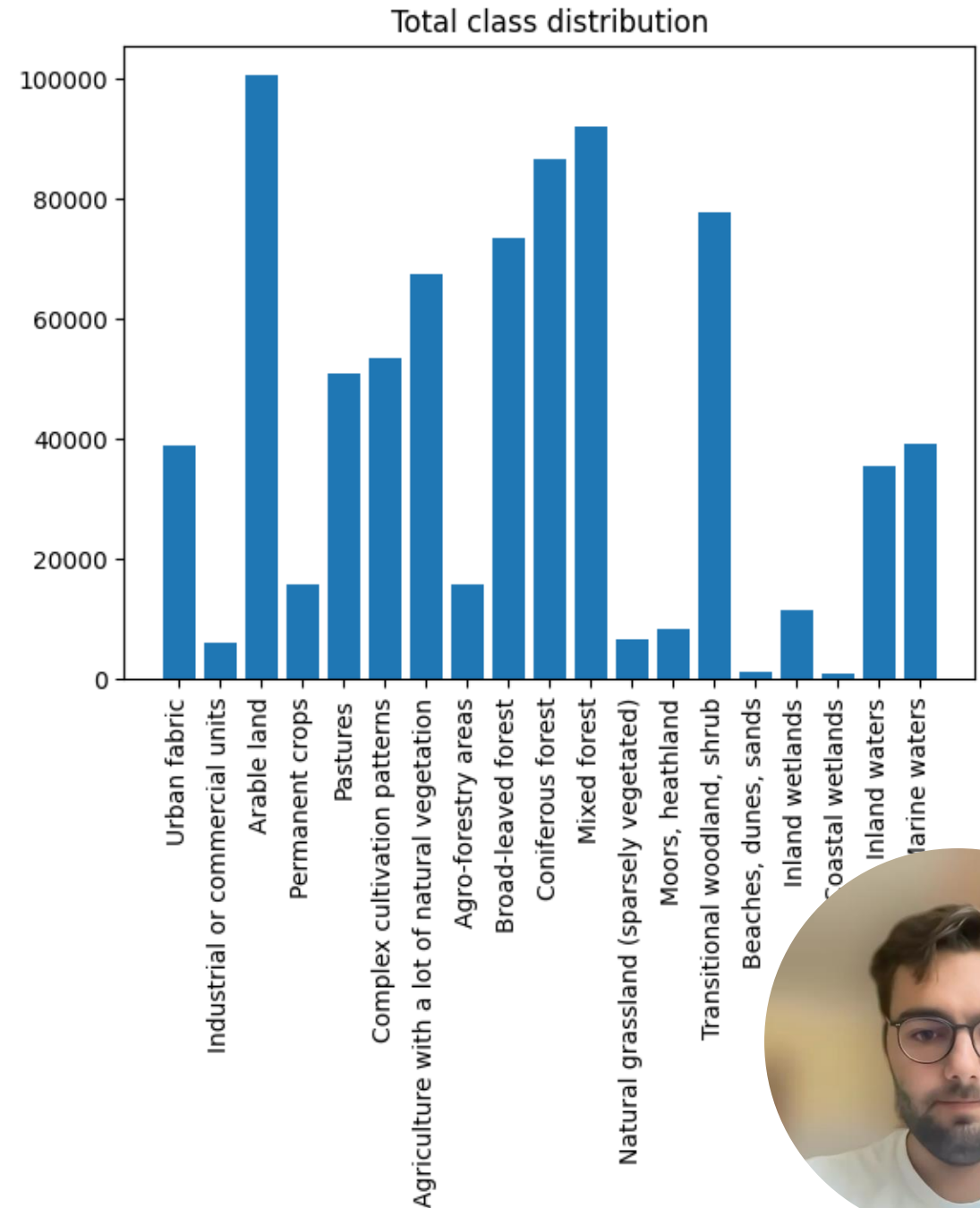


Methodology



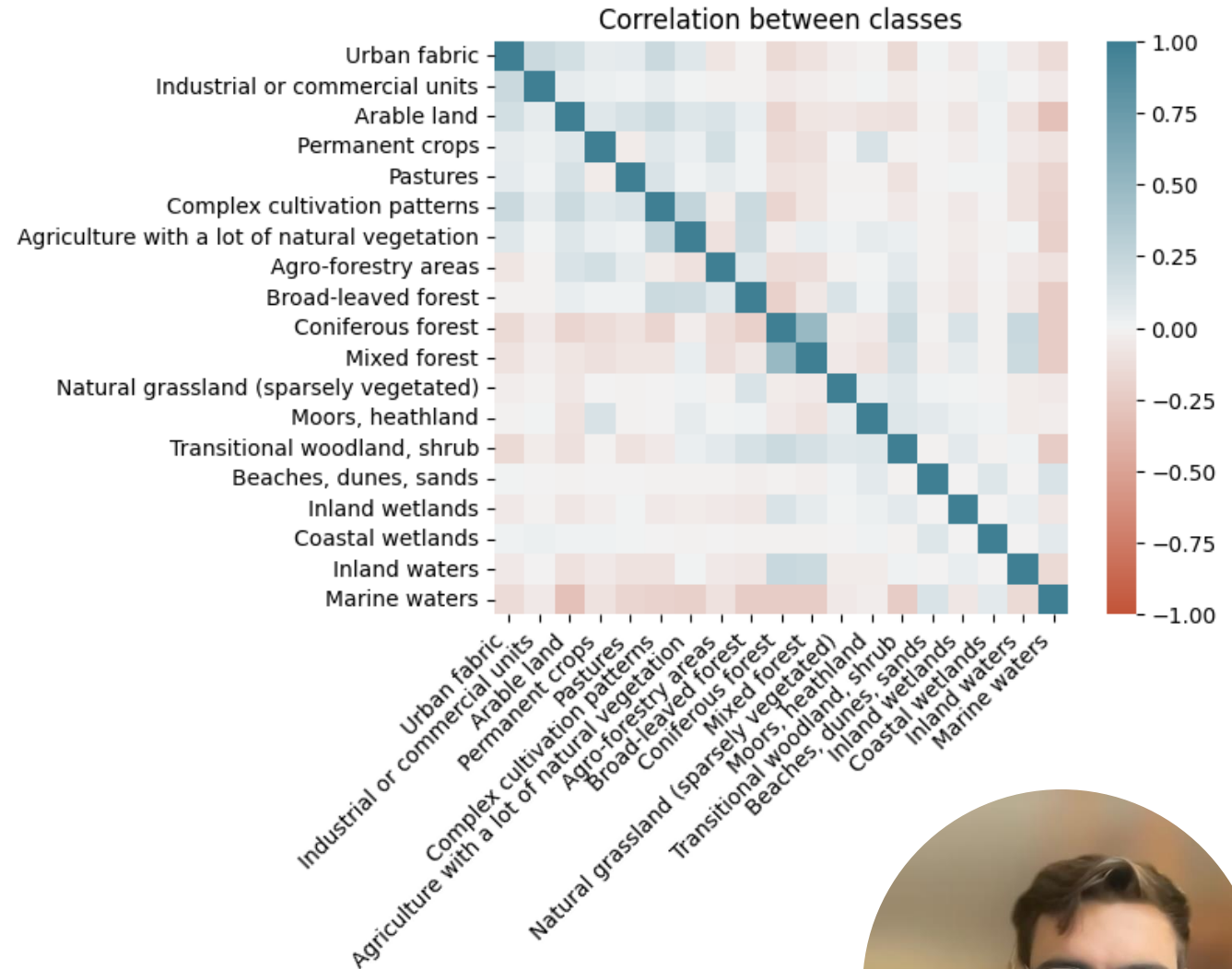
About the Dataset

- 550.000 labeled image patches from **Sentinel-I** and Sentinel-II
- 19 classes or 43 classes
- Split into train and test set by torchgeo repository



Dataset

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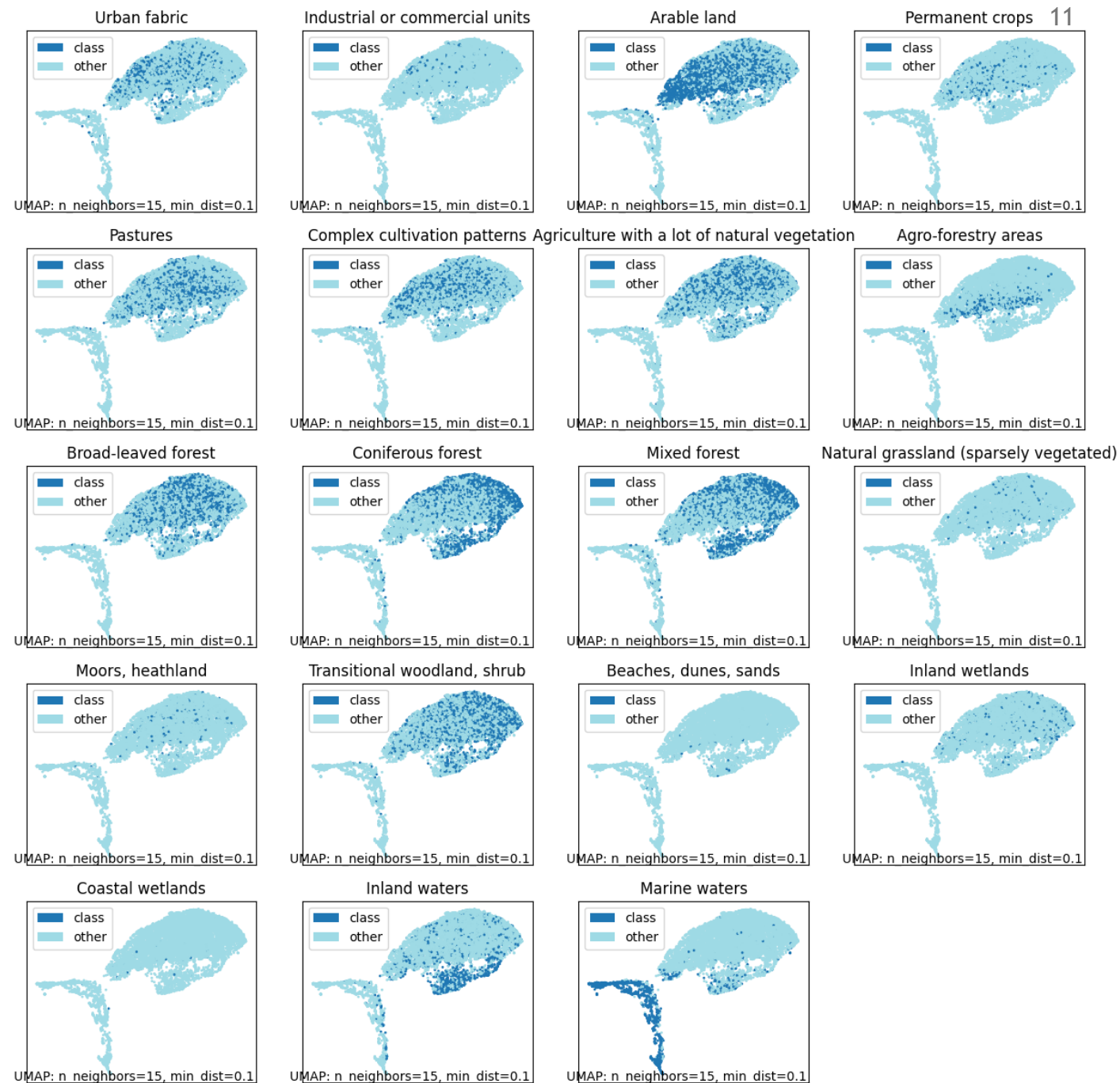


UMAP feature analysis

- UMAP transformation on DOFA features
- labelled with 19 classes



UMAP visualization of BigEarthNet features (One vs Rest labels)



Classification results

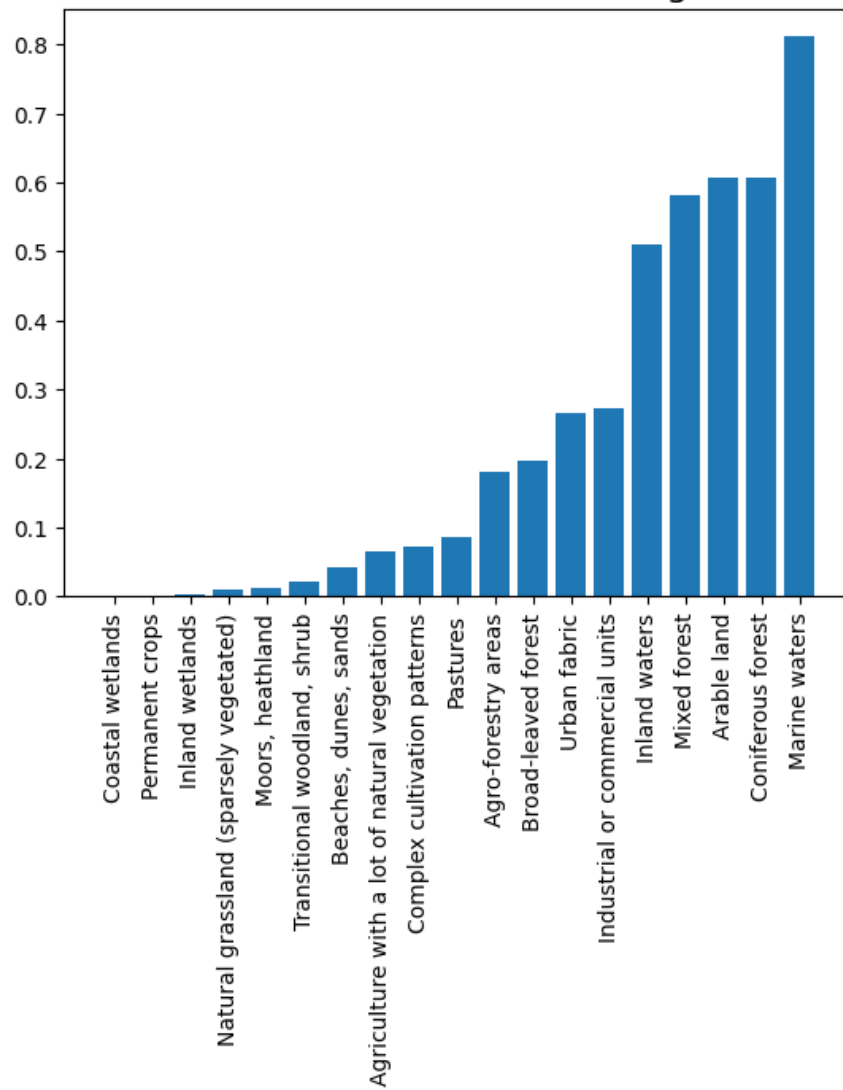
Tests results on BigEarthNet using DOFA feature vectors and 19 classes:

	$F^2_{macro} (\%)$	$F^2_{micro} (\%)$	hamming loss	$P_{macro} (\%)$	$P_{micro} (\%)$
Random Forest	21.4	35.8	0.123	52	72
Linear Probing	22.9	35.3	0.124	49	70
MLP	34.5	47.8	0.113	58	71

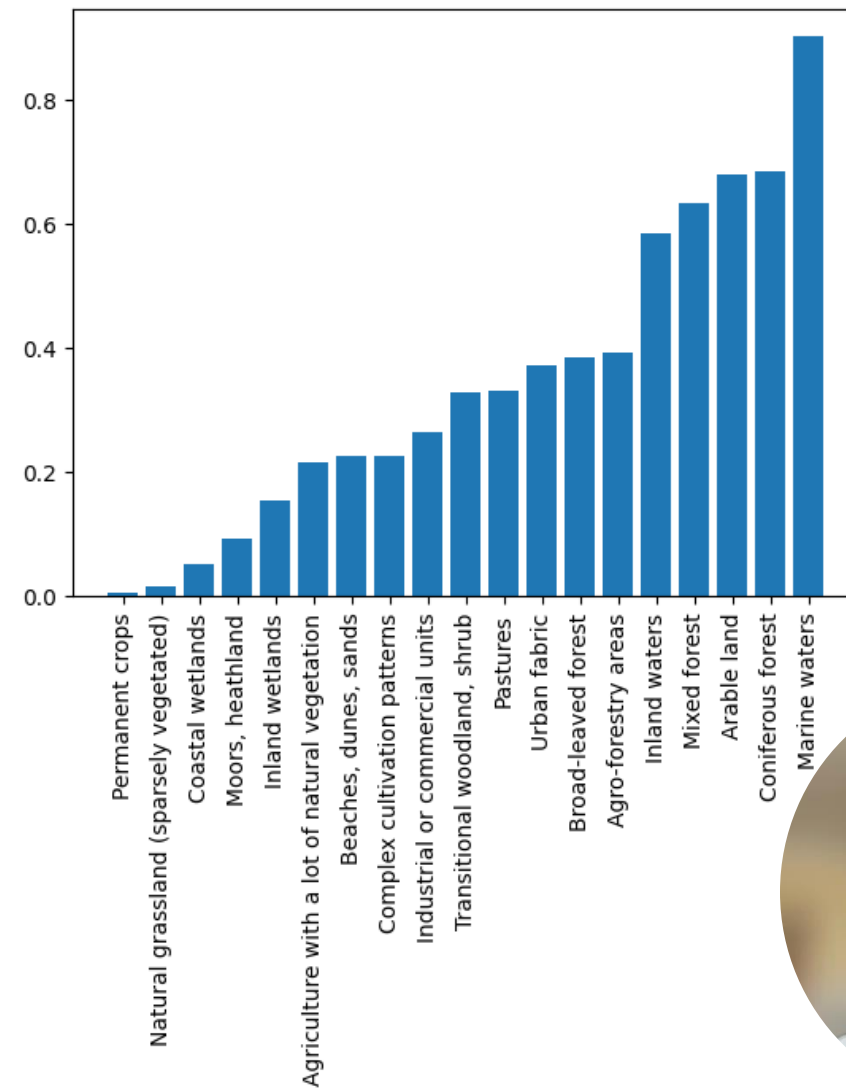


Classification results

F2-Scores - Linear Probing



F2-Scores - MLP



Key takeaway

- DOFA is able to produce meaningful features from the BigEarthNet
- DOFA has low performance for low-data-availability classes



References

- Zhitong Xiong, Yi Wang, Fahong Zhang, Adam J. Stewart, Joëlle Hanna, Damian Borth, Ioannis Papoutsis, Bertrand Le Saux, Gustau Camps-Valls, and Xiao Xiang Zhu. Neural plasticity-inspired multimodal foundation model for earth observation, 2024
- [https://www.esa.int/ESA_Multimedia/Missions/Sentinel-1/\(sortBy\)/view_count/\(result_type\)/images](https://www.esa.int/ESA_Multimedia/Missions/Sentinel-1/(sortBy)/view_count/(result_type)/images)
- https://eoimages.gsfc.nasa.gov/images/imagerecords/153000/153149/princecharles_oli2_20240715_lrg.jpg
- <https://cs.stanford.edu/people/karpathy/cnnembed/>
- Zhenda Xie, Zheng Zhang, Yue Cao, Yutong Lin, Jianmin Bao, Zhuliang Yao, Qi Dai, and Han Hu. Simmim: A simple framework for masked image modeling, 2022
- Leland McInnes, John Healy, and James Melville. Umap: Uniform manifold approximation and projection for dimension reduction,

