

SyntEO: Synthetic training data generation for offshore wind energy infrastructure detection in Sentinel-1 imagery

ECMWF-ESA 14-17 November 2022

Workshop on Machine Learning for Earth Observation and Prediction

Thorsten Hoeser

Prof. Dr. Claudia Kuenzer

German Aerospace Center

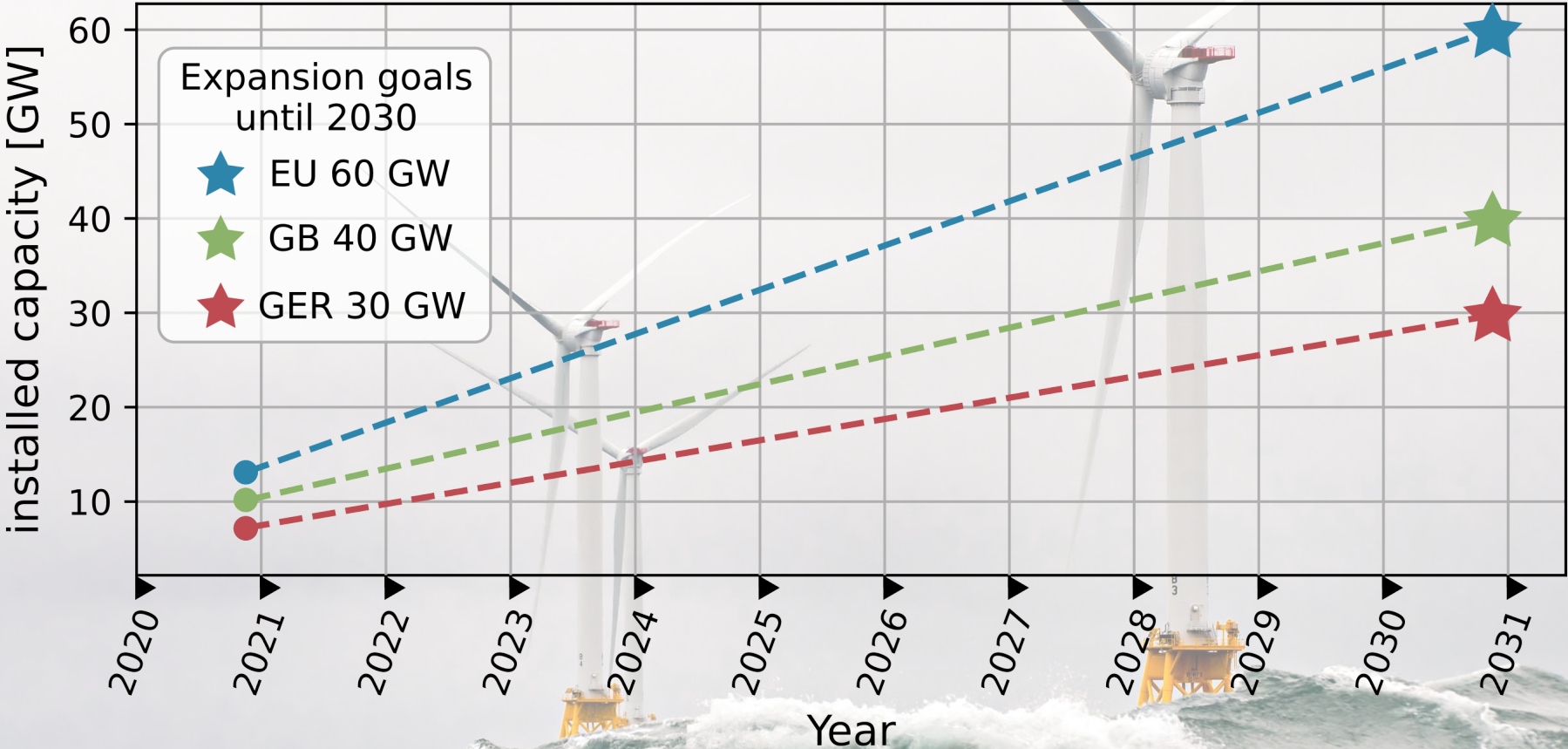
German Remote Sensing Data Center



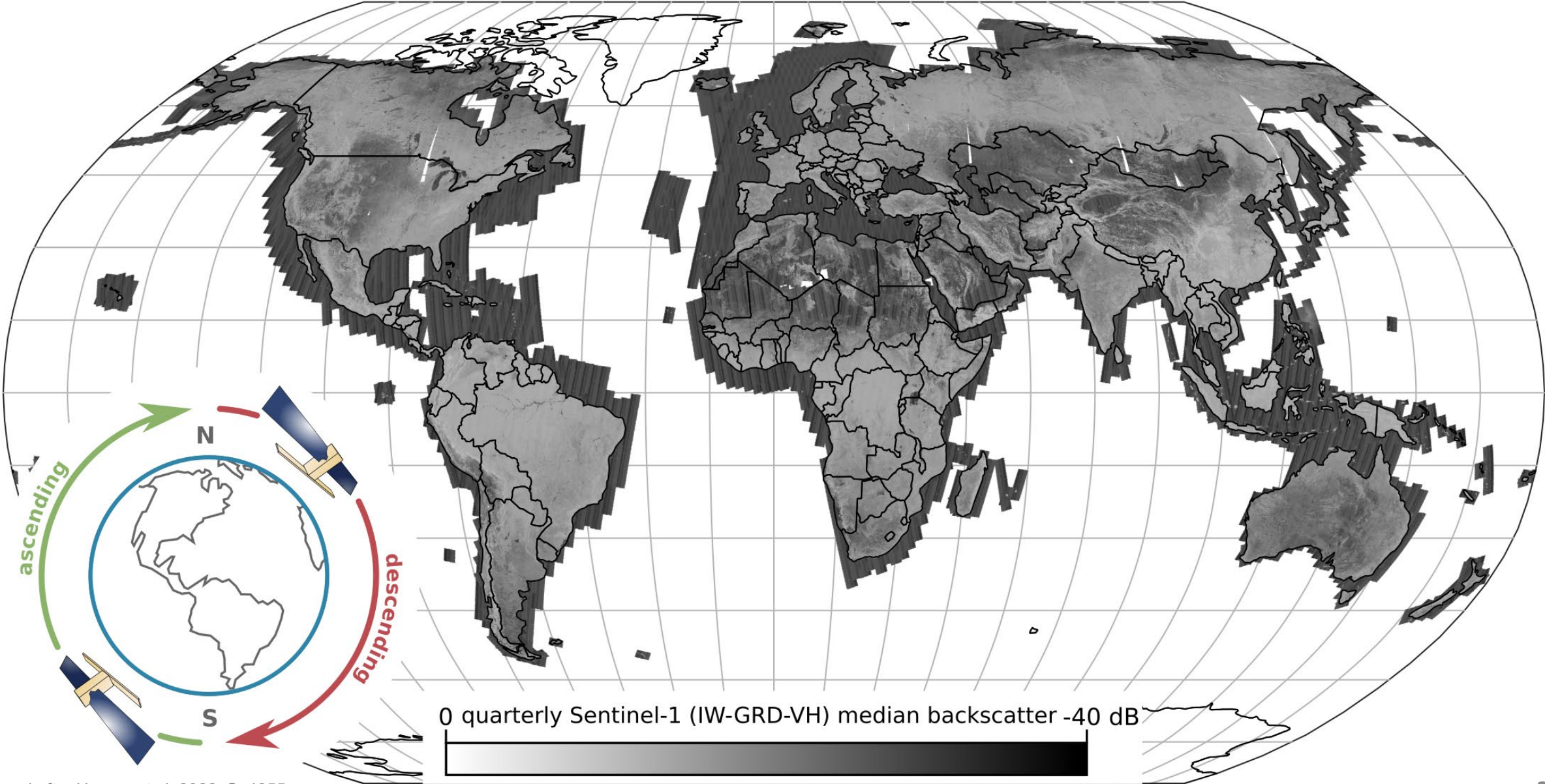
Knowledge for Tomorrow



Offshore Wind Energy Expansion

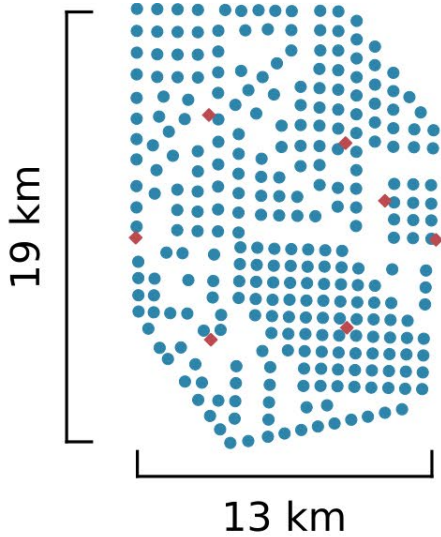


Sentinel-1 Median Composite

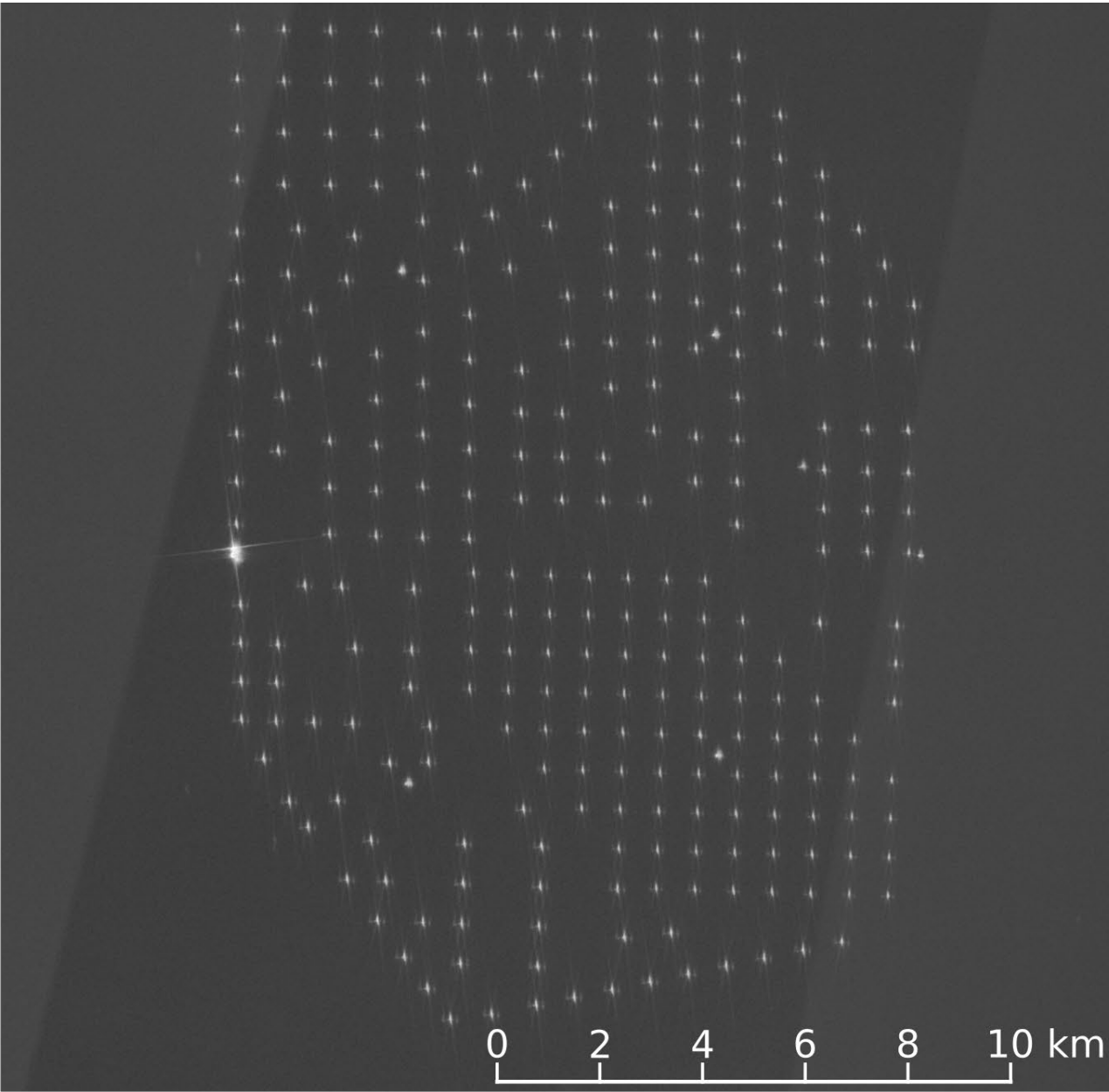


Changed after Hoerer et al. 2022, S. 4255

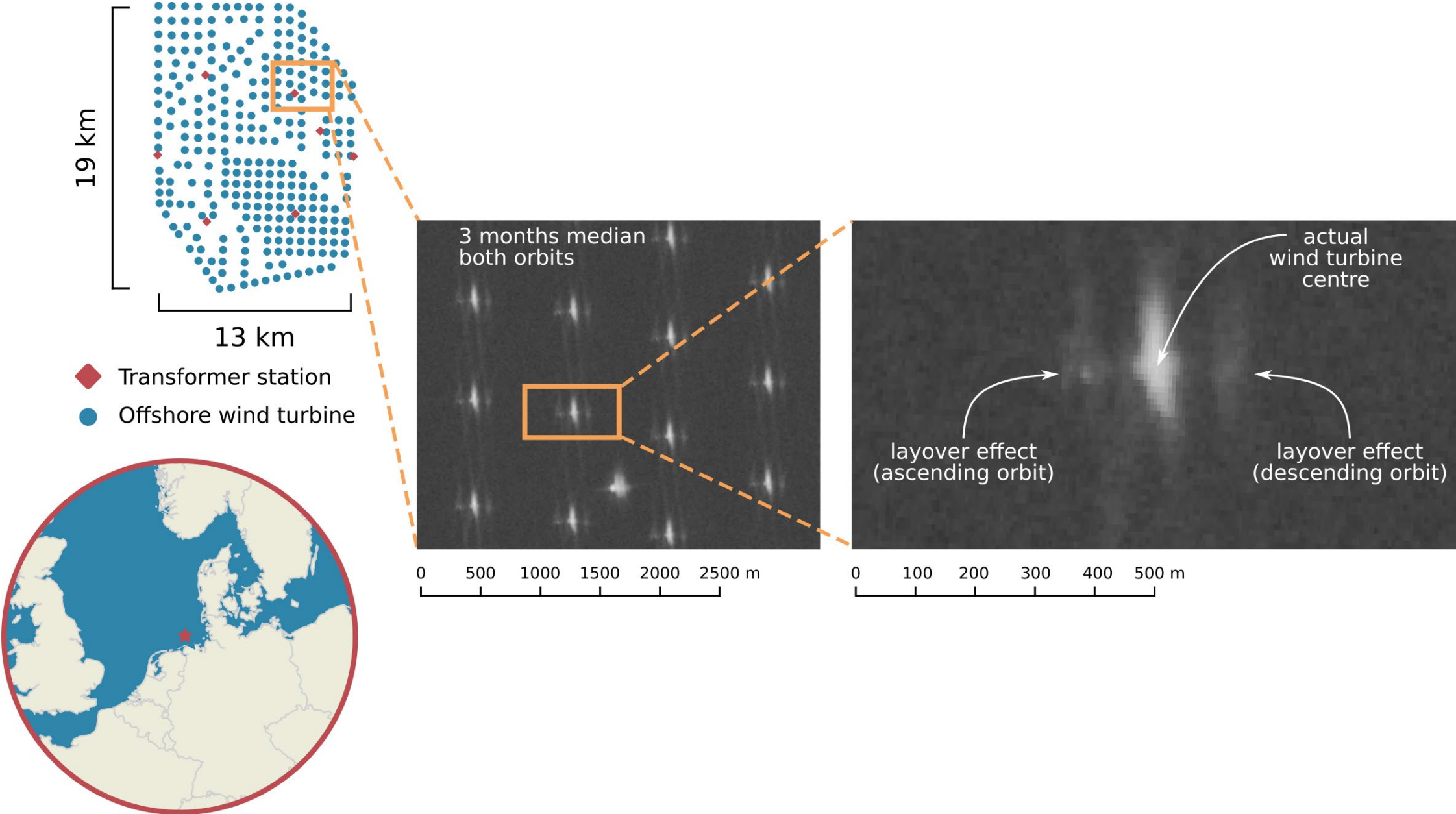
Offshore Wind Energy Infrastructure in Sentinel-1 imagery



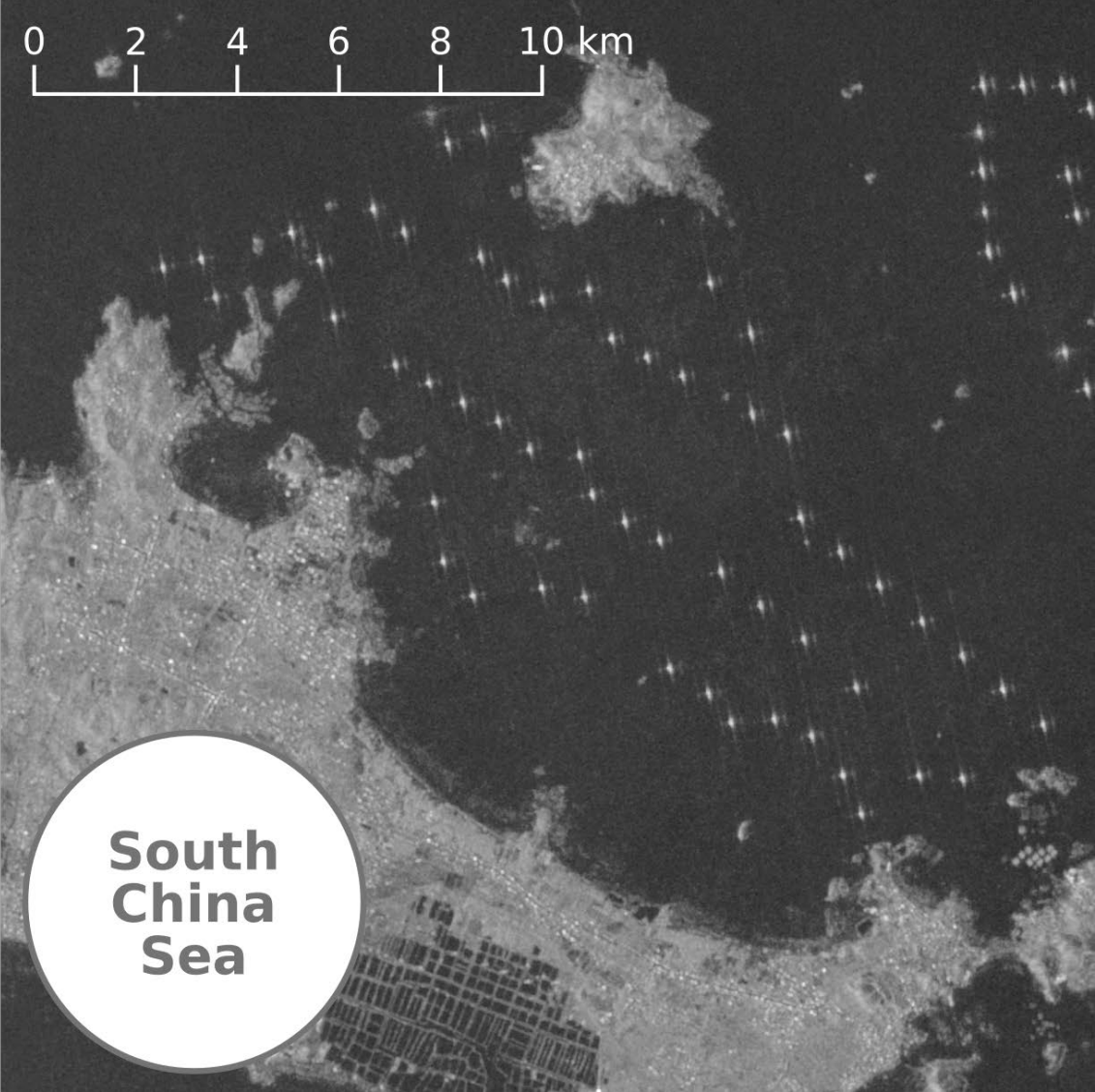
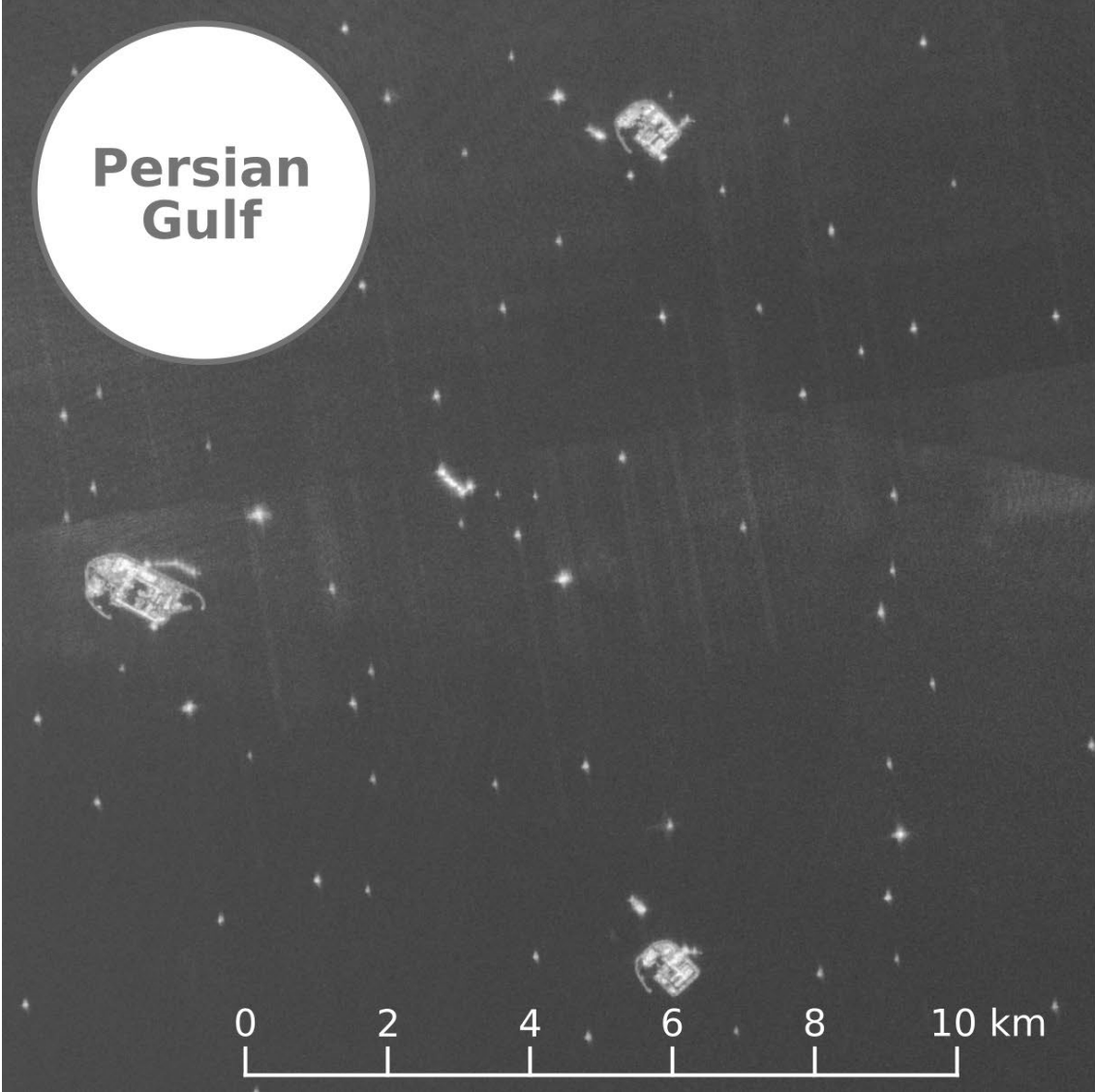
- ◆ Transformer station
- Offshore wind turbine



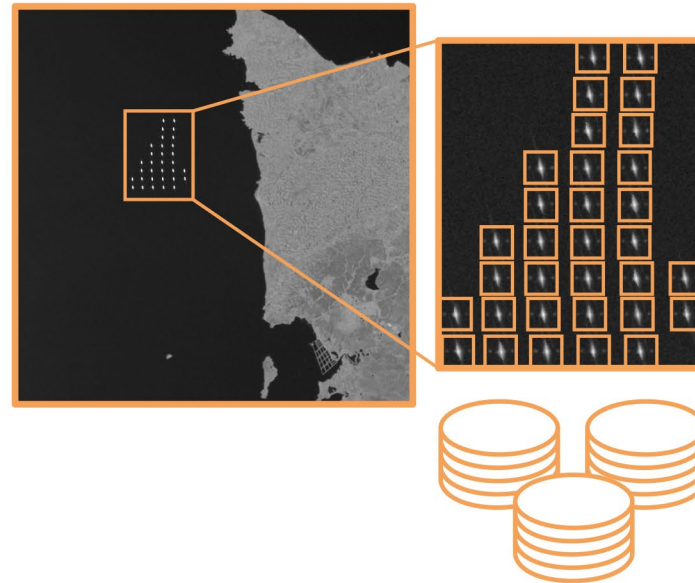
Offshore Wind Energy Infrastructure in Sentinel-1 imagery



Offshore Wind Energy Infrastructure in Sentinel-1 imagery



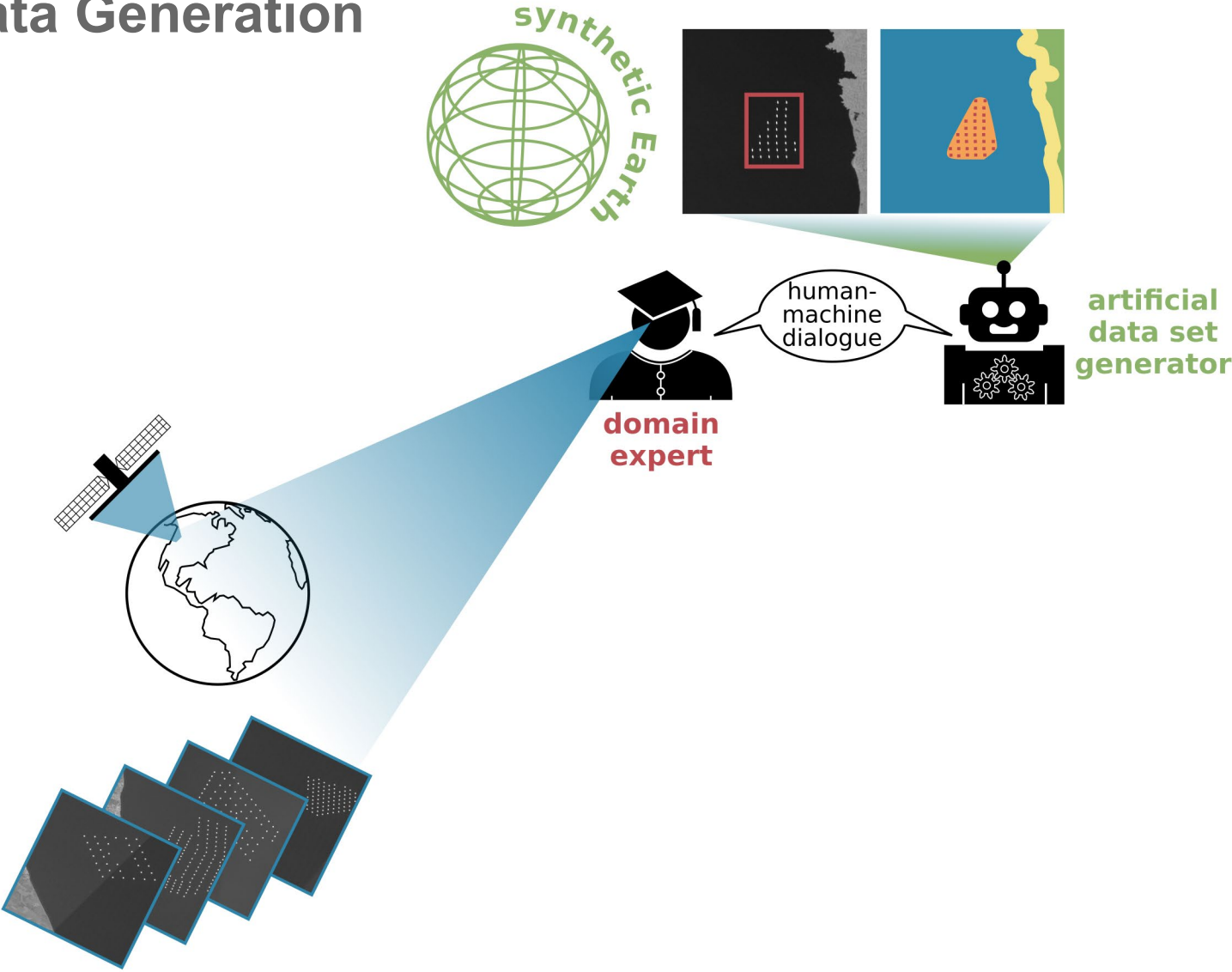
Supervised Deep Learning for Offshore Wind Energy Infrastructure detection



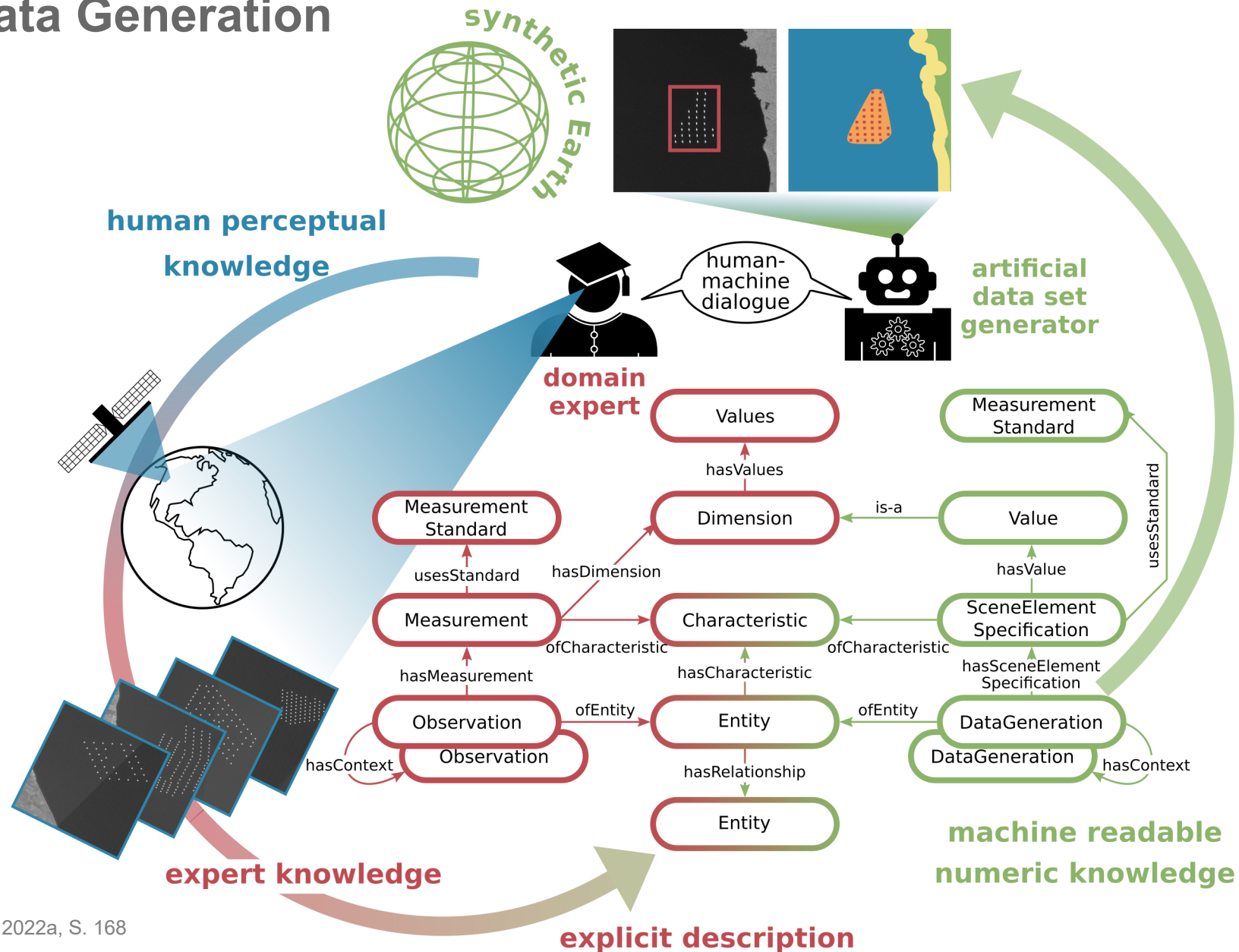
"With enough computation and enough data, learning beats programming for complicated tasks that require the integration of many different, noisy cues."

Krizhevsky, Sutskever and Hinton (2017, S. 84)

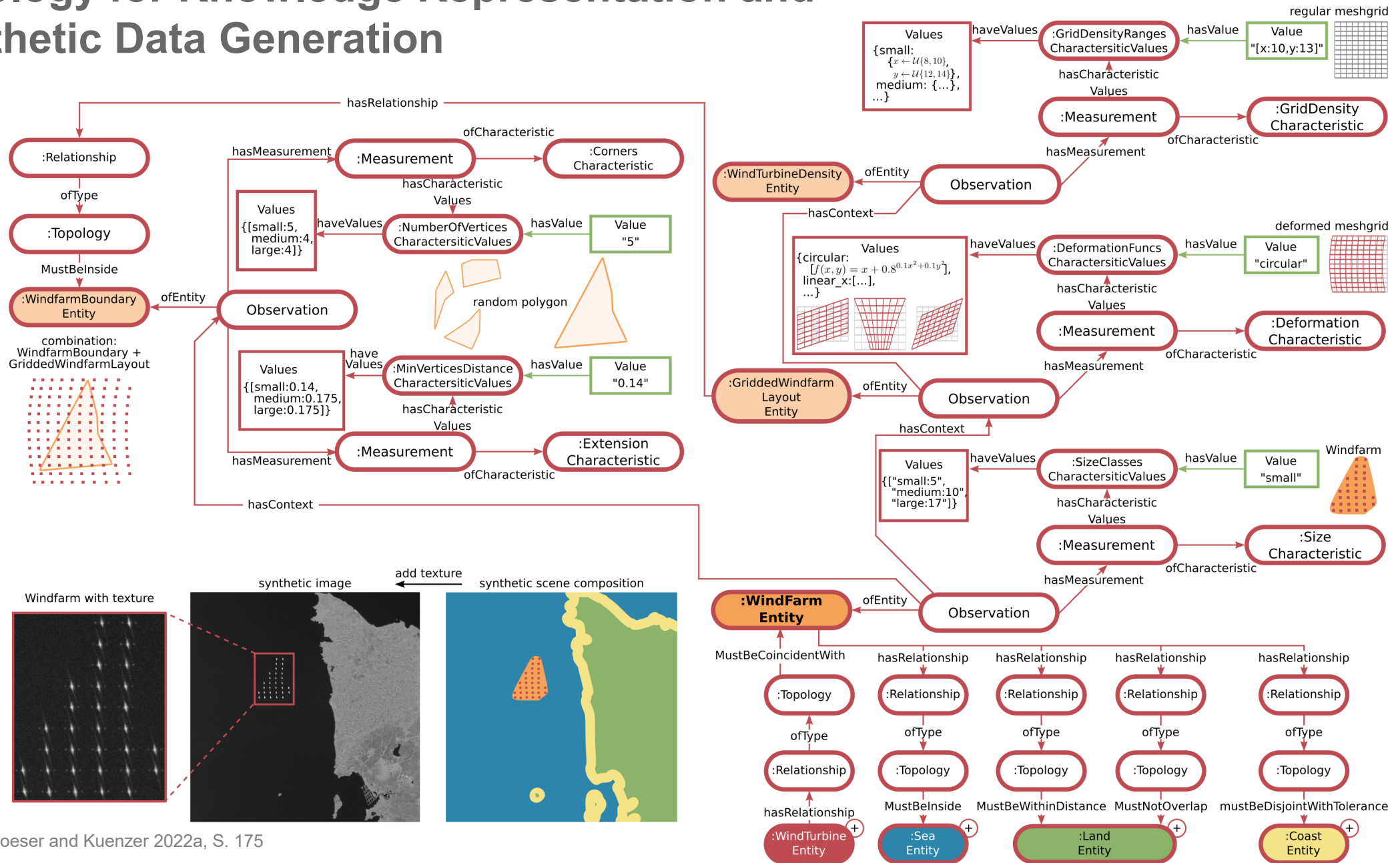
Ontology for Knowledge Representation and Synthetic Data Generation



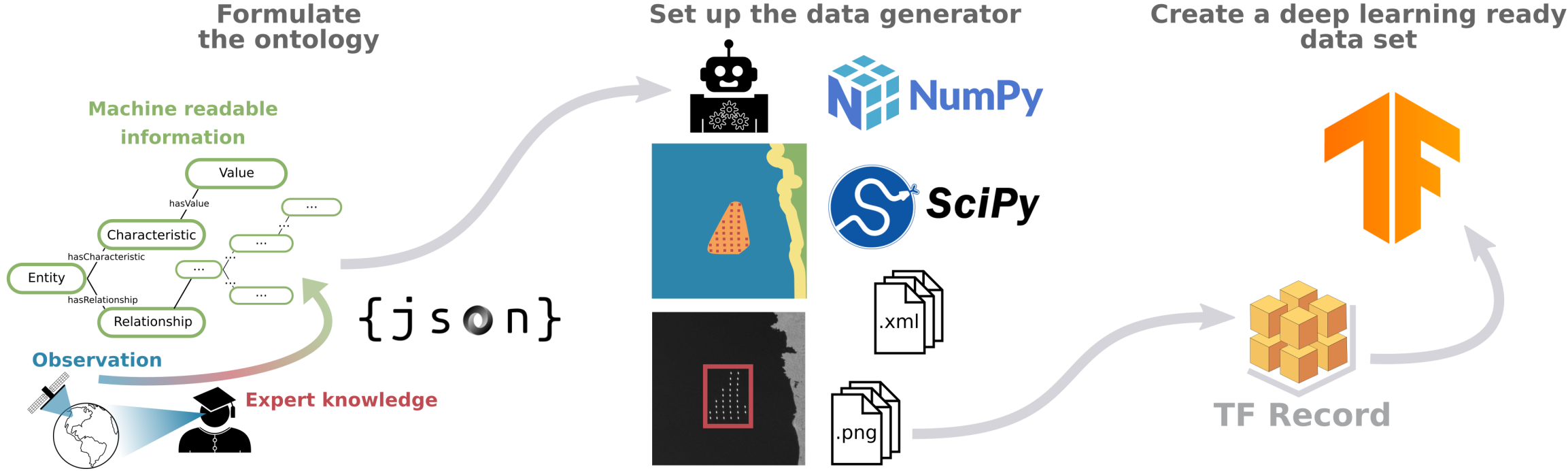
Ontology for Knowledge Representation and Synthetic Data Generation



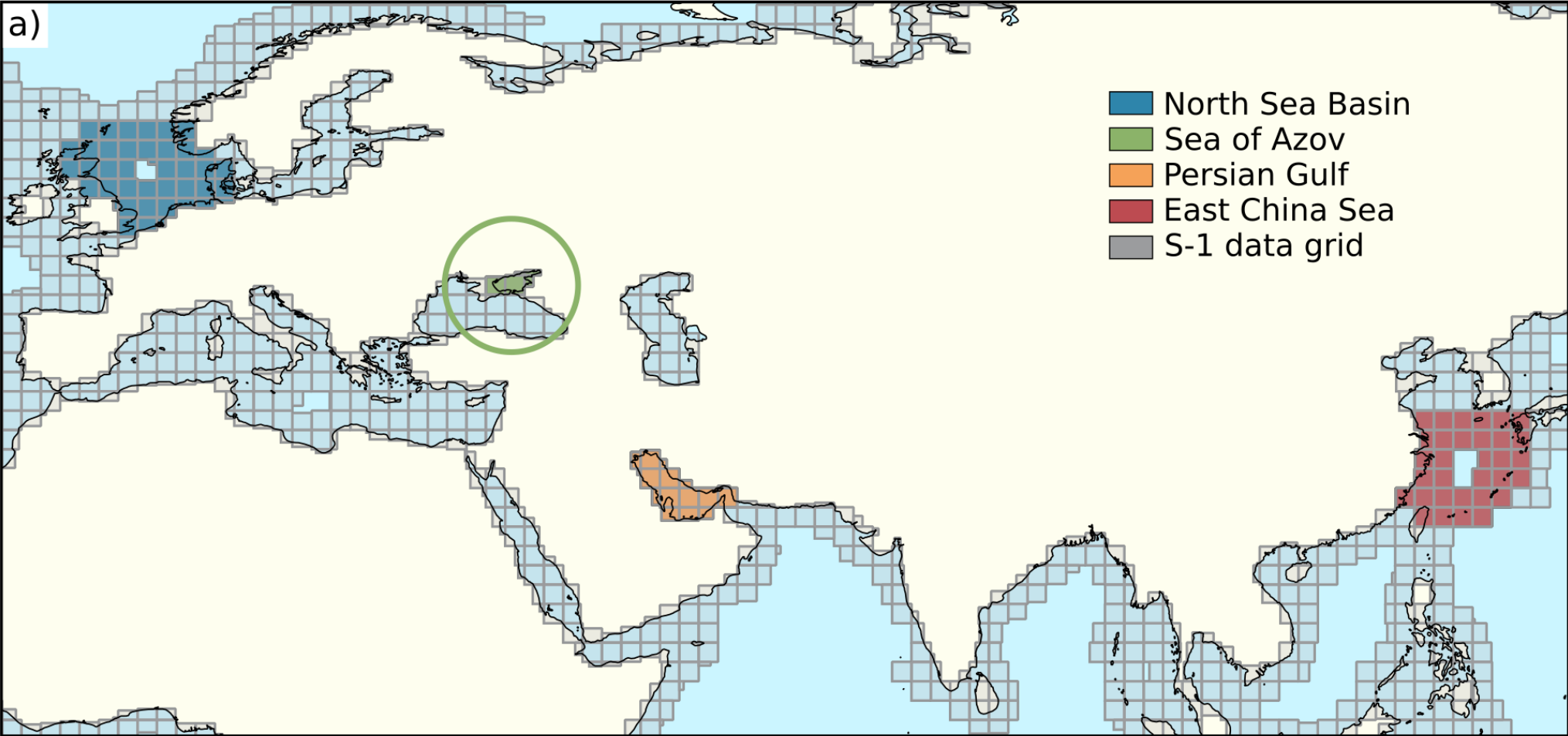
Ontology for Knowledge Representation and Synthetic Data Generation



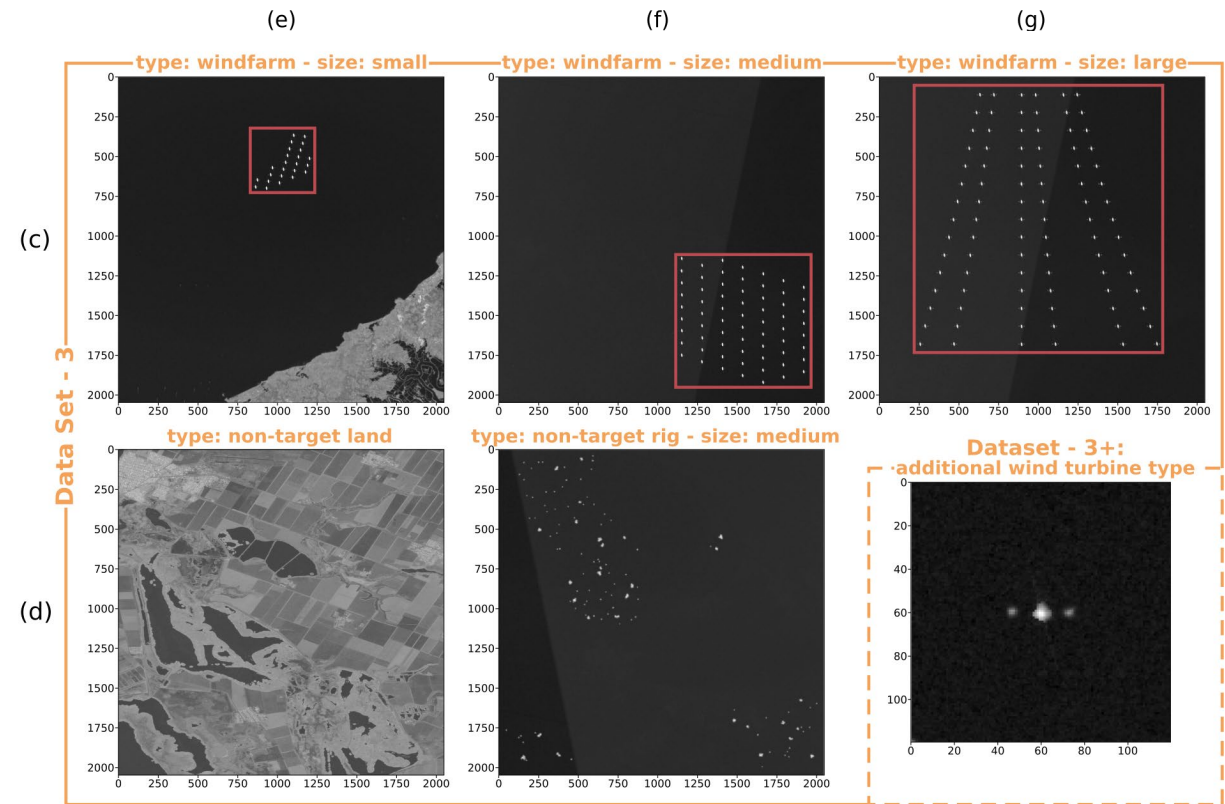
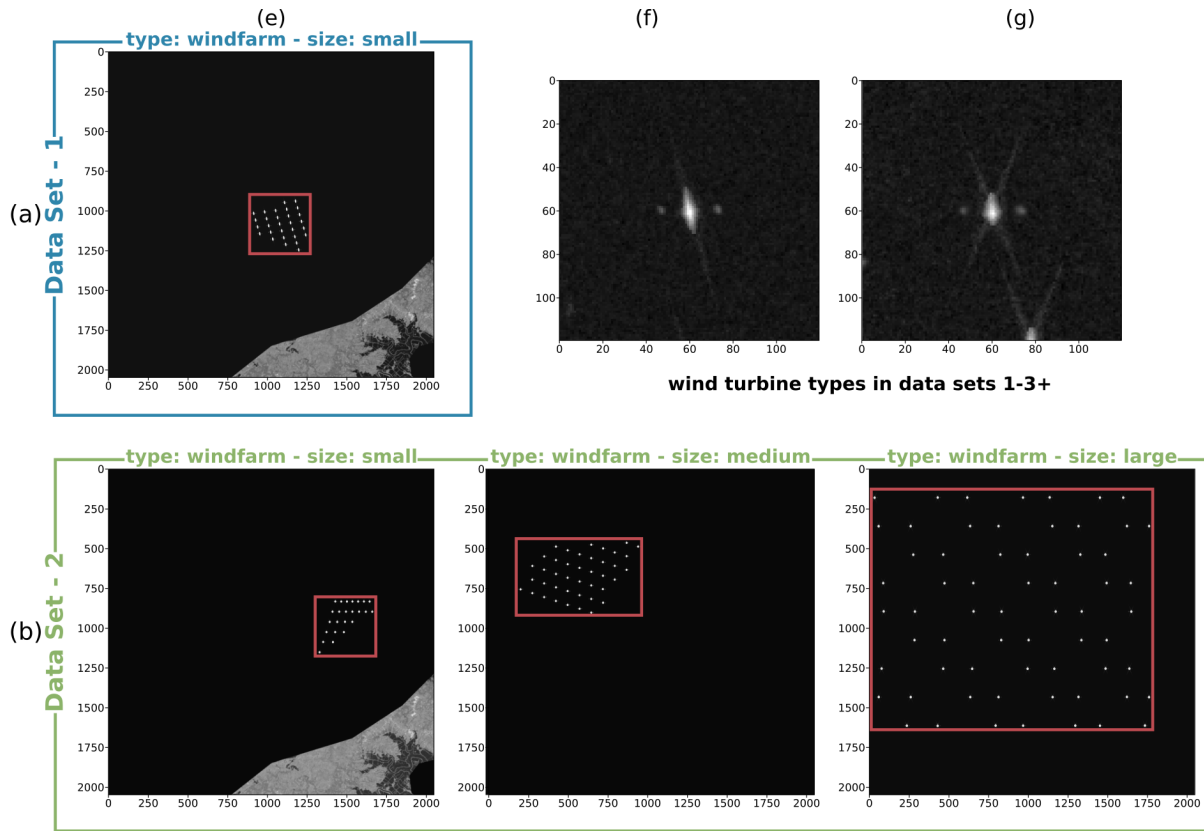
Ontology for Knowledge Representation and Synthetic Data Generation



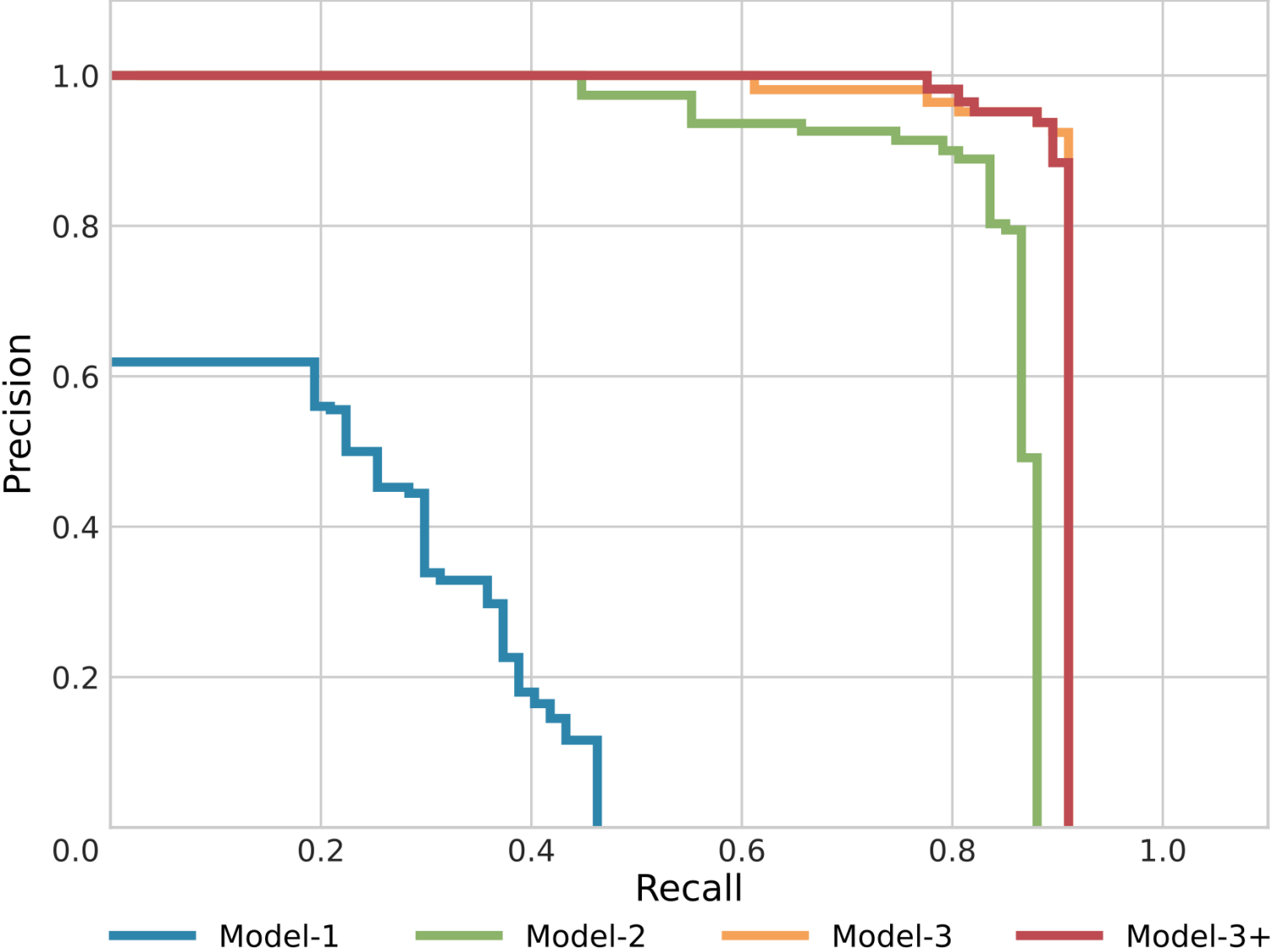
Real World Test Sites



Synthetic Training Data Sets

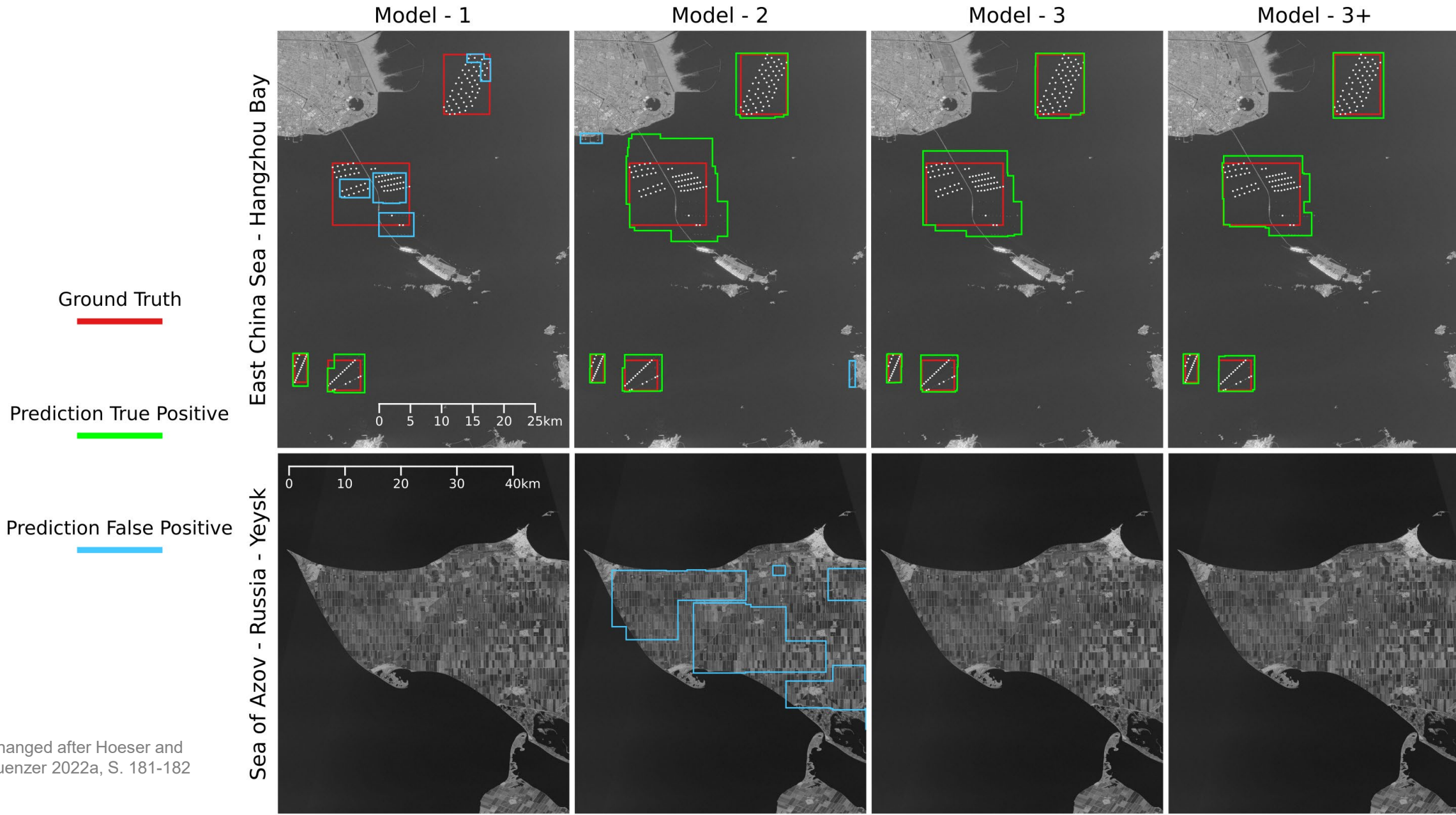


ResNet-50 Faster R-CNN Detection performances



Changed after Hoeser and Kuenzer 2022a, S. 179

Detection Results – East China Sea and Sea of Azov



Changed after Hoerer and Kuenzer 2022a, S. 181-182

Detection Results – East China Sea and Persian Gulf

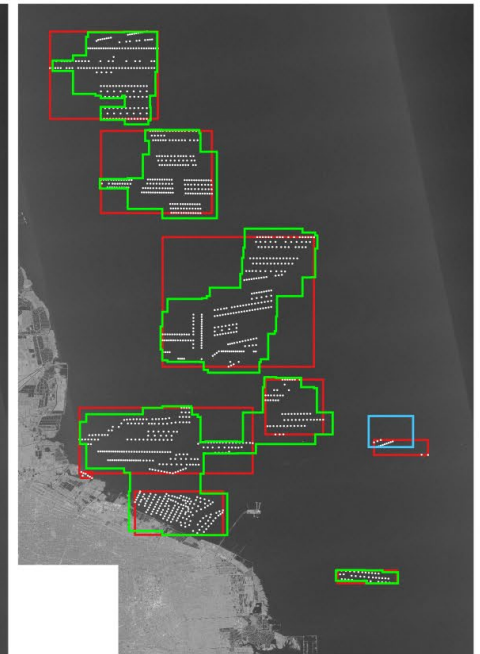
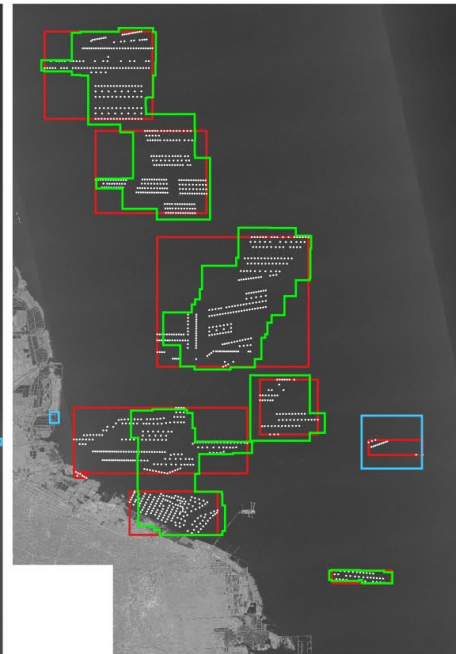
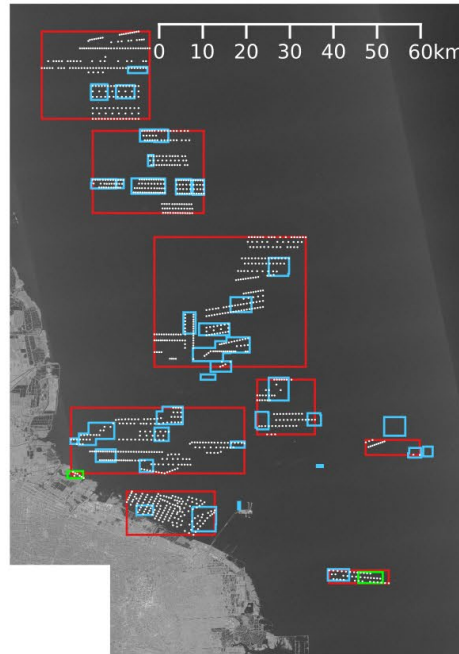
Model - 1

Model - 2

Model - 3

Model - 3+

East China Sea - jiangsu



Ground Truth



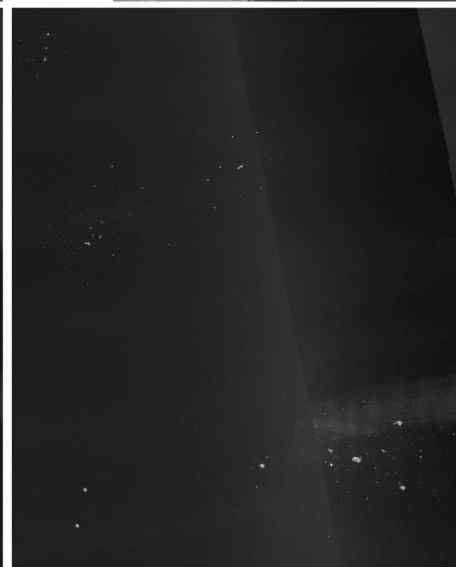
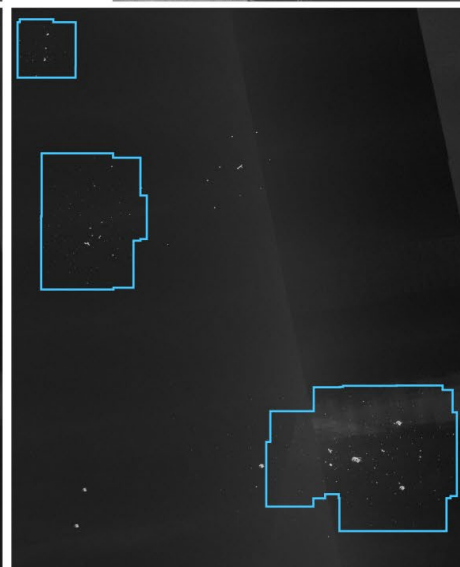
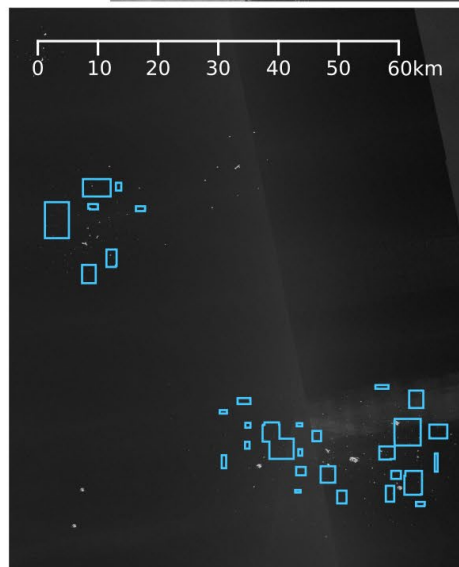
Prediction True Positive



Prediction False Positive



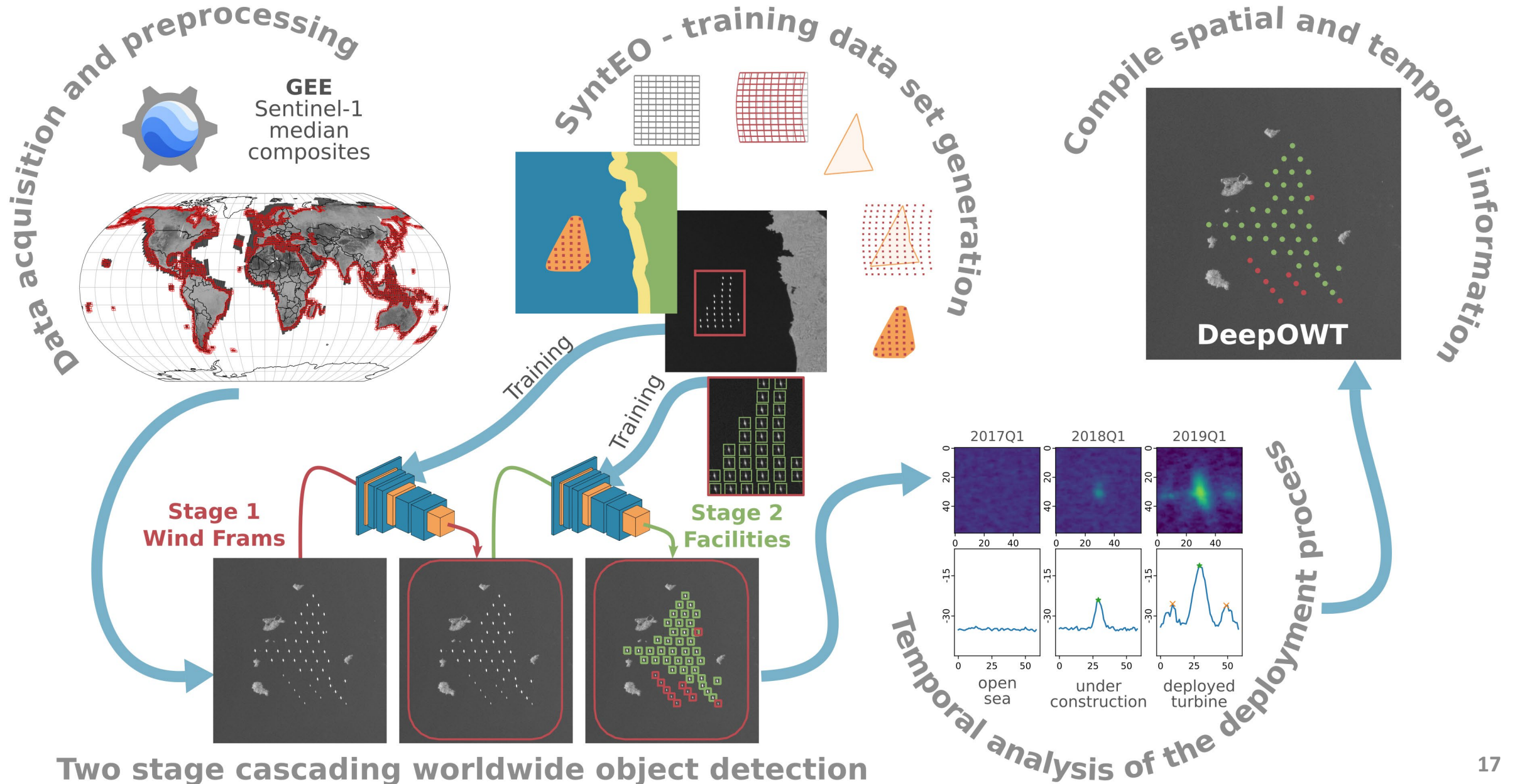
Persian Gulf - U.A.E near Abu Dhabi



Changed after Hoerer and Kuenzer 2022a, S. 181-182

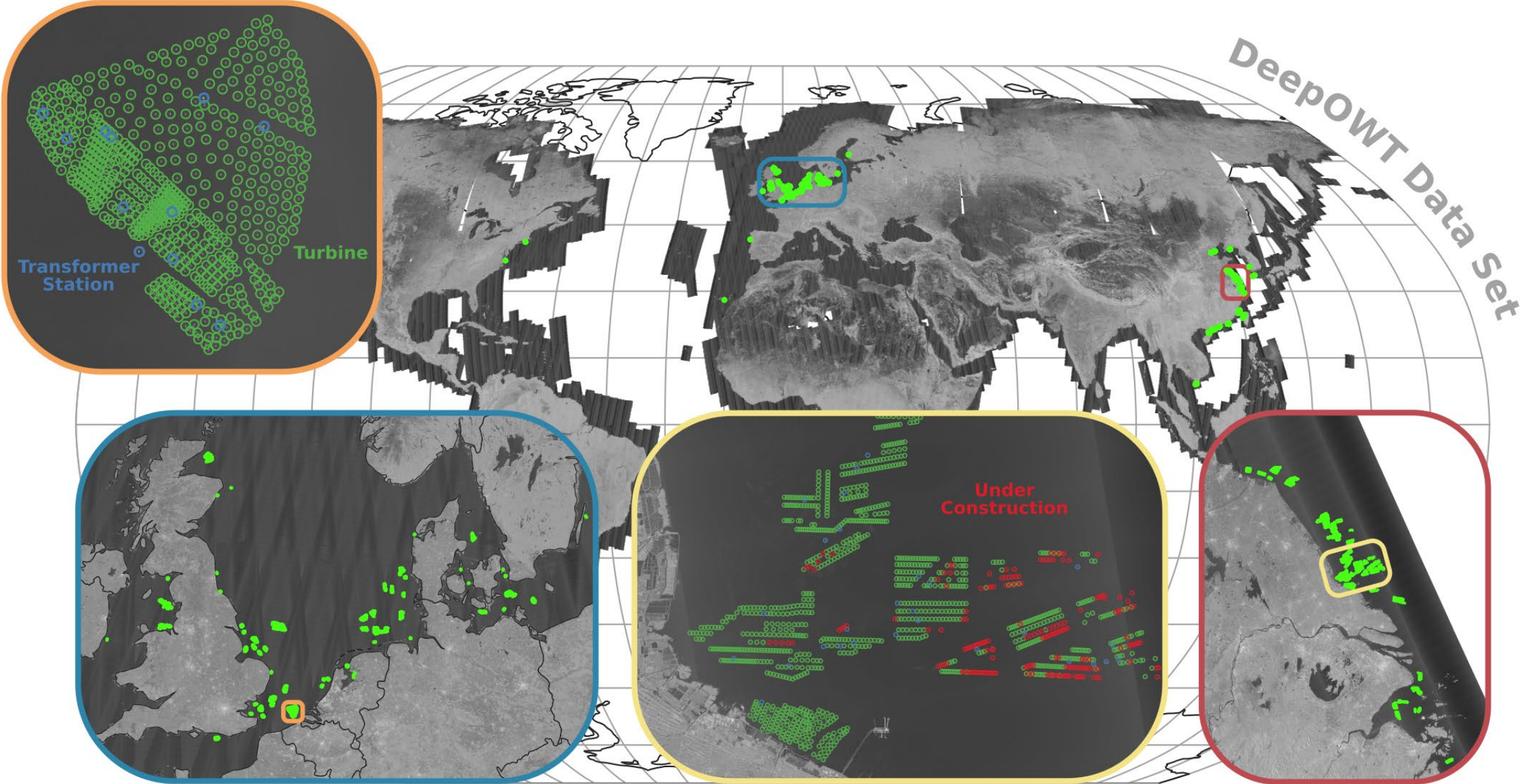
Detection Results – East China Sea and Persian Gulf

Changed after Hoerer et al. 2022, S. 4265

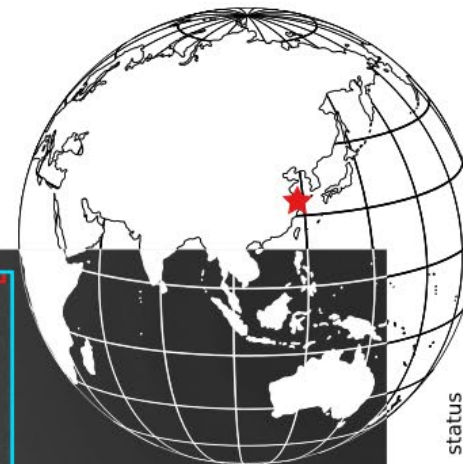


Two stage cascading worldwide object detection

Global offshore wind turbine data set derived with deep learning - DeepOWT



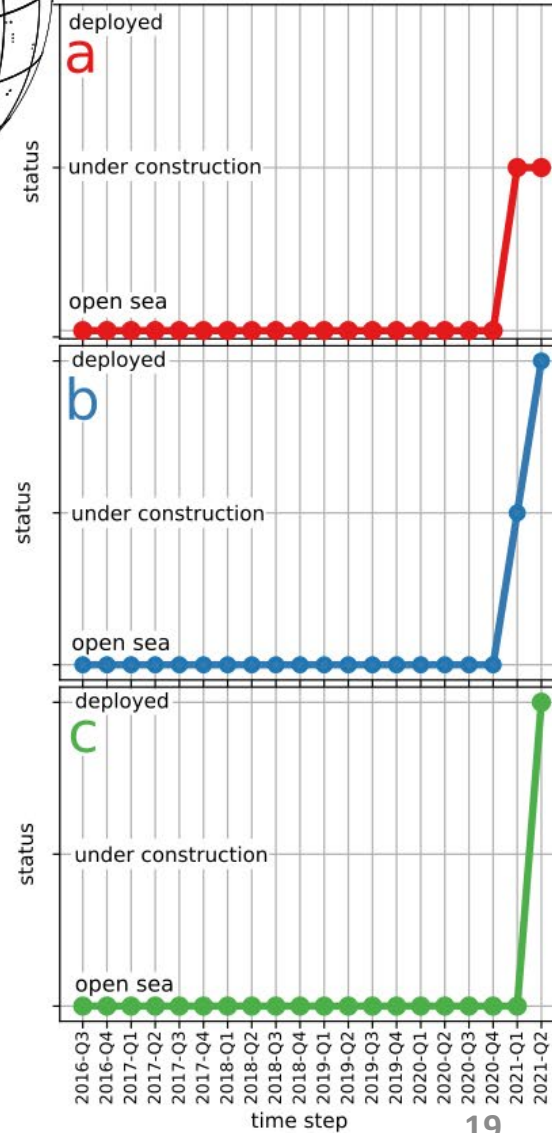
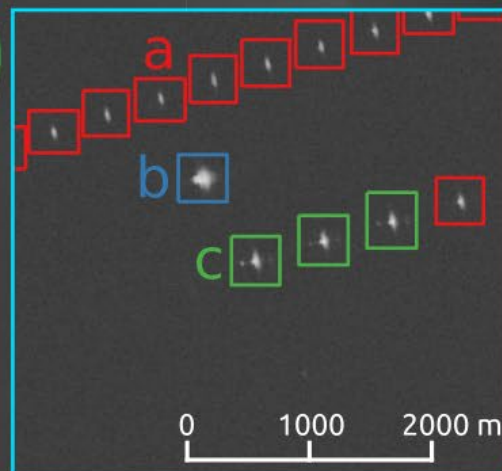
Offshore wind turbine deployment - China



status monitor 2016-Q3 - 2021-Q2

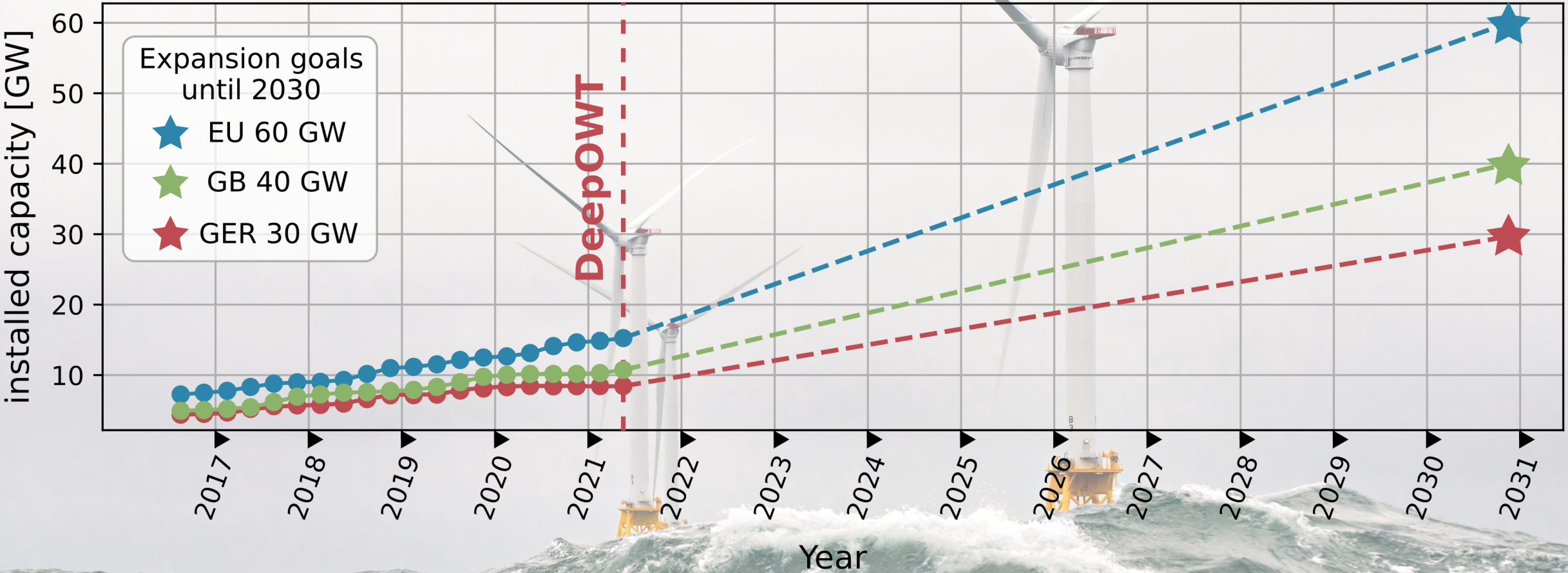
Jiangsu - China - 2021-Q2

- deployed substation
- deployed OWT
- under construction
- wind farm RoI



0 10 20 km

Offshore Wind Energy Expansion



References

Hoeser, T., Feuerstein, S., Kuenzer, C., 2022. DeepOWT: A global offshore wind turbine data set derived with deep learning from Sentinel-1 data. *Earth System Science Data* 14, 4251-4270. doi: <https://doi.org/10.5194/essd-14-4251-2022>.

Hoeser, T., Kuenzer, C., 2022a. SyntEO: Synthetic Dataset Generation for Earth Observation and Deep Learning - Demonstrated for Offshore Wind Farm Detection. *ISPRS Journal of Photogrammetry and Remote Sensing* 189, 163-184. doi: <https://doi.org/10.1016/j.isprsjprs.2022.04.029>.

Hoeser, T., Kuenzer, C., 2022b. Global dynamics of the offshore wind energy sector monitored with Sentinel-1: Turbine count, installed capacity and site specifications. *International Journal of Applied Earth Observation and Geoinformation* 112, 102957. doi: <https://doi.org/10.1016/j.jag.2022.102957>.

Krizhevsky, A., Sutskever, I., Hinton, G.E., 2017. ImageNet Classification with Deep Convolutional Neural Networks. *Communications of the ACM* 60, 84–90. doi: <https://doi.org/10.1145/3065386>.



Thank you for your attention

Contact: felix.bachofer@dlr.de claudia.kuenzer@dlr.de

