

Historical Reconstruction and Future Projection of Land Surface Boundary Conditions

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Why dynamic LSBC?

Land Surface Boundary Conditions (LSBC): Land Use (LU), Land Cover (LC), and Leaf Area Index (LAI), define land-atmosphere interactions, regulating energy, water, and carbon exchanges, and playing a key role in the carbon cycle.

- Improved representation of land-atmosphere interactions ensures more realistic exchanges of water, energy, and carbon between land and atmosphere.
- Enhanced forecasts and projections improve near-surface weather and climate predictions for better risk management, adaptation, and mitigation.

Figure is generated by ChatGPT

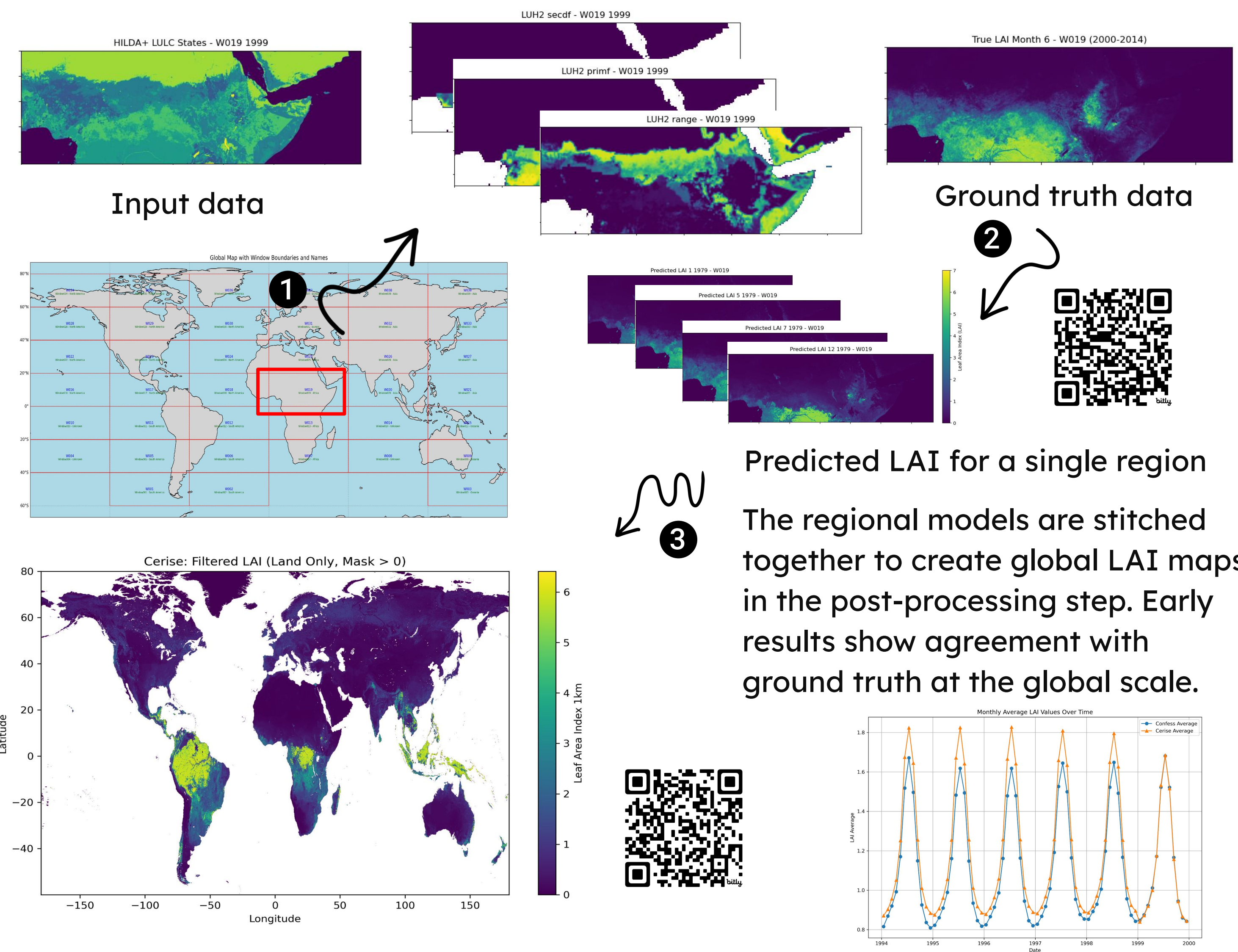
Datasets

- ESA CCI LC yearly [1992–2020] maps at 300 m resolution
- CONFESS LAI 5-days intervals [1980–2020] maps at 1 km resolution
- HILDA+ LU yearly [1899–2020] maps at 1 km resolution
- LUH2 LU yearly [950–2100] maps at 0.25° x 0.25° resolution
- Köppen-Geiger climate classification [present–future] maps at 1 km resolution
- Soil and Topography from PNNL Global at 1 km resolution
- LSM for ESM

LAI

Historical reconstruction of LAI

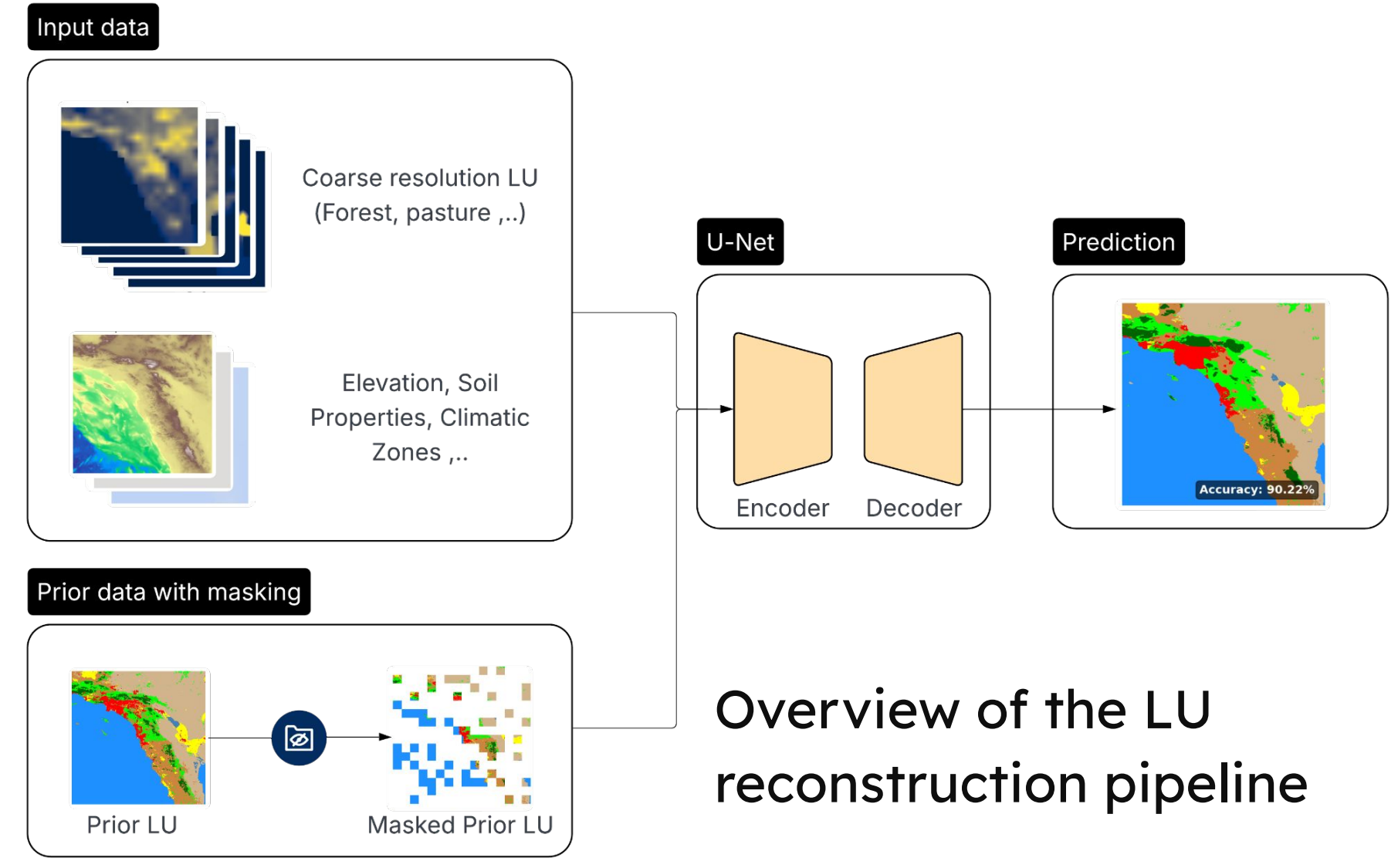
We developed a multi-scale approach for monthly LAI reconstruction at 1 km resolution using LUH2 and HILDA+ LU data. The global domain was partitioned into 39 independent regional tiles of 20° x 60°, with each model trained using Extreme Gradient Boosting on 14 years of data. The pipeline included re-gridding to target resolution, regional model training on GPU, and post-processing to generate seamless global outputs. Models achieved approximately 90% average misfit reduction.



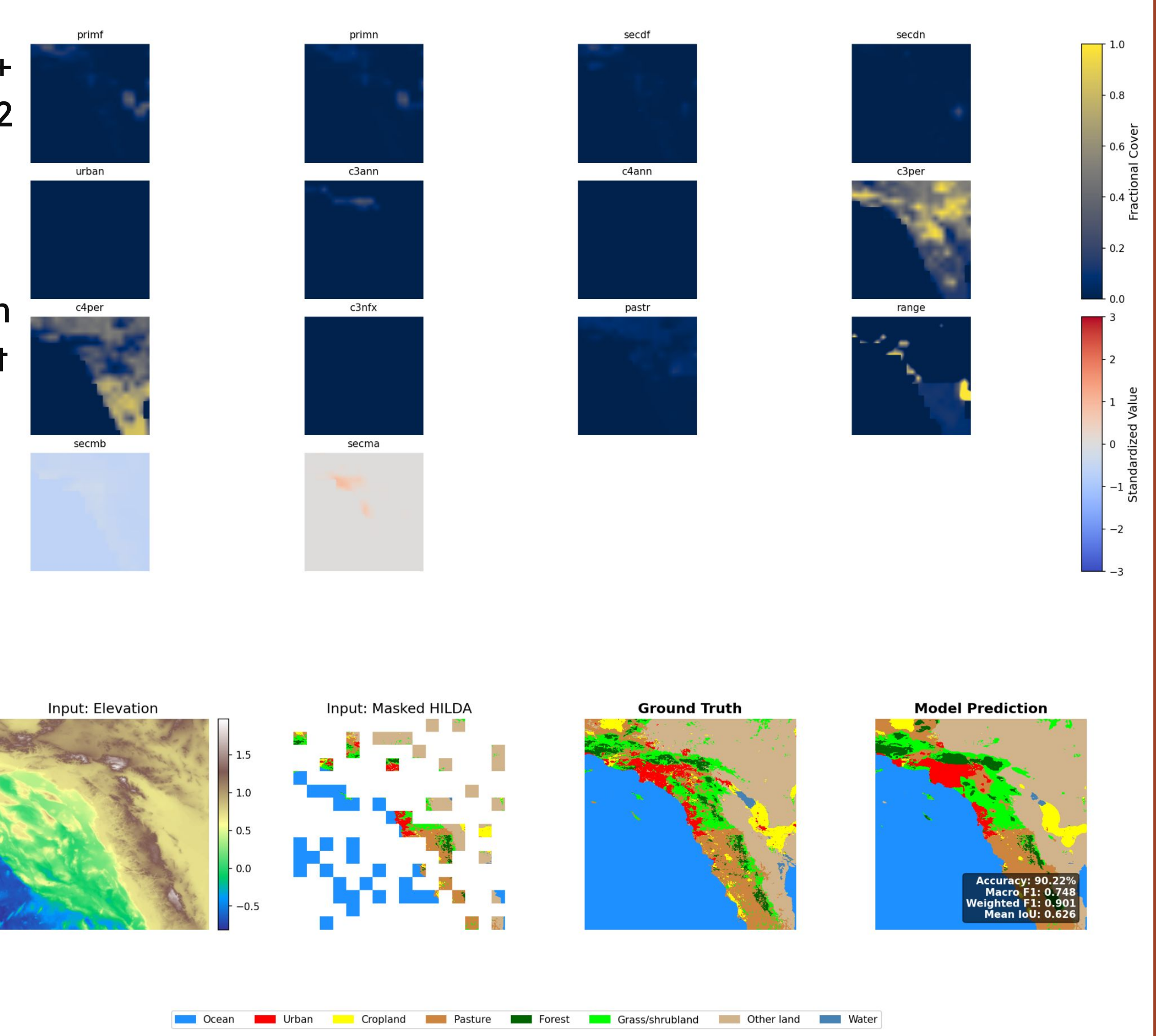
LU

Historical reconstruction & future projection of LU

We developed a U-Net architecture for LU reconstruction with a preprocessing pipeline aligning datasets to HILDA+ at 1 km resolution. Inputs included coarse-resolution LUH2 fractional LU variables, topographical data, and a partially masked prior LU map. The encoder-decoder U-Net with skip connections was trained on 30,000 samples, achieving 90% accuracy under 75% masking with mean Intersection over Union of 0.626. This reflects robust performance on dominant classes but confusion between similar vegetation types and difficulty detecting small features. Average mIoU of ~0.42 across test regions highlighted challenges with underrepresented LU types.



Prediction for San Diego (Year 2001) | Model Epoch: 3



Reconstruction of the LU map for the city of San Diego for the year 2001, using LUH2 fractional LU with 0.25° x 0.25° resolution, an elevation map, and a prior LU map randomly 75 percent masked with 32 by 32 pixel patches.

What is next?

- Expand LU reconstruction: Incorporating climatic zone and soil information as additional input channels to improve model accuracy.
- Develop LC prediction: Expanding to high-resolution LC reconstruction using ESA CCI dataset with climate variables from ERA5 data.
- Implement LAI modeling: Extending framework to predict monthly LAI maps, exploring sequential models and joint multi-task approaches.
- Adopt models as emulators: Integrating developed LU, LC, and LAI models as emulators within IFS and ICON weather-climate modeling frameworks.

