

LEAF AREA INDEX RETRIEVAL IN SHANXI PROVINCE OF CHINA USING SENTINEL-1 DATA

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Context

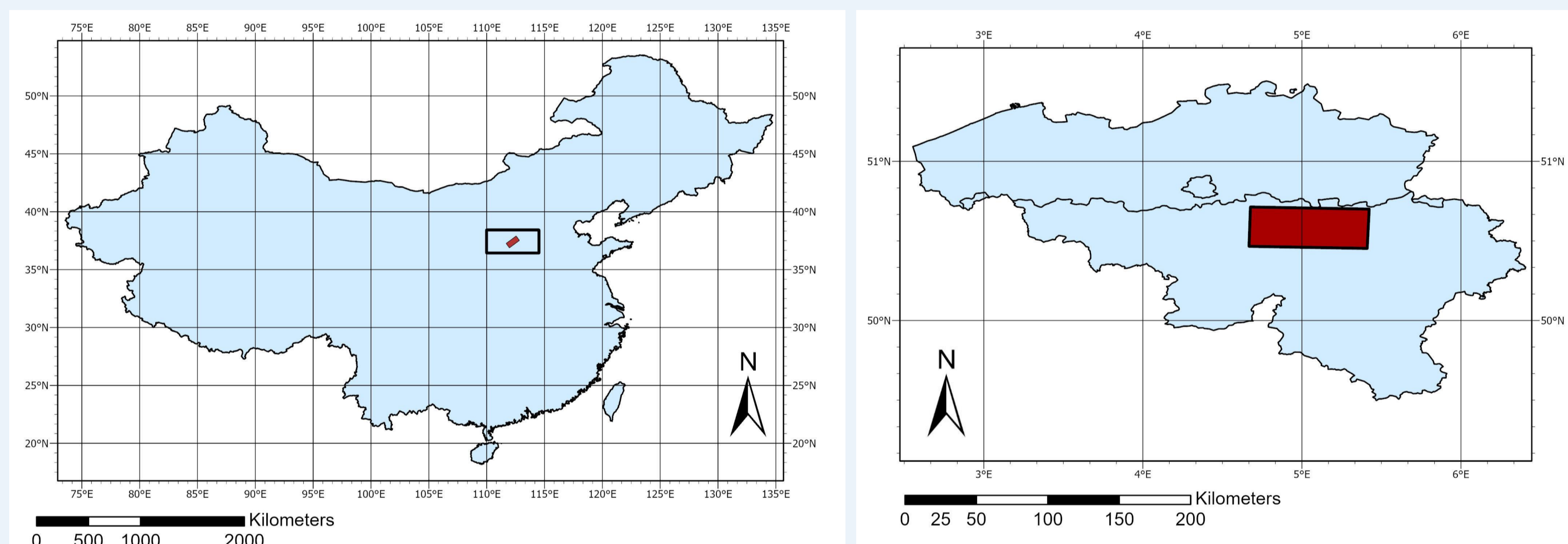
- Leaf area index (**LAI**) is a key variable for agricultural monitoring
- The most accurate LAI estimation methods exploit optical data (e.g., [1]) and are rendered ineffective in the case of frequent cloud cover
- Synthetic aperture radar (**SAR**) can allow the remote estimation of LAI at the parcel-level, on a large scale, regardless of cloud cover [2]
- SAR-to-optical regression methods have shown promising results for NDVI monitoring [3]

Objectives

1. Development of a reliable SAR-to-optical **LAI estimation** method for **maize** based on recurrent neural network and dual-pol **SAR** data
2. Assessment of its temporal and spatial (Belgium and China) **transferability**

Data

- Two geographically distinct regions: Shanxi province of **China** and Hesbaye region of **Belgium**



- **Field-average** values

Training & testing

- SAR backscatter data (σ_{VH}^0 , σ_{VV}^0 , θ) from **Sentinel-1**, from 2019 to 2023
- **LAI** derived from **Sentinel-2** optical imagery [1], from 2019 to 2023

External validation

- **LAI** measured **in situ** in 5 sites in the Shanxi province of China in 2023 (ongoing)



Methods

Model

- **Recurrent neural network**
- Time series (May to mid-October) of field-average values

Features (Sentinel-1)

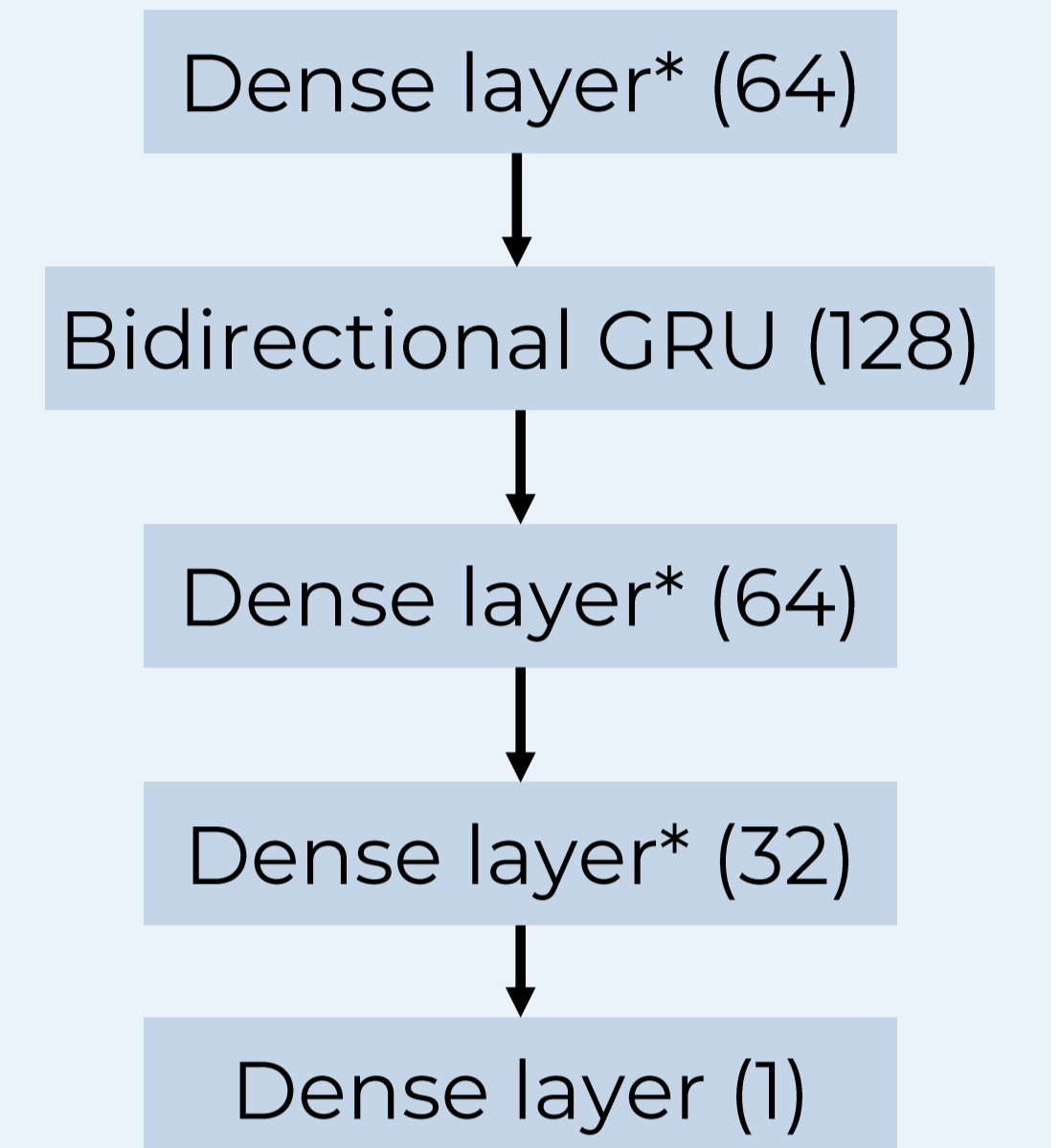
- Mean, std, min, and max for VV and VH
- VV/VH ratio
- Incidence angle
- Dual Pol Radar Vegetation Index (dpRVI)

Labels (Sentinel-2)

- LAI derived from Sentinel-2
- Interpolated with a nearest neighbor algorithm on Sentinel-1 DSC acquisition grid
- Label masked if > 5 days apart from nearest Sentinel-1 acquisition

Training and validation

- 10-to-1 train/test split
- External validation against LAI measured in situ with Licor LAI-2000

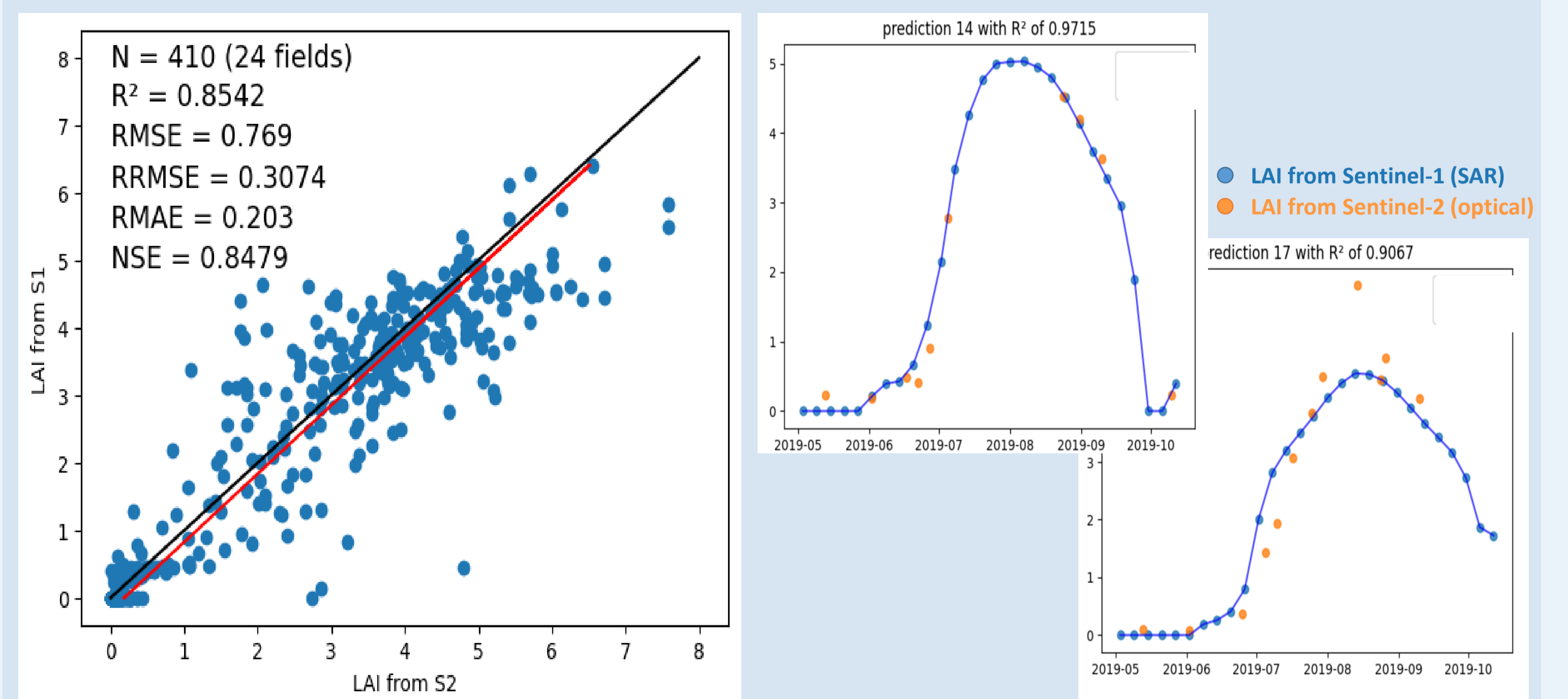


*batch normalizations, ReLU activations, and dropouts

Results

- **Field campaign** in Shanxi still **ongoing**
- **Preliminary study** conducted on 231 maize fields in **Belgium** with Sentinel-1 and -2 images from **2019**

LAI estimation results for Hesbaye region, Belgium, in 2019



References

- (1) Delloye, C., Weiss, M., & Defourny, P. (2018). Retrieval of the canopy chlorophyll content from Sentinel-2 spectral bands to estimate nitrogen uptake in intensive winter wheat cropping systems. *Remote Sensing of Environment*, 216, 245-261.
- (2) Bouchat, J., Tronquo, E., Orban, A., Neyt, X., Verhoest, N. E., & Defourny, P. (2022). Green area index and soil moisture retrieval in maize fields using multi-polarized C- and L-Band SAR data and the water cloud model. *Remote Sensing*, 14(10), 2496.
- (3) Garioud, A., Valero, S., Giordano, S., & Mallet, C. (2021). Recurrent-based regression of Sentinel time series for continuous vegetation monitoring. *Remote Sensing of Environment*, 263, 112419.