

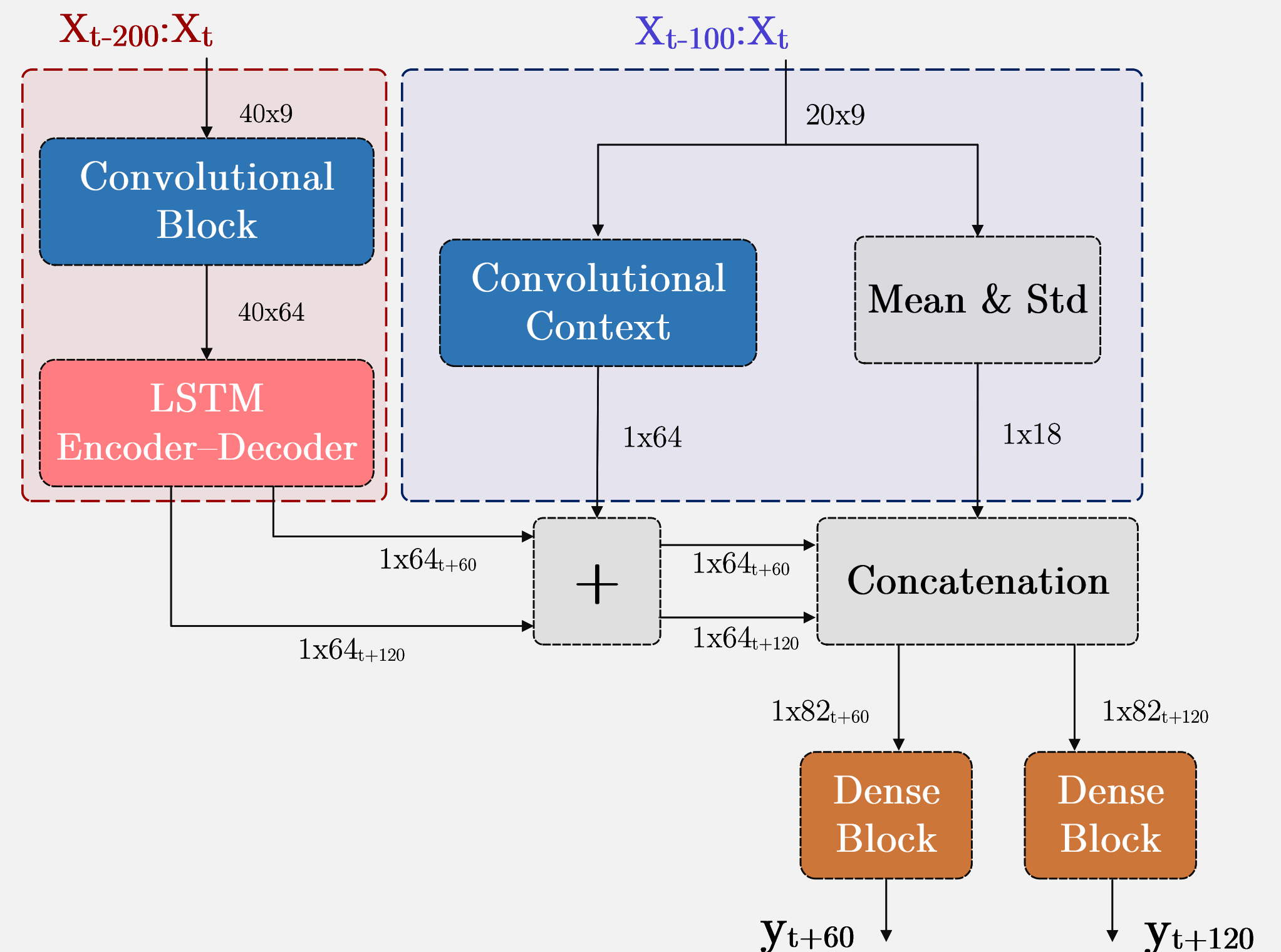
1. Work description

- DNN combining Convolutional and LSTM layers to forecast the SYM-H and ASY-H geomagnetic indices.
- Two networks with the same architecture, one for each index, for better accuracy.
- The ASY-H index is significantly harder to forecast.
- Full description: <https://doi.org/10.1029/2021sw002748>

2. Database

- 42 intense geomagnetic storms between 1998 and 2018, SYM-H achieved values lower than -100 nT.
- IMF data measured by ACE (1-second resolution) and SYM-H and ASY-H indices values (5-min averages).
- IMF data grouped to the indices resolution computing mean and standard deviation.
- Plasma is not used due to large amount of missing data.
- Input features (last 200 minutes):
 - 5-min avg and std of the IMF strength.
 - 5-min avg and std of the X, Y, and Z GSM components of the IMF.
 - SYM-H and ASY-H indices.
- Output features:
 - Forecast of SYM-H and ASY-H for next 1 and 2 hours.

3. Neural Network Architecture

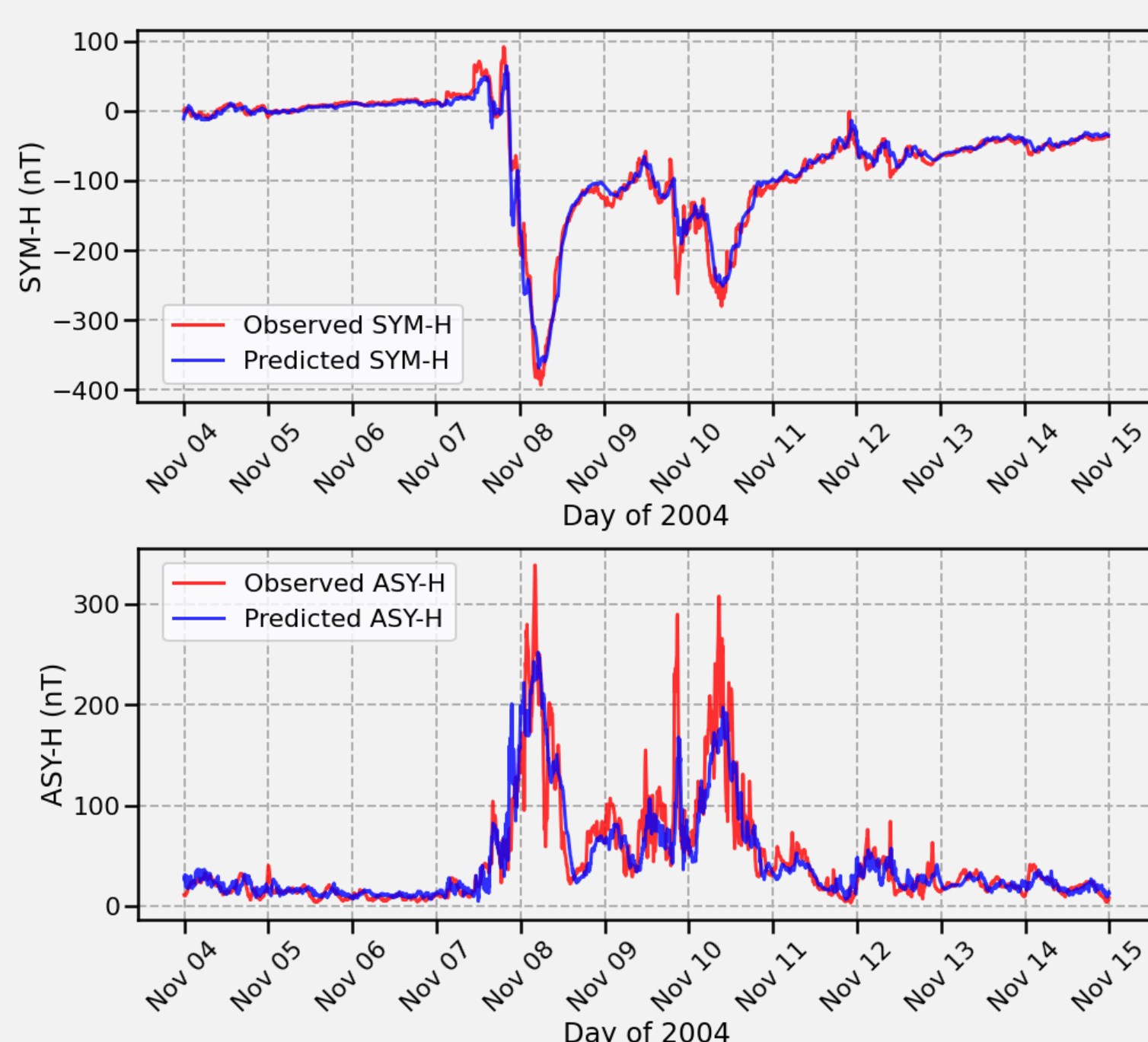


4. Metrics

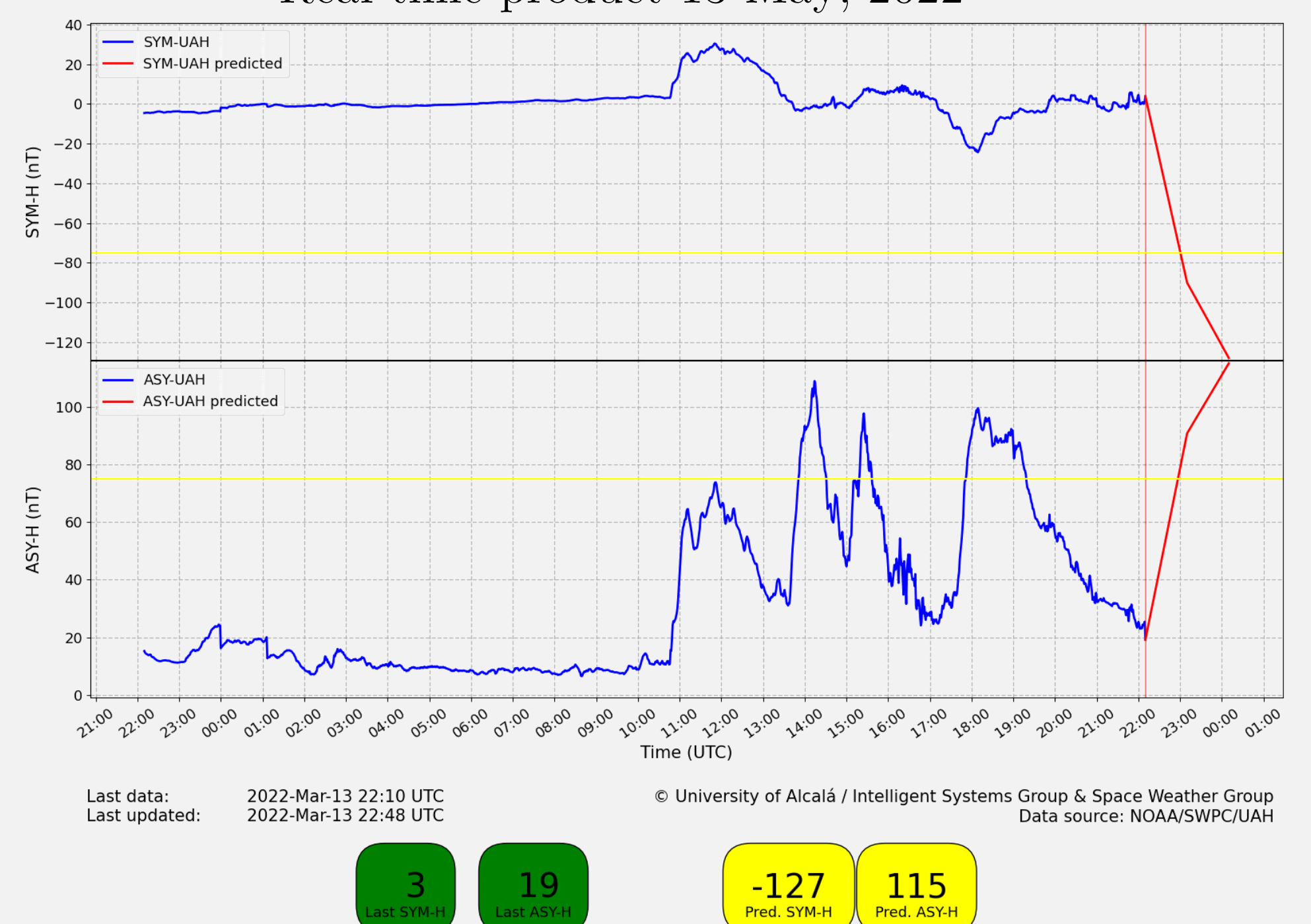
Dataset	Index	Mean RMSE	Mean R2
Test Storms	SYM-H 1h	7.861	0.930
	SYM-H 2h	10.527	0.875
	ASY-H 1h	13.300	0.727
	ASY-H 2h	15.213	0.647
Quiet Period 2013 - 2020	SYM-H 1h	4.717	0.888
	SYM-H 2h	6.208	0.805
	ASY-H 1h	8.459	0.545
	ASY-H 2h	9.786	0.389

5. Forecasting results

Storm of November, 2004 observed and prediction.



Real time product 13 May, 2022¹



Acknowledgements

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