T-FG-RS-

T-FORS 1st Innovation Day

LSTID Forecasting

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Outline

- LSTIDs occurrence chain of events
- LSTID forecasting based on ML: general strategy
- Developing the ML models
 - Features and labels
 - First approach: catalogue-based forecasting
 - Second approach: LSTID indices-based forecasting
- \circ $\,$ Remarks and way forward $\,$



Large-scale Travelling Ionospheric Disturbances





143 3

GNSS TEC observations.



LSTIDs occurrence chain of events from the auroral oval to middle latitudes



LSTID forecasting based on ML: general strategy





LSTID forecasting based on ML: general strategy





LSTID detection model development: General strategy





Developing the ML models: features & labels

Features

- IE (IL, IU) values
- GNSS TEC Gradients over Europe
- F10.7 value of the previous day
- current hour (ch) (cos(2pi/24 ch) + 1)/2
- current month (cm) (cos(2pi/12 cm) + 1)/2





Developing the ML models: features & labels

Features

- IE (IL, IU) values
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- current month (cm) (cos(2pi/12 cm) + 1)/2

Labels

- SPCont time series (based on MUF from single ionosondes)
- LSTID catalogue provided by Ebro Observatory based on HF-INT method
- HF-EU index (Activity index at European Level)



Developing the ML models: catalogue-based forecasting



- > The problem is handled as a binary classification
- HF-INT refined LSTID catalogue provided by Ebro Observatory
- In the catalogue, there are 760 TIDs events detected and recorded above Europe between FEB 2014 to DEC 2022
- The database is generated by leveraging a network of ionosondes covering the European sector

Parameter	Example
Start time	2022 01 11 21:00
Duration	2.0 hrs
Period	119.74 min
Amplitude	0.72 MHz
Velocity	597.47 m/s
Azimuth	202.39°
Detection technique	D2D

Catalogue example



Developing the ML models: catalogue-based forecasting

- The dataset is incomplete: (misdetections related to the technique used to create it)
- The class are severely unbalanced: 3% of Yes and 97% of No
- The input is shifted in space and time respect to the output
- Built ML dataset made of 157.777 couples (X(T),y(T))
 for each T every 30min between FEB 2014 and DEC
 2022, where:



$$\mathbf{X}(T) = \begin{bmatrix} X_1(T-6.5h) & X_1(T-6h) & X_1(T-5.5h) & X_1(T-5h) & \dots & X_1(T-1h) & X_1(T-0.5h) \\ X_2(T-6.5h) & X_2(T-6h) & X_2(T-5.5h) & X_2(T-5h) & \dots & X_2(T-1h) & X_2(T-0.5h) \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ X_N(T-6.5h) & X_N(T-6h) & X_N(T-5.5h) & X_N(T-5h) & \dots & X_N(T-1h) & X_N(T-0.5h) \end{bmatrix}$$

 $y(T) = \begin{cases} 1 & \text{TID detected in 3h starting from T,} \\ 0 & \text{else.} \end{cases}$



Developing the ML models: catalogue-based forecasting

- > We run different configurations of the models with different hyperparameters
- Best performances are obtained using neural networks
- Results are not satisfactory: we are not able yet to classify correctly the two classes based on external drivers
- > This suggests no clear correlation between the classes, given the features used





Developing the ML models: LSTID indices-based forecasting

- The LSTID forecasting is treated as a binary classification problem
- If *SPcont* > *Treshold* then an LSTID is detected (otherwise no LSTID is detected).
- *IL*, *IU*, *GNSS*, *LT* and *SPCont* values are considered as features
- Results (so far) only for Juliusruh ionosonde

3 Different models:

- **Prediction** of LSTID based **exclusively** on the **most recent** *SPcont* values.
- **Prediction** of LSTID based **exclusively** on the **most recent** *IL*, *IU*, *GNSS*, *LT* values.
- **Prediction** of LSTID based on **both** (a) the **most recent** *SPcont* values and (b) the **most recent** *IL*, *IU*, *GNSS*, *LT* values.

Classifiers employed:

- Feedforward Neural Network classifier FNN
- Block Recurrent Neural Network classifier RNN



Developing the ML models: LSTID indices-based forecasting

Model 1: Prediction of LSTID based **exclusively** on the **most recent** *SPcont* values.

Remarks:

- Trained FNN features very high accuracy results, even for higher time horizon forecasts.
- > In principle, as the time horizon increases, the **performance** degrades.



Model 2: Prediction of LSTID based exclusively on the most recent *IL*, *IU*, *GNSS*, *LT* values.

Remarks:

The results of the FNN classifier are less accurate than in the case where only the SPcont past values are considered (Model 2).



Developing the ML models: LSTID indices-based forecasting

Model 3: Prediction of LSTID based on both (a) the most recent SPcont values and (b) the most recent IL, IU, GNSS, LT values.

Remarks:

- The results of Model 3 are more accurate than in the case where only IL, IU, GNSS,
 LT past values are exclusively considered as inputs (Model 2).
- The results of Model 3 are less accurate than in the case where only the SP cont past values are considered (Model 1) but it can handle missing values



- If all the most recent SPcont values are available, Model 1 should be used.
- If none of the most recent SPcont values are available, Model 2 should be used.
- > If **some** of the most recent **SPcont** values are available, **Model 3** should be used.





Remarks and way forward

Catalogue-based forecasting:

- Add new input features
- Investigate other time-delays\input time window to be used for the catalogue-based model

Indices-based forecasting:

- > Application of the model to other stations (only Juliusruh was considered).
- Utilization of larger data sets (longer time periods).
- Intensive study of the data (e.g., the time periods where LSTIDs are encountered).
- > Dealing with the missing data issue (e.g., the cases where the **SPcont** computation fails).
- Performing classification at a specific station utilizing data from other stations that are at higher latitudes and/or develop a model leveraging the TID activity index dataset (HF-INT)



TEGERS

Thanks for your attention

LSTID Forecasting

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