



# Analysis of lonospheric Anomalies Associated with the Maduo Ms7.4 Earthquake on 22 May 2021 in China Using GIM-TEC Data

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**CONTENTS** 



# 01 Intoduction and Motivation

## Introduction 1



Pre-earthquake ionospheric TEC anomalies may be related to the preparation process of earthquakes, and studying these anomalies holds potential value for earthquake forecasting.

• **Group 1** Setting **upper and lower boundaries** using mean and median values/ quartiles.



J.Y.Liu et al. (2009)

Upper Bound =  $\mu_T + k\sigma$ Lower Bound =  $\mu_T - k\sigma$ This traditional statistical bounds always rely on empirical parameters.

Liu, J. Y., Chuo, Y. J., Shan, S. J., Tsai, Y. B., Chen, Y. I., Pulinets, S. A., & Yu, S. B. (2004, April). Pre-earthquake ionospheric anomalies registered by continuous GPS TEC measurements. In Annales geophysicae (Vol. 22, No. 5, pp. 1585-1593). Göttingen, Germany: Copernicus Publications.





 Group 2 Deep Learning-based TEC Anomaly Detection, such as the LSTM (long short temporal memory) etc.



- Usually refer to Temporal waveform learning via neural networks.
- However, after predicting TEC through a neural network, it always requires a reasonable threshold to extract anomalies.

Feng, J., Xiao, Y., Chen, J., Sun, S., & Ke, F. (2023). A method for detecting ionospheric TEC anomalies before earthquake: the case study of Ms 7.8 earthquake, February 06, 2023, Türkiye. Remote Sensing, 15(21), 5175.

### **Motivation**



Can we develop an **end-to-end model** that can generate prediction intervals while autonomously learning TEC time dependence ?





# 02

# Methods

#### Methods —— Overview

• This is the LSTM-LUBE framework. We combine the two mainstream methods, by using the advantages of neural networks, followed by a lower upper bound estimation (LUBE), so that the output of the network is directly the upper and lower bounds of predicted TEC.



## Methods —— LSTM



LSTM (Long Short-Term Memory) Long-Term Dependency Modeling

#### **Gating Mechanism**

- Forget Gate  $f_t = \sigma(W_f \cdot [h_{t_{-}}, x_t] + b_f)$
- Input Gate
- $i_t = \sigma(W_i \cdot [h_{t-1}, x_t] + b_i)$   $\widetilde{C}_t = tanh(W_c \cdot [h_{t-1}, x_t] + b_c)$  $C_t = f_t \cdot C_t - 1 + i_t \cdot \widetilde{C}_t$
- Output Gate
- $o_t = \sigma(W_o \cdot [h_{t-1}, x_t] + b_o)$  $h_t = o_t \cdot tanh(C_t)$



### Methods —— LUBE

LUBE (Lower Upper Bound Estimation)

Automatically generate a prediction interval by optimizing both PICP and PINAW.

Prediction Interval Coverage Probability .

Average Width

 $\operatorname{PICP} = \frac{1}{n} \sum_{i=1}^{n} c_i$ • n : Prediction sample size  $c_i = \begin{cases} 1, if y_i \in [L_i, U_i] \\ 0, else \end{cases}$ 

$$\frac{\text{PINAW} = \frac{1}{nR} \sum_{i=1}^{n} (U_i - L)}{\text{Normalized} \cdot R : \text{target value range}}$$
Average
$$\cdot U_i : \text{Upper bound}$$

 $L_i$ : Lower bound

**Higher PICP Lower PINAW** 





The loss of our model is achieved through the minimization of an interval objective function, which covers both two parameters.

$$Loss = \frac{1}{N} \sum_{l=1}^{N} [ReLU(y_{lower} - t_{rue}) + ReLU(y_{true} - u_{pper})] + \alpha \cdot (ReLU(q - \frac{1}{N} \sum_{l=1}^{N} ||y_{lower} \le y_{true} \le y_{upper})^{2} + \frac{1}{N} \sum_{i=1}^{N} (y_{upper} - t_{lower}) PICP PINAW$$



# 03

# Results

### **Results**—**Case Study**





 Earthquake
 2021 Maduo Earthquake, China, on May 22, 2021

- Magnitude and Focal Depth
   Mw 7.4, focal depth of 17 km
- ✓ Epicenter Location
  - 34.59° N, 98.34° E
- ✓ Test Period

May 1 to May 30

# **Results** —— Temporal Results 1







# **Results** — — Temporal Results 2



- After excluding solar and geomagnetic disturbances
- Dates of anomaly occurrence







张学民,刘静,周煜林,等.2021年云南漾濞MS6.4地震和青海玛多MS7.4地震前后的电离层扰动[J].地震,2022,42(03):1-20.



#### UT 8:00 to UT 10:00 on May 5





#### Lei Dong et al. (2022)

Dong, Lei, Xuemin Zhang, and Xiaohui Du. "Analysis of ionospheric perturbations possibly related to Yangbi Ms6. 4 and Maduo Ms7. 4 earthquakes on 21 May 2021 in China using GPS TEC and GIM TEC data." Atmosphere 13, no. 10 (2022): 1725.



#### UT 12:00 to UT 14:00 on May 7



Dong, Lei, Xuemin Zhang, and Xiaohui Du. "Analysis of ionospheric perturbations possibly related to Yangbi Ms6. 4 and Maduo Ms7. 4 earthquakes on 21 May 2021 in China using GPS TEC and GIM TEC data." Atmosphere 13, no. 10 (2022): 1725.



#### UT 12:00 to UT 14:00 on May 8



Methods	Anomaly on May 8
LSTM-LUBE (ours)	V
CNN-LUBE	V
GRU-LUBE	×
BILSTM-LUBE	V
Attention-LUBE	V
Transformer- LUBE	V
Sliding Quartile	×



#### UT 10:00 to UT 12:00 on May 21



# **Results** —— **Statistical Studies**



#### 30 global earthquake events with Ms≥6.5 between 2016-2020









The anomaly is colored in red.

It seems that TEC anomalies are increasing before earthquakes.

## **Results** —— Statistical Results 2



2016-2022 TEC



The statistical results reveal a correlation: TEC anomalies begin to increase 20 days before the earthquake, and are most significant 10 days before the earthquake.



# 03

# Conclusions

## Conclusions



Our study establishes three key findings:

- LSTM-LUBE is simpler and more direct way to detect TEC anomalies, which achieves more accurate identification of anomalies before Maduo Earthquake compared with traditional methods.
- LSTM-LUBE can be extended to various NN- LUBE, which can be applied to any problem related to prediction intervals.
- TEC perturbations exhibit spatiotemporal anomaly patterns before major quakes: The 20-day pre-earthquake window shows statistically significant ionospheric responses.





# Thank you for your listening!

**Zining Yu**