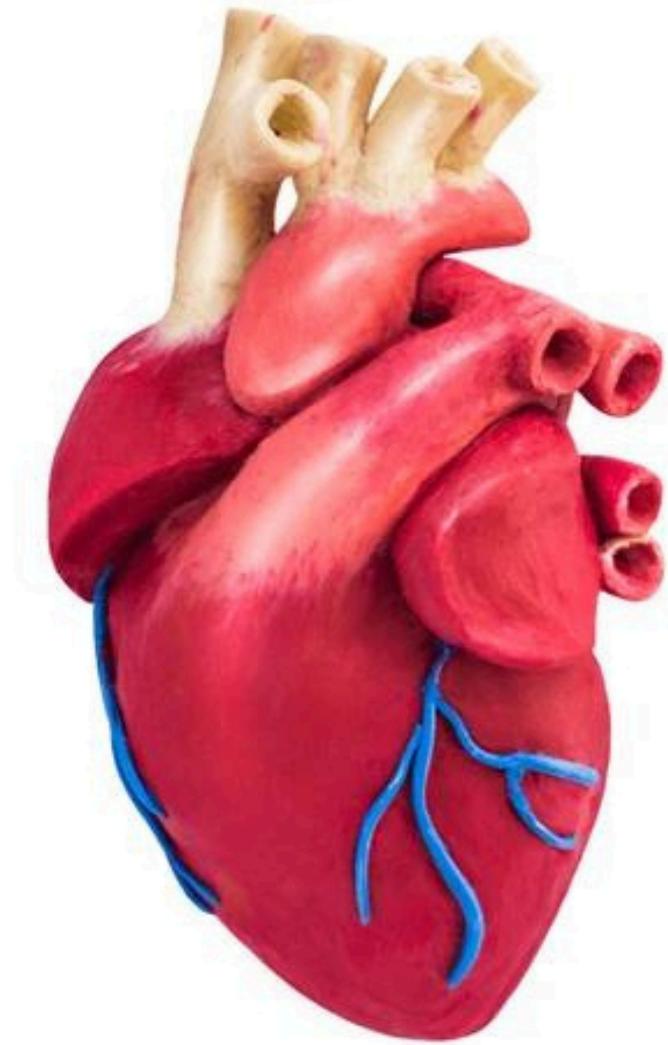




UNIVERSITAT DE
BARCELONA



DEVELOPING AN EARLY-WARNING SYSTEM FOR ACUTE MYOCARDIAL INFARCTIONS IN CATALONIA



● THESIS DEFENCE

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July 2024

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BACKGROUND AND AIMS

"ACUTE MYOCARDIAL INFARCTION IS ONE OF THE LEADING CAUSES OF MORTALITY."

- Explain the association between environmental variables and the incidence of AMI
- Identify disparities or differential susceptibility to environmental variables across different population segments.
- Implementation and comparison of models for series prediction.
 - Seasonal Autoregressive Integrated Moving Average (SARIMAX)
 - Long Short-Term Memory (LSTM)
- Assess the reliability of the predictions with increasing lead-time for effective EWS



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INTRODUCTION

"THE INTERACTION BETWEEN ENVIRONMENTAL FACTORS HAVING AN EFFECT ON THE INCIDENCE OF AMI GAINED AN INCREASED ATTENTION IN RECENT YEARS."

- Myocardial necrosis
- Existing literature indicates that both hot and cold temperatures have an impact on the incidence of AMI
- Catalonia's diverse geographical and demographic landscape
 - temperature fluctuations,
 - humidity levels, and
 - pollution levels



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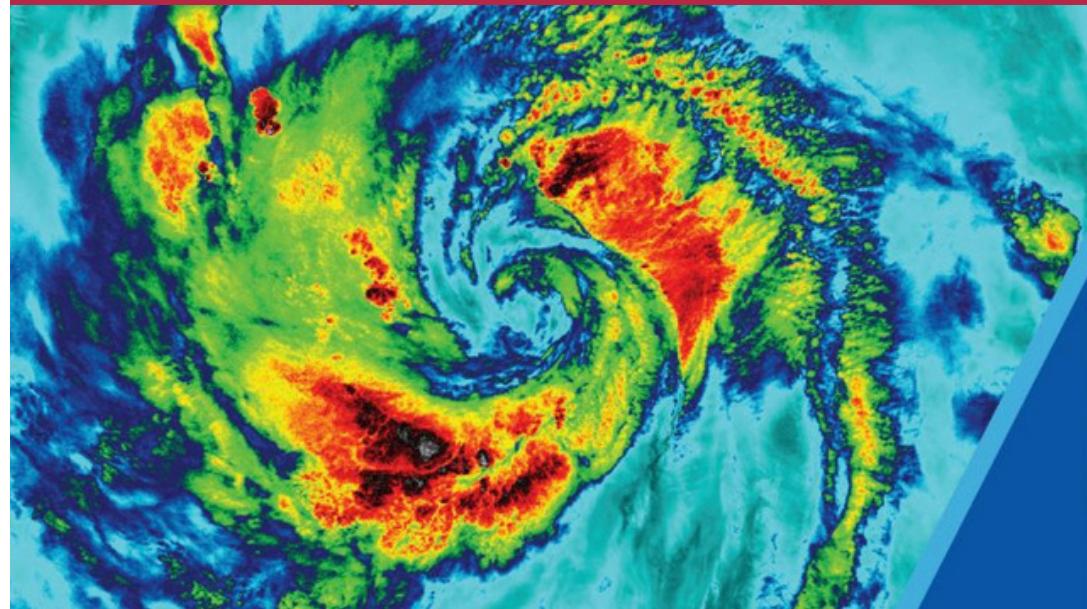
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DATASET & DATA PREPARATION

Statistical Institute of Catalonia



METEOROLOGICAL

- **239 meteorological stations**
- 30min readings
- Relative Humidity
- Temperature

AIRQUALITY

- **90 monitoring stations**
- Hourly readings
- Carbon monoxide (CO),
- Nitrogen dioxide (NO₂),
- Ozone (O₃),
- Particulate matter (PM),
- Sulfur dioxide (SO₂).



Hospitals



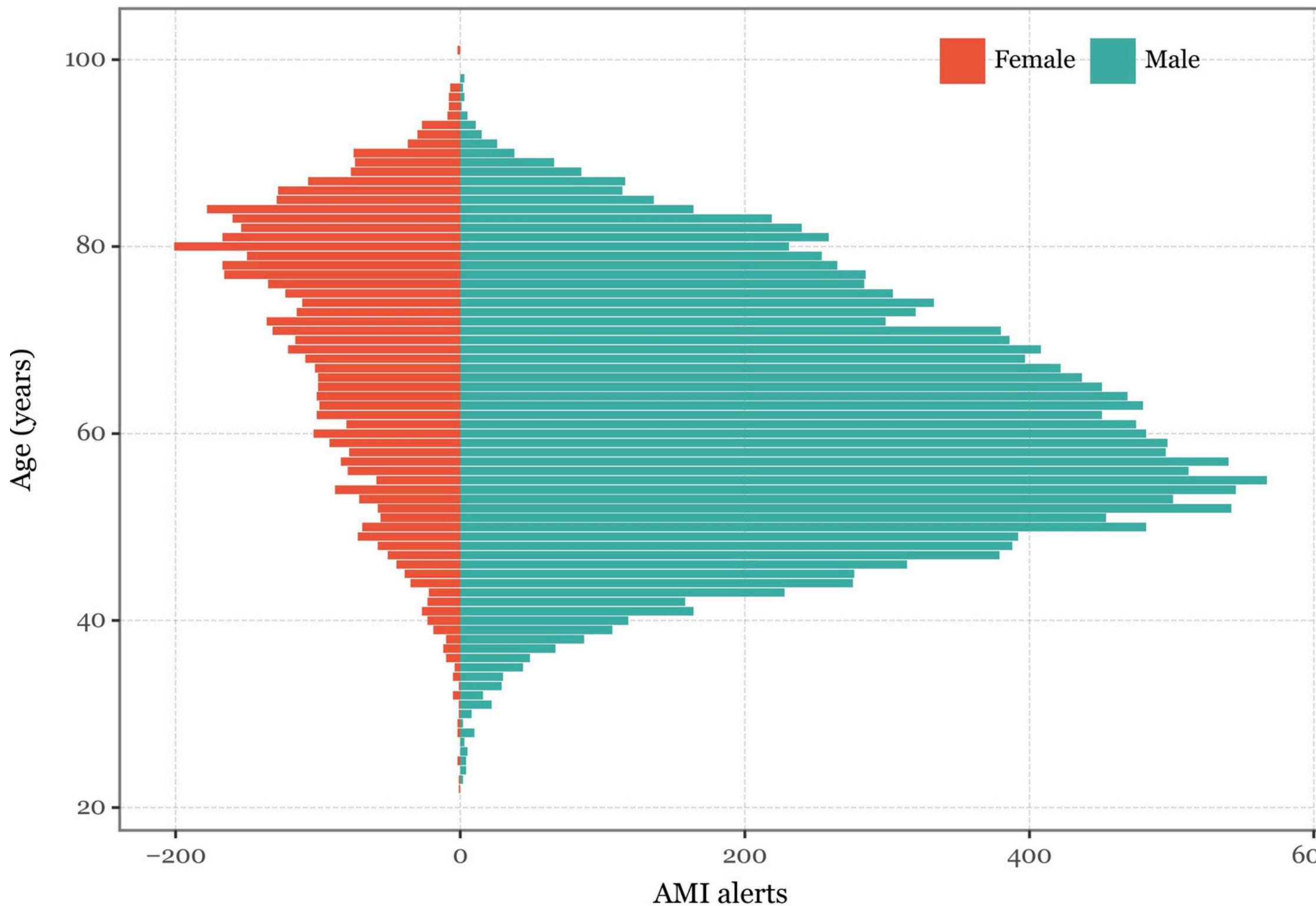
HEALTH

- **10 hospitals**
- Daily counts
- 22,812 admissions
- 948 municipalities
- stratified - province, sex
age

THE CHOICE OF ASIR



Age and sex distribution of all AMI alerts (2010-2018)



THE CHOICE OF ASIR



*"RESEARCHERS OFTEN FACE THE
CHALLENGE OF COMPARING INCIDENCE
RATES ACROSS POPULATIONS WITH
DIFFERENT AGE DISTRIBUTIONS."*

- Ensures comparability of events across populations.
- Ensures comparability across different age distributions.
- Adjusts for age as a confounding factor, providing a more accurate representation of AMI incidence.
- Allows comparisons between regions or over time periods.

THE CHOICE OF SPATIAL LEVEL

"THE GEOGRAPHICAL DISTRIBUTION OF AMI INCIDENCE WITHIN CATALUNYA IS INFLUENCED BY A MYRIAD OF FACTORS, INCLUDING URBANISATION, SOCIOECONOMIC STATUS, HEALTHCARE INFRASTRUCTURE, AND ENVIRONMENTAL EXPOSURES."

- **Higher regions offer more stable geographical units** for examination than counties level.
- **High sparsity** even on higher region level **on daily basis**
- Given that the yearly incidence of AMI alerts around 35 cases per year per 100k inhabitants, **the signal-to-noise ratio in the least populated areas becomes notably skewed**, rendering analysis at the daily scale impractical.

THE CHOICE OF TARGET



Daily Age Standardised Incidence Rate for AT01

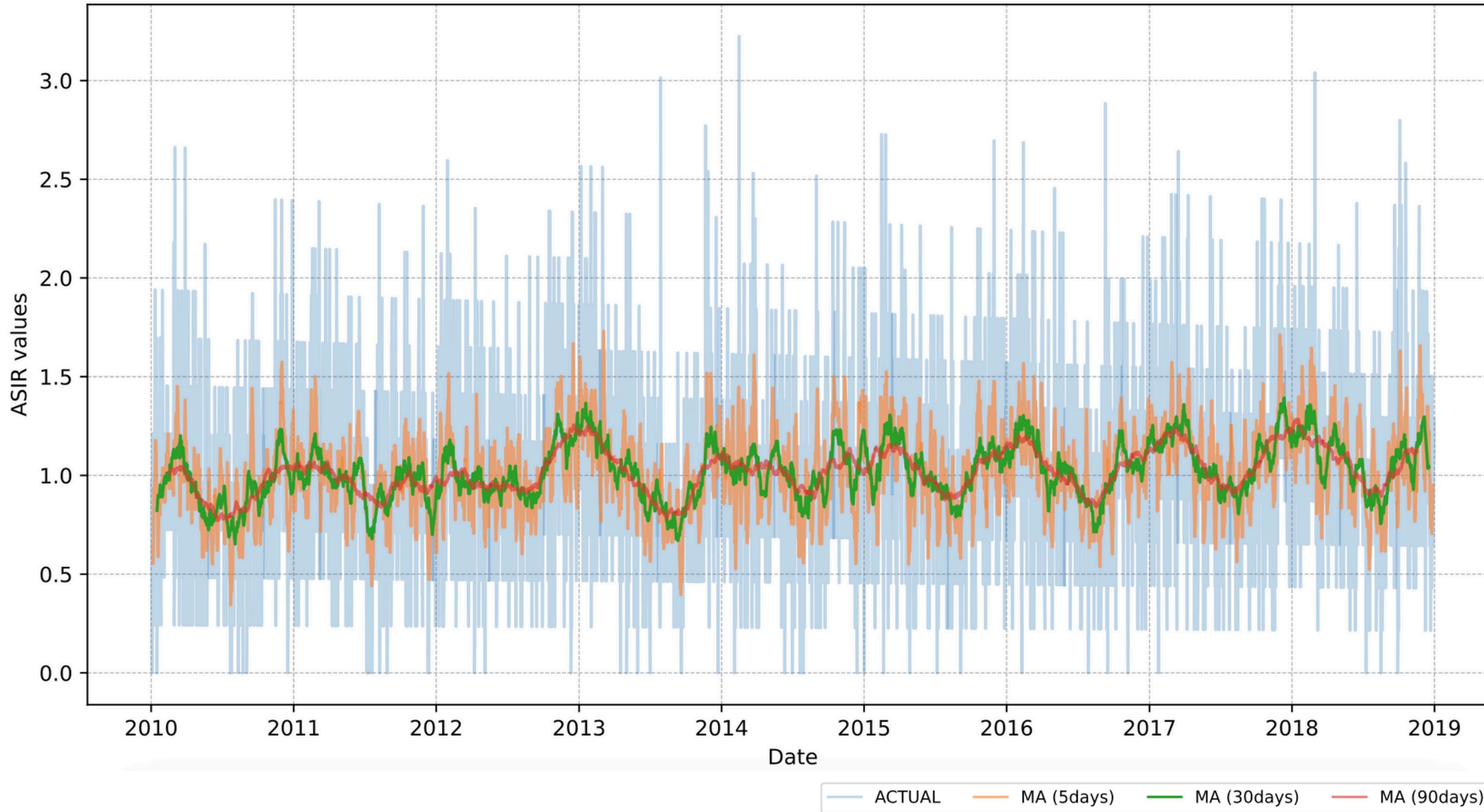


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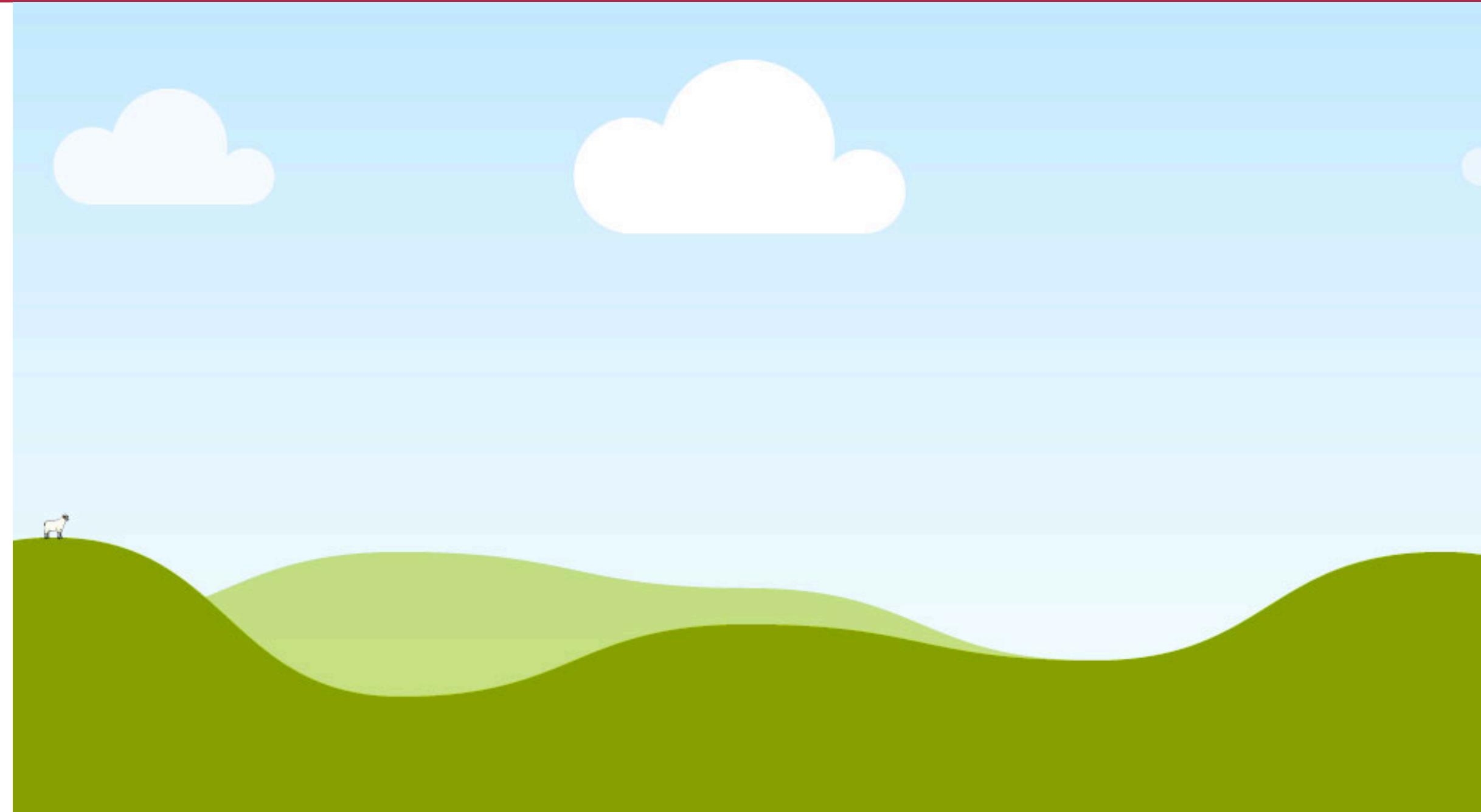
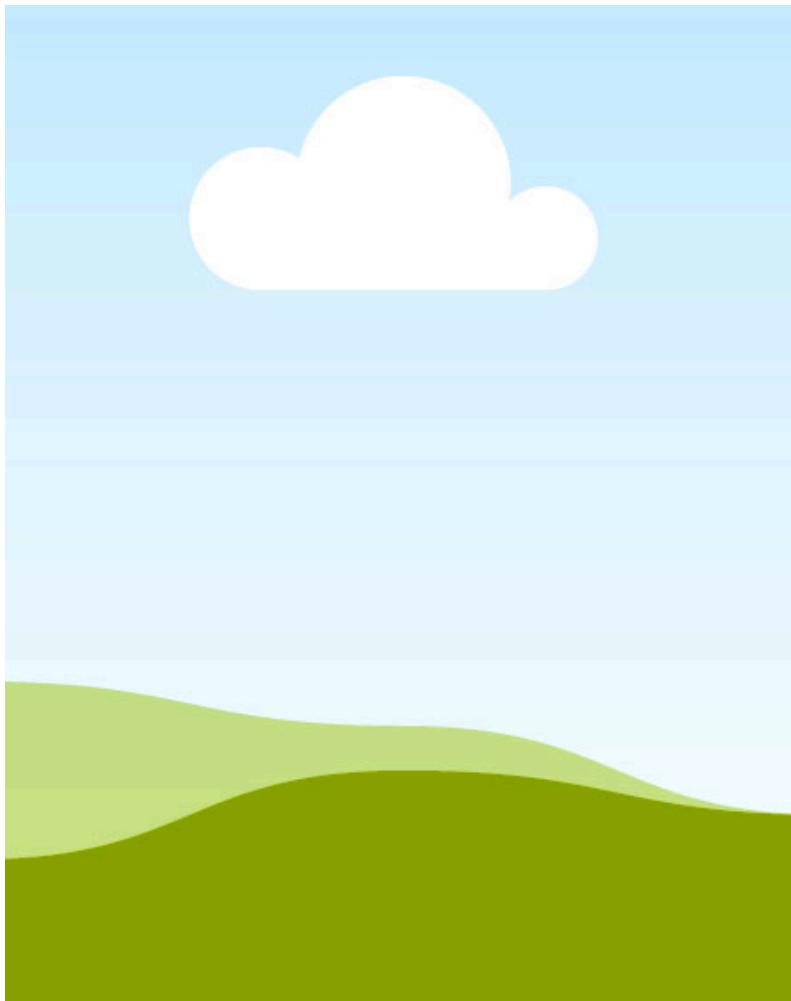
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METHODOLOGY

SEASONAL AUTOREGRESSIVE INTEGRATED MOVING AVERAGE



METHODOLOGY

LONG SHORT-TERM MEMORY

- A type of RNN architecture specifically designed to model sequence data while **addressing the vanishing gradient problem**.
- The LSTM cell has a memory cell and **three gates**: input gate, forget gate, and output gate
- Capable of learning **long-term dependencies**
- Capture **complex patterns** and relationships

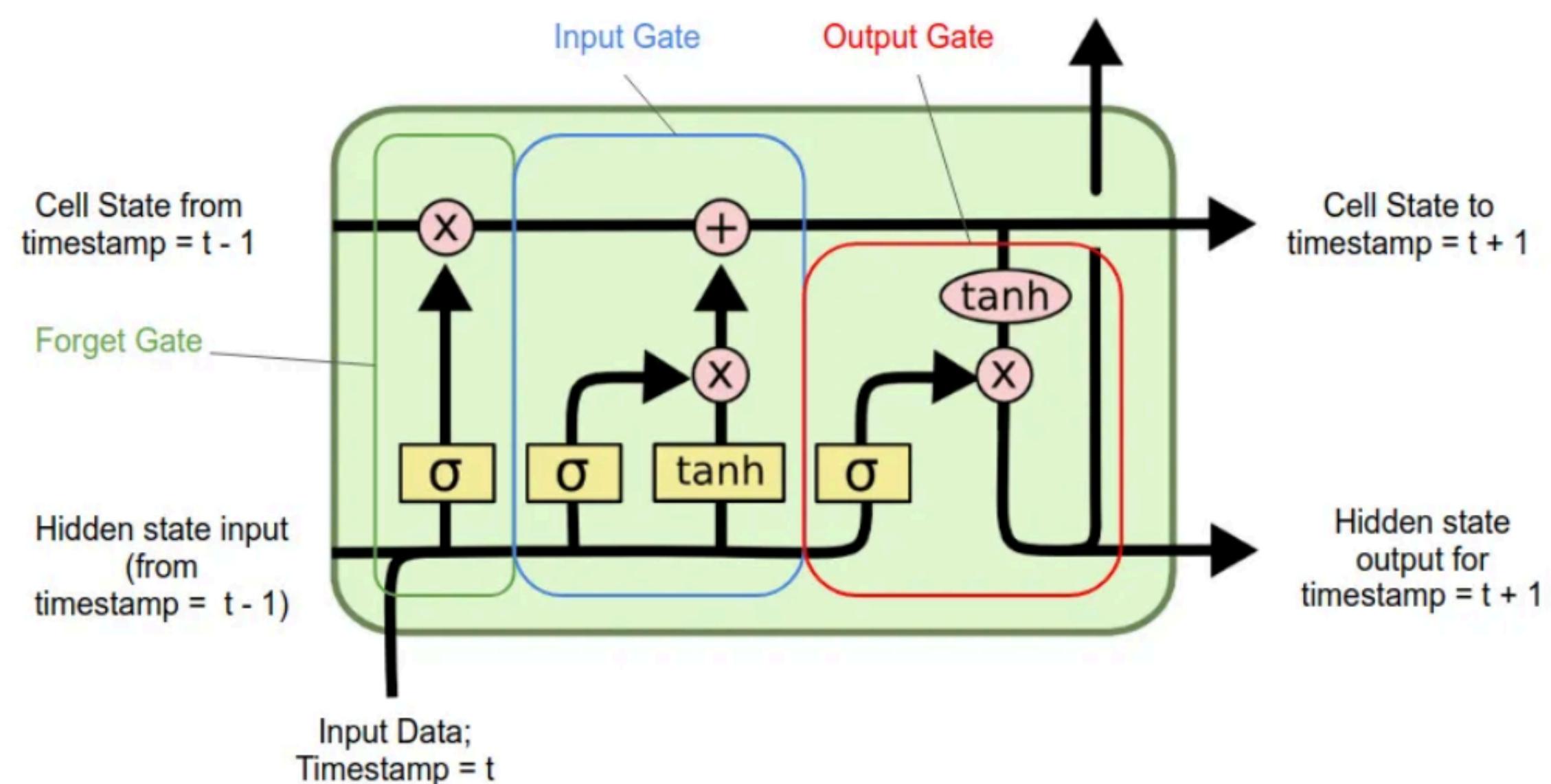


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RESULTS & LIMITATIONS



01 THE OPTIMAL MODEL

SARIMAX(1, 1, 1) x (2, 0, 0, 52)

02 EXOGENOUS VARIABLES

Better fit than SARIMA

03 LSTM

Similar performance to SARIMAX

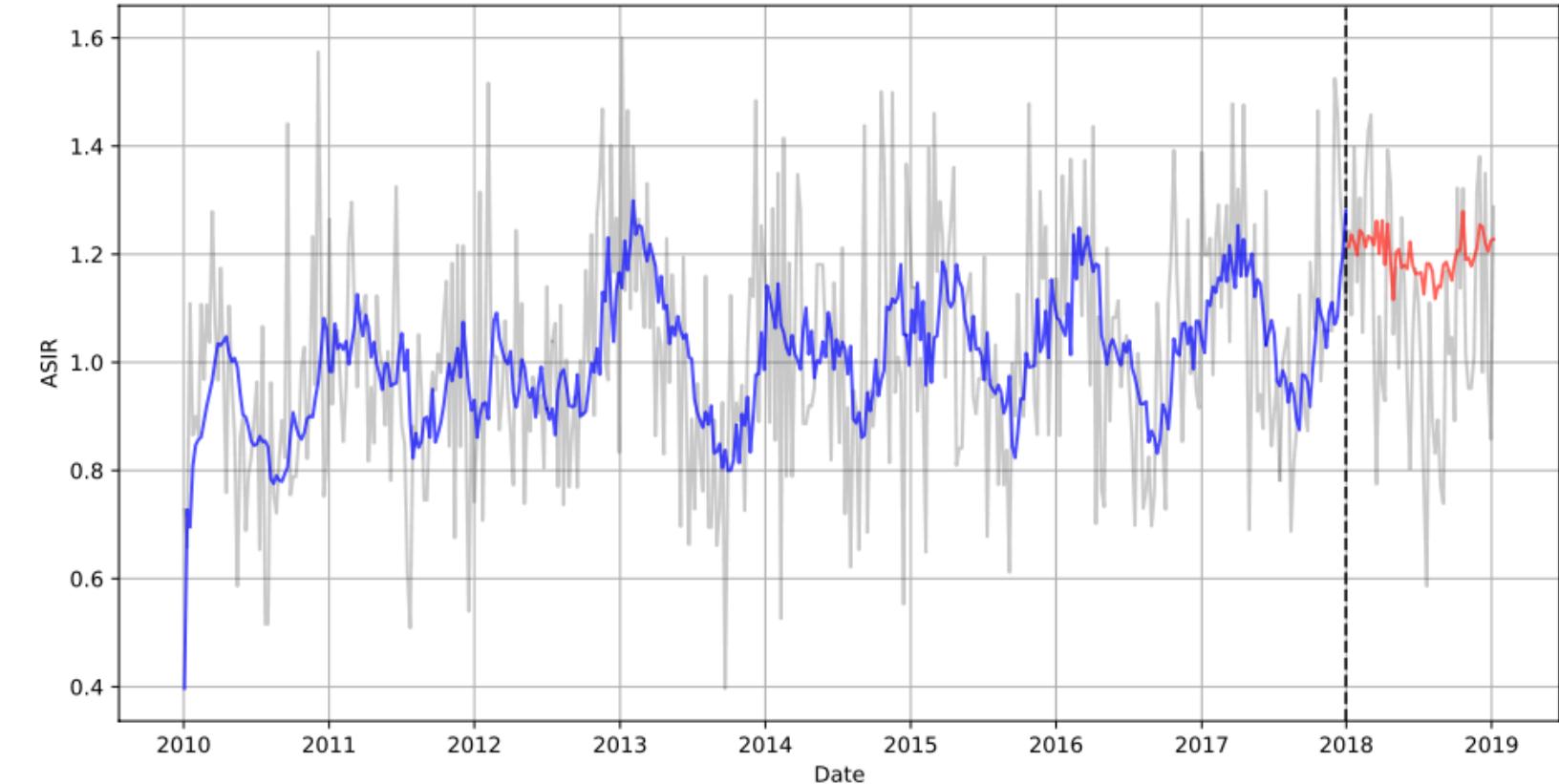
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01

THE OPTIMAL MODEL

SARIMAX(1, 1, 1) x (2, 0, 0, 52)

(A) Time Series Fitted by SARIMA



02

EXOGENOUS VARIABLES

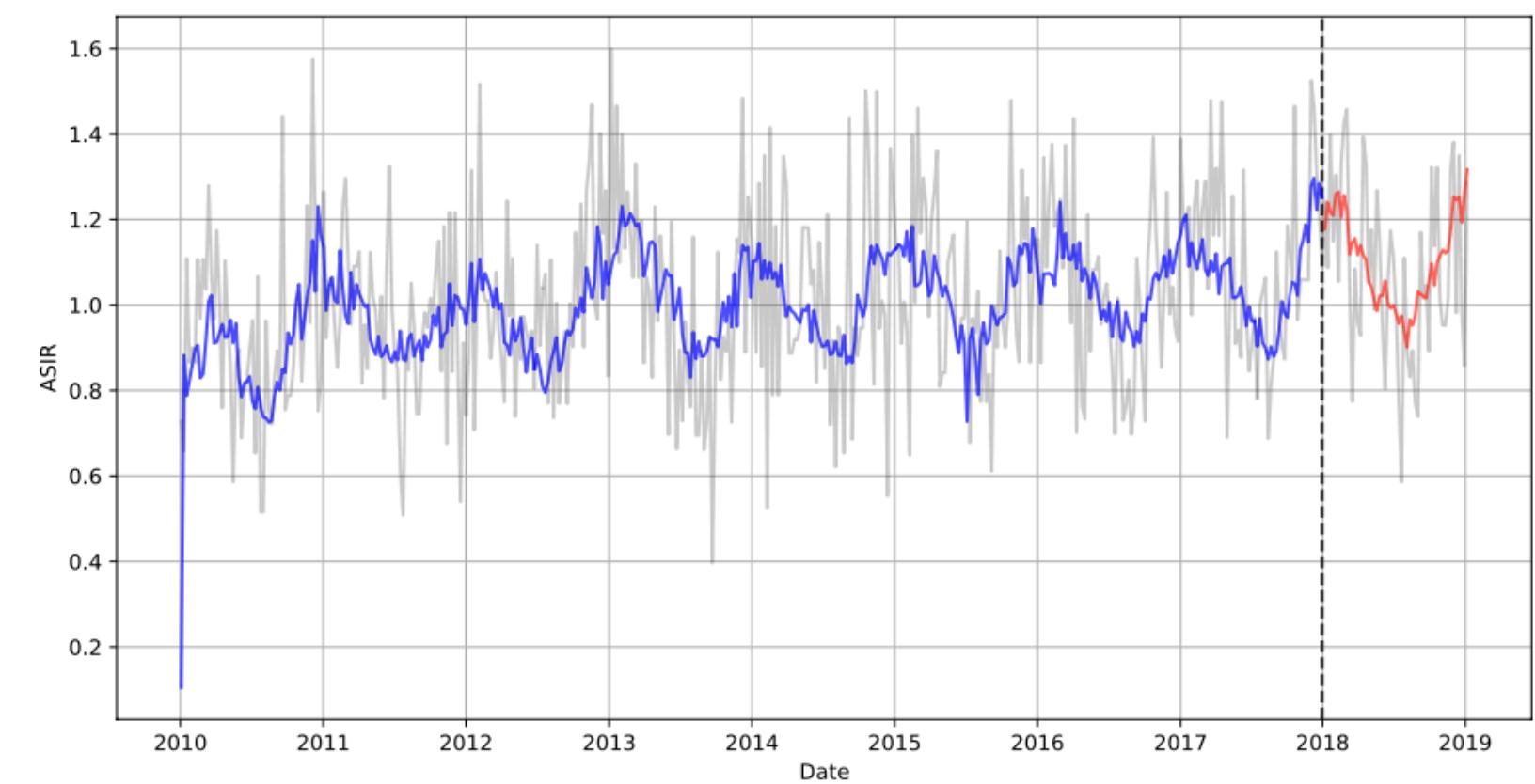
Better fit than SARIMA

03

LSTM

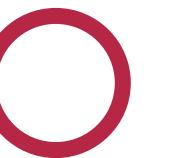
Similar performance to SARIMAX

(B) Time Series Fitted by SARIMAX



— Actual Time Series --- End of Training — Training Predictions — Test Predictions

RESULTS & LIMITATIONS



01 THE OPTIMAL MODEL

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02 EXOGENOUS VARIABLES

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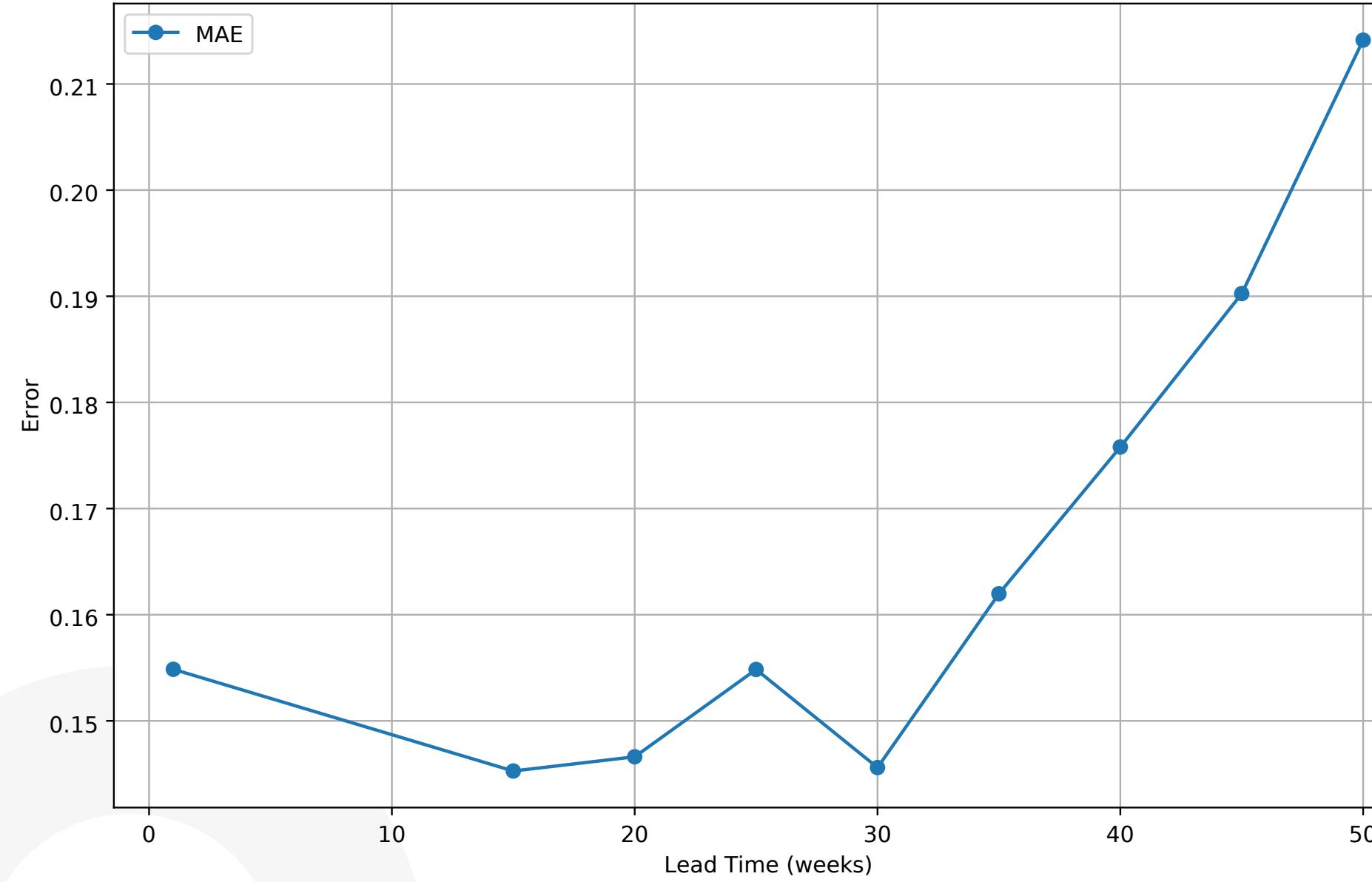
Similar performance as SARIMAX

Hyperparameter tuning - **2 layers, dropout rate 0.2, Adam**

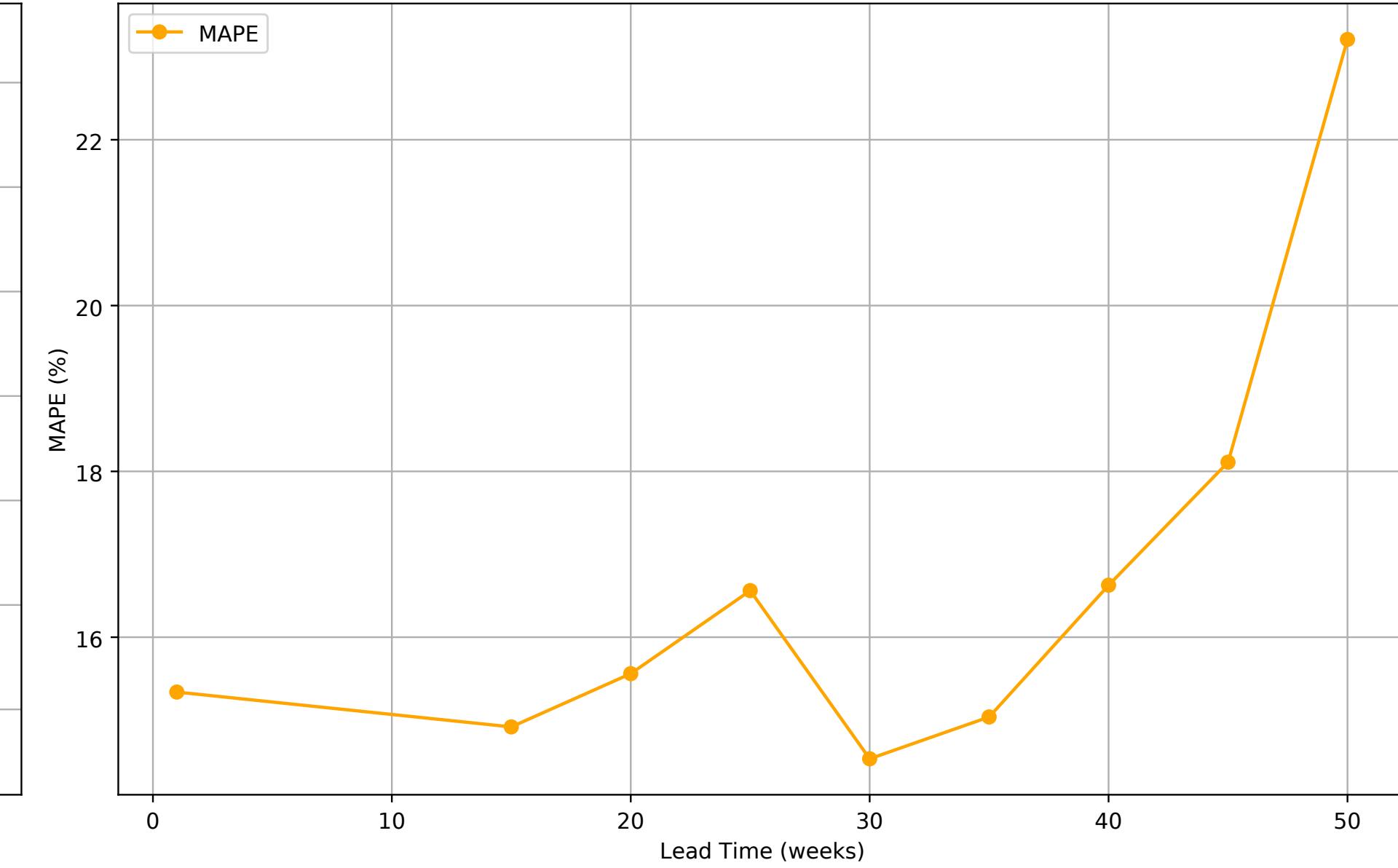
PREDICTION SKILL VS. LEAD-TIME

*"TO MAINTAIN PREDICTION ACCURACY OVER A MORE EXTENDED FORECAST, IT IS ESSENTIAL TO UPDATE THE MODEL AND **ADD NEW DATA REGULARLY TO MAINTAIN THE ACCURACY AND EFFECTIVENESS OF THE EWS.**"*

Prediction Skill vs Lead-time (MAE)



Prediction Skill vs Lead-time (MAPE)



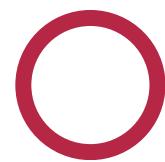
FEATURE IMPORTANCE

"THERE ARE VARIOUS METHODS THAT EXTRACT AND ANALYSE THE IMPORTANCE SUCH AS PERMUTATION FEATURE IMPORTANCE, FEATURE IMPORTANCE FROM **MODEL COEFFICIENT**, **SHAPLEY ADDITIVE EXPLANATIONS (SHAP) VALUES OR LOCAL INTERPRETABLE MODEL- AGNOSTIC EXPLANATIONS (LIME)**."

Feature	MAE	% Increase
<i>Baseline MAE</i>	0.1590	–
Temperature	0.1800	13.21%
Humidity	0.1689	6.23%
SO2	0.1640	3.14%
NO2	0.1635	2.83%
CO	0.1629	2.45%
PM10	0.1624	2.14%

- The **permutation feature importance** measures the importance of the feature by comparing the error of the model with and without permuting the feature's values.
- To ensure robustness in our feature importance analysis, we employed **K-fold cross-validation**
- Other techniques that could be used are **adding noise, integrating gradients or ablation**

LIMITATIONS



01

Models require accurate values of exogenous variables for prediction

02

Interpretability of LSTM models and requirement of large amount of data for tuning

03

SARIMAX captures only linear relationship between the predictor and response variables

04

Use of linear interpolation can cause minor disruptions

05

Precision of the environment data

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CONCLUSION

AS **CLIMATE CHANGE CONTINUES TO ALTER WEATHER PATTERNS**, THE INCIDENCE OF AMI MAY BE INFLUENCED BY NEW AND CHANGING ENVIRONMENTAL FACTORS, MAKING TIMELY AND ACCURATE PREDICTIONS EVEN MORE CRITICAL.

01

IMPORTANCE OF ENVIRONMENTAL FACTORS

- **18% improvement** of forecast accuracy
- By identifying main environmental predictors of AMI, health authorities can implement interventions that would be targeted **on high-risk areas and time**

02

LEAD TIME UP TO HALF YEAR AHEAD

03

TEXT

TEXT



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THANK YOU

● FOR YOUR ATTENTION

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