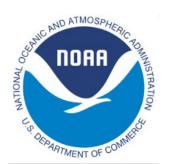
## Deep Learning Based Long Short-Term Memory (LSTM) Prediction System for the Indian Ocean Dipole

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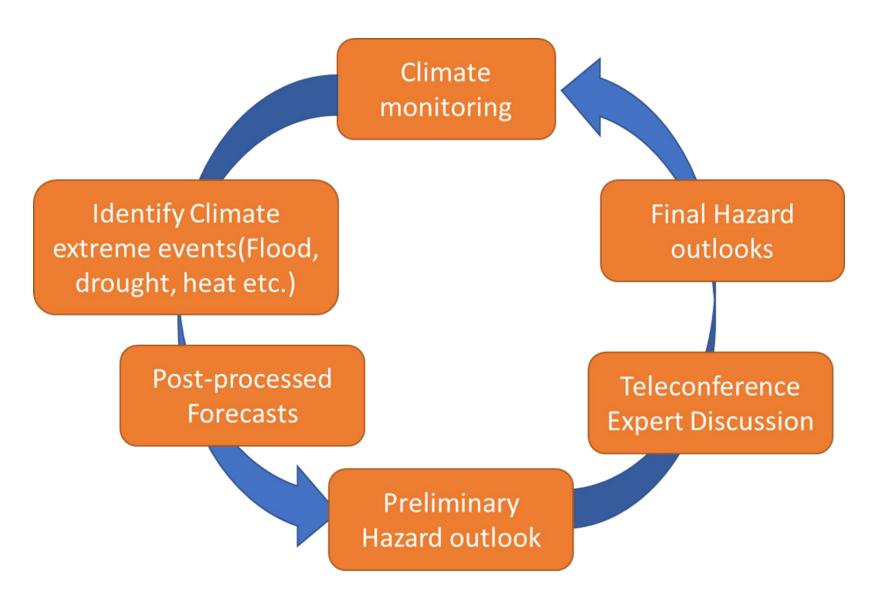




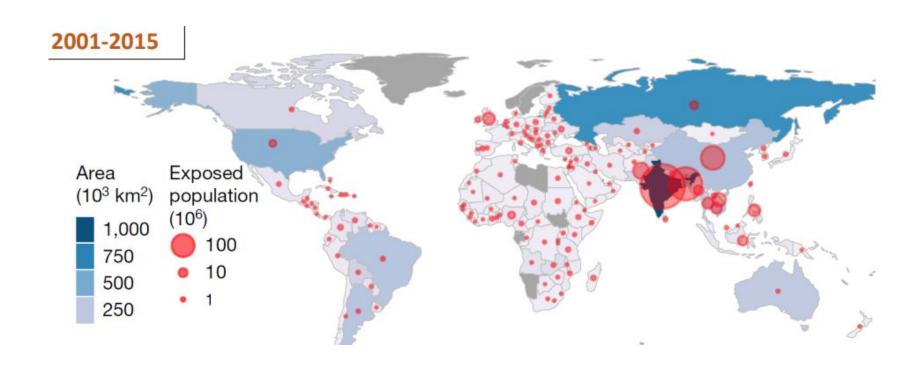
#### **Motivation**

- Estimation of extreme events, such as flood & drought, remains a significant management challenge.
- The Indian Ocean Dipole (IOD), a climate phenomenon characterized by sea surface temperature anomalies in the Indian Ocean, plays a significant role in driving extreme weather events
- Traditional methods for estimating the Indian Ocean Dipole (IOD) have limitations, such as random and systematic errors, which reduce their effectiveness in water resources planning.
- Artificial intelligence (AI) models offer promising avenues for enhancing the estimation of the Indian Ocean Dipole (IOD).

#### Goal



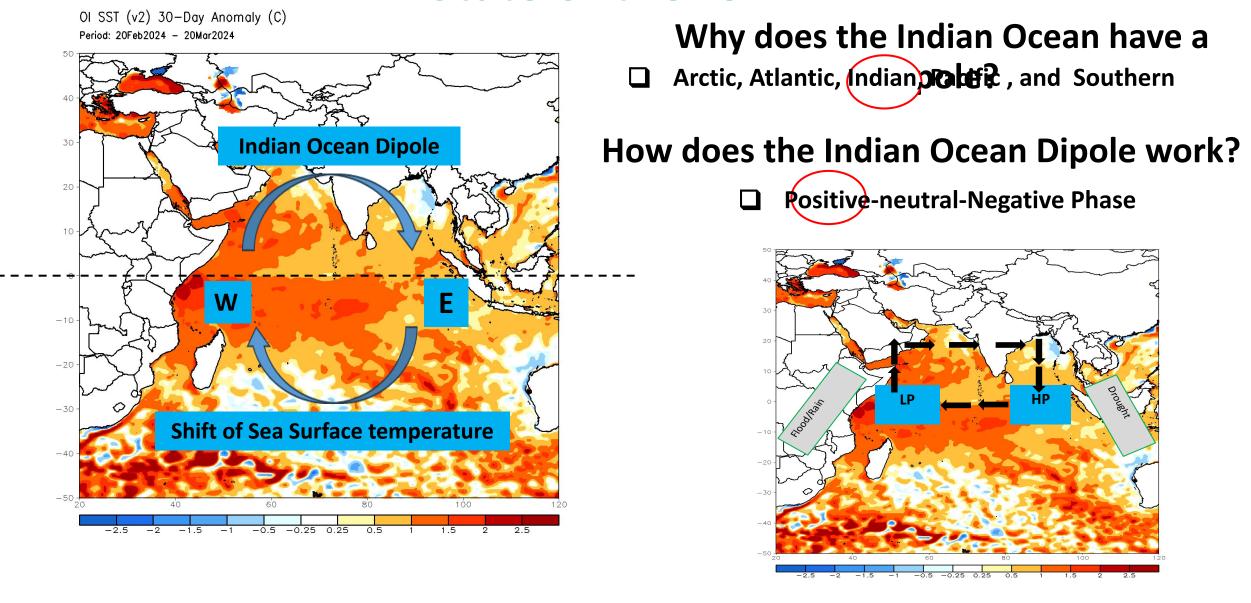
#### **Global Flood Exposure Map**



[Tellman et. al, 2021, Nature]

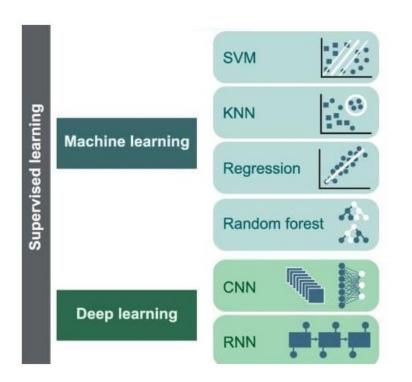
Intensity of the IOD is represented by anomalous SST gradient between the western equatorial Indian Ocean (50E-70E and 10S-10N) and the southeastern equatorial Indian Ocean (90E-110E and 10S-0N). This gradient is named as Dipole Mode Index (DMI).

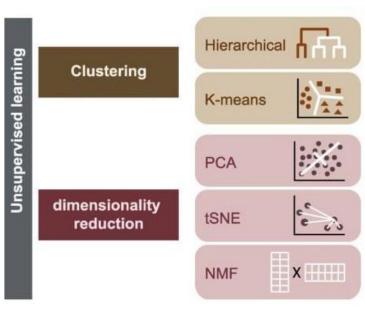
#### **State of the IOD**



During a positive phase, warm waters are brought up to the western part of the Indian Ocean, and in the eastern Indian Ocean, cold, deep waters rise to the surface.

#### AI Techniques

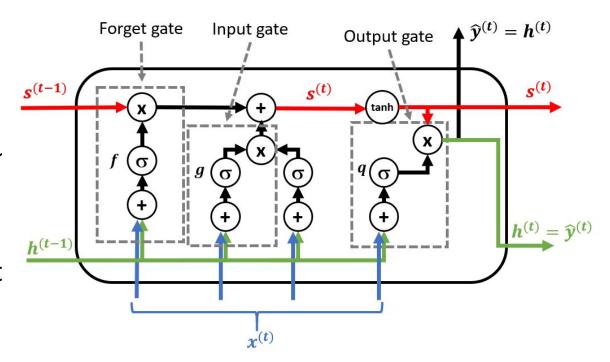




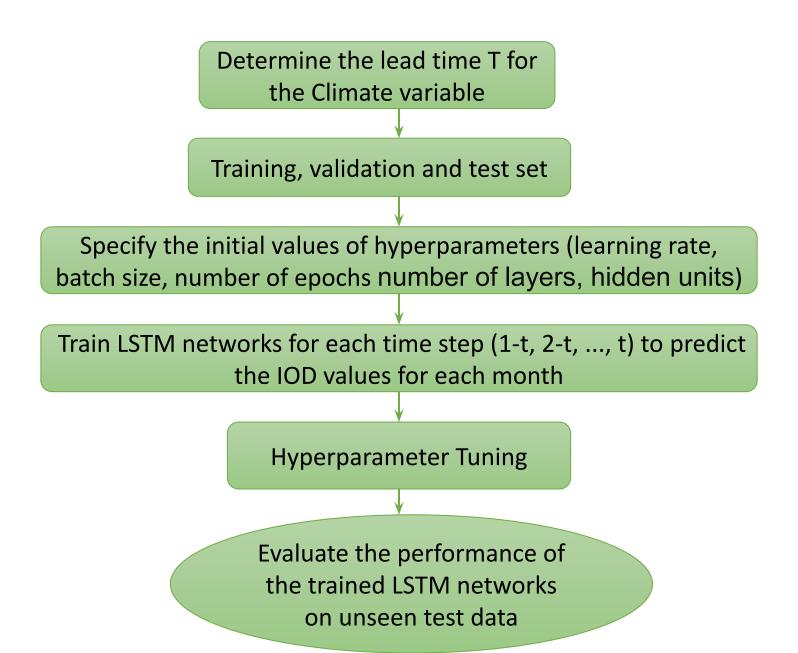
- Decision Tree Regressor (DT
- Random Forest Regressor (RF)
- Gradient Boosting Regressor (GB)
- Deep Neural Network (DNN)
- Recurrent Neural Networks (RNN)
- Bayesian Regression Tree(BART)

#### LONG SHORT-TERM MEMORY(LSTM)

- ☐ There are three types of gates within a unit:
  - <u>Forget Gate</u>: conditionally decides what information to throw away from the block.
- <u>Input Gate</u>: conditionally decides which values fror the input to update the memory state.
- <u>Output Gate</u>: conditionally decides what to output based on input and the memory of the block.
- ☐ Each unit is like a mini-state machine where the gates of the units have weights that are learned during the training procedure.



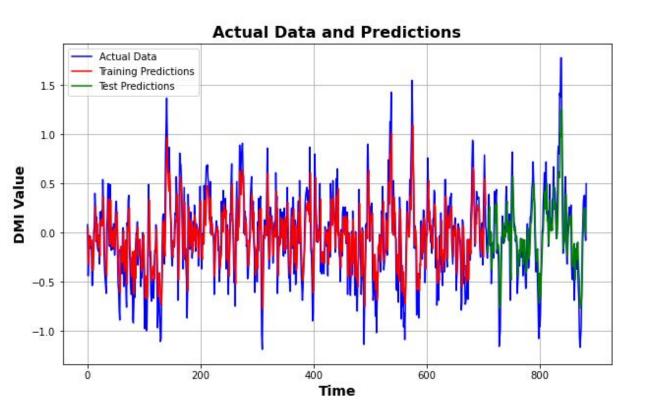
#### **Research Framework**

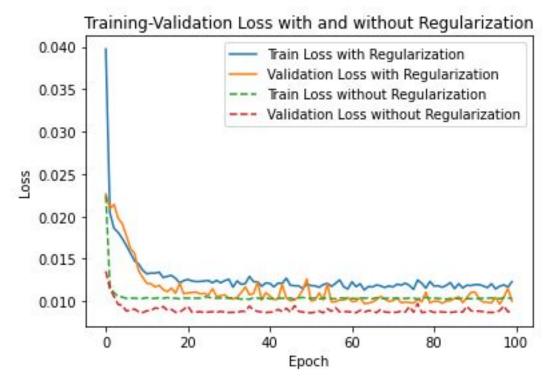


#### **Research Framework**

Dataset:
□ Dipole Mode Index (DMI) which is based ERSSTv5( NOAA Extended Reconstructed SST V5 )
□ Date period: 1950-2024
□ Temporal resolution: Monthly
Methodology:
LSTM Model setup
Model optimization
Model evaluation:
☐ Trained period:1950- 1990 and Testing period:1991-2021
☐ Systematic/random error

#### **Model Optimization**

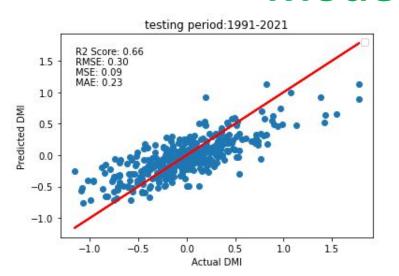


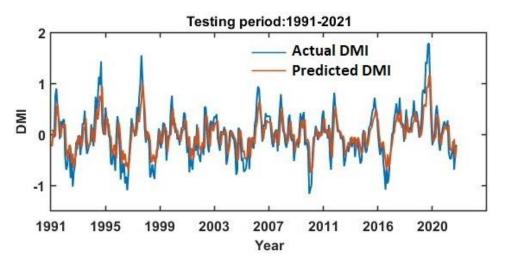


Train Score: 0.30 RMSE

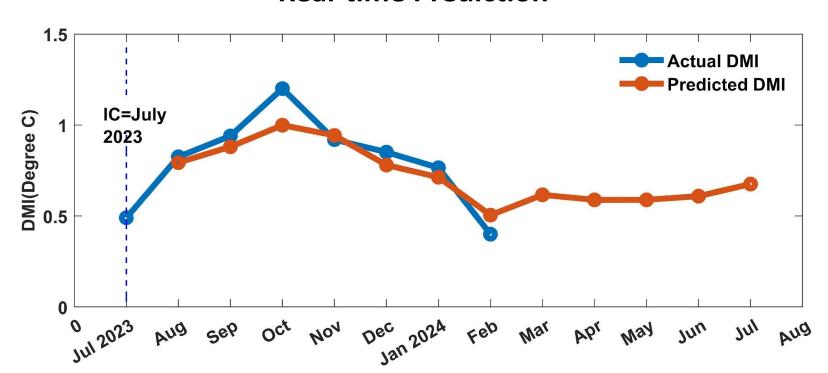
Test Score: 0.28 RMSE

#### **Model Evaluation**

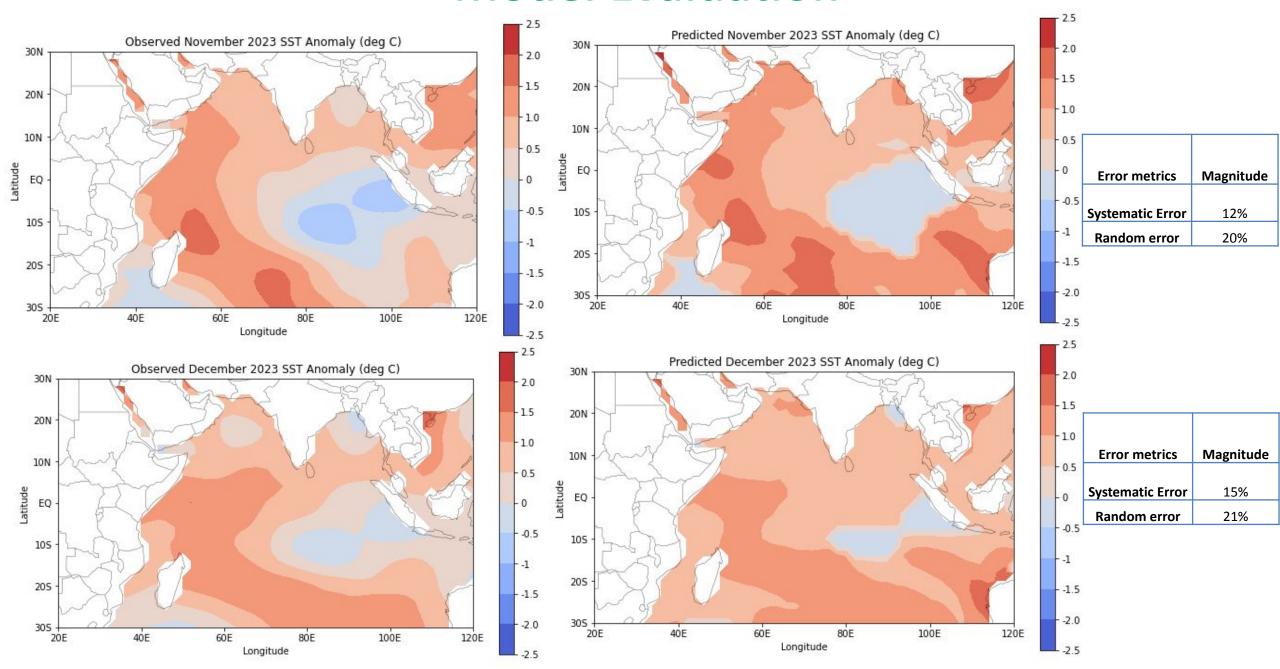


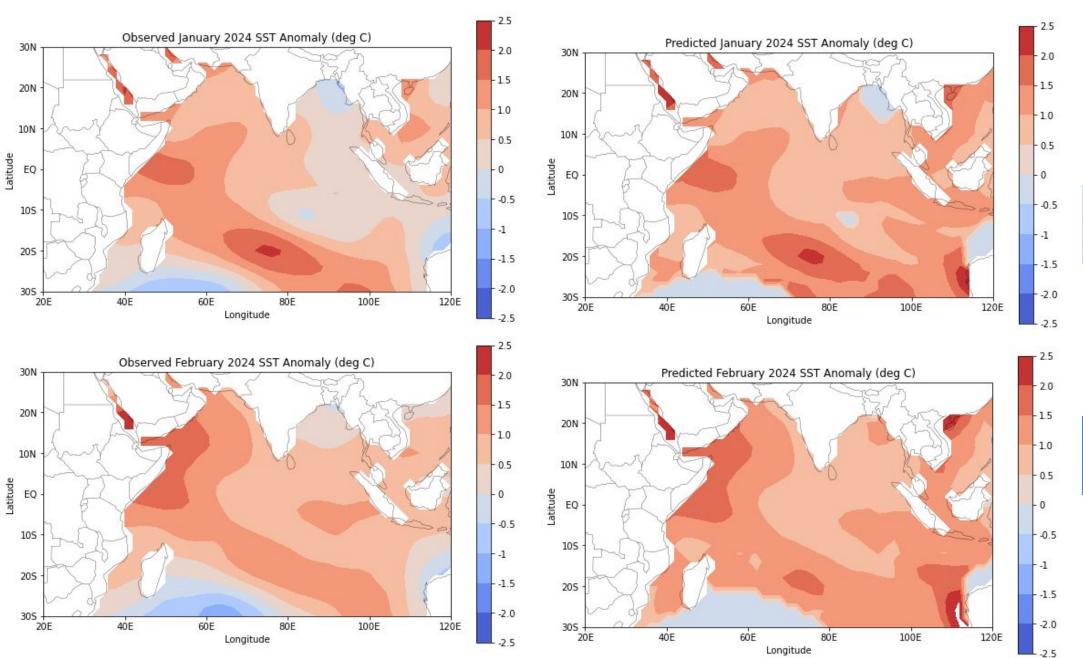


#### **Real-time Prediction**



#### **Model Evaluation**





Error metrics	Magnitude
Systematic Error	15%
Random error	23%

Error metrics	Magnitude
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Systematic Error	15%
Random error	18%

### Summary

- The model evaluation results indicated that the LSTM technique was able to reduce significantly the random and systematic error with high correlation coefficients.
- Realtime -12-month IOD forecasts results indicated the deep learning-based LSTM model to be capable of forecasting the IOD index (DMI) well in advance with excellent skills.
- The LSTM model forecasts are solely dependent on past observed data and hence have higher skills in forecasting the IOD index.
- The study demonstrates promising results in SST Anomaly forecasting, showcasing the potential of LSTM model as a reliable tool for climate prediction.
- Overall, the development of SST Anomaly forecasting system contributes to the advancement of climate science and contributes to building climate resilience in the Indian Ocean region

# Thank you ehsan.bhuiyan@noaa.gov