FogNet: 3D CNN for Coastal Fog Prediction

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https://gridftp.tamucc.edu/fognet/





- Fog substantially affects transportation, including air and sea
- Accurate fog prediction \rightarrow actionable insights for safety & efficiency



Image: Bumble Bee Productions (https://youtu.be/5Z_Hx0nvR9w)



Image: zzathras777 (https://www.flickr.com/photos/zzathras777/)

- Developed a fog prediction model for the Texas Coastal Bend (6, 12, & 24-hour)
- ullet Currently analyzing model o want longer lead times, extension to coastal CONUS

Related publications

Kamangir, Hamid, et al. "FogNet: A multiscale 3D CNN with double-branch dense block and attention mechanism for fog prediction." Machine Learning with Applications 5 (2021): 100038.

Kamangir, Hamid, et al. "Importance of 3D convolution and physics on a deep learning coastal for model." Environmental Modelling & Software (2022): 105424.

Krell, Evan, et al. "Explainable artificial intelligence for geoscience models with highly correlated, high-dimensional inputs: a case study with FogNet, a 3D CNN for predicting coastal fog." In-progress.

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FogNet



- 3D CNN with attention, dense block, & dilated convolution
- Imbalanced dataset: very few fog cases
- Input data: fusion of NWP output & satellite imagery

Architecture overview

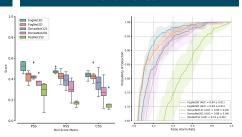


Physics-based channel groups



- G1 wind
- G2 turbulence kinetic energy & humidity
- G3 lower atmosphere thermodynamic profile
- G4 surface atmospheric moisture & microphysics
- G5 sea surface temperature

Performance



Explainable Artificial Intelligence (XAI): what strategies has FogNet learned?

- Which spatio-temporal features are most influential for FogNet's decisions? Do these strategies match forecaster knowledge?
- Model performs better for Advection fog (A) than the other types
- How do strategies differ between fog types? We customize & apply Shapley-based XAI techniques to investigate the model

Hits (R. A-R)





























XAI results

- Channel-wise superpixels Spatial features: where is FogNet looking?
- Top channels Which features are most influential?

























0.06

- 1. Vertical velocity, 950 mb 6 hours Surface T minus dew point temperature
- 12 hours Vertical velocity, 875mb.
 - at initialization















