

642943: Residual prediction to improve the meteorological based sea surface temperature forecasts using ANN

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Introduction

- Sea surface temperature (SST) is utmost important for locating fishing zones
- SST prediction is largely based on numerical models, but its output often largely deviate from ground truth due to many implicit assumptions
- This study attempts SST forecasts using meteorological parameters as inputs
- Further it also improves the forecasts by residual prediction

Methodology

Meteorological inputs

Meteorological based forecasts

Data

- Target data: JCOPE SST hourly data
- Input data: ECMWF reanalysis hourly data (ERA5)
- Ground truth: Buoy and drifter data
- Buoy: KEO buoy (32.3°N, 144.6°E)
- Drifter: JMA drifting ocean data buoys $(32.3^{\circ}N)$, $140^{\circ}E$ 140'E 145'E 150'E 155'E 160'E 165'E 170'E 175'E





Updated hourly SST predictions

Conclusions

- ANN based meteorological forecasts can be an alternative to physics based models for small spatial regions
- Moving average method improves the prediction accuracy significantly
- Residual prediction is an effective technique to further improvise this ANN based predictions like an offline data assimilation
- Buoy predictions shown more accuracy in ANN based and residual predictions
- Drifter predictions were showing low accuracy as they were not temporally correlated at same place

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0	100	200	300	400	500	600	700	
Time in hours								

RMSE before - 0.7188

RMSE for Drifter

Leadtime	Without moving average	Moving average (1 to 48)						
1 Hr	0.6804 (°C)	0.6364 (°C)						
8 Hr	0.6735 (°C)	0.6313 (°C)						
24 Hr	0.6834 (°C)	0.6213 (°C)						
After residual prediction								
1 Hr	0.1224 (°C)	0.1067 (°C)						
8 Hr	0.5034 (°C)	0.5534 (°C)						
24 Hr	0.8432 (°C)	0.8454 (°C)						
() () () () () () () () () () () () () (observed SST before residual corr	rection						

correlated at same place

Time in hours

150

200

after residual correction

100

Beneficiary of the study
Fishery industries
Meteorological organization
Future scope
Deep learning for SST prediction from larger spatial domain of meteorological parameters

Prof. Masaaki liyama lab's other research

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SST image in-painting

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- AI meets fisheries
 - High-resolution Bathymetry Data Acquisition



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