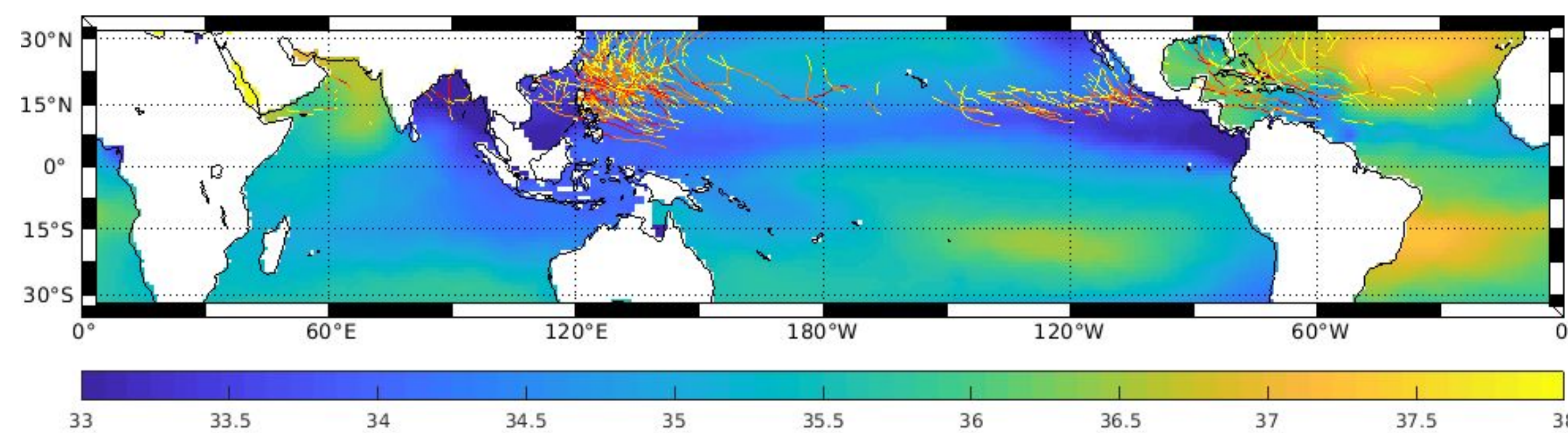


# A GRIDDED UPPER-OCEAN STRATIFICATION DATASET FOR TROPICAL WEATHER AND CLIMATE STUDIES



Nguyen Dac Da<sup>(1,2)</sup>, Gregory Foltz<sup>(1)</sup>  
Corresponding: [gregory.foltz@noaa.gov](mailto:gregory.foltz@noaa.gov)

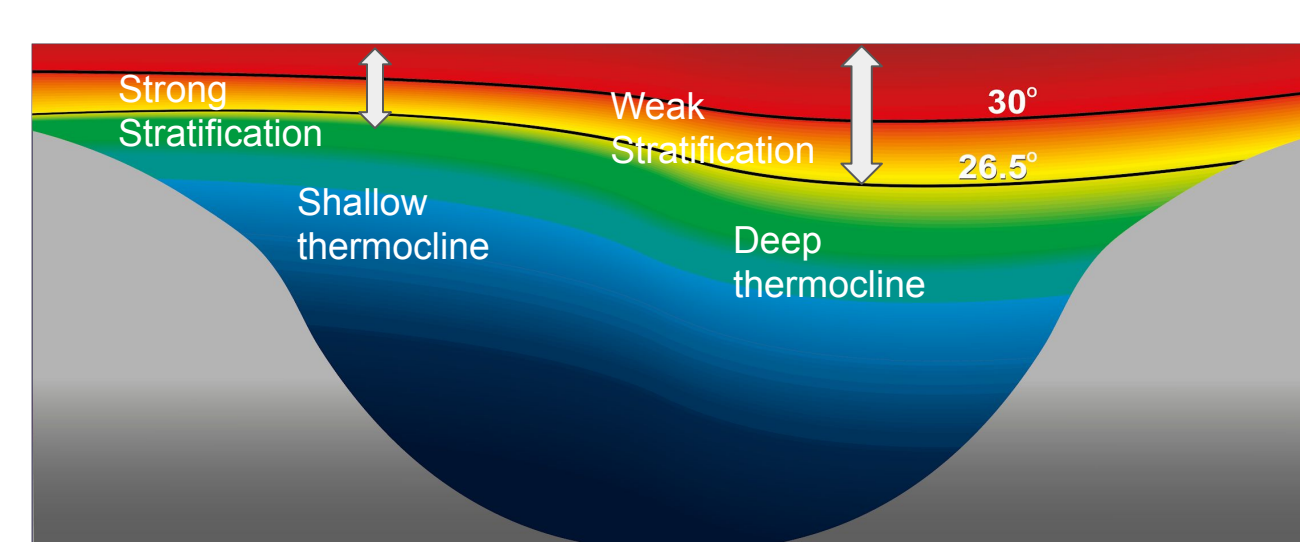
(1) Atlantic Oceanographic Meteorological Laboratory (AOML) - NOAA  
(2) Cooperative Institute for Marine & Atmospheric Studies (CIMAS) - University of Miami



Tropical Cyclone Tracks (category 3 and higher, from Hurdad) and SSS (August-Oct) from 2004 - 2017. (Data from Hurdad and WOD)

## I. Introduction

- Ocean provides the major heat source for Tropical Cyclones (TC) via surface enthalpy (turbulent heat) flux.
- Ocean stratification also impacts TC intensity by modulating SST via:
  - stronger stratification resists vertical mixing
  - > keeping **warmer SST** -> favorable for TC intensification



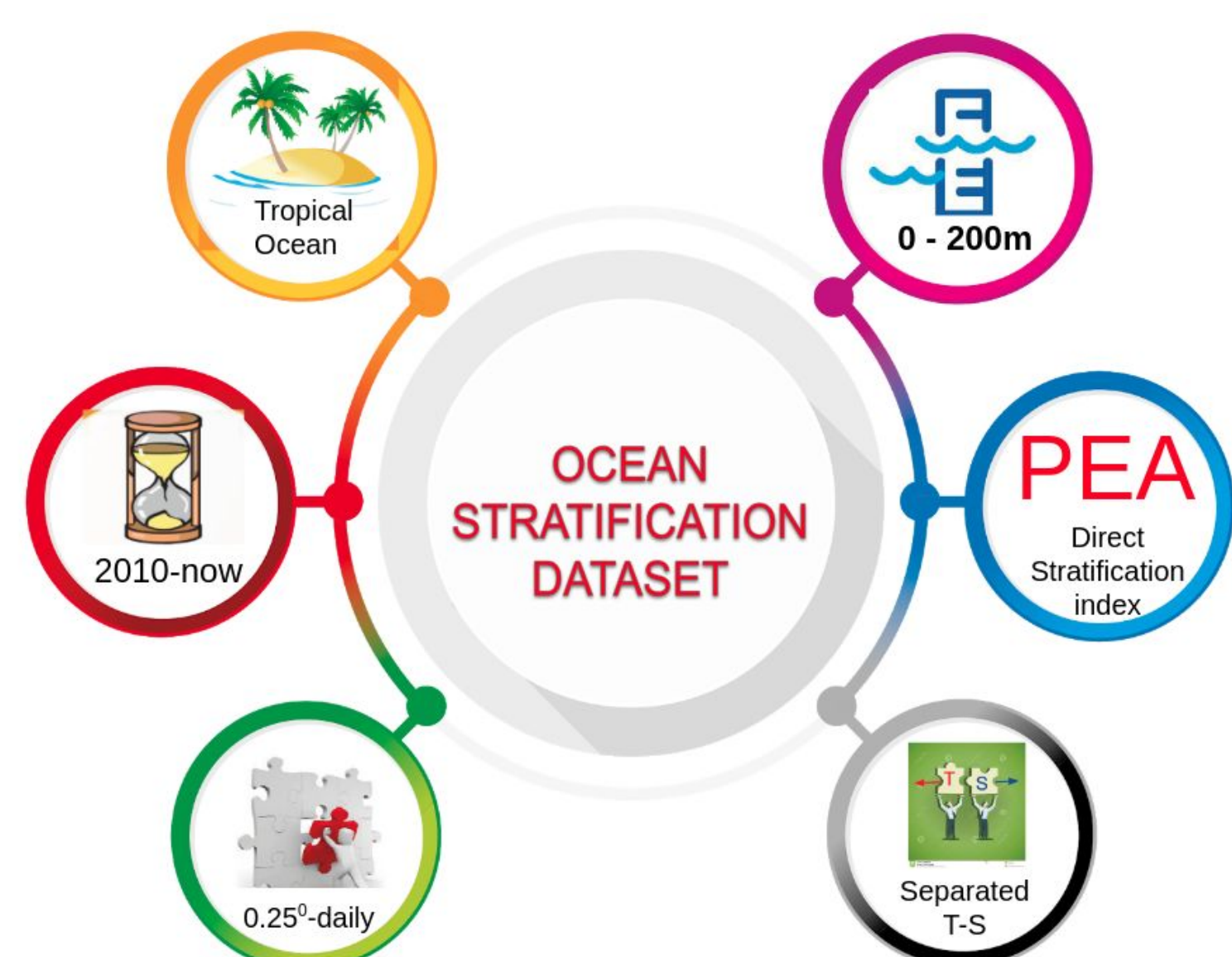
Ocean vertical temperature distribution

- stronger stratification associated with shallower thermocline -> cooler temperature near surface -> mixing cause **stronger SST cooling** -> TC weaken

- While stratification due to temperature can increase or decrease TC-induced SST cooling, stratification due to salinity can only reduce cooling. There is growing evidence on the significant role of salinity to TC rapid intensifications.

## Motivation:

- Previous stratification datasets based purely on in-situ observations have low spatio temporal resolutions and used indirect stratification indexes (Mixed Layer Depth, Ocean Heat Content ...)
- Respective contributions of temperature and salinity to ocean stratification were not explored

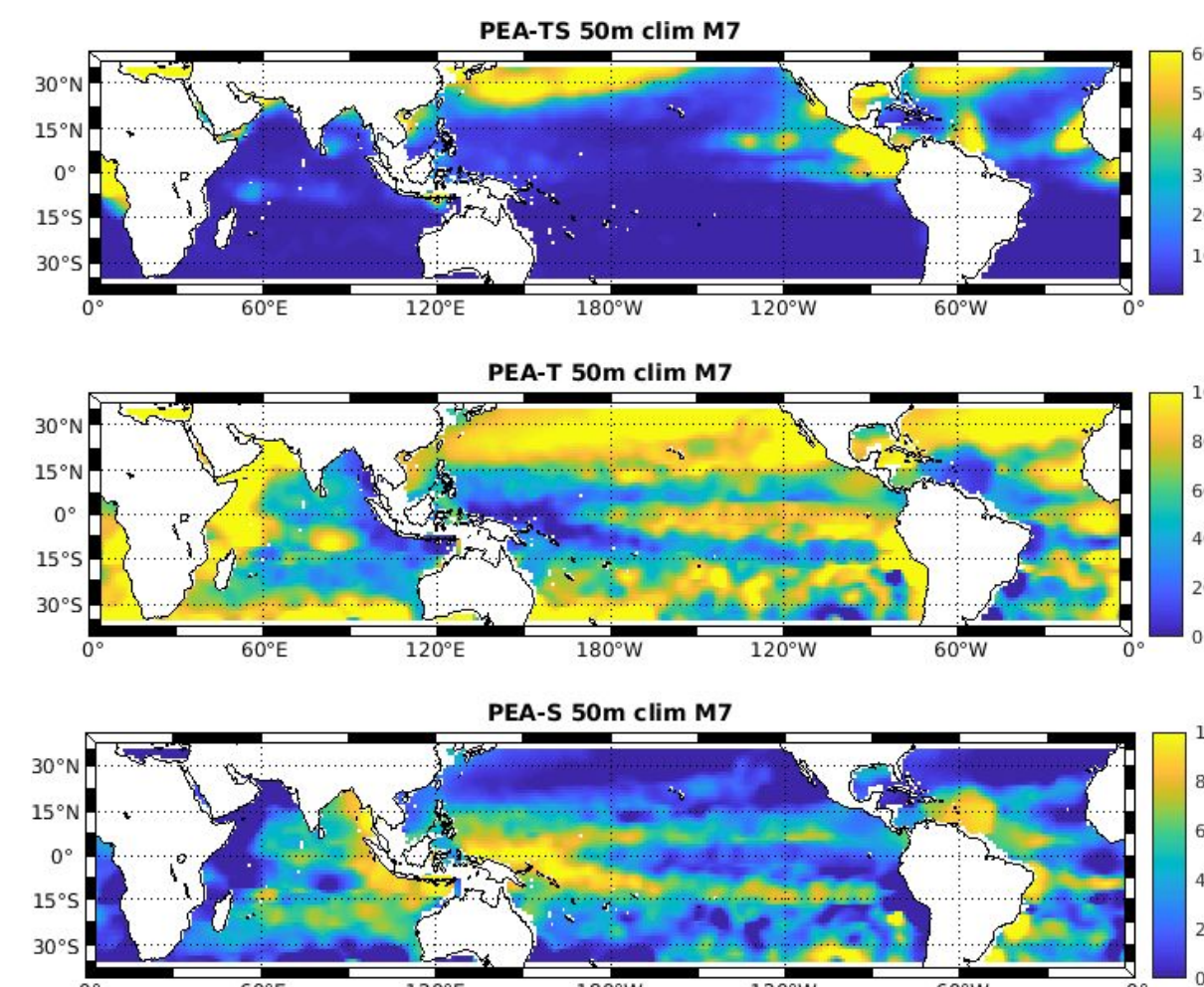


## II. Method

### 2.1 Stratification index

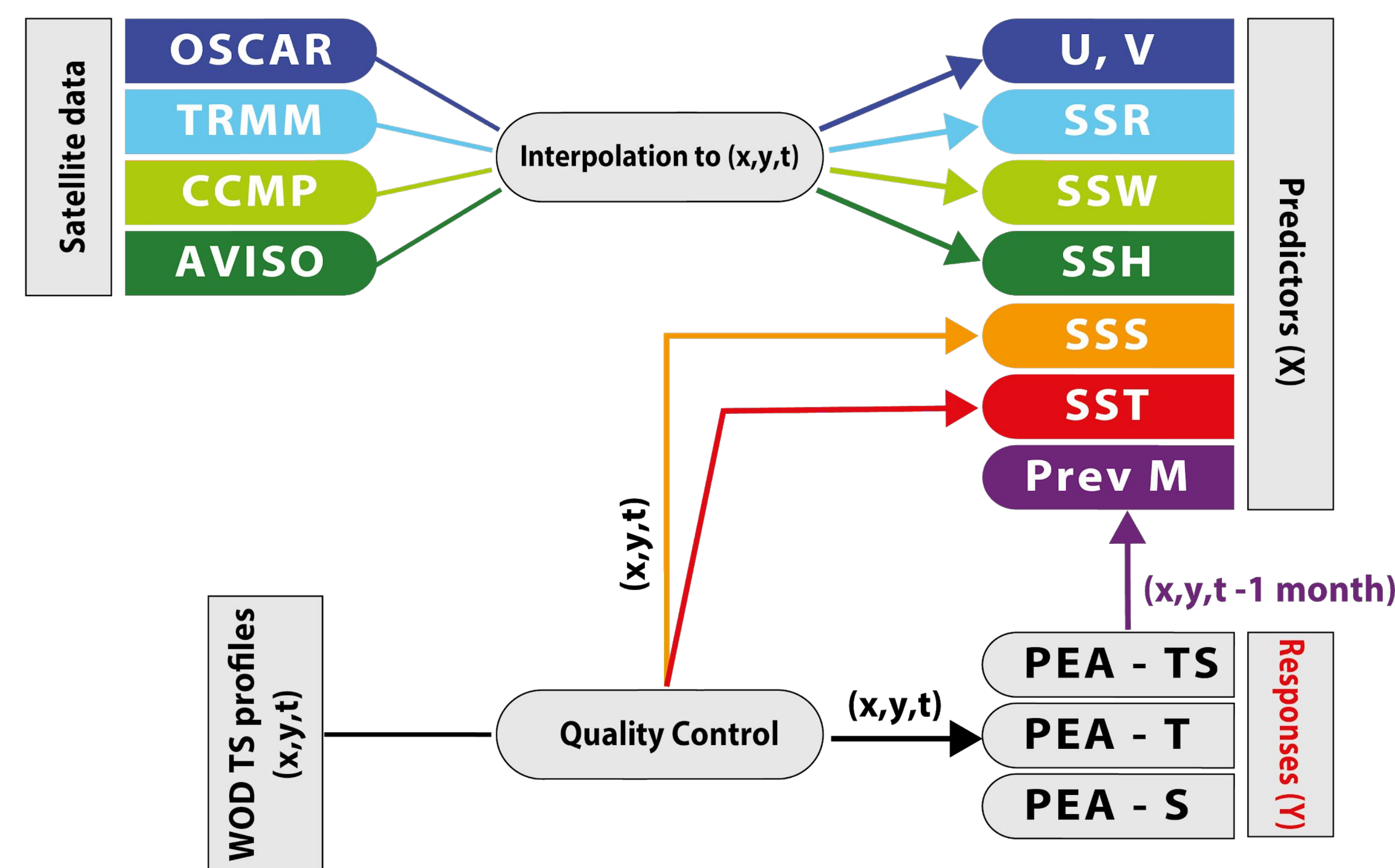
#### Integrated Potential Energy Anomaly

(PEA) is the total energy (J/m<sup>2</sup>) required to mix the whole water column from surface to a certain depth (h).

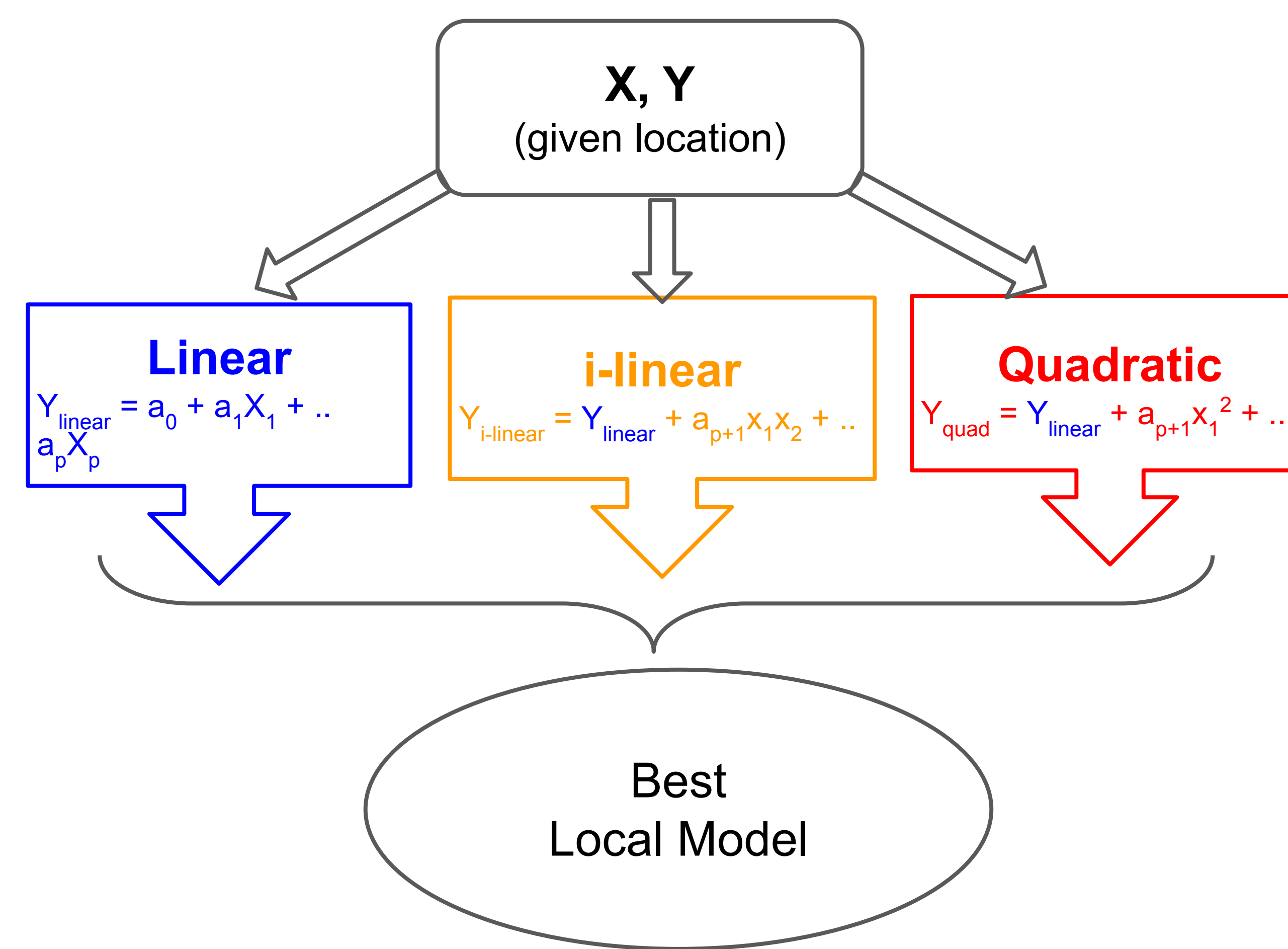


July climatology of total stratification and relative contribution of temperature, and salinity stratification (derived from WOD18)

## 2.2 Data Preprocessing



## 2.3 Building statistical model

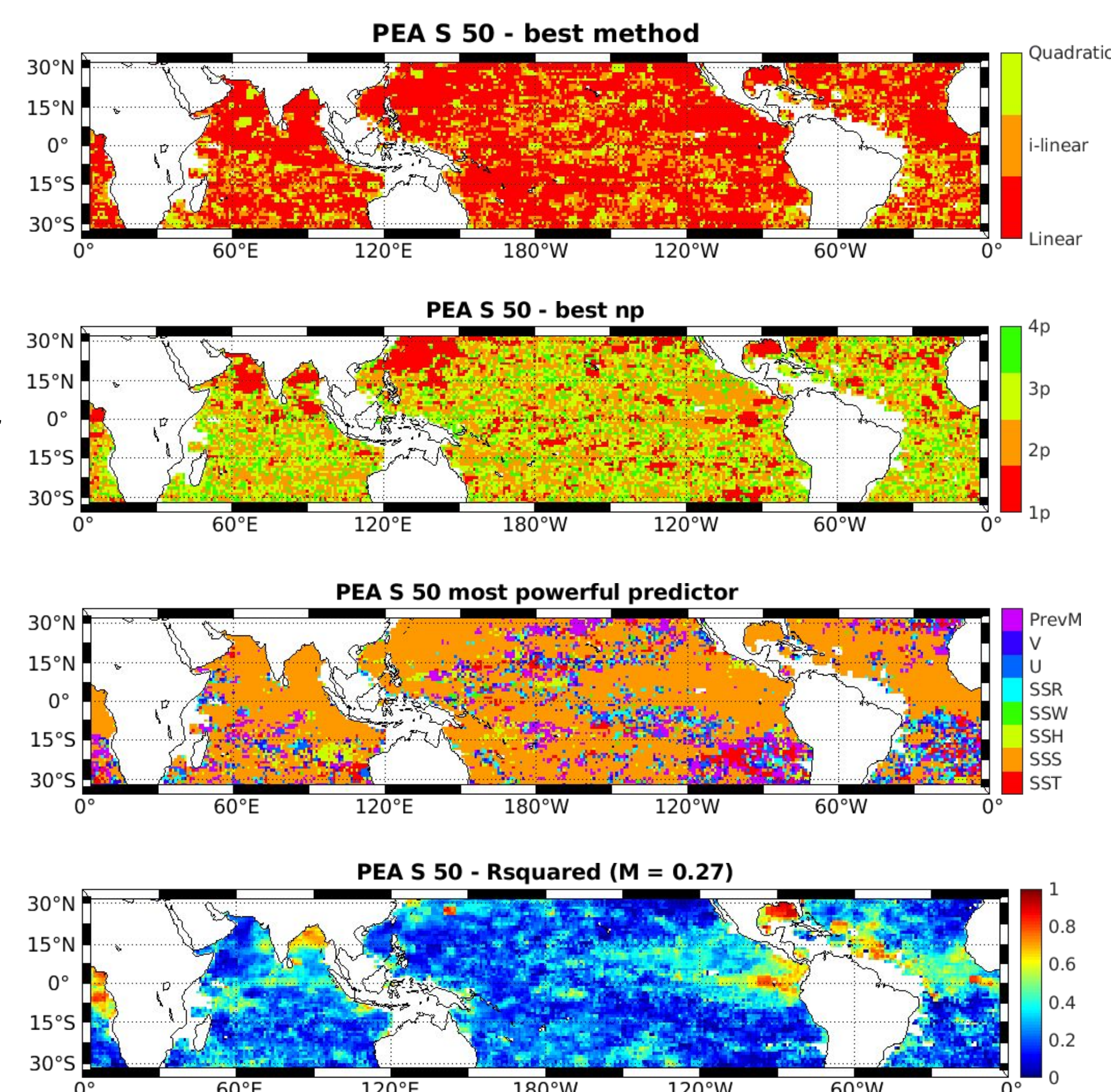


## III. Results

### 3.1 Best models

#### a) PEA-S 50m:

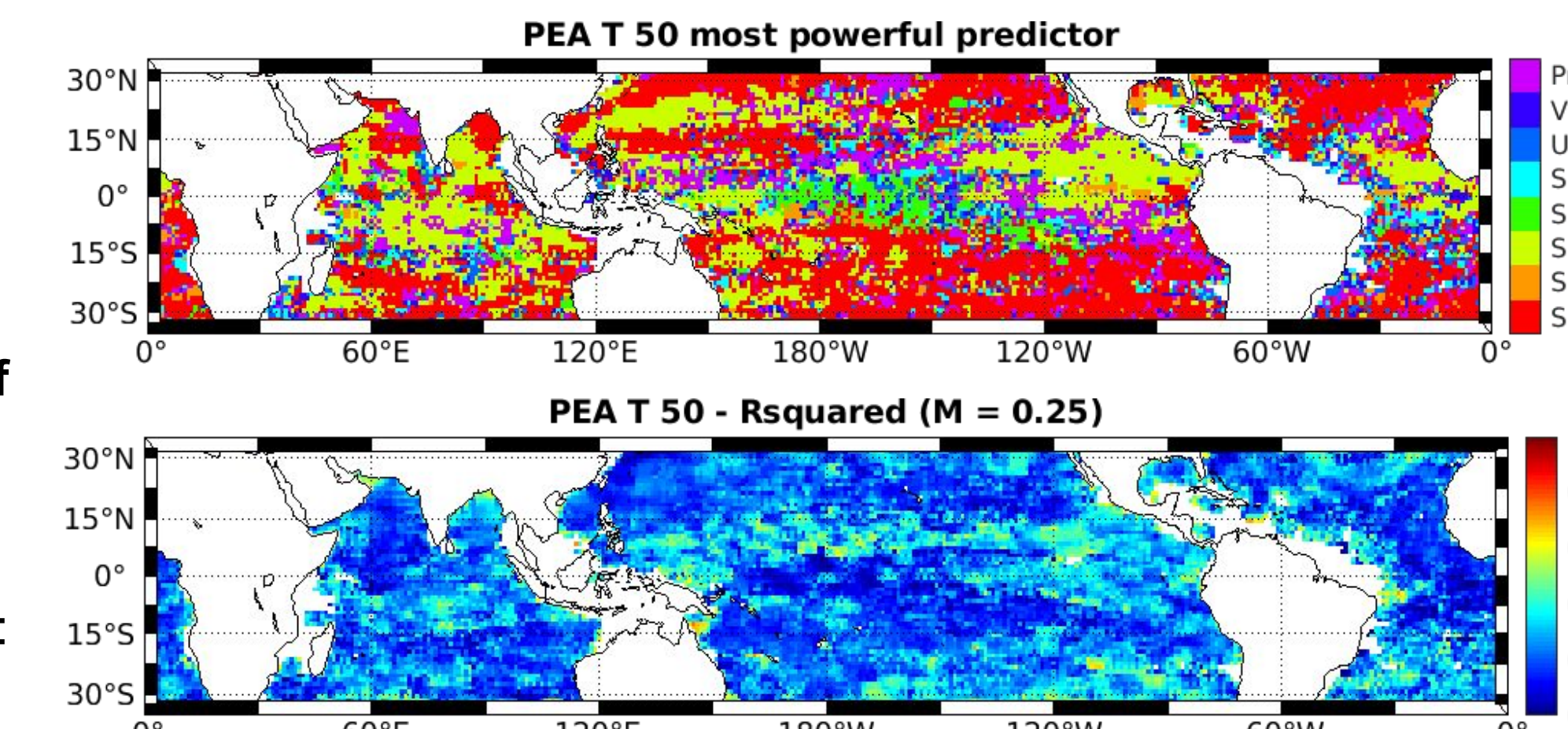
- Best method:** Simple linear regression
- Best number of predictor (np):** 1p - 3p
- Most powerful predictor:** SSS, PrevM
- Global mean R<sup>2</sup>: 0.27



Best configuration of regression model for PEA-S 50m and respective predictive performance (R-squared).

#### b) PEA-T 50m:

- Best method:** Simple linear regression
- Best number of predictor (np):** 1p - 3p
- Most powerful predictor:** SST, SSH, PrevM.
- Global mean R<sup>2</sup>: 0.25

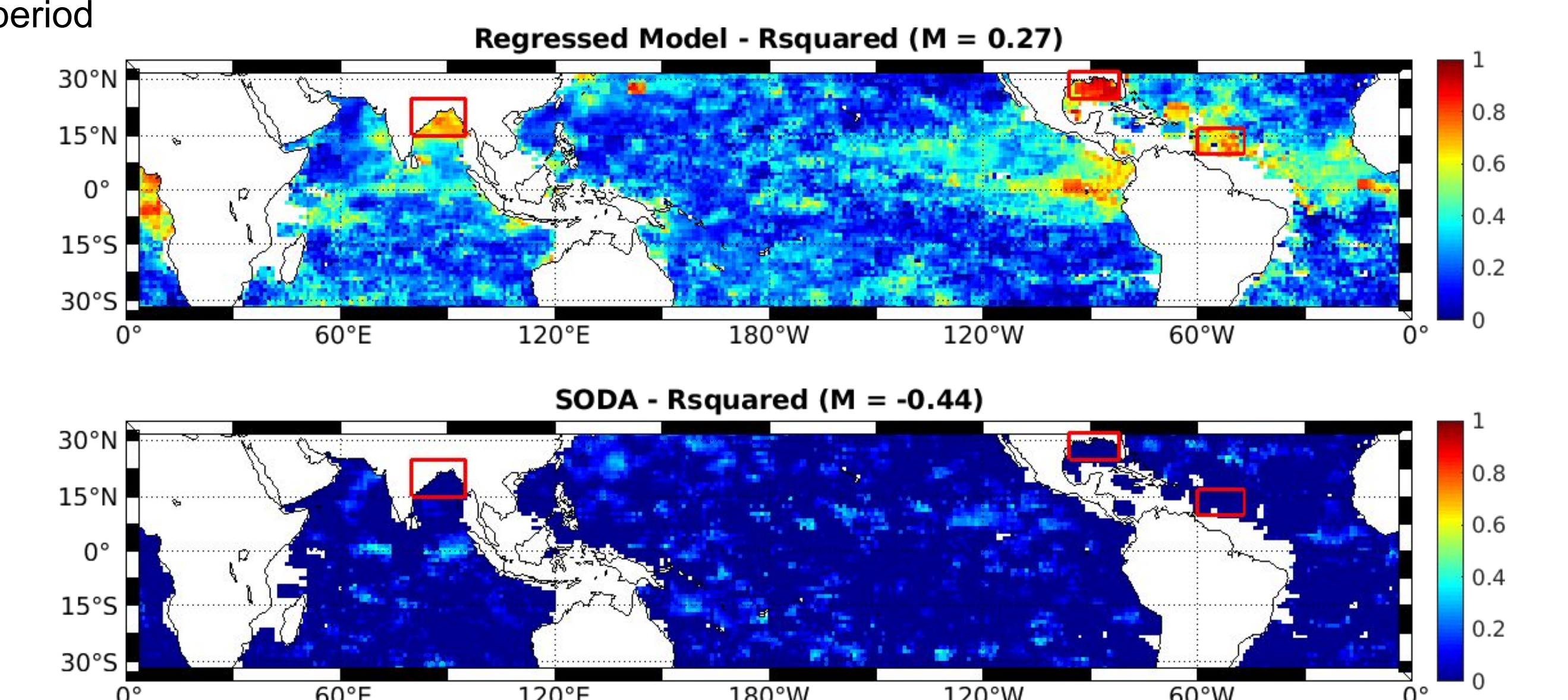


Best configuration of regression model for PEA-T 50m and respective predictive performance (R-squared). The best method and number of predictors map are similar to PEA-S 50m and not shown.

## 3.2 Regression Model vs. SODA

#### a) SODA data sampling

- SODA data at 0.5 deg, 5-day resolution is interpolated to WOD profiles locations for 2004-2016 period

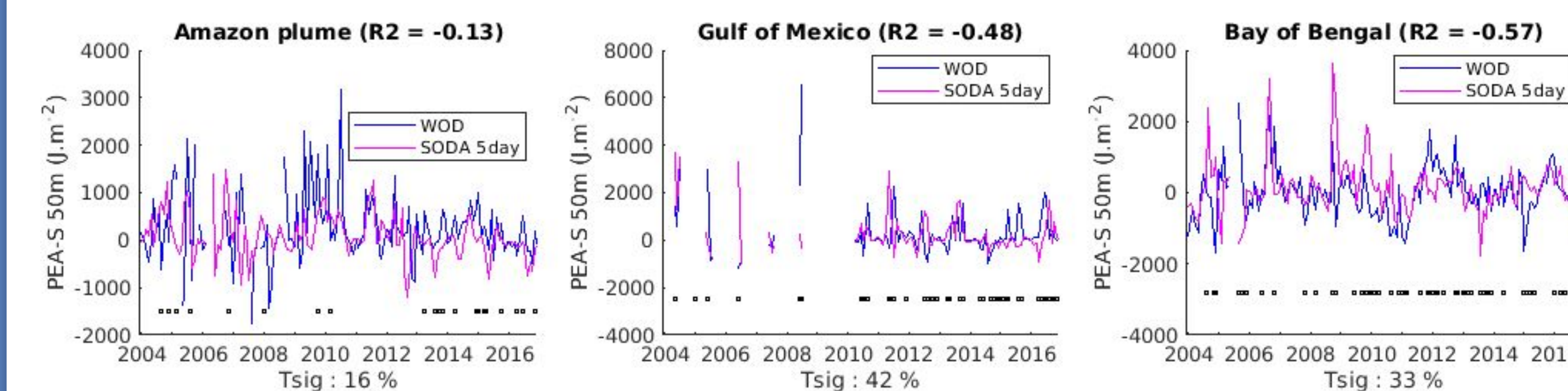


R-squared computed between PEA-S 50m of Regressed model predictions and SODA reanalysis vs. WOD18 data. Red boxes shows the locations where regressed model outperform SODA, from left to right: Bay of Bengal, Gulf of Mexico and Amazon Plume

#### b) Performance comparison

- Regressed models have better performance especially over areas of low SSS such as Bay of Bengal, Gulf of Mexico, Eastern Pacific
- This performance holds for 100m layer as well as for PEA-T in all layers (figure not shown)

#### c) Reasons for poor performance of SODA taking the red boxes as example



Monthly mean of all individual PEA-S 50m anomalies within the red boxes (solid lines). Black dots represents the months where the mean values are significantly different between SODA and WOD (using Student t-test). Tsig indicates percentage of black dots over the whole period.

- The monthly mean of PEA-S 50m between SODA and WOD are out of phase.
- Partly because WOD data contains high frequencies whereas SODA has 5-day average as well as due to the representativeness of number of profiles in the areas
- About 20-40% of the time SODA significantly failed to reproduce the monthly mean anomalies of PEA-S

## IV. Conclusions

- Globally, regression model can predict 25-30% of PEA-T and PEA-S within 50m, and up to 40% within 100m layer (figure not shown).
- The predictability of PEA-S is particularly high (R<sup>2</sup> = 0.7 - 0.9) in some areas (Bay of Bengal, Gulf of Mexico etc.)
- Statistical model outperform SODA

## V. Future Work

- Comparison with other high resolution reanalysis (HYCOM)
- High resolution stratification dataset based on satellite data
- Using more sophisticated machine learning methods: Neural Network, Random Forest
- Application for improving TC's intensity forecast