



# In Situ database Analyses Report

Sea mammals

*prepared by the Pi-MEP Consortium*

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## Acronym

<b>Aquarius</b>	NASA/CONAE Salinity mission
ASCAT	Advanced Scatterometer
ATBD	Algorithm Theoretical Baseline Document
BLT	Barrier Layer Thickness
<b>CMORPH</b>	CPC MORPHing technique (precipitation analyses)
<b>CPC</b>	Climate Prediction Center
CTD	Instrument used to measure the conductivity, temperature, and pressure of seawater
DM	Delayed Mode
EO	Earth Observation
<b>ESA</b>	European Space Agency
FTP	File Transfer Protocol
<b>GOSUD</b>	Global Ocean Surface Underway Data
<b>GTMBA</b>	The Global Tropical Moored Buoy Array
<b>Ifremer</b>	Institut français de recherche pour l'exploitation de la mer
<b>IPEV</b>	Institut polaire français Paul-Émile Victor
IQR	Interquartile range
ISAS	<i>In Situ</i> Analysis System
Kurt	Kurtosis (fourth central moment divided by fourth power of the standard deviation)
L2	Level 2
<b>LEGOS</b>	Laboratoire d'Etudes en Géophysique et Océanographie Spatiales
<b>LOCEAN</b>	Laboratoire d'Océanographie et du Climat : Expérimentations et Approches Numériques
<b>LOPS</b>	Laboratoire d'Océanographie Physique et Spatiale
MDB	Match-up Data Base
<b>MEOP</b>	Marine Mammals Exploring the Oceans Pole to Pole
MLD	Mixed Layer Depth
<b>NCEI</b>	National Centers for Environmental Information
NRT	Near Real Time
<b>NTAS</b>	Northwest Tropical Atlantic Station
OI	Optimal interpolation
<b>Pi-MEP</b>	Pilot-Mission Exploitation Platform
<b>PIRATA</b>	Prediction and Researched Moored Array in the Atlantic
QC	Quality control
$R_{sat}$	Spatial resolution of the satellite SSS product
<b>RAMA</b>	Research Moored Array for African-Asian-Australian Monsoon Analysis and Prediction
$r^2$	Square of the Pearson correlation coefficient
RMS	Root mean square
RR	Rain rate
<b>SAMOS</b>	Shipboard Automated Meteorological and Oceanographic System
Skew	Skewness (third central moment divided by the cube of the standard deviation)
<b>SMAP</b>	Soil Moisture Active Passive (NASA mission)
<b>SMOS</b>	Soil Moisture and Ocean Salinity (ESA mission)
<b>SPURS</b>	Salinity Processes in the Upper Ocean Regional Study
SSS	Sea Surface Salinity
$SSS_{in situ}$	<i>In situ</i> SSS data considered for the match-up

$SSS_{SAT}$	Satellite SSS product considered for the match-up
$\Delta SSS$	Difference between satellite and <i>in situ</i> SSS at colocalized point ( $\Delta SSS = SSS_{SAT} - SSS_{in\,situ}$ )
SST	Sea Surface Temperature
Std	Standard deviation
Std*	Robust Standard deviation = $\text{median}(\text{abs}(x - \text{median}(x))) / 0.67$ (less affected by outliers than Std)
Stratus	Surface buoy located in the eastern tropical Pacific
Survostral	SURVeillance de l'Océan AuSTRAL (Monitoring the Southern Ocean)
TAO	Tropical Atmosphere Ocean
TSG	ThermoSalinoGraph
WHOI	Woods Hole Oceanographic Institution
WHOTS	WHOI Hawaii Ocean Time-series Station
WOA	World Ocean Atlas

# 1 Overview

This report presents some characteristics of the Sea mammals in situ dataset used by the Pi-MEP to validate SMOS, SMAP and Aquarius satellite SSS products. A series of plots is proposed showing:

- Number of SSS data as a function of time and distance to coast
- Histogram of shallowest salinity and pressure (if relevant)
- Temporal mean of shallowest salinity pressure measurements (if relevant)
- Spatial density of shallowest salinity
- Spatial Maps of the Time-mean and temporal Std of *in situ* and satellite SSS and of the  $\Delta$ SSS
- Time series of the monthly median and Std of *in situ* and satellite SSS and of the  $\Delta$ SSS
- Zonal mean and Std of *in situ* and satellite SSS and of the  $\Delta$ SSS
- Scatterplots of ISAS vs *in situ* SSS by latitudinal bands
- Time series of the monthly median and Std of the  $\Delta$ SSS sorted by latitudinal bands
- $\Delta$ SSS sorted as function of geophysical parameters
- $\Delta$ SSS maps and statistics for different geophysical conditions

## 1.1 In situ dataset

### 1.1.1 Sea mammals

Instrumentation of southern elephant seals with satellite-linked CTD tags proposes unique temporal and spatial coverage. This includes extensive data from the Antarctic continental slope and shelf regions during the winter months, which is outside the conventional areas of Argo autonomous floats and ship-based studies. The use of elephant seals has been particularly effective to sample the Southern Ocean and the North Pacific. Other seal species have been successfully used in the North Atlantic, such as hooded seals. The marine mammal dataset ([MEOP-CTD database](#)) is quality controlled and calibrated using delayed-mode techniques involving comparisons with other existing profiles as well as cross-comparisons similar to established protocols within the Argo community, with a resulting accuracy of  $\pm 0.03$  °C in temperature and  $\pm 0.05$  in salinity or better ([Treasure et al. \(2017\)](#)). The marine mammal data were collected and made freely available by the International MEOP Consortium and the national programs that contribute to it (<http://www.meop.net>). This dataset is updated once a year and can be downloaded [here](#) ([Roquet et al. \(2018\)](#)). A preprocessing stage is applied to the database before being used by the Pi-MEP which consist to keep only profile with salinity, temperature and pressure quality flags set to 1 or 2 and if at least one measurement is in the top 10 m depth. Marine mammal SSS correspond to the top (shallowest) profile salinity data provided that profile depth is 10 m or less.

## 1.2 Auxiliary geophysical datasets

Additional EO datasets are used to characterize the geophysical conditions at the *in situ* measurement locations and time, and 10 days prior the measurements to get an estimate of the geophysical condition and history. As discussed in [Boutin et al. \(2016\)](#), the presence of vertical gradients in, and horizontal variability of, sea surface salinity indeed complicates comparison of satellite and *in situ* measurements. The additional EO data are used here to get a first estimates of conditions for which L-band satellite SSS measured in the first centimeters of the upper ocean within a 50-150 km diameter footprint might differ from pointwise *in situ* measurements performed in general between 10 and 5 m depth below the surface. The spatio-temporal variability of SSS within a satellite footprint (50-150 km) is a major issue for satellite SSS validation in the vicinity of river plumes, frontal zones, and significant precipitation. Rainfall can in some cases produce vertical salinity gradients exceeding  $1 \text{ pss m}^{-1}$ ; consequently, it is recommended that satellite and *in situ* SSS measurements less than 3–6 h after rain events should be considered with care when used in satellite calibration/validation analyses. To identify such situation, the Pi-MEP platform is first using [CMORPH](#) products to characterize the local value and history of rain rate and [ASCAT](#) gridded data are used to characterize the local surface wind speed and history. For validation purpose, the [ISAS](#) monthly SSS *in situ* analysed fields at 5 m depth are collocated and compared with the *in situ* SSS value. The use of ISAS is motivated by the fact that it is used in the SMOS L2 official validation protocol in which systematic comparisons of SMOS L2 retrieved SSS with ISAS are done. In complement to ISAS, annual std climatological field from the World Ocean Atlas ([WOA13](#)) at the *in situ* location are also used to have an a priori information of the local SSS variability.

### 1.2.1 CMORPH

Precipitation are estimated using the [CMORPH](#) 3-hourly products at  $1/4^\circ$  resolution ([Joyce et al. \(2004\)](#)). CMORPH (CPC MORPHing technique) produces global precipitation analyses at very high spatial and temporal resolution. This technique uses precipitation estimates that have been derived from low orbiter satellite microwave observations exclusively, and whose features are transported via spatial propagation information that is obtained entirely from geostationary satellite IR data. At present NOAA incorporate precipitation estimates derived from the passive microwaves aboard the DMSP 13, 14 and 15 (SSM/I), the NOAA-15, 16, 17 and 18 (AMSU-B), and AMSR-E and TMI aboard NASA's Aqua, TRMM and GPM spacecraft, respectively. These estimates are generated by algorithms of [Ferraro \(1997\)](#) for SSM/I, [Ferraro et al. \(2000\)](#) for AMSU-B and [Kummerow et al. \(2001\)](#) for TMI. Note that this technique is not a precipitation estimation algorithm but a means by which estimates from existing microwave rainfall algorithms can be combined. Therefore, this method is extremely flexible such that any precipitation estimates from any microwave satellite source can be incorporated.

With regard to spatial resolution, although the precipitation estimates are available on a grid with a spacing of 8 km (at the equator), the resolution of the individual satellite-derived estimates is coarser than that - more on the order of  $12 \times 15 \text{ km}$  or so. The finer "resolution" is obtained via interpolation.

In effect, IR data are used as a means to transport the microwave-derived precipitation features during periods when microwave data are not available at a location. Propagation vector matrices are produced by computing spatial lag correlations on successive images of geostationary satellite IR which are then used to propagate the microwave derived precipitation estimates. This process governs the movement of the precipitation features only. At a given location, the shape and intensity of the precipitation features in the intervening half hour periods between microwave scans are determined by performing a time-weighting interpolation between microwave-derived

features that have been propagated forward in time from the previous microwave observation and those that have been propagated backward in time from the following microwave scan. NOAA refer to this latter step as "morphing" of the features.

For the present Pi-MEP products, we only considered the 3-hourly products at 1/4 degree resolution. The entire CMORPH record (December 2002-present) for 3-hourly, 1/4 degree lat/lon resolution can be found at: [ftp://ftp.cpc.ncep.noaa.gov/precip/CMORPH\\_V1.0/CRT/](ftp://ftp.cpc.ncep.noaa.gov/precip/CMORPH_V1.0/CRT/). CMORPH estimates cover a global belt (-180°W to 180°E) extending from 60°S to 60°N latitude and are available for the complete period of the Pi-MEP core datasets (Jan 2010-now).

### 1.2.2 ASCAT

Advanced SCATterometer (ASCAT) daily data produced and made available at [Ifremer/CERSAT](http://ifremer.cersat.fr/) on a 0.25°x0.25° resolution grid ([Bentamy and Fillon \(2012\)](#)) since March 2007 are used to characterize the mean daily wind at the match-up pair location as well as the wind history during the 10-days period preceding the in situ measurement date. These wind fields are calculated based on a geostatistical method with external drift. Remotely sensed data from ASCAT are considered as observations while those from numerical model analysis (ECMWF) are associated with the external drift. The spatial and temporal structure functions for wind speed, zonal and meridional wind components are estimated from ASCAT retrievals. Furthermore, the new procedure includes a temporal interpolation of the retrievals based on the complex empirical orthogonal function (CEOF) approach, in order to enhance the sampling length of the scatterometer observations. The resulting daily wind fields involves the main known surface wind patterns as well as some variation modes associated with temporal and spatial moving features. The accuracy of the gridded winds was investigated through comparisons with moored buoy data in [Bentamy et al. \(2012\)](#) and resulted in rms differences for wind speed and direction are about 1.50 m.s<sup>-1</sup> and 20°.

### 1.2.3 ISAS

The In Situ Analysis System (ISAS), as described in [Gaillard et al. \(2016\)](#) is a data based re-analysis of temperature and salinity fields over the global ocean. It was initially designed to synthesize the temperature and salinity profiles collected by the Argo program. It has been later extended to accommodate all type of vertical profile as well as time series. ISAS gridded fields are entirely based on *in situ* measurements. The methodology and configuration have been conceived to preserve as much as possible the data information content and resolution. ISAS is developed and run in a research laboratory (LOPS) in close collaboration with Coriolis, one of Argo Global Data Assembly Center and unique data provider for the Mercator operational oceanography system. At the moment the period covered starts in 2002 and only the upper 2000 m are considered. The gridded fields were produced over the global ocean 70°N–70°S on a 1/2° grid by the ISAS project with datasets downloaded from the Coriolis data center (for more details on ISAS see [Gaillard et al. \(2009\)](#)). In the Pi-MEP, the product in used is the [INSITU\\_GLO\\_TS\\_OA\\_NRT\\_OBSERVATIONS.013.002.a](#) v6.2 NRT derived at the Coriolis data center and provided by the Copernicus Marine Environment Monitoring Service (CMEMS). The major contribution to the data set is from Argo array of profiling floats, reaching an approximate resolution of one profile every 10-days and every 3-degrees over the satellite SSS period (<http://www.umr-lops.fr/SNO-Argo/Products/ISAS-T-S-fields/>); in this version SSS from ship of opportunity thermosalinographs are not used, so that we can consider SMOS SSS validation using these measurements independent of ISAS. The ISAS optimal interpolation involves a structure function modeled as the sum of two Gaussian functions, each associated with specific time and space scales, resulting in a smoothing over typically 3 degrees. The smallest scale which can

be retrieved with ISAS analysis is not smaller than 300–500 km (Kolodziejczyk et al. (2015)). For validation purpose, the ISAS monthly SSS fields at 5 m depth are collocated and compared with the satellite SSS products and included in the Pi-MEP Match-up files. In addition, the "percentage of variance" fields (PCTVAR) contained in the ISAS analyses provide information on the local variability of *in situ* SSS measurements within  $1/2^\circ \times 1/2^\circ$  boxes.

#### 1.2.4 World Ocean Atlas Climatology

The World Ocean Atlas 2013 version 2 (WOA13 V2) is a set of objectively analyzed ( $1^\circ$  grid) climatological fields of *in situ* temperature, salinity and other variables provided at standard depth levels for annual, seasonal, and monthly compositing periods for the World Ocean. It also includes associated statistical fields of observed oceanographic profile data interpolated to standard depth levels on  $5^\circ$ ,  $1^\circ$ , and  $0.25^\circ$  grids. We use these fields in complement to ISAS to characterize the climatological fields (annual mean and std) at the match-up pairs location and date.

## 2 In Situ Database Analyses

### 2.1 Introduction

Instrumentation of southern elephant seals with satellite-linked CTD tags proposes unique temporal and spatial coverage. This includes extensive data from the Antarctic continental slope and shelf regions during the winter months, which is outside the conventional areas of Argo autonomous floats and ship-based studies. The use of elephant seals has been particularly effective to sample the Southern Ocean and the North Pacific. Other seal species have been successfully used in the North Atlantic, such as hooded seals. The marine mammal dataset (MEOP-CTD database) is quality controlled and calibrated using



delayed-mode techniques involving comparisons with other existing profiles as well as cross-comparisons similar to established protocols within the Argo community, with a resulting accuracy of  $\pm 0.03^\circ\text{C}$  in temperature and  $\pm 0.05$  in salinity or better (Treasure et al. (2017)). The marine mammal data were collected and made freely available by the International MEOP Consortium and the national programs that contribute to it (<http://www.meop.net>). This dataset is updated once a year and can be downloaded here (Roquet et al. (2018)). A preprocessing stage is applied to the database before being used by the Pi-MEP which consist to keep only profile with salinity, temperature and pressure quality flags set to 1 or 2 and if at least one measurement is in the top 10 m depth. Marine mammal SSS correspond to the top (shallowest) profile salinity data provided that profile depth is 10 m or less.

### 2.2 Number of SSS data as a function of time and distance to coast

Figure 1 shows the time (a) and distance to coast (b) distributions of the Sea mammals *in situ* dataset.



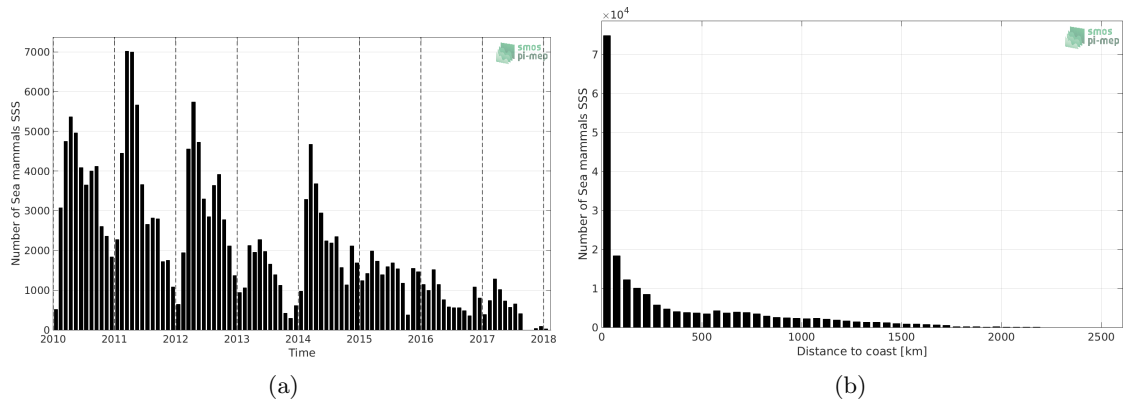


Figure 1: Number of SSS from Sea mammals as a function of time (a) and distance to coast (b).

### 2.3 Histograms of SSS

Figure 2 shows the SSS distribution of the Sea mammals (a) and colocalized ISAS (b) dataset.

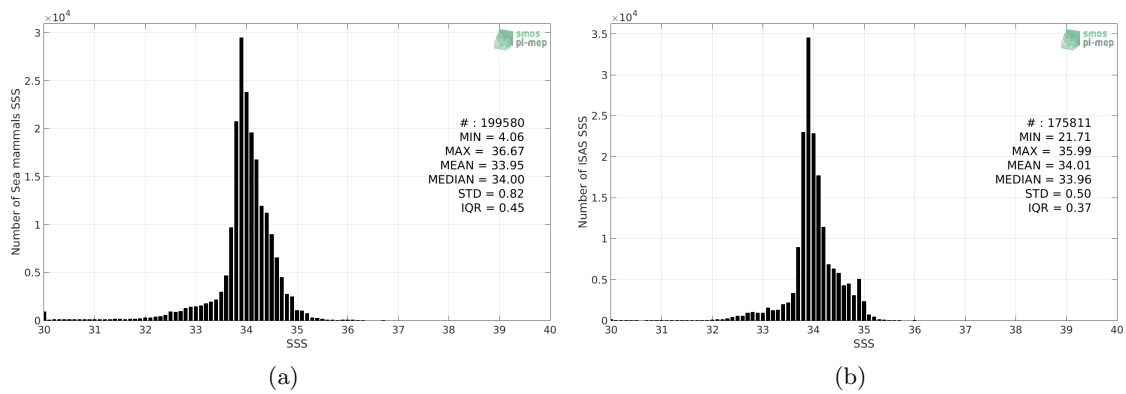


Figure 2: Histograms of SSS from Sea mammals (a) and ISAS (b) per bins of 0.1.

### 2.4 Distribution of *in situ* SSS depth measurements

In Figure 3, we show the depth distribution of the *in situ* salinity dataset (a) and the spatial distribution of the depth temporal mean in  $1^\circ \times 1^\circ$  boxes and considering the full *in situ* dataset period (b).

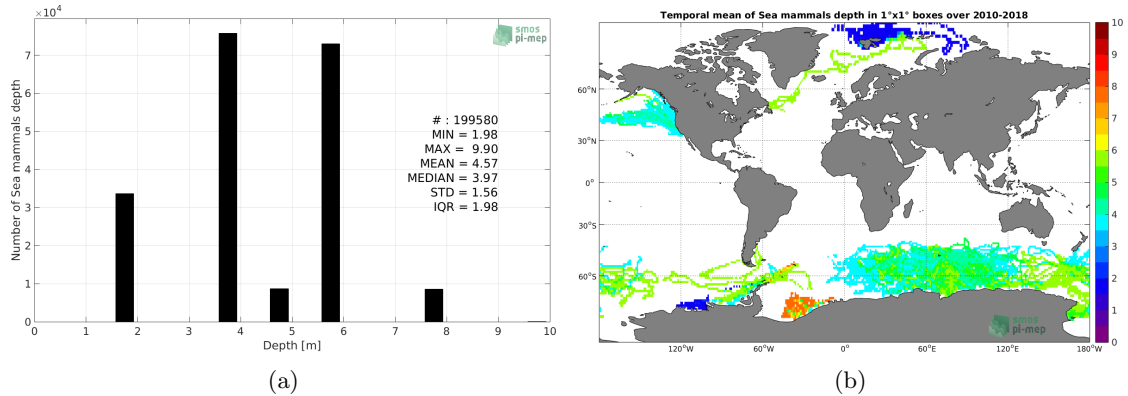


Figure 3: Depth distribution of the upper level SSS measurements from Sea mammals (a) and spatial distribution of the *in situ* SSS depth measurements showing the mean value in  $1^\circ \times 1^\circ$  boxes and considering the full *in situ* dataset period (b).

## 2.5 Spatial distribution of SSS

In Figure 4, the number of Sea mammals SSS measurements in  $1^\circ \times 1^\circ$  boxes is shown.

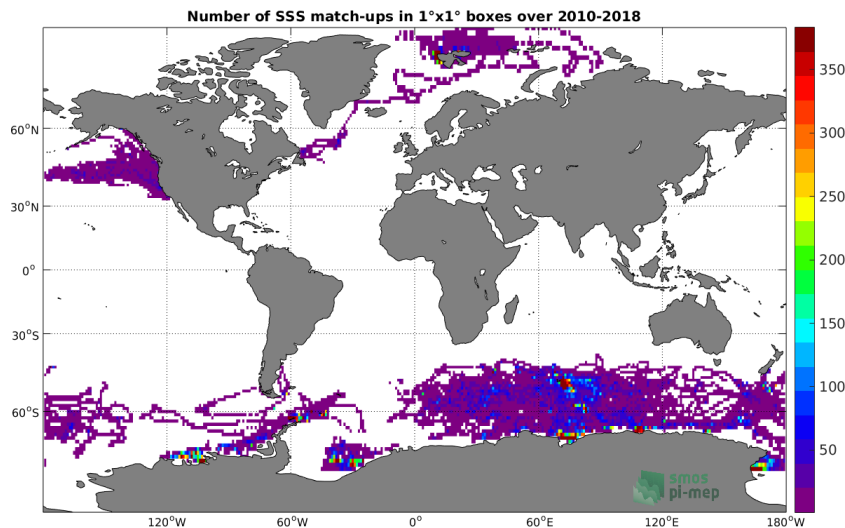


Figure 4: Number of SSS from Sea mammals in  $1^\circ \times 1^\circ$  boxes.

## 2.6 Spatial Maps of the Temporal mean and Std of *in situ* and ISAS SSS and of the difference ( $\Delta$ SSS)

In Figure 5, we show maps of temporal mean (left) and standard deviation (right) of ISAS (top), Sea mammals *in situ* dataset (middle) and the difference  $\Delta$ SSS (ISAS - Sea mammals) (bottom). The temporal mean and std are calculated using all match-up pairs falling in spatial boxes of size  $1^\circ \times 1^\circ$  over the full Sea mammals dataset period.

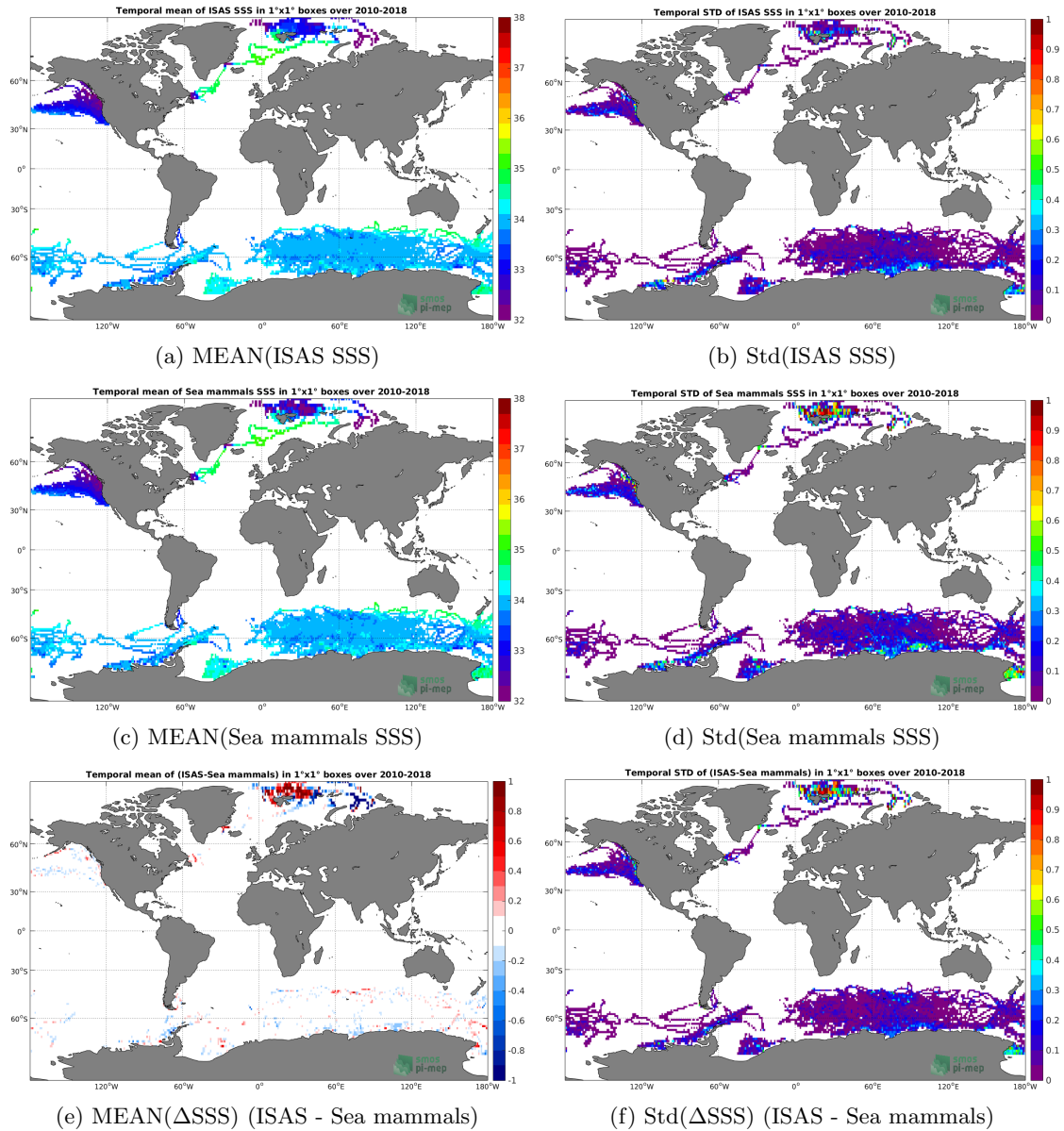


Figure 5: Temporal mean (left) and Std (right) of SSS from ISAS (top), Sea mammals (middle), and of  $\Delta$ SSS (ISAS - Sea mammals). Only match-up pairs are used to generate these maps.

## 2.7 Time series of the monthly median and Std of *in situ* and ISAS SSS and of the difference ( $\Delta$ SSS)

In the top panel of Figure 6, we show the time series of the monthly median SSS estimated for both ISAS SSS product (in black) and the Sea mammals *in situ* dataset (in blue) at the collected Pi-MEP match-up pairs.

In the middle panel of Figure 6, we show the time series of the monthly median of  $\Delta$ SSS (ISAS - Sea mammals) for the collected Pi-MEP match-up pairs.

In the bottom panel of Figure 6, we show the time series of the monthly standard deviation of the  $\Delta$ SSS (ISAS - Sea mammals) for the collected Pi-MEP match-up pairs.

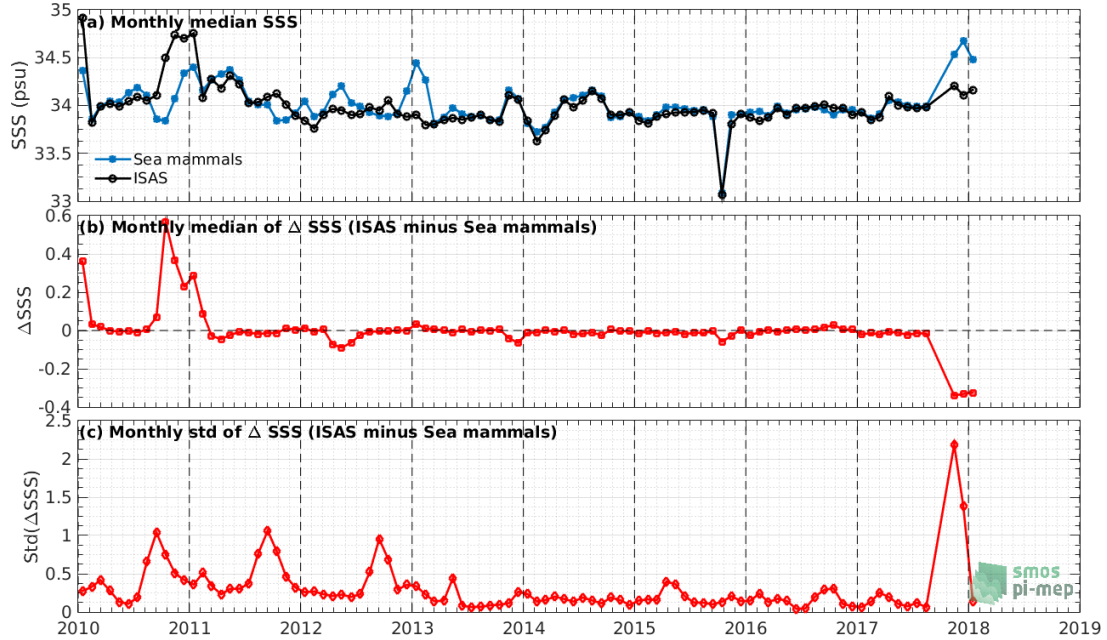


Figure 6: Time series of the monthly median SSS (top), median of  $\Delta$ SSS (ISAS - Sea mammals) and Std of  $\Delta$ SSS (ISAS - Sea mammals) considering all match-ups collected by the Pi-MEP.

## 2.8 Zonal mean and Std of *in situ* and ISAS SSS and of the difference $\Delta$ SSS

In Figure 7 left panel, we show the zonal mean SSS considering all Pi-MEP match-up pairs for both ISAS SSS product (in black) and the Sea mammals *in situ* dataset (in blue). The full *in situ* dataset period is used to derive the mean.

In the right panel of Figure 7, we show the zonal mean of  $\Delta$ SSS (ISAS - Sea mammals) for all the collected Pi-MEP match-up pairs estimated over the full *in situ* dataset period.

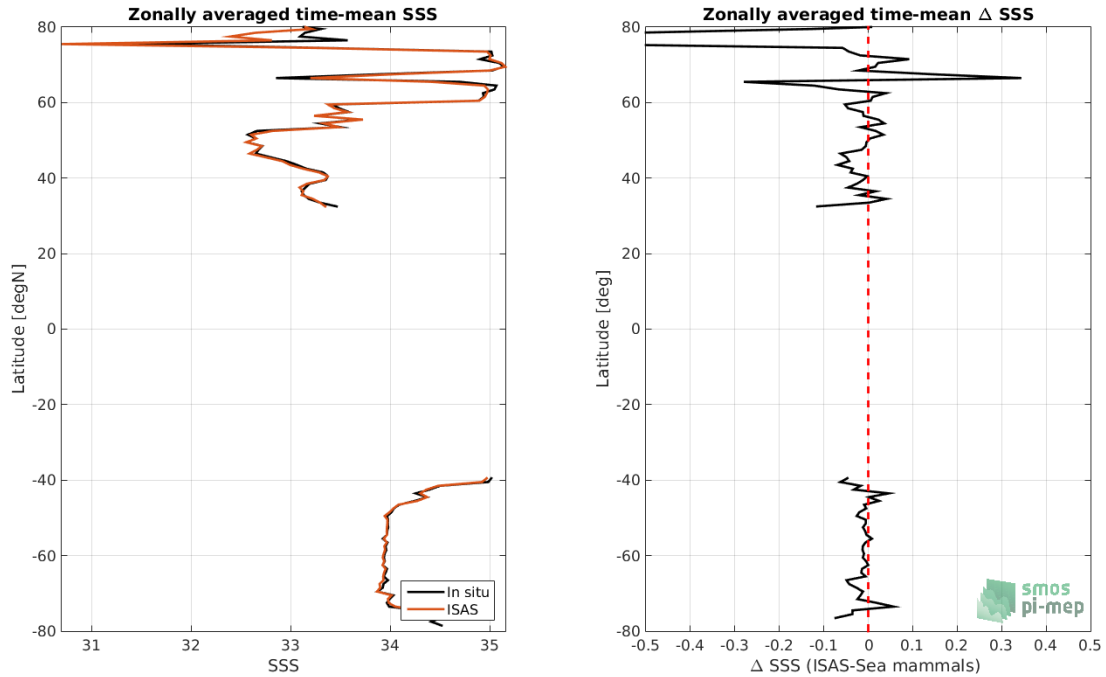


Figure 7: Left panel: Zonal mean SSS from ISAS product (black) and from Sea mammals (blue). Right panel: Zonal mean of  $\Delta$ SSS (ISAS - Sea mammals) for all the collected Pi-MEP match-up pairs estimated over the full *in situ* dataset period.

## 2.9 Scatterplots of ISAS vs *in situ* SSS by latitudinal bands

In Figure 8, contour maps of the concentration of ISAS SSS (y-axis) versus Sea mammals SSS (x-axis) at match-up pairs for different latitude bands: (a) 80°S-80°N, (b) 20°S-20°N, (c) 40°S-20°S and 20°N-40°N and (d) 60°S-40°S and 40°N-60°N. For each plot, the red line shows  $x=y$ . The black thin and dashed lines indicate a linear fit through the data cloud and the  $\pm 95\%$  confidence levels, respectively. The number match-up pairs  $n$ , the slope and  $R^2$  coefficient of the linear fit, the root mean square (RMS) and the mean bias between ISAS and *in situ* data are indicated for each latitude band in each plots.

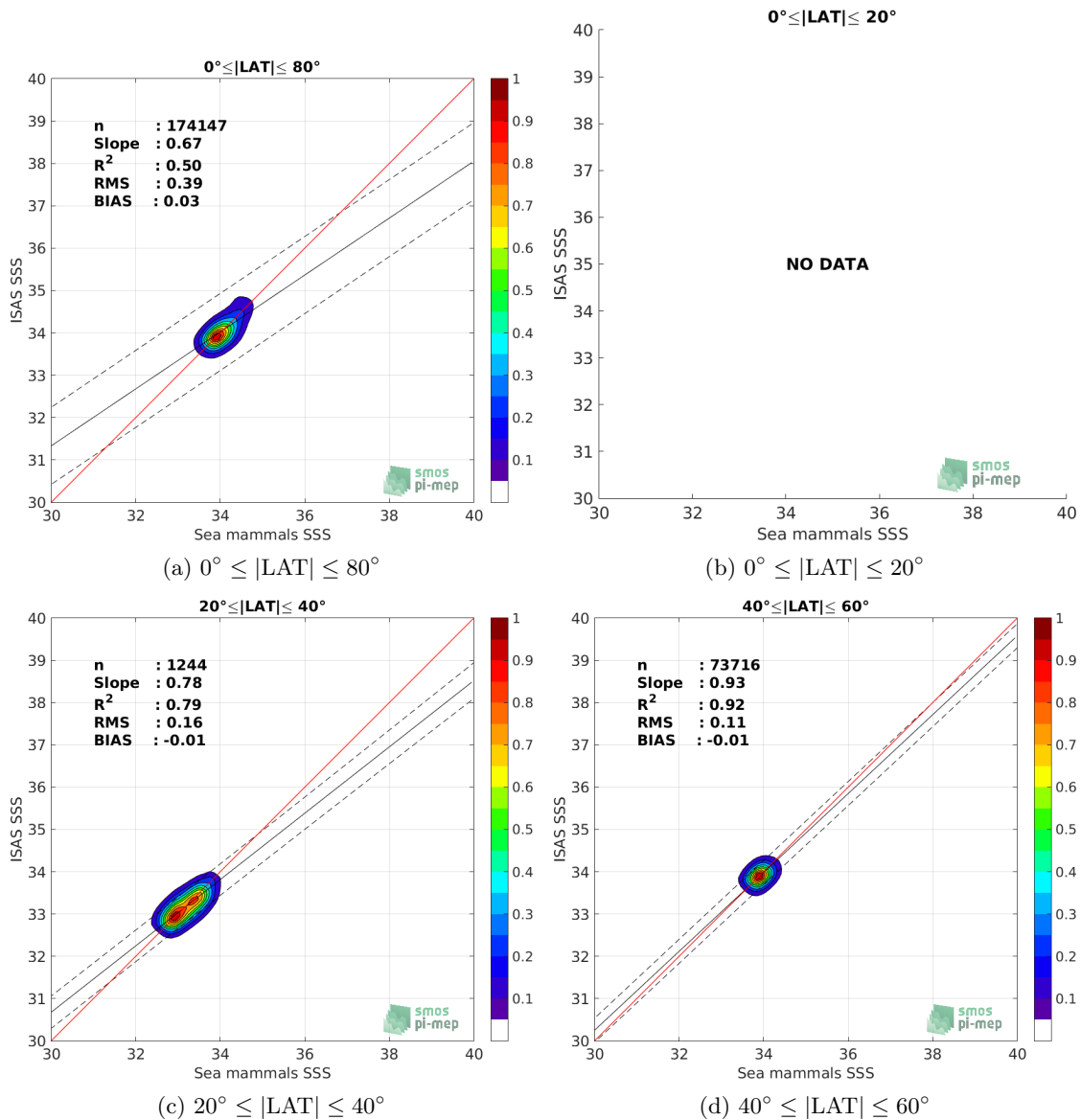


Figure 8: Contour maps of the concentration of ISAS SSS (y-axis) versus Sea mammals SSS (x-axis) at match-up pairs for different latitude bands. For each plot, the red line shows  $x=y$ . The black thin and dashed lines indicate a linear fit through the data cloud and the  $\pm 95\%$  confidence levels, respectively. The number match-up pairs  $n$ , the slope and  $R^2$  coefficient of the linear fit, the root mean square (RMS) and the mean bias between ISAS and *in situ* data are indicated for each latitude band in each plots.

## 2.10 Time series of the monthly median and Std of the difference $\Delta\text{SSS}$ sorted by latitudinal bands

In Figure 9, time series of the monthly median (red curves) of  $\Delta\text{SSS}$  (ISAS - Sea mammals) and  $\pm 1$  Std (black vertical thick bars) as function of time for all the collected Pi-MEP match-up

pairs estimated for the full *in situ* dataset period are shown for different latitude bands: (a) 80°S-80°N, (b) 20°S-20°N, (c) 40°S-20°S and 20°N-40°N and (d) 60°S-40°S and 40°N-60°N.

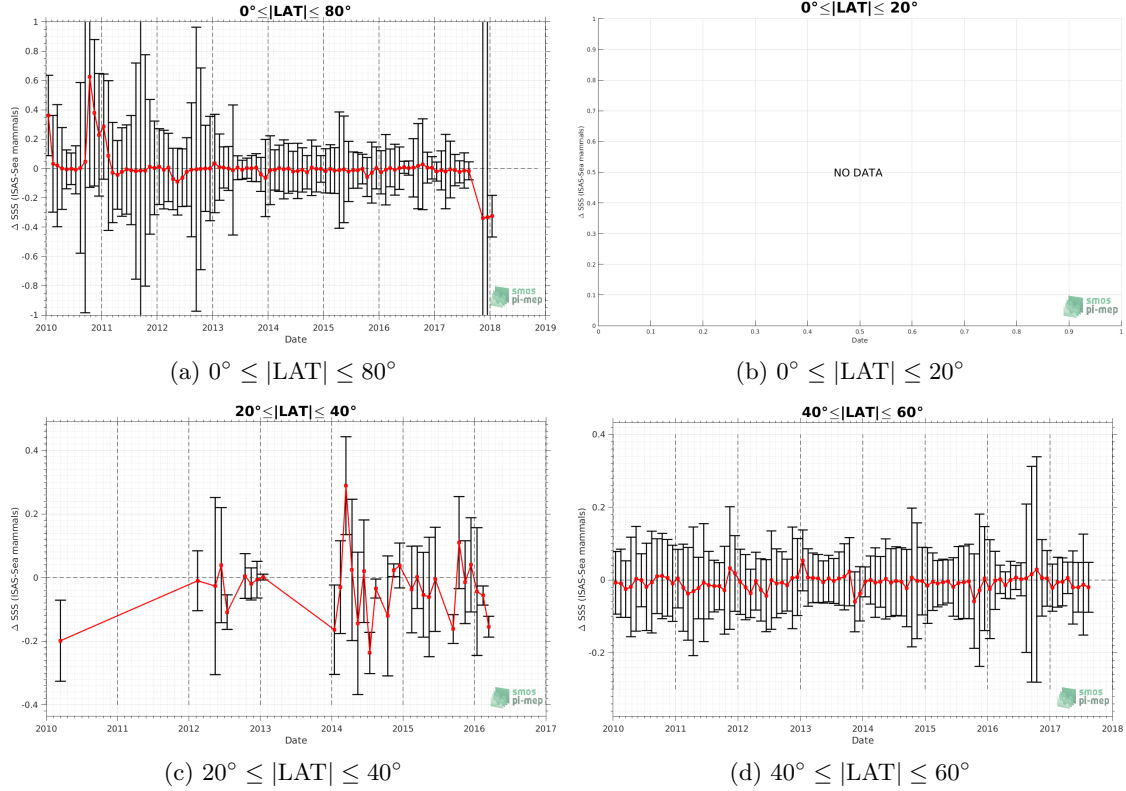


Figure 9: Monthly median (red curves) of  $\Delta\text{SSS}$  (ISAS - Sea mammals) and  $\pm 1$  Std (black vertical thick bars) as function of time for all the collected Pi-MEP match-up pairs for the full *in situ* dataset period are shown for different latitude bands: (a) 80°S-80°N, (b) 20°S-20°N, (c) 40°S-20°S and 20°N-40°N and (d) 60°S-40°S and 40°N-60°N.

## 2.11 $\Delta\text{SSS}$ sorted as geophysical conditions

In Figure 10, we classify the match-up differences  $\Delta\text{SSS}$  (ISAS - *in situ*) as function of the geophysical conditions at match-up points. The mean and std of  $\Delta\text{SSS}$  (ISAS - Sea mammals) is thus evaluated as function of the

- *in situ* SSS values per bins of width 0.2,
- *in situ* SST values per bins of width 1°C,
- ASCAT daily wind values per bins of width 1 m/s,
- CMORPH 3-hourly rain rates per bins of width 1 mm/h, and,
- distance to coasts per bins of width 50 km,
- *in situ* measurement depth (if relevant).

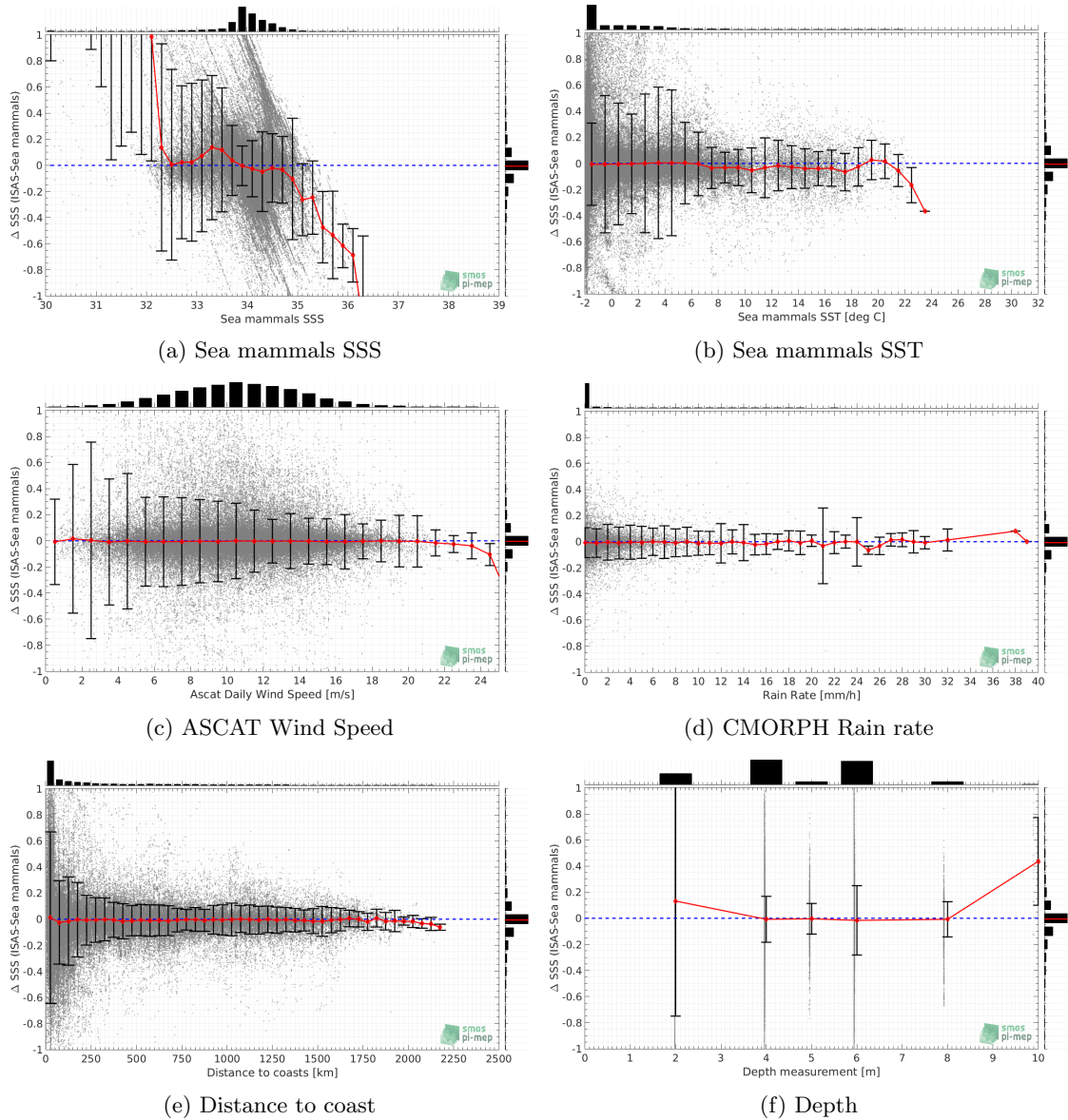


Figure 10:  $\Delta$  SSS (ISAS - Sea mammals) sorted as geophysical conditions: Sea mammals SSS a), Sea mammals SST b), ASCAT Wind speed c), CMORPH rain rate d), distance to coast (e) and depth measurements (f).

## 2.12 $\Delta$ SSS maps and statistics for different geophysical conditions

In Figures 11 and 12, we focus on sub-datasets of the match-up differences  $\Delta$ SSS (ISAS - *in situ*) for the following specific geophysical conditions:

- **C1**: if the local value at *in situ* location of estimated rain rate is zero, mean daily wind is in the range [3, 12] m/s, the SST is  $> 5^\circ\text{C}$  and distance to coast is  $> 800$  km.
- **C2**: if the local value at *in situ* location of estimated rain rate is zero, mean daily wind is



in the range [3, 12] m/s.

- **C3**:if the local value at *in situ* location of estimated rain rate is high (ie.  $> 1$  mm/h) and mean daily wind is low (ie.  $< 4$  m/s).
- **C4**:if the mixed layer is shallow with depth  $< 20$ m.
- **C5**:if the *in situ* data is located where the climatological SSS standard deviation is low (ie. above  $< 0.2$ ).
- **C6**:if the *in situ* data is located where the climatological SSS standard deviation is high (ie. above  $> 0.2$ ).

For each of these conditions, the temporal mean (gridded over spatial boxes of size  $1^\circ \times 1^\circ$ ) and the histogram of the difference  $\Delta$ SSS (ISAS - *in situ*) are presented.

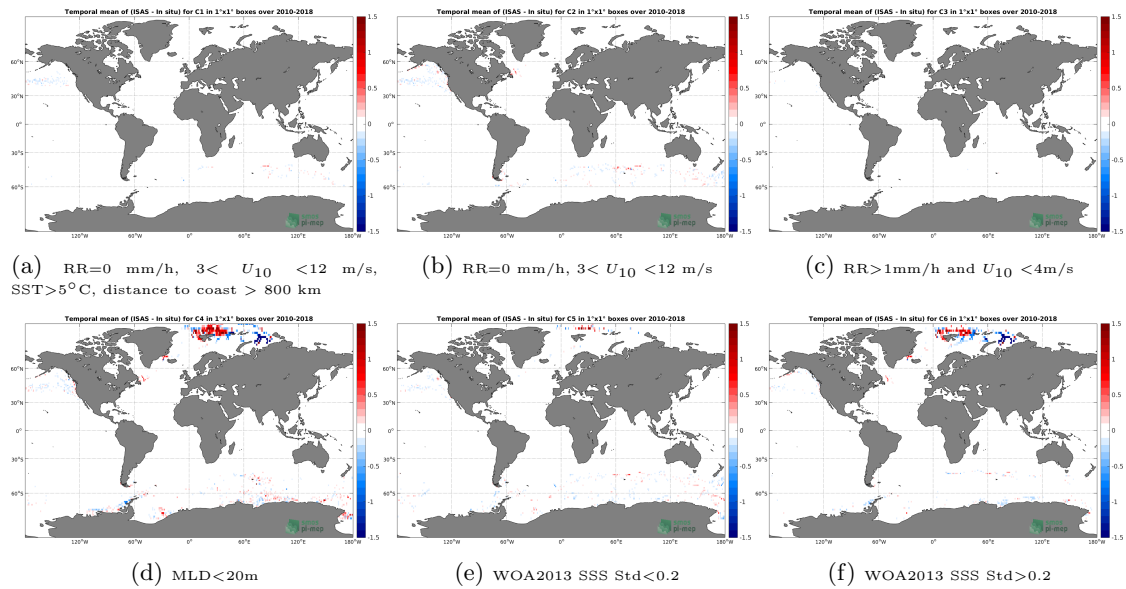


Figure 11: Temporal mean gridded over spatial boxes of size  $1^\circ \times 1^\circ$  of  $\Delta$ SSS (ISAS - Sea mammals) for 6 different subdatasets corresponding to:RR=0 mm/h,  $3 < U_{10} < 12$  m/s, SST $>5^\circ$ C, distance to coast  $> 800$  km (a), RR=0 mm/h,  $3 < U_{10} < 12$  m/s (b), RR $>1$ mm/h and  $U_{10} < 4$ m/s (c),MLD $<20$ m (d),WOA2013 SSS Std $<0.2$  (e),WOA2013 SSS Std $>0.2$  (f).

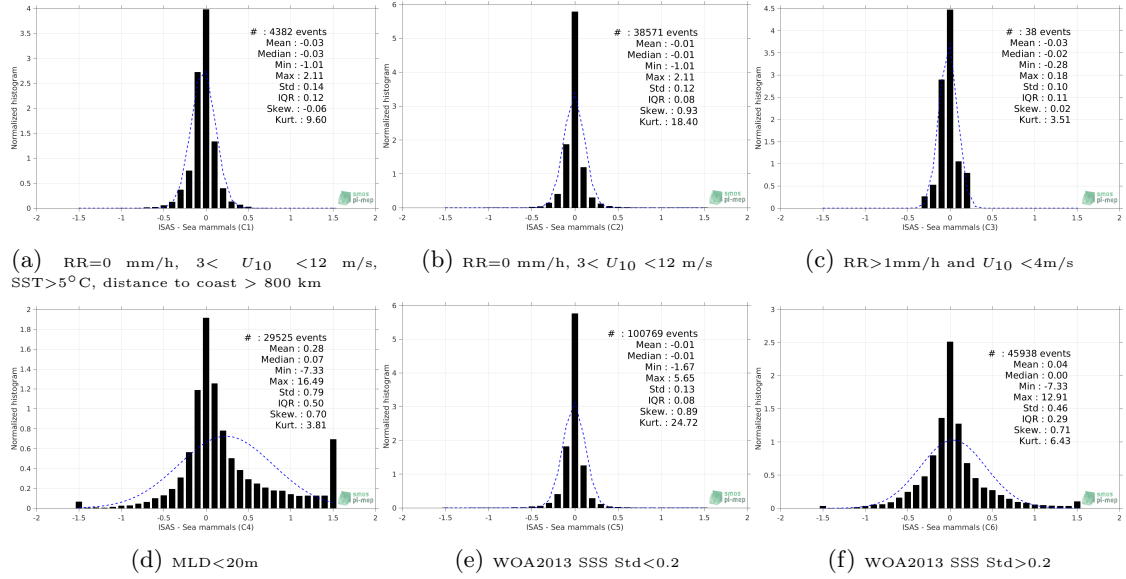


Figure 12: Normalized histogram of  $\Delta SSS$  (ISAS - Sea mammals) for 6 different subdatasets corresponding to: RR=0 mm/h,  $3 < U_{10} < 12$  m/s, SST>5°C, distance to coast > 800 km (a), RR=0 mm/h,  $3 < U_{10} < 12$  m/s (b), RR>1mm/h and  $U_{10} < 4$ m/s (c), WOA2013 SSS Std<0.2 (d), WOA2013 SSS Std>0.2 (e).

## 2.13 Summary

Table 1 shows the mean, median, standard deviation (Std), root mean square (RMS), interquartile range (IQR), correlation coefficient ( $r^2$ ) and robust standard deviation (Std\*) of the match-up differences  $\Delta SSS$  (ISAS - Sea mammals) for the following conditions:

- all: All the match-up pairs satellite/in situ SSS are used to derive the statistics
- C1: only pairs where RR=0 mm/h,  $3 < U_{10} < 12$  m/s, SST>5°C, distance to coast > 800 km
- C2: only pairs where RR=0 mm/h,  $3 < U_{10} < 12$  m/s
- C3: only pairs where RR>1mm/h and  $U_{10} < 4$ m/s
- C4: only pairs where MLD<20m
- C5: only pairs where WOA2013 SSS Std<0.2
- C6: only pairs where WOA2013 SSS Std>0.2
- C7a: only pairs where distance to coast is < 150 km.
- C7b: only pairs where distance to coast is in the range [150, 800] km.
- C7c: only pairs where distance to coast is > 800 km.
- C8a: only pairs where in situ SST is < 5°C.
- C8b: only pairs where in situ SST is in the range [5, 15]°C.

- C8c: only pairs where in situ SST is  $> 15^{\circ}\text{C}$ .
- C9a: only pairs where in situ SSS is  $< 33$ .
- C9b: only pairs where in situ SSS is in the range [33, 37].
- C9c: only pairs where in situ SSS is  $> 37$ .

**Table 1: Statistics of  $\Delta\text{SSS}$  (ISAS - Sea mammals)**

Condition	#	Median	Mean	Std	RMS	IQR	$r^2$	Std*
all	175811	-0.01	0.03	0.40	0.40	0.12	0.49	0.09
C1	4382	-0.03	-0.03	0.14	0.14	0.12	0.94	0.09
C2	38571	-0.01	-0.01	0.12	0.12	0.08	0.93	0.06
C3	38	-0.02	-0.03	0.10	0.10	0.11	0.97	0.09
C4	29525	0.07	0.28	0.79	0.84	0.50	0.36	0.29
C5	100769	-0.01	-0.01	0.13	0.13	0.08	0.87	0.06
C6	45938	0.00	0.04	0.46	0.46	0.29	0.63	0.21
C7a	81623	0.00	0.07	0.56	0.57	0.24	0.32	0.18
C7b	63015	-0.01	-0.01	0.17	0.17	0.09	0.86	0.06
C7c	31148	-0.01	-0.01	0.10	0.10	0.07	0.92	0.05
C8a	154749	0.00	0.03	0.42	0.42	0.12	0.38	0.09
C8b	18630	-0.02	-0.01	0.22	0.22	0.12	0.90	0.09
C8c	2432	-0.04	-0.04	0.14	0.15	0.16	0.91	0.12
C9a	7650	0.08	0.60	1.29	1.42	0.98	0.11	0.29
C9b	168161	-0.01	0.00	0.28	0.28	0.12	0.51	0.09
C9c	0	NaN	NaN	NaN	NaN	NaN	NaN	NaN

Table 1 numerical values can be downloaded as a csv file [here](#).

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