



Match-up database Analyses Report

$\begin{array}{c} \text{SMOS-L3-CATDS-CECOS-IFREMER-V2-1DAY-}\\ 0.5\text{DEG} \end{array}$

Argo

Amazon and Orinoco river plumes

prepared by the Pi-MEP Consortium March 15, 2019

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Acronym

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



SSS_{SAT}	Satellite SSS product considered for the match-up
ΔSSS	Difference between satellite and in situ SSS at colocalized point (Δ SSS =
	SSS_{SAT} - SSS_{insitu})
SST	Sea Surface Temperature
Std	Standard deviation
$\operatorname{Std}^{\star}$	Robust Standard deviation = $median(abs(x-median(x)))/0.67$ (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

In this report, we present systematic analyses of the Match-up DataBase (MDB) files generated by the Pi-MEP platform within the following Pi-MEP region and for the below pair of Satellite/In situ SSS data:

- Pi-MEP region: Amazon and Orinoco river plumes (download the corresponding mask here)
- SSS satellite product (SSS_{SAT}): SMOS-L3-CATDS-CECOS-IFREMER-V2-1DAY-0.5DEG
- In situ dataset (SSS_{Insitu}): Argo (download the corresponding report here)

In the following, $\Delta SSS = SSS_{SAT}$ - SSS_{Insitu} denotes the difference between the satellite and in situ SSS at the colocalized points that form the MDB.

This report presents successively:

The MDB file DataSets (Section 2)

- A short description of the satellite SSS product considered in the match-up (2.1)
- A short description of the In situ SSS dataset considered in the match-up (2.2)
- A short description of the auxiliary geophysical datasets co-localized with SSS pairs (2.3)
- An overview of how the Match-ups were evaluated (2.4)
- An overview of the MDB characteristics for the particular in situ/satellite pairs (2.5)

The major results of the MDB file Analyses (Section 3)

- Spatial Maps of the Time-mean and temporal Std of in situ and satellite SSS and of the $\Delta {\rm SSS}$ (3.1)
- Time series of the monthly averaged mean and Std of in situ and satellite SSS and of the Δ SSS (3.2)
- Zonally-averaged Time-mean and temporal Std of in situ and satellite SSS and of the Δ SSS (3.3)
- Scatterplots of satellite vs in situ SSS by latitudinal bands (3.4)
- Time series of the monthly averaged mean and Std of the Δ SSS sorted by latitudinal bands (3.5)
- Δ SSS sorted as function of geophysical parameters (3.6)
- Δ SSS maps and statistics for different geophysical conditions (3.7)

All analyses are conducted over the Pi-MEP Region specified above and over the full satellite SSS product period.



2 The MDB file datasets

2.1 Satellite SSS product

2.1.1 SMOS-L3-CATDS-CECOS-IFREMER-V2-1DAY-0.5DEG

The version V02 of the SMOS L3 salinity Ifremer research products from the CATDS/CEC Expertise center includes reprocessed data from May 2010 to December 2017 with a daily temporal window. Compared to the previous version V01, the V02 version includes daily composite at 0.5° resolution without the previous ± 5 days running window average. These products are the daily composite from Ascending and Descending passes for the considered day. The Level 3 SSS algorithm is now starting using as input the L1B ESA products (while we used L1A products for the V01 version).

In addition, each product contain also:

- ECMWF SST (GHRSST product)
- ECMWF wind speed module: These additional information in the products were first colocated with SMOS swath data and then re-gridded as the level 3 SSS space-time resolution.
- An estimate for the strong Radio Frequency Interference (RFI) contamination probability is also included at the SSS retrieved pixel and for the level 3 SSS space-time window.
- Residual temporal drifts (at seasonal time scales) and land contamination signals are still found in the up-to-date SMOS brightness temperature level 1 products. These are mostly due to yet unsolved calibration (thermal) and image reconstruction issues (solar effects in L-band imaging, antenna patterns uncertainties). These impacts translate into significant latitudinal and seasonal biases in the retrieved SSS with different amplitudes in ascending and descending passes. An attempt to correct for such large-scale biases was applied at Level 3: the L3 SSS data for ascending and descending passes were bias-adjusted using a large-scale correction based on a 5°x5° spatial filter derived from the World Ocean Atlas 2001 monthly SSS climatology, temporally interpolated to the SMOS acquisition date. Not as well that only one fixed Ocean Target Transformation (used to correct for spatial image reconstruction biases) was used to generate all the products.
- a new empirical correction for the galactic glint contamination has been implemented from a Geometric Optic model which shall improve the quality of SSS retrievals in Sep-Oct months for descending passes and in March-April months for Ascending passes.
- a filter at the brightness temperature level has been applied to remove data reconstructed in the proximity of the aliased sun-disk, in order to best mitigate solar contamination effects.
- Radio Frequency Interferences are still a big issue for SMOS SSS retrieval quality. In order to tentatively best minimize the impact of RFI in our level 3 products, a weighting average of the swath retrieved SSS based on the RFI probability at the considered pixels and time window has been applied. This probability being now incorporated in the products, the user can now chose by him or herself only those pixels with 0 probability of (detected) RFI. Pixels with RFI probability higher than 80% over the considered space-time window have been removed.

Note that V02 data were not generated before May 2010 because of the reduced quality of the datasets during the initial period of the satellite commissioning phase. User still interested in that period can use our previous V01 data.



SMOS-L3-CATDS-CECOS-IFREMER-V2-1DAY-0.5DEG							
Spatial resolution	0.5°						
Temporal resolution	Daily						
Temporal coverage	From 2010-05-01 to 2017-12-31						
Data Provider	CATDS						
Release Date	2017-09-08						
Version	2						
Data access	sss_smos_13_V02						
User Guide	CATDS_CECOS_SMOS_Level3Products_UserDocument_V02.pdf						
Validation report	CATDS_CECOS_SMOS_Level3Products_Validation.pdf						
Documentation	$OS_L3_products_Differences_and_ProsCons.pdf$						

These daily Composite, half degree resolution SMOS SSS data were obtained from the Ocean Salinity Expertise Center (CECOS) of the CNES-IFREMER Centre Aval de Traitemenent des Donnees SMOS (CATDS), at IFREMER, Plouzane (France).

2.2 In situ SSS dataset

Argo is a global array of 3,000 free-drifting profiling floats that measures the temperature and salinity of the upper 2000 m of the ocean. This allows continuous monitoring of the temperature and salinity of the upper ocean, with all data being relayed and made publicly available within hours after collection. The array provides around 100,000 temperature/salinity profiles per year distributed over the global oceans at an average of 3-degree spacing. Only Argo salinity and temperature float data with quality index set to 1 or 2 and data mode set to real time (RT), real time adjusted (RTA) and delayed mode (DM) are considered in the Pi-MEP. Argo floats which may have problems with one or more sensors appearing in the grey list maintained at the Coriolis/GDACs are discarded. Furthermore, Pi-MEP provides an additional list of ~1000 "suspicious" argo salinity profiles that are also removed before analysis. The upper ocean salinity and temperature values recorded between 0m and 10m depth are considered as Argo sea surface salinities (SSS) and sea surface temperatures (SST). These data were collected and made freely available by the international Argo project and the national programs that contribute to it (Argo (2000)).

2.3 Auxiliary geophysical datasets

Additional EO datasets are used to characterize the geophysical conditions at the in situ/satellite SSS pair 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 spatiotemporal 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 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 test 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 satellite SSS products. 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, monthly std climatological fields from the World Ocean Atlas (WOA13) at the match-up pairs location and date are also used to have an a priori information of the local SSS variability.

2.3.1 CMORPH

Precipitation are estimated using the CMORPH 3-hourly products at 1/4° 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 $12 \ge 15$ 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. O/RAW/. 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).

2.3.2 ASCAT

Advanced SCATterometer (ASCAT) daily data produced and made available at Ifremer/CERSAT on a 0.25°x0.25° resolution grid (Bentamy and Fillon (2012)) since March 2007 are used to char-



acterize 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⁻¹ and 20°.

2.3.3 ISAS

The In Situ Analysis System (ISAS), as described in Gaillard et al. (2016) is a data based reanalysis 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 $70^{\circ}N-70^{\circ}S$ on a $1/2^{\circ}$ 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 Copernicus (www.marine.copernicus.eu/documents/ PUM/CMEMS-INS-PUM-013-002-ab.pdf). 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.

2.3.4 World Ocean Atlas Climatology

The World Ocean Atlas 2013 version 2 (WOA13 V2) is a set of objectively analyzed (1° 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° , 1° , and 0.25° 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.

2.4 Overview of the Match-ups generation method

The match-up production is basically a three steps process:

- 1. preparation of the input in situ and satellite data, and,
- 2. co-localization of satellite products with in situ SSS measurements.
- 3. co-localization of the in situ/satellite pair with auxiliary information.

In the following, we successively detail the approaches taken for these different steps.

2.4.1 In Situ/Satellite data filtering

The first step consist in filtering Argoin situ dataset using the quality flags as described in 2.2 so that only valid salinity data remains in the produced match-ups.

For high-spatial resolution in situ SSS measurements such as the Thermo-SalinoGraph (TSG) SSS data from research vessels, Voluntary Observing Ships (VOS) or sailing ships, as well as SSS data from surface drifters, an additional spatial-filtering step is performed on the in situ data that will be in fine compared to the satellite SSS products. If R_{sat} is the spatial resolution of the satellite SSS product (L2 to L3-L4), we keep the in situ data at the original spatial resolution but we also estimate for all spatio-temporal samples a running median filtered SSS applied to all neighbouring in situ SSS data acquired within a distance of $R_{sat}/2$ from a given in situ acquisition. Both the original and the filtered data are finally stored in the MDB files.

Only for satellite L2 SSS data, a third step consist in filtering spurious data using the flags and associated recommendation as provided by the official data centers and described in 2.1.

2.4.2 In Situ/Satellite Co-localization

In this step, each SSS satellite acquisition is co-localized with the filtered in situ measurements. The method used for co-localization differ if the satellite SSS is a swath product (so-called Level 2-types) or a time-space composite product (so-called Level 3/level 4-types).

• For L2 SSS swath data :

If R_{sat} is the spatial resolution of the satellite swath SSS product, for each in situ data sample collected in the Pi-MEP database, the platform searches for all satellite SSS data found at grid nodes located within a radius of $R_{sat}/2$ from the in situ data location and acquired with a time-lag from the in situ measurement date that is less or equal than ± 12 hours. If several satellite SSS samples are found to meet these criteria, the final satellite SSS match-up point is selected to be the closest in time from the in situ data measurement date. The final spatial and temporal lags between the in situ and satellite data are stored in the MDB files.

• For L3 and L4 composite SSS products :

If R_{sat} is the spatial resolution of the composite satellite SSS product and D the period over which the composite product was built (e.g., periods of 1, 7, 8, 9, 10, 18 days, 1 month, etc..) with central time to, for each in situ data sample collected in the Pi-MEP database



during period D, the platform searches for all satellite SSS data of the composite product found at grid nodes located within a radius of $R_{sat}/2$ from the in situ data location. If several satellite SSS product samples are found to meet these criteria, the final satellite SSS match-up point is chosen to be the composite SSS with central time to which is the closest in time from the in situ data measurement date. The final spatial and temporal lags between the in situ and satellite data are stored in the MDB files.

2.4.3 MDB pair Co-localization with auxiliary data and complementary information

MDB data consist of satellite and in-situ SSS pair datasets but also of auxiliary geophysical parameters such as local and history of wind speed and rain rates, as well as various information (climatology, distance to coast, mixed layer depth, barrier layer thickness, etc) that can be derived from in situ data and which are included in the final match-up files. The collocation of auxiliary parameters and additional information is done for each filtered in-situ SSS measurement contained in the match-up files as follows :

If t_{insitu} is the time/date at which the in situ measurement is performed, we collect:

- The ASCAT wind speed product of the same day than t_{insitu} found at the ASCAT $1/4^{\circ}$ grid node with closest distance from the in situ data location and the time series of the ASCAT wind speed at the same node for the 10 days prior the in situ measurement day.
- If the in situ data is located within the 60°N-60°S band, we select the CMORPH 3-hourly product the closest in time from tin situ and found at the CMORPH 1/4° grid node with closest distance from the in situ data location. We then store the time series of the CMORPH rain rate at the same node for the 10 days prior the in situ measurement time.

For the given month/year of the in situ data, we select the ISAS and WOA fields for the same month (and same year for ISAS fields) and take the SSS analysis (monthly mean, std) found at the closest grid node from the in situ measurement.

The distance from the in situ SSS data location to the nearest coast is evaluated and provided in km. We use a distance-to-coast map at $1/4^{\circ}$ resolution where small islands have been removed.

When vertical profiles of salinity (S) and temperature (T) are made available from the in situ measurements used to build the match-up (Argo or sea mammals), the following variables are included into each satellite/in situ match-up file:

- 1. The vertical distribution of pressure at which the profile were measured,
- 2. The vertical S(z) and T(z) profiles,
- 3. The vertical potential density anomaly profile $\sigma_0(z)$,
- 4. The Mixed Layer Depth (MLD). The MLD is defined here as the depth where the potential density has increased from the reference depth (10 meter) by a threshold equivalent to 0.2° C decrease in temperature at constant salinity: $\sigma_0 = \sigma_{010m} + \Delta \sigma_0$ with $\Delta \sigma_0 = \sigma_0(\theta_{10m} 0.2, S_{10m}) \sigma_0(\theta_{10m}, S_{10m})$ where θ_{10m} and S_{10m} are the temperature and salinity at the reference depth (i.e. 10 m) (de Boyer Montégut et al. (2004), de Boyer Montégut et al. (2007)).
- 5. The Top of the Thermocline Depth (TTD) is defined as the depth at which temperature decreases from its 10 m value by 0.2°C.



- 6. The Barrier Layer if present, is defined as the intermediate layer between the top of the thermocline and the bottom of the density mixed-layer and its thickness (BLT) is defined as the difference between the MLD and the TTD.
- 7. The vertical profile of the buoyancy frequency $N^2(z)$

The resulting match-ups files are serialized as NetCDF-4 files whose structure depends on the origin of the in-situ data they contain.

2.4.4 Content of the Match-Up NetCDF files

```
netcdf pimep-mdb_smos-l3-catds-ifr-v2-1d-05deg_argo_TIMEID_v01 {
dimensions:
     N_{-}prof = 944;
     N_{\text{LEVELS}} = 499;
     N_DAYS_WIND = 10;
     N_3H_RAIN = 80;
     TIME\_Sat = UNLIMITED; // (1 currently)
variables:
float DATE_ARGO(N_prof);
    DATE_ARGO:long_name = "Date of Argo profile";
    DATE_ARGO:units = "days since 1990-01-01 00:00:00";
    DATE_ARGO:standard_name = "time";
    DATE_ARGO:_FillValue = -999.f;
float LATITUDE_ARGO(N_prof);
    LATITUDE_ARGO:long_name = "Latitude of Argo profile";
    LATITUDE_ARGO:units = "degrees_north";
    LATITUDE_ARGO:valid_min = -90.;
    LATITUDE_ARGO:valid_max = 90. ;
    LATITUDE_ARGO:standard_name = "latitude";
    LATITUDE_ARGO:_FillValue = -999.f;
float LONGITUDE_ARGO(N_prof);
    LONGITUDE_ARGO:long_name = "Longitude of Argo profile";
    LONGITUDE_ARGO:units = "degrees_east";
    LONGITUDE_ARGO:valid_min = -180.;
    LONGITUDE_ARGO:valid_max = 180.;
    LONGITUDE_ARGO:standard_name = "longitude";
    LONGITUDE_ARGO:_FillValue = -999.f;
float SSS_DEPTH_ARGO(N_prof);
    SSS_DEPTH_ARGO:long_name = "Sea water pressure at Argo float location (equals 0 at
sea level)";
    SSS_DEPTH_ARGO:units = "decibar";
    SSS_DEPTH_ARGO:standard_name = "sea_water_pressure";
    SSS_DEPTH_ARGO:_FillValue = -999.f;
float SSS_ARGO(N_prof);
    SSS_ARGO:long_name = "Argo SSS";
    SSS\_ARGO:units = "1";
    SSS\_ARGO:salinity\_scale = "Practical Salinity Scale(PSS-78)";
    SSS_ARGO:standard_name = "sea_water_salinity";
    SSS_ARGO:_FillValue = -999.f;
```



float SST_ARGO(N_prof); $SST_ARGO:long_name = "Argo SST";$ $SST_ARGO:units = "degree Celsius";$ SST_ARGO:standard_name = "sea_water_temperature"; $SST_ARGO:FillValue = -999.f;$ float DELAYED_MODE_ARGO(N_prof); $DELAYED_MODE_ARGO:long_name = "Argo data mode (delayed mode = 1, real time)$ =0)"; DELAYED_MODE_ARGO:units = "1"; $DELAYED_MODE_ARGO:_FillValue = -999.f;$ float DISTANCE_TO_COAST_ARGO(N_prof) : DISTANCE_TO_COAST_ARGO:long_name = "Distance to coasts at Argo float location" ; $DISTANCE_TO_COAST_ARGO:units = "km";$ $DISTANCE_TO_COAST_ARGO:_FillValue = -999.f;$ float PLATFORM_NUMBER_ARGO(N_prof) ; PLATFORM_NUMBER_ARGO:long_name = "Argo float unique identifier"; PLATFORM_NUMBER_ARGO:conventions = "WMO float identifier : A9IIIII"; $PLATFORM_NUMBER_ARGO:units = "1";$ PLATFORM_NUMBER_ARGO:_FillValue = -999.f; float PSAL_ARGO(N_prof, N_LEVELS); PSAL_ARGO:long_name = "Argo salinity profile"; $PSAL_ARGO:units = "1";$ PSAL_ARGO:salinity_scale = "Practical Salinity Scale (PSS-78)"; PSAL_ARGO:standard_name = "sea_water_salinity"; $PSAL_ARGO:_FillValue = -999.f;$ float TEMP_ARGO(N_prof, N_LEVELS); TEMP_ARGO:long_name = "Argo temperature profile"; $TEMP_ARGO:units = "degree Celsius";$ TEMP_ARGO:standard_name = "sea_water_temperature"; TEMP_ARGO:_FillValue = -999.f; float PRES_ARGO(N_prof, N_LEVELS); PRES_ARGO:long_name = "Argo pressure profile"; PRES_ARGO:units = "decibar"; PRES_ARGO:standard_name = "sea_water_pressure"; $PRES_ARGO:_FillValue = -999.f;$ float RHO_ARGO(N_prof, N_LEVELS); RHO_ARGO:long_name = "Argo in-situ density profile"; $RHO_ARGO:units = "kg/m";$ $RHO_ARGO:_FillValue = -999.f;$ float SIGMA0_ARGO(N_prof, N_LEVELS); SIGMA0_ARGO:long_name = "Argo potential density anomaly profile"; SIGMA0_ARGO:units = "kg/m³"; $SIGMA0_ARGO:_FillValue = -999.f;$ float N2_ARGO(N_prof, N_LEVELS); N2_ARGO:long_name = "Argo buoyancy frequency profile"; N2_ARGO:units = $"1/s^{2"}$; $N2_ARGO:_FillValue = -999.f;$ float MLD_ARGO(N_prof);



```
MLD_ARGO:long_name = "Mixed Layer Depth (MLD) calculated from Argo profile (depth
where \sigma_0 = \sigma_{010m} + \Delta \sigma_0 with \Delta \sigma_0 = \sigma_0(\theta_{10m} - 0.2, S_{10m}) - \sigma_0(\theta_{10m}, S_{10m}) )";
     MLD_ARGO:units = "m";
     MLD_ARGO:FillValue = -999.f;
float TTD_ARGO(N_prof);
     TTD_ARGO:long_name = "Top of Thermocline Depth (TTD) calculated from Argo profile
(depth where \theta = \theta_{10m} - 0.2)";
     TTD_ARGO:units = "m";
     TTD_ARGO:_FillValue = -999.f;
float BLT_ARGO(N_prof);
     BLT_ARGO:long_name = "Barrier Laver Thickness (TTD-MLD)";
     BLT_ARGO:units = "m";
     BLT_ARGO:_FillValue = -999.f;
float DATE_Satellite_product(TIME_Sat);
     DATE_Satellite_product:long_name = "Central time of satellite SSS file";
     DATE_Satellite_product:units = "days since 1990-01-01 00:00:00";
     DATE_Satellite_product:standard_name = "time";
float LATITUDE_Satellite_product(N_prof);
     LATITUDE_Satellite_product:long_name = "Satellite product latitude at Argo float loca-
tion";
     LATITUDE_Satellite_product:units = "degrees_north";
     LATITUDE_Satellite_product:valid_min = -90.;
     LATITUDE_Satellite_product:valid_max = 90.;
     LATITUDE_Satellite_product:standard_name = "latitude";
     LATITUDE_Satellite_product:_FillValue = -999.f;
float LONGITUDE_Satellite_product(N_prof);
     LONGITUDE_Satellite_product:long_name = "Satellite product longitude at Argo float lo-
cation";
     LONGITUDE_Satellite_product:units = "degrees_east";
     LONGITUDE_Satellite_product:valid_min = -180.;
     LONGITUDE_Satellite_product:valid_max = 180.;
     LONGITUDE\_Satellite\_product:standard\_name = "longitude";
     LONGITUDE_Satellite_product:_FillValue = -999.f;
float SSS_Satellite_product(N_prof) ;
     SSS_Satellite_product:long_name = "Satellite product SSS at Argo float location";
     SSS\_Satellite\_product:units = "1";
     SSS\_Satellite\_product:salinity\_scale = "Practical Salinity Scale(PSS-78)";
     SSS_Satellite_product:standard_name = "sea_surface_salinity";
     SSS\_Satellite\_product:\_FillValue = -999.f;
float SST_Satellite_product(N_prof) ;
     SST_Satellite_product:long_name = "Satellite product SST at Argo float location";
     SST_Satellite_product:units = "degree Celsius";
     SST_Satellite_product:standard_name = "sea_surface_temperature";
     SST_Satellite_product:_FillValue = -999.f;
float Spatial_lags(N_prof);
     Spatial_lags:long_name = "Spatial lag between Argo float location and satellite SSS product
pixel center";
     Spatial_lags:units = "km";
     Spatial_lags:_FillValue = -999.f;
```



float $Time_lags(N_prof)$; Time_lags:long_name = "Temporal lag between Argo float time and satellite SSS product central time"; $Time_{ass:units} = "days";$ Time_lags:_FillValue = -999.f; float ROSSBY_RADIUS_at_ARGO(N_prof); ROSSBY_RADIUS_at_ARGO:long_name = "Baroclinic Rossby radius of deformation (Chelton et al., 1998) at Argo float location"; $ROSSBY_RADIUS_at_ARGO:units = "km";$ $ROSSBY_RADIUS_at_ARGO:_FillValue = -999.f;$ float Ascat_daily_wind_at_ARGO(N_prof); Ascat_daily_wind_at_ARGO:long_name = "Daily Ascat wind speed module at Argo float location"; $Ascat_daily_wind_at_ARGO:units = "m/s";$ $Ascat_daily_wind_at_ARGO:_FillValue = -999.f;$ float CMORPH_3h_Rain_Rate_at_ARGO(N_prof); CMORPH_3h_Rain_Rate_at_ARGO:long_name = "3-hourly CMORPH rain rate at Argo float location"; $CMORPH_3h_Rain_Rate_at_ARGO:units = "mm/3h";$ CMORPH_3h_Rain_Rate_at_ARGO:_FillValue = -999.f; float Ascat_10_prior_days_wind_at_ARGO(N_prof, N_DAYS_WIND); Ascat_10_prior_days_wind_at_ARGO:long_name = "Prior 10 days time series of Ascat wind speed module at Argo float location"; $Ascat_10_prior_days_wind_at_ARGO:units = "m/s";$ Ascat_10_prior_days_wind_at_ARGO:_FillValue = -999.f; float CMORPH_10_prior_days_Rain_Rate_at_ARGO(N_prof, N_3H_RAIN); CMORPH_10_prior_days_Rain_Rate_at_ARGO:long_name = "Prior 10 days times series of 3-hourly CMORPH Rain Rate at Argo float location"; $\label{eq:cmorphi} CMORPH_10_prior_days_Rain_Rate_at_ARGO:units = "mm/3h" ;$ CMORPH_10_prior_days_Rain_Rate_at_ARGO:_FillValue = -999.f; float SSS_ISAS_at_ARGO(N_prof); SSS_ISAS_at_ARGO:long_name = "ISAS SSS (5m depth) at Argo float location"; $SSS_ISAS_at_ARGO:units = "1";$ $SSS_ISAS_at_ARGO:salinity_scale = "Practical Salinity Scale(PSS-78)";$ SSS_ISAS_at_ARGO:standard_name = "sea_water_salinity"; $SSS_ISAS_at_ARGO:_FillValue = -999.f;$ float SSS_PCTVAR_ISAS_at_ARGO(N_prof); SSS_PCTVAR_ISAS_at_ARGO:long_name = "Error on ISAS SSS (5m depth) at Argo float location (% variance)"; $SSS_PCTVAR_ISAS_at_ARGO:units = "\%";$ $SSS_PCTVAR_ISAS_at_ARGO:_FillValue = -999.f;$ float SSS_WOA13_at_ARGO(N_prof); SSS_WOA13_at_ARGO:long_name = "WOA 2013 (DECAV-1deg) SSS (0m depth) at Argo float location"; $SSS_WOA13_at_ARGO:units = "1";$ SSS_WOA13_at_ARGO:salinity_scale = "Practical Salinity Scale(PSS-78)"; SSS_WOA13_at_ARGO:standard_name = "sea_surface_salinity"; $SSS_WOA13_at_ARGO:_FillValue = -999.f;$ float SSS_STD_WOA13_at_ARGO(N_prof);



SSS_STD_WOA13_at_ARGO:long_name = "WOA 2013 (DECAV-1deg) SSS STD (0m depth) at Argo float location "; $SSS_STD_WOA13_at_ARGO:units = "1";$ $SSS_STD_WOA13_at_ARGO:_FillValue = -999.f;$ float SSS_ISAS15_at_ARGO(N_prof); SSS_ISAS15_at_ARGO:long_name = "Monthly ISAS-15 SSS (5m depth) at Argo float location"; $SSS_ISAS15_at_ARGO:units = "1";$ SSS_ISAS15_at_ARGO:salinity_scale = "Practical Salinity Scale (PSS-78)"; SSS_ISAS15_at_ARGO:standard_name = "sea_water_salinity"; $SSS_ISAS15_at_ARGO:_FillValue = -999.f;$ float SSS_PCTVAR_ISAS15_at_ARGO(N_prof); SSS_PCTVAR_ISAS15_at_ARGO:long_name = "Error on monthly ISAS-15 SSS (5m depth) at Argo float location (% variance)"; $SSS_PCTVAR_ISAS15_at_ARGO:units = "\%";$ $SSS_PCTVAR_ISAS15_at_ARGO:_FillValue = -999.f;$ float SSS_WOA18_at_ARGO(N_prof); SSS_WOA18_at_ARGO:long_name = "Monthly WOA 2018 (DECAV-1deg) SSS (0m depth) at Argo float location"; $SSS_WOA18_at_ARGO:units = "1";$ SSS_WOA18_at_ARGO:salinity_scale = "Practical Salinity Scale (PSS-78)"; SSS_WOA18_at_ARGO:standard_name = "sea_surface_salinity"; $SSS_WOA18_at_ARGO:_FillValue = -999.f;$ float SSS_STD_WOA18_at_ARGO(N_prof); SSS_STD_WOA18_at_ARGO:long_name = "Monthly WOA 2018 (DECAV-1deg) SSS STD (0m depth) at Argo float location "; $SSS_STD_WOA18_at_ARGO:units = "1";$ $SSS_STD_WOA18_at_ARGO:_FillValue = -999.f;$ float SEA_ICE_CONCENTRATION_at_ARGO(N_prof); SEA_ICE_CONCENTRATION_at_ARGO:long_name = "Daily sea ice area fraction (EU-METSAT OSI-SAF OSI-450) at Argo float location (%)"; $SEA_ICE_CONCENTRATION_at_ARGO:units = "1";$ SEA_ICE_CONCENTRATION_at_ARGO:standard_name = "sea_ice_area_fraction"; SEA_ICE_CONCENTRATION_at_ARGO:_FillValue = -999.f; float CCMP_6h_Wind_Speed_at_ARGO(N_prof); CCMP_6h_Wind_Speed_at_ARGO:long_name = "6-hourly CCMP wind speed at Argo float location"; $CCMP_6h_Wind_Speed_at_ARGO:units = "m s-1";$ CCMP_6h_Wind_Speed_at_ARGO:standard_name = "wind_speed"; $CCMP_6h_Wind_Speed_at_ARGO:_FillValue = -999.f;$ float CCMP_10_prior_days_Wind_Speed_at_ARGO(N_prof, N_DAYS_WIND_CCMP); CCMP_10_prior_days_Wind_Speed_at_ARGO:long_name = "Prior 10 days time series of CCMP wind speed at Argo float location"; $CCMP_{10}$ -prior_days_Wind_Speed_at_ARGO:units = "m s-1"; CCMP_10_prior_days_Wind_Speed_at_ARGO:standard_name = "wind_speed"; CCMP_10_prior_days_Wind_Speed_at_ARGO:_FillValue = -999.f; float CDM_GLOBCOLOUR_at_ARGO(N_prof); CDM_GLOBCOLOUR_at_ARGO:long_name = "8-day Coloured dissolved and detrital organic materials - mean of the binned pixels at Argo float location";



 $CDM_GLOBCOLOUR_at_ARGO:units = "m-1";$; $CDM_GLOBCOLOUR_at_ARGO:_FillValue = -999.f;$ float CHL1_GLOBCOLOUR_at_ARGO(N_prof); CHL1_GLOBCOLOUR_at_ARGO:long_name = "8-day Chlorophyll concentration - mean of the binned pixels at Argo float location"; $CHL1_GLOBCOLOUR_at_ARGO:units = "mg m-3";$ $CHL1_GLOBCOLOUR_at_ARGO:standard_name = "mass_concentration_of_chlorophyll_a_in_sea_water"$; $CHL1_GLOBCOLOUR_at_ARGO:_FillValue = -999.f;$ float EVAPORATION_OAFLUX_at_ARGO(N_prof); EVAPORATION_OAFLUX_at_ARGO:long_name = "Daily mean evaporation rate (OAFlux) at Argo float location"; $EVAPORATION_OAFLUX_at_ARGO:units = "cm year-1";$ EVAPORATION_OAFLUX_at_ARGO:_FillValue = -999.f; float SSS_SCRIPPS_at_ARGO(N_prof); SSS_SCRIPPS_at_ARGO:long_name = "Argo gridded monthly mean SSS (0m depth) from SCRIPPS (Roemmich-Gilson) at Argo float location"; $SSS_SCRIPPS_at_ARGO:units = "1";$ SSS_SCRIPPS_at_ARGO:salinity_scale = "Practical Salinity Scale (PSS-78)"; SSS_SCRIPPS_at_ARGO:standard_name = "sea_water_salinity"; $SSS_SCRIPPS_at_ARGO:_FillValue = -999.f;$ float SSS_IPRC_at_ARGO(N_prof); SSS_IPRC_at_ARGO:long_name = "Argo gridded monthly mean SSS (0m depth) from IPRC at Argo float location"; $SSS_IPRC_at_ARGO:units = "1";$ SSS_IPRC_at_ARGO:salinity_scale = "Practical Salinity Scale (PSS-78)"; SSS_IPRC_at_ARGO:standard_name = "sea_water_salinity"; $SSS_IPRC_at_ARGO:_FillValue = -999.f;$ float SST_AVHRR_at_ARGO(N_prof); SST_AVHRR_at_ARGO:long_name = "Daily OI AVHRR-only v2 SST (Reynolds et al., 2007) at Argo float location"; $SST_AVHRR_at_ARGO:units = "degree Celsius";$ SST_AVHRR_at_ARGO:standard_name = "sea_water_temperature"; $SST_AVHRR_at_ARGO:FillValue = -999.f;$ float U_EKMAN_GLOBCURRENT_at_ARGO(N_prof); U_EKMAN_GLOBCURRENT_at_ARGO:long_name = "15m depth Ekman current velocity: zonal component at Argo float location"; U_EKMAN_GLOBCURRENT_at_ARGO:units = "m s-1"; U_EKMAN_GLOBCURRENT_at_ARGO:_FillValue = -999.f; float V_EKMAN_GLOBCURRENT_at_ARGO(N_prof); V_EKMAN_GLOBCURRENT_at_ARGO:long_name = "15m depth Ekman current velocity: meridian component at Argo float location"; $V_EKMAN_GLOBCURRENT_at_ARGO:units = "m s-1";$ V_EKMAN_GLOBCURRENT_at_ARGO:_FillValue = -999.f; float U_GEOSTROPHIC_GLOBCURRENT_at_ARGO(N_prof); U_GEOSTROPHIC_GLOBCURRENT_at_ARGO:long_name = "Absolute geostrophic velocity: zonal component at Argo float location";



```
U_GEOSTROPHIC_GLOBCURRENT_at_ARGO:units = "m s-1";
     U_GEOSTROPHIC_GLOBCURRENT_at_ARGO:_FillValue = -999.f;
float V_GEOSTROPHIC_GLOBCURRENT_at_ARGO(N_prof);
     V_GEOSTROPHIC_GLOBCURRENT_at_ARGO:long_name = "Absolute geostrophic ve-
locity: meridian component at Argo float location";
     V_GEOSTROPHIC_GLOBCURRENT_at_ARGO:units = "m s-1";
     V_GEOSTROPHIC_GLOBCURRENT_at_ARGO:_FillValue = -999.f;
   // global attributes:
     :Conventions = "CF-1.6";
     :title = "ARGO Match-Up Database";
     :Satellite_product_name = "SMOS-L3-CATDS-CECOS-IFREMER-V2-1DAY-0.5DEG";
     :Satellite_product_spatial_resolution = "0.50 deg" ;
     :Satellite_product_temporal_resolution = "Daily";
    :Satellite_product_filename = " Daily_Composite/Half_degree/SSS_SMOS_L3_Daily_0.5deg_CATDS_CECOS_2010.05.01_V02.nc"
;
    :Match-Up_spatial_window_radius_in_km = 27.5;
     :Match-Up_temporal_window_radius_in_days = 0.5;
     :start_time = "20100114T000005Z";
     :stop_time = "20100118T235026Z";
     :northernmost_latitude = 77.676f;
     :sourthenmost_latitude = -66.423f;
     :westernmost_longitude = -179.219 f;
     :easternmost_longitude = 179.199f;
     :geospatial\_lat\_units = "degrees north";
     :geospatial_lat_resolution = "0.50 deg";
     :geospatial_lon_units = "degrees east";
     :geospatial_lon_resolution = "0.50 deg";
     :institution = "ESA-IFREMER-ODL";
     :project_name = "SMOS Pilote Mission Exploitation Platfrom (Pi-MEP) for salinity";
     :project_url = "https://pimep-project.odl.bzh";
     :license = "Pi-MEP data use is free and open";
     :product_version = "1.0";
     :keywords = "Oceans > Ocean Salinity > Sea Surface Salinity" :
    :acknowledgment = "Please acknowledge the use of these data with the following statement:
These data were provided by SMOS Pilote Mission Exploitation Platfrom (Pi-MEP) for salinity"
;
     :Source = "Daily_Composite/Half_degree/SSS_SMOS_L3_Daily_0.5deg_CATDS_CECOS_2010.05.01_V02.nc";
     :In_situ_data_source = "ftp://ftp.ifremer.fr/ifremer/argo/geo/";
     :references = "https://pimep-project.odl.bzh";
     :history = "Processed on 2018-04-18 using MDB_generator";
     :date\_created = "2018-04-18\ 17:09:30";
```

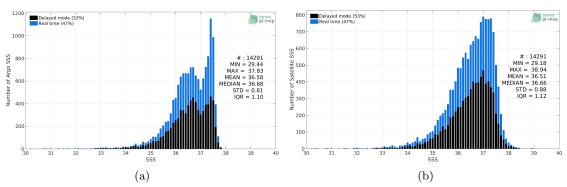
```
}
```



400 Delayed mode (53%) smos pi-mep 500 350 450 400 ا 300 م با Sd 350 of Match-u 200 700 - of Match 520 Number Number 200 150 100 100 50 50 2010 Distance to coast [km] (a) (b)

2.5 MDB characteristics for the particular in situ/satellite pairs2.5.1 Number of paired SSS data as a function of time and distance to coast

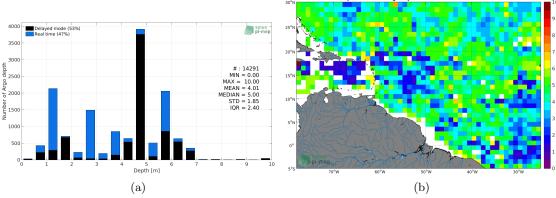
Figure 1: Number of match-ups between Argo and SMOS-L3-CATDS-CECOS-IFREMER-V2-1DAY-0.5DEG SSS as a function of time (a) and as function of the distance to coast (b) over the Amazon and Orinoco river plumes Pi-MEP region and for the full satellite product period.



2.5.2 Histograms of the SSS match-ups

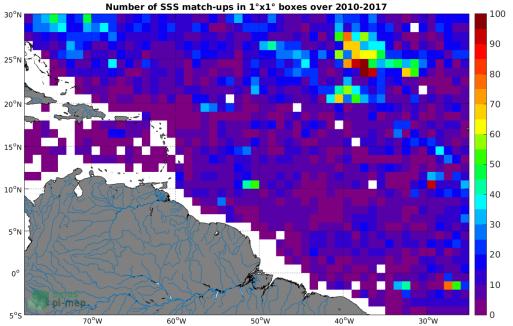
Figure 2: Histograms of SSS from Argo (a) and SMOS-L3-CATDS-CECOS-IFREMER-V2-1DAY-0.5DEG (b) considering all match-up pairs per bins of 0.1 over the Amazon and Orinoco river plumes Pi-MEP region and for the full satellite product period.





2.5.3 Distribution of in situ SSS depth measurements

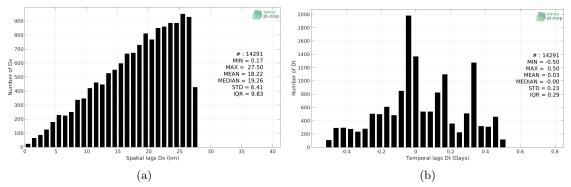
Figure 3: Histograms of the depth of the upper level SSS measurements from Argo in the Matchup DataBase for the Amazon and Orinoco river plumes Pi-MEP region (a) and temporal mean spatial distribution of pressure of the in situ SSS data over $1^{\circ}x1^{\circ}$ boxes and for the full satellite product period (b).



2.5.4 Spatial Distribution of Match-ups

Figure 4: Number of SSS match-ups between Argo SSS and the SMOS-L3-CATDS-CECOS-IFREMER-V2-1DAY-0.5DEG SSS product for the Amazon and Orinoco river plumes Pi-MEP region over $1^{\circ}x1^{\circ}$ boxes and for the full satellite product period.





2.5.5 Histograms of the spatial and temporal lags of the match-ups pairs

Figure 5: Histograms of the spatial (a) and temporal (b) lags between the time of the Argo measurements and the date of the corresponding SMOS-L3-CATDS-CECOS-IFREMER-V2-1DAY-0.5DEG SSS product.

3 MDB file Analyses

3.1 Spatial Maps of the Temporal mean and Std of in situ and satellite SSS and of the difference (Δ SSS)

In Figure 6, we show maps of temporal mean (left) and standard deviation (right) of the SMOS-L3-CATDS-CECOS-IFREMER-V2-1DAY-0.5DEG satellite SSS product (top) and of the Argo in situ dataset at the collected Pi-MEP match-up pairs. The temporal mean and std are gridded over the full satellite product period and over spatial boxes of size 1°x1°.

At the bottom of Figure 6, the temporal mean (left) and standard deviation (right) of the differences between the satellite SSS product and in situ data found at match-up pairs, namely Δ SSS(Satellite -Argo), is also gridded over the full satellite product period and over spatial boxes of size 1°x1°.



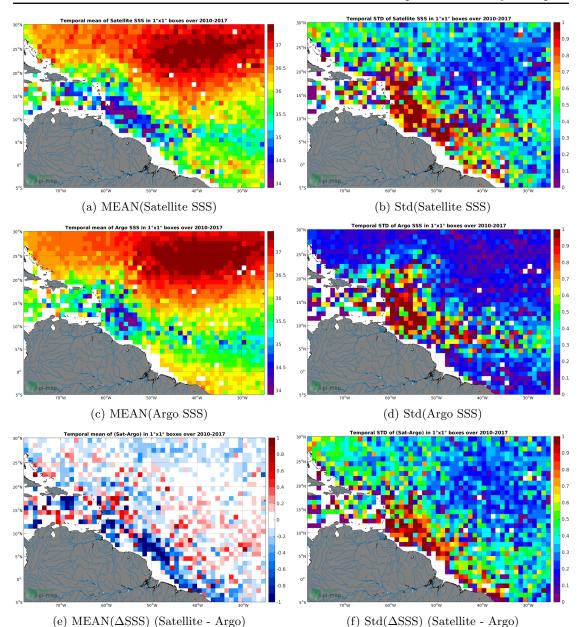


Figure 6: Temporal mean (left) and Std (right) of SSS from SMOS-L3-CATDS-CECOS-IFREMER-V2-1DAY-0.5DEG (top), Argo (middle), and of Δ SSS (Satellite - Argo). Only matchup pairs are used to generate these maps.

3.2 Time series of the monthly averaged mean and Std of in situ and satellite SSS and of the (Δ SSS)

In the top panel of Figure 7, we show the time series of the monthly averaged SSS estimated over the full Amazon and Orinoco river plumes Pi-MEP region for both SMOS-L3-CATDS-CECOS-IFREMER-V2-1DAY-0.5DEG satellite SSS product (in black) and the Argo in situ dataset (in



blue) at the collected Pi-MEP match-up pairs.

In the middle panel of Figure 7, we show the time series of the monthly averaged Δ SSS (Satellite - Argo) for the collected Pi-MEP match-up pairs and estimated over the full Amazon and Orinoco river plumes Pi-MEP region.

In the bottom panel of Figure 7, we show the time series of the monthly averaged standard deviation of the Δ SSS (Satellite - Argo) for the collected Pi-MEP match-up pairs and estimated over the full Amazon and Orinoco river plumes Pi-MEP region.

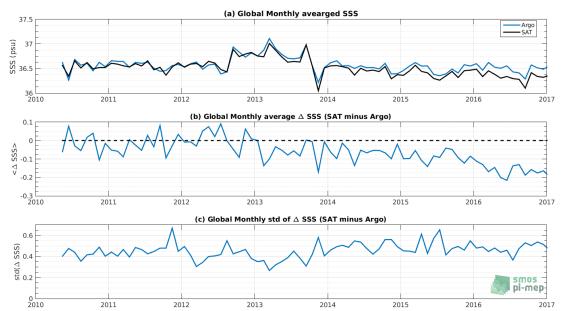


Figure 7: Time series of the monthly averaged mean SSS (top), mean Δ SSS (Satellite - Argo) and Std of Δ SSS (Satellite - Argo) over the Amazon and Orinoco river plumes Pi-MEP region considering all match-ups collected by the Pi-MEP platform.

3.3 Zonally-averaged Time-mean and temporal Std of in situ and satellite SSS and of the Δ SSS

In Figure 8 left panel, we show the zonally averaged time-mean SSS estimated at the collected Pi-MEP match-up pairs for both SMOS-L3-CATDS-CECOS-IFREMER-V2-1DAY-0.5DEG satellite SSS product (in black) and the Argo in situ dataset (in blue). The time mean is evaluated over the full satellite SSS product period.

In the right panel of Figure 8, we show the zonally averaged time-mean Δ SSS (Satellite - Argo) for all the collected Pi-MEP match-up pairs estimated over the full satellite product period.



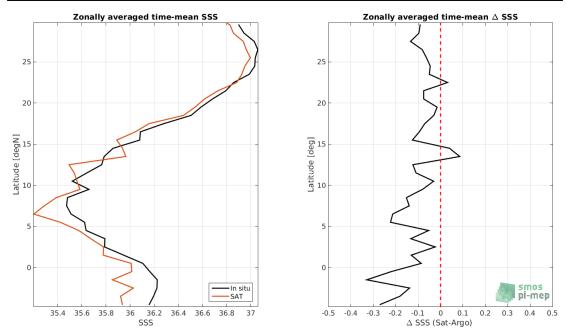
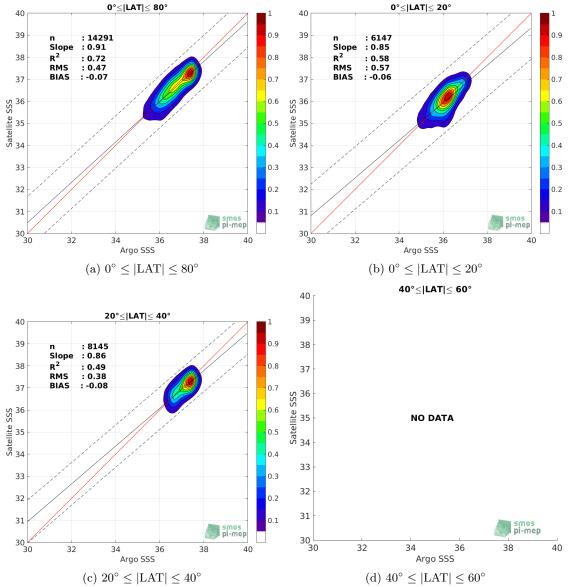


Figure 8: Left panel: Zonally averaged time mean SSS from SMOS-L3-CATDS-CECOS-IFREMER-V2-1DAY-0.5DEG (black) and from Argo (blue). Right panel: zonally averaged time-mean Δ SSS (Satellite - Argo) for all the collected Pi-MEP match-up pairs estimated over the full satellite product period.

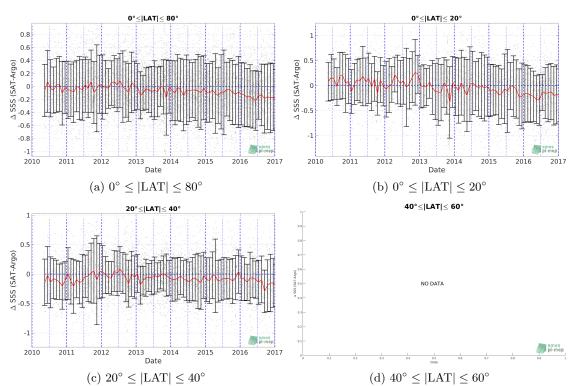




3.4 Scatterplots of satellite vs in situ SSS by latitudinal bands

Figure 9: Contour maps of the concentration of SMOS-L3-CATDS-CECOS-IFREMER-V2-1DAY-0.5DEG SSS (y-axis) versus Argo 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 \mathbb{R}^2 coefficient of the linear fit, the root mean square (RMS) and the mean bias between satellite and in situ data are indicated for each latitude band in each plots.





3.5 Time series of the monthly averaged mean and Std of the Δ SSS sorted by latitudinal bands

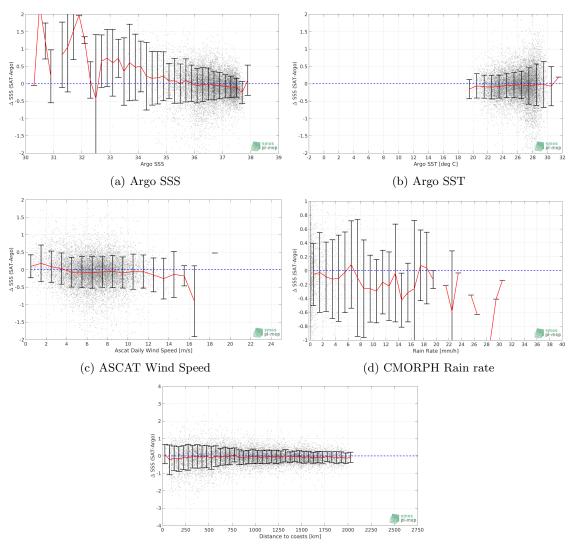
Figure 10: Monthly-average mean (red curves) Δ SSS (Satellite - Argo) and ±1 Std (black vertical thick bars) as function of time for all the collected Pi-MEP match-up pairs estimated over the Amazon and Orinoco river plumes Pi-MEP region and for the full satellite product period are shown for different latitude bands: (a) Latitude band 80°S-80°N, (b) latitude band 20°S-20°N, (c) Mid Latitude bands 40°S-20°S and 20°N-40°N and (d) Latitude bands 60°S-40°S and 40°N-60°N.

3.6 Δ SSS sorted as function of geophysical parameters

In Figure 11, we classify the match-up differences Δ SSS (Satellite - in situ) between SMOS-L3-CATDS-CECOS-IFREMER-V2-1DAY-0.5DEG and Argo SSS as function of the geophysical conditions at match-up points. The mean and std of Δ SSS (Satellite - Argo) 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.





(e) Distance to coast

Figure 11: Δ SSS (Satellite - Argo) sorted as function of Argo SSS values a), Argo SST b), ASCAT Wind speed c), CMORPH rain rate d) and distance to coast (e). In all plots the mean and Std of Δ SSS for each bin is indicated by the red curves and black vertical thick bars (±1 Std)

3.7 Δ SSS maps and statistics for different geophysical conditions

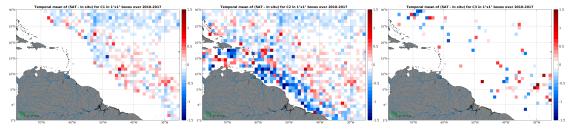
In Figures 12 and 13, we focus on sub-datasets of the match-up differences Δ SSS (Satellite - in situ) between SMOS-L3-CATDS-CECOS-IFREMER-V2-1DAY-0.5DEG and Argo for the following specific geophysical conditions:

- 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°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 <20m.
- 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).

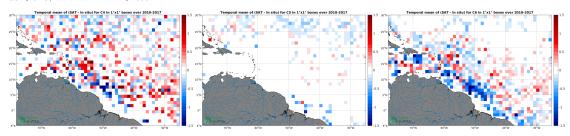
For each of these conditions, the temporal mean (gridded over spatial boxes of size $1^{\circ}x1^{\circ}$) and the histogram of the difference Δ SSS (Satellite - in situ) are presented.



(a) RR=0 mm/h, 3< U_{10} <12 m/s, SST>5°C, distance to coast > 800 km

(b) RR=0 mm/h, $3 < U_{10} < 12$ m/s

(C) RR>1mm/h and $U_{10} < 4m/s$



(d) MLD < 20m

(e) woa2013 SSS std<0.2

(f) woa2013 SSS std>0.2

Figure 12: Temporal mean gridded over spatial boxes of size $1^{\circ}x1^{\circ}$ of Δ SSS (SMOS-L3-CATDS-CECOS-IFREMER-V2-1DAY-0.5DEG - Argo) 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),MLD<20m (d),WOA2013 SSS Std<0.2 (e),WOA2013 SSS Std>0.2 (f).



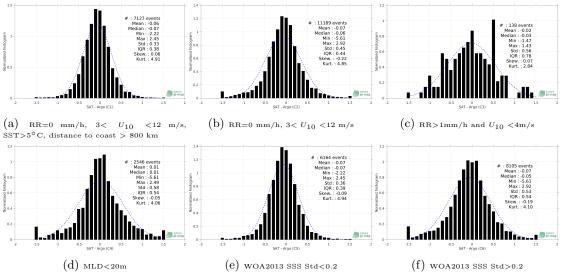


Figure 13: Normalized histogram of Δ SSS (SMOS-L3-CATDS-CECOS-IFREMER-V2-1DAY-0.5DEG - Argo) for 6 different subdatasets corresponding to: RR=0 mm/h, $3 < U_{10} < 12$ m/s, SST>5°C, distance to coast > 800 km (a), RR=0 mm/h, $3 < U_{10} < 12$ m/s (b), RR>1mm/h and $U_{10} < 4$ m/s (c), MLD<20m (d), WOA2013 SSS Std<0.2 (e), WOA2013 SSS Std>0.2 (f).

4 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 Δ SSS (Satellite - in situ) between SMOS-L3-CATDS-CECOS-IFREMER-V2-1DAY-0.5DEG and Argo derived over the Amazon and Orinoco river plumes Pi-MEP region and for the full satellite product period and 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 $U_{10} < 4m/s$
- C4: only pairs where MLD<20m
- C5: only pairs where WOA2013 SSS Std<0.2
- C6: only pairs at 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 $< 5^{\circ}$ C.



- C8b: only pairs where in situ SST is in the range [5, 15]°C.
- C8c: only pairs where in situ SST is $> 15^{\circ}$ 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 \triangle SSS (Satellite - Argo)

	Tuble	i. Statisti		22) 22				
Condition	#	Median	Mean	\mathbf{Std}	\mathbf{RMS}	IQR	\mathbf{r}^2	\mathbf{Std}^{\star}
all	14291	-0.06	-0.07	0.47	0.47	0.47	0.72	0.35
C1	7127	-0.07	-0.06	0.33	0.34	0.38	0.75	0.28
C2	11189	-0.06	-0.07	0.45	0.45	0.44	0.73	0.33
C3	138	-0.03	-0.02	0.56	0.56	0.78	0.60	0.59
C4	2546	0.01	0.01	0.58	0.58	0.54	0.76	0.40
C5	6164	-0.07	-0.07	0.36	0.36	0.39	0.58	0.29
C6	8105	-0.05	-0.07	0.53	0.54	0.54	0.67	0.40
C7a	387	-0.15	-0.22	0.77	0.80	0.89	0.45	0.65
C7b	4771	-0.04	-0.08	0.60	0.61	0.64	0.57	0.47
C7c	9133	-0.06	-0.05	0.35	0.36	0.40	0.76	0.30
C8a	0	NaN	NaN	NaN	NaN	NaN	NaN	NaN
C8b	0	NaN	NaN	NaN	NaN	NaN	NaN	NaN
C8c	14287	-0.06	-0.07	0.47	0.47	0.47	0.72	0.35
C9a	55	0.84	0.84	0.92	1.24	1.20	0.38	0.94
C9b	9142	-0.02	-0.05	0.53	0.53	0.54	0.56	0.40
C9c	5094	-0.10	-0.11	0.30	0.32	0.35	0.18	0.26

Table 2 presents statistics of Δ SSS (Satellite - Argo) as Table 1 but for Argo delayed mode only.

Table 2: Statistics of \triangle SSS (Satellite - Argo) - Delayed mode								
Condition	#	Median	Mean	\mathbf{Std}	RMS	IQR	\mathbf{r}^2	\mathbf{Std}^{\star}
all	7636	-0.04	-0.05	0.47	0.48	0.48	0.70	0.36
C1	3848	-0.06	-0.05	0.35	0.36	0.40	0.74	0.30
C2	5818	-0.05	-0.05	0.45	0.45	0.46	0.70	0.34
C3	92	-0.01	0.00	0.58	0.58	0.80	0.54	0.63
C4	1481	0.03	0.04	0.59	0.59	0.57	0.72	0.41
C5	3339	-0.07	-0.07	0.38	0.38	0.41	0.57	0.30
C6	4285	-0.01	-0.03	0.54	0.54	0.54	0.65	0.40
C7a	179	-0.15	-0.18	0.81	0.83	1.04	0.44	0.79
C7b	2389	-0.02	-0.06	0.61	0.61	0.62	0.53	0.47
C7c	5068	-0.05	-0.04	0.37	0.37	0.42	0.75	0.31
C8a	0	NaN	NaN	NaN	NaN	NaN	NaN	NaN
C8b	0	NaN	NaN	NaN	NaN	NaN	NaN	NaN
C8c	7635	-0.04	-0.05	0.47	0.48	0.48	0.70	0.36
C9a	17	1.12	1.14	0.76	1.36	1.39	0.52	1.06
C9b	5191	0.01	-0.01	0.52	0.52	0.53	0.56	0.39
C9c	2428	-0.13	-0.13	0.31	0.34	0.35	0.18	0.26

ble 2: Statistics of Δ SSS (Satellite - Argo) - Delayed mode



	Table 3	3: Statisti	cs of ΔS	SS (Sa	tellite -	ISAS)		
Condition	#	Median	Mean	Std	RMS	IQR	\mathbf{r}^2	\mathbf{Std}^{\star}
all	14285	-0.05	-0.07	0.45	0.45	0.47	0.74	0.35
C1	7127	-0.05	-0.06	0.32	0.33	0.37	0.75	0.28
C2	11186	-0.05	-0.07	0.43	0.44	0.44	0.74	0.33
C3	138	-0.03	-0.07	0.54	0.54	0.81	0.60	0.60
C4	2543	-0.04	-0.08	0.56	0.57	0.52	0.77	0.39
C5	6163	-0.06	-0.07	0.35	0.36	0.39	0.59	0.29
C6	8100	-0.04	-0.07	0.51	0.51	0.53	0.69	0.40
C7a	384	-0.09	-0.15	0.77	0.79	0.92	0.45	0.67
C7b	4769	-0.03	-0.08	0.58	0.58	0.62	0.60	0.46
C7c	9132	-0.05	-0.05	0.34	0.35	0.40	0.76	0.30
C8a	0	NaN	NaN	NaN	NaN	NaN	NaN	NaN
C8b	0	NaN	NaN	NaN	NaN	NaN	NaN	NaN
C8c	14281	-0.05	-0.07	0.45	0.45	0.47	0.74	0.35
C9a	55	-0.87	-0.91	1.17	1.48	1.77	0.17	1.23
C9b	9136	-0.02	-0.05	0.50	0.50	0.54	0.59	0.40
C9c	5094	-0.08	-0.09	0.30	0.32	0.36	0.19	0.27

For the same conditions, Table 3 presents statistics of Δ SSS (Satellite - ISAS). Only ISAS SSS values with PCTVAR<80% are used to derive the statistics.

Numerical values can be downloaded as csv files for Table 1, Table 2 and Table 3.

References

- Argo. Argo float data and metadata from Global Data Assembly Centre (Argo GDAC), 2000. doi: 10.17882/42182.
- Abderrahim Bentamy and Denis Croize Fillon. Gridded surface wind fields from Metop/ASCAT measurements. Int. J. Remote Sens., 33(6):1729–1754, March 2012. ISSN 1366-5901. doi: 10.1080/01431161.2011.600348.
- Abderrahim Bentamy, Semyon A. Grodsky, James A. Carton, Denis Croizé-Fillon, and Bertrand Chapron. Matching ASCAT and QuikSCAT winds. J. Geophys. Res., 117(C2), February 2012. ISSN 0148-0227. doi: 10.1029/2011JC007479. C02011.
- Jaqueline Boutin, Y. Chao, W. E. Asher, T. Delcroix, R. Drucker, K. Drushka, N. Kolodziejczyk, T. Lee, N. Reul, G. Reverdin, J. Schanze, A. Soloviev, L. Yu, J. Anderson, L. Brucker, E. Dinnat, A. S. Garcia, W. L. Jones, C. Maes, T. Meissner, W. Tang, N. Vinogradova, and B. Ward. Satellite and In Situ Salinity: Understanding Near-Surface Stratification and Sub-footprint Variability. *Bull. Am. Meterol. Soc.*, 97(8):1391–1407, 2016. ISSN 1520-0477. doi: 10.1175/bams-d-15-00032.1.
- Clément de Boyer Montégut, Gurvan Madec, A. S. Fischer, A. Lazar, and D. Ludicone. Mixed layer depth over the global ocean: An examination of profile data and a profile-based climatology. J. Geophys. Res., 109(C12):C12003, December 2004. ISSN 0148-0227. doi: 10.1029/2004jc002378.
- Clément de Boyer Montégut, Juliette Mignot, Alban Lazar, and Sophie Cravatte. Control of salinity on the mixed layer depth in the world ocean: 1. General description. J. Geophys. Res., 112(C6):C06011, June 2007. ISSN 0148-0227. doi: 10.1029/2006jc003953.



- Ralph R. Ferraro. SSM/I derived global rainfall estimates for climatological applications. J. Geophys. Res., 1021:16715–16736, 07 1997. doi: 10.1029/97JD01210.
- Ralph R. Ferraro, Fuzhong Weng, Norman C. Grody, and Limin Zhao. Precipitation characteristics over land from the NOAA-15 AMSU sensor. *Geophys. Res. Lett.*, 27(17):2669–2672, 2000. doi: 10.1029/2000GL011665.
- Fabienne Gaillard, E. Autret, V. Thierry, P. Galaup, C. Coatanoan, and T. Loubrieu. Quality Control of Large Argo Datasets. J. Atmos. Oceanic Technol., 26(2):337–351, 2012/10/10 2009. doi: 10.1175/2008JTECH0552.1.
- Fabienne Gaillard, Thierry Reynaud, Virginie Thierry, Nicolas Kolodziejczyk, and Karina von Schuckmann. In Situ-Based Reanalysis of the Global Ocean Temperature and Salinity with ISAS: Variability of the Heat Content and Steric Height. J. Clim., 29(4):1305–1323, February 2016. ISSN 1520-0442. doi: 10.1175/jcli-d-15-0028.1.
- Robert J. Joyce, John E. Janowiak, Phillip A. Arkin, and Pingping Xie. CMORPH: A Method that Produces Global Precipitation Estimates from Passive Microwave and Infrared Data at High Spatial and Temporal Resolution. J. Hydrometeorol., 5(3):487–503, June 2004. ISSN 1525-7541. doi: 10.1175/1525-7541(2004)005(0487:camtpg)2.0.co;2.
- Nicolas Kolodziejczyk, Gilles Reverdin, and Alban Lazar. Interannual Variability of the Mixed Layer Winter Convection and Spice Injection in the Eastern Subtropical North Atlantic. J. Phys. Oceanogr., 45(2):504–525, Feb 2015. ISSN 1520-0485. doi: 10.1175/jpo-d-14-0042.1.
- Christian Kummerow, Y. Hong, W. S. Olson, S. Yang, R. F. Adler, J. McCollum, R. Ferraro, G. Petty, D-B. Shin, and T. T. Wilheit. The Evolution of the Goddard Profiling Algorithm (GPROF) for Rainfall Estimation from Passive Microwave Sensors. J. Appl. Meteorol., 40(11): 1801–1820, 2001. doi: 10.1175/1520-0450(2001)040(1801:TEOTGP)2.0.CO;2.