Evaluation of Arctic Ocean surface salinities from SMOS against a regional reanalysis and in situ data

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Abstract

Recently two gridded Sea Surface Salinity (SSS) products that cover the Arctic Ocean have been derived from the European Space Agency's (ESA) Soil Moisture and Ocean Salinity (SMOS) mission: one developed by the Barcelona Expert Centre (BEC) and the other developed by the Ocean Salinity Expertise Center of the Centre Aval de Traitement des Données SMOS at IFREMER (CEC). The uncertainties of these two SSS products are quantified during the period of 2011-2013 against other SSS products: one data assimilative regional reanalysis, one data-driven reprocessing in the framework of the Copernicus Marine Environment Monitoring Services (CMEMS), two climatologies: the 2013 World Ocean Atlas (WOA) and the Polar science center Hydrographic Climatology (PHC), and in-situ datasets, both assimilated and independent. The CMEMS reanalysis comes from the TOPAZ4 system which assimilates a large set of ocean and sea-ice observations using an Ensemble Kalman Filter (EnKF). Another CMEMS product is the Multi-OBservations reprocessing (MOB), a multivariate objective analysis combining in-situ data with satellite SSS. The monthly root mean squared deviations (RMSD) of both SMOS products, compared to the TOPAZ4 reanalysis, reach 1.5 psu in the Arctic summer, while in the winter months the BEC SSS is closer to TOPAZ4 with a deviation of 0.5 psu. The comparison of CEC satellite SSS against in-situ data shows too fresh Atlantic waters in the Barents Sea, the Nordic seas, and in the northern North Atlantic Ocean, consistently with the abnormally fresh deviations against TOPAZ4. When compared against independent in-situ data in the Beaufort Sea, the BEC product shows the smallest bias (<0.1 psu) in summer and the smallest RMSD (1.8 psu), although all six SSS products share a common challenge to represent fresher water masses (<24 psu). Along the Norwegian coast and at the southwestern coast of Greenland, the BEC SSS shows smaller errors than TOPAZ4 and indicates the potential value of assimilating the satellite-derived salinity in this system.

Keywords: Arctic Ocean; sea surface salinity; SMOS; reanalysis;

1. Introduction

The sea surface salinity (SSS) plays a key role in tracking processes in the global water cycle through precipitation, evaporation, runoff, and sea-ice thermodynamics (Vialard and Delecluse, 1998; Sumner and Belaineh, 2005; Vancoppenolle et al., 2009; Yu, 2011). SSS is known to impact the oceanic upper mixing significantly (Latif et al., 2000; de Boyer Montegut et al., 2004; Maes et al., 2006; Furue et al., 2018) and via its effect on the surface layer density (Johnson et al, 2012). The SSS also affects the decadal variability of hydrography in the upper waters of the North Atlantic (Reverdin et al., 1997). Using a coupled atmosphere-ocean model and an observed SSS climatology dataset, Mignot and Frankignoul (2003) attributed the interannual variability of the Atlantic SSS to two factors: anomalous Ekman advection and the freshwater flux. Additionally, the increased melting of glaciers and sea-ice in the Arctic (McPhee et al., 1998; Macdonald et al., 1999) leads to significant changes in the salinity distribution and fresh water pathways (Steele and Ermold, 2004; Morison et al., 2012). The freshwater flux is regarded as one of the least constrained parameters in ocean models due to poorly known river discharge, precipitation, and glacial/sea-ice melt (e.g., Tseng et al., 2016; Furue et al., 2018). In ocean models the sea-surface freshwater flux is often adjusted directly or the SSS is restored to its corresponding climatological value to avoid salinity drift.

Monitoring SSS from space is crucial for understanding the global water cycle and the ocean dynamics, especially in the Arctic Ocean where our knowledge of the SSS variability is limited due to non-homogenous and sparse in-situ data. The European Space Agency's (ESA) Soil Moisture and Ocean Salinity (SMOS) satellite, launched in November 2009, consists of the Microwave Imaging Radiometer using Aperture Synthesis (MIRAS) instrument, a passive 2-D interferometric radiometer operating in L-band (1.4 GHz, 21 cm), that measures the brightness temperature (BT) emitted from the Earth. The L-band microwave is highly sensitive to water salinity, which influences the dielectric constants in the sea, and is less susceptible to atmospheric or vegetation-induced attenuation than higher frequency measurements (Font et al., 2010; Kerr et al., 2010; Mecklenburg et al., 2012). Committed to provide global salinities averaged over 10-30 days with an accuracy of 0.1 psu in the open ocean, ESA provides the MIRAS data into SMOS Level 1 (L1) and Level 2 (L2) products through a set of sequential processors (Mecklenburg et al., 2012; ESA, 2017).

Over the ocean, Level 2 products (L2OS) are comprised of three different ocean salinities, together with the BTs at the top of atmosphere and at the sea surface, distributed by ESA with swath-based format (e.g., SMOS Team, 2016; ESA, 2017). As a result of the efforts of the national agencies in France and Spain respectively, two Level 3 (L3) data products of SSS are freely available, which are independently developed by the Ocean Salinity Expertise Center (CECOS) of the Centre Aval de Traitement des Données SMOS at IFREMER and the Barcelona Expert Centre. These two SMOS products have successfully resolved the Agulhas salinity front (D'Addezio et al., 2016) and proven useful for the estimating precipitation (Supply et al., 2018). The work of Olmedo et al. (2018) quantitatively evaluate the accuracy of the SMOS Arctic and sub-Arctic SSS to less than 0.35 psu, but this evaluation against Argo data was limited by the lack of data in the Arctic proper. The present study thus investigates the accuracy of these two SMOS SSS products in the Arctic Ocean.

A good estimate of surface salinity is a necessary step towards the knowledge of the three-dimensional water mass properties, for which data assimilation and optimal interpolation methods must be invoked. In a recent study, Uotila et al. (2018) investigated the Arctic salinity in ten reanalysis products and found disagreements within them regarding the seasonal cycle in the upper layer (0-100 m; Figure 12 of Uotila et al., 2018). Note that the full assessment of the Arctic SSS products has been hindered by the extreme paucity of in-situ data in the Arctic. The SSS data from the SMOS mission should in principle allow the evaluation of salinity on a basin scale. In this study, we use two SSS products available from the Copernicus Marine Environment Monitoring Service (CMEMS). The first is the regional Arctic CMEMS reanalysis (ARCTIC-REANALYSIS-PHYS-002-003) from the TOPAZ4 assimilation system, which is a coupled ocean and sea-ice data assimilation system using the Ensemble Kalman filter (EnKF) to assimilate the various ocean and sea-ice observations (e.g., Xie et al., 2017). The second is the CMEMS multivariate optimal interpolation reprocessing (MULTIOBS_GLO_PHY_REP_015_002, Droghei et al., 2018). The latter product directly merges in-situ data with satellite measurements including SMOS without the use of a model and is therefore a reprocessing rather than a reanalysis. There are four other global reanalysis products under CMEMS, but understanding well their differences requires an intimate knowledge of their setup, and is out of scope of the present study.

We assess the quantitative deviations of Arctic SSS among the two SMOS products and the two CMEMS products, together with two climatology datasets: WOA13 (version 2.0 of World Ocean Atlas of 2013; Zweng et al., 2013) and the older PHC (Polar Science Center Hydrographic Climatology version 3.0; Steele et al., 2001). We further extend the evaluation using available in-situ salinity observations during the years 2011-2013 from different data sources. Can the evaluation against the in-situ data also shed light on the uncertainty of the SMOS products? Can it also give useful information needed for the assimilation of the SMOS SSS products into an Arctic ocean forecast/reanalysis system?

The paper is organized as follows: Section 2 describes all SSS products and the in-situ datasets. The monthly mean SSS from these six products are intercompared and monthly differences from the TOPAZ SSS are analyzed in Section 3. Section 4 evaluates the SSS products against in-situ data, which are divided between assimilated and independent data. A summary of this study is provided in Section 5.

2. Data description

2.1 Sea surface salinity from SMOS

The SSS retrieval from SMOS is subject to biases originating from various non-geophysical sources such as the so-called land-sea contamination and the latitudinal biases, mainly caused by the thermal drift of the instrument. A particular challenge in the Arctic is the sea-ice edge because of ice-ocean contamination. Based on different statistical approaches, match-up criteria, and SMOS data filtering flags, two centers have developed separate processing chains producing a Level 3 SSS product on a regular grid. These two SSS products are hereafter named respectively CEC and BEC in this study, evaluated during the three years of 2011-2013 (see Table 1).

- *The BEC product*

The latest regional Arctic product (version 2.0) from BEC is available from http://bec.icm.csie.es since December 2018 (last access: March 2019). The BEC SSS product was generated from ESA L1B (v620) products, and accumulates salinity data over 9 days with a spatial grid resolution of 25 km. With respect to its previous version, a systematic bias in the retrieved salinity is corrected by computing the SMOS climatology (the most probable value for a given lat-lon, incidence angle and across-swath distance) which is substituted by a reference value from WOA13. In addition, a temporal bias correction has been refined in this version using near-surface Argo salinity to compute regional averages (see the details in Olmedo et al., 2018).

- *The CEC product*

The third version of LOCEAN SMOS SSS L3 maps (L3_DEBIAS_LOCEAN_v3) was released by the CECOS in July 2018. Every 4 days, the SSS maps averaged over 9 days are released on ftp.ifremer.fr (last access: December 2018). This product uses the Equal-Area Scalable Earth Grid (EASE-Grid) which has limited grid distortion and a spatial resolution of 25km. Using a Bayesian retrieval approach (Kolodzejczyk et al., 2016), the SMOS systematic errors in the vicinity of continents are discarded o improve the product quality. Further, a 'de-biasing' method (Boutin et al., 2018) has been applied in this version of the CEC product, in which the non-Gaussian distribution of SSS is taken into account, refining the latitudinal correction at high latitude, and preserving the naturally seasonal variability of SSS.

2.2 Sea surface salinity from two CMEMS products

- *The TOPAZ4 Arctic MFC reanalysis*

TOPAZ4 uses the version 2.2 of Hybrid Coordinate Ocean Model (HYCOM, Chassignet et al., 2003; Bertino and Lisæter, 2008) coupled with a simple thermodynamic sea ice model (Drange and Simonsen, 1996) in which the elastic-viscous-plastic rheology describes the sea ice dynamics (Hunke and Dukowicz, 1997). The model domain covers the Arctic Ocean and the north Atlantic Ocean with a horizontal resolution of 12-16 km. In order to obtain an accurate and dynamically consistent reanalysis in the Arctic Ocean, the deterministic EnKF (DEnKF; Sakov and Oke, 2008) was implemented in TOPAZ with a dynamical ensemble of 100 members all driven by perturbed 6-hourly atmosphere forcing from ERA interim (Simmons et al., 2007). The perturbations of precipitations are following a log-normal probability distribution and conserve the ensemble-average total precipitation.

Along the model lateral boundaries in the South Atlantic and in Bering Strait, the temperature and salinity are relaxed to a combined climatology data from PHC and

WOA. The river discharges are treated as an additional mass and a negative salinity flux. Near the surface, to avoid the salinity drift (Tseng et al., 2016; Furue et al., 2018), a weak relaxation to the same combined climatological SSS with 30 days decay is used as most ocean models, but restricted to the areas where the difference to climatology is smaller than 0.5 psu. The EnKF assimilates various ocean and sea-ice observations (e.g., Xie et al., 2016, 2018) into a multivariate state update of the HYCOM model.

The understanding for the uncertainty of the TOPAZ4 SSS has been hindered by poor coverage of in-situ data over the Arctic domain, although Xie et al. (2017) had comprehensively assessed the TOPAZ4 reanalysis during 1991-2013 against various types of ocean and sea-ice observations. For the sake of brevity, the TOPAZ4 reanalysis SSS is named TP4 hereafter.

- *SSS from the Multi-OBservations dataset*

The CMEMS product of MULTIOBS_GLO_PHY_REP_015_002 combines the SSS observations from in-situ and satellite data, using optimal interpolation (OI, Buongiorno Nardelli et al., 2016; Verbrugge et al., 2018) at weekly interval on a 0.25° x 0.25° regular grid. The main datasets used during the OI processing are: 1) the quality controlled in-situ data, COriolis dataset for Re-Analysis (CORA, Cabanes et al., 2013) distributed through CMEMS; 2) the objectively analyzed SSS and SST data generated from CORA, also distributed by CMEMS, which uses the WOA 2013 climatology as first guess and has been upscaled to the MOB grid as another first guess of the multidimensional OI; 3) The SMOS L3 binned (L3bin) data reprocessed by SMOS-BEC at 0.25° grid, although the previous version 1.0 of the product mentioned above; 4) The daily Reynolds L4 AVHRR_OI Global blended SST product on a 0.25° grid. This product is called MOB hereafter.

2.3 Surface salinity from in-situ data

The in-situ SSS data are acquired here from three quality-controlled datasets. The first data source is CORA from CMEMS (product id: INSITU_GLO_TS_REP_OBSERVATIONS_013_001_b), also used in the MOB SSS. CORA contains temperature and salinity profiles from various in-situ data sources (Cabanes et al., 2013). Since 2013, the CORA dataset has been updated every year and includes all the Argo float profiles, moorings, gliders, Ice-Tethered Profilers (ITP; Toole et al., 2011), XBT, CTD, and XCTD data. The latest version of the dataset, CORA5.1, covers the period of 1950-2016. Figure 1a shows the distribution of SSS (averaged over 0-8 m depth) observations from CORA5.1 (total 69,246 observations) over the domain north of 52°N during the years 2011-2013.

The second source of in-situ data is from the Beaufort Gyre Experiment Project (BGEP, http://www.whoi.edu/website/beaufortgyre/background, last access: 14th December 2018). In order to monitor the natural variabilities of the Beaufort Sea in the Canada Basin, BGEP maintains moorings since 2003 and acquires in-situ measurements over the Beaufort Sea region every summer. Symbols (anti-triangle, square, and star) shown in Fig. 1b indicate the locations of valid SSS observations obtained from BGEP. The in-situ dataset used in this study is obtained from the GO-SHIP (the Global Ocean Ship-based Hydrographic Investigations Program, Talley et al., 2017) database under the Climate Variability and Predictability Experiment (CLIVAR). The SSS observations in the Beaufort Sea are extracted from CLIVAR/GO-SHIP data with EXPOCODE (33HQ20111003 and 33HQ20121005, ref. Mathis and Monacci, 2014), which are available from https://cdiac.ess-dive.lbl.gov/ftp/oceans/CARINA/Healy/ (last access: 18th December 2018). All the valid salinity profiles are averaged within the upper 8 m layer, in order to match at best with the satellite SSS measurements. Contrarily to the CORA data, both BGEP and CLIVAR data are independent from all the evaluated datasets.

3. Intercomparison of monthly SSS fields

Prior to the intercomparison of different SSS products, all the gridded products from satellite, reanalysis and climatology have been mapped on the same grid used in the TP4 model by a “nearest neighbor” interpolation. To quantitatively evaluate the SSS deviation in the Arctic, the bias and the root mean square deviation (RMSD) are defined by

$$\mathrm{Bias} = \frac{1}{\mathbf{p}}\sum_{\mathbf{i=1}}^{\mathbf{p}}(\mathbf{H_i x_i^f} - \boldsymbol{s}_{\mathbf{i}}) \qquad (1)$$

$$\mathrm{RMSD} = \sqrt{\frac{1}{\mathbf{p}}\sum_{\mathbf{i=1}}^{\mathbf{p}}(\mathbf{H_i x_i^f} - \mathbf{s_i})^2} \qquad (2)$$

Where p is the length of the time series, $\mathbf{x}_i^{\mathrm{f}}$ is the valid salinity from different sources at the *i*th time, compared to the reference salinity field $\mathbf{s}_i$. $\mathbf{H}_i$ is the observation operator projecting $\mathbf{x}_i^{\mathrm{f}}$ onto $\mathbf{s}_i$.

- *Monthly mean comparison of SSS*

Figure 2 shows the monthly mean Arctic SSS in March from the six products. Notable differences in the two SMOS products appear in the Nordic Seas, Barents Sea, and around the Labrador Sea in the northern North Atlantic Ocean. At first sight, the large-scale SSS features from SMOS products are similar to the other products. However, the CEC SSS is fresher (as shown by the isolines of 35 psu) compared to the BEC, TP4, MOB and both climatologies. The location of the sea-ice edge in the two SMOS products match comparatively well with the TP4 reanalysis (Fig. 2a, d). In sea-ice covered region, TP4 shows a gradual decrease in SSS from the European to the American sector, with two minima near the Beaufort Sea and the East Siberian Sea (ESS; Fig. 2b) consistently with the PHC (Fig. 2c). Those are unclear in the MOB and WOA (Fig. 2e, f), especially the SSS minimum in the Beaufort Sea. The latter two products also show artificial projection artefacts around the North Pole. Figure 3 shows the corresponding SSS fields in September. In comparison to the March situation, the BEC and CEC SSS in the Nordic Seas are both less saline, indicated by the 35 psu isoline. The sea ice masking of the two SMOS products differ considerably in the Canadian Basin and in the Arctic marginal seas. Although the SSS of TP4, MOB, PHC and WOA agree relatively well in the northern Atlantic Ocean, the discrepancies become dramatic in ice-covered areas. Below the ice or near the sea-ice edge (denoted by the brown thick line in Fig. 2 and 3), TP4 and PHC share common features, which can be explained by the model restoring to PHC. On the other hand, the MOB and WOA differ significantly in spite of WOA being used as input to the MOB. Short of a universal reference for Arctic SSS, the monthly mean SSS deviations will be quantified using TP4 as a reference.

- *Deviation analysis of monthly SSS referred to TP4*

Figure 4 and Figure 5 show the deviations of the monthly mean SSS of the five products with reference to the TP4 SSS in August and September respectively. In August, the two SMOS products (Fig. 4a, c) show coherently negative deviations (~2 psu) in the marginal seas of the Beaufort Sea, the ESS, the Laptev Sea, and the Kara Sea. In the North Atlantic Ocean, away from the sea-ice edge, the deviation of the BEC from TP4 is lower (bias less than 0.5 psu). Focusing on the Arctic domain (>60°N), the mean deviation of the BEC SSS is -0.87 psu and its root mean square is 1.75 psu. The CEC SSS shows considerable negative deviations over 1 psu in the northern Atlantic, from north of Denmark Strait to the west coast of Ireland. This is remarkably different from the BEC, and does not discern the subpolar from the subtropical waters there (Hátún et al., 2005). The deviations of MOB and the two climatology products are comparatively small in the open ocean of the northern Atlantic (Fig. 4b, e). Near and below the sea-ice cover, the deviations are much larger, particularly both the MOB and WOA show strong saline anomalies (> 1 psu) in the Eurasian basin and low anomalies in the American basin.

In September, the SSS deviations of BEC, MOB, PHC and WOA show similar fresher patterns as in August, but the CEC deviations becomes surprisingly positive around the ice edge. The SSS deviation of CEC, averaged over the Arctic domain (>60°N), swaps from -0.42 to 0.42 psu from one month to the next one. The seasonal evolution of monthly SSS deviations from TP4 for all five remaining products, averaged over the Arctic, are shown in Fig. 6. Among the five products, the MOB shows the strongest seasonality with the RMSD higher than 4 psu in July and August (Fig. 6a), and close to 2 psu in winter. The spatially averaged deviation is much fresher than TP4, over -2 psu in summer and -0.5 psu in winter (Fig. 6b). The deviations of the two SMOS SSS show a relatively smaller seasonality (Fig. 6a). During summer months, their RMSDs reach 1.5 psu (Fig. 6a) in summer, and they decrease to 0.5 and 1.0 psu (for BEC and CEC respectively). Throughout the whole year, the BEC RMSDs (Fig. 6a) are consistently smaller than that of CEC, and the seasonal cycles are different. This shows that the BEC SSS is closest to TP4, although it is overall fresher in the Summer.

4. Evaluation against in-situ observations

The misfits of the six SSS products from SMOS, CMEMS and climatologies are calculated as in Eqs. (1) and (2) against the pointwise in-situ observations described in Section 2.3. For TP4, the SSS evaluation is conducted on the same model day as the in-situ observations. Owing to the fact that the SSS from BEC, CEC and MOB are averaged over either 9 days or one week (see Table 1), the product dates at the center of the averaging window lag 5 or 4 days compared to the observation date. For PHC and WOA, the in-situ observations are sorted to monthly bins and evaluated for each month. The quantitative evaluation is divided into two main sections starting with dependent and then independent observations.

4.1 Against SSS from CORA5.1

As shown in Fig. 1a, the distribution of SSS observations from CORA5.1 over the Arctic is very inhomogeneous during the three years. Due to this, the evaluation of the gridded SSS products against in-situ observations is restricted to the observation-rich regions. The SSS misfits bias and RMSD for the six products are reported in Table 2 according to the eight Arctic sub-regions defined previously (Figure 1a). The observations are displayed on scatterplots (Figure 7 and 8) to exhibit their uncertainties for fresh and saline waters in different areas.

- *Central Arctic*

Figure 7 shows the SSS products compared with discrete observations in the central Arctic (sub-regions S0, S1, S2, and S3). The observed SSS in S0 and S1 are mainly from the ITP at a minimal depth of 8 m. Around the North Pole (S0), where the satellite SSS are absent, the TP4 reanalysis and MOB reprocessing show opposite biases: +0.48 psu and -0.52 psu respectively (Table 2). The two climatologies used by them, PHC and WOA respectively, also show opposite biases. Considering the latter climatologies, both SSS scatterplots shows a fresh bias for high salinity water (>33 psu) and a saline bias for low salinity water (<31 psu).
In the Canadian basin (in S1), the two climatological SSSs show an obvious gap in comparison to the ITP observations. Comparing to the fresh in-situ SSS from 24 to 30 psu, the PHC has strong saline bias (from 2 to more than 5 psu). On the other hand, the WOA shows both a fresh bias for relatively high salinity water (>28 psu) and saline bias for fresher water (<26 psu). Owing to the different time periods (Table 1) of the in-situ data they used, this result confirms the freshening of the Canadian basin since in the 1990s (Morison et al., 2012).
In the S1 sub-region, the satellite SSS from BEC and CEC have only 20 and 42 data points for evaluation respectively. The resulting scatterplots show a significantly

positive salinity bias (>4 psu) for fresh waters (<27 psu). For relatively higher salinity water (> 27 psu), the CEC has a stronger saline bias than the BEC.

In the Kara Sea (sub-region S2), the TP4 SSS has the smallest RMSD at 1.7 psu, which is significantly smaller than other products. The scatterplot also shows a good linear relationship between the TP4 and the in-situ SSS, while other products generally show fresh biases, indicating that the SSS variability in the Kara Sea is well captured by TP4. In the Barents Sea (sub-region S3), TP4 gives as well the smallest misfit (RMSD: 0.34 psu; bias: -0.14 psu). The SSS scatterplots exhibits linear relationships for all products except the CEC, which underestimates the Atlantic water SSS.

- *Northern North Atlantic and Nordic Seas*

Figure 8 shows the paired scatterplots of the six SSS products in the subpolar seas from sub-regions S4 to S7 (see Fig. 1a). In S4 and S5, the bias of SSS products is relatively small, less than 0.15 psu (Table 2), except for CEC in S4 and TP4 in S5, both too saline by 0.2 psu. The scatterplots further indicate that low salinity waters are too saline in all SSS products in S4 (<31 psu) and in S5 (<28 psu). Meanwhile, the respective bias and RMSD of the SSS products are less than 0.1 psu and 0.43 psu respectively, except for the CEC in S6 and S7. The MOB SSS has the smallest salinity bias. Among the eight regions compared here (S0 to S7), the SSS bias is lowest in S6 (Irminger Sea).

Over the northern North Atlantic and the Nordic seas, Fig. 9 shows maps of the mean SSS deviation for each product during the period 2011-2013. Considerable negative biases (<-0.2 psu) are found in the CEC, whereas the MOB and WOA have the smallest bias, less than 0.02 psu (Fig. 9 d, e, f). The SSS products from BEC, TP4 and PHC (Fig. 9 a, b, c) have slightly higher bias (~0.05 psu) in comparison to the MOB and WOA. On average, the BEC bias is only -0.04 psu, much smaller than that of the CEC (<-0.2 psu). Focusing on the BEC SSS, Fig 9a shows that while a fresh bias dominates the Nordic Seas, the product is too saline in the northern North Atlantic and the North Sea.

The inter-comparison of the biases against the in-situ data in Fig. 9a and 9b exhibits two strong positive biases of TP4 along the Norwegian coast and along the West Greenland coast. Notably, the BEC has smaller bias along both coasts, although it has a slightly saline bias offshore. This indicates potential benefits of the BEC SSS for the TOPAZ system along the Norwegian and Greenland coasts, were it successfully assimilated into the system. Figure 10 shows RMSDs of SSS for all the products over the northern North Atlantic Ocean and the Nordic Seas. On average, the largest uncertainty is found with the CEC (~1.0 psu; Fig. 10d), with RMSDs as large as 1.5 psu in the Greenland Sea and the Barents Sea. The SSS RMSDs for the five other SSS products are much smaller (~0.5 psu).

4.2 Independent SSS in the Beaufort Sea

Independent in-situ data from BGEP and CLIVAR are used during the summer months of 2011-2013 in the Beaufort Sea for the evaluation of the six SSS products (Fig. 11). The in-situ SSS observations range from 15 to 32 psu. The range of BEC SSS is limited to 24 to 31 psu with a minor bias of 0.09 psu and a RMSD of 1.82 psu. On the other hand, the range of TP4 SSS is even shorter from 19 to 32 psu, with a large saline bias of 2.59 psu and a RMSD of 3.63 psu. The linear regression coefficients for BEC and TP4 are 0.57 and 0.07 respectively. Looking at the low-salinity observations (~27 psu) collected at (136.4°W, 70.5°N) on 15th August 2011, marked by anti-triangles (Fig. 1b) near the Mackenzie River estuary, TP4 has a significant negative bias (< -4 psu) visible as the outliers above the dashed-black line in Fig. 11a. This hints to a lack of fresh water signatures from river discharge. The range of PHC SSS climatology is only reaching from 24 to 31 psu, similar to TP4, with a saline bias of 1.65 psu and RMSD of 2.85 psu. Compared to the TP4 deviation at the Makenzie River basin, the PHC saline bias is present, but smaller. The strong positive bias in TP4 at these points can then be partly attributed to the SSS relaxation of the TOPAZ model towards the PHC climatology, albeit rather weak. The range of the WOA is much wider, from 12 to 31 psu. Among the six products, the WOA bias is the smallest (~0.02 psu) over the Beaufort Sea during all three summers. However, it should be noted that the variability of in-situ observations is very large for salinities lower than 24 psu, which contributes to the large RMSD (>3.0 psu) of both PHC and WOA. It confirms that the two climatologies have a sizable uncertainty over low salinity regions (<24 psu) in the Arctic Ocean. The CEC SSS ranges from 13 psu to 34 psu, which is much wider than the range of the BEC SSS. The saline bias of CEC is however larger at 2.38 psu and its RMSD is about quite large at 3.77 psu. Futhermore, the CEC deviations from the in-situ observations are larger in waters fresher than 27 psu. The MOB combined product performs poorly with the largest negative bias (>5 psu) and an RMSD in excess of 8 psu. In contrast to the other five SSS products, the anomalously fresh SSS observed around the point (140°W, 71°N) near the Mackenzie River estuary are represented by even fresher values of 12 psu in MOB, which may hint at an amplification of the anomalies.

In order to characterize dependencies of the bias for the six SSS products against the in-situ data, their absolute differences are plotted as a function of observed SSS in Fig. 12. In general, all products show considerable deviations with the maxima reaching 8 to 14 psu. While the absolute misfits of most of the SSS products monotonically increase towards lower salinity, the bias of MOB shows its peak around 20 psu shown in Fig. 12c. The fourth-order polynomial curve function,

$$F(S) = p_1 S^4 + p_2 S^3 + p_3 S^2 + p_4 S + p_5 \quad (3)$$

is then fitted to the absolute bias for each of the SSS products, where S represents the in-situ salinity. The fitting coefficients, p_1 to p_5, for each product are listed in Table 3. The norm residuals are displayed on each panel in Fig. 12 and clearly show that fitting for MOB has the largest uncertainty, while the minimal norm residuals are about 10 and 7 psu^2 respectively for BEC and TP4. This suggests the derived fitting curves for BEC and TP4 have credible skill in charactering its error distribution as a function of the observed SSS. Both curves monotonically decrease towards the salinity higher than 28 (30) psu for BEC (TP4) and increase slightly afterwards. The absolute bias in TP4 is consistently larger than that in BEC. The fitted curves of PHC and WOA have the similar functional forms to TP4 and BEC, but with lower amplitudes.

5. Conclusions

To understand the uncertainties in the Arctic SSS, our study evaluates the two gridded SMOS SSS products (BEC and CEC), two CMEMS products (TP4 and MOB), and two climatology products (PHC and WOA) by their inter-comparison and comparisons against both of dependent and independent in-situ datasets during the years of 2011-2013.

The differences in the spatial coverage of the two SMOS SSS were clearly shown in the monthly mean (Fig. 2 and Fig.3), due to the different retrieval applied in these two datasets. The spatial distributions of SSS from TP4 and PHC are considerably close to each other, mainly as for the fact that the SSS in the TOPAZ model is relaxed towards PHC at each time step. Relative to TP4, the SSS deviations of the four products (BEC, MOB, WOA and PHC) in summer show similar magnitude over the open waters. On the contrary, the CEC SSS shows a negative bias (<-1 psu) over the region extending from the Iceland towards the western side of Ireland (Fig. 4, 5), but clearly the BEC SSS has a slightly negative bias over the region. In general, the most significant differences in the SSS deviations relative to TP4 are found under the sea-ice cover and in its surrounding marginal ice zones.

Furthermore, the intercomparison of the SSS products shows that the BEC SSS in August and September (Fig. 4, 5) has consistent negative deviations along the sea-ice edge in the Beaufort Sea and the Chukchi Sea, but the CEC SSS has opposite deviations in these two months. Thus, it may be arguable that the two SMOS products would give rise to significantly different effects to the upper ocean state in the TOPAZ system if it to be assimilated into. Hence the SSS quantitative evaluations of two products for optimal selection or blending would be worth of investigating further.

Focusing on the wide Arctic domain (>60°N), the deviations of the five SSS products relative to TP4 show diverse seasonal characteristics (Fig. 6). Although the SSS products of BEC and CEC have the similar deviation of about 1.5 psu (Fig. 6a) in summer, the BEC deviations in winter months are clearly lower (~0.5 psu). The deviations of MOB and WOA (Fig. 6a) varies from over 1.5 psu in winter to around 4 psu in summer, which suggests a considerable gap with the TP4. Consequently, the intercomparison suggests that the BEC SSS has the most consistent pattern with the TP4 SSS among all other SSS products.

Against the in-situ data from CORA5.1 which were used in both TP4 and MOB, the quantitative evaluations of the six SSS products were investigated in the eight sub-regions (Fig. 1a). It was divided into two parts: in the central Arctic Ocean; the northern North Atlantic Ocean and the Nordic Seas. Due to the limited coverage of BEC and CEC in S1, the scatterplots (Fig. 7) show a positive saline bias (>4 psu) for low salinity water (< 27 psu). However, the salinity bias of BEC is slightly reduced for relative high salinity water (> 27 psu). In the Kara Sea and the Barents Sea, the TP4 SSS has the minimal RMSD compared with others (Table 2). The BEC scatterplots in S2 and S3 (Fig. 7) have similar distributions with respect to TP4.

In the northern North Atlantic Ocean and the Nordic Seas (S6, S4, and S3; Fig. 8), the scatterplots of the CEC SSS show that it underestimates the Atlantic water salinity, which also is consistent with the intercomparison results (low salinity deviation) shown in Fig. 4 and 5. The misfits of mean and RMSDs shown in Fig. 9 and 10, suggest the CEC SSS has considerable uncertainty (RMSD of about 1 psu), especially in the Nordic Seas with obvious low salinity biases. On the other hand, the SSS uncertainties of the BEC are significantly lower in comparison to the CEC, but are equivalent compared with TP4 and PHC. Two notable regions where the BEC SSS has lower uncertainties referred to the in-situ observations than the TP4 are along the Norwegian coast and near the west coast of Greenland Island. It is reasonable to expect that they are the most beneficial region in the Nordic Seas if the BEC SSS is successfully assimilated into the TOPAZ system.

Against independent in-situ observations from BGEP and CLIVAR, the SSS evaluation in the Beaufort Sea is performed in the summers of the three years. The linear regression against these independent SSS observations (Fig. 11) suggests the BEC SSS has the smallest RMSD of 1.8 psu with a positive bias of 0.1 psu, and the CEC SSS has larger RMSD of about 3.8 psu with a larger positive bias of 2.4 psu (Fig. 11). On the other hand, the TP4 SSS also shows large RMSD of about 3.6 psu with large positive bias of 2.6 psu. They are smaller than MOB which has the RMSD of 8.2 psu and larger negative bias (-5.0 psu). As for the two climatology products, the RMSDs of WOA and PHC both are more than 2.8 psu, but with significantly smaller bias in WOA. Overall, the large uncertainty found in the linear regression of all products is attributed to large product-observation mismatch against in-situ salinity data of less than 24 psu, which are observed over the continental shelf near the estuary of the Mackenzie River.

In order to characterize the product-data misfits of all six products against in-situ data, a 4th order polynomial function is fitted to the absolute deviation as a function of observed salinity (Fig.12). The absolute deviations of most of the products except for MOB monotonically decrease as observed salinity increases. The norm residuals for TP4 and BEC and are the smallest of 10.2 and 7.0, respectively, among all six products and the fitted curves give certain confidence in estimating the size of the error in the each SSS product. The fitted curve reaches its smallest value of less than 1.0 psu at the in-situ salinity of 28 psu and 30 psu for BEC and TP4 respectively. Both the fitted curves for CEC and MOB have large norm residuals of 18.1 and 68.8 psu^2 respectively. Note that special attention must be paid in usage of MOB in the

Arctic Ocean due to its large negative bias and the RMSD in regions where the product is based on limited number of observations.
Evaluations of the SSS products against TP4 product and in situ data conducted above suggest certain benefit can be expected in assimilating one of the SMOS salinity products, the BEC SSS, into the TOPAZ Arctic ocean analysis-forecast system. The knowledge of error structure in the SSS products provided in this study will assist to reasonably estimate the observation error for the SMOS product, which is required by a data assimilation system. We recommend that due to the poor spatial coverages of CORA in situ data in the Arctic Ocean, more data - especially from the Arctic Ocean marginal seas - should be compiled from independent data source for validating the SMOS SSS products. In addition, when comparing the two climatology products, PHC and WOA, the SSS scatterplots of the PHC in the central Arctic (Fig. 7) show salinity bias for low saline water. Considering the different time periods of their compiled in-situ data sources (Table 1), it independently verifies that the freshening in the Canada Basin since 1990s is rather significant as discussed by Morison et al. (2012). Based on this evaluation, the next TOPAZ system will use the WOA to replace the PHC as the target relaxation field.

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Captions of Table and Figures:

Table 1. Details of the six products evaluated during 2011-2013.

Product	*Data source*	*Resolution*	*Provider*	*Website or CMEMS id*	*Release year*
BEC	SMOS	9 days; 25 km	Barcelona Expert Centre, Spain	http://bec.icm.csie.es	2018
CEC	SMOS	9 days; 25 km zonal	Ocean Salinity Expertise Center, IFREMER	FTP: ftp.ifremer.fr	2018
TP4	Reanalysis	Daily; 12~16 km	CMEMS	ARCTIC-REANALYSIS-PHYS-002-003	2015
MOB	In situ + SMOS	7 days; 1/4x1/4°;	CMEMS	MULTIOBS_GLO_PHY_REP_015_002	2016
PHC	In situ (1950-1994)	Monthly; 1x1°	*Polar Science Center, University of Washington*	http://psc.apl.washington.edu/	2005
WOA	In situ (1955~2012)	Monthly; 1/4x1/4°	*NODC, NOAA*	https://www.nodc.noaa.gov/OC5/woa13/	2013

Table 2. Misfits of SSS relative to in-situ CORA5.1 observations during 2011-2013 in each sub-region. Bold numbers denote the smallest error among the six products.

	Bias (psu)						RMSD (psu)					
Region	BEC	CEC	TP4	MOB	PHC	WOA	BEC	CEC	TP4	MOB	PHC	WOA
S0	-	-	.48	-.52	.48	***-.11***	-	-	1.25	1.78	1.28	***.70***
S1	4.03	3.18	3.29	1.63	3.29	***.42***	4.23	3.70	3.47	2.22	3.43	***1.37***
S2	-1.76	***-.44***	-.97	2.96	-3.30	-2.93	2.16	2.57	***1.70***	3.68	3.87	3.62
S3	***-.14***	-.70	***-.14***	-.21	-.29	-.25	.45	1.17	***.34***	.42	.51	.44
S4	-.09	-.20	.12	.11	***-.02***	***.02***	.91	1.21	.89	.86	.94	***.84***
S5	-.07	.06	.20	***.01***	.02	.07	1.47	1.52	1.42	1.44	1.39	***1.30***
S6	***-.01***	.15	***.01***	***-.01***	-.09	.05	.25	.66	.14	***.12***	.28	.16
S7	.05	.34	.04	***-.03***	-.23	***-.03***	.31	.88	.33	***.22***	.43	.27

Table 3. Optimal coefficients for the 4th order polynomial fit of the errors (see Eq. 3) as a function of in-situ SSS for each product.

	$F(p_1, p_2, p_3, p_4, p_5, s)$					Residual norm	In situ samples
Product	$p_1(x10^{-3})$	p_2	p_3	p_4	p_5		
BEC	0.168	-0.016	0.614	-11.345	87.097	7.03	91
CEC	0.225	-0.033	-1.550	-29.886	205.179	18.13	121
TP4	0.993	-0.096	3.430	-54.552	335.197	10.17	232
MOB	-1.080	0.128	-5.469	99.824	-645.087	68.81	163
PHC	1.257	-0.120	4.235	-65.938	388.808	13.98	232
WOA	-0.121	0.010	-0.322	3.998	-10.847	38.91	232

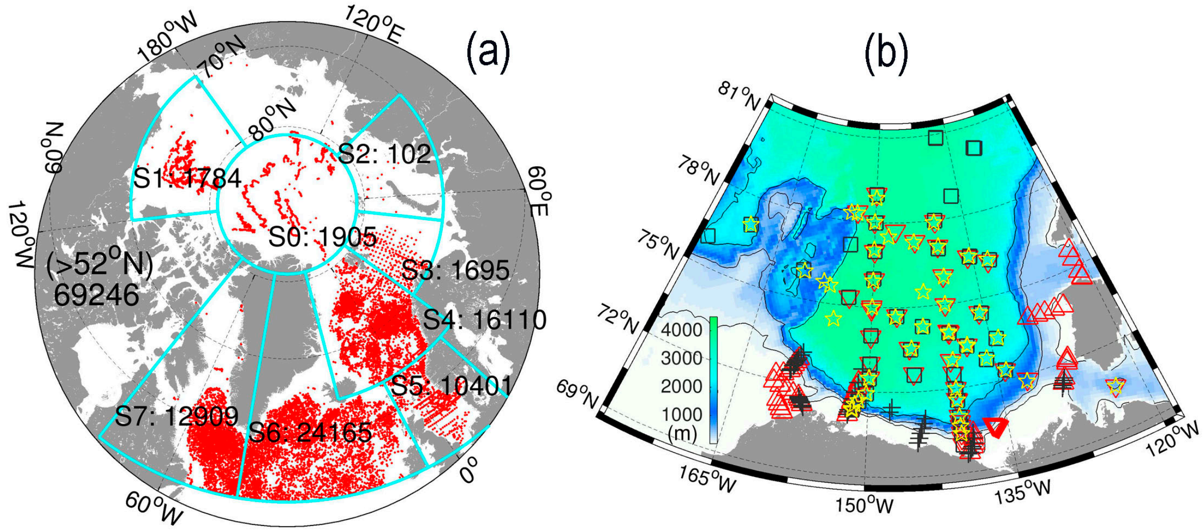

Fig. 1 **(a)**: SSS locations of the in-situ observations north of 52°N in CORA5.1 during the years 2011-2013. 8 sub-regions divide the Arctic Ocean, with the number of observations indicated in each region. **(b)**: Independent SSS observations in the Beaufort Sea during the summer months of 2011-2013 from the BGEP (marked by anti-triangles, squares, and starts) and the CLIVAR (marked by triangles and crosses). Different colors (red, black and yellow) indicate the years (2011, 2012 and 2013 resp.).

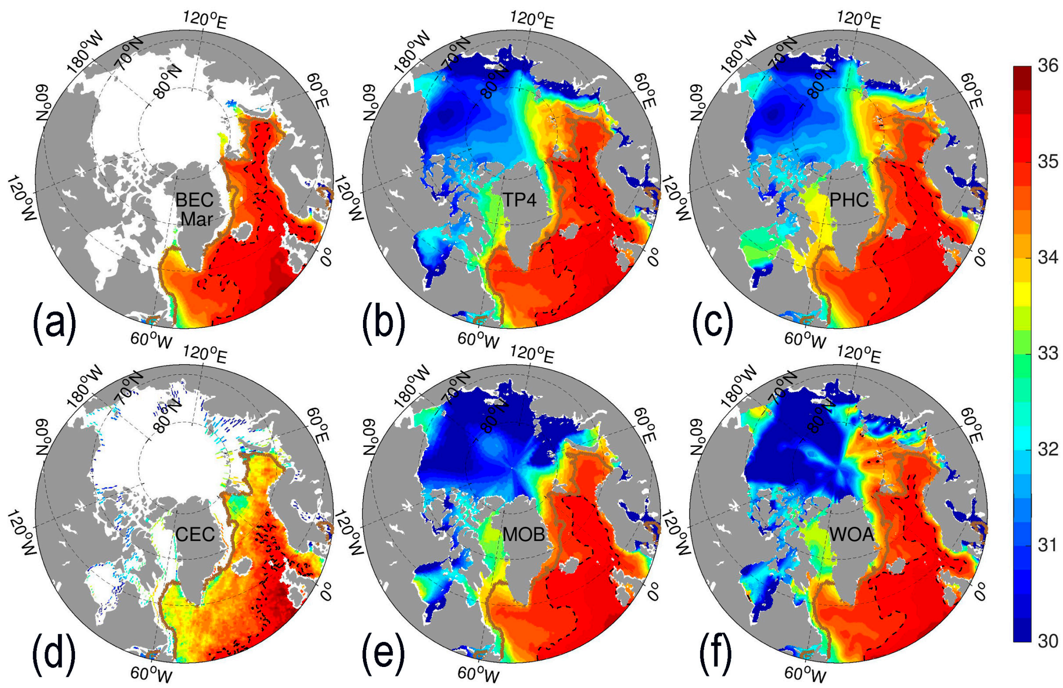

Fig. 2 Monthly SSS (unit: psu) in March from satellite products (BEC and CEC, *left column*), reanalysis/reprocessing (TP4 and MOB, *middle column*), and climatology (PHC and WOA, *right column*). White areas are masked by sea ice. The thick brown line represents the sea ice edge (15% concentration from TP4), and the black shaded isoline represents the 35 psu salinity near the surface.

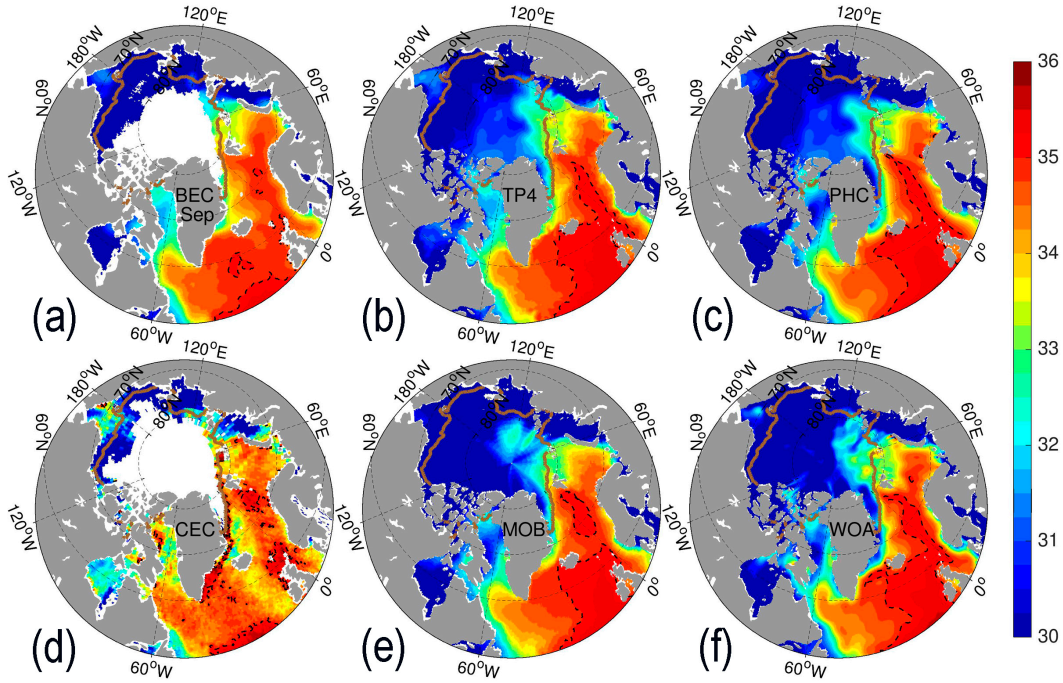

Fig. 3 Similar to previous figure in September.

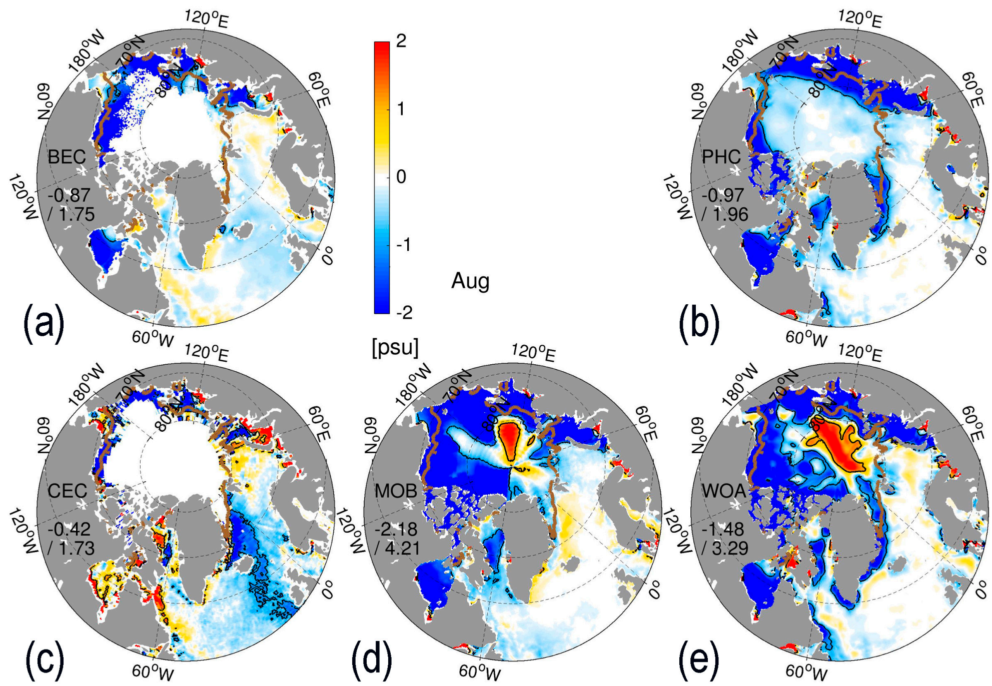

Fig. 4 Deviations of monthly SSS (unit: psu) in August for (a) BEC; (b) PHC; (c) CEC; (d) MOB; and (e) WOA relative to TP4. The thick brown line represents sea ice edge (15% concentration from TP4), the black lines represent ±1 psu.

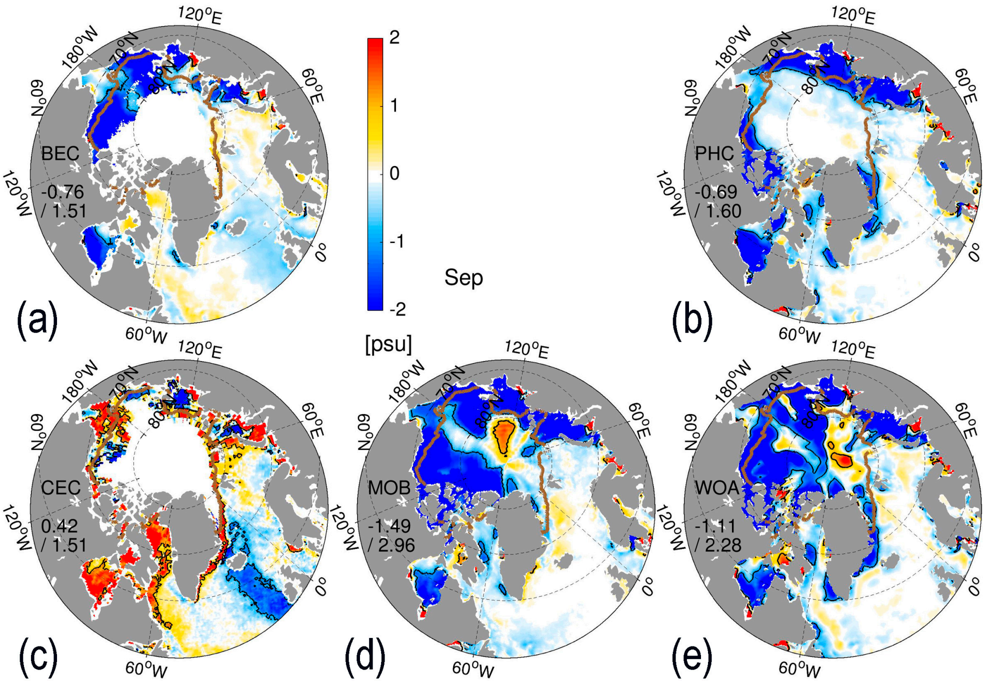

Fig. 5 Same as previous for September.

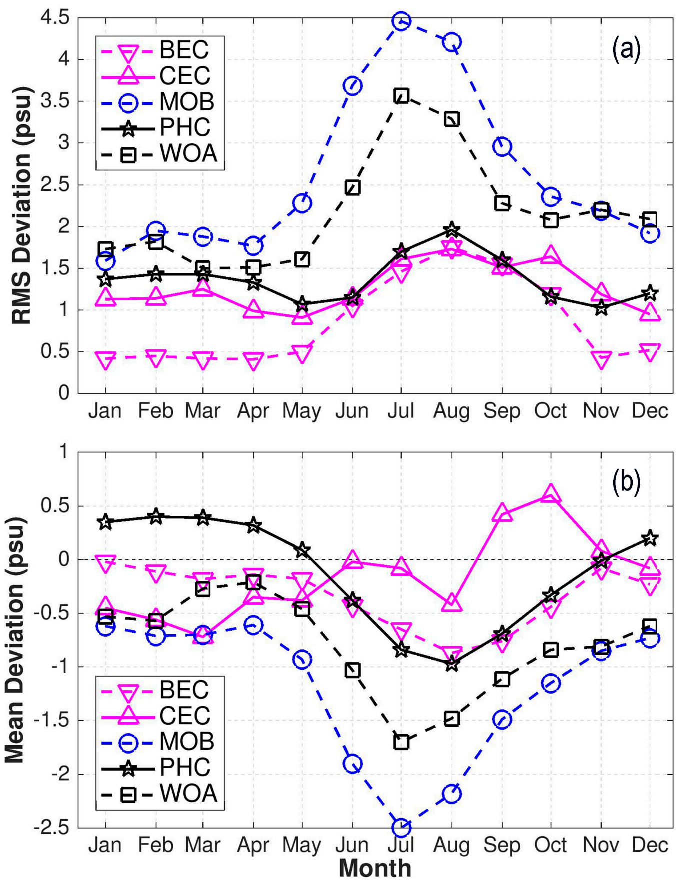

Fig. 6 Monthly deviations in the Arctic Ocean (>60N) of (a) the RMS and (b) the spatial average during the period 2011-2013 for the five SSS products referred to TP4. The anti-triangle (triangle, circle, star and square) line represents the SSS deviations from BEC (CEC, MOB, PHC and WOA respectively).

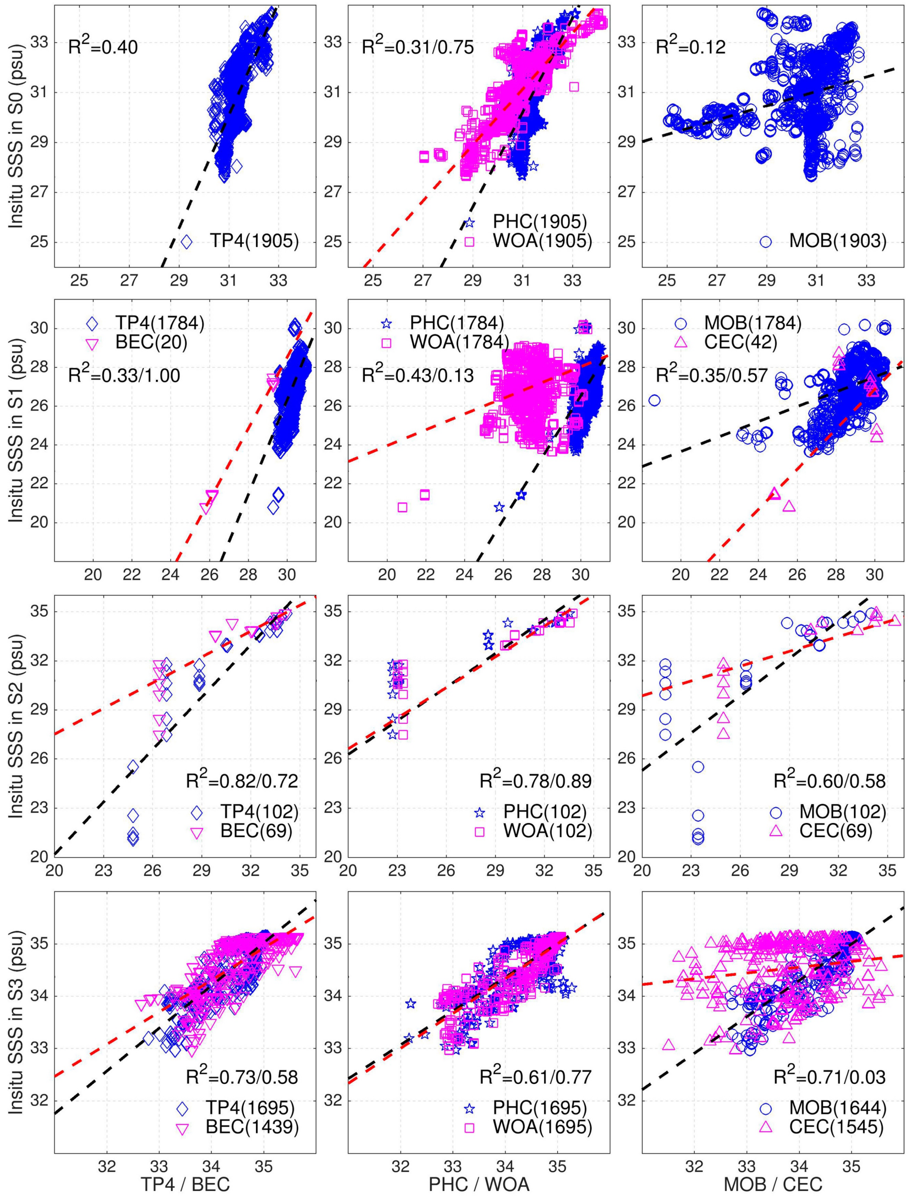

Fig. 7 Scatterplots of SSS compared to the CORA5.1 in-situ observations with respect to the S0-S3 regions in the Arctic. The diamonds (anti-triangles, stars, squares, circles, and triangles) represents the SSS from TP4 (BEC, PHC, WOA, MOB, and CEC respectively). The black (red) lines are the linear regressions of the blue (purple) dots in each panel, and the coefficient R^2 is indicated in the panel together with the number of observations in parentheses.

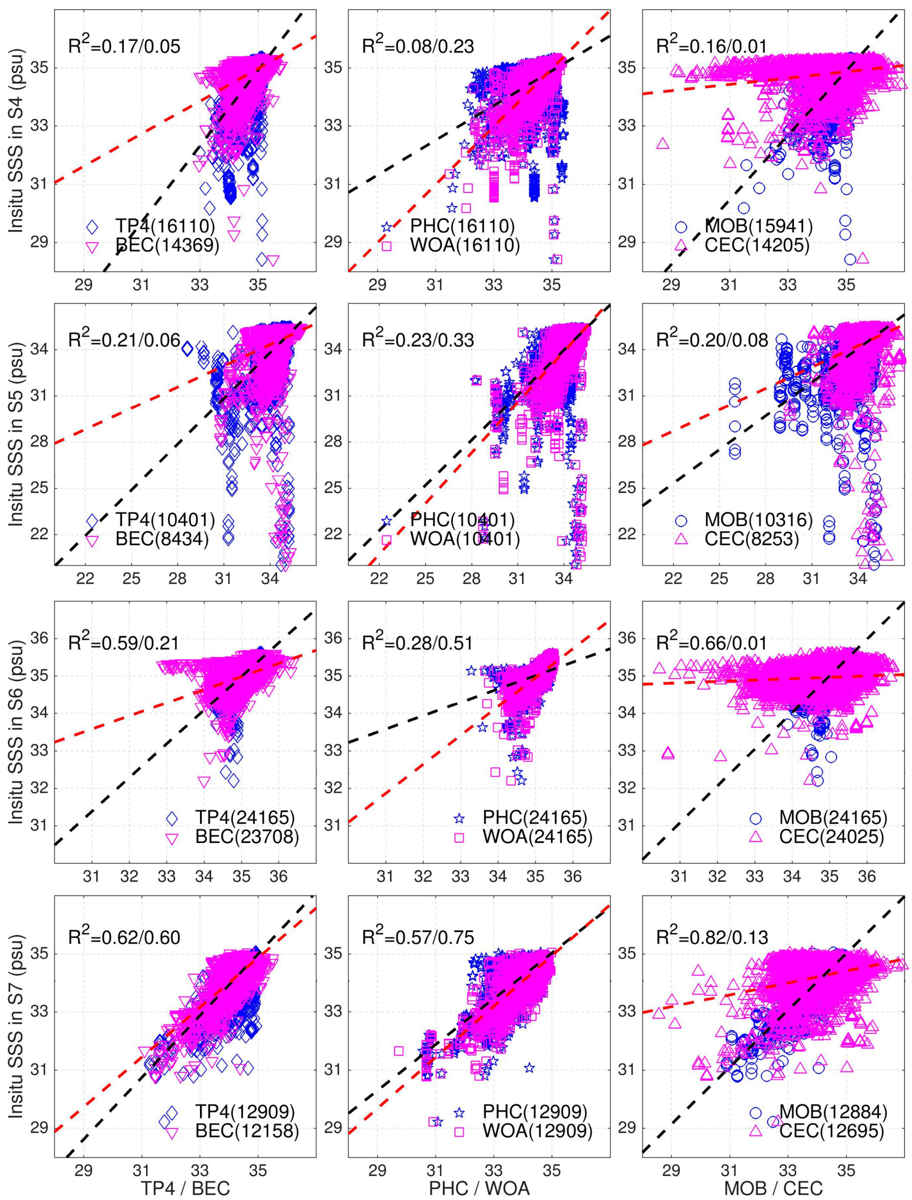

Fig. 8 Same as Fig. 7 but for the subpolar regions S4-S7.

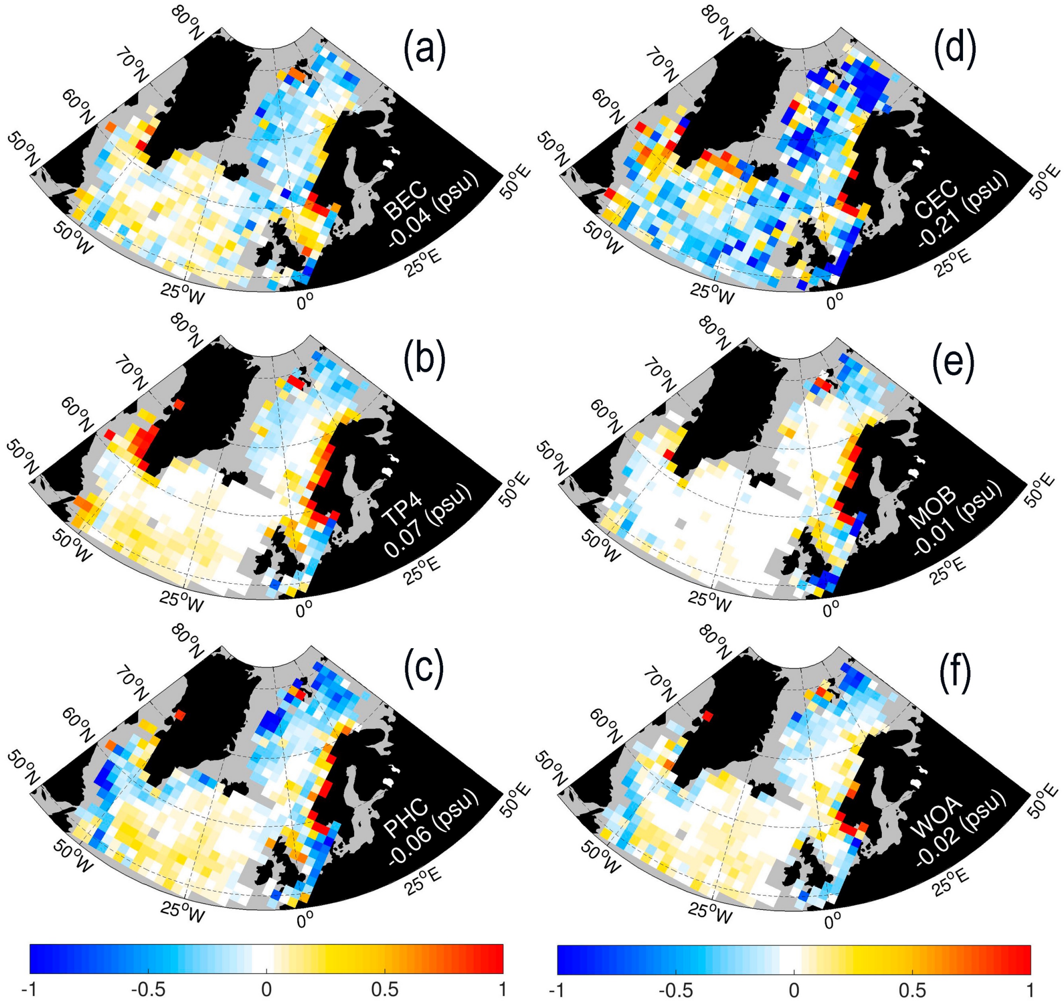

Fig. 9 The mean deviation of SSS for the six datasets compared to in situ observations from CORA 5.1 during the three years of 2011-2013 in the northern North Atlantic and the Nordic seas. The SSS observations are distributed into the coarse grid cells of 9x9 grids in TP4, with a gray mask if the valid observations less than 10.

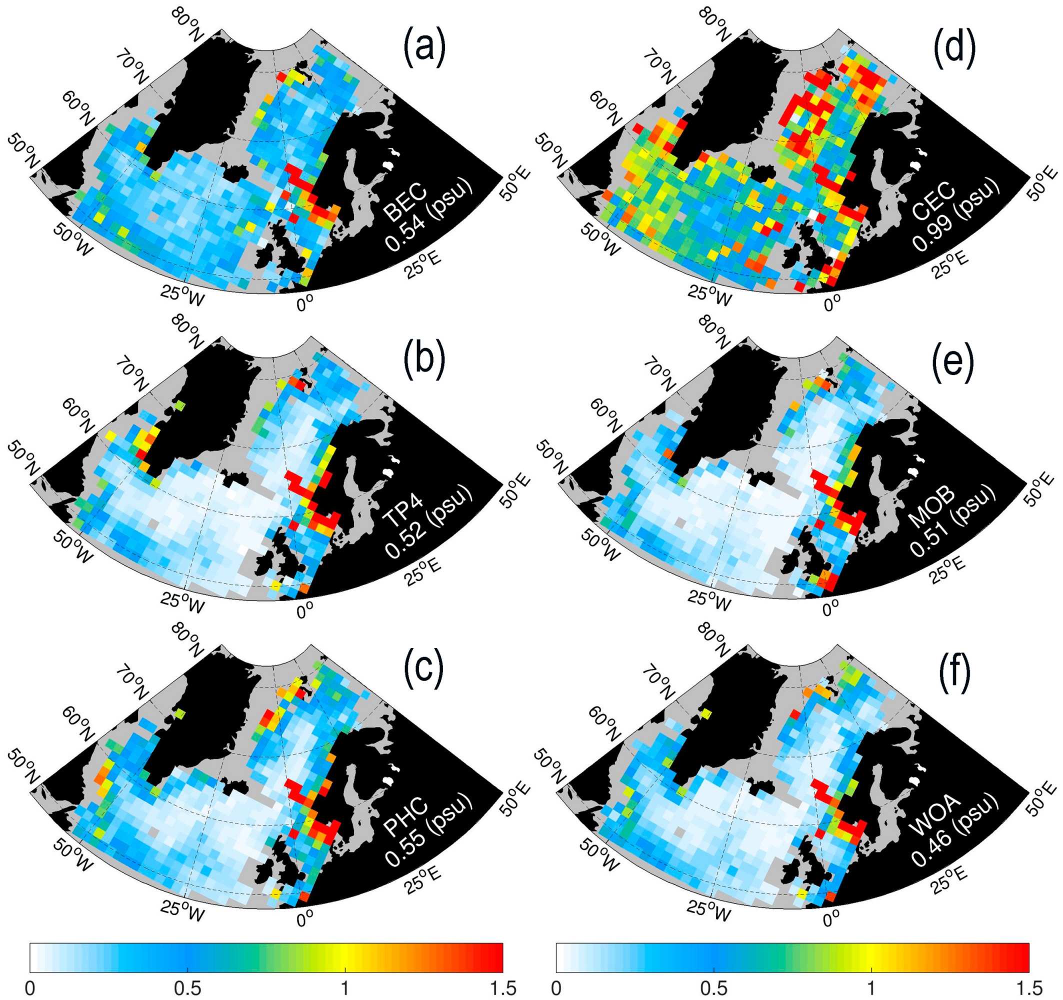

Fig. 10 The Root Mean Square deviation of SSS for six datasets compared to in situ observations from CORA 5.1 during the three years of 2011-2013 in the northern North Atlantic and the Nordic seas. The SSS observations are distributed into the coarse grid cells of 9x9 grids in TP4, with a gray mask if the valid observations less than 10.

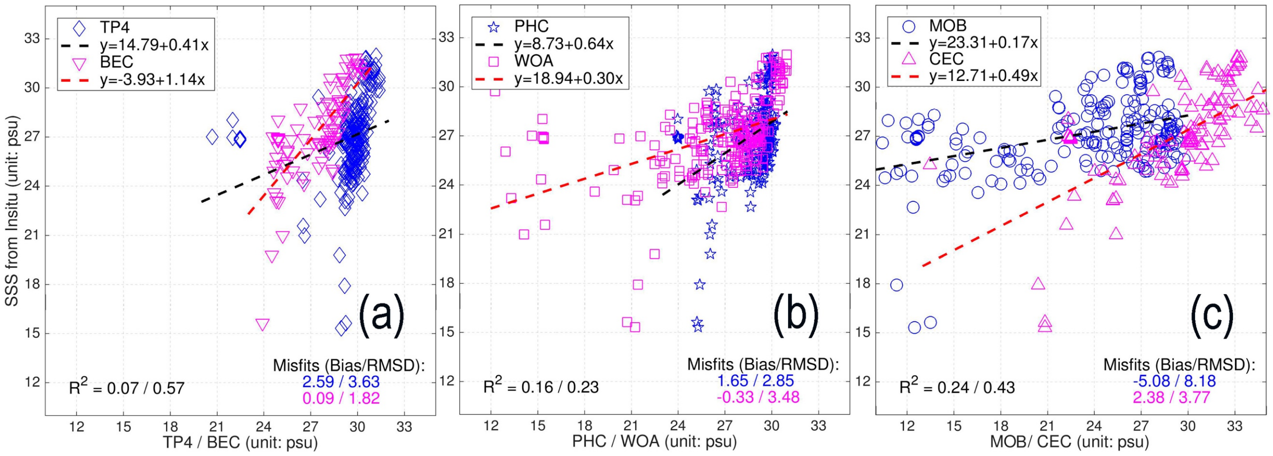

Fig. 11 Scatterplots of SSS compared to the in-situ observations in Beaufort Sea during the summer months of 2011-2013: (a) The diamond (anti-triangle) represents the SSS from TP4 (BEC) with blue (purple), and the linear regression is denoted by the dashed black (red) line. (b) The star (square) from the climatology of PHC (WOA). (c) The circle (triangle) represents from MOB (CEC). The coefficient R^2 is the squared linear relationship, and the misfits also shown on the panels.

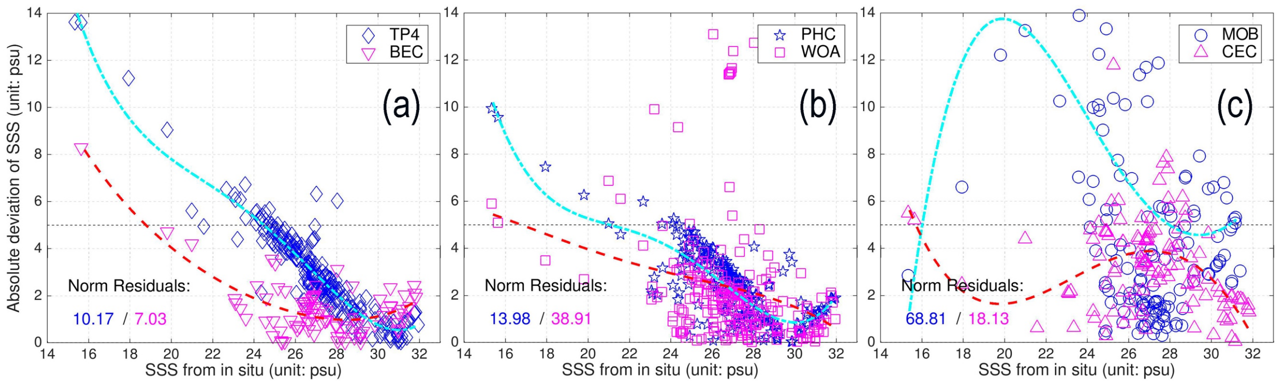

Fig. 12 Scatterplots of SSS uncertainty compared to the in-situ observations in Beaufort Sea as a function of the observed salinity. The black dashed line represents the absolute deviation of 5 psu. (a) The diamond (anti-triangle) represents from TP4 (BEC) with blue (purple). (b) The star (square) from the climatology of PHC (WOA). (c) The circle (triangle) represents from MOB (CEC). The thick dashed curves are fitted by the fourth order polynomial function, and the norm residuals are marked on panel respectively.