



# Sea ice export through the Fram Strait derived from a combined model and satellite data set

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**Abstract.** Sea ice volume export through the Fram Strait plays an important role in the Arctic freshwater and energy redistribution. The combined model and satellite sea ice thickness (CMST) data set assimilates CryoSat-2 and soil moisture and ocean salinity (SMOS) thickness products together with satellite sea ice concentration. The CMST data set closes the gap of stand-alone satellite-derived sea ice thickness in summer and therefore allows us to estimate sea ice volume export during the melt season. In this study, we first validate the CMST data set using field observations, and then we estimate the continuous seasonal and interannual variations in Arctic sea ice volume flux through the Fram Strait from September 2010 to December 2016. The results show that seasonal and interannual sea ice volume export vary from about  $-240(\pm 40)$  to  $-970(\pm 60)$  km<sup>3</sup> and  $-1970(\pm 290)$  to  $-2490(\pm 280)$  km<sup>3</sup>, respectively. The sea ice volume export reaches its maximum in spring and about one-third of the yearly total volume export occurs in the melt season. The minimum monthly sea ice export is  $-11$  km<sup>3</sup> in August 2015, and the maximum ( $-442$  km<sup>3</sup>) appears in March 2011. The seasonal relative frequencies of sea ice thickness and drift suggest that the Fram Strait outlet in summer is dominated by sea ice that is thicker than 2 m with relatively slow seasonal mean drift of about 3 km d<sup>-1</sup>.

## 1 Introduction

The sea ice extent and volume in the Arctic region undergo a decline for the past decades and will likely continue to decrease (Comiso and Hall, 2014; Meier et al., 2014; Stroeve and Notz, 2015). The decline of ice extent changes the surface albedo, and as a consequence, the absorption of solar shortwave radiation increases. The variability of ice volume, however, exerts influence on heat, freshwater budget and weather systems in the lower latitudes (Gregory et al., 2002; Tilling et al., 2015). Correspondingly, both the thermodynamic processes and dynamic processes can affect Arctic sea ice mass budget (Ricker et al., 2018). The sea ice outflow driven by atmospheric circulation is an important component of dynamic processes. The Fram Strait serves as the primary outlet of the Arctic sea ice export (Krumpal et al., 2016). Moreover, the ice outflow through the strait into the Nordic Seas covers approximately 25 % of the total Arctic freshwater export (Lique et al., 2009; Serreze et al., 2006).

Variations in satellite-based Arctic sea ice volume and sea ice export through the Fram Strait have been estimated by numerous studies (Bi et al., 2018; Kwok and Cunningham, 2015; Ricker et al., 2018; Spreen et al., 2009). Nevertheless, in terms of the volume flux, the primary focus of these studies is the variations during the freezing season (October–April). This is due to the limitations in retrieving sea ice thickness

and motion by satellite remote sensing during the melt season (May–September). It is mainly caused by more melt ponds and saturated water vapor in the sea ice surface, which restrains satellite-based ice thickness limited to the cold season only (Mu et al., 2018a). The speedup of sea ice drift usually accompanies thin sea ice; meanwhile the faster the sea ice drift, the larger the retrieving errors (Spreen et al., 2011; Sumata et al., 2014). Melting sea ice with a less scattering surface could significantly suppress the signal-to-noise ratio and obstruct the employment of satellite imagery to retrieve ice drift. For the abovementioned reasons, the spaceborne sea ice drift data usually induce more uncertainties in the melt season. All these deficiencies make the estimate of the Arctic sea ice thickness and drift variations all year round difficult with only satellite sea ice data.

Sea ice volume flux, compared to area flux, could reflect the sea ice mass balance in a more comprehensive way. However, the amounts of Fram Strait sea ice volume export during the freezing season do not demonstrate a conspicuous growth or decline trend (Ricker et al., 2018; Spreen et al., 2009). And the variation in the melt season ice volume flux through the Fram Strait still remains a query owing to the fact that sea ice thickness observations are sparse in the melt season, with the yearly total amount of ice volume flux also being unknown. In terms of sea ice volume flux, Ricker et al. (2018), Bi et al. (2018) and Zamani et al. (2019) point out that the variation in ice drift plays the major role in determining the annual and interannual ice volume export variability. Due to thermodynamic growth and deformation, sea ice thickness on the other hand drives the increase in the seasonal cycle of the exported volume. For this reason, an accurate data set of sea ice drift and thickness is crucial to better estimate sea ice volume output.

Employing the benefits of both the CryoSat-2 (CS2) and the Soil Moisture and Ocean Salinity (SMOS) satellite sea ice thickness products, the new data set (combined model and satellite thickness, CMST) that assimilates these data together with satellite-derived sea ice concentration (Mu et al., 2018a, b) provides the daily sea ice thickness, concentration and drift estimates simultaneously. Moreover, taking advantage of model dynamics and sea ice concentration assimilation, the new sea ice data set extends to cover the melt season when satellite thickness data are limited (Mu et al., 2018a). Previous results reveal that CMST data even have some advantages among the statistically merged satellite data CS2–SMOS and Pan-Arctic Ice–Ocean Modeling and Assimilation System (PIOMAS) thickness product when comparing with the in situ observations (Mu et al., 2018a). Therefore, the CMST sea ice product enables us to examine the all-year-round changes in sea ice volume export through the Fram Strait for 2010–2016, during a time when Arctic sea ice is undergoing dramatic changes. Further, we also calculate the sea ice thickness, concentration and drift frequency distributions along the main sea ice export gate all year round.

This paper is organized as follows. Section 2 describes the data used to validate the CMST data set and the method used to derive the volume flux. In Sect. 3, firstly, we evaluate the performance of CMST data. Then, we estimate the continuous seasonal and interannual variation in sea ice thickness, concentration and drift in the Fram Strait. Also, the all-year-round variability of sea ice volume export through the Fram Strait is calculated. Uncertainty in our volume flux estimate is discussed in Sect. 4. Concluding remarks are given in Sect. 5.

## 2 Data and methods

### 2.1 CMST sea ice data

The CMST sea ice data (<https://doi.org/10.1594/PANGAEA.891475>, Mu et al., 2018c, and <https://doi.org/10.1594/PANGAEA.906973>, Mu et al., 2019) in addition to ice thickness and concentration also provide the modeled ice drift. They are generated by an Arctic regional ice–ocean model accompanied by CS2, SMOS sea ice thickness and Special Sensor Microwave Imager/Sounder (SSMIS) sea ice concentration assimilation. This Arctic regional model (Losch et al., 2010; Mu et al., 2017; Nguyen et al., 2011; Yang et al., 2014) is configured on the basis of the Massachusetts Institute of Technology generation circulation model (MITgcm) (Marshall et al., 1997). To reflect the impacts of atmospheric uncertainties on the sea ice data assimilation, the atmospheric ensemble forecasts of the United Kingdom Met Office (UKMO) Ensemble Prediction System (EPS; <https://www.ecmwf.int/en/research/projects/tigge>, Bougeault et al., 2010, last access: 1 October 2019) are used as atmospheric forcing (Mu et al., 2018b; Yang et al., 2015; Yang et al., 2016). The Parallel Data Assimilation Framework (PDAF; Nerger and Hiller, 2013; <http://pdaf.awi.de>, last access: 1 October 2019) is applied to assimilate satellite thickness (e.g., SMOS thickness data thinner than 1 m and weekly mean CS2 thickness data) and concentration data (provided by the Integrated Climate Data Center, <http://icdc.cen.uni-hamburg.de>, Ricker et al., 2014; Tian-Kunze et al., 2014, last access: 1 October 2019). More details about this assimilation process can be found in previous studies (Mu et al., 2018a, b). CMST provides grid-cell-averaged ice thickness, i.e., the effective ice thickness (Mu et al., 2018a; Schweiger et al., 2011), with a resolution about 18 km. Further taking advantage of model dynamics and ice concentration assimilation, the daily CMST thickness data in summer are also available from September 2010 to December 2016. Although the CMST data do not span the most recent 2 years (i.e., 2017 and 2018), it does cover the years of the lowest sea ice extent record at that time (i.e., 2012 and 2016) (Parkinson and Comiso, 2013; Petty et al., 2018).

## 2.2 OSI SAF drift data

As suggested by Sumata et al. (2014), the merged Ocean and Sea Ice Satellite Application Facility (OSI SAF) sea ice drift product (OSI-405) reveals a better performance than other low-resolution sea ice drift products in the Fram Strait. Thus, we use it for comparison with CMST drift data when calculating sea ice volume export. The merged drift data can be downloaded from the Ocean and Sea Ice Satellite Application Facility (OSI SAF, [ftp://osisaf.met.no/archive/ice/drift\\_lr](ftp://osisaf.met.no/archive/ice/drift_lr), EUMETSAT, 2010–2016). The merged drift products are retrieved from multiple sensors and channels (shown in Table 1) in order to supplement data gaps in the single-sensor products. A more detailed description can be seen in the Low Resolution Sea ice Drift Product User's Manual ([http://osisaf.met.no/p/ice/lr\\_ice\\_drift.html](http://osisaf.met.no/p/ice/lr_ice_drift.html), last access: 1 January 2019).

## 2.3 NSIDC sea ice drift

The latest released Polar Pathfinder Daily 25 km EASE-Grid sea ice drift data from the National Snow and Ice Data Center (NSIDC, <https://doi.org/10.5067/INAWUWO7QH7B>, Tschudi et al., 2019, last access: 2 May 2019) are also used to evaluate the CMST drift. These data cover both the melt season and the freezing season and are widely used in the modeling and data assimilation (Miller et al., 2006; Stark et al., 2008). The input sea ice motion data sets are obtained from AVHRR, AMSR-E, SMMR, SSM/I, SSM/I, International Arctic Buoy Program (IABP) buoys and National Center for Environmental Prediction (NCEP)/National Center for Atmospheric Research (NCAR) Reanalysis wind data. More descriptions can be seen in the NSIDC ice motion user guide (<https://doi.org/10.5067/INAWUWO7QH7B>).

## 2.4 Sentinel-1 SAR sea ice drift

To further validate the CMST sea ice drift in the Fram Strait, the sea ice drift data retrieved from Sentinel-1 Synthetic Aperture Radar (SAR) images are used as the reference products (<https://www.ncbi.nlm.nih.gov/pmc/articles/PMC5999601/>, Muckenhuber and Sandven, 2018). Based on the different polarization channels, thousands of HH (horizontal transmittance, receive horizontally) and HV (horizontal transmittance, receive vertically) polarization images are calculated as monthly mean sea ice drift at 79° N along the gate from 15° W to 5° E (Muckenhuber and Sandven, 2018). These SAR drift data are derived by an open-source feature-tracking algorithm (Muckenhuber et al., 2016). Owing to the better performance of the HV polarization channel (Muckenhuber et al., 2016), we only use the southward velocity component of HV polarization for the validation. More information about the Sentinel-1 SAR sea ice drift can be obtained in previous studies (Muckenhuber et al., 2016; Muckenhuber and Sandven, 2018).

## 2.5 HEM sea ice thickness

For the purpose of evaluating the performance of CMST sea ice thickness, the helicopter-borne electromagnetic induction sounding (HEM) sea ice thickness (<https://doi.org/10.21334/npolar.2017.1ed8c57e>, King et al., 2017) is utilized for intercomparison. This HEM measurement campaign consists of nine separate flights implemented in the Fram Strait from August to September 2014. The helicopter-measured sea ice thickness is named “total thickness” including snow layer. Thus, following Krumpfen et al. (2016), we assume the thickness of snow or weathered ice is 0.1 m; i.e., we subtract the 0.1 m snow thickness from the total thickness in the later calculation. Sea ice concentration is low in the operational areas during this period and the data have not been adjusted with sea ice concentration. Because the CMST model thickness is effective thickness (i.e., mean thickness over the model grid), for easy comparison, and as recommended by the data providers (<https://doi.org/10.21334/npolar.2017.1ed8c57e>), we adjust these data with the CMST ice concentration to obtain daily mean ice floe thickness.

## 2.6 ULS sea ice thickness

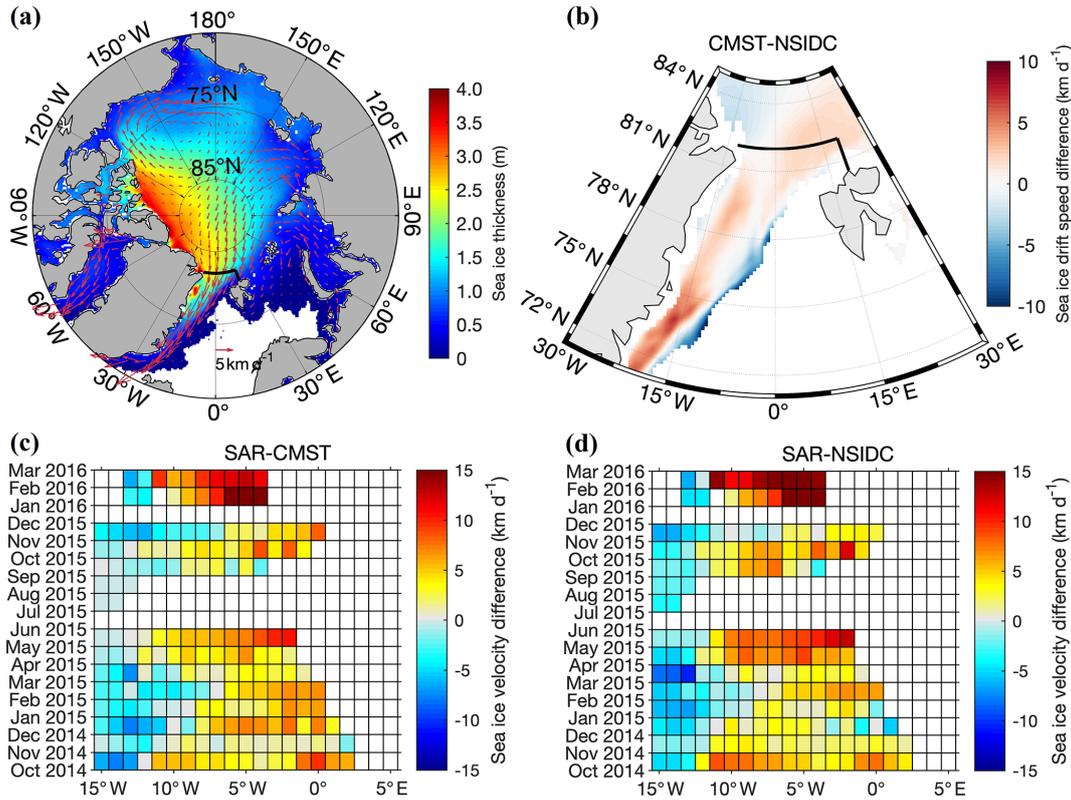
The upward-looking sonar (ULS) measurement (moored at 79° N, 5° W) in the Fram Strait is deployed and maintained by the Norwegian Polar Institute. Since ULS measures sea ice draft, the derived sea ice thickness is less affected by uncertainties in the snow layer depth and ice density. Moreover, the ULS provides year-round measurements and is therefore used to validate the CMST thickness. More details about the ULS data can be found in previous work (Hansen et al., 2013). In this study, we use a 1-year data set of monthly mean sea ice thickness from September 2010 to August 2011.

## 2.7 Retrieving methods in sea ice volume export

The sea ice thickness, concentration and drift in the CMST data set are provided on a cubed-sphere grid with a resolution of 18 km. Both sea ice variables in CMST and the OSI-405 merged data are projected to the geographic coordinates at first. Following Krumpfen et al. (2016) and Ricker et al. (2018), we define the Fram Strait export gate with zonal and meridional components as shown in Fig. 1. The zonal gate is situated at 82° N between 12° W and 20° E, and the meridional gate is located at 20° E between 80.5 and 82° N. The chosen gates are dedicated to decreasing errors and bias in low-resolution drift data and thickness data from satellites (Krumpfen et al., 2016; Ricker et al., 2018). Secondly, we use the linear interpolation method to interpolate the CMST data and OSI SAF data onto the zonal gate with a spatial resolution of 1° and onto the meridional gate with a spatial resolution of 0.15°, which is of the purpose to better match the model grids with the interpolated grids.

**Table 1.** OSI SAF drift data used in this study for comparison.

| Name    | Product          | Original data   | Algorithm | Temporal resolution | Spatial resolution | Period    |
|---------|------------------|---|-----------|---------------------|--------------------|-----------|
| OSI SAF | OSI-405 (merged) | SSMIS (91 GHz, DMSP F17), ASCAT (Metop-B), AMSR-2 (18.7 and 37 GHz) | CMCC      | 2 d                 | 62.5 km            | 2010–2016 |



**Figure 1.** (a) The mean CMST sea ice drift and thickness averaged from September 2010 to December 2016. (b) The differences between CMST drift speed and NSIDC drift speed, where the background color represents the magnitudes of ice velocity difference during the same period. The thick black line represents zonal and meridional sea ice export gates to derive sea ice volume flux through the Fram Strait. (c) Meridional velocity difference between SAR drift and CMST drift at the Fram Strait (79° N). (d) Meridional velocity difference between Sentinel-1 SAR drift and NSIDC drift at the Fram Strait (79° N).

Following Ricker et al. (2018), we also define the negative values as representing ice volume loss from the Arctic Basin through the outlet, and the sea ice volume flux can be estimated as following the formulas

$$Q_x = L_x H_x v, \tag{1}$$

$$Q_y = L_y H_y u, \tag{2}$$

where  $L_x$  is the size of the zonal interpolated grid and  $L_y$  is the size of the meridional interpolated grid.  $H_x$  and  $v$  are the interpolated effective ice thickness and meridional velocity at the zonal gate.  $H_y$  and  $u$  are the interpolated effective ice thickness and zonal velocity at the meridional gate. Note that ice concentration information is not involved in Eqs. (1)

and (2) because the calculation process of CMST effective ice thickness has already taken ice concentration information into account.

The total sea ice volume export ( $Q_{EX}$ ) through the Fram Strait is obtained by adding the zonal ice volume flux ( $Q_x$ ) and meridional ice flux ( $Q_y$ ) together:

$$Q_{EX} = Q_x + Q_y. \tag{3}$$

Uncertainties of sea ice volume export ( $\delta_{Q_x}$ ) are evaluated as

$$\delta_{Q_x} = L_x \sqrt{(H \delta_v)^2 + (v \delta_H)^2}. \tag{4}$$

This strategy is used to estimate the expected uncertainties of volume flux via the zonal gate.  $\delta v$  and  $\delta_H$  represent ice drift

uncertainty and ice thickness uncertainties, respectively. Expected sea ice volume flux uncertainties along the meridional gate can be determined by the similar method of Eq. (4).

Sea ice volume export derived from CMST thickness and drift is represented by M2 in detail in Table 2 (Sect. 3.2). The results derived from CS2 thickness and OSI SAF drift for Ricker et al. (2018) are represented by *R*. To investigate the flux biases due to the existing deviations between the CMST and the CS2 thickness data, CMST thickness and OSI SAF drift are also used to calculate the flux that is shown by M1.

### 3 Results

#### 3.1 Validation of CMST data

Firstly, the field and satellite-based observations are used to evaluate the performance of CMST sea ice data in the Fram Strait. The mean sea ice drift and thickness of nearly 6 years of CMST data are shown in Fig. 1a. The mean sea ice thickness is distributed as expected (Tilling et al., 2015; Kwok and Cunningham, 2015); e.g., the relatively thicker ice, which is more than 2.5 m, is mainly distributed in the north of Greenland and the Canadian Arctic Archipelago, and the sea ice becomes thinner towards the Eurasian coasts (Fig. 1a). We then compare the mean difference between the CMST drift and the latest released sea ice drift data (V4) from the NSIDC. The circulation patterns (the transpolar drift and the Beaufort Gyre) and magnitude distributions of the two sea ice drift data (CMST vs. NSIDC) are quite similar (figure not shown). The relatively larger differences of sea ice drift speed are found along the coast of Greenland and the ice edge, which is shown in Fig. 1b. It is noticeable that the mean sea ice drift speed of CMST is larger than the NSIDC in most areas. This may suggest that the CMST sea ice drift performs better than NSIDC drift data in the Fram Strait because NSIDC drift data usually contain underestimations in sea ice velocity (Sumata et al., 2014, 2015). For further validation of CMST sea ice velocity, we compare the CMST southward velocities that affect sea ice volume flux most with high-resolution Sentinel-1 SAR sea ice drift data. Results (Fig. 1c and d) show that both CMST drift and NSIDC drift generally overestimate the southward velocities near Greenland but underestimate the velocity far away from Greenland. Nevertheless, monthly mean CMST drift data show a better performance than NSIDC drift data, especially near Greenland.

Further assessments of CMST thickness and drift data are shown in Fig. 2. The geography map (Fig. 2a) shows the trajectories of HEM measurement campaigns and the site of ULS. Helicopter-borne daily mean sea ice thickness is used to evaluate the CMST thickness data in the Fram Strait. Monthly CMST sea ice thickness is also compared with the thickness derived from the ULS data (shown in Fig. 2c). Note that the comparison period for CMST thickness and ULS thickness is from September 2010 to August 2011 since

the ULS data afterwards have not been available for this study. Monthly mean CMST sea ice drift over the entire Fram Strait gate is evaluated with OSI SAF drift used in Ricker et al. (2018) within the same period from September 2010 to December 2016 and the same domain defined before. The correlation coefficient (CC), the relative bias (RB) and the root-mean-square error (RMSE) are explored to quantify the comparison. These statistic metrics are calculated as follows (Chen et al., 2013; Zhang et al., 2019):

$$CC = \frac{\text{Cov}(\text{CMST}, \text{OBS})}{\sigma_{\text{CMST}}\sigma_{\text{OBS}}}, \quad (5)$$

$$RB = \frac{\sum(\text{CMST} - \text{OBS})}{\sum \text{OBS}}, \quad (6)$$

$$RMSE = \sqrt{\frac{(\text{CMST} - \text{OBS})^2}{N}}, \quad (7)$$

where Cov represents the covariance operator,  $\sigma$  is the standard deviation and the number of the observations (OBS) is indicated by *N*.

Statistical analysis between CMST and HEM sea ice thickness shows that the CC, RB and RMSE are 0.59, 15.13 % and 0.66 m, respectively. Furthermore, statistics indicate that the CMST data are comparable to ULS measurements with a CC of 0.68 and a low RB (1.74 %) and RMSE (0.328 m). Note that the CMST thickness has already been quantitatively evaluated with more observation records by a previous study (Mu et al., 2018a) and exhibits some advantages over the widely used CS2–SMOS and PIOMAS thickness data. The CC between CMST drift and OSI SAF drift shows a high correlation of 0.93 (Fig. 2d) in the freezing season (October–April). The RB (−6.05 %) and RMSE (0.985 km d<sup>−1</sup>) are also relatively quite low. These statistical metrics suggest a good performance of CMST over the Fram Strait outlet in reproducing the real sea ice drift and thickness.

#### 3.2 Sea ice thickness, concentration and drift variation

In this study, the spring, summer, autumn and winter span from March to May, June to August, September to November and December to February, respectively. The all-year-round seasonal variation in Arctic sea ice thickness and concentration is shown in Fig. 3 and 4. During the study period, both the Arctic sea ice thickness and concentration show a significant seasonal variation; e.g., the sea ice thickness reaches its maximum in spring (except for 2013), while the sea ice concentration has a peak value in spring and winter.

As shown in Fig. 3, the distribution of sea ice thickness along the Fram Strait zonal gate features thicker sea ice in the east of Greenland than that in the west of Svalbard, showing a gradually thinning trend from west to east. And along with the meridional gate, sea ice is thickening from northern Svalbard to the central Arctic Ocean, which is in line with other studies (Hansen et al., 2013; Kwok et al., 2004; Krumpfen et al., 2016; Vinje et al., 1998). Note that the sea

**Table 2.** Monthly Arctic sea ice volume export through the Fram Strait in cubic kilometers per month. Note that *R* is calculated by CS2 thickness and OSI SAF drift, M1 is calculated by CMST thickness and OSI SAF drift, and M2 is calculated by CMST thickness and CMST drift. The bold values represent the minimum and maximum values in this study.

|      |          | Jan  | Feb        | Mar         | Apr  | May  | Jun  | Jul  | Aug        | Sep  | Oct  | Nov  | Dec  |
|------|----------|------|------------|-------------|------|------|------|------|------------|------|------|------|------|
| 2010 | <i>R</i> | –    | –          | –           | –    | –    | –    | –    | –          | –    | –    | –227 | –275 |
|      | M1       | –    | –          | –           | –    | –    | –    | –    | –          | –    | –    | –209 | –258 |
|      | M2       | –    | –          | –           | –    | –    | –    | –    | –          | –148 | –222 | –195 | –239 |
| 2011 | <i>R</i> | –267 | <b>–21</b> | <b>–540</b> | –279 | –    | –    | –    | –          | –    | –164 | –214 | –354 |
|      | M1       | –238 | <b>–24</b> | <b>–478</b> | –255 | –    | –    | –    | –          | –    | –149 | –163 | –293 |
|      | M2       | –238 | <b>–34</b> | <b>–442</b> | –230 | –278 | –185 | –115 | –64        | –28  | –151 | –175 | –290 |
| 2012 | <i>R</i> | –129 | –381       | –379        | –487 | –    | –    | –    | –          | –    | –203 | –182 | –187 |
|      | M1       | –109 | –299       | –287        | –428 | –    | –    | –    | –          | –    | –207 | –157 | –125 |
|      | M2       | –137 | –300       | –267        | –372 | –334 | –218 | –187 | –131       | –100 | –160 | –149 | –136 |
| 2013 | <i>R</i> | –103 | –163       | –299        | –318 | –    | –    | –    | –          | –    | –215 | –400 | –231 |
|      | M1       | –80  | –122       | –254        | –254 | –    | –    | –    | –          | –    | –212 | –372 | –211 |
|      | M2       | –78  | –109       | –217        | –219 | –194 | –140 | –107 | –98        | –26  | –228 | –367 | –191 |
| 2014 | <i>R</i> | –78  | –195       | –345        | –452 | –    | –    | –    | –          | –    | –200 | –165 | –373 |
|      | M1       | –49  | –105       | –240        | –401 | –    | –    | –    | –          | –    | –203 | –122 | –307 |
|      | M2       | –61  | –114       | –282        | –425 | –232 | –161 | –112 | –184       | –194 | –170 | –162 | –283 |
| 2015 | <i>R</i> | –160 | –425       | –429        | –354 | –    | –    | –    | –          | –    | –52  | –261 | –275 |
|      | M1       | –129 | –358       | –328        | –284 | –    | –    | –    | –          | –    | –72  | –215 | –243 |
|      | M2       | –129 | –355       | –339        | –308 | –171 | –240 | –114 | <b>–11</b> | –107 | –78  | –192 | –244 |
| 2016 | <i>R</i> | –177 | –352       | –348        | –310 | –    | –    | –    | –          | –    | –129 | –151 | –307 |
|      | M1       | –129 | –272       | –255        | –264 | –    | –    | –    | –          | –    | –98  | –90  | –243 |
|      | M2       | –150 | –267       | –287        | –289 | –196 | –194 | –113 | –198       | –75  | –97  | –72  | –222 |

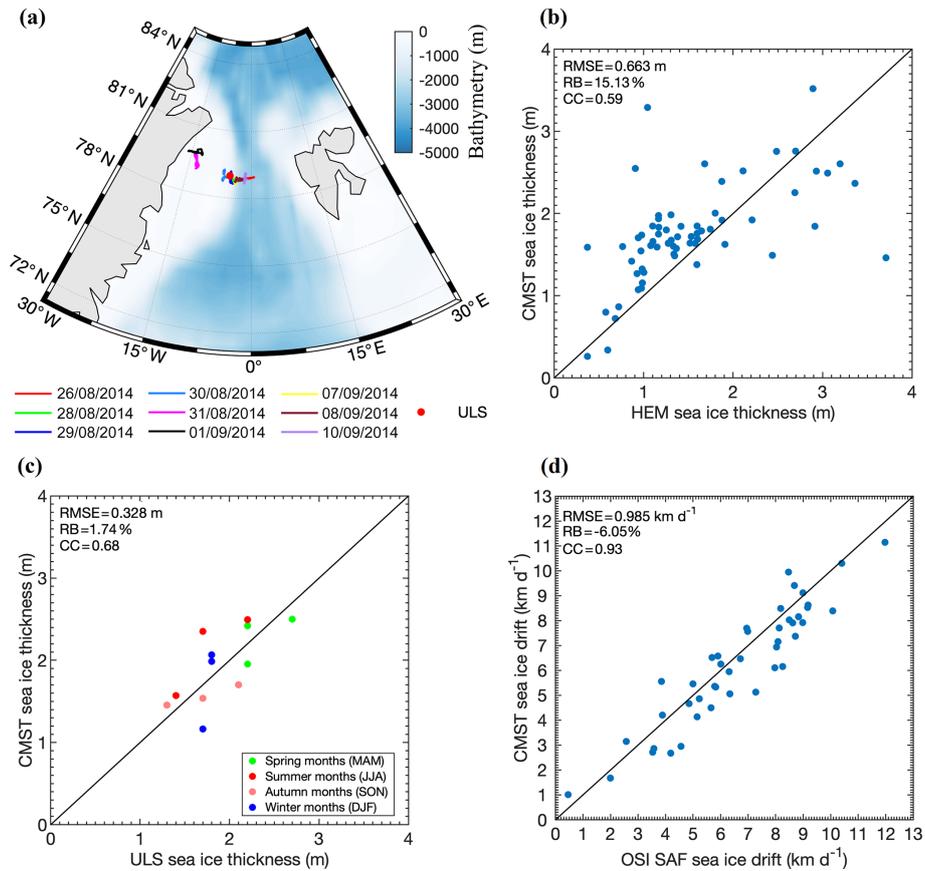
ice thickness hits its minimum in the autumn of 2011, and such an anomaly is also found in previous studies based on sea ice satellite data (Kwok and Cunningham, 2015; Tilling et al., 2015). Also, it is notable that the mean thickness of winter 2013 shows a significant thickening compared with other winters. This remarkable thickening may be linked to the anomalous cooling in 2013, which enhances the thermodynamic ice growth (Tilling et al., 2015).

Further analysis on the sea ice volume in the Arctic Basin shows a typical seasonal variation with the minimum in autumn and the maximum in spring. The Arctic sea ice volume undergoes a minimum season in the autumn of 2011 ( $6.93 \times 10^3 \text{ km}^3$ ) and reaches a maximum of  $20.19 \times 10^3 \text{ km}^3$  in the spring of 2014. A maximum (minimum) sea ice extent does not correspond to a maximum (minimum) volume. For instance, the sea ice extent minimum ( $5.17 \times 10^6 \text{ km}^2$ ) and maximum ( $10.87 \times 10^6 \text{ km}^2$ ) are each found in autumn of 2012 and in spring of 2013, while the sea ice volume minimum ( $6.93 \times 10^3 \text{ km}^3$ ) happens in autumn of 2011 and the maximum of  $20.19 \times 10^3 \text{ km}^3$  occurs in spring of 2014. The trends of the temporal variation in Arctic ice volume and extent are similar to the results from Tilling et al. (2015) and Kwok and Cunningham (2015).

The sea ice thickness, concentration and drift averaged over the entire Fram Strait gate are shown in Figs. 5, 6 and 7,

respectively. We also compare these sea ice variables with Ricker et al. (2018). The results show that the CMST ice thickness and drift are smaller than that of CS2 and OSI SAF, while the CMST ice concentration is a little larger than OSI SAF ice concentration. Such a thinner CMST sea ice thickness found in the Fram Strait is discussed to be reasonable in Mu et al. (2018b) because of the assimilation of SMOS thickness data. The previous study shows that the mean Arctic-wide OSI SAF drift is slightly larger than IABP/D buoy ice drift (Sumata et al., 2014), which suggests the slight underestimation of CMST drift seems also tenable. Further validation with more ice drift data over the Arctic Basin (e.g., buoy drift data) is needed; however, it is beyond the scope of this work. In terms of variation trend, they are in good agreement with those of Ricker et al. (2018). As shown in Figs. 5 and 7, the averaged sea ice thickness and drift reveal a significant seasonal cycle. That is, the variations of sea ice thickness and motion always accompany with spring increase and autumn decrease. The analysis of ice concentration shows a steadily low values in the melt season. And the 6-year mean sea ice thickness, concentration and drift averaged over the entire Fram Strait gate are about 1.7 m, 85 % and  $5 \text{ km d}^{-1}$ .

Following Ricker et al. (2018), the relative standard deviation ( $\text{RSD} = \text{SD} / \text{mean}$ ) is used to measure the effects of different sea ice variables on the variability of the ice volume



**Figure 2.** (a) The trajectories of nine separate flights of HEM measurement campaigns carried out in the Fram Strait and the red dot denotes the site of ULS. Scatter plots of (b) daily mean sea ice thickness derived from CMST and HEM data, (c) monthly mean sea ice thickness derived from CMST and ULS data, and (d) monthly average sea ice drift based on CMST and OSI SAF.

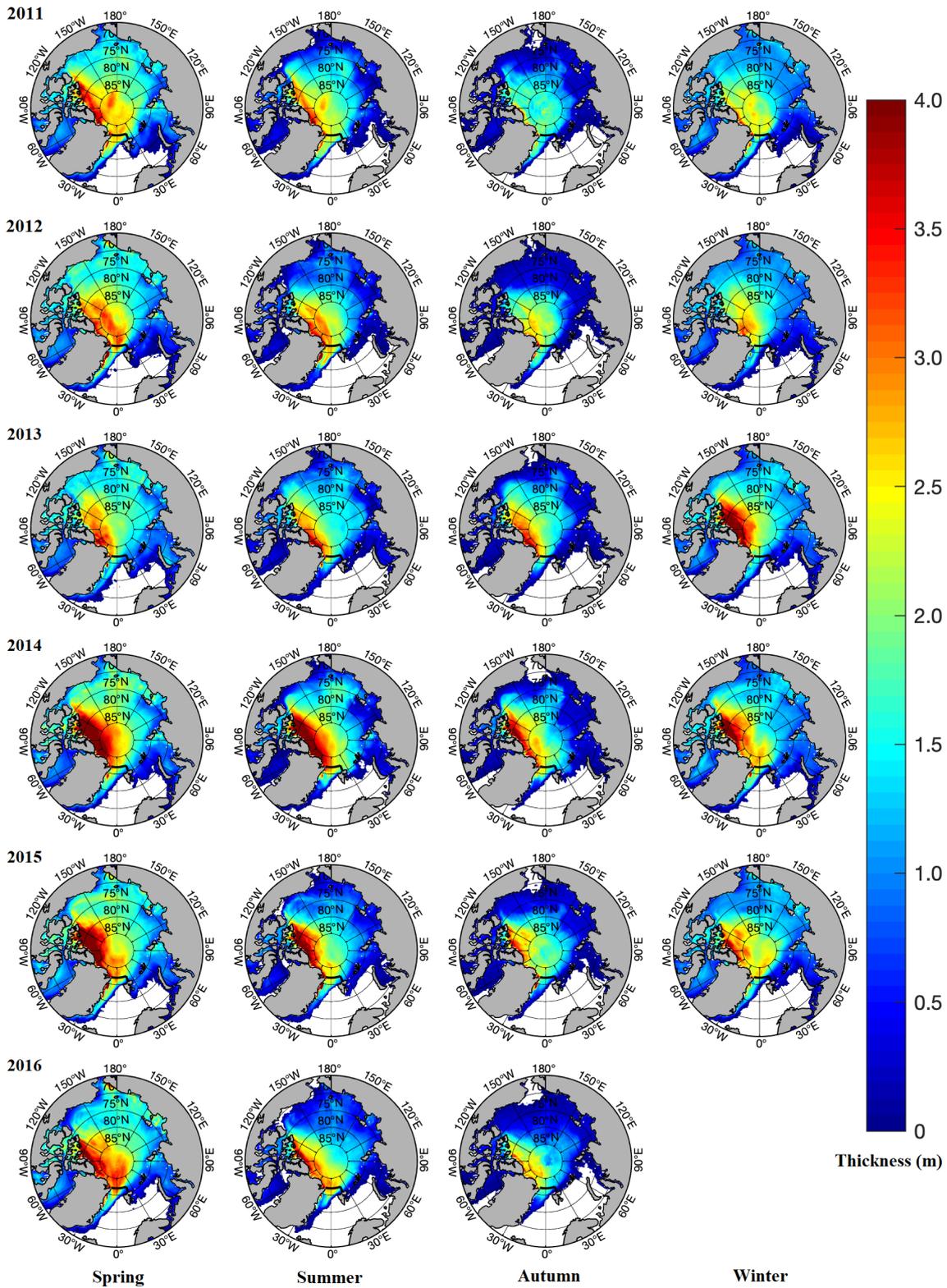
output. Variables with a larger RSD contribute to a greater impact on the volume variation. As shown in Figs. 5, 6 and 7, the RSD of ice thickness is 0.30, which is about twice that of ice concentration (0.14). The ice drift is the largest contributor with an RSD of 0.50. It is shown that the ice drift with maximal RSD is more likely to affect variations in sea ice volume flux, which corresponds to the previous findings in Kwok and Rothrock (1999), Ricker et al. (2018) and Bi et al. (2018).

To analyze the respective contributions of ice drift and ice thickness to the seasonal variation in sea ice export, the frequency distributions of seasonal sea ice thickness (Fig. 8), drift (Fig. 9) and concentration (not shown owing to the minimum RSD) along the Fram Strait outlet are further calculated. Specifically, we define the relative frequency (RF) as following

$$RF = \frac{n}{N_{\text{grids}}}, \quad (8)$$

where  $n$  represents the number of the grid cells accounted for in different thickness bins, and  $N_{\text{grids}}$  is the sum of  $n$ . As suggested by Fig. 8, the thickness along the zonal gate is much

thicker than the meridional gate. Thin ice is more observed in autumn over the zonal gate according to the RF distribution in Fig. 8. Although the maximum thickness over the entire Fram Strait occurs in May and June (Fig. 5), higher RF values in thick ice bins are found in summer (June, July and August in our definition) over the zonal gate. Over the meridional gate, the ice thickness in summer and spring is almost uniformly distributed, while in autumn and winter, high RFs are more often found in thin ice bins. In statistics, the seasonal mean sea ice thicknesses are 2.06 m for spring, 2.11 m for summer, 1.32 m for autumn and 1.43 m for winter over the entire outlet, respectively. Nevertheless, the mean relative frequency of sea ice drift distribution (Fig. 9) shows that the ratio of summer sea ice drift lower than 6 km d<sup>-1</sup> is in the majority (more than 90% of zonal gate), indicating that the sea ice drift is much slower than other seasons. Also, the ice drift along the zonal gate is usually faster than the meridional gate and the meridional sea ice velocities are slower than 6 km d<sup>-1</sup> during summer. The seasonal mean sea ice velocity over the entire gate is larger than 5 km d<sup>-1</sup>, except in summer when it is 3 km d<sup>-1</sup>. And it can be found that the



**Figure 3.** Seasonal variation in Arctic sea ice thickness. The thick black line represents the sea ice flux gate in the Fram Strait used in this study.

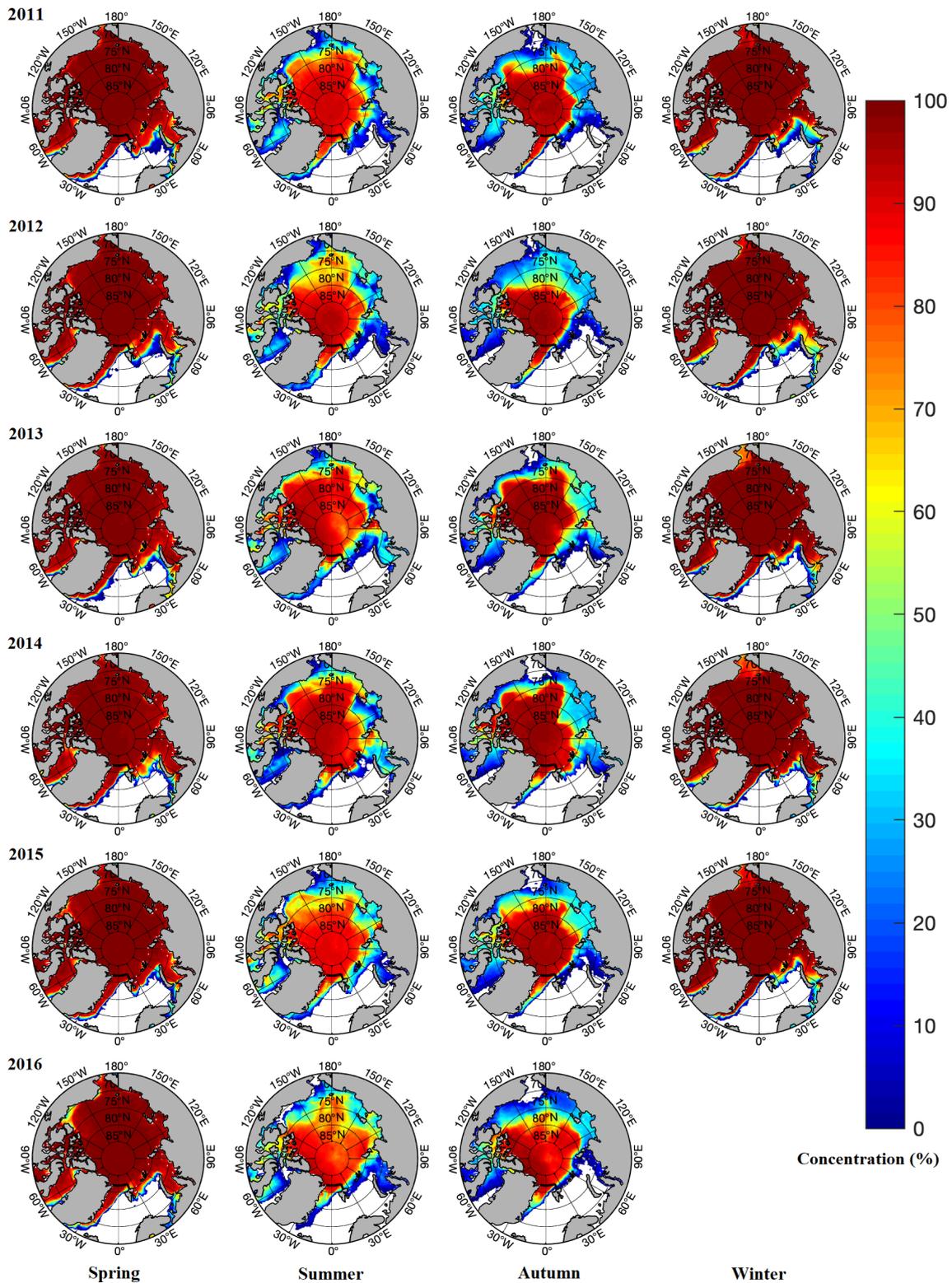
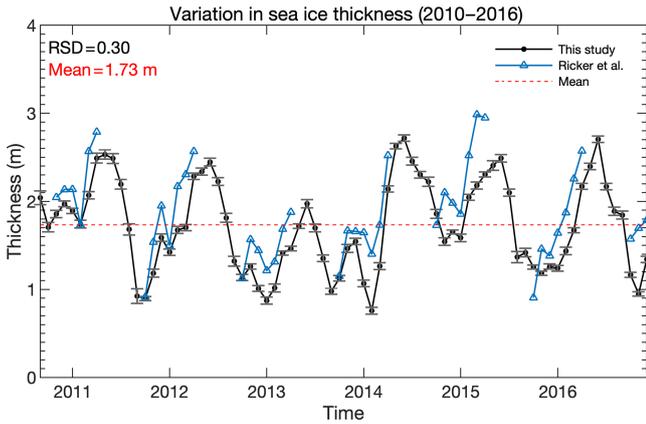
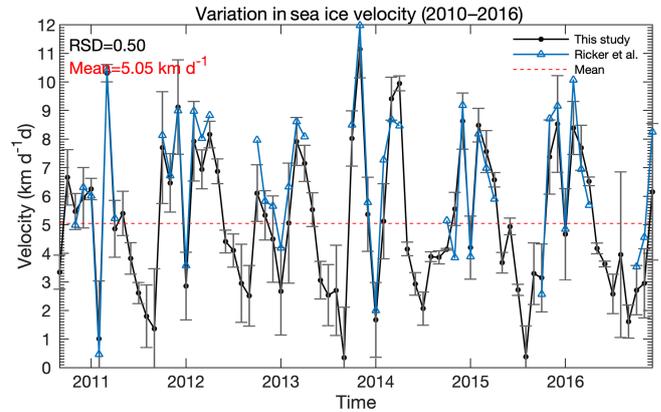


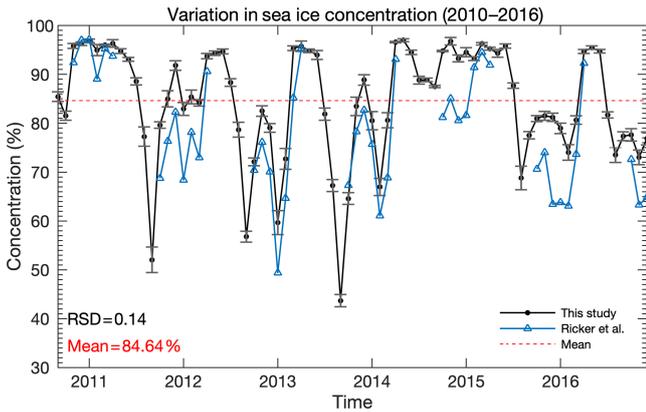
Figure 4. Seasonal variation in Arctic sea ice concentration. The thick black line represents the sea ice flux gate in the Fram Strait.



**Figure 5.** CMST sea ice thickness averaged over the entire Fram Strait gate, from September 2010 to December 2016. The black dotted line denotes monthly mean ice thickness based on CMST data with corresponding standard deviations while the blue dotted line represents monthly mean effective sea ice thickness of Ricker et al. (2018).



**Figure 7.** CMST sea ice drift averaged over the entire Fram Strait gate, from September 2010 to December 2016. The black dotted line represents monthly mean ice drift based on CMST data with corresponding standard deviations while the blue dotted line shows the monthly mean ice drift of OSI SAF.

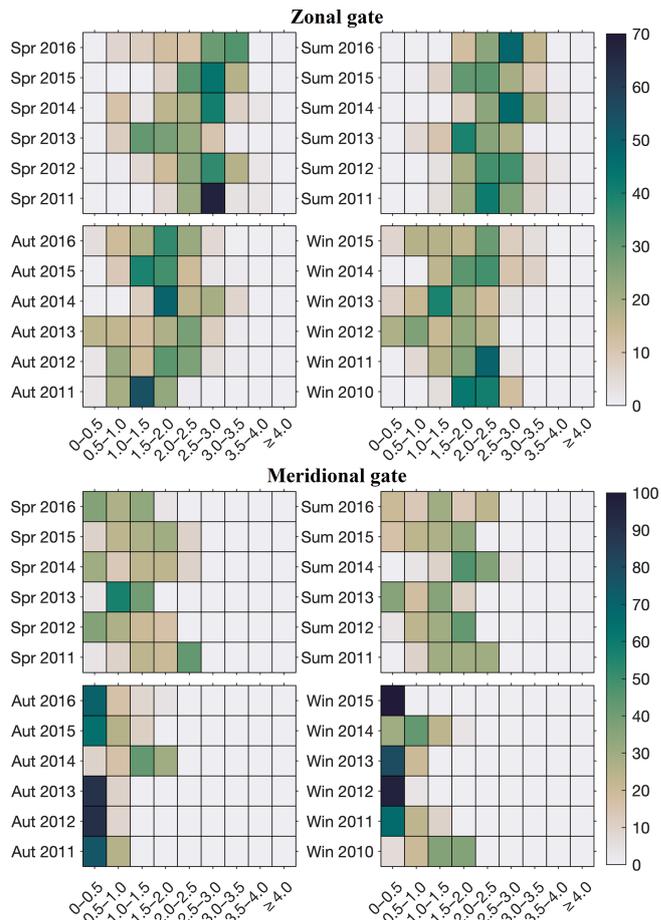


**Figure 6.** CMST sea ice concentration averaged over the entire Fram Strait gate, from September 2010 to December 2016. The black dotted line represents monthly mean ice concentration based on CMST data with corresponding standard deviations while the blue dotted line represents monthly mean ice concentration of OSI SAF.

spring and winter ice concentration along the zonal gate is larger than that of summer and autumn.

### 3.3 Sea ice volume export through the Fram Strait

In this section, sea ice volume export over all seasons is investigated. Firstly, the examination of monthly Arctic sea ice volume export through the Fram Strait is shown in Table 2. Both our results and Ricker et al. (2018) find that the maximum monthly sea ice export takes place in March 2011. The maximum of CMST data is  $-442 \text{ km}^3$ , which is less than that ( $-540 \text{ km}^3$ ) of Ricker et al. (2018). Consistently, the lowest sea ice output for each study occurs in February 2011 when excluding the melt season (May–September).



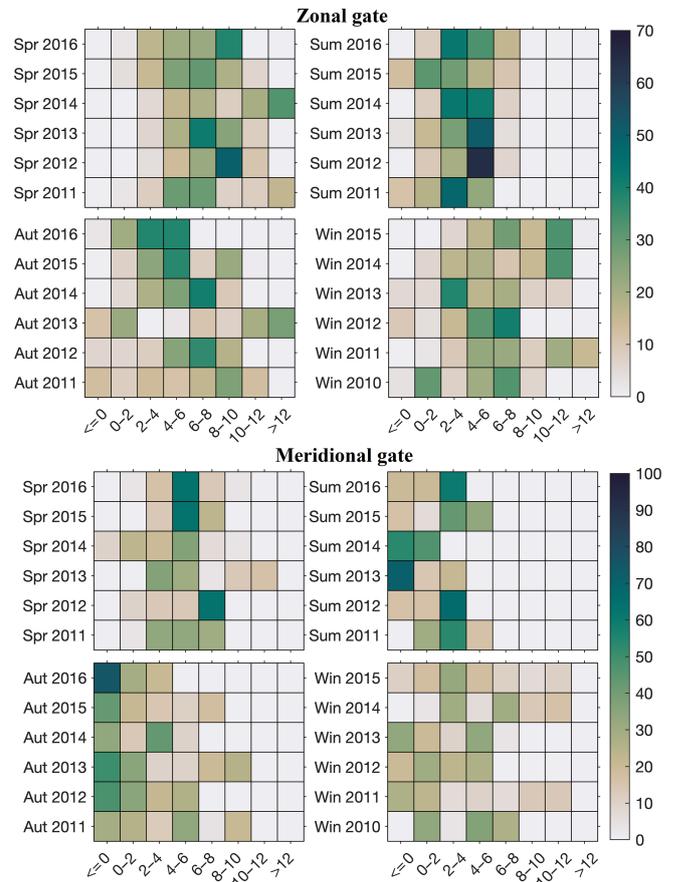
**Figure 8.** Seasonal variation in relative frequency (unit: %) of CMST sea ice thickness (unit: m) over the Fram Strait gate.

The minimum of the results shown in Ricker et al. (2018) is  $-21 \text{ km}^3$  while that is  $-34 \text{ km}^3$  in CMST data. Although there are some differences in flux calculated based on CMST data and CryoSat-2 thickness and OSI SAF drift data, both the estimations show a similar trend in annual cycle. Furthermore, the CMST data can provide sea ice variables (e.g., sea ice thickness, concentration and drift) in the melt season that remote sensing retrieval data cannot cover. Taking advantage of CMST data, this study is trying to fill the research gap in the summer sea ice volume export. It is found that another minimum of ice export occurs in August 2015 because of the rather slow mean sea ice velocity (shown in Fig. 11) during the study period. The minimum value for CMST is  $-11 \text{ km}^3$ , which is  $10 \text{ km}^3$  less than  $-21 \text{ km}^3$  ( $R$ ) in February 2011 and  $23 \text{ km}^3$  less than that for M2.

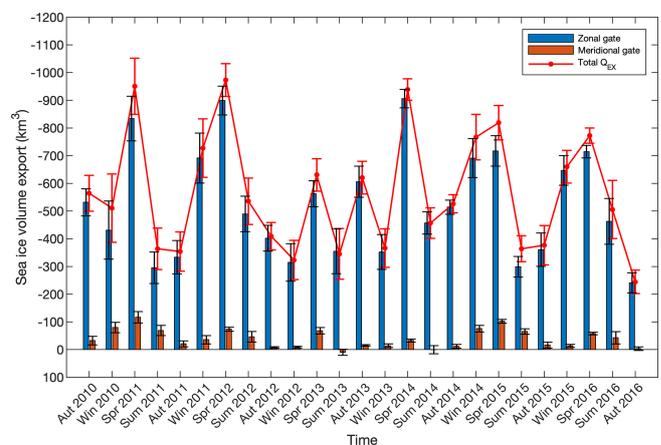
Moreover, the seasonal variation in sea ice export through Fram Strait is shown in Fig. 10. The ice volume output shows a significant seasonal variation. The seasonal maximums are found in spring of all years (2011–2016), and the low values usually occur in summer and autumn. The maximum seasonal ice export of  $-970(\pm 60) \text{ km}^3$  (sea ice volume export has been rounded off to significant figures in seasonal and interannual timescales) takes place in the spring of 2012 owing to both simultaneously faster ice drift and thicker ice thickness, while the minimum flux of  $-240(\pm 40) \text{ km}^3$  occurs in autumn of 2016 caused by simultaneously rather slower ice motion and thinner ice thickness. Unlike in other autumns, the ice volume export of autumn 2013 abnormally increases and reaches  $-620(\pm 60) \text{ km}^3$ . This abnormal increase can also be explained by the faster ice drift (shown in Fig. 9).

Furthermore, we standardize the sea ice volume export, ice drift and thickness and then calculate the correlations of determination ( $R^2$ ) between monthly sea ice volume export and thickness and also for sea ice drift (shown in Fig. 11). For monthly mean sea ice flux and drift,  $R^2$  is 0.77, which is much higher than that against thickness (0.16). This result shows that the sea ice drift variation contributes more to sea ice flux variation. However, when averaged over seasonal timescale, both the sea ice drift and thickness become significant factors for their close  $R^2$  within the range of 0.36–0.46. Analogously, this conclusion was pointed out by Ricker et al. (2018) and Bi et al. (2016). In addition, the Arctic Oscillation (AO) and North Atlantic Oscillation (NAO) index are used to analyze the possible links between atmospheric circulation and sea ice volume flux through the Fram Strait (Fig. 12). The AO and NAO indexes are both obtained from National Oceanic and Atmospheric Administration (NOAA). We calculate the seasonal mean AO and NAO index and find that the correlation of ice volume flux with AO index (0.55) is higher than that with the NAO index (0.34). Both our study and Ricker et al. (2018) find that the AO may influence the sea ice export (2011–2016) more directly.

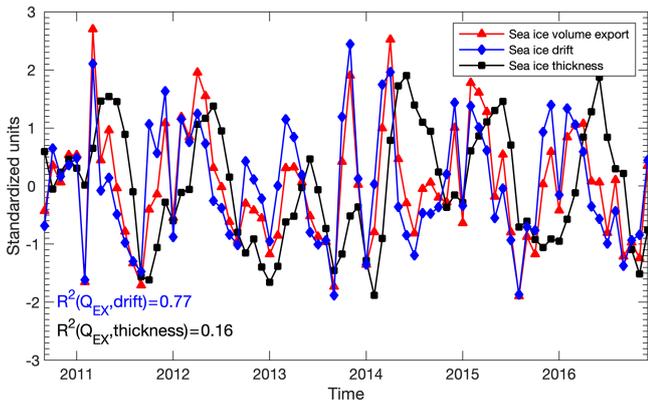
The CMST-based sea ice volume during both the melt season and the freezing season is first reported in this study. The estimations show that the mean ice volume export during the



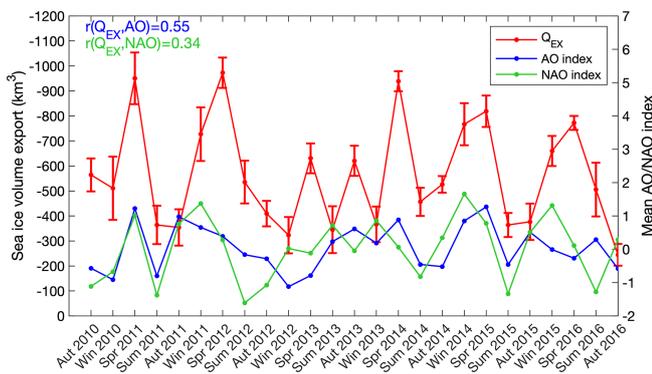
**Figure 9.** Seasonal variation in relative frequency (unit: %) of CMST sea ice drift (unit:  $\text{km d}^{-1}$ ) over the entire Fram Strait gate.



**Figure 10.** CMST seasonal Arctic sea ice volume export (unit:  $\text{km}^3$ ) through the Fram Strait with corresponding uncertainty.  $Q_{EX}$  represents the sea ice volume export based on CMST thickness and drift (in the following maps as well).



**Figure 11.** Time series of standardized monthly mean sea ice volume export ( $Q_{EX}$ , red line) and corresponding monthly mean sea ice drift (blue line) and sea ice thickness (black line), including correlation of determination ( $R^2$ ).

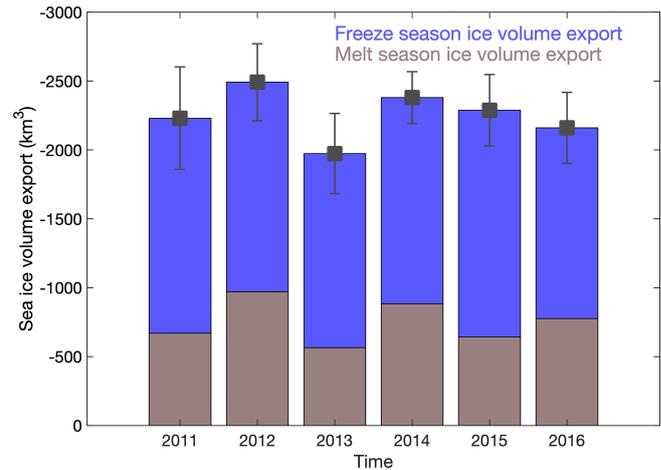


**Figure 12.** Time series of seasonal mean sea ice volume export ( $Q_{EX}$ , unit:  $\text{km}^3$ , red line) and corresponding mean seasonal AO (blue line) and NAO (green line) index, including correlation coefficient ( $r$ ).

melt season is  $-750(\pm 120) \text{ km}^3$ , which is about half of that during the freezing season ( $-1500 \pm 160 \text{ km}^3$ ). Annually, sea ice volume export (Fig. 13) is  $-2250 \text{ km}^3$  and varies from  $-1970(\pm 290)$  to  $-2490(\pm 280) \text{ km}^3$ . It is verified again that the annual sea ice volume export through the Fram Strait does not show a significant growth or decline trend (Ricker et al., 2018; Spreen et al., 2009). And the maximum yearly ice volume export occurs in the year of 2012, while the ice volume export reaches its minimum in 2013. This decline in ice volume export derives from the decreases in both thickness and drift through the Fram Strait.

#### 4 Discussions

The ensemble standard deviation (SD) map of CMST ice concentration, thickness and drift shows that uncertainties are larger downstream to the east of Greenland (Fig. 14). Therefore, following Krumpen et al. (2016) and Ricker et

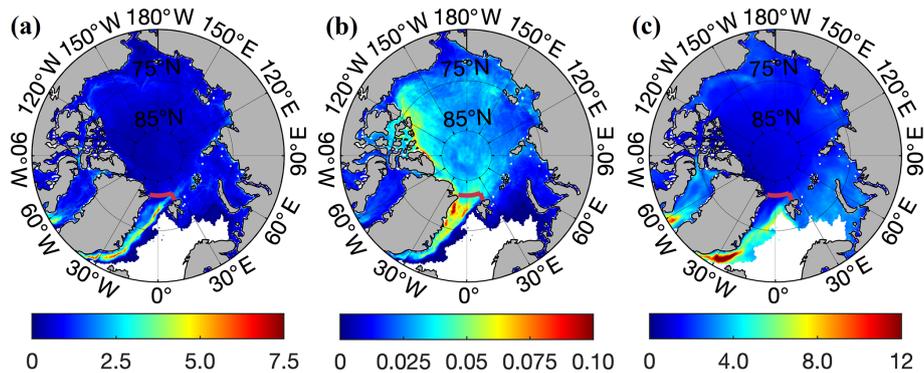


**Figure 13.** CMST interannual Arctic sea ice volume export (unit:  $\text{km}^3$ ) through the entire Fram Strait with corresponding uncertainty. The freezing season represents the months from October to April, and the melt season is from May to September.

al. (2018), a different gateway over the Fram Strait that consists of a zonal gate and a meridional gate located at a slightly higher latitude compared to previous studies is chosen (Kwok et al., 2004; Kwok and Rothrock, 1999; Spreen et al., 2009). Alternatively, the choice of a lower-latitude gate at  $79^\circ \text{ N}$  (e.g., the ULS moored sites) is suggested to utilize the ULS thickness for rough volume flux calculation. It should be noted that studies at different Fram Strait gates and over different periods will introduce deviations in the final ice volume estimation (Krumpen et al., 2016; Kwok et al., 2004; Kwok and Rothrock, 1999; Ricker et al., 2018; Spreen et al., 2009). For example, Ricker et al. (2018) investigated the sea ice flux in the Fram Strait and pointed out that the maximum ( $-540 \text{ km}^3$ ) occurs in March of 2011 and the minimum ( $-21 \text{ km}^3$ ) appears in February of 2011 from 2010 to 2017. However, for a different gate and period, Spreen et al. (2009) showed a relatively low maximum volume export of  $-420 \text{ km}^3$  and relatively high minimum flux ( $-92 \text{ km}^3$ ) in the freezing season.

We investigate a similar period with Ricker et al. (2018), but we further extend the sea ice volume flux estimation to the melt season. Also, the CMST sea ice thickness data used in this study are evaluated to be reasonable when compared with in situ observations (Mu et al., 2018a). The other important driver (sea ice drift) of ice volume export has also been compared with OSI SAF drift used in former estimations (Ricker et al., 2018) and Sentinel-1 SAR sea ice drift. The monthly mean CMST ice drift shows a better performance than NSIDC drift (Fig. 1), and meanwhile a good consistency is found between CMST and OSI SAF (Fig. 2d and 7) drift. Overall, the estimation of volume export in this study reveals a reasonable sea ice volume export all year round.

The nearly 6 years of ice volume export through the Fram Strait is calculated and shown in Table 2. In addition to the

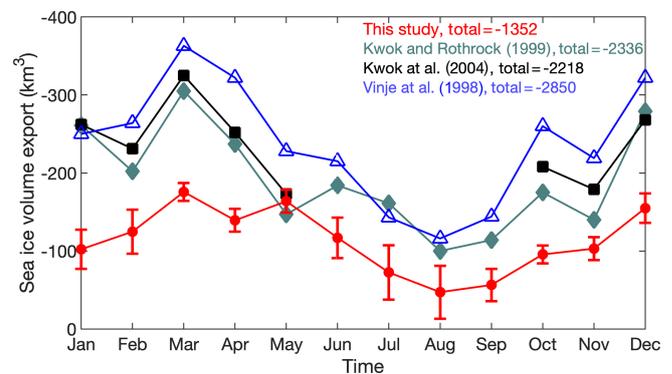


**Figure 14.** The mean ensemble standard deviation (SD) map of CMST (a) sea ice concentration (unit: %), (b) sea ice thickness (unit: m) and (c) sea ice drift (unit:  $\text{km d}^{-1}$ ) from September 2010 to December 2016. The thick red line represents zonal and meridional sea ice export gates to derive sea ice volume flux through the Fram Strait.

ice volume export ( $R$ ) of Ricker et al. (2018), we also calculate the export using OSI SAF drift and CMST thickness (M1) and CMST thickness and drift (M2), respectively. It can be concluded that  $R$  is larger than M1 and M2 because  $R$  is derived from thicker CS2 thickness (Fig. 5) and relatively faster OSI SAF drift (Fig. 7). In addition, M1 is generally larger than M2 also due to the faster OSI SAF ice motion for most periods. There are also cases when M2 is larger than M1 when CMST has higher ice motion than OSI SAF, for example, in March, April and November of 2014. Ricker et al. (2018) gave their multi-year averaged volume export of  $-1711 \text{ km}^3$  in the freezing season. Our average estimate (M2) based on the CMST ice thickness and drift is  $-1580 \text{ km}^3$ , while the volume flux (M1) derived from CMST thickness and OSI SAF drift is  $-1600 \text{ km}^3$  in the freezing season. The similar results between M1 and M2 are because the CMST drift deviates minorly compared to OSI SAF drift in the cold seasons. But more reliable validations of CMST ice drift need more in situ records and more systematic evaluations.

To further validate the sea ice volume export in the melt season, we compare our CMST-based volume flux (M2) with the relative short-term summer ice volume flux that Krumpfen et al. (2016) derived from airborne ice thickness and NSIDC ice drift data on the same export gates. The intercomparison shows that the sea ice volume export in August 2011 and July 2012 estimated by Krumpfen et al. (2016) is smaller than in this study. The underestimation of summer sea ice volume may derive from a general underestimation of NSIDC drift during the melt season (Krumpfen et al., 2016; Sumata et al., 2014, 2015).

Through the Fram Strait gate located at  $79^\circ \text{N}$ , Kwok and Rothrock (1999) and Kwok et al. (2004) investigated the summer sea ice export by using ULS thickness and area flux in the freezing season. The average annual ice volume flux is  $-2218 \text{ km}^3 \text{ yr}^{-1}$  from 1991 to 1998, while the mean sea ice volume export from 1990 to 1995 is  $-2366 \text{ km}^3 \text{ yr}^{-1}$  (Kwok et al., 2004; Kwok and Rothrock, 1999). To com-



**Figure 15.** Monthly mean sea ice volume export (unit:  $\text{km}^3$ ) at the  $79^\circ \text{N}$  transect in the Fram Strait from this study (red line), Kwok and Rothrock (1999, dark green line), Kwok et al. (2004, black line), and Vinje et al. (1998, blue line).

pare with previous studies (Kwok et al., 2004; Kwok and Rothrock, 1999; Vinje et al., 1998), we also calculate the sea ice volume flux through the same gate located at  $79^\circ \text{N}$ . Results (Fig. 15) show that our annual mean sea ice volume export ( $-1352 \text{ km}^3$ ) is smaller than that of these studies, which is expected because of the decline of sea ice thickness in recent decades. All these works show consistent seasonality with maximum export in March and minimum export in August. In a recent study, Wei et al. (2019) calculated the annual mean sea ice volume export ( $-3216 \text{ km}^3 \text{ yr}^{-1}$ ) through the Fram Strait from their simulation from 1979 to 2012. Their estimations give a long period of sea ice volume export through the Fram Strait, which can serve as an important reference when focusing on the long-term trend and the variations in the volume flux. However, this estimation seems to overestimate the volume flux owing to the overestimations of sea ice drift and thickness (Wei et al., 2019).

## 5 Conclusions

The CMST data over all seasons are first used to estimate ice volume export through the Fram Strait. Also, benefitting from the advantage of CMST data, the melt season (e.g., summer season and autumn season) ice volume export can be derived to fill the satellite data gap over such periods. The entire seasonal and interannual variations in Arctic sea ice volume are helpful for communities that focus on climate teleconnection between polar regions and low latitudes and Arctic freshwater transport and ocean circulation. Conclusions of this study can be drawn as follows.

1. The Arctic sea ice thickness and volume show a significant seasonal variation. The thickness and volume maximum usually occur in spring and the Arctic sea ice volume hits its minimum in autumn 2011 during the study period.
2. Along the entire Fram Strait gate, the relative standard deviation (RSD) of ice drift (0.50) is higher than the RSD of ice thickness (0.30) and concentration (0.14), demonstrating that ice drift is a main driver of ice volume export through the Fram Strait. The correlations of determination ( $R^2$ ) also show that sea ice drift is a much more important contributor for sea ice volume export on a monthly scale.
3. The mean sea ice volume export during the melt season is around  $-750(\pm 120)$  km<sup>3</sup>, which is about 50% of that during the freezing season ( $-1500 \pm 160$  km<sup>3</sup>). The lowest and largest annual sea ice volume export occur in 2013 and 2012, respectively. Seasonal sea ice volume export varies from  $-240(\pm 40)$  to  $-970(\pm 60)$  km<sup>3</sup>, while the monthly sea ice export varies between  $-11$  km<sup>3</sup> (August of 2015) and  $-442$  km<sup>3</sup> (March of 2011) during this study period. The abnormal ice volume export increase in autumn 2013 is primarily associated with the faster ice motion.
4. The relative frequency (RF) of seasonal variation in CMST sea ice thickness shows that sea ice thicker than 2 m in spring and summer is more than that in other seasons. The summer mean ice drift that is lower than  $6 \text{ km d}^{-1}$  is the majority in each year.

An updated and improved CMST version 2 sea ice data set will be developed in the near future to obtain a long-term record for climate research.

*Data availability.* The CMST sea ice thickness and drift data are available at <https://doi.org/10.1594/PANGAEA.891475> (Mu et al., 2018c) and <https://doi.org/10.1594/PANGAEA.906973> (Mu et al., 2019), respectively (Mu et al., 2018a). The OSI SAF drift data can be downloaded at [ftp://osisaf.met.no/archive/ice/drift\\_lr](ftp://osisaf.met.no/archive/ice/drift_lr) (EUMETSAT, 2010–2016). The latest released Polar

Pathfinder Daily 25 km EASE-Grid sea ice drift data are provided by the National Snow and Ice Data Center (NSIDC, <https://doi.org/10.5067/INAWUWO7QH7B>, Tschudi et al., 2019). The Sentinel-1 SAR sea ice drift data can be downloaded at (<https://www.ncbi.nlm.nih.gov/pmc/articles/PMC5999601/>, Muckenhuber and Sandven, 2018). The helicopter-borne electromagnetic induction sounding (HEM) sea ice thickness data are available at <https://doi.org/10.21334/npolar.2017.1ed8c57e> (King et al., 2017; last access: 3 May 2019). The AO and NAO indices can be downloaded at [https://www.cpc.ncep.noaa.gov/products/precip/CWlink/daily\\_ao\\_index/ao.shtml](https://www.cpc.ncep.noaa.gov/products/precip/CWlink/daily_ao_index/ao.shtml) (Thompson and Wallace, 2019) and <https://www.cpc.ncep.noaa.gov/products/precip/CWlink/pna/nao.shtml> (Hurrell, 2019).

*Author contributions.* LM and QY conceptualized this study and provided the CMST sea ice data. CM conducted this study, performed the calculation and drafted the paper. RR supplied the sea ice data of Ricker et al. (2018) for intercomparison. LM, QY and RR polished this paper and improved the readability. QS, RW, BH and JL reviewed this paper.

*Competing interests.* The authors declare that they have no conflict of interest.

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