

Estimation of duration and its changes in Lagrangian observations relying on ice floes in the Arctic Ocean utilizing sea ice motion product

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Abstract: Since the 1890s, buoy- and camp-based Lagrangian observations relying on ice floes have been indispensable for data acquisition in the difficult-to-access central Arctic Ocean in winter. Evaluating the potential observation duration and its changes associated with changes in Arctic climate system, is crucial for the planning of ice camp/buoy deployment. Using remote sensing sea ice motion product, we reconstructed sea ice drift trajectories for each annual cycle from 1979–1980 to 2022–2023 and identified the ideal areas of ice camp/buoy deployment in the central Arctic Ocean. The results show that, based on the setup time of October 1, the areas centered at 82°N and 160°E near north of the East Siberian and Laptev seas, with a size of $7.0 \times 10^5 \text{ km}^2$, could ensure Lagrangian observations for at least 9 months with the drifting maintaining in the ice zone and not entering the exclusive economic zones (EEZs) of Arctic coastal countries, with the probability of 75.0%–90.9% during 44 years. The potential deployment areas favored ice advection to the Transpolar Drift (TPD) region relative to the Beaufort Gyre (BG) region. Ice trajectory terminal points did not reveal an obvious long-term tendency, but were regulated by large-scale atmospheric circulation patterns, especially the atmospheric patterns in the early drifting stage of autumn (OND). In particular, the autumn east-west surface air pressure gradient across the central Arctic and the Arctic Dipole Anomaly indices significantly influenced terminal points of ice trajectories after 9 months and can expand ideal deployment areas under scenarios with their extreme positive phases. The increasing rate of autumn–spring near-surface air temperatures along the trajectories was more pronounced in the TPD region than that in the BG region. The sea ice response to wind stress significantly intensified in recent Lagrangian observations, suggesting stronger ice dynamic processes as the sea ice thins. Geopolitical boundaries of EEZs have a significant impact on the sustainability of the Lagrangian observations, making it rarely exceed 10 months. Without this restriction, the potential Lagrangian observations in the BG and TPD regions would expand southward, with the increased duration by 20.5 and 5.0 days, respectively, compared to those with the EEZ restriction.

KEYWORDS: Arctic Ocean; Sea ice; Lagrangian observation; Buoy; Ice camp; Transpolar Drift; Beaufort Gyre

1. Introduction

Arctic sea ice, a crucial indicator and amplifier for climate change (Kwok, 2018), has experienced pronounced to become progressively thinner and younger, with its extent in September declining by 13% per decade during the satellite observation era since 1979 (Parkinson and DiGirolamo, 2021; Meier and Stroeve, 2022; Babb et al., 2023). The state-of-the-art earth system models still have an obvious spread to project the evolution of Arctic sea ice (Holland and Hunke, 2022), mainly due to insufficient observational data for parameterizations of sea ice thermodynamic and dynamic processes (Smith et al., 2022), a severe absence of reliable observation data available for assimilation (Liu et al., 2019), and the rough treatment of Arctic snow and sea ice processes by atmospheric reanalysis data (Batrak and Müller, 2019). The frozen ocean and extremely harsh weather limit the accessibility of the central Arctic Ocean, exacerbating data scarcity of ship-based oceanography measurements. This situation is even worse in the freezing season (Rabe et al., 2022).

Lagrangian measurements based on ice camp or buoy deployed on ice floes provide an alternative for the observations of interactions among atmosphere, ice, and ocean in the Arctic. In the 1890s, Fridtjof Nansen and his companions pioneered Lagrangian observations in the central Arctic Ocean using the ice camp and wooden galleon, which finally provided the first basic depiction of the Arctic sea ice and oceanic physical regimes. Subsequent ice-camp-based campaigns, including the Ice Station Alpha (Cabaniss et al., 1965), the Arctic Ice Dynamics Joint Experiment (AIDJEX; Coon, 1980), the Surface Heat Budget of the Arctic Ocean (SHEBA) campaign (Uttal et al., 2002), as well as the Norwegian young sea ICE (N-ICE2015) Expedition (Granskog et al., 2016). These efforts provided vital observation data for the construction of the theoretical framework of sea ice physics, as well as the parameterizations of sea ice thermodynamic and dynamic processes, and heat and/or salt exchanges with lower atmosphere and/or upper ocean, promoting the developing of sea ice numerical models. The Soviet Union-Russia Arctic ice-camp project, lasting for several decades since the 1930s, have provided climatological characteristics of extensive snow and sea ice geophysical variables over the central Arctic Ocean (Frolov, 2005), supporting numerical simulations (e.g., Tian et al., 2024) and remote sensing retrieval algorithms of Arctic sea ice (e.g., Lavergne et al., 2010). Recently, the Multidisciplinary drifting Observatory for the Study of the Arctic Climate (MOSAiC), fully leverages the advantages of multidisciplinary observations on the ice floes as an intermediate medium (Nicolaus et al., 2022; Rabe et al., 2022; Shupe et al., 2022), marking a milestone for Arctic drifting observation campaigns.

However, the implementation of ice camp, accompanied by a modern icebreaker as the MOSAiC, requires a significant logistical budget; or without the icebreaker supporting, as the Soviet Union-Russia ice camps, faces with risks including those from ice floe fragmentation, storms, and polar bears. These factors all limit the sustainable implementation of ice camp. It is gratifying that Arctic ice floes also provide a broad platform without the need for extra floating support for deploying

57 buoys or other observation instruments. Various types of buoys are designed and deployed in the Arctic Ocean to measure
58 sea ice kinematics (Lukovich et al., 2011), snow and sea ice mass balance processes (Richter-Menge et al., 2006; Jackson et
59 al., 2013; Nicolaus et al., 2021; Lei et al., 2022), meteorological parameters and heat exchanges over ice surface (Cox et al.,
60 2023), as well as oceanic temperature and salinity profiles or turbulence heat flux underneath the ice (Shaw et al., 2008;
61 Toole et al., 2011). Co-deployment of various types of buoys on the same floe to obtain comprehensive observation matrix of
62 multiple media (e.g., Morison et al., 2002), or at a local scale of tens of kilometers (e.g., Rabe et al., 2024) to match the grid
63 scales of satellite remote sensing (e.g., Koo et al., 2021) and numerical models (e.g., Pithan et al., 2023). Such task is
64 extremely hard to achieve in open water.

65 The effective duration of Lagrangian observations relying on the ice floes, and the observation regions they may drift
66 through, is highly dependent on the setup or deployment location of ice camp or buoy. This is because the Arctic sea ice
67 advection is primarily regulated by two surface ocean circulation systems of the Beaufort Gyre (BG) and Transpolar Drift
68 (TPD) (Kwok et al., 2013). Remote sensing sea ice motion (SIM) products can be used to simulate forward (backward) ice
69 drift trajectories to track the destinations (origins) of ice floe (Lei et al., 2019) or estimate ice age by tracking the duration of
70 ice drifting (Tschudi et al., 2020). Krumpen et al. (2019) analyzed the interannual variations in the advection patterns based
71 on the reconstructed ice trajectories using a Lagrangian tracking approach, and investigated the impacts of ice advection
72 patterns on ice mass or ice-associated material transmission in the TPD region. On this basis, Krumpen et al. (2020)
73 estimated the original region of ice floe for the setup of the MOSAiC ice camp, in order to support interdisciplinary
74 observations at the ice camp. In the practical process of establishing MOSAiC ice camp, the reconstructed ice drift trajectory
75 using the SIM product is also an important basis for selecting the deployment location of ice camp. From the perspective of
76 logistics, Granskog et al. (2016), Nicolaus et al. (2022), and Rabe et al. (2024) emphasized the impacts of dynamic
77 deformation of ice floes on the safety of observation equipment deployed over the ice camps. However, there is still a lack of
78 a comprehensive assessment of how deployment locations influence the subsequent drifting and destinations of ice camps
79 and buoys at the basin scale in the Arctic Ocean. This study aims to address this gap by identifying the ideal deployment
80 regions for ice camp or buoy using SIM products. The goal is to ensure uninterrupted observations while avoiding risks such
81 as the drifting to the ice edge or to the exclusive economic zones (EEZs) of one country that is not involved in observation
82 experiments within a shorter period than the expected duration.

83 During the Lagrangian observations, the atmospheric thermodynamic and dynamic forcing not only determine the
84 seasonal evolution of sea ice itself, but also affect the heat and momentum exchanges between the atmosphere and sea ice.
85 They can provide important backgrounds supporting the interdisciplinary studies based on Lagrangian observational data
86 (e.g., Krumpen et al., 2021; Rinke et al., 2021). Under the background of Arctic amplification and sustained loss of Arctic

87 sea ice, understanding the climatological characteristics and long-term trends of atmospheric forcing and sea ice kinematics
88 along the subsequent potential drifting trajectory is also essential for planning the deployments of ice camp or buoy.
89 Moreover, it has gained widespread consensus, some atmospheric circulation patterns play a critical role in regulating Arctic
90 sea ice advection. The Arctic Oscillation (AO) (Thompson and Wallace, 1998) regulates the axis alignment of the TPD and
91 the extent of BG. At positive (negative) AO phases, the axis alignment of the TPD tends to shift westward (eastward) and the
92 BG shrinks (expands) (Rigor et al., 2002). The wind anomalies induced by Dipole Anomaly (DA) (Wu et al., 2006) exhibit
93 strong meridional forcing in the TPD region, with positive (negative) phases accelerating (decelerating) the sea ice drift
94 along TPD (Wang et al., 2009). The Central Arctic air pressure-gradient Index (CAI), defined as the east-west gradient of sea
95 level air pressure (SLP) across the central Arctic Ocean could regulate partly meridional wind forcing parallel to TPD
96 (Vihma et al., 2012). The Beaufort High (BH) (Moore et al., 2018) is closely associated with sea ice circulation in the BG
97 region (Proshutinsky and Johnson, 1997). These atmospheric circulation patterns affect sea ice drift trajectory and advection
98 direction through various mechanisms and consequently affect the duration of Lagrangian observations relying on the ice
99 floes. Thus, their regulatory mechanisms and seasonal variations needs further clarification regarding the evaluation of
100 duration of the Lagrangian observations using Arctic ice floes.

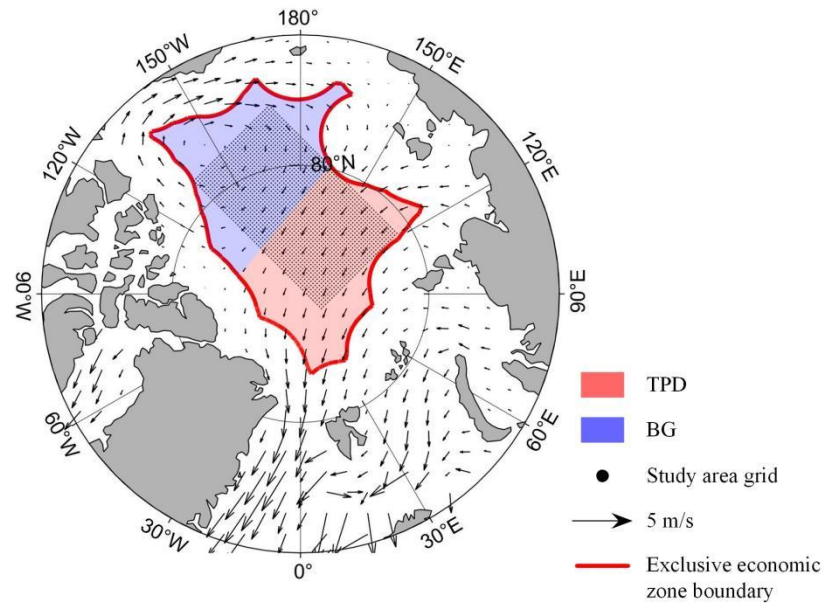
101 In this study, we organized the sections as follows. The datasets and methods used to reconstruct the sea ice drift
102 trajectory and estimate the changes in atmospheric and ice conditions along the trajectory are briefly described in Sect. 2.
103 The ideal deployment areas of Lagrangian observations, as well as changes in the atmospheric thermodynamic and dynamic
104 forcing along the potential ice trajectories during each annual cycle from 1979–1980 to 2022–2023 (hereinafter referred to as
105 1979–2022) are presented in Sect. 3. The performance of the reconstructed method, the connection of the ice trajectories
106 with the atmospheric circulation patterns, and the impact of EEZ boundary and deployment time on the sustainability of
107 Lagrangian observations are discussed in Sect. 4. Conclusions are given in the last section. This study provides important
108 cognitive foundation for the planning and implementation of Lagrangian observations relying on ice floes in the central
109 Arctic Ocean.

110 **2. Data and methods**

111 **2.1 Study area**

112 Our study focuses on the reconstruction of sea ice drift trajectory in the central Arctic Ocean because it is difficult to
113 maintain sufficient drifting observation time in the peripheral sea. Here, the central Arctic Ocean is defined as the high Arctic
114 that excluded from the EEZs of any country, using the maritime boundary polylines (version 12) of the geodatabase provided

115 by the Flanders Marine Institute. To define the potential areas for identifying preferred deployment sites, we identified a
 116 rectangular area of $1.44 \times 10^6 \text{ km}^2$, consisting of 2294 pixels on the 25-km polar stereographic grid, with area corners aligned
 117 with the EEZ boundary polylines, which covers approximately 53.5% of the central Arctic Ocean we defined (Fig. 1). The
 118 definition of such a rectangular area can effectively reduce computational cost relative to surveying the entire central Arctic
 119 Ocean. To demonstrate its rationality, we conducted surveys for the potential preferred area for ice camp or buoy deployment
 120 over the entire central Arctic Ocean using 1979–2023 climatology of SIM field. Results reveal that the effective starting
 121 points for deployment of ice camp or buoy with the subsequent drifting trajectory ≥ 9 months, avoiding drifting into the
 122 EEZ or beyond ice zone, occupy 91.6% of the defined rectangular area, which is higher than that (36.0%) for other regions
 123 within the central Arctic Ocean. Even during the most recent decade of 2013–2023, a period facing more severe and complex
 124 challenges due to thinner and younger ice, the rectangular area still contained 25.6% effective starting points, while other
 125 regions had none. Thus, in other regions within the central Arctic Ocean beyond our defined rectangular area, it is extremely
 126 difficult to identify suitable areas to deploy ice camps or buoys to maintain subsequent sufficiently long Lagrangian
 127 observations. Considering to save computational cost, we believe that definition of this rectangular area for reconstruction
 128 and analysis of sea ice drift trajectories in specific years is reasonable. On the basis of the 1979–2023 climatology of Arctic
 129 SIM field, we also roughly defined boundaries to separate the BG and TPD regions, as shown in Fig. 1.



130
 131 **Figure 1.** Study area. The black dots indicate grid points defined to identify the optimal area for the buoy or camp deployment. The arrows
 132 depict the mean SIM vectors from 1979 to 2023. The region delineated by the red lines represents the central Arctic Ocean, which is
 133 defined as the high Arctic that excluded the EEZs. The shaded blue and red areas roughly denote the Beaufort Gyre and Transport Drift
 134 regions within the central Arctic Ocean.

2.2 Data

a. Sea ice data

Due to the difficulty of obtaining long-time series, large-coverage SIM fields from high-resolution remote sensing images (e.g., Li et al., 2022; Fang et al., 2023), we used the 25-km Polar Pathfinder version 4.1 Sea Ice Motion Vectors from the U.S. National Snow and Ice Data Center (NSIDC; Tschudi et al., 2020) to reconstruct sea ice drift trajectories originating from the study area in 1979–2022. The Global Sea Ice Concentration Climate Data Records from the European Organization for the Exploitation of Meteorological Satellites Ocean and Sea Ice Satellite Application Facility (EUMETSAT OSI SAF; Lavergne et al., 2019) is utilized for evaluating ice conditions along the trajectory. This sea ice concentration (SIC) data is derived from the Scanning Multichannel Microwave Radiometer (SMMR), Special Sensor Microwave Imager (SSM/I), and Special Sensor Microwave Imager/Sounder (SSMIS) passive microwave satellite series sensors. The SIM and SIC data are projected onto the 25-km polar stereographic grid.

b. Buoy data

The trajectories of the buoys deployed over the Arctic ice were utilized to validate the reconstructed ice trajectories. To ensure the quality of validated SIM product, we constrained buoy selection to those situated 100 km offshore within the Arctic Ocean and excluded buoys south of the Fram Strait. The data of these buoys were derived from the International Arctic Buoy Programme (IABP), the German Arctic Research Expedition, and the Chinese National Arctic Research Expedition (CHINARE) during the summers of 2014, 2016, and 2018. The data collected by 30 buoys from 2010 to 2023, with 15 in each of the BG and TPD regions, were utilized for the validation of reconstructed ice trajectories in this study, with the details given in Table A1.

c. Atmospheric data

Atmospheric conditions were examined using atmospheric reanalysis data from the European Centre for Medium-Range Weather Forecasts Reanalysis v5 (ERA5; Hersbach et al., 2020). Hourly near-surface (2 m) air temperature, 10-m wind, snowfall, total precipitation and surface longwave radiation at about 30-km horizontal resolution are bilinearly interpolated to derive daily atmospheric conditions along the trajectories. ERA5 data is the next generation of the ERA-interim product. It has significant improvements in both temperature and wind fields in the Arctic Ocean (Graham et al., 2019b), and is able to reasonably characterize snowfall and precipitation (Wang et al., 2019), and the surface net longwave radiative flux under various weather conditions (Graham et al., 2019a).

Seasonal (autumn-OND, winter-JFM and spring-AMJ) indices of AO, DA, CAI, and BH were used to characterize the regulatory mechanism of atmospheric circulation patterns on ice trajectories. The AO and DA indices were calculated from the first and second empirical orthogonal functions of the SLP anomalies north of 70°N, utilizing monthly SLP from the National Centre for Environmental Prediction/National Centre for Atmospheric Research to maintain consistency with previous studies (Wu et al., 2006; Wang et al., 2009). Hourly SLP from ERA5 reanalysis was used to calculate the monthly CAI (Vihma et al., 2012), defined as the difference between SLPs at 90°W, 84°N, and 90°E, 84°N. According to Moore et al., (2018), the ERA5 SLP anomaly over the region of 75°–85° N and 170°E–150°W were utilized to define the BH index, which is more compatible with the BG from the perspective of sea ice circulation.

2.3 Methods

To assess the effective Lagrangian observation time, a survival time (ST) threshold for floes still drifting within the Arctic ice region and avoiding entering EEZs is crucial. Based on a given ST threshold, regular grids (Fig. 1) were established as the starting point for ice trajectory reconstruction to identify the preferred potential deployment areas of ice buoy or camp. Reconstructed ice trajectories from these areas start on October 1, aligning with the approximate onset of ice surface freezing season (Markus et al., 2009) and the setup time (October 3) of MOSAiC ice camp (Nicolaus et al., 2022). According to Lei et al., (2019), the ice drift trajectories were reconstructed as follows:

$$X(t) = X(t-1) + U(t-1) \cdot \delta_t, \quad (2)$$

$$\text{and } Y(t) = Y(t-1) + V(t-1) \cdot \delta_t, \quad (3)$$

where X and Y are the zonal and meridional coordinates of ice trajectories, $U(t)$ and $V(t)$ are the ice motion components at the time t along the ice trajectories, and the δ_t is the calculation time step of one day. The ice drift speed is interpolated using bilinear interpolation.

When ice floes enter region with SIC < 15% or the EEZ of one country, the reconstructed ice trajectory is terminated, and the time from October 1 to the terminal point is defined as the ST of ice floe, corresponding to the effective working duration of an ice camp or a buoy that is deployed on it. Note that this study truncated reconstructed ice trajectories exceeding one year (after 30 September of the following year) to focus on identifying areas unsuitable for deploying ice camp or buoy where the ST of reconstructed trajectory did not meet the threshold.

Due to the fact that the SIM products were projected onto the 25-km polar stereographic grid, and the rectangular area based on this grid was used as the starting point for ice trajectory reconstruction, the proportion of effective starting points relative to the total grid points, corresponding to the a given threshold of ST, can be equivalent to the area ratio over the

study region. As shown in Fig. 2a, with a 10-month ST threshold, the available deployment areas in the central Arctic Ocean, having the ST not less than 10 months with the probability >75% during 44 study years, are very limited (21.6% of the rectangular study region), which is less than the 48.6% area when the ST threshold is set as 9 months. The probability with a relative short duration less than 180 days was 1.9% for the effective region corresponding to the ST threshold of 6 months, which was reduced to a negligible value of 0.8% (1.1%) for the 10-month (9-month, not shown) ST threshold; while that with a relative long duration of 365 days or beyond was 70.9% for the ST threshold of 6 months, which increased to 82.6% (77.8%) for the 10-month (9-month, not shown) ST threshold (Fig. 2b). Therefore, to ensure a broad range of deployment areas, i.e., with a probability of > 50% across the entire study region, and ensure sufficient duration for Lagrangian observations, we used a 9-month ST threshold for the subsequent analyses. It means the ST lasts until the end of June of following year after the deployments, which is close the first drifting phase of MOSAiC ice camp (Nicolaus et al., 2022). With this ST threshold, almost all the peripheral area of our defined rectangular region are identified as unsuitable for ice camps and buoy deployment (Fig. 2a). This again proves that it is difficult to identify suitable sites for ice camp or buoy deployments in other regions beyond our defined rectangular region in the central Arctic Ocean, and also confirms the rationality of our definition of this rectangular region.

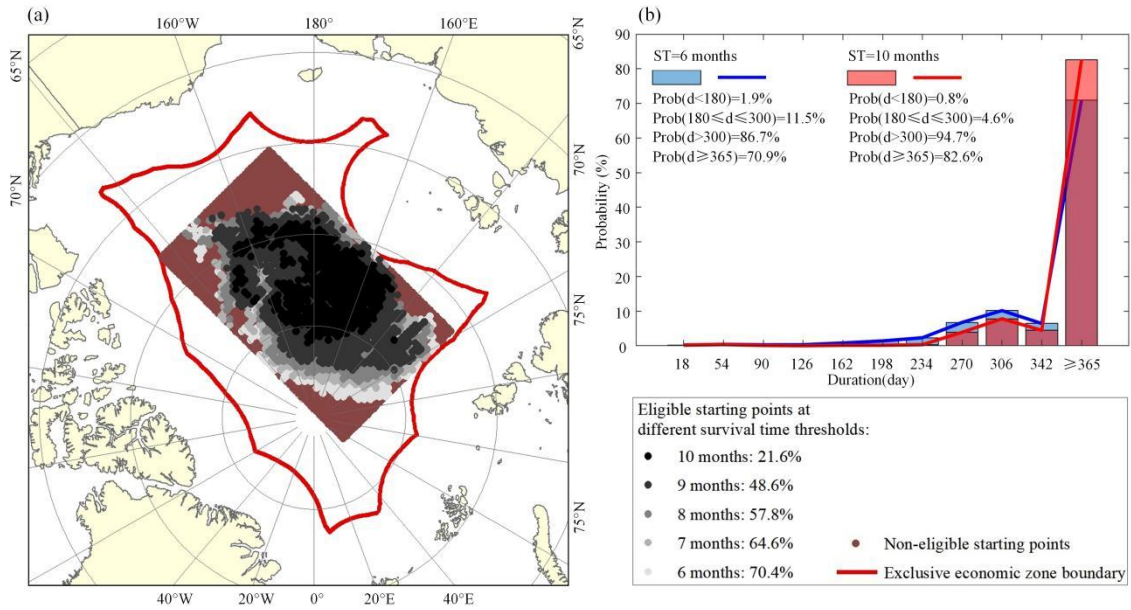


Figure 2. (a) Spatial distribution of eligible starting points and (b) probability distribution of duration according to different thresholds of survival time in 1979–2022. Note that the truncation of reconstructed ice trajectory in this study is one year until 30 September next year. Thus, the proportions with the duration ≥ 365 days means the ice trajectory is still valid by 30 September of following year.

To verify the reliability of reconstructed ice trajectories, the Euclidean distance and cosine similarity against the buoy observations are used to quantify their distance and direction deviations. The Euclidean distance (D_f) is defined as follows:

$$D_f = \sqrt{(x_{buoy} - x_{cal})^2 + (y_{buoy} - y_{cal})^2}, \quad (3)$$

where the subscripts of *buoy* and *cal* denote the coordinates of buoy trajectory and reconstructed ice trajectory, respectively. Cosine similarity is an effective metric for assessing the geometrical similarity between the reconstructed trajectories and buoy trajectories (Gui et al., 2020), with a value approaching one denoting a high similarity between them. The cosine similarity (S_c) between the coordinate vectors of the reconstructed trajectory ($\overrightarrow{Q_{cal}}$) and buoy measurement ($\overrightarrow{Q_{buoy}}$) is calculated as follows:

$$S_c = \frac{\overrightarrow{Q_{cal}} \cdot \overrightarrow{Q_{buoy}}}{||\overrightarrow{Q_{cal}}|| ||\overrightarrow{Q_{buoy}}||}, \quad (4)$$

where $\overrightarrow{Q_{cal}} = (x_{cal}(i), y_{cal}(i))(i = 1, 2, 3, \dots)$ and $\overrightarrow{Q_{buoy}} = (x_{buoy}(i), y_{buoy}(i))(i = 1, 2, 3, \dots)$.

To characterize regional differences between BG and TPD regions, we defined the starting points using the geometric centers of grid points with a probability of reaching the BG or TPD region greater than 90% (BG: 80.04°N 157.71°W; TPD: 83.31°N 155.94°E) and that having ambiguous destination with a probability of reaching both regions between 40% and 60% (Both: 81.10°N, 177.89°E), and reconstructed the ice trajectory originating from these regions over 9 months for each cycle from 1979 to 2022 (Fig. 3). The atmospheric thermodynamic forcing, including the freezing degree days (FDDs) and thawing degree days (TDDs), snowfall, and precipitation, closely related to the ice thermodynamic growth and melting processes (Ricker et al., 2017a; Bigdeli et al., 2020), as well as the surface net longwave radiative flux, related to the feedbacks of clouds and sea ice itself on the near-surface atmosphere (Graham et al., 2017), were estimated along the reconstructed ice trajectories. FDD (TDD) refers to the integral of daily average near-surface air temperatures below -1.8°C (above 0°C) over the study period, derived from ERA5 hourly air temperatures, with a negligible bias (7.18×10^{-7} K) compared to the daily air temperatures obtained directly from ERA5. The dynamic response of sea ice to atmospheric forcing is characterized using the ice-wind speed ratio (e.g., Herman and Glowacki, 2012).

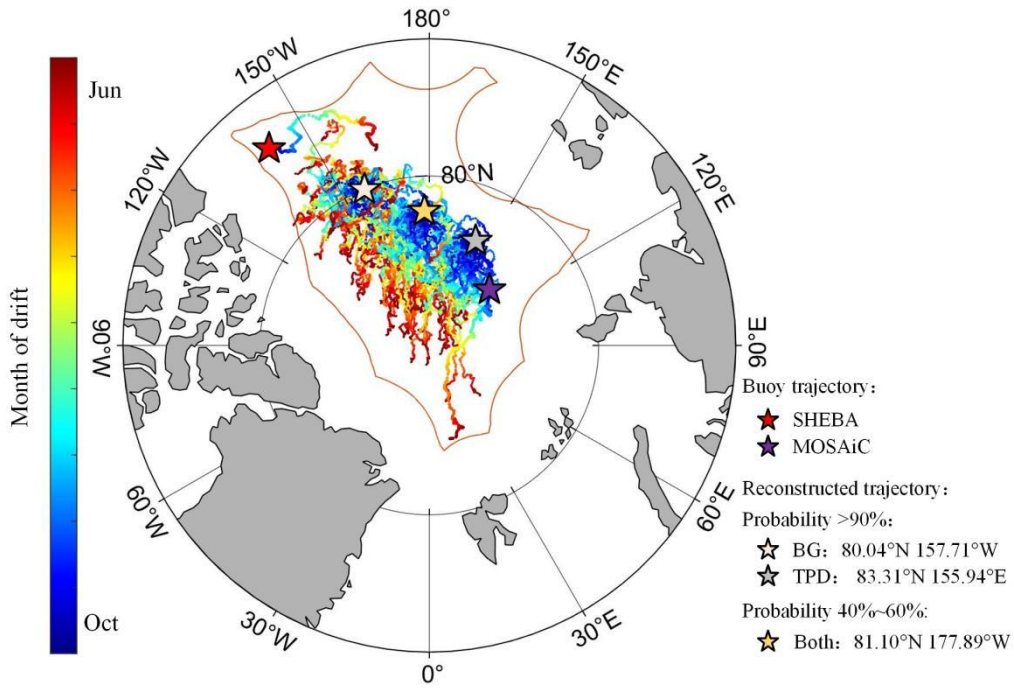


Figure 3. Reconstructed 9-month sea ice trajectories in 1979–2022, starting from three geometric centers of grid points with a probability of reaching the BG or TPD region greater than 90% and that having ambiguous destination with a probability of reaching both regions between 40% and 60%. For comparison, the partial drifting trajectories of SHEBA and MOSAiC ice camps started from October 3 of 1997 and 2019 to the time of 9 months after deployment are also shown.

3. Results

3.1 Spatial distribution of the effective starting points of reconstructed ice trajectories with 9-month ST

Using the reconstructed ice trajectories for each ice season from 1979 to 2022, the influence of the specific starting point on the ST and its destination is assessed here. The results reveal that the effective probabilities (the ratio between the effective years to all study years from 1979 to 2022) of starting points with the reconstructed ice trajectories having sufficient ST of no less than 9 months ranged from 9.1% to 90.9%. Drifting from the region close to north of the East Siberian and Laptev seas (centered at about 82°N and 160°E), the probability over the 44 years is relatively high than other regions, generally exceeding 75%. The likelihood of sea ice drifting into the EEZs or beyond the ice zone increased when the starting point approached the corners of rectangular study region, particularly in the downstream region of the TPD, where the probability is notably less than 20.0%. The eligible points shown in Fig. 4b indicated that the locations (with a size of $7.0 \times 10^5 \text{ km}^2$) as the starting point of reconstructed trajectories with the $ST \geq 9$ months within 33 years (or 75%) or beyond from 1979 to 2022. Among them, the blue dots represent moderate-recommendation zones in the central Arctic Ocean for ice camp or buoy deployment, where Lagrangian observations have a 75–85% probability of lasting sufficiently long, with these zones occupying 80.4% of the eligible area. The red dots, on the other hand, having a probability of $> 85\%$, are

249 high-recommendation zones, occupying 19.6% of the eligible area.

250 The probabilities of termination of reconstructed ice trajectory reaching the BG or TPD region during the study period
251 are illustrated in Fig. 4c–d. From 1979 to 2022, as expected, the ice floes that tend to drift to the BG region (Fig. 4c) are
252 mainly originating from the southwest part of the defined rectangular area; while the ice floes that tend to drift to the TPD
253 region (Fig. 4d) are mainly originating from the northeast part of the rectangular area. However, there is also a large
254 overlapping area between these two regions, and the magnitude of the probabilities exhibits a regular regional variability
255 pattern for the specific regions. This suggests that the location of the starting point has a crucial influence on the subsequent
256 ice advection. Thus, the deployment areas of the ice camp or buoy would determine their drift trajectory and final destination
257 to a high degree. The number of eligible starting points, whose reconstructed trajectory reached the TPD region with ST of
258 no less than 9 months, accumulated over 75% years of the 44-year study period, was 2.2 times that of such starting points
259 that reached the BG region. This indicates that sea ice originating from eligible starting points is more likely to reach the
260 TPD region. For the ice floes originating from the junction zone between the BG and TPD regions (yellow strip in Fig. 4b),
261 defined using the climatological SIM field, the probability of reconstructed ice trajectories reaching these two regions ranges
262 from 42.5% to 52.5%, without obvious regional tendency for ice advection destination.

263 Noting the symbolic shift in the physical nature of Arctic sea ice after 2007 (Sumata et al., 2023), we further calculated
264 the probability distribution of the starting point with the termination of the reconstructed ice trajectory reaching the BG or
265 TPD region for the sub periods prior to and after 2007, as shown in Fig. 5. The probabilities of starting points having the
266 sufficient ST of no less than 9 months ranged between 14.3% and 92.9% in 1979–2006, which changed to 7.3%–87.5% since
267 then, indicating a greater variability after 2007 (Fig. 5a–b). The size of ideal deployment area (red and blue dots), with a
268 probability > 75% in Fig. 5c–d, was reduced obviously in 2007–2022 ($2.1 \times 10^5 \text{ km}^2$) by 81.3% compared to that derived
269 from 1979–2006 ($11.2 \times 10^5 \text{ km}^2$). In particular, the area of high-recommendation zones (red dots) has shrunk from 6.9×10^5
270 km^2 in 1979–2006 to an almost negligible area of $1.1 \times 10^4 \text{ km}^2$ in 2007–2022. Such a conspicuous reduce in the moderate-
271 and high- recommendation zones suggests that the deployments of ice camp or buoy, with an intention to obtain sufficiently
272 long Lagrangian drift observations, in the central Arctic Ocean become more challenging as sea ice decreases.

273 The spatial distributions of the probabilities of reaching the BG and TPD regions in two sub periods prior to or after
274 2007 are similar to those derived from the whole study period (Fig. 5e–h); and the spatial proportions of starting points
275 with > 75% probability of reaching the two regions varied slightly for two sub periods, with changes ranging from 0.7% to
276 4.8% relative to the full period. This suggests the destination of ice floe advection is relatively stable, which is mainly
277 associated with the Arctic sea ice circulation patterns (detailed analysis will be provided in Section 4.2).

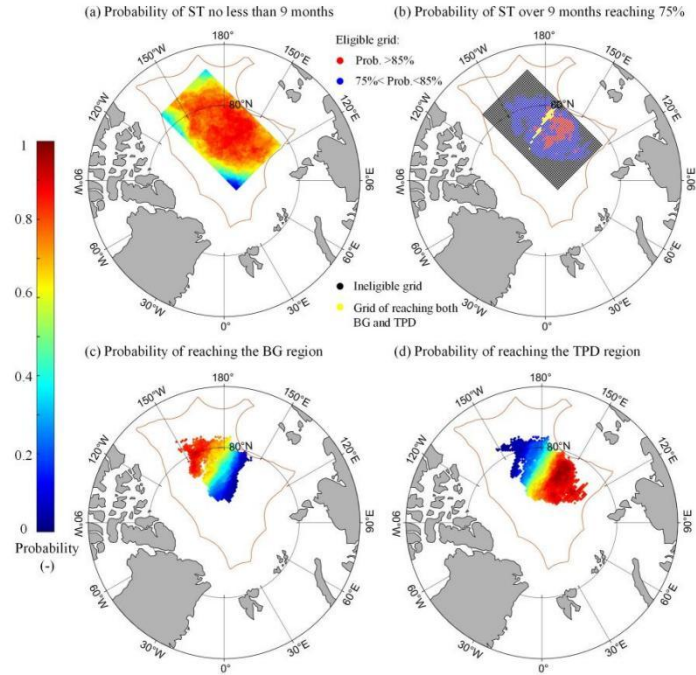


Figure 4. Spatial distribution of probability that sea ice drifting from a defined grid point satisfies the following conditions in 1979–2022: (a) the ST within the central Arctic Ocean for no less than 9 months; (b) the region with the probability of ST over 9 months reaching 75% (red and blue dot), also shown is the junction zone between the BG and TPD regions (yellow strip), defined using the climatological SIM field; and the probabilities with the destinations of reconstructed trajectories reaching the (c) BG or (d) TPD region.

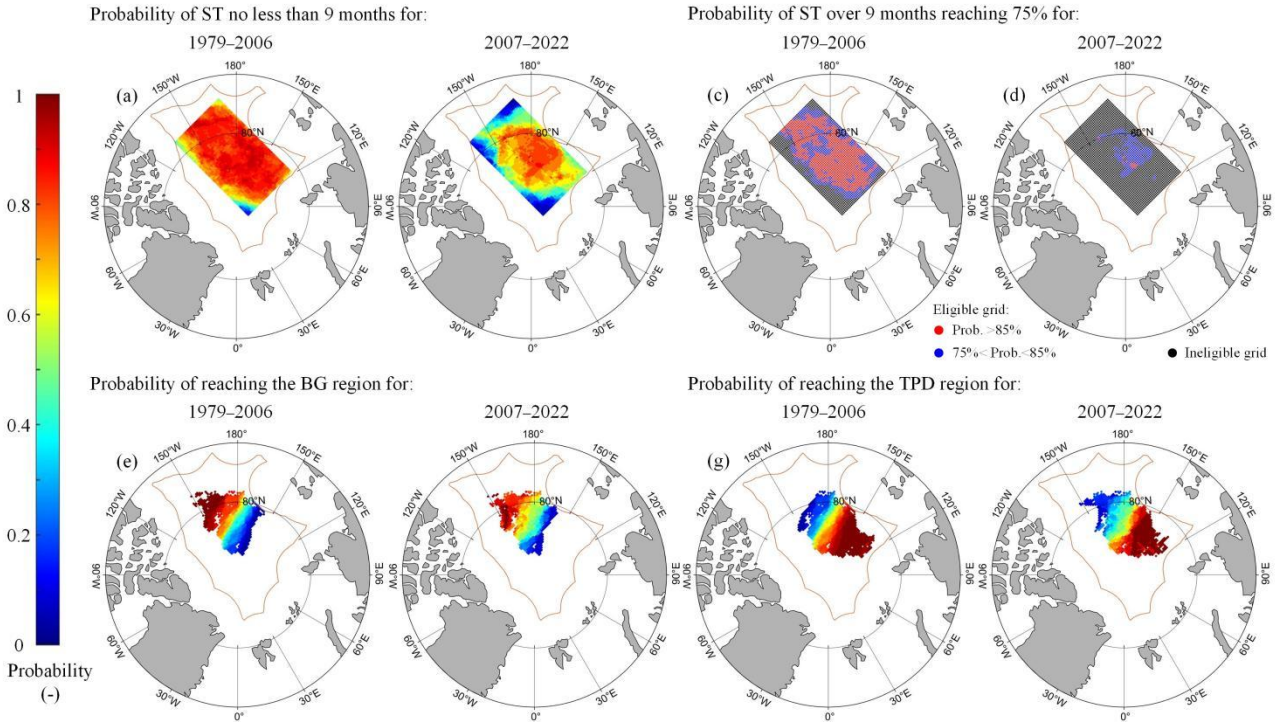


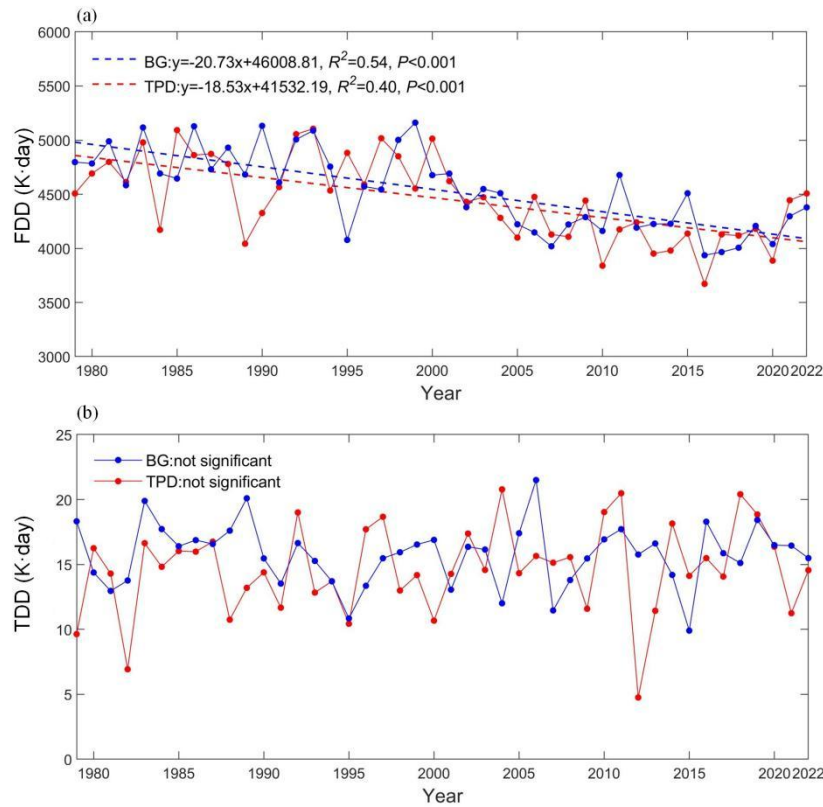
Figure 5. Spatial distribution of probability that sea ice drifting from a defined grid point in two sub periods of 1979–2006 and 2007–2022 under the following conditions: (a–b) probability of ST of no less than 9 months; (c–d) the region with the probability of ST over 9 months reaching 75% (red and blue dot); and probability of reaching the (e–f) BG or (g–h) TPD region.

3.2 Changes in atmospheric thermodynamic and dynamic forcing along the trajectories

Sea ice thermodynamic growth is regulated by both atmospheric and oceanic forcing. Since the oceanic heat flux underneath the ice is relatively weak during the freezing season (Lei et al., 2022), near-surface air temperature could be considered as the most decisive parameter regulating ice growth and is a major atmospheric forcing factor for the sea ice growth analysis model (Leppäranta, 1993). Thus, the Arctic Amplification can reduce ice growth during the freezing season (Ricker et al., 2017a) and trigger an earlier onset of sea ice melting (Stroeve et al., 2014). In 1979–2022, the average air temperature along ice trajectories was higher in the TPD region (-17.6°C) than in the BG region (-18.2°C), with a slightly higher increasing trend (0.096°C/yr) than that in the BG region (0.082°C/yr). This is consistent with the results given by Rantanen et al., (2022), which revealed a relatively high warming trend in the Atlantic sector compared to other regions in the Arctic Ocean, mainly due to the enhanced atmosphere-ocean heat flux caused by the reduction of the ice cover, increased warm air mass intrusions, and changes in atmospheric circulations. Despite significant increases in air temperatures in both regions, the occurrence of extremely high air temperatures exceeding 90th percentile of daily mean from 1979 to 2022, also defined as hot days by Vautard et al., (2013), did not change significantly. These hot days occurred mainly in June, with negligible regional variation in frequency between the BG (7.3%–12.4%) and TPD (6.6%–13.2%) regions. This implies that extremely events with high near-surface air temperature are largely concentrated in the initial stage of ice melting (Markus et al., 2009). These events are often accompanied by the process of rainfall (e.g., Robinson et al., 2021). The average precipitation along the trajectories reaching the BG and TPD regions increased significantly by 0.13 and 0.10 mm, respectively ($P < 0.05$) from 1979 to 2022. Despite the significant increase in precipitation in both regions, a significant increasing trend in accumulated snowfall was only observed in the BG region ($0.50\text{ mm water equivalent/yr}$, $P < 0.05$). In contrast, the TPD region is more susceptible to a shift in precipitation form snowfall to rainfall due to warmer conditions (Cohen et al., 2020), resulting in an insignificant change in snowfall. Especially in late spring and summer, the increased precipitation in the form of rain would accelerate the melting of the sea ice surface, promote melt ponds formation (e.g., Feng et al., 2021), and trigger positive albedo feedbacks (e.g., Goosse et al., 2018).

To further investigate the potential impact of near-surface temperature on sea ice melting or freezing, we also calculated FDD and TDD along the trajectories in 1979–2022. As shown in Fig. 6a, the FDD of the ice trajectories reaching the BG region was generally higher compared to those reaching the TPD from 1979 to 2022, which suggests warmer conditions during the freezing season in the TPD region, although the ice trajectories in this region were located in a relatively high-latitude area. Furthermore, the FDD along the trajectories reaching the BG region showed a significant decreasing trend ($P < 0.05$), which was slightly larger relative to that in the TPD region. However, the magnitude of TDD along the trajectories

316 reaching both regions of BG and TPD did not differ considerably and did not reveal a clear trend due to the insignificant
 317 warming trend for the summer in the Arctic Ocean. In winter, the average surface net longwave radiative flux along the
 318 trajectory in the TPD region was upward, indicating the heat loss from the sea ice-ocean system to the low atmosphere, with
 319 the peak of probability distribution in its absolute value decreasing from about 57.5 W/m² prior to 2007 to about 52.5 W/m²
 320 after 2007. However, such shift in the BG region was relatively weak from about 52.5 W/m² to about 50.0 W/m² (Fig. 7).
 321 This indicates that the weakened radiational cooling effect from the surface in the TPD region under the clear-sky conditions
 322 was more pronounced compared to that in the BG region, which also can be attributed the difference in the winter warming
 323 trend between two regions. Moreover, the frequency with the net longwave radiation feature under the opaquely cloudy state
 324 during the winter, having the typical value of > -10 W/m² (Graham et al., 2017), increased from 3.6% in 1979–2006 to 4.3%
 325 in 2007–2022 in the TPD region, while it decreased from 4.3% to 4.2% in the BG region. This can, at least in part, be
 326 attributed to the more prominent trend of enhanced cyclonic activity in the Atlantic sector of Arctic Ocean than the western
 327 Arctic Ocean (e.g., Zhang et al., 2023).



328
 329 **Figure 6.** Changes in (a) FDD and (b) TDD during October to June in two regions from 1979 to 2022.

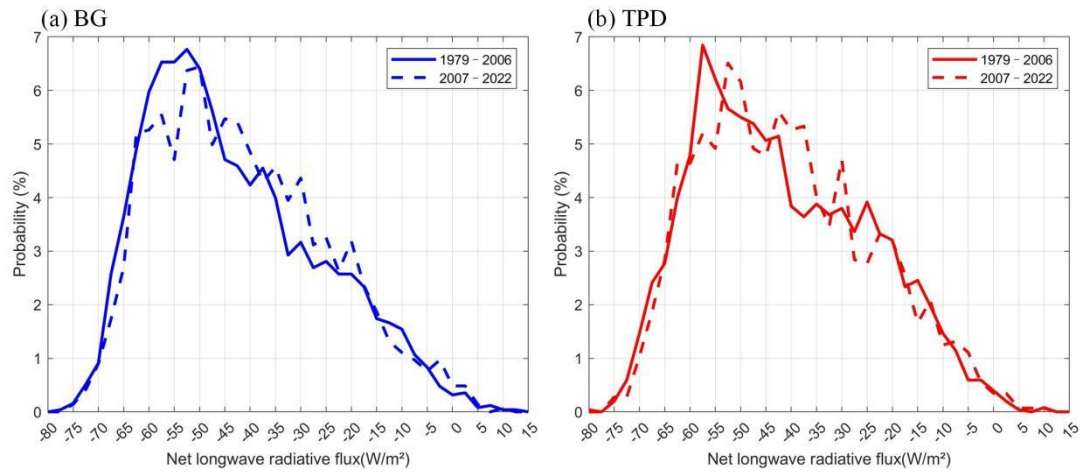


Figure 7. Probability distribution of winter surface net longwave radiative flux along the trajectories in the BG and TPD regions for two sub periods of 1979–2006 and 2007–2022, with negative values denoting the upward flux from the surface.

The dynamic response of sea ice to wind forcing can be characterized using the ice-wind speed ratio (Herman and Glowacki, 2012). The seasonal average ice-wind speed ratio along the ice trajectories was largest in autumn, which may be due to the relatively weak sea ice consolidation at that time (e.g., Lund-Hansen et al., 2020). In this season, the ratio was slightly larger in the TPD region (1.55%) than in the BG region (1.52%), which is consistent with the TPD region being generally considered as a region with a higher ice-wind speed ratio than other regions in the Arctic Ocean (Haller et al., 2014). Note that, the ice-wind speed ratio obtained from this study was slightly lower than those obtained from buoy observations close to the North Pole by Haller et al., (2014), as the remote sensing SIM product typically underestimates SIM speeds due to the low temporal resolution (Gui et al., 2020). This ratio is also overall lower than the typical values in free-drift analytical solution (Leppäranta, 2011). This difference is likely associated with SIC. As SIC increases, the internal stress among ice floes increases, leading to enhanced resistance against SIM, thereby reducing the response of sea ice to the wind. Since 2007, as the acceleration of SIM has been more apparent (Sumata et al., 2023), the seasonal average ice-wind speed ratio in the BG region increased to $1.6 \pm 0.3\%$ for autumn, winter and spring seasons, which results in its values already overwhelmed those of the TPD region by about 10.0%. This indicates that Lagrangian observations in the BG region would experience an ongoing enhancement of dynamic response of sea ice to wind forcing compared to the TPD region. It is related to the fact that the thick multi-year ice with weak mobility has been gradually replaced by the thin seasonal ice in the BG region, especially in its southern part (Babb et al., 2023). Therefore, from the perspective of sea ice dynamics response, the observation data of the SHEBA campaign (e.g., Lindsay, 2002) cannot be considered representative for the current ice state in the BG region.

4. Discussions

4.1 Assessment of reconstructed sea ice drift trajectories

To validate the reliability of the ice trajectory reconstruction method using remote sensing SIM product, we used data from 15 buoys in each of the BG and TPD regions (Table A1). Within the initial 100 days, most misalignment distances were less than 50 km (Fig. 8a). However, the reconstructed trajectories from the deployment sites of 5 buoys were misaligned with the buoys trajectories by a relatively large distance (146.3–173.0 km after 100 days), likely due to the complex wind conditions at the early drifting stage of these buoys. These buoys experienced relatively high wind speeds (6.8 ± 2.7 m/s) in the first 10 days, with over 60.0% of days exceeding 6.0 m/s, which is higher than the climatological average since 1979 (5.6 ± 0.4 m/s). Conversely, the distribution of initial 10-day wind speeds in our reconstructed trajectories in the BG and TPD regions over the 44-year period is mainly concentrated in the range up to 6.0 m/s ($>58.0\%$), with modal wind speeds ranging from 4 to 5 m/s (Fig. A1). This indicates that most reconstructed trajectories in this study did not experience such strong wind conditions that affect the accuracy.

Excluding the 5 buoys with complex initial wind conditions, the average misalignment distances between the reconstructed trajectories and the buoy trajectories (10 cases) after 9 months are 59.0 ± 41.4 km (about 2.5 pixels of ice motion product), with cosine similarities all above 0.92, and the average endpoint Euclidean distances is 114.4 ± 87.5 km. This high geometric similarity ensures appropriate advection direction and destination of the reconstructed trajectories. Regionally, there is a better reconstruction performance in the BG region than that in the TPD region. After 9 months, the average misalignment distance in the BG region (3 cases) is about 23.0 km, which is 30.9% of that in the TPD region (7 cases), consistent with the visual comparison of drift trajectories shown in Fig. 8b. This may be due to the larger SIM speed and its meridional gradient in the TPD region, especially in the southern region. All reconstructed trajectories in the TPD region terminate at a further northward location compared to the corresponding buoy trajectories. This leads to a relatively adventurous estimate of the effective duration of ice trajectories in the TPD region. However, we estimate this uncertainty to be approximately 1.6% (or 4.3 days) compared with the buoy measurement, which looks trivial.

In addition, the semi-Lagrangian method also has been proposed by combining the Lagrangian and Eulerian aspects to reconstruct ice trajectories through a time-stepping procedure (Sagawa and Yamaguchi, 2006). Therefore, we also compared the reconstructed trajectories obtained using the Lagrangian or semi-Lagrangian methods. For two starting points with $>90\%$ probability of reaching the BG and TPD regions, the reconstructed trajectories of the two methods showed high similarity, with average cosine similarity reaches above 0.99 and average misalignment distances of 3.5 ± 1.6 km in the BG region and 2.5 ± 1.2 km in the TPD region, respectively. However, relative to the validation data measured by 30 buoys, the

semi-Lagrangian method exhibited larger errors compared to the Lagrangian method used in this study. The endpoint distances between the trajectories reconstructed using the semi-Lagrangian method and the buoy trajectories after 9 months (10 cases) were 118.6 ± 88.2 km, with the misalignment distances of 61.1 ± 40.9 km and the cosine similarities of above 0.92. These statistical matrices are larger than those obtained by the Lagrangian method by 3.67% and 3.56%, with comparable cosine similarities, respectively. It is likely is because the semi-Lagrangian method does not directly track individual particles as precisely as the Lagrangian method (Subich et al., 2020), affecting its accuracy in reconstructing ice trajectories.

To further assess the reliability, we also conducted a closed-loop examination by reconstructing backward ice drift trajectories from endpoints of the original forward-reconstructed trajectories. This verification test required reconstruction of backward trajectories to return to their starting points after the same time period. Specifically, we initiated backward reconstructions from hotspot regions (defined by a density-based clustering method with 75% probability of ST >9 months during 1979–2022) at the endpoints of forward trajectories. Consequently, 66.3% of backward trajectories are projected to return to recommended regions, with 49.2% and 17.1% reaching the moderate- and high- recommendation zones, respectively (Fig. 9). Thus, this closed calibration testing also demonstrates the high confidence in the reconstruction method of ice trajectory.

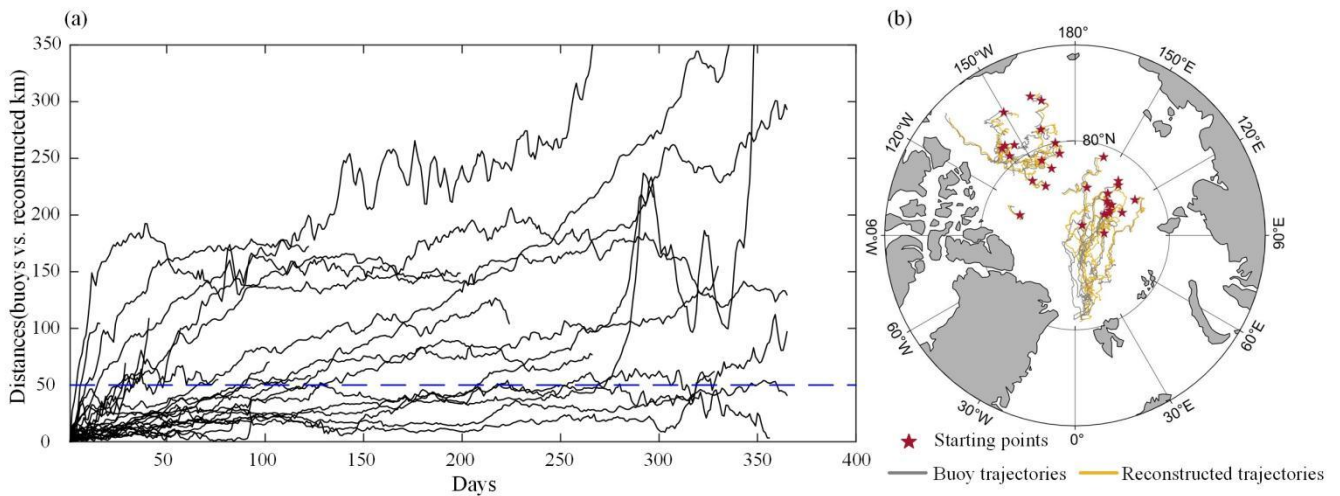


Figure 8. Comparison of reconstructed ice trajectories with buoy trajectories: (a) misalignment distances over time for trajectory pairs; (b) trajectory pairs from the deployment site of buoys.

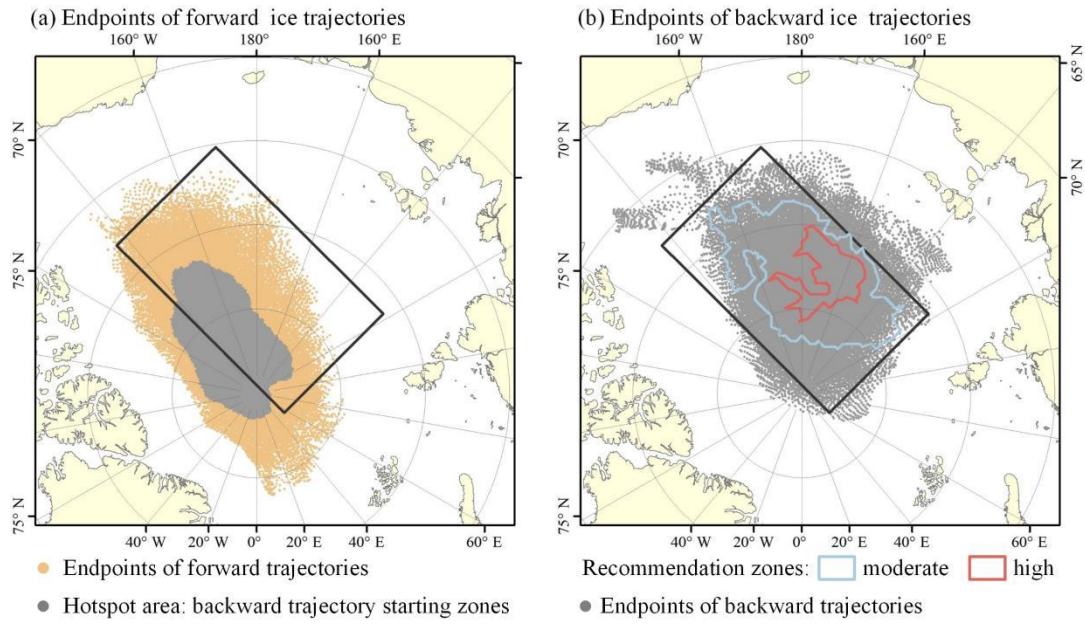


Figure 9. Spatial distribution of endpoints of forward- and backward- reconstructed trajectories: (a) forward trajectory endpoints and hotspot regions; (b) endpoints of backward-reconstructed trajectories from hotspot regions.

4.2 Links of endpoints of ice trajectories to atmospheric circulation patterns

Based on three scenarios for ice trajectory endpoints, i.e., > 90% probability of reaching the BG or TPD region and 40–60% probability of reaching both regions, we explored the links of endpoints of ice trajectories to atmospheric circulation patterns (Table 1). We also analyzed the statistical relationship (Table 2) between the distance from the endpoint to Fram Strait (80°N) for the ice trajectories reaching the TPD region and atmospheric circulation indices, which allows exploring the potential impact mechanisms of atmospheric circulation patterns on the sea ice outflow from the central Arctic Ocean.

For ice trajectories reaching the TPD region, autumn CAI, DA, and BH, can explain 22.5%–26.1% ($P < 0.05$) of the endpoint latitude after 9 months. The impact of CAI on the latitude of the ice trajectory was more pronounced because sea ice in the TPD region mainly advected meridionally. These indices were also significantly correlated with the distance between the endpoint of the ice trajectory after 9 months and the Fram Strait (R^2 : 21.8%–32.5%, $P < 0.05$). When the BH index is positive, the BG squeezes the axis alignment of the TPD eastward, lengthening the distance that ice advects along the TPD toward the Fram Strait. Compared to BH, the autumn CAI and DA were more strongly correlated with the latitude of endpoint or the distance between the endpoint and Fram Strait after 9 months. As their positive phases imply enhanced meridional wind forcing in the TPD region, which exacerbates the transpolar sea ice drift (Wu et al., 2006; Vihma et al., 2012). Therefore, it is necessary to take the autumn CAI and DA index into consideration for predicting the subsequent trajectory of ice floe, as well as the buoy or camp deployed on it in the TPD region.

For the ice trajectory reaching the BG region, most atmospheric circulation indices in autumn had a significant impact on the endpoints after 9 months, with the longitude of endpoint being significantly influenced by AO, CAI, and BH (R^2 : 14.9%–21.8%, $P<0.05$) and the latitude of endpoint being significantly correlated with CAI and DA (R^2 : 9.5%–19.1%, $P<0.05$). As sea ice motion in the BG region is mainly driven by the anticyclonic circulation, CAI mainly affects the longitude of the trajectory rather than the latitude. Although sea ice circulation in the BG region is driven by anticyclonic wind stress curl associated with the positive BH index (Proshutinsky et al., 2002), the BH did not reveal more effective interpretability for the location of the ice drift endpoints in this region than other indices. This may be because the influence of BH on the ice advection in this region is susceptible to interference from the impacts by other atmospheric circulation patterns, in particular, the DA and the associated meridional wind anomalies (e.g., Wang et al., 2009). Moreover, for sea ice that has the potential to reach both regions of BG and TPD, all atmospheric circulation indices in autumn also had a significant explanatory level for the latitude of endpoint and its distance to the Fram Strait after 9 months ($P<0.05$).

Since atmospheric circulation patterns during the start stage of ice drift in autumn, especially for the CAI and DA, had a strong influence on the endpoints of ice trajectories after 9 months in both regions of BG and TPD, we further analyzed scenarios where these indices exhibit extreme positive (negative) anomalies, defined with the value higher (lower) than the 1979–2022 climatology by one standard deviation (Fig. 10). When CAI and DA are at extreme positive (negative) phases, the spatial proportions of starting points with a 9-month ST threshold for more than 75% years are 75.2% (46.3%) and 86.0% (44.1%), respectively, which are greater (less) than the spatial proportions obtained from the mean field in 1979–2022 (48.6% as shown in Fig. 2). This suggests that the extreme scenarios of autumn CAI and DA have a pronounced modulating effect on the ideal deployment areas for Lagrangian observations, with a wider range of ideal areas at their extreme positive phases. Under extremely positive phases of CAI and DA, the preferred area of deployment tends to extend to the Chukchi Sea and the Canada Basin, while at the negative phase it prefers the northern Laptev Sea. However, the extremely positive phase of the autumn CAI only favors a trivial increase (by 2.2%) in the spatial proportion of points with > 75% probability of reaching the TPD region compared to the average state over 44 years. The extreme negative phase of the autumn DA, on the other hand, significantly increases the probability of reaching the BG region, and the spatial proportion > 75% is 1.5 times that obtained from the whole study period.

Table 1. Coefficient of determination (R^2) between atmospheric circulation indices and location (longitude/latitude) of sea ice trajectory endpoint after 9 months in 1979–2022.

Regional tendency		Autumn CAI	Autumn AO	Autumn DA	Autumn BH
Longitude of endpoint	TPD	n.s.	n.s.	n.s.	n.s.
	BG	0.218	0.181	n.s.	0.149

	TPD/BG	<i>0.215</i>	<i>0.158</i>	n.s.	0.145
	TPD	<i>0.225</i>	n.s.	0.261	<i>0.216</i>
Latitude of endpoint	BG	0.095	n.s.	<i>0.191</i>	n.s.
	TPD/BG	0.255	0.092	0.333	<i>0.210</i>

Note: Significance levels are $P < 0.001$ (bold), $P < 0.01$ (italic) and $P < 0.05$ (plain); n.s. denotes insignificant at the 0.05 level.

Table 2. Coefficient of determination (R^2) of autumn atmospheric circulation indices for the distance from the sea ice trajectory endpoint after 9 months to the Fram Strait in 1979–2022.

Regional tendency	Autumn CAI	Autumn AO	Autumn DA	Autumn BH
TPD	0.309	n.s.	0.325	<i>0.218</i>
TPD/BG	0.303	0.128	0.239	0.247

Note: Consistent with Table 1.

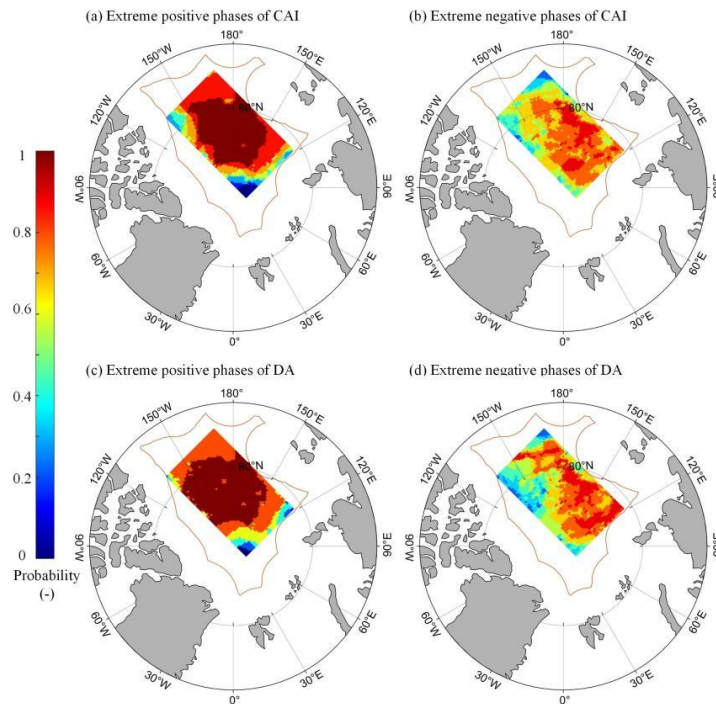


Figure 10. Spatial distribution of the probability that the ST of sea ice drifting from a defined grid point is no less than 9 months at the extreme positive and negative phases of the autumn CAI and DA for 1979–2022.

4.3 Impact of the EEZ boundary and deployment time on the ST of ice trajectory

The ST of reconstructed ice drift trajectories, or the potential Lagrangian observations on the basis of ice camp and buoy is limited to a high extent by the EEZ boundary. To quantitatively evaluate the impact in this regard, we hypothesized a desirability of enhanced international cooperation to reduce the impact of geopolitical boundaries on this type of observations, and identified the ideal deployment areas for Lagrangian observations under this scenario. It is found that the probability of ST exceeding 9 months for ice trajectories reconstructed from all grids in the rectangular study region ranged from 45.5% to 90.9% between 1979 and 2022 without the limitation of the EEZ boundary. The spatial range with

probabilities > 75% (i.e., 33 years) for the ST threshold of 9 months extends to 83.7% of the whole rectangular study region, much larger compared to that (48.6%) with the limitation of the EEZ boundary. Disregarding the EEZ boundary, the increase in eligible starting points with > 75% probability is proportional to the used ST threshold (Table 3). Especially for the 10-month ST threshold, the eligible area increases by over 150% compared to that with the limitation of the EEZ boundary. Disregarding the EEZ boundary, the increase in eligible starting points in the rectangular study region with > 75% probability of reaching the BG or TPD regions is also proportional to the ST threshold. Particularly for the 10-month ST, the number of eligible starting points reaching the BG or TPD region increases by over 100% through removing the limitation of EEZ boundary. For starting points with a close probability of reaching both regions of BG and TPD, the spatial proportion of eligible starting points would instead be suppressed compared to that estimated with the consideration of the EEZ boundary. This is because these eligible starting points are primarily located at the junction of two regions of BG and TPD and relatively far from the EEZ boundary.

For the period 1979–2022, the average Lagrangian observation duration in the rectangular study region disregarding the EEZ boundary is about 298.0 ± 25.7 days, which extends by about 34.3 days compared to those estimated with the consideration of the EEZ boundary. Regionally, the Lagrangian observations located in the BG and TPD regions would be further extended by about 20.5 days (286.1 ± 128.1 days) and 5.0 days (298.3 ± 114.4 days), respectively. This suggests that the EEZ boundary has a slightly larger impact on the observation duration in the BG region compared to the TPD region, because the EEZ boundary in the downstream of TPD is overall close to the marginal ice zone. Spatially, for sea ice reaching the BG region, the added eligible starting points are located in the southern part of the BG, as shown in Fig. 11. Sea ice originating from these locations might be more strongly affected by the clockwise ice circulation of the BG and cross beyond the EEZ boundary in the south more easily. For the ice trajectories reaching the TPD region, the added eligible starting points are located in the south of the study region or in the sector facing the Fram Strait. Sea ice originating from these areas might have been advected more rapidly to cross the EEZ boundary in the Atlantic sector.

Table 3. Increased spatial percentage in eligible starting points without considering the EEZ constraints compared to those estimated with the constraints.

ST threshold in the ice zone (months)	6	7	8	9	10
Case 1: probability of ST over corresponding ST > 75%	21.5%	32.0%	46.3%	72.0%	157.3%
Case 2: probability of reaching the BG region > 75%	10.8%	17.2%	21.5%	72.4%	111.1%
Case 3: probability of reaching the TPD region > 75%	3.1%	11.5%	21.2%	49.0%	144.0%
Case 4: probability of reaching the BG or TPD region ranging between 40% and 60%	−35.4%	−44.1%	−47.7%	−54.5%	−72.7%

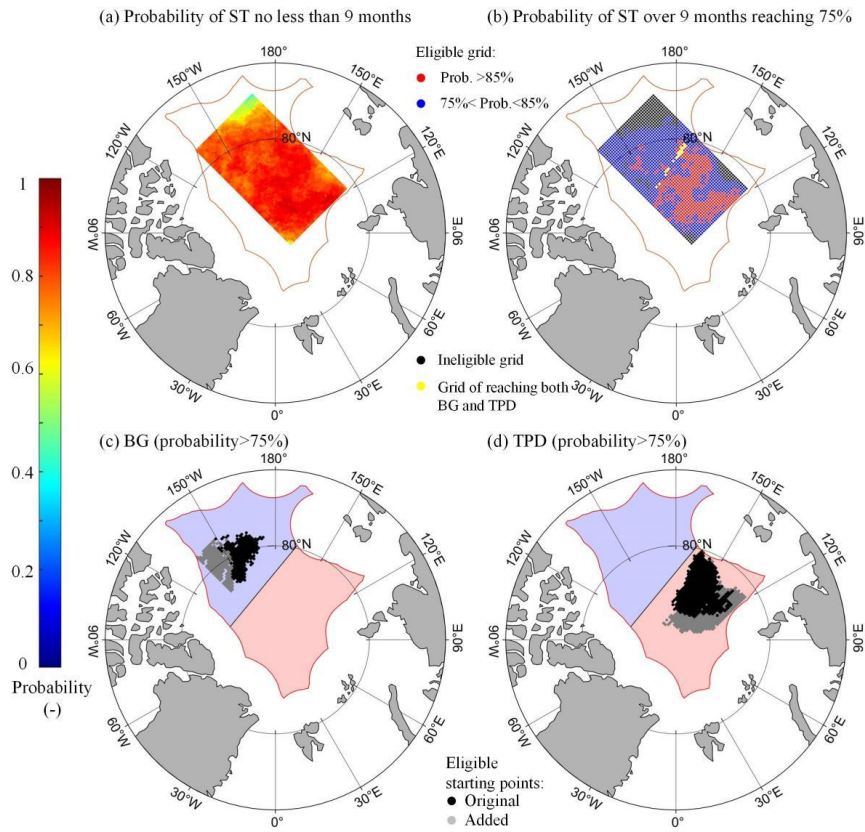


Figure 11. Assuming that the EEZ boundary constraints are not considered in 1979–2022: (a) spatial distribution of the probability of ST in the ice region no less than 9 months; (b) the region with the probability of ST over 9 months reaching 75% (red and blue dot), also shown is the junction zone between the BG and TPD regions (yellow strip); and (c–d) the added eligible starting point (gray) with > 75% probability of reaching the BG or TPD region, compared to those estimated with the EEZ constraints (black).

To further assess the influence of setup time of ice camp or buoy on the potential duration of subsequent observations, using the starting points reaching the BG or TPD region over 90% (Fig. 3), we calculated the duration with the deployment date ranging from August 15 to November 1 (Fig. 12). The mean duration of Lagrangian observation in the two regions decreases gradually from 296.1 ± 11.8 days for the deployment on August 15 to 279.9 ± 17.1 days according to the deployment on November 1. Although the advanced deployment of ice camp or buoy based on ice floes to August 15 may result in longer observation time, approximately by 10.0 days, compared that derived from the deployment on October 1. We still argue that it is more appropriate to implement the deployments of ice camp or buoy over the Arctic ice floes in early October if the logistics support allows, because there is often a risk that the ice holes drilled for the equipment deployment are hard to refreeze, and the risk of floe fragmentation would increase by the end of ice melt season in August and September. In these situations, the equipment is prone to collapse, causing observation interruptions. As expected, the duration in both regions is longer in the case of disregarding the EEZs relative to that derived with the EEZ restriction. Even in the BG region, with a shorter duration of observation compared to that in the TPD region, the potential duration for Lagrangian observation is estimated to reach 255.4 days with the EEZs and 283.3 days without the EEZs, respectively, with the deployment on

November 1. This suggests that the deployment of buoys or camps on the floes in the central Arctic Ocean, even by the end of October with the EEZ constraint, is still able to guarantee a observation duration of at least 8 months. Furthermore, the influence of EEZ on the potential Lagrangian observation duration would gradually strengthen with later deployment dates.

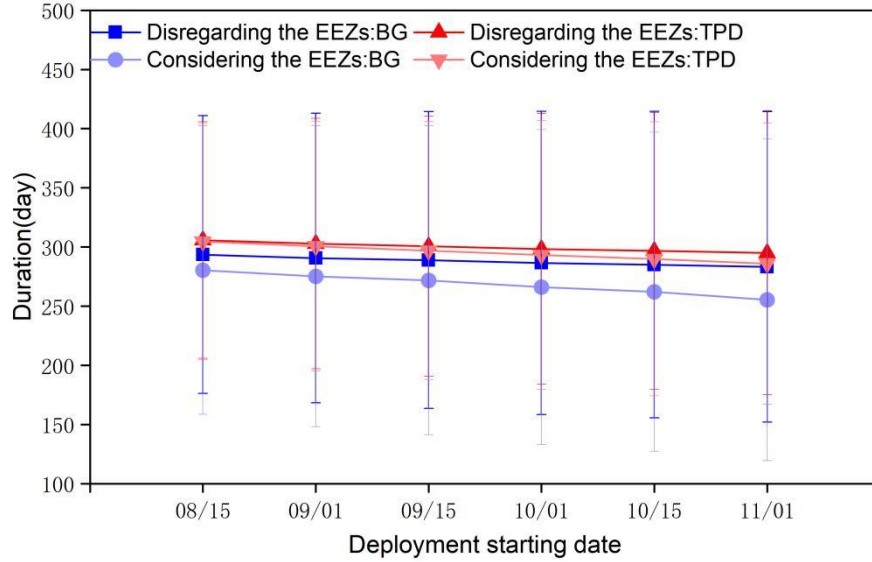


Figure 12. Changes in the mean duration of Lagrangian observations in 1979–2022 for various deployment dates from the starting points reaching the BG or TPD region over 90% , for both cases of taking into account or disregarding the EEZs.

5. Conclusions

From a rectangular study region defined in the central Arctic Ocean excluding the EEZs, we reconstructed the sea ice trajectories from 1979 to 2022 and determined the ideal deployment areas for the subsequent Lagrangian observations with a an expected duration. On this basis, regional differences in the atmospheric conditions along the trajectories were assessed. Subsequently, we explored the regulation mechanisms of atmospheric circulation patterns on sea ice advection and the influence of EEZ boundary constraints and deployment time on the duration of sustained Lagrangian observation.

Deployment of Lagrangian observations at locations centered around 82°N and 160°E, near north of the East Siberian and Laptev seas, can ensure at least 9 months of drifting observation time. The probabilities of remaining in the central Arctic ice region during the 44-year study period ranges from 75.0% to 90.9%, with mainly moderate-recommendation zones (75%–85%) with a spatial proportion of 80.4%, and small proportion of high-recommendation zones (>85%). Ice floes originating from this area of $7.0 \times 10^5 \text{ km}^2$ are more likely to reach the TPD region.

There are obvious regional differences in the atmospheric and sea ice conditions during ice drifting between the BG and TPD regions. Near-surface (2 m) air temperatures in both regions of BG and TPD show a significant warming trend in

1979–2022, with a higher increasing rate in the TPD region than in the BG region due to its proximity to the Atlantic sector of the Arctic Ocean. The significant decrease in FDD in the BG and TPD regions suggests that sea ice has experienced warmer conditions during the freezing season in recent years. Lagrangian observations in the TPD region would experience increased days of cloud opacity during the winter 1979–2022 by 19.4% compared to that in 1979–2006, because the cyclone activities are more frequent in the TPD region in recent years. The observations in the TPD region in early years would experience a relatively strong dynamic response of sea ice to wind forcing, with a higher ice-wind speed ratio than in the BG region. However, this response has been enhanced more prominently in the BG region due to the larger loss of multiyear ice, especially for the south part of BG region. Large-scale atmospheric circulation patterns at the early stage of ice drifting in autumn have a significant influence on the terminal location of ice trajectories. Thus, compared to the 1979–2022 average, the extreme positive phases of CAI and DA indices in autumn would expand the ideal deployment area to the Chukchi Sea and the Canada Basin. On the contrary, at the extreme negative phase of these indices, it is preferred to expand to the northern Laptev Sea.

In addition to natural conditions, the EEZ boundary has a great constraint on the Lagrangian observations. The absence of these constraints would increase the number of eligible starting points in the study region. Disregarding the EEZ boundary constraints, the eligible starting points with the trajectories toward the BG region expands southward, while for those toward the TPD region it would expand in the areas facing to the Fram Strait. The advanced deployment start time to mid August may result in a longer duration of Lagrangian observations, by 10.0 days compared to that obtained from the deployment on October 1. However, in order to reduce the failure risk of observation instruments deployed on the floes, in particular in the later ice melt season, we still consider the deployments in October are more appropriate for Lagrangian observation relying on ice floes in the central Arctic Ocean. Using the buoy data as a validation, the reconstructed trajectories have high geometrical similarity and accuracy, and the reconstruction in the BG region is better than that of the TPD region. Meanwhile, the accuracy of the Lagrangian method in ice trajectory reconstruction is higher than that of the semi-Lagrangian method, and the high-probability closure of the reconstructed forward and backward trajectories supports the high confidence of the reconstruction method.

In this study, daily SIM product is the main data source used to reconstruct sea ice drift trajectories and evaluate the ST of Lagrangian observations relying on ice floe. We acknowledge this as a primary evaluation, ignoring operational safety risks. The main challenges for survival and maintaining continuous observation for the specific devices deployed on the Arctic ice floes include the breakage or compression of sea ice, the formation of melt ponds, and the intrusion of polar bear, etc. As Arctic warming continues, the combined effects of accelerated melting and limited replenishment of multi-year ice will eventually trigger the complete loss of multiyear ice and a shift to a seasonally ice-free Arctic ocean (Babb et al., 2023).

549 This change puts forward greater demands on ice floe-based observational campaigns and on the development of more
550 adaptive observational techniques and equipment to cope with future extreme ice and atmospheric environments. Our work
551 mainly provides supporting cognition for the site selection for the deployments of ice buoy and ice camp. The preferred areas
552 identified in this study still require adaptable adjustments, associated with the changes in Arctic sea ice itself in the future.
553 From a practical perspective, once reaching the preferred deployment area, the specific conditions of the ice floe, such as ice
554 thickness, floe size, and distribution of ice ridge and melt pond, need to be further surveyed using high resolution satellite
555 remote sensing images and helicopters or ice-based measurements.

556 Appendix

557 **Table A1.** Basic information on buoy data used for validation of reconstructed ice drift trajectories

Number	Start date (YY/MM/DD)	Start location (°N, °E)	End Date (YY/MM/DD)	End location (°N, °E)	Duration (Day)	Buoys type
1	18/10/01	78.49, -146.12	19/08/24	71.29, -133.35	328	Snow_Buoy
2	20/11/04	83.93, -149.12	20/12/30	82.53, -144.07	57	iSVP
3	20/11/04	83.77, -110.26	20/12/30	82.81, -115.35	57	iSVP
4	20/11/04	82.50, -160.67	20/12/30	81.18, -154.30	57	iSVP
5	20/10/01	79.12, -140.50	20/12/26	76.66, -141.98	87	ITP
6	18/08/13	81.19, -169.34	19/02/27	80.88, -134.24	199	SIMBA
7	14/09/01	77.96, -141.98	15/05/24	75.67, -151.84	266	iSVP
8	14/09/01	81.32, -156.03	15/08/31	77.85, -138.64	365	iSVP
9	14/09/01	78.24, -162.07	15/08/31	79.50, -151.95	365	iSVP
10	16/09/01	82.67, -142.03	16/12/31	77.99, -132.51	122	iSVP
11	12/06/29	75.34, -165.97	12/08/08	77.74, -159.63	41	ITP
12	21/08/25	75.00, -149.90	21/09/09	75.67, -146.56	16	AXIB
13	14/09/12	79.99, -167.96	15/09/03	81.36, -160.42	356	-

14	22/07/14	74.61, -162.14	22/08/11	75.21, -162.29	29	IceCadet
15	11/08/05	78.01, -139.94	11/10/21	78.84, -134.63	76	-
16	15/10/01	85.06, 136.82	16/09/30	83.28, 8.21	366	PAWS
17	15/10/01	84.46, 115.64	16/09/12	81.13, 5.95	330	iSVP
18	15/10/01	85.06, 136.92	16/09/30	83.27, 8.20	366	Snow_Buoy
19	18/10/01	82.63, 141.50	19/08/26	82.42, 11.34	330	iSVP
20	18/10/01	81.17, 159.90	19/08/24	87.18, 13.64	328	Snow_Buoy
21	19/10/01	82.62, 120.56	20/09/29	83.30, 8.73	364	iSVP
22	19/10/01	86.18, 125.61	20/06/08	81.05, 3.78	252	iSVP
23	19/10/10	85.13, 133.02	20/07/14	81.04, -0.10	279	SIMBA
24	19/10/01	85.71, 123.25	20/07/14	81.06, -0.67	288	SVP5S 003
25	19/03/26	86.90, 94.19	19/12/08	81.11, 4.56	258	iSVP
26	10/04/19	88.69, 145.18	10/07/29	87.51, -0.59	102	AOFB
27	11/09/09	84.79, 165.96	11/12/17	85.29, -162.785	100	ITP
28	13/01/11	85.49, 125.51	13/08/22	85.47, 29.31	224	ITP
29	23/08/31	84.93, 129.60	23/12/31	86.01, 104.04	123	Thermistor String
30	12/12/01	82.98, 129.08	13/08/15	86.58, 20.20	255	SIMB

Note: buoys without specified types indicate that the IABP does not provide information on the buoy type.

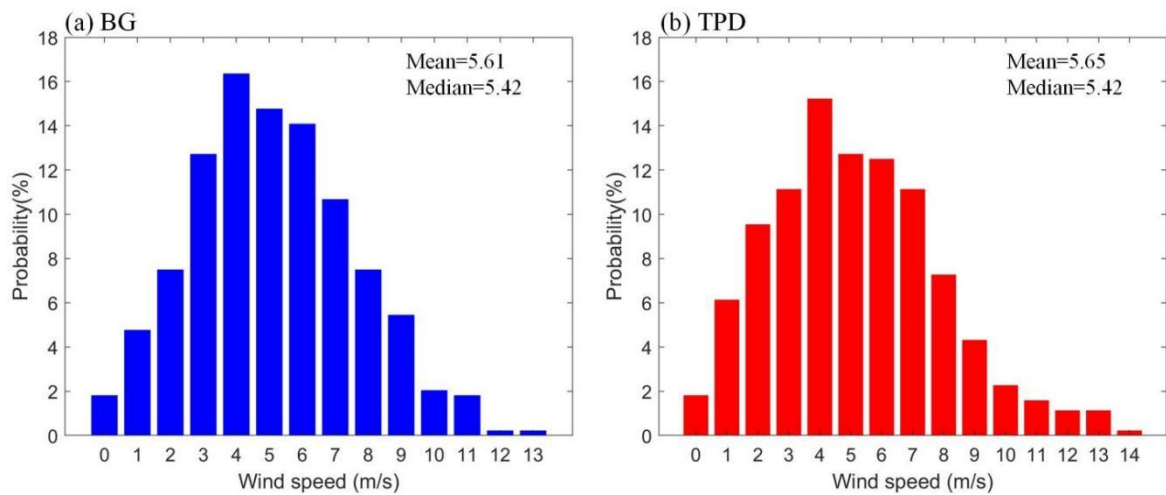


Figure A1 Probability distribution of wind speed in the initial 10 days along the reconstructed trajectories in the BG and TPD regions in 1979–2022.

Data Availability

Sea ice motion, concentration data from NSIDC is available at <https://nsidc.org/data/NSIDC-0116/versions/4> and <https://nsidc.org/data/G02202/versions/4>. Sea ice thickness data is downloaded from https://data.seaiceportal.de/data/cs2smos_awi/v204/. Shapefiles of maritime boundaries and EEZs are publicly available online (<https://www.marineregions.org/>). The ERA5 reanalysis data are downloaded from <https://cds.climate.copernicus.eu/cdsapp#!/dataset/reanalysis-era5-single-levels>. Buoy data is available at <https://www.meereisportal.de/> or <https://iabp.apl.uw.edu/TABLES/ArcticTable.html>.

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Competing interests

The contact author has declared that none of the authors has any competing interests.

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