

1 **Validation of OSCAR Surface Currents in the Western Arctic Marginal Seas against**
2 **Saildrone Observations**

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9 **Key Points:**

- 10 • This study validates OSCAR satellite derived surface currents in the western Arctic,
11 including the very shallow waters, by saildrones.
- 12 • High vector correlation but larger vector difference between OSCAR and saildrone
13 currents often occur in strong topography guided currents.
- 14 • Low vector correlation occurs at weaker currents, over the shallow Hanna Shoal, and
15 near fresher waters due to ice melt and river discharge.

16

17 **Key Words:**

18 **Surface currents, saildrone, OSCAR, western Arctic marginal seas, validation against in**
19 **situ observation**

20

21 **Abstract:**

22 The western Arctic marginal seas undergo large seasonal variation, but are very challenging to
23 observe directly due to sea ice and shallow depths. Deployments of several saildrone uncrewed
24 surface vehicles (USVs) in the summers of 2018 and 2019 provided unique opportunities to

25 validate the satellite-derived near surface currents, Ocean Surface Current Analysis Real-time
26 (OSCAR), in the western Arctic marginal seas against in situ upper ocean current measurements.
27 Overall, OSCAR current is biased low with significant noise. Higher vector correlation and
28 speed difference often occur where stronger currents (often topography-steered) are observed.
29 Such differences reveal that the dataset resolvability depends on spatial and temporal resolutions,
30 smoothing, and latitudes, suggesting that OSCAR is able to depict the major current systems but
31 significantly underestimates their strength. Poorer vector correlation occurs at weaker current
32 regimes (< 10 cm/s), over the shallow Hanna Shoal, near fresher water due to ice melt and river
33 discharge. The latter two water class regimes highlight the importance of salinity contribution to
34 the buoyancy force which is neglected in the OSCAR formulation.

35
36 **Plain Language Summary:**

37 It is challenging to make direct measurements in the western Arctic marginal seas, which
38 undergo large seasonal swings, because of the sea ice and shallow depths. We validate the
39 satellite-derived surface current product, Ocean Surface Current Analysis Real-time (OSCAR),
40 against observations from saildrone uncrewed surface vehicles (USVs) in the western Arctic
41 marginal seas cruising in summers of 2018 and 2019. Overall, OSCAR current is biased low and
42 noisy. Higher current direction relation and speed differences often occur where stronger
43 currents are observed. Such differences suggest that the reliability of OSCAR depends on several
44 factors. It also suggests that OSCAR is able to depict the major current systems but significantly
45 underestimates their strength. Poorer current direction relation occurs at weaker current regimes,
46 over the shallow Hanna Shoal, near fresher water due to ice melt and river discharge. The latter
47 two regimes highlight the importance of salinity information, which however, is neglected in the
48 OSCAR model.

49 **1. Introduction**

50 The near surface currents in the western Arctic marginal seas (i.e., the Bering, Chukchi,
51 and Beaufort Seas) play important roles in various geophysical phenomena, such as the transport
52 of heat, salt (e.g., Woodgate et al., 2018, Woodgate and Peralta-Ferriz, 2021), and sea ice
53 (Krumpfen et al., 2019, DeRepentigny et al., 2020). Monitoring the surface current variation is
54 therefore essential for studying the fate of the Pacific-origin waters into the Arctic basin and its
55 impact on the Arctic ecosystem (Stabeno, 2019).

56 Accurate knowledge of the upper surface currents systems in the western Arctic marginal
57 seas is important in estimating the transport and fluxes of various physical properties and
58 matters. The state-of-art global near surface ocean currents products (i.e., OSCAR) are not
59 directly measured by satellites but derived from the satellite measurements using simplified
60 formulation (Bonjean and Lagerloef, 2002). As the sea surface height measurements in the high
61 latitudes become available after 2010s from emergence of new satellite products and now being
62 used (i.e., Cryosat-2) in the latest OSCAR near surface current products, these large-scale near
63 surface current estimates are extended to polar oceans. On the other hand, satellite observations
64 show declining Arctic sea ice extent for all months (Serreze and Stroeve, 2015, Stabeno and Bell,
65 2019), shortened ice season (Wang et al., 2018) and related to the greater rising of surface
66 temperatures than the global mean surface temperature (Serreze and Francis, 2006, Richter-
67 Menge et al., 2019). Climate forecast models also suggest that in the Arctic the surface air
68 temperature will continue to rise much faster and the summertime sea-ice extent will continue to
69 decline (Alexander et al., 2018, Jefferies et al., 2013). Therefore, the spatial extent and the length
70 of direct air-sea interaction in the ice-free part of the Arctic, which could be measured from the
71 satellites, is increasing and may eventually become normal during summer in a warmer climate.

72 Validating the high latitude satellite-derived near surface products against in situ measurements
73 is therefore important in advancing the understanding of aforementioned geophysical phenomena
74 towards higher latitudes oceans, including the western Arctic marginal seas.

75 The western Arctic marginal seas feature many shallow shelf regions and mobile sea ice
76 which not only prevent Argo float measurements but also limit research ships from complete
77 surveys, therefore very challenging for in situ observations. Sailandrone uncrewed surface vehicles
78 (USVs) provide a unique opportunity to measure air-sea interaction over the very shallow waters
79 on the shelf and ice edge in western Arctic marginal seas. Sailandrones are wind and solar powered
80 vehicles that allow deployments lasting up to 12 months and provide high quality, near real-time,
81 multivariate upper ocean and atmospheric observations (Zhang et al. 2019). The primary goal of
82 this work is using the saildrone arrays to validate the satellite-derived ocean surface current
83 products, OSCAR, in the western Arctic marginal seas. Our analysis should be taken as a
84 preliminary step in the comparison of large scale gridded upper ocean current data to encourage
85 future research and application in this region.

86 In this study we focus on subsurface current velocities measured by current profilers
87 described in section 2.1. These saildrone provide valuable and rare in situ current observation in
88 the seasonal ice zone or gaps of ship observations. These observations are instrumental in
89 advancing our knowledge for further development and verification of satellite observation,
90 satellite-derived data products, and numerical models. Recent studies used two saildrones to
91 validate various satellite SST products and SMAP SSS products in the western Arctic (Vazquez-
92 Cuervo et al., 2021, Vazquez-Cuervo et al., 2022). As the satellite-based products evolve
93 constantly with emergence of new satellites and frequent changes of algorithms for improving
94 their absolute accuracy, it is vital to do periodic validation against in situ data. Among eight of

95 L4 satellite SST products, NOAA/NCEI DOISST and the RSS MWOI SST are shown to have
96 better relative accuracy against saildrone. Also, the SMAP SSS products are shown to resolve the
97 runoff signal associated with the Yukon River discharge with high correlation between SMAP
98 products and saildrone 0.5 m salinity. Both MWOI SST and SMAP SSS products from Remote
99 Sensing System (RSS) described in 2.3 are therefore chosen to be used in this study.

100 The manuscript is organized as follows. Section 2 describes the data and methods.
101 Section 3 presents the results. Section 4 summarizes this paper.

102 **2. Data and Methods**

103 **2.1 In-Situ Saildrone Data**

104 In situ data used in this study are from two saildrones deployed in July - September of
105 2018 (1020, 1021) and three saildrones (1035, 1036, 1037) deployed in May - September 2019
106 (Chiodi et al., 2021). They were launched from Dutch Harbor, Alaska crossed the Bering Strait
107 into the Chukchi Sea (and Beaufort Sea in 2019) and headed south for recovery as the sunlight
108 hours became short (Figure 1,2).

109 The surface current is measured by the downward looking 300 kHz Workhorse
110 WHM300-I-UG1 acoustic Doppler current profilers (ADCP) mounted on the keels of the five
111 saildrones. The vertical resolution of the ADCP data is 2 m. The temporal resolution of the 5-
112 minute average ADCP currents analyzed here are on 1020, 1021, 1035, and 10 minutes on 1036,
113 1037, on which ADCPs were turned on and off for 5 minutes every 10 minutes to save power.

114 The saildrone ADCP data is often easier to process than those of the ship-board ADCP.
115 The saildrone's transiting speed is slow – on average speed of 0.96 m/s, or ~18% of the average
116 wind speed at 5.4 m/s during the 2019 deployment (Chiodi et al., 2021). This quiet vehicle is less
117 affected by bubble issues (Joseph, 2014). The ADCP data is generally reliable ~5 m below the

118 sea surface. Depending on the echo intensity, ADCP data typically extend to 60-100 m depth,
119 ideal for survey of the entire water column on the shelf. The saildrone ADCP has onboard
120 motion correction for preliminary quality control before the data are sent to the data center in
121 near real time. However, additional steps, including removing data below the ocean floor, or
122 where echo intensity is too small, or where the vertical velocity is too large, are necessary to
123 remove unrealistic current estimates and to achieve an accuracy of 2-3 cm/s.

124 The near surface temperature and salinity were measured at a nominal depth of 0.5 m by
125 2 saildrones in summer 2018 and 3 saildrones in summer 2019. The unpumped RBR sensors
126 were available in both years and the pumped SBE sensors were available in 2019. Here we use
127 RBR measurements in 2018 and SBE measurements in 2019. The RBR data are available every
128 10 minute on 1020, 1021 and the SBE data are available every 5 minute on 1035, 1036, 1037.
129 The temperature and salinity data analyzed are 1-minute averages of 1-Hz measurements.
130 Simple QC are applied to the temperature and salinity data by removing those with large 1-
131 minute standard deviation (0.1 degC for temperature and 0.05 psu for salinity).

132 **2.2 OSCAR Data**

133 OSCAR (Ocean Surface Current Analysis Real-time) is a global near-surface (nominal
134 depth at 15 m) ocean current product derived from sea surface height, ocean surface vector winds
135 and sea surface temperature observed by various satellites and in situ instruments. The model
136 formulation combines geostrophic, Ekman and thermal wind dynamics (e.g., the local
137 acceleration and non-linearities are not represented. Bonjean and Lagerloef, 2002). Its version
138 2.0, on $\frac{1}{4}$ degree with a 1-day resolution, is used in this study.

139 The OSCAR calibration and validation (<https://www.esr.org/research/oscar/validation/>)
140 using both 15 m drogued drifters and moorings suggest that its known problem areas fall in 3

141 main categories: eddy-dead regions such as the Alaskan Gyre, the meridional component around
142 the equator and near coasts. Smoothing in creation of the regular gridded source data and during
143 the calculation of spatial gradients would consistently underestimate the speeds compared to
144 those of drifters, generally by 50-60% globally. Parts of the western Arctic marginal seas could
145 fall in the last case where the model is not accurate very close to coastlines or ice, since source
146 satellite signals can be corrupted there.

147 **2.3 Satellite Sea Surface Temperature and Salinity Data**

148 Sea surface temperature (SST) and sea surface salinity (SSS) data are both from Remote
149 Sensing Systems (RSS) (Meissner et al., 2022). The SST product is the RSS Microwave (MW)
150 OI SST version 5.1 (daily mean). The SSS data is SMAP version 5.0 Level 3 8-day running
151 mean by RSS on 1-day nominal resolution. Both SST and SSS products are distributed in a 0.25°
152 rectangular projection. This latest SMAP RSS version uses a new sea ice flag and sea ice
153 correction including detection of large drifting icebergs.

154 **2.4 Satellite Ancillary Data – AVISO FES 2014 Tide Database**

155 AVISO FES 2014 tides database is used to remove the barotropic tidal current from the
156 in situ saildrone measured current. FES2014 was produced by Noveltis, Legos and CLS and
157 distributed by Aviso+, with support from Cnes (<https://www.aviso.altimetry.fr/>). It is a global
158 tide solution that uses finite element mesh, T-UGO barotropic model and data assimilation of
159 altimetry and tidal gauges.

160 Removing tidal currents are necessary in order to compare with OSCAR, the non-tidal
161 near surface currents products. Previous studies have documented the regional-dependent
162 characteristics of tides along the western Arctic marginal seas (i.e., Foreman et al., 2006, Huang
163 et al., 2011, Mofjeld et al., 1986). The northeastern Chukchi Sea appears to be non-tidal and is

164 often dominated by currents driven by synoptic weather patterns. Tidal currents are also weak in
165 the eastern Chukchi shelf (amplitude $< O(5 \text{ cm/s})$). In the Bering Sea tidal current amplitudes are
166 significant (on $O(40 \text{ cm/s})$) especially near coastlines of the Bristol Bay, Kuskokwim Bay and
167 Norton Sound.

168 **2.5 Collocation and Statistical Evaluation of OSCAR**

169 We use nearest-neighbor interpolation to first match each saildrone measurements (5-
170 minute or 10-minute) with the OSCAR gridded data in time and space. For each unique gridded
171 OSCAR data point, all saildrone data are averaged within that grid cell for a single match-up
172 saildrone data point. Figure 3 shows the time series of the collocated saildrone 10-20 m layer
173 averaged current vectors in 2019 around the OSCAR grids and their vector differences.

174 The evaluation of the differences between OSCAR and saildrone current speed are
175 quantified objectively by several statistical metrics including bias, root-mean-square error
176 (RMSE), standard deviation of the errors (SDE), and signal-to-noise ratio (SNR). The definitions
177 of the statistical metrics follow Vazquez-Cuervo et al. (2022). The vector correlation is the
178 cosine similarity as the cosine of the angle between the collocated OSCAR and saildrone current
179 vectors.

180 **3. Results**

181 Figure 1ab shows 3 saildrone tracks during summer 2019. They were deployed from
182 Dutch Harbor in mid-May and made their way off Alaska's west and northwestern coastline
183 through the Bering Sea and up through Bering Strait around June 5th 2019. Then the saildrone
184 stayed in the Chukchi Sea shelf until July 2019 and reached the farthest northern latitude ~ 75.5
185 N in the Beaufort Sea in August before returning to Dutch Harbor in early October. The near
186 surface waters are remarkably colder and fresher in the Beaufort Sea than those on the Bering

187 and Chukchi shelves (Figure 1ab, 2). The saildrones also measured the very warm and fresh
188 surface waters, associated with seasonal warming and freshening, west of Yukon-Kuskokwim
189 delta (Y-K delta) in early June 2019 (Vazquez-Cuervo et al., 2021, 2022). The surface waters on
190 the shelf are significantly warmed in 3 months of the saildrone deployment (Figure 1,2); the
191 Bering Sea shelf is freshened especially near the Alaskan coast.

192 The saildrones in summer 2018 went similar routes in the Bering Sea and Chukchi Sea
193 from July to the end of September but did not reach the Beaufort Sea. The fresh and warm
194 signals associated with Yukon River discharge are not as obvious as in July 2019. The difference
195 is consistent with the climatology of the seasonal variation west of the Y-K delta (Figure 9 of
196 Vazquez-Cuervo et al., 2022), with SSS dipping seasonally around mid-May to mid-June and
197 increasing afterwards, and SST increasing seasonally from April to July.

198 **3.1 Overall Comparison between Saildrone and OSCAR datasets**

199 Figure 4 summarizes the overall data distribution in direction bins of every 22.5° and
200 several specific subsets of the collocated saildrone 10-20 m layer averaged and OSCAR 15 m
201 current data, the current speed difference and vector correlation between the two current datasets.
202 The spokes represent the direction of which current vectors are towards. Colors along the spokes
203 indicate the parameters specified (i.e., speed, speed difference, or vector correlation). The length
204 of each spoke and its colored segment represents the percentage occurrence of the currents (as
205 numbered) flowing towards a particular direction at a given parameter range. Their overall speed
206 difference is notable (Figure 4abc), with OSCAR currents weaker than saildrones by 5.3 cm/s
207 (Table 1). Underestimation can be due to smoothing and spatial resolution of satellite data in
208 creating the regular-grid OSCAR products. 67% of the collocated velocity pairs are fairly
209 correlated (with direction difference within 67.5°), and is consistent between using 2018 current

210 data only (67%) and 2019 only (70%). The slope of the linear regression line indicates that
211 OSCAR in general underestimates the current speeds and their zonal or meridional velocities by
212 50-100% (Figure 5). The RMSD, which characterizes the variability in the difference of the two
213 datasets, is 11 cm/s. The SDE, in which the mean bias (5.3 cm/s) was removed, is 9.6 cm/s.
214 RMSD and SDE are both more profound than the mean bias, and on the same order of magnitude
215 as the observed SD, suggesting notable speed difference from observation. The signal-to-noise
216 ratio (SNR), which is inversely proportional to SDE, is 0.8. In fact, the SNR of all subsets is
217 smaller than 1, indicating that the noise of OSCAR current speed overwhelms the real signals.

218 **3.2 Comparisons between Saildrone and OSCAR datasets in different current speed** 219 **regimes**

220 The strong current regime (current speed > 30 cm/s) is observed mostly in bathymetry
221 guided flows, i.e., in the Bering Strait where currents flow northward in early June 2019,
222 upstream Barrow Canyon where currents flow eastward in July 2019, and on the Chukchi Slope
223 where Chukchi Slope Currents flow northwestward in the end of July 2019 (Figure 1,3). Their
224 vector correlation is well above average (64% of vector correlation is > 0.92, i.e., less than 22.5-
225 degree deviation from each other). The negative bias and RMSD of the OSCAR current speed
226 are of the same order of magnitude (O(20 cm/s) and significantly larger than average (Figure
227 4efg, Table 1). This phenomenon is consistent with the lower-than-1 linear regression slope and
228 small intercept (Figure 5a). The SDE is half of the RMSE and larger than the observed SD,
229 indicating the differences in current speed of the strong current regime is significant and
230 considerably contributed by underestimation.

231 For the weak current regime (current speed < 10 cm/s), the overall vector correlation is
232 lower than average, more than 2/3 is either orthogonal or negatively correlated. In contrast,

233 OSCAR current speed is positively biased by 0.8 cm/s, also indicated by the positive intercept of
234 y-axis in Figure 5a. The RMSD and SDE are twice the observed SD. It suggests that for the
235 weaker current regime the OSCAR current speed is significantly different from observation and
236 with slight overestimation.

237 **3.3 Comparisons between Saildrone and OSCAR datasets in different water classes or** 238 **areas**

239 The collocated OSCAR and saildrone currents show poorer vector correlation than
240 average at the cold fresh water lenses from ice melting and warm fresh Yukon River discharge
241 (Figure 4pt, Table 1). The RMSD and SDE of current speed for both water classes are slightly
242 larger than the observed SD, indicating significant current speed difference. The SNR is reduced
243 to half compared to the overall statistics, indicating the OSCAR current speeds in these two
244 surface water classes are excessively overwhelmed by noise. These lighter surface water lenses
245 from ice melting or river outflow that increase the near surface stratification are commonly
246 present in summertime. However, the buoyancy force is only a function of SST but not SSS in
247 OSCAR formulation (Bonjean and Lagerloef, 2002). Since salinity dominates the density
248 variation in the cold-water regime, it suggests that the salinity contribution to the buoyancy
249 gradient to the Ekman current velocities could be significant. Large stratification near the surface
250 could also induce large shear near the surface, which could potentially modulate the ocean
251 response to the winds, and their air-sea momentum transfer.

252 On the other hand, there are areas with very shallow bathymetry on the Chukchi Sea shelf
253 near the west Alaskan coast where the saildrones had chances to make measurements in 2018
254 and 2019. The vector correlation at the shallow bathymetry on the eastern Chukchi shelf is
255 similar to the overall statistics (Table 1). The vector correlation is poorer around Hanna Shoal

256 where the bathymetry is also shallow and current speed is slightly weaker. The RMSD and SDE
257 of both subsets are slightly larger than the observed SD, and with reduced SNR compared to the
258 overall statistics. Similar to the aforementioned two water classes, the OSCAR current speeds in
259 the shallow water are of significant difference from observation and excessively overwhelmed by
260 noise. Larger uncertainties near the coastlines, within 100 km, are known problems in the
261 OSCAR model (section 2.2).

262 **4. Summary**

263 This work presents rare upper ocean current measurements from saildrone USVs in the
264 Bering Sea shelf, Chukchi Sea shelf and slope regions, and Beaufort Sea, where shallow depths
265 prevent most research vessels from completing surveys. We use saildrone in situ data from the
266 summers of 2018 and 2019 to validate the OSCAR satellite derived current products. This work
267 quantifies the uncertainty of the OSCAR near surface currents and highlights the regimes to
268 expect larger negative biases of current speed or lower vector correlation. The overall current
269 speed difference for the collocated OSCAR and saildrones is significant (SDE larger than the
270 observed SD), with OSCAR under-estimating the observed speed. Particularly, the OSCAR
271 satellite derived currents data in the western Arctic is 50-100% lower than the observations. For
272 comparison, globally the derived currents data is 50-60% lower than in situ data observed from
273 15 m drogued drifters and moorings from the global network of OceanSITES. Such differences
274 reveal that the dataset resolvability depends on spatial and temporal resolution, smoothing, and
275 latitudes. In addition, the signal-to-ratio (SNR) of the OSCAR current speeds is lower than 1 in
276 western Arctic marginal seas, indicating an overwhelming noise level in the dataset.

277 Higher vector correlation occurs at the strong current regime where currents are guided
278 by the bathymetry. The negative speed bias of the strong current regime is $O(20 \text{ cm/s})$. This

279 implies that OSCAR is able to depict the major current systems in the western Arctic marginal
280 seas but significantly underestimates their strength. Low vector correlation is often observed for
281 weaker currents, for example, in the Hanna Shoal area, or in stratified surface waters from either
282 the warm fresh river discharge or the cold fresh surface melt waters. The SNR of OSCAR is also
283 significantly reduced in the Hanna Shoal area and the aforementioned lighter surface waters. The
284 poor statistical results for the two water classes highlight the importance of the salinity
285 information especially in cold water regimes for the vertical momentum diffusion, though
286 neglected in the OSCAR model.

287 This analysis serves as a preliminary step toward comparing and validating large scale
288 gridded upper ocean current products to encourage future research and application in the western
289 Arctic and other parts of the world's oceans using saildrone data. In particular, repeated saildrone
290 missions in some key areas in the western Arctic marginal seas would be beneficial in
291 monitoring and quantification of the Pacific water inflow rates and routes. On the other hand,
292 saildrones can be used to validate and improve the satellite-derived surface current data in high
293 latitude oceans so that not only the Pacific water surface routes but in the other parts of high
294 latitude surface water routes and properties can be better estimated under the warming climate.

295

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297 This is PMEL Contribution #5406.

298

299 **Open Research:**

300 The NASA saildrone data is accessed from <https://podaac->

301 tools.jpl.nasa.gov/drive/files/allData/insitu and the NOAA saildrone data is from

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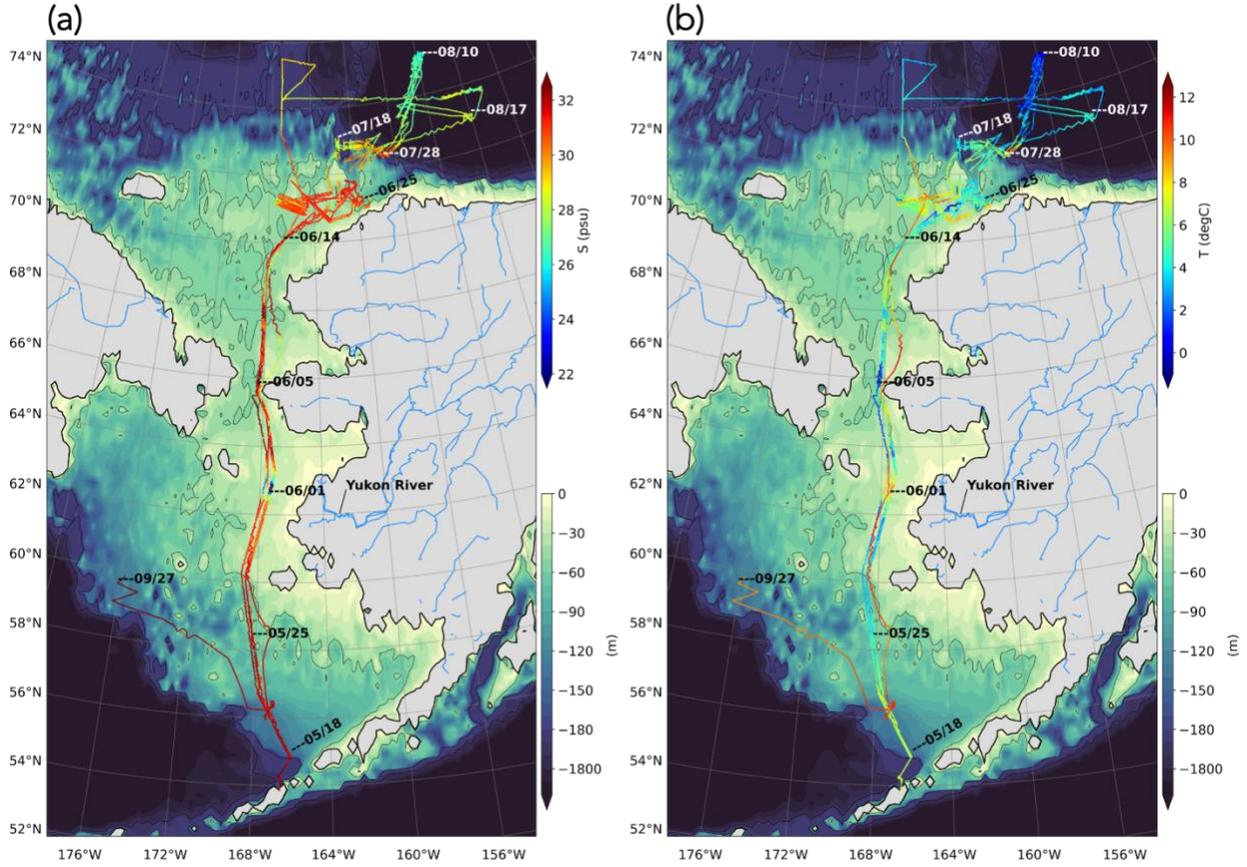
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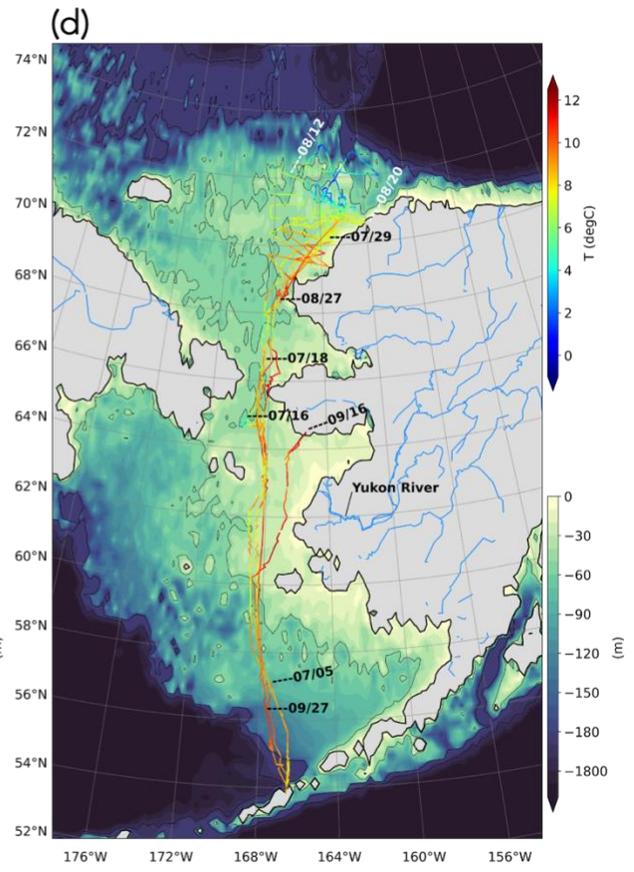
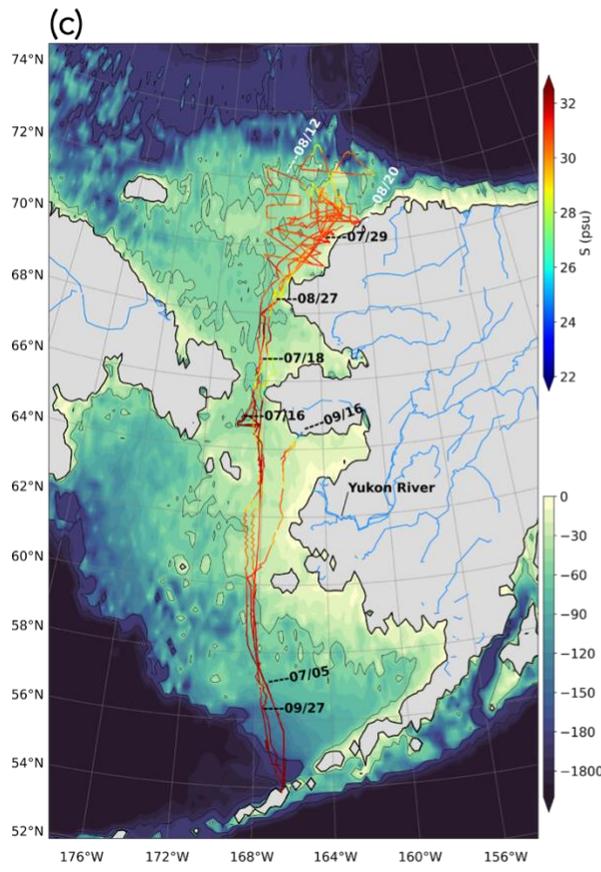
411 **Table 1:** Summary of statistical differences between the collocated OSCAR and saildrone
412 current speed and vector correlation under several specific subsets described in the leftmost
413 column with the number of pairs in the parenthesis. Obs. SD is the observed current speed by
414 saildrones. RMSD is root-mean-square-difference. Bias is the mean residual difference, OSCAR-
415 SD. SDE is the standard deviation of errors in which the mean bias error is removed. SNR is the
416 signal-to-noise ratio. R is the vector correlation coefficient described in Figure 4. $R > 0.38$, $|R| <$
417 0.38 , and $R < -0.38$ refers to the angle between collocated vectors smaller than 67.5° , between
418 67.5° and 112.5° , and between 112.5° and 180° respectively.
419

	Obs. SD (cm/s)	Bias (cm/s)	RMSD (cm/s)	SDE (cm/s)	SNR (cm/s)	Percentage of current vector correlation (%)		
						$R > 0.38$	$ R < 0.38$	$R < -0.38$
Overall (2267)	11.5	-5.3	11.0	9.6	0.80	67	14	17
Current > 30 cm/s (237)	10.3	-20.6	24.2	12.8	0.80	88	5	7
Current < 10 cm/s (896)	2.4	0.8	4.8	4.8	0.92	56	19	25
Cold fresh $T < 2C$, $S < 27\text{psu}$ (127 from 2019)	7.1	-3.2	8.0	7.3	0.38	43	24	32
Yukon River Discharge $T > 5C$, $S < 28\text{psu}$ (12 from 2019)	4.2	-4.8	7.2	5.4	0.40	58	33	8
Hanna Shoal > -30m (44)	4.9	-4.8	7.1	5.3	0.60	55	14	32
> -30m (96)	8.7	-6.0	10.9	9.0	0.46	67	13	20

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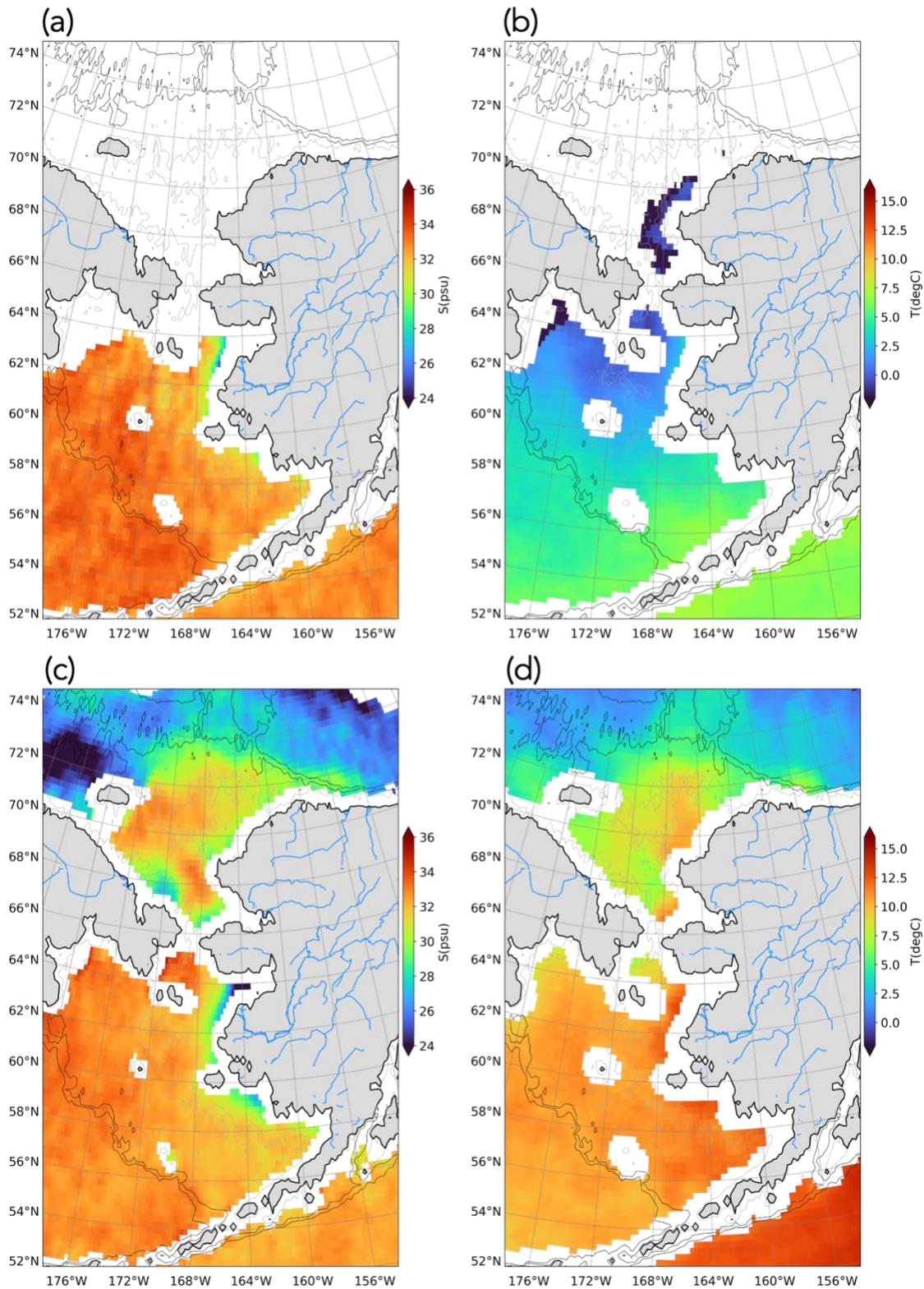
421 **Figure 1:** Three saildrone tracks (1035, 1036, 1037) in 2019 colored by (a) saildrone 0.5 m
 422 salinity and (b) saildrone 0.5 m temperature. Two saildrones (1020, 1021) in 2018 colored by (c)
 423 saildrone 0.5 salinity and (d) saildrone 0.5 m temperature. The contour background is the
 424 bathymetry. The thin black contours are depth contours of -50, -200 and -1000 m. The magenta
 425 texts are the dates (mm/dd) of the locations of 1036 in 2019 and of 1020 in 2018.



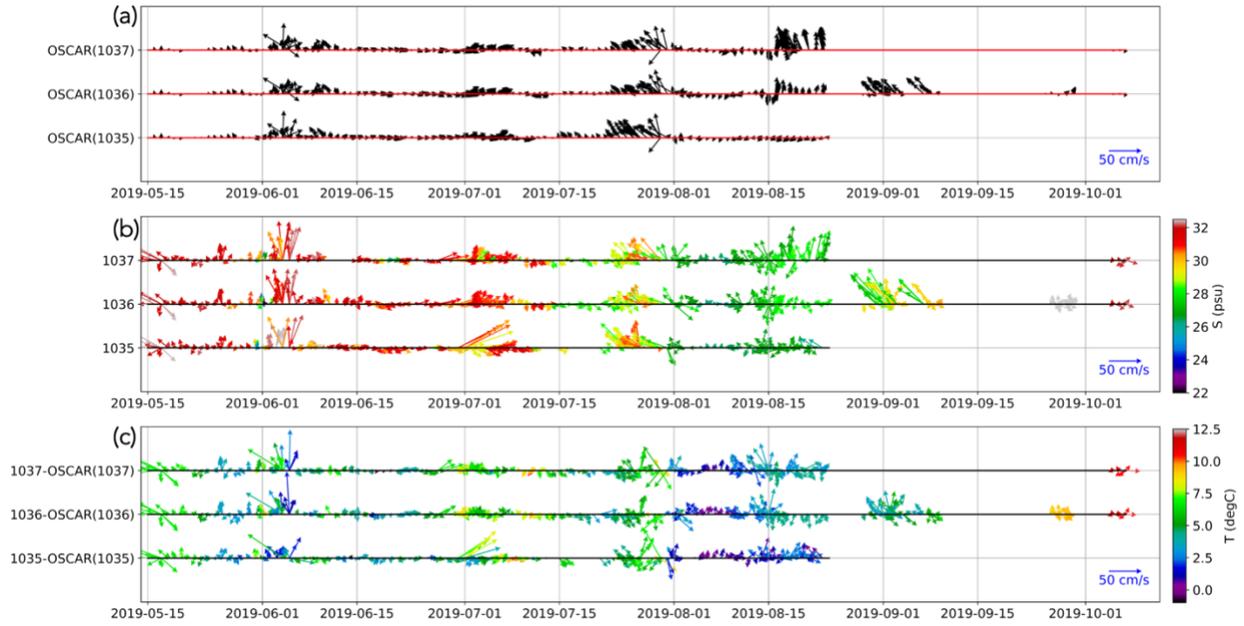


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429 **Figure 2:** Satellite maps of (a,c) SSS and (b,d) SST on (a,b) 2019/05/15 and (c,d) 2019/09/15.
430 The bathymetry contour levels are the same as in Figure 1. Data with sea ice concentration larger
431 than 0.1, or uncertainties larger than 1 psu for SSS and 1 °C for SST are not plotted.

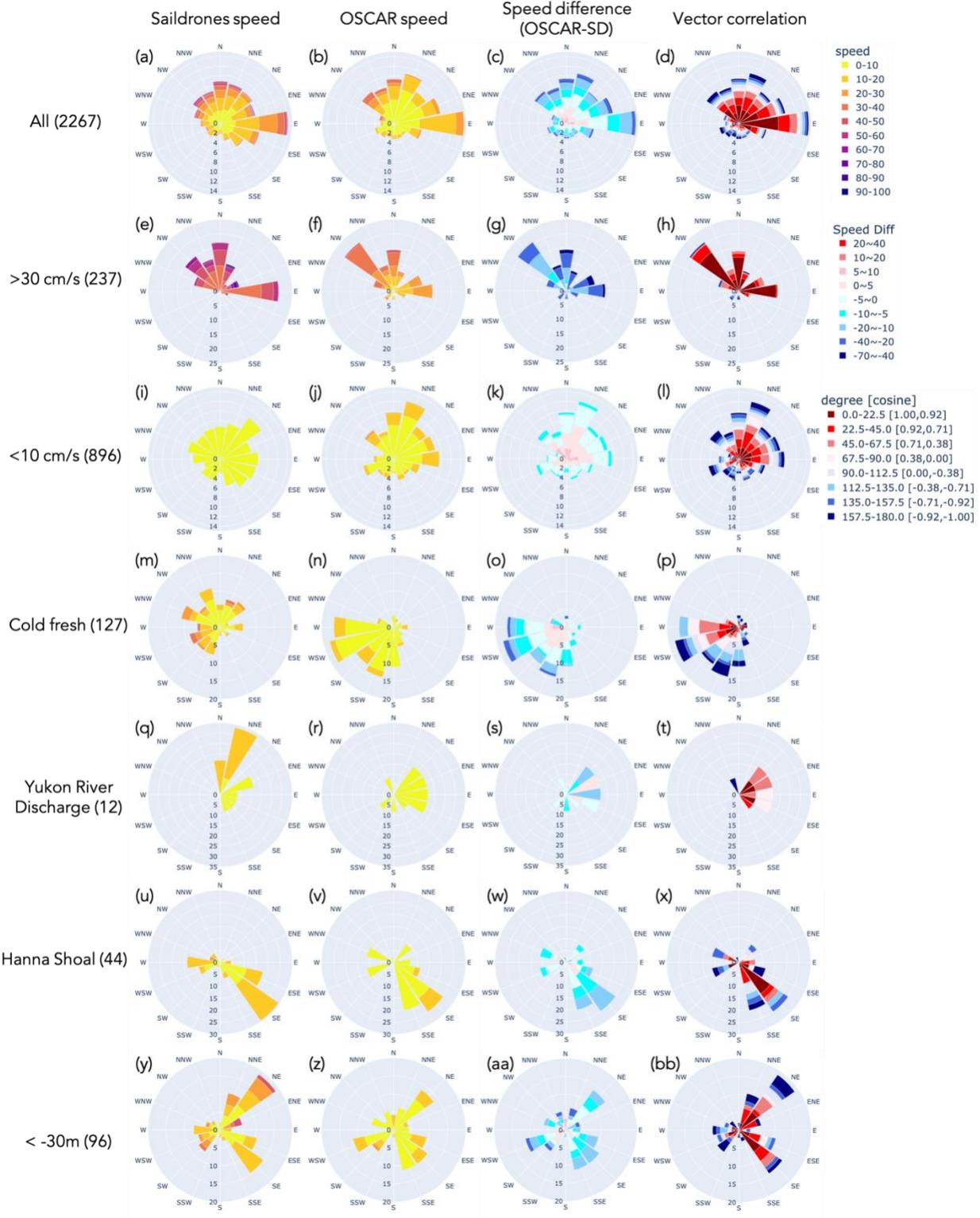


433 **Figure 3:** The time series of current vectors of (a) OSCAR 15m current in time and space along
434 the saildrone track, (b) saildrones 10-20 m current vectors averaged around each OSCAR ¼
435 degree data grid within 1 days to match OSCAR grids colored with saildrone 0.5 m salinity, and
436 (c) their vector differences (saildrone - OSCAR) colored with saildrone 0.5 m temperature.
437 OSCAR(1035), OSCAR(1036), and OSCAR(1037) are the corresponding OSCAR current in
438 time and space along the saildrone track.

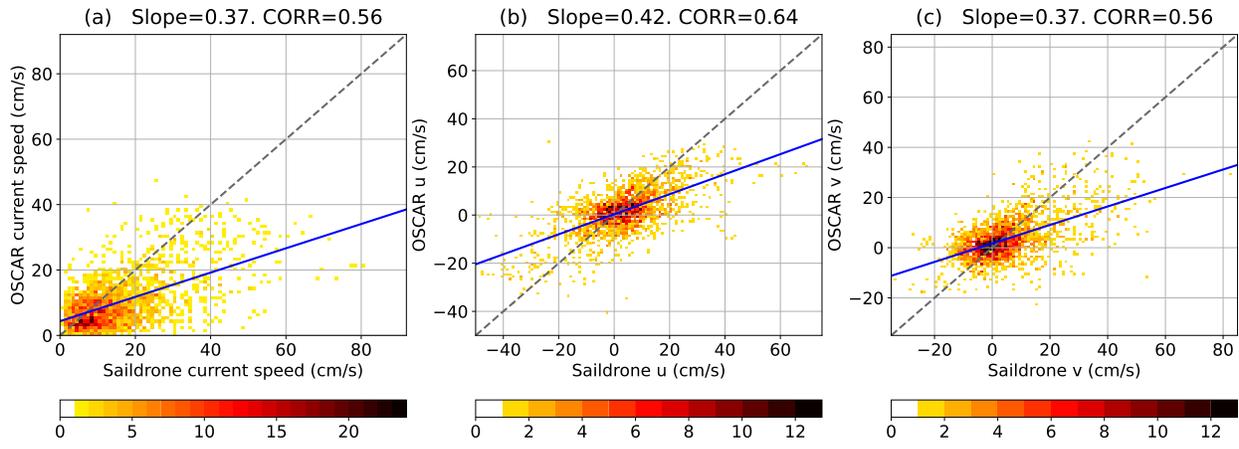


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441 **Figure 4:** The wind rose plot summarizing the collocated vectors for (first column: a,e,i,m,q,u,y)
442 saildrone 10-20 m averaged current vectors, (second column: b,f,j,n,r,v,z) OSCAR 15-m current
443 vectors, (third column: c,g,k,o,s,w,aa) current speed difference (OSCAR - saildrone), and (last
444 column: d,h,l,p,t,x,bb) vector correlation of the collocated pairs between OSCAR and saildrones.
445 Each row represents the results of a specific subset of collocated vectors described in the
446 subtitles on the left with the number of pairs in the parenthesis. The convention here is direction
447 towards; i.e., the “spoke” towards “N” represents current flowing northward. The degrees in the
448 rightmost column subplots (c,f,i,l,o,r,u) are the angle between the collocated OSCAR and
449 saildrone current vectors. The values in the brackets for vector correlation colorbar are the limits
450 of the cosine similarity. The unit of current speed is cm/s.



453 **Figure 5:** 2D histogram of (a) current speed (b) zonal current velocity and (c) meridional current
454 velocity between the collocated saildrones and OSCAR data. The blue solid line is the linear
455 regression line. The unit of the colors is the number.
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