1	Towards an assessment of riverine dissolved organic carbon in surface waters of the Western				
2	Arctic Ocean based on remote sensing and biogeochemical modeling				
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# 24 Abstract

Future climate warming of the Arctic could potentially enhance the load of terrigenous dissolved 25 organic carbon (tDOC) of Arctic rivers due to increased carbon mobilization within watersheds. A 26 greater flux of tDOC might impact the biogeochemical processes of the coastal Arctic Ocean (AO) 27 and ultimately its capacity of absorbing atmospheric CO<sub>2</sub>. In this study, we show that sea surface 28 29 tDOC concentrations simulated by a physical-biogeochemical coupled model in the Canadian Beaufort Sea for 2003-2011 compare favorably with estimates retrieved by satellite imagery. Our 30 31 results suggest that, over spring-summer, tDOC of riverine origin contributes to 35 % of primary production and that an equivalent of  $\sim 10$  % of tDOC is exported westwards with the potential for 32 33 fueling the biological production of the eastern Alaskan nearshore waters. The combination of model and satellite data provide promising results to extend this work to the entire AO so as to 34 quantify, in conjunction with in-situ data, the expected changes in tDOC fluxes and their potential 35 36 impact on the AO biogeochemistry at basin scale.

### 38 **1. Introduction**

The Arctic Ocean (AO) receives ~10% of the global freshwater discharge (Opsahl et al., 1999 and 39 references therein) of which the larger part (~54-64 %) originates from six main pan-Arctic rivers 40 41 (Haine et al., 2015; Holmes et al., 2012; Aagaard and Carmack, 1989). Over the past 30 years, the 42 Arctic freshwater cycle intensified as reflected by changes in snow cover (Bring et al., 2016), evapotranspiration from terrestrial vegetation (Bring et al., 2016), and precipitation (Vihma et al., 43 44 2016). It resulted into an increase of the freshwater discharge from North American and Eurasian 45 rivers by ~2.6 % and ~3.1 % per decade, respectively (Holmes et al., 2015). More than half the soil 46 organic carbon stock on Earth is contained in the permafrost of the Arctic watersheds (Tarnocai et 47 al., 2009). With the warming of the lower atmosphere, the permafrost undergoes a substantial thawing (Romanovsky et al., 2010) likely to alter the organic carbon content and quality of inland 48 49 waters. In the past decades, the flux of dissolved organic carbon (DOC) decreased in the Yukon 50 River (40 %; Striegl et al., 2005) while it increased at the Mackenzie River mouth (~39 %; Tank et al., 2016). These contrasting responses to climate change suggest that the direction of future trends 51 52 of DOC concentrations and fluxes to the AO are very uncertain (Abbott et al., 2016).

53 The coastal AO influenced by large river plumes is hence exposed to changing conditions. Coastal waters are supplied in riverine organic carbon all year round with a maximal flux in spring-early 54 55 summer when the freshwater discharge reaches a seasonal maximum. In river waters, DOC is present in higher concentration than the particulate form (Le Fouest et al., 2013; Dittmar et al., 56 2003). It accounts for more ~82 % of the flux of total riverine organic carbon (McGuire et al., 2009). 57 The pan-Arctic flux of riverine DOC to the AO is estimated to 33-37.7 TgC yr<sup>-1</sup> (Holmes et al., 58 59 2012; Manizza et al., 2009; McGuire et al., 2009; Raymond et al., 2007). As the organic carbon formed by phytoplankton, terrigenous DOC (tDOC) can be considered as new carbon fueling 60 61 annually the upper AO. To that respect, and regardless of its distinct nature and fate, the flux of riverine DOC would be equivalent to 10-19 % of the AO primary production (Stein and Macdonald, 62 2004; Bélanger et al., 2013). In the oligotrophic Beaufort Sea, this proportion would reach ~34 % (S. 63

64 Bélanger, pers. comm.). Riverine DOC is hence a significant pool in the Arctic carbon cycle that can markedly modify the biological production and biogeochemistry of the AO waters. Within the 65 66 pelagic food web, riverine DOC can be assimilated and transformed, promoting both phytoplankton 67 and bacterioplankton production (Le Fouest et al., 2015; Tank et al., 2012). Riverine DOC can also modulate the air-sea fluxes of CO<sub>2</sub>. In present climatic conditions, Manizza et al. (2011) suggest 68 that the mineralization of riverine DOC into dissolved inorganic carbon would induce a 10 % 69 70 decrease of the net oceanic CO<sub>2</sub> uptake at the pan-Arctic scale. In East Siberian shelves, the 71 degradation of terrestrial organic carbon would be partly responsible for the sea surface 72 acidification (Semiletov et al., 2016).

73 In recent studies, riverine DOC flux data were used in a 3D ocean-biogeochemical coupled model to investigate the fate of riverine DOC within surface Arctic waters (Le Fouest et al., 2015; Manizza 74 et al., 2013, 2011, 2009). However, simulated spatial and temporal changes in riverine DOC 75 76 concentrations have not yet been compared with remote sensing data to assess the model predictive 77 ability. Such a model-satellite comparison allows validating the model and then using it with 78 confidence to resolve the annual cycle of riverine DOC, a prerequisite for a robust assessment of the 79 riverine DOC contribution to the Arctic carbon cycle. To this end, riverine DOC concentrations at 80 the sea surface obtained from a previous model run described in Le Fouest et al. (2015) and tDOC 81 concentrations derived from remote sensing data were analyzed for the Canadian Beaufort Sea. As riverine DOC accounts for more than 99 % of the total tDOC exported to the AO (McGuire et al., 82 2009), we will use the term tDOC for both the model and remotely sensed data. Our goals are to 83 84 compare tDOC data derived from the model and from remote sensing using skill metrics, to assess 85 the model capacity to reproduce the observed seasonal and spatial variability in tDOC, and to provide bulk estimates of the seasonal tDOC stock and lateral fluxes within the surface coastal 86 87 waters using a combination of these two approaches.

The paper is organized as follows. First, we describe the two different approaches used to quantify
tDOC within the AO, i.e. a semi-analytical method based on remote sensing and a regional ocean-

90 biogeochemical coupled model that includes explicit fluxes of riverine DOC to the AO. Second, we
91 compare the distribution and export flux of tDOC within surface waters of the Beaufort Sea
92 estimated by the model and remote sensing. Finally, we discuss future developments of
93 biogeochemical models necessary to simulate successfully the carbon budget of Arctic coastal
94 waters in a warming world.

95

## 96 2. Material and methods

#### 97 2.1 Remote sensing data

Level 1A scene images acquired from the MODerate-resolution Imaging Spectroradiometer 98 99 (MODIS) aboard Aqua satellite were downloaded from the NASA ocean color website 100 (https://oceandata.sci.gsfc.nasa.gov/MODIS-Aqua/L1/). After geometric correction, remote sensing reflectance. Rrs( $\lambda$ ) data at 412, 443, 488, 531, 555, and 667 nm were obtained by applying 101 102 atmospheric correction proposed by Wang and Shi (2009) with modifications adapted to Arctic environments (Doxaran et al., 2015; Matsuoka et al., 2016). The light absorption coefficients of 103 104 colored dissolved organic matter at 443 nm ( $a_{CDOM}(443)$ ) were derived from the Rrs( $\lambda$ ) data using 105 the gsmA algorithm (Matsuoka et al., 2017) that optimizes the difference between satellite  $Rrs(\lambda)$ 106 and  $Rrs(\lambda)$  calculated using parameterization of absorption and backscattering coefficients for Arctic waters (Matsuoka et al., 2011, 2013). tDOC concentrations were estimated from the 107  $a_{CDOM}(443)$  data using an empirical relationship between DOC and  $a_{CDOM}(443)$  established in the 108 109 Southern Beaufort Sea (Matsuoka et al., 2013). Since DOC concentrations estimated using ocean color data are based on a highly significant DOC versus  $a_{CDOM}(443)$  relationship ( $R^2 = 0.97$ ; 110 Matsuoka et al., 2012), the DOC is considered to be of terrestrial origin. Errors of intercept, slope, 111 and  $a_{CDOM}(443)$  were propagated into the in-situ (empirical) DOC versus  $a_{CDOM}(443)$  relationship. It 112 113 resulted into a mean uncertainty of the tDOC concentration estimates of 28 % (see Appendix A2 of Matsuoka et al., 2017). Scene images of tDOC concentrations were used to make monthly 114 115 composite images at 1 km horizontal resolution.

### 117 **2.2 3D physical-biogeochemical model data**

We used sea surface tDOC concentrations and ocean currents simulated over 2003-2011 by a 118 119 previous pan-Arctic model run ("RIV run") whose setup is fully detailed in Le Fouest et al. (2015). 120 The pan-Arctic model data were extracted on the remote sensing geographical domain focused on 121 the southern Beaufort Sea. We provide here a brief description of the physical-biogeochemical coupled model used to generate the "RIV run". The MITgcm (MIT general circulation model) 122 123 ocean-sea ice model (Nguyen et al., 2011, 2009; Losch et al., 2010; Condron et al., 2009) has a 124 variable horizontal resolution of ~18 km and covers the Arctic domain with open boundaries at 125 55°N on the Atlantic Ocean and Pacific Ocean sides. The open ocean boundaries are constrained by 126 potential temperature, salinity, flow, and sea-surface elevation derived from integrations of a global 127 configuration of the MITgcm model (Menemenlis et al., 2005). Atmospheric forcings (10 m winds, 128 2 m air temperature and humidity, and downward long and short-wave radiation) are taken from the six-hourly data sets of the Japanese 25 year ReAnalysis (JRA-25) (Onogi et al., 2007). In addition 129 130 to precipitations, the hydrologic forcing includes a monthly climatology of freshwater discharge 131 from 10 pan-arctic watersheds (Manizza et al., 2009). Monthly mean estuarine fluxes of freshwater 132 are based on an Arctic Runoff database (Lammers et al., 2001; Shiklomanov et al., 2000). For each 133 watershed, the river discharge forcing is associated with a monthly climatology of riverine DOC concentration (Manizza et al., 2009). The total annual load of tDOC in the model is 37.7 TgC yr<sup>-1</sup>. It 134 is consistent with previous values reported in Raymond et al. (36 TgC yr<sup>-1</sup>; 2007) and Holmes et al. 135 (34 TgC yr<sup>-1</sup>; 2012) and obtained by using load estimation models linking riverine DOC 136 137 concentrations to river discharge data. The physical model is coupled with a 10-compartment 138 biogeochemical model (Lee et al., 2016; Le Fouest et al, 2015). The biogeochemical model 139 explicitly accounts for dissolved inorganic nutrients (nitrate and ammonium), small and large phytoplankton, protozooplankton, mesozooplankton, bacterioplankton, detrital particulate and 140 dissolved organic nitrogen, and tDOC (Lee et al., 2016; Le Fouest et al., 2015). The tDOC 141

142 compartment couples the marine and terrestrial cycling of organic matter though tDOC recycling 143 into inorganic nutrients by bacterioplankton. We set to 15 % the percentage of tDOC entering the 144 model as usable by the bacterioplankton compartment. This value was estimated based on the mean 145 yearly percentages of the total load of riverine DOC considered as biodegradable DOC for six 146 major Arctic rivers given in Wickland et al. (2012).

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# 148 **2.3 Analysis**

149 Remotely sensed and simulated tDOC data were binned for the months of June, July, August and September over the 9-year period (2003-2011) to get the best areal coverage in the satellite 150 151 composites. The remotely sensed tDOC concentrations were regridded on the model horizontal grid. Skill metrics were used to compare the remotely sensed estimates of tDOC with their simulated 152 counterparts. The metrics included the correlation coefficient (r), the unbiased root mean square 153 154 error (RMSE), the Nash-Sutcliffe model efficiency index (MEF), the geometric bias, and the geometric RMSE (see Stow et al., 2009; Doney et al., 2009; Nash and Sutcliffe, 1970). The metrics 155 156 are computed as follows:

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$$r = \frac{\sum_{n=1}^{N} (sat_n - )(mod_n - \overline{mod})}{\sqrt{\sum_{n=1}^{N} (sat_n - \overline{sat})^2 \sum_{n=1}^{N} mod_n - \overline{mod}^2}}$$
(Eq. 1)

unbiased RMSE = 
$$\sqrt{\frac{1}{N}\sum_{n=1}^{N} \left(mod_n - sat_n - \left(\overline{mod} - \overline{sat}\right)\right)^2}$$
 (Eq. 2)

$$MEF = \frac{\sum_{n=1}^{N} (sat_n - \overline{sat})^2 - \sum_{n=1}^{N} (sat_n - mod_n)^2}{\sum_{n=1}^{N} (sat_n - \overline{sat})^2}$$
(Eq. 3)

geometric bias = 
$$e^{(\overline{mod} - \overline{sat})}$$
 (Eq. 4)

geometric RMSE = 
$$\sqrt{e^{\left(\frac{1}{N}\sum_{n=1}^{N}(mod_n - sat_n)^2\right)}}$$
 (Eq.5)

where N is the number of tDOC data, and  $\overline{sat}$  and  $\overline{mod}$  are the remotely sensed and the simulated tDOC averages, respectively. Monthly fluxes of tDOC were calculated and summed along two cross-shelf transects (see upper-middle panel in Fig. 1). At each grid cell, the model flux estimate was computed as the product of the simulated sea surface current velocity with the simulated tDOC concentration. The remote sensing flux estimate was computed as the product of the simulated sea surface current velocity with the remotely sensed tDOC concentration.

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### 166 **3. Results and discussion**

# 167 **3.1 tDOC concentrations and distribution**

Over the Mackenzie shelf, the plume of high-tDOC (> 120 mmolC m<sup>-3</sup>) had a maximal areal extent 168 in June for both the model and the satellite data (Fig. 1). This coincided with the seasonal peak of 169 river discharge in June as parameterized in the model and generally depicted by in-situ time series 170 171 (Yang et al., 2015). From July to September, the high-tDOC areal extent progressively decreased following the seasonal pattern of riverine freshwater discharge (see Yang et al., 2015; Manizza et al., 172 2009). This seasonal pattern was observed both in the model and satellite data. The simulated tDOC 173 174 concentrations were lower than in the satellite record in Mackenzie Bay and east of the Mackenzie Bay, especially in June (by 44 % in average) and July (by 27 % in average). In the Beaufort and 175 Chukchi seas, first year sea ice represents a carbon flux to the ocean of  $2 \times 10^{-4}$  TgC yr<sup>-1</sup> (Rachold 176 et al., 2004). This flux is 4 orders of magnitude lower than the tDOC supply from the Mackenzie 177 River specified as boundary conditions in the model (2.54 TgC yr<sup>-1</sup>). Similarly, tDOC eroded from 178 permafrost stored in the North American shores would account for only ~ $0.5-1.6 \times 10^{-4}$  TgC yr<sup>-1</sup> 179 (Tanski et al., 2016; Ping et al., 2011, using a DOC:POC ratio of 1:900 as in Tanski et al., 2016) to 180  $\sim 2 \times 10^{-3}$  TgC yr<sup>-1</sup> (McGuire et al., 2009). With regard to these flux values, tDOC originating from 181 182 both melted sea ice and eroded permafrost, not taken into account in the model, are hence not believed to explain the model-satellite discrepancies (Fig. 1). Other factors might contribute to these 183 model-satellite differences observed nearshore. First, the model does not distinguish between the 184

185 two main pathways of the Mackenzie River discharge entering the shallow delta zone. In June, the Mackenzie Bay receives most of the fresh and turbid river water ( $\sim 66 \%$ ) while the remaining  $\sim 33 \%$ 186 spreads east of the delta in Kugmallit Bay (Davies, 1975). This pattern was particularly well 187 188 captured by the remotely sensed data in June-July (Fig. 1). Second, the inner Mackenzie shelf (< 20 m depth) is bounded during winter by a thick ridged ice barrier grounded on the sea floor called 189 stamukhi (Macdonald et al., 1995). The stamukhi retains the turbid river water within the inner shelf 190 191 in winter. When sea ice breaks up and the freshet reaches its seasonal maximum in spring, the 192 retained turbid waters spread farther within the coastal zone. Contrary to the model, the remote sensing data could resolve this particular feature explaining the higher tDOC concentrations 193 194 observed nearshore in June (see Fig. 1). Such a pattern observed for tDOC is also reported for 195 terrigenous particulate organic matter (Doxaran et al., 2015). Further offshore on the Mackenzie shelf, as delimited by the 300 m isobaths both remotely sensed and simulated concentrations of 196 tDOC were within the range of values measured in spring (~110-230 mmolC m<sup>-3</sup>; Osburn et al., 197 2009) and summer (~60-100 mmolC  $m^{-3}$ ; Para et al., 2014). Overall, the model and the satellite data 198 captured the seasonal cycle and spatial distribution of tDOC concentrations in the study area. 199

200 Skill metrics were computed over the whole study area (see Fig. 1) to provide a quantitative 201 comparison of tDOC simulated with the model and satellite data (Table 1). For all months, the 202 correlation coefficient was relatively high (0.78<r<0.82) within the range of values obtained for sea surface dissolved inorganic nutrients simulated by global models (r>0.75; Doney et al., 2009). 203 Regardless of amplitude, the r values showed that the simulated and remotely sensed tDOC 204 205 concentrations presented similar patterns of variation. The size of the model-satellite discrepancies 206 was given by the unbiased RMSE. Overall, the unbiased RMSE decreased from June (41.4 mmolC  $m^{-3}$ ) to September (29.3 mmolC  $m^{-3}$ ). This result suggested that the model accuracy increased from 207 spring to summer. The model capability for predicting tDOC relative to the average of the remote 208 209 sensing counterparts was estimated by the model efficiency index (- $\infty < MEF \le 1$ ) (Nash and

210 Sutcliffe, 1970). The MEF is a normalized statistic that relates the residual variance between the 211 simulated and remotely sensed tDOC concentrations to the variance within the remotely sensed tDOC data (see Eq. 3). A MEF value near zero means that the residual variance compares to the 212 213 remotely sensed variance, i.e. that the model predictions are as accurate as the mean of the satellite 214 data. As the MEF increases towards a value of one, the residual variance becomes increasingly lower than the observed variance. For all months, the MEF was positive (0.26-0.60) suggesting that 215 216 tDOC concentrations simulated by the model were an acceptable predictor relative to tDOC 217 concentrations derived from remote sensing, especially in June-July. In order to give a more even 218 weight to all of the data and to limit the skewness towards the higher tDOC concentrations, metrics 219 based on log-transformed tDOC data were also computed. For all months, the geometric RMSE was 220 close to one and span between 1.02 and 1.12. It suggested that the model-satellite data dispersion 221 was relatively small when the positive skewness was reduced. In June, the relatively high unbiased 222 RMSE could be partly due to high tDOC concentrations as suggested by the relatively low geometric RMSE (1.07). Finally, the computed geometric bias informs on the direction of the 223 224 model-satellite discrepancies. For all months, the geometric bias (1.07-1.32) was higher than one 225 meaning that the model tended, on average, to overestimate the observations over the whole domain. 226 The highest geometric bias was reported in August (1.32), when the river discharge was low, 227 suggesting that tDOC removal was likely underestimated in the model in late summer. A Taylor 228 diagram (Taylor, 2001) was produced to provide a synthetic and complementary overview of how 229 the simulated and remotely sensed tDOC concentrations compared seasonally in terms of 230 correlation, amplitude of variations (given by the standard deviations), and normalized model-231 satellite discrepancies (Fig. 2). All months differed by their normalized RMSE and amplitude of 232 variations while the correlation coefficient was close to  $\sim 0.8$  (see Table 1). The model best 233 performed in simulating tDOC in July, just after the seasonal peak of river discharge, followed by 234 the months of June and August. June and August were very close months despite distinct seasonal patterns of river discharge (high and low, respectively), whereas September showed the highest 235

236 model-satellite data dispersion. With respect to satellite estimates, the skill metrics overall
237 suggested that the model could reliably simulate tDOC concentrations in surface waters over a wide
238 range of river discharge and tDOC load.

239

## 240 **3.2 tDOC stock and lateral export fluxes**

The overall agreement between the model and the satellite tDOC concentrations allowed the 241 242 assessment of the mean areal stock and lateral fluxes of tDOC using the mean surface ocean circulation simulated by the MITgcm (Table 2). The monthly-averaged (June to September) areal 243 244 stock of tDOC over the Mackenzie shelf as delimited by the 300 m isobaths was estimated to 1.37 TgC (Table 2). The bias between the model and the satellite data was the highest in August but did 245 246 not exceed +8.2 % (0.1 Tg C). This result is consistent with the highest geometric bias reported in 247 August (Table 1). In the model, the removal of tDOC through photo-oxidation (Bélanger et al., 2006) was not taken into account. Assuming an annual mean mineralization rate of tDOC of ~0.02 248 249 TgC (Bélanger et al., 2006), this process would explain <2 % of the reported tDOC difference. In 250 the model, bacterioplankton consumed tDOC to produce ammonium usable in turn by phytoplankton. In the Beaufort Sea, this pathway contributed to primary production by 35 % on 251 252 average over 2003-2011. However, the simulated rates of bacterioplankton production ( $\leq 30 \text{ mgC}$  $m^{-2} d^{-1}$ ) still remained in the lower range of those measured in the Beaufort Sea (25-68 mgC  $m^{-2} d^{-1}$ ; 253 Ortega-Retertua et al., 2012; Vallières et al., 2008). The likely underestimation of the tDOC 254 255 removal by bacterioplankton in the model during summer months might also contribute to the reported bias between the model and the satellite data. Nevertheless, the bias remained moderate 256 257 with respect to values reported for June, July and September (-1.5 % to -2.8 %) (Table 2).

258 Combining the modeling and remote sensing approaches allowed for the reconstruction of the 259 dominant surface pattern in lateral tDOC fluxes in the Canadian Beaufort Sea from June to 260 September (Fig. 3). Two north-south transects were defined east (Cape Bathurst) and west

261 (Mackenzie Trough) of the Mackenzie shelf (see upper-middle panel in Fig. 1). The net seasonal flux was westward along the two transects following the anticyclonic circulation pattern of the 262 263 Beaufort gyre (Mulligan et al., 2010) and was maximum in June and September. The flux was at 264 least three times higher along the western transect near the Mackenzie Through than east at Cape Bathurst. This suggests a net export of tDOC towards the Alaskan part of the Beaufort Sea. In 265 contrast, whilst the flux in July and August remained oriented westward near the Mackenzie Trough, 266 267 it was reversed at Cape Bathurst. In July, the tDOC flux was still 1.3 to 1.7 times higher along the 268 western transect. In August, however, there was more tDOC (~1.4-fold) exported eastward at Cape 269 Bathurst than exported westward near the Mackenzie Through.

Along the two transects, the simulated fluxes were higher than those derived from remotely sensed 270 271 tDOC concentrations (Fig. 3). The monthly bias between the model and the satellite flux estimates varied between 0 % and +18.2 %. The bias on the seasonal net flux was moderate (+8.3 %) near the 272 Mackenzie Trough but reached +25 % at Cape Bathurst. The seasonal mean flux however was one 273 274 order of magnitude lower than near the Mackenzie Trough. The flux estimates suggested that, 275 despite discrepancies in tDOC concentrations, the modeling and remote sensing approaches provided robust estimates of the lateral transport of tDOC in surface waters in late spring-summer. 276 277 Because of sea ice and cloud cover, the satellite retrieval was limited to a temporal window covering a third of a year only, i.e. from June to September. The yearly mean lateral flux of tDOC 278 was computed from the simulated data along the Mackenzie Trough transect and it reached 0.31 279 TgC. The flux of tDOC cumulated over June to September along this transect (0.12-0.13 TgC) 280 281 represented ~42 % of this annual flux (0.31 TgC), which is consistent with the fraction of the 282 annual discharge of freshwater by the Mackenzie that occurs during spring-summer (~50 %; 283 McClelland et al., 2012). Using stable isotope techniques on pelagic particulate organic matter, Bell et al. (2016) showed that OC originating from the Mackenzie outflow in summer was incorporated 284 285 within bentho-pelagic food webs as far as the eastern Alaskan shelf. In nearshore waters of this part 286 of the Beaufort Sea, the study of Dunton et al. (2006) using stable isotopes also suggested that

tDOC from the Mackenzie River could add to the local terrigenous carbon inputs mediated by coastal erosion and smaller rivers to fuel the biological production in summer. Using the model and satellite data, we report that an equivalent of  $\sim 10 \%$  (0.12-0.13 TgC) of the cumulated flux of tDOC delivered by the Mackenzie River over spring-summer (1.32 TgC) was exported westward in the Alaskan Beaufort Sea along the Mackenzie Trough transect.

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## 293 4. Perspectives

294 The results of our study suggest that the model is in fair agreement with the surface tDOC fields 295 remotely sensed in spring-summer when most of the riverine flux occurs. The comparison allows an 296 evaluation of the model and justifies its use to resolve the annual cycle of tDOC. Because satellite 297 imagery provides data only during spring-summer, further uncertainties still remain in the model in 298 fall-winter in terms of tDOC concentrations and spatial distribution. In addition, the model involves 299 some limitations mostly due to the biogeochemical processing of tDOC, which is complex to 300 translate into robust mechanistic equations as highly dependent on the availability of in-situ data in 301 Arctic waters. For instance, the riverine tDOC compartment is split in the model into a labile and a 302 non-labile fraction (see Le Fouest et al., 2015). This parameterization strongly constrains the 303 removal of tDOC by bacterioplankton and therefore the tDOC concentrations simulated within 304 surface waters. In natural waters, however, tDOC is made of a complex mixture of compounds that 305 differ by their chemical composition and age (Mann et al., 2016) and so along the seasons (Wickland et al., 2012, Mann et al., 2012). The chemical nature of tDOC impacts its bioavailability 306 307 estimated to average 6 % to 46 % of the total tDOC pool with marked disparities amongst the 308 seasons and the rivers (Mann et al., 2012). Nevertheless, the general trend for the six major Arctic 309 rivers (Kolyma, Yukon, Mackenzie, Ob, Yenisey and Lena) is a more labile tDOC pool in winter 310 than in spring and summer (Wickland et al., 2012). In the Kolyma River, Mann et al. (2012) report a higher labile fraction in spring ( $\sim 20$  %) than in summer (< 10 %) as the exported tDOC is younger 311 during the freshet. Such a pattern is, however, not clearly evidenced in the Mackenzie River (e.g. 312

Wickland et al., 2012). We suggest that a more realistic representation in the model of the nature of the organic matter entering the coastal waters including the riverine flux of both dissolved organic carbon and nitrogen along with an improved C:N stoichiometry for bacterioplankton uptake (see Le Fouest et al., 2015) might improve the tDOC concentrations simulated in surface AO waters.

In the model, the seasonal forcing of tDOC was based on DOC measurements gathered hundreds 317 kilometers upstream the rivers' mouths. This prevented any DOC enrichment of the Mackenzie 318 319 River water as it flows through the delta (see Emmerton et al., 2008) with, as a consequence, a 320 likely underestimation of tDOC concentrations simulated in nearshore waters. Therefore, the 321 quantification of the tDOC flux from the watersheds to the coastal AO poses as another key issue to 322 addressing the role of tDOC in the biogeochemistry of shelf waters. Recently, watersheds models were developed to assess this tDOC flux (Tank et al., 2016; Kicklighter et al., 2013; Holmes et al., 323 2012). Such models provide realistic estimates but still require improvements as watersheds 324 325 properties and mechanistic processes underlying the tDOC mobilization and riverine transport are complex to set up (see Kicklighter et al., 2013). The remote sensing of high resolution ocean color 326 327 data is increasingly used to assess tDOC concentrations in large pan-Arctic rivers during the open 328 water season (Herrault et al., 2016; Griffin et al., 2011). Ocean color techniques could then prove 329 useful in the future to improve the tDOC time series set at models boundaries by accounting for 330 instance for year-to-year variations of tDOC concentrations during the freshet period.

331 In our study, the remotely sensed tDOC concentrations retrieved in shelf waters provide the advantage of already integrating the effect of the watersheds processes such as mobilization, 332 333 transformation and transport at the seasonal and synoptic time scales. However, we acknowledge 334 that the temporal coverage of the remote sensing data is restricted to spring and summer. Because of 335 clouds and sea ice, we miss the winter season when tDOC is the most labile (e.g. Wickland et al., 336 2012) and likely subject to remineralization. In the Mackenzie River, about 25 % of the annual load of labile tDOC occurs during winter (Wickland et al., 2012). Despite this limitation, and in regard to 337 the model-satellite data comparison, the assimilation of remotely sensed tDOC data into Arctic 338

339 models could still offer an interesting perspective as it might result in more realistic simulated fields of tDOC in spring and summer when the river discharge and tDOC export is the highest. Physical 340 341 and biological data have already been assimilated into Arctic predictive models to make the simulated sea surface temperature, salinity, sea ice extent and thickness, and chlorophyll more 342 reliable (Simon et al., 2015; Massonnet et al., 2015). We may hence expect the assimilation of 343 remotely sensed tDOC concentrations to mitigate, at least partly, the issues linked to setting up 344 345 realistic tDOC forcings within predictive models. For instance, the assimilation of remotely sensed 346 tDOC data in open waters might help accounting for the interannual variations of tDOC delivered 347 by rivers, which are not resolved by the coupled model that is constrained by a monthly climatology 348 of tDOC load (see Manizza et al., 2009).

Improving the capability of Arctic models to resolve the fate and pathways of tDOC in the AO will 349 350 require certain limitations to be unlocked. To this purpose, future model developments should lie on 351 the always increasing observational effort realized by mean of field campaigns and new remote sensing techniques. As a prerequisite, we can reasonably encourage improvements of the riverine 352 353 forcings to better encompass the seasonal to interannual variability of the terrigenous dissolved 354 organic matter exported to the coastal AO, both in qualitative and quantitative terms. We also 355 suggest bacterioplankton dynamics to be better represented in biogeochemical models. In particular, 356 the processes related to the competition for resources as dissolved organic carbon and nitrogen of both allochtonous and autochtonous origin are likely to play an important role in mediating 357 bacterioplankton growth and tDOC remineralization in Arctic coastal waters impacted by river 358 plumes. Realistic fields of tDOC simulated by Arctic ocean-biogeochemical coupled models would 359 help for a more accurate assessment of CO<sub>2</sub> fluxes at the ocean-atmosphere interface. Arctic models 360 that would combine realistic terrestrial fluxes of organic matter along with a robust representation 361 362 of the pathways and processes responsible for its transformation in the AO would open an interesting perspective to address the effect on the Arctic carbon cycle of ongoing and future 363 changes in the land-ocean continuum. The increase in seawater temperature of the AO due to global 364

365 warming (Timmermans, 2016) might promote in the future the metabolism and respiration rates of 366 marine bacterioplankton (Vaquer-Sunyer et al., 2010; Kritzberg et al., 2010). This enhanced 367 microbial activity could then liberate extra nutrients provided by the remineralization of terrigenous 368 organic matter that will then be available for primary production. This process might have an 369 impact not only on the seasonal cycle of PP in the AO but also implications for the higher levels of 370 the marine food webs of the AO, both benthic and pelagic.

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# 372 Data availability

373 Data used in this study are available at http://www.obs-lienss.cnrs.fr/Publications/BGD
374 \_data\_nc.tar.

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386 Figure 1. Monthly climatology (2003-2011) of surface tDOC concentration (mmolC m<sup>-3</sup>) in the Beaufort Sea estimated from remotely sensed ocean color data (left panels) and by the 387 biogeochemical model (middle panels) for June, July, August and September. The Mackenzie Bay 388 389 (MB), Mackenzie delta (MD) and Cape Bathurst (CB) cited in the text are shown on the left panels. 390 The isolines of tDOC concentration are overlaid (black full lines). In the middle panels, simulated surface currents (m s<sup>-1</sup>) are overlaid. The two straight lines in the upper-middle panel refer to 391 392 transects along which surface tDOC fluxes were computed. The right panels show the model over satellite tDOC data ratio with the 200 m and 500 m isobaths overlaid. 393



**Figure 2.** Taylor diagram displaying a statistical comparison between the simulated and remotely sensed tDOC concentrations. The x-axis and y-axis show the model standard deviation relative to the satellite standard deviation. The open circle on the x-axis represents the reference point. The model-satellite correlation is represented in polar coordinates (angle from the x-axis). The light grey full lines indicate the RMSE relative to the satellite standard deviation.



Figure 3. Monthly flux of surface tDOC (TgC month<sup>-1</sup>) computed along transects located west of
the Mackenzie Trough (139°W ; 69.5°N-71°N) (upper panel) and at Cape Bathurst (128°W ;
69.5°N-71°N) (lower panel). Transects are shown in figure 1 in the upper-middle panel. Negative
values indicate a westward flux. Percentages refer to the model data relative to the satellite data.
The seasonal flux refers to the 4-month net flux.

**Table 1.** Skill metrics of comparison computed based on the 2003-2011 monthly climatologies of

409 tDOC.

Metric	June	July	August	September			
Correlation coefficient	0.79	0.82	0.78	0.79			
Unbiased RMSE (mmolC m <sup>-3</sup> )	41.4	29.4	26.0	29.3			
Model efficiency	0.49	0.60	0.26	0.38			
Geometric statistics using log-transformed data							
Model bias	1.24	1.07	1.32	1.21			
RMSE	1.07	1.02	1.12	1.06			

- **Table 2.** Areal stock (TgC) of sea surface tDOC computed over the Mackenzie shelf (delimited by
- 414 the 300 m isobaths) from the model and satellite data. The bias (%) refers to the model data relative
- 415 to the satellite data. The seasonal areal stock refers to the 4-month average  $\pm$  standard deviation.

	June	July	August	September	Seasonal
Model	1.48	1.40	1.32	1.28	$1.37 \pm 0.07$
Satellite	1.51	1.44	1.22	1.30	1.37±0.11
Bias	-2	-2.8	+8.2	-1.5	0

### 418 **References**

- Aagaard, K. and Carmack, E. C.: The role of sea ice and other fresh water in the Arctic circulation,
  J. Geophys. Res., 94, doi:10.1029/JC094iC10p14485. Issn: 0148-0227, 1989.
- 421 Abbott, B. W., Jones, J. B., Schuur, E. A. G., Chapin, F. S., III, Bowden, W. B., Bret-Harte, M. S.,
- 422 et al.: Biomass offsets little or none of permafrost carbon release from soils, streams, and
- wildfire: an expert assessment, Environmental Discussion paper Research Letters, 11(3),
  034014–14, doi:10.1088/1748-9326/11/3/034014, 2016.
- Bélanger, S., Xie, H., Krotkov, N., Larouche, P., Vincent, W. F., and Babin, M.:
  Photomineralization of terrigenous dissolved organic matter in Arctic coastal waters from 1979
  to 2003: Interannual variability and implications of climate change, Global Biogeochem.
  Cycles, 20, GB4005, doi:10.1029/2006GB002708, 2006.
- Bélanger S., Babin, M., and Tremblay, J.-E.: Increasing cloudiness in Arctic damps the increase in
  phytoplankton primary production due to sea ice receding, Biogeosciences, 10, 4087–4101,
  doi:10.5194/bg-10-4087-2013, 2013.
- Bell, L. E., Bluhm, B. A., and Iken, K.: Influence of terrestrial organic matter in marine food webs
  of the Beaufort Sea shelf and slope, Mar. Ecolo. Prog. Ser., 550, 1–24, doi:10.3354/meps11725,
  2016.
- Bring, A., Fedorova, I., Dibike, Y., Hinzman, L., Mård, J., Mernild, S. H., Prowse, T., Semenova,
  O., Stuefer, S. L., and Woo M.-K.: Arctic terrestrial hydrology: A synthesis of processes,
  regional effects, and research challenges, J. Geophys. Res. Biogeosci., 121, 621–649,
  doi:10.1002/2015JG003131, 2016.
- Buesseler K. O.: The decoupling of production and particulate export in the surface ocean, Global
  Biogeochem. Cycles, 12, 297–310, 1998.
- 441 Condron, A., Winsor, P., Hill, C. N., and Menemenlis, D.: Response of Arctic freshwater budget to
  442 extreme NAO forcing, J. Clim., 22, 2422–2437, 2009.

- 443 Davies, K. F.: Mackenzie River input to the Beaufort Sea, Beaufort Sea Project, Technical Report
  444 15, Institute of Ocean Sciences, Sidney, British, Columbia, 72 p, 1975.
- Dittmar, T., and Kattner, G.: The biogeochemistry of the river and shelf ecosystem of the Arctic
  Ocean: a review, Mar. Chem., 83, 103–120, doi:10.1016/S0304-4203(03)00105-1, 2003.
- Doney, S.C., Lima, I., Moore, J. K., and Takahashi, T.: Skill metrics for confronting global upper
  ocean ecosytem-biogeochemistry models against field and remote sensing data, J. Mar. Syst.,
  76, 95–112, doi:10.1016/j.jmarsys.2008.05.015, 2009.
- Doxaran, D., Devred, E., and Babin, M.: A 50 % increase in the mass of terrestrial particles
  delivered by the Mackenzie River into the Beaufort Sea (Canadian Arctic Ocean) over the last
  10 years, Biogeosci., 12, 3551–3565, doi:10.5194/bg-12-3551-2015, 2015.
- Dunton, K. H., Weingartner, T., and Carmack, E. C.: The nearshore western Beaufort Sea
  ecosystem: circulation and importance of terrestrial carbon in arctic coastal food webs, Progr.
  Oceanogr., 71, 362–378, doi:10.1016/j.pocean.2006.09.011, 2006.
- 456 Emmerton, C. A., Lesack, L. F. W., and Vincent, W. F.: Nutrient and organic matter patterns across
- the Mackenzie River, estuary and shelf during the seasonal recession of sea-ice, J. Marine Syst.,
  74, 741–755, doi:10.1016/j.jmarsys.2007.10.001, 2008.
- Griffin, C. G., Frey, K. E., Rogan, J., and Holmes, R. M.: Spatial and interannual variability of
  dissolved organic matter in the Kolyma River, East Siberia, observed using satellite imagery, J.
  Geophys. Res., 116, G03018, doi:10.1029/2010JG001634, 2011.
- 462 Haine, T. W. N., Curry, B., Gerdes, R., Hansen, E., Karcher, M., Lee, C., Rudels, B., Spreen, G., de
- 463 Steur, L., Stewart, K. D., and Woodgate R.: Arctic freshwater export: Status, mechanisms, and
- 464 prospects, Global and Planetary Change, 125, 13–35,doi:/10.1016/j.gloplacha.2014.11.013,
  465 2015.

- Herrault, P.-A., Gandois, L., Gascoin, S., Tananaev, N., Le Dantec, T., and Teisserenc, R.: Using
  high spatio-temporal optical remote sensing to monitor dissolved organic carbon in the Arctic
  river Yenisei, Remote Sens., 8, 803, doi:10.3390/rs8100803, 2016.
- 469 Holmes, R. M., McClelland, J. W., Peterson, B. J., Tank, S. E., Bulygina, E., Eglinton, T. I.,
- 470 Gordeev, V. V., Gurtovaya, T. Y., Raymond, P. A., Repeta, D. J., Staples, R., Striegl, R. G.,
- 471 Zhulidov, A. V., and Zimov, S. A.: Seasonal and annual fluxes of nutrients and organic matter
- 472 from large rivers to the Arctic Ocean and surrounding seas, Estuar. Coasts, 35, 369–382,
  473 doi:10.1007/s12237-011-9386-6, 2012.
- Holmes, R. M., Shiklomanov, A. I., Tank, S. E., McClelland, J. W., and Tretiakov, M.: River
  discharge, Arctic Report Card: Update for 2015, <u>http://www.arctic.noaa.gov/Report-</u>
  Card/Report-Card-2015/ArtMID/5037/ArticleID/227/River-Discharge, 2015.
- Kicklighter, D. W., Hayes, D. J., MacClelland, J. W., Peterson, B. J., McGuire, A. D., and Melillo,
  J. M.: Insights and issues with simulating terrestrial DOC loading of Arctic river networks,
  Ecol. App., 23, 1817–1836, doi:10.1890/11-1050.1, 2013.
- 480 Kritzberg, E., Duarte, C. M., and Wassmann, P.: Changes in Arctic marine bacterial carbon
  481 metabolism in response to increasing temperature, Polar Biol., 33, 1673–1682,
  482 doi:10.1007/s00300-010-0799-7, 2010.
- Lammers, R. B., Shiklomanov, A. I., Vörösmarty, C. J., Fekete, B. M., and Peterson, B. J.:
  Assessment of contemporary Arctic river runoff based on observational discharge records, J.
  Geophys. Res., 106(D4), 3321–3334, doi:10.1029/2000JD900444, 2001.
- 486 Lee, Y. J., Matrai, P. A., Friedrichs, M. A. M., Saba, V. S., Aumont, O., Babin, M., Buitenhuis, E.
- 487 T., Chevallier, M., de Mora, L., Dessert, M., Dunne, J. P., Ellingsen, I., Feldman, D., Frouin, R.,
- 488 Gehlen, M., Gorgues, T., Ilyina, T., Jin, M., John, J. G., Lawrence, J., Manizza, M., Menkes, C.
- 489 E., Perruche, C., Le Fouest, V., Popova, E., Romanou, A., Samuelsen, A., Schwinger, J.,
- 490 Séférian, R., Stock, C. A., Tjiputra, J., Tremblay, B. L., Ueyoshi, K., Vichi, M., Yool, A., and

- 491 Zhang, J.: Net primary productivity estimates and environmental variables in the Arctic Ocean:
- 492 An assessment of coupled physical-biogeochemical models, J. Geophys. Res.,
  493 doi:10.1002/2016JC011993, 2016.
- Le Fouest V., Babin, M., and Tremblay, J.-E.: The fate of riverine nutrients on Arctic shelves,
  Biogeosciences, 10, 3661–3677, doi:10.5194/bg-10-3661-2013, 2013.
- Le Fouest, V., Manizza, M., Tremblay, B., and Babin, M.: Modeling the impact of riverine DON
  removal by marine bacterioplankton on primary production in the Arctic Ocean,
  Biogeosciences, 12, 3385–3402, doi:10.5194/bg-12-3385-2015, 2015.
- 499 Losch, M., Menemenlis, D., Campin, J.-M., Heimbach, P., and Hill, C.: On the formulation of sea-
- ice models. Part 1: Effects of different solver implementations and parameterizations, Ocean
  Model., 33, 129–144, doi:10.1016/j.ocemod.2009.12.008, 2010.
- 502 McClelland, J. W., Holmes, R. M., Dunton, K. H., and Macdonald, R. W.: The Arctic Ocean 503 estuary, Estuar. Coast., 35, 353–368,doi:10.1007/s12237-010-9357-3, 2012.
- McGuire A. D., Anderson, L. G., Christensen, T. R., Dallimore, S., Guo, L., Hayes, D. J., Heimann,
  M., Lorenson, T. D., Macdonald, R.W., and Roulet, N.: Sensitivity of the carbon cycle in the
- 506 Arctic to climate change, Ecol. Monogr., 79, 523–555, doi:10.1890/08-2025.1, 2009.
- Macdonald, R. W., Paton, D. W., Carmack, E. C., and Omstedt, A.: The freshwater budget and
  under-ice spreading of Mackenzie River water in the Canadian Beaufort Sea based on salinity
  and 18 O / 16 O measurements in water and ice, J. Geophys. Res., 100, 895–919, 1995, 1995.
- Manizza, M., Follows, M. J., Dutkiewicz, S., McClelland, J. W., Menemenlis, D., Hill, C. N.,
  Townsend-Small, A., and Peterson, B. J.: Modeling transport and fate of riverine dissolved
  organic carbon in the Arctic Ocean, Global Biogeochem. Cycles, 23, GB4006,
  doi:10.1029/2008GB003396, 2009.

- 514 Manizza, M., Follows, M., Dutkiewicz, S., Menemenlis, D., McClelland, J. W., Hill, C. N.,
- 515 Peterson, B. J., and Key, R. M.: A model of the Arctic Ocean carbon cycle, J. Geophys. Res.,
  516 116(C12), C12020, doi:10.1029/2011JC006998, 2011.
- 517 Manizza, M., Follows, M. J., Dutkiewicz, S., Menemenlis, D., Hill, C. N., and Key, R. M.: Changes
- 518 in the Arctic Ocean  $CO_2$  sink (1996–2007): A regional model analysis, Global Biogeochem.
- 519 Cycles, 27, 1108–1118, doi:10.1002/2012GB004491, 2013.
- 520 Mann, P. J., Spencer, R. G. M., Hernes, P. J., Six, J., Aiken, G. R., Tank, S. E., McClelland, J. W.,
- Butler, K. D., Dyda, R. Y., and Holmes, R. M.: Pan-Arctic Trends in Terrestrial Dissolved
  Organic Matter from Optical Measurements, Front. Earth Sci., 4, 25, 10.3389/feart.2016.00025,
  2016.
- Massonnet, F., Fichefet, T., and Goosse, H.: Prospects for improved seasonal Arctic sea ice
  predictions from multivariate data assimilation, Ocean Model., 88, 16–25,
  doi:10.1016/j.ocemod.2014.12.013, 2015.
- Matsuoka, A., Hill, V., Huot, Y., Babin, M., and Bricaud, A.: Seasonal variability in the light 527 528 absorption properties of Western Arctic waters: parameterization of the individual components 529 absorption for ocean color applications, J. Geophys. Res., 116. C02007, of 530 doi:10.1029/2009JC005594, 2011.
- Matsuoka, A., Bricaud, A., Benner, R., Para, J., Sempéré, R., Prieur, L., Bélanger, S., and Babin, M.:
  Tracing the transport of colored dissolved organic matter in water masses of the Southern
  Beaufort Sea: relationship with hydrographic characteristics, Biogeosciences, 9, 925–940,
  doi :10.5194/bg-9-925-2012, 2012.
- Matsuoka, A., Hooker, S. B., Bricaud, A., Gentili, B., and Babin, M.: Estimating absorption
  coefficients of colored dissolved organic matter (CDOM) using a semi-analytical algorithm for
  southern Beaufort Sea waters : application to deriving concentrations of dissolved organic
  carbon from space, Biogeosci., 10, 917–927, doi :10.5194/bg-10-917-2013, 2013.

- Matsuoka, A., Babin, M., Doxaran, D., Hooker, S. B., Mitchell, B. G., Bélanger, S., and Bricaud, A.:
  A synthesis of light absorption properties of the Pan-Arctic Ocean: application to semianalytical estimates of dissolved organic carbon concentrations from space, Biogeosci., 11,
  3131–3147, doi:10.5194/bg-11-3131-2014, 2014.
- Matsuoka, A., Babin, M., and Devred, E. C.: A new algorithm for discriminating water sources
  from space: a case study for the southern Beaufort Sea using MODIS ocean color and SMOS
  salinity data, Remote Sens. Env., 184, 124–138, http://dx.doi.org/10.1016/j.rse.2016.05.006,
  2016.
- Matsuoka, A., Boss, E., Babin, M., Karp-Boss, L., Hafez, M., Chekalyuk, A., Proctor, C. W.,
  Werdell, P. J., and Bricaud, A.: Pan-Arctic optical characteristics of colored dissolved organic
  matter: Tracing dissolved organic carbon in changing Arctic waters using satellite ocean color
  data, Remote Sens. Env., 200, 89–101, doi:10.1016/j.rse.2017.08.009, 2017.
- Menemenlis, D., Hill, C., Adcroft, A., Campin, J.-M., Cheng, B., Ciotti, B., Fukumori, I., Heimbach,
  P., Henze, C., Kohl, A., Lee, T., Stammer, D., Taft, J., and Zhang, J.: NASA supercomputer
  improves prospects for ocean climate research, Eos Trans. AGU, 86, 89–96, 2005.
- Mulligan, R. P., Perrie, W., and Solomon, S.: Dynamics of the Mackenzie River plume on the inner
  Beaufort shelf during an open water period in summer, Estuar. Coast Shelf Sci., 89, 214–220,
  doi:10.1016/j.ecss.2010.06.010, 2010.
- Nash, J., and Sutcliffe, J.: River flow forecasting through conceptual models, part 1 a discussion
  of principles, J. Hydrol., 10, 282–290, 1970.
- Nguyen, A. T., Menemenlis, D., and Kwok, R.: Improved modeling of the Arctic halocline with a
  subgrid-scale brine rejection parameterization, J. Geophys. Res., 114, C11014,
  doi:10.1029/2008JC005121, 2009.

- Nguyen, A. T., Menemenlis, D., and Kwok, R.: Arctic ice-ocean simulation with optimized model
  parameters: Approach and assessment, J. Geophys. Res., 116(C4), C04025,
  doi:10.1029/2010JC006573, 2011.
- Onogi, K., Tsutsui, J., Koide, H., Sakamoto, M., Kobayashi, S., Hatsushika, H., Matsumoto, T.,
  Yamazaki, N., Kamahori, H., Takahashi, K., Kadokura, S., Wada, K., Kato, K., Oyama, R.,
  Ose, T., Mannoji, N., and Taira, R.: The JRA-25 Reanalysis, J. Meteor. Soc. Japan, 85, 369–
- 568 432, doi:10.2151/jmsj.85.369, 2007.
- Opsahl, S., Benner, R., and Amon, R. M.: Major flux of terrigenous dissolved organic matter
  through the Arctic Ocean, Limnol. Oceanogr., 44, 2017–2023, doi:10.4319/lo.1999.44.8.2017,
  1999.
- Ortega-Retuerta, E., Jeffrey, W. F., Babin, M., Bélanger, S., Benner, R., Marie, D., Matsuoka, A.,
  Raimbault, P., and Joux, F.: Carbon fluxes in the Canadian Arctic: patterns and drivers of
  bacterial abundance, production and respiration on the Beaufort Sea margin, Biogeosciences, 9,
  3679–3692, doi:10.5194/bg-9-3679-2012, 2012.
- Osburn, C. L., Retamal, L., and Vincent, W. F.: Photoreactivity of chromophoric dissolved organic
  matter transported by the Mackenzie River to the Beaufort Sea, Mar. Chem., 115,
  doi:10.1016/j.marchem.2009.05.003, 2009.
- Para, J., Charrière, B., Matsuoka, A., Miller, W. L., Rontani, J. F., and Sempéré, R.: UV/PAR
  radiation and DOM properties in surface coastal waters of the Canadian shelf of the Beaufort
  Sea during summer 2009, Biogeosciences, 10, 2761-2774, doi:10.5194/bg-10-2761-2013, 2013.
- 582 Ping, C.-L., Michaelson, G. J., Guo, L., Torre Jorgenson, M., Kanevskiy, M., Shur, Y., Dou, F., and
- Liang, J.: Soil carbon and material fluxes across the eroding Alaska Beaufort Sea coastline, J.
  Geophys. Res., 116, G02004, doi:10.1029/2010JG001588, 2011.
- 585 Rachold, V., Eiken, H., Gordeev, V. V., Grigoriev, M. N., Hubberten, H.-W., Lisitzin, A. P.,
- 586 Shevchenko, V. P., and Schirmeister, L.: Modern terrigenous organic carbon input to the Arctic

- 587 Ocean, in The Organic Carbon Cycle in the Arctic Ocean, edited by R. S. Stein and R. W.
  588 Macdonald, 33–55, Springer, New York, 2004.
- 589 Rawlins, M. A., Steele, M., Holland, M., Adam, J., Cherry, J., Francis, J., Groisman, P., Hinzman,
- 590 L., Huntington, T., Kane, D., Kimball, J., Kwok, R., Lammers, R., Lee, C., Lettenmaier, D.,
- 591 McDonald, K., Podest, E., Pundsack, J., Rudels, B., Serreze, M., Shiklomanov, A., Skagseth,
- 592 O., Troy, T., Vorosmarty, C., Wensnahan, M., Wood, E., Woodgate, R., Yang, D., Zhang, K.,
- and Zhang, T.: Analysis of the Arctic System for Freshwater Cycle Intensification:
  Observations and Expectations, J. Climate, 23, 5715–5737, doi:10.1175/2010JCLI3421.1, 2010.
- Raymond, P. A., McClelland, J. W., Holmes, R. M., Zhulidov, A. V., Mull, K., Peterson, B. J.,
  Striegl, R. G., Aiken, G. R., and Gurtovaya, T. Y.: Flux and age of dissolved organic carbon
  exported to the Arctic Ocean: A carbon isotopic study of the five largest arctic rivers, Global
  Biogeochem. Cycles, 21, GB4011, doi:10.1029/2007GB002934, 2007.
- Romanovsky, V. E., Drozdov, D. S., Oberman , N. G., Malkova, G. V., Kholodov, A. L.,
  Marchenko, S. S., Moskalenko, N. G., Sergeev, D. O., Ukraintseva, N. G., Abramov, A. A.,
  Gilichinsky, D. A., and Vasiliev, A. A.: Thermal State of Permafrost in Russia, Permafrost and
  Periglacial Process., 21, 136–155. doi:10.1002/ppp.683, 2010.
- Semiletov, I., Pipko, I., Gustafsson, Ö., Anderson, L. G., Sergienko, V., Pugach, S., Dudarev, O.,
  Charkin, A., Gukov, A., Bröder, L., Andersson, A., Spivak, E., and Shakhova, N.: Acidification
  of East Siberian Arctic Shelf waters through addition of freshwater and terrestrial carbon,
  Nature Geosci., 9, 361–365, doi:10.1038/ngeo2695, 2016.
- 607 Serreze, M., Barret, A. P., Slater, A. G., Woodgate, R. A., Aagard, K., Lammers, R. B., Steele, M.,
- 608 Mortitz, R., Meredith, M., and Lee, C. M.: The large-scale fresh water cycle of the Arctic, J.
- 609 Geophys. Res., 111, C11010, doi:10.1029/2005JC003424, 2006.

- 610 Shiklomanov, I., Shiklomanov, A., Lammers, R., Peterson, B., and Vorosmarty, C.: The dynamics
- of river water inflow to the Arctic Ocean, in The Freshwater Budget of the Arctic Ocean, edited
  by E. Lewis, 281–296, Kluwer Acad., Boston, Mass, 2000.
- 613 Simon, E., Samuelsen, A., Bertino, L., and Mouysset, S.: Experiences in multiyear combined state-
- 614 parameter estimation with an ecosystem model of the North Atlantic and Arctic Oceans using
- 615 the Ensemble Kalman Filter, J. Mar. Syst., 152, 1–17, doi:10.1016/j.jmarsys.2015.07.004.
- 616 Stein, R., and Macdonald, R. W.: The Organic Carbon Cycle in the Arctic Ocean, Springer,
  617 Heidelberg, Germany, 2015.
- 618 Stow, A. C., Jolliff, J., McGillicuddy Jr., D. J., Doney, S. C., Allen, J. I., Friedrichs, M. A. M., Rose,
- K. A., and Wallhead, P.: Skill Assessment for coupled biological/physical models of marine
  systems, J. Mar. Syst., 76, 4–15, doi:10.1016/j.jmarsys.2008.03.011, 2009.
- Striegl, R. G., Aiken, G. R., Dornblaser, M. M., Raymond, P. A., and Wickland, K. P.: A decrease
  in discharge-normalized DOC export by the Yukon River during summer through autumn,
  Geophysical Research Letters, 32(21), L21413, doi:10.1029/2005GL024413, 2005.
- Tank, S. E., Manizza, M., Holmes, R. M., McClelland, J. W., and Peterson, B. J.: The Processing
  and Impact of Dissolved Riverine Nitrogen in the Arctic Ocean, Estuar. Coasts, 35,
  doi:10.1007/s12237-011-9417-3, 2012.
- Tank, S. E., Striegl, R. G., McClelland, J. W., and Kokelj, S. V: Multi-decadal increases in
  dissolved organic carbon and alkalinity flux from the Mackenzie drainage basin to the Arctic
  Ocean, Environ. Res. Lett., 11, 054015, doi:10.1088/1748-9326/11/5/054015, 2016.
- Tanski, G., Couture, N., Lantuit, H., Eulenburg, A., and Fritz, M.: Eroding permafrost coasts release
  low amount of dissolved organic carbon from ground ice into the nearshore zone of the Arctic
  Ocean, Global Biogeochem. Cycles, 30, 1054–1068, doi:10.1002/2015GB005337, 2016.

- Tarnocai, C., Canadell, J. G., Schuur, E. A. G., Kuhry, P., Mazhitova, G., and Zimov, S.: Soil
  organic carbon pools in the northern circumpolar permafrost region, Global Biogeochem.
  Cycles, 23, GB2023, doi:10.1029/2008gb003327, 2009.
- Taylor, K. E.: Summarizing multiple aspects of model performance in a single diagram, J. Geophys.
  Res., 106, 7183–7192, 2001.
- Timmermans, M.-L.: Sea Surface Temperature, Arctic Report Card: Update for 2016,
   <u>http://www.arctic.noaa.gov/Report-Card/Report-Card-2016/ArtMID/5022/ArticleID/285/Sea-</u>
   <u>Surface-Temperature</u>, 2016.
- Vallières, C., Retamal, L., Osburn, C., and Vincent, W. F.: Bacterial production and microbial food
  web structure in a large Arctic river and the coastal Arctic Ocean, J. Mar. Syst., 74, 756–773,
  doi:10.1016/j.jmarsys.2007.12.002, 2008.
- Vaquer-Sunyer, R., Duarte, C. M., Santiago, E., Wassmann, P., and Reigstad, M.: Experimental
  evaluation of planktonic respiration response to warming in the European Arctic sector, Polar
  Biol., 33, 1661–1671, doi:10.1007/s00300-010-0788-x, 2010.
- Vihma, T., Screen, J., Tjernström, M., Newton, B., Zhang, X., Popova, V., Deser, C., Holland, M.,
  and Prowse, T.: The atmospheric role in the Arctic water cycle: A review on processes, past
  and future changes, and their impacts, J. Geophys. Res. Biogeosci., 121, 586–620,
  doi:10.1002/2015JG003132, 2016.
- Wang, M. and Shi, W.: The NIR-SWIR combined atmospheric correction approach for MODIS
  ocean color data processing, Opt. Express, 15, 15722–15722, 2007.
- Yang, D., Shi, X., and Marsh, P.: Variability and extreme of Mackenzie River daily discharge
  during 1973-2011, Quatern. Internat., 380–381, 159–168, doi:10.1016/j.quaint.2014.09.023,
  2015.
- 656