MODIS observed increase in duration and spatial extent of sediment plumes in Greenland fjords

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Abstract

We test the hypothesis that increased meltwater runoff from the Greenland Ice Sheet (GrIS) has elevated the suspended sediment concentration (SSC) of six river plumes in three fjords in southwest Greenland. A SSC retrieval algorithm was developed using the largest in situ SSC dataset for Greenland known and applied to all cloud-free NASA Moderate Resolution Imaging Spectrometer (MODIS) reflectance values in the Terra image archive (2000 to 2012).

Melt-season mean plume SSC has not increased as anticipated, with the exception of one river. However, positive statistically significant trends involving metrics that described the duration and the spatial extent of river plumes were observed in many locations. Zones of sediment concentration > 50 mgL\(^{-1}\) expanded in three river plumes, with potential consequences for biological productivity. The high SSC cores of river plumes (> 250 mgL\(^{-1}\)) expanded in one-third of study locations. When data from study rivers was aggregated, higher volumes of runoff were associated with higher melt-season mean plume SSC values, but this relationship did not hold for individual rivers. High spatial variability between proximal plumes highlights the complex processes operating in Greenland’s glacio-fluvial-fjord systems.

1 Introduction

The Greenland Ice Sheet (GrIS) is the largest landbound ice mass in the Northern Hemisphere (Serreze and Barry, 2005). It stores the equivalent of 7 m of global sea level (Bamber et al., 2013). Freshwater discharge from the GrIS is an important contributor to the Arctic hydrological cycle (Mernild et al., 2011), though few multi-year runoff observations exist (Cowton et al., 2012; Hasholt et al., 2013; Rennermalm et al., 2013).

The GrIS has lost approximately 142 ± 49 Gtyr\(^{-1}\) of mass between 1992 and 2011 (Shepherd et al., 2012) or ∼ 0.005 % of its total mass per year assuming a volume of...
2.96 × 10^6 km^3 (Bamber et al., 2013), at a density of 917 kg m^-3. This mass loss has accelerated since the 1990s. Using reconciled mass balance estimates from multiple sources, Shepherd et al. (2012) estimated that the GrIS lost 51 ± 65 Gtyr^-1 from 1992 to 2000 and 211 ± 37 Gtyr^-1 from 2000 to 2011.

The GrIS has two different types of outlets where most mass is lost: (1) approximately two-thirds of the mass is lost at marine terminating outlets (e.g. calving of ice); (2) the GrIS loses the remaining one-third of loss via liquid runoff at land terminating outlets of the GrIS using the mean value estimates for 1991 to 2010 (Bamber et al., 2012). This paper focuses on rivers fed by the latter. Meltwater production on the GrIS is thought to have increased due to expansion of the ablation zone (McGrath et al., 2013). Though no direct runoff estimates exist for the GrIS (Shepherd et al., 2012), modeling studies suggest that ice sheet runoff has increased 3% annually (1990 through 2007, Ettema et al., 2009).

Rivers convey not only runoff but also sediment from the GrIS to the ocean. The relationship between these two constituents is largely unstudied in Greenlandic rivers, but there often exists a power law relationship between the suspended sediment concentration of a river, \( C_s \), and its discharge, \( Q \): \( C_s = aQ^b \) (Morehead et al., 2003; Syvitski et al., 2000): \( a \) and \( b \) are rating parameters. This relationship is often valid for braided rivers with glacierized catchments. Church (1972) found this relationship for the braided Lewis River (90% glacierized catchment including the Barnes Ice Cap) and the Upper South River, Ekalugad Fjord (50% glacierized catchment on Baffin Island). In contrast Østrem (1975) found no simple relationship between melt from Norwegian glaciers and their suspended sediment load. Fenn et al. (1985) noted that the relationship between sediment concentration and water discharge is not simple and that relationships only including \( Q \) and \( C_s \) terms exclude important explanatory variables. O’Farrell et al. (2009) found for the Matanuska Glacier, Alaska large variability between annual discharge and suspended sediment flux. In southwest Greenland, high discharge years led to increased sediment load in the Watson River (Hasholt et al., 2013). Chu et al. (2012) found that an approximation for melt water production on the GrIS, Polar MM5 modeled
positive degree-days, was the most significant driver of regional suspended sediment concentration variability in fjords.

We hypothesize that increased freshwater discharge due to increased ice sheet melt has elevated the melt-season suspended sediment concentration of Greenlandic fjords. While we recognize that the suspended sediment concentration (SSC) of Greenland’s fjords are an understudied problem, it offers the potential to be an important indicator of environmental change. The sediment exported by a glacier is a function of the efficiency of its subglacial drainage network, its sediment production rate, and its sediment storage ability (Fausto et al., 2012).

Fjord SSC is not only diagnostic of subglacial sediment dynamics but also of fjord health. Suspended sediment impacts the availability of light (Retamal et al., 2008). High levels of SSC in fjords may impact biological productivity, such as pelagic zooplankton feeding efficiency and production rates (Arendt et al., 2011). However, sediment delivers nutrients to the coastal ocean at the same time (Statham et al., 2008).

Widespread in situ observations of SSC in Greenland are logistically infeasible; in fact Hasholt (1996) and Hasholt et al. (2006) identified < 15 locations where sediment dynamics had been studied, even for a short period of time. Instead, recent efforts have focused on using orbital remote sensing to monitor GrIS sediment dynamics (Chu et al., 2009, 2012; McGrath et al., 2010; Tedstone and Arnold, 2012). A goal of this paper is to provide an improved SSC retrieval algorithm, based on in situ observations, for future investigations and to better characterize the variability of SSC in fjords in southwest Greenland.

2 Study areas and approach

Six proglacial river systems draining into three major fjords in southwest Greenland were selected: Pakitsuup, Kangerlussuaq, and Ameralik fjords (Fig. 1). To determine the portions of the GrIS that contribute melt to each river we calculated the local hydrostatic pressure field of the ice sheet, which combines surface elevation data with
basal topography (Lewis and Smith, 2009; Cuffey and Patterson, 2010). We used surface elevation of the Greenland Mapping Project (GIMP) digital elevation model (Howat et al., 2013) and the University of Kansas Center for the Remote Sensing of Ice Sheets (CRESIS) Jacobshavn composite dataset for the Pakitsuup region (https://data.cresis.ku.edu/data/grids/) and the coarser basal topography from Bamber et al. (2001) for the Kangerlussuaq and Ameralik regions. Flow routing and catchment delineation was performed following a D-infinity approach using RiverTools 3.0 (Peckham, 2009). Catchment areas describe only the portion of the watershed covered by glacial ice, as these areas are the major sources of water and sediment to the systems studied.

2.1 Pakitsuup Fjord

Pakitsuup Fjord (also spelled Pakitsoq) is located near the town of Ilulissat, and receives melt from two small river systems. The “Pakitsuup North” River (69°31′45.5″ N, 50°13′23.7″ W) has two branches, respectively 8 and 10 km long. It drains a catchment of the GrIS ∼302 km² (Table 1), which reaches a maximum height of ∼1169 meters above sea level (ma.s.l.). The “Pakitsuup South” River (69°26′5.9″ N, 50°27′53.3″ W) flows 7 km from the margin of the GrIS that feeds it along its southern tributary. A northern tributary flows 20 km through a series of lakes, before combining with the southern tributary approximately 4 km from the river mouth. It is the smallest catchment studied (∼143 km²) with the lowest maximum elevation (∼1020 ma.s.l.). Mernild et al. (2010) estimate that the equilibrium line altitude (ELA) for the region is approximately 1125 ma.s.l.

2.2 Kangerlussuaq Fjord

Three major river systems discharge water and sediment into Kangerlussuaq Fjord (Søndre Strømfjord).
The Watson River (66°57’53.7’’ N, 50°51’49.8’’ W) enters at Kangerlussuaq Fjord's northeast head. A discharge gauging station has operated on the river since 2007 at the town of Kangerlussuaq (asterisk on Fig. 1), 10 km from the river mouth (Hasholt et al., 2013). The Watson River receives water from two tributaries, each flowing ~34 km, and has a combined GrIS catchment area of 3639 km² on the GrIS. This catchment has a maximum elevation of 1860 m a.s.l. (Table 1). Van de Wal et al. (2012) estimated that the 21 yr mean ELA for this region of the GrIS is 1553 m a.s.l.

The Umiiviit River (66°50’1.6’’ N, 50°48’37.3’’ W) flows into the southeastern head of Kangerlussuaq Fjord and drains a catchment area of 6320 km² with a maximum elevation of ~2390 m a.s.l. In comparison, the Umiiviit River is >85 km in length, more than double that of the Watson River branches.

The Sarfartoq River (66°29’29.6’’ N, 52°1’29.5’’ W) discharges into Kangerlussuaq Fjord ~79 km down fjord from the Watson River. From its mouth, the river stretches 53 km in a narrow valley to a lake, Tasersiaq (“T” on Fig. 1). Tasersiaq is ~71 km long and 14 m deep (Goldthwait et al., 1964), and bordered by the Sukkertoppen Ice Cap, lobes of the GrIS, and an 18 km long lake, Tasersiaq Qalia (“TQ” on Fig. 1). The Sarfartoq River has 5385 km² of glaciated catchment with a maximum height of 2157 m a.s.l. Approximately 919 km² of this area is covered by minor ice caps and glaciers not part of the GrIS (e.g. the Sukkertoppen Ice Cap). It is the only river in the study that receives melt from both local glaciers and the GrIS.

2.3 Ameralik Fjord

Ameralik Fjord (also known as Lysefjord, or Ameralgda Fjord if considering only the innermost part of the fjord) is located close to Greenland's capitol city, Nuuk. The Naujat Kuat River (64°12’37.5’’ N, 50°12’31.0’’ W) flows 26 km from the margin of the GrIS, receiving melt from a catchment 460 km², with a maximum height of 1342 m a.s.l. Mote (2000) estimated the regional ELA near 1450 m a.s.l. Meltwater discharge originates from two lobes of the GrIS. The southern lobe discharges water into an 8 km long lake, Isortuarssuk (“I” on Fig. 1), a short distance from its terminus. The northern lobe,
Kangaussarssup Sermia ("KS" on Fig. 1), discharges water to the river uninterrupted by lakes. On the north side of Kangaussarssup Sermia two ice margin lakes exist. These lakes historically drained down a valley (the Austmannadalen) to a point approximately 7 km from the river mouth. Since 2004, the lakes drain north into Kangiata Nunaata Sermia ("KNS" on Fig. 1, Weidick and Citterio, 2011). As the MODIS record straddles this drainage realignment, Table 1 provides two delineations, one with the lakes (Naujat Kuat) and one without (Naujat Kuat – Small). At its smallest, the on-ice catchment area is 356 km$^2$. This uncertainty in drainage basin delineation does not affect the maximum elevation of the drainage or the general shape of the hypsometric curve.

3 Methods

3.1 Suspended sediment samples

We collected surface water samples as part of our oceanographic surveys in the three fjords during the 2008, 2010, 2011, and 2012 summer seasons (Table 2). In total, one hundred and forty three samples were utilized for ground-truthing work, which collectively covered 12 days in all major melt months. The SSC retrieval algorithm dataset also contained three additional water samples collected from Orpigsoq Fjord (68°38′30.9″ N, 50°54′45.5″ W) on 2 July 2011 though this paper did not analyze that fjord.

As Fig. 1 shows, sampling followed variably spaced transects. Water samples were collected by dipping a 1000 mL bottle into the surface layer while a handheld GPS noted the location. Samples were then vacuum filtered through pre-weighed Whatman GF/F filters, rinsed with distilled water to remove salt, freeze dried, and re-weighed to determine SSC. Suspended sediment concentrations ranged between 1.2 and 716 mgL$^{-1}$ with a mean SSC of 73 mgL$^{-1}$. One sample collected from the shallow river mouth of the Watson river on 9 July 2011 had an SSC of 1581 mgL$^{-1}$ but was excluded from
analysis because it could not be determined if the sampling boat had disturbed bottom sediments.

3.2 SSC retrieval algorithm based on MODIS reflectance

MODIS Terra MOD02 250 m imagery and the MOD03 geolocation data product for each swath image were downloaded from NASA’s Reverb Earth Science Data Discovery Tool (http://reverb.echo.nasa.gov/reverb/). Exelis ENVI 4.8 software, including the MODIS Conversion Toolkit (http://www.exelisvis.com) georeferenced, bow tie, and atmospherically corrected imagery (using dark object subtraction). Each scene covered a geographic area that included cloud-free open ocean pixels required for accurate dark object subtraction. (Kirk, 2011; Miller and McKee, 2004). A Danish National Survey and Cadastre (DNSC) Greenland coastline vector file confirmed the geopositional accuracy of each image. If needed, MODIS imagery was manually shifted to agree with the DNSC coastline.

SSC values were compared to co-located band one (620 to 670 nm) and two (841 to 876 nm) reflectance values. Barbieri et al. (1997) defined the MOD02 reflectance data product as the bidirectional reflectance factor of the Earth ($\rho_{ev}$) multiplied by cosine of the solar zenith angle at the earth scene ($\theta_{ev}$)

$$\rho_{ev} \times \cos \theta_{ev}.$$ (1)

With $\rho_{ev}$ defined as

$$\rho_{ev} = \frac{L_{ev}}{E_{i} \cos \theta_{ev}},$$ (2)

where $L_{ev}$ is the earth view spectral radiance (W m$^{-2}$ sr$^{-1}$ μm$^{-1}$), and $E_{i}$ is solar irradiance (W m$^{-2}$ μm$^{-1}$). Combing these two equations, reflectance is calculated as

$$R_{rs} = \frac{L_{ev}}{E_{i}}.$$ (3)
Ocean color remote sensing literature often define $R_{rs}$ as radiance reflectance (Kirk, 2011) or remote sensing reflectance (Mobley, 1999), though technically $E_i$ should be downwelling irradiance directly above the surface of the water ($E_d$), not from the top of atmosphere ($E_i$). Despite this difference, we use the $R_{rs}$ notation. Following Miller and McKee (2004), variations in scene solar zenith angle were corrected on an individual pixel basis to arrive at

$$R_{rsc} = R_{rs} \times \frac{1}{\cos \theta_{ev}}. \quad (4)$$

We develop a SSC retrieval algorithm using both MODIS band one and two and the co-located in situ sediment samples (Fig. 2). Unlike previous MODIS algorithms which became saturated above 80 mg L$^{-1}$ (e.g. Chu et al., 2009), the addition of band two, which measures reflectance in the near infrared, remains sensitive to higher sediment concentrations. The final retrieval algorithm used was

$$SSC = 1.80 \times \exp(19.11 \times (R_{rsc,B1} + R_{rsc,B2})). \quad (5)$$

The retrieval dataset included samples from four different fjords with variable salinity regimes. Using a Seabird 19 conductivity, temperature, and depth probe, we measured salinities as high as 25 PSU (Ameralik Fjord) and as low as 1 PSU (Kangerlussuaq Fjord) in the top meter of the surface layer where samples were taken. The variable salinity during our oceanographic surveys, and the independence of the relationship on this parameter, suggests that this retrieval technique can be applied to disparate locations with variable oceanographic conditions.

Following algorithm development, we processed ~300 sea ice-free swath scenes per year (spanning day of the year 135 to 295) to map fjord SSC values over the entire melt season. Appendix A describes the full processing details.
3.3 Error from sampling irregularities caused by clouds

Since the year 2000, MODIS Terra satellite observations provide an ongoing daily record in this region, although, as common in polar regions, the presence of clouds compromise the imagery’s utility as an observational tool. In total, clear-sky existed for between 5 and 21% of all available fjord pixels in a given summer season.

Computational experiments following Gregg and Casey (2007) assessed this sampling bias. A “truth” dataset based on the entire 13 yr data record was created and then masked with MODIS observed cloud cover to understand how cloud obscuration impacted SSC statistics.

To create the “truth” dataset we calculated the 13 yr mean SSC map on a pixel-by-pixel basis as

\[
13 \text{ year } SSC = \frac{\sum_{\text{year}=2000}^{2012} \sum_{\text{day}=135}^{295} SSC_{i,j}}{\sum_{\text{year}=2000}^{2012} N_{i,j}}. \tag{6}
\]

Where \( N \) was the number of times the pixel passed all tests and was used in the analysis, \( i \) is pixel location in the horizontal direction, and \( j \) is pixel location in the vertical direction. The cloud masking scheme set SSC values to equal not-a-number if a cloud was detected, and thus excluded cloud influenced reflectance values from the statistics.

We then used the 13 yr mean SSC map as a spatial template to mimic seasonal SSC fluctuations by multiplying it by a randomly perturbed sinusoidal factor, the scene intensity, \( I_s \) that peaked near day 190 of each melt-season simulation.

\[
SSC_{i,j} = 13 \text{ year } SSC_{i,j} \times I_s \tag{7}
\]
Where

\[ I_s = x \times \sin \left( \frac{y}{55} \right), \tag{8} \]

\( x \) was a number between two and four randomly generated for each scene, and \( y \) began at one and increased by one with each scene used.

This range was selected by tuning “truth” SSC values to the range observed by MODIS. As an example, Fig. 3 plots how mean SSC was allowed to vary seasonally during computational experiments.

Computational experiments explored cloud bias in a number of annual metrics (mean, median, etc.) by imposing the MODIS derived 13 yr record of cloud cover on the “truth” dataset (Fig. 4). In addition, these experiments evaluated how metrics calculating the percent of available fjord pixels in a year observed above a specified SSC threshold (50 or 250 \( \text{mg L}^{-1} \)) performed. These \( P_{>t} \) metrics describe the duration and the spatial extent of a river plume for a given region of interest (ROI). They were calculated with two steps. First, for each scene a threshold mask (STM) was created, where all pixels in a ROI greater than a given SSC threshold \( t \) were assigned a value of one and all others a value of zero. Second, STM values were summed for an entire melt season and corrected for pixels classified as fjord \( (F_p) \) and the total number of scenes in a year \( (S) \) with Eq. (9):

\[
P_{>t} = \frac{\sum_{\text{day}=135} \sum_i \sum_j \text{STM}_{i,j} F_p}{F_p \times S}. \tag{9} \]

Variable plume area and fjord shape caused each ROI to vary in size and fjord area. The Pakitsuup North ROI contained a fjord area of 4.437 \( \text{km}^2 \) and Pakitsuup South 4.25 \( \text{km}^2 \). Watson, Umiiviit, and Sarfartoq ROIs contained 41.375, 42.812, and 65.125 \( \text{km}^2 \) of fjord area respectively. The Naujat Kuat ROI contained a fjord area of 13 \( \text{km}^2 \). Tables 3 through 5 report ROI metrics normalized by each ROI 13 yr mean to account for this variability.
3.4 Selection of metrics

Three characteristic metrics were selected to describe plume dynamics and trends over the 13 yr record: melt-season mean SSC (SSC_{msm}), percent available fjord pixels in a melt season above 50 mg\(L^{-1}\) (\(P_{>50}\)), and above 250 mg\(L^{-1}\) (\(P_{>250}\)). The melt season was defined as day of the year 135 through day 295 (~15 May through 22 October). SSC_{msm} was selected because values were closest to “truth” values. \(P_{>50}\) was selected because it matches the threshold Arendt et al. (2011) used in a recent study near Nuuk in Godthåb fjord for significant impact on copepods and secondary production. \(P_{>250}\) was selected because SSC above 250 mg\(L^{-1}\) approximately maps the plume core or zone of flow establishment (Albertson et al., 1950). \(P_{>50}\) and \(P_{>250}\) were also the most consistent metrics between years.

SSC_{msm} deviated the least from the “truth” dataset. At most, it underestimated “truth” values by 16 % and overestimated by 38 %. On average it underestimated by 16 % and overestimated by 12 %. The maximum range of percent errors found for SSC_{msm} was 54 % (Naujat Kuat), with a mean of 36 %.

\(P_{>50}\) estimates were between 67 and 91 % below “truth” values. The maximum range in the deviation from the “truth” using \(P_{>50}\) for a plume was 22 % (Umiiviit), with an average range of 15 %. As Fig. 5d–f shows, \(P_{>250}\) estimates were similar, between 63 and 92 % below “truth” values. The maximum range in the deviation from the “truth” was 20 % (Sarfartoq), with an average range of 16 %. Cloud obscuration prevents \(P_{>t}\) metrics from correctly calculating the true number of times fjord pixels exceed a given value. Thus, absolute values reported may not be reliable. However, we argue this bias did not compromise the usefulness of the metric to detect trends.

We assigned error bars that symmetrically divided the mean range of variability found (SSC_{msm} ± 18 %, \(P_{>50}\) ± 7.5 %, \(P_{>250}\) ± 8 %). The large bias in \(P_{>t}\) values should also be remembered when interpreting results. We then added error associated with the SSC retrieval to bring overall error assigned to melt-season metrics to: ±19 % for SSC_{msm} and ±10 % for \(P_{>t}\) metrics.
3.5 RACMO2 runoff data

In Greenland, only the Watson River discharge record is longer than 5 yr (Hasholt et al., 2013). To extend our analysis, we used yearly catchment scale RACMO2 modeled ice sheet runoff volume (Bamber et al., 2012, hereafter referred to as RACMO2 runoff) as a substitute for observed discharge. Watson River observed discharge and RACMO2 runoff were correlated ($r^2 = 0.5$) for the 4 yr a comparison is possible, but RACMO2 runoff overestimates observed discharge values by 0.79 to 2.95 km$^3$ yr$^{-1}$. We did not use RACMO2 runoff for either the Sarfartoq catchment or the Umiiviit River. For the former, no geographically appropriate outlet could be identified and for the latter, the misfit between RACMO2 runoff values and in situ field discharge measurements was too great to be reliable.

4 Results and discussion

4.1 Trends in mean suspended sediment concentration

We hypothesize that increased freshwater discharge due to increased ice sheet melt has elevated the melt-season suspended sediment concentration of Greenlandic fjords. Overall, SSC$_{msm}$ has increased for most plumes, although this increase is only statistically significant for the Sarfartoq River plume.

The Sarfartoq River plume SSC$_{msm}$ increased $2.1 \pm 0.4$ mgL$^{-1}$ yr$^{-1}$ ($p = 0.013$). We offer two explanations for why only this river is statistically significant: (1) The Sarfartoq ice sheet catchment may be most sensitive to climate warming because it has the largest below ELA catchment area studied (3291 km$^2$), and hence large volumes of meltwater production are likely to occur, and are more likely to exit the catchment area, rather than be retained in a firn aquifer (Harper et al., 2012); (2) The Sarfartoq watershed also contains two long lakes, Tasersiaq and Tasersiaq Qalia, that may smooth the SSC signal. Tasersiaq’s length ($\sim 71$ km) suggests it has a high sediment trapping
efficiency for all but the finest particles (Brune, 1953) and that sediment particles that
do pass through the lake have a longer transit time than particles that are only trans-
ported in rivers. These two characteristics may have dampened the magnitude of SSC
variability year-to-year.

Further, the relatively short 13 yr record studied, and high interannual variability in
this natural system, may impede the detection of statistically significant trends. In ad-
dition, variable cloud obscuration makes establishment of statistically significant trends
difficult, though clouds obscured the Sarfartoq plume similarly.

Additionally, mass balance estimates (Shepherd et al., 2012; van den Broeke et al.,
2009; Rignot et al., 2008) for Greenland have shown that acceleration in GrIS mass
loss began in the 1990s. Resultantly, MODIS imagery only captured the later half of
the response to recent mass balance changes.

4.2 Trends in plume area metrics

Though SSC_{msm} largely displayed no statistically significant upward trend over the
MODIS record, \( P_{>t} \) metrics showed that both areas potentially detrimental to fjord biol-
ogy (\( P_{>50} \)) and the core of plumes (\( P_{>250} \)) have grown larger and/or more persistent.

\( P_{>50} \) and \( P_{>250} \) significantly increased for all river plumes except for the Pakitsuup
South River. Within a given plume ROI, \( P_{>50} \) increased between 0.08 and 0.25 % yr\(^{-1}\)
of its 13 yr mean and \( P_{>250} \) increased between 0.05 and 0.08 % yr\(^{-1}\) of its 13 yr mean.
Expressed as the sum of the area of pixels exceeding these thresholds for all images
in a melt-season (where 1 pixel = 0.0625 km\(^2\)) \( P_{>50} \) zones increased between 1.4 and
43.2 km\(^2\) yr\(^{-1}\). \( P_{>250} \) zones increased between 0.9 and 15.5 km\(^2\) yr\(^{-1}\).

\( P_{>50} \) trends are within the 90 % confidence interval for the Umiiviit (\( p = 0.061 \))
and Naujat Kuat (\( p = 0.095 \)) plumes, the 95 % confidence interval for the Watson
(\( p = 0.026 \)), and 99 % interval for the Pakitsuup North plume (\( p=0.006 \)), and the Sar-
fartoq plume (\( p = 0.002 \)). For \( P_{>250} \), trends are within the 90 % confidence interval
for the Watson (\( p = 0.087 \)), Umiiviit (\( p = 0.089 \)) and Naujat Kuat (\( p = 0.099 \)) plumes
and 99% interval for the Pakitsuup North plume ($p = 0.010$), and the Sarfartoq plume ($p = 0.006$).

### 4.3 Comparison of fjord SSC to river discharge and ice sheet runoff

Implicit in this paper’s hypothesis was that plume SSC and discharge would be correlated, as they often are for the suspended sediment concentration of a river and its discharge. Instead, we find plume $SSC_{\text{msm}}$ does not directly correlate to the yearly volume of water a river delivered to the ocean.

Total annual river discharge did not closely correspond to $SSC_{\text{msm}}$ using the Watson gauge data (Hasholt et al., 2013). For example, in 2007 moderate yearly discharge ($3.5^{+1.68}_{-0.52}$ km$^3$ yr$^{-1}$) corresponded with the highest $SSC_{\text{msm}}$ recorded ($216 \pm 41$ mgL$^{-1}$). Yet in 2011, similar or larger discharge ($3.9^{+1.76}_{-0.56}$ km$^3$ yr$^{-1}$) produced one of the lowest $SSC_{\text{msm}}$ values ($127 \pm 24$ mgL$^{-1}$).

The poor correlation between these two variables also exists when RACMO2 annual runoff totals were used to extend the analysis in both time and to additional rivers. Of note, the Naujat Kuat plume $SSC_{\text{msm}}$ was negatively correlated to RACMO2 runoff ($r^2 = -0.6$, $p = 0.050$).

However, when data from multiple river plumes was aggregated we found larger discharge catchments predicted by the RACMO2 model tended to have higher $SSC_{\text{msm}}$ values (Fig. 9; $r^2 = 0.64$, $p < 0.001$, $n = 44$). This result agrees with Chu et al. (2012) that found melt (as determined from a positive degree day model) was correlated to SSC for large, spatially aggregated 100 by 100 km grid cells.

### 4.4 Trends between rivers

Melt-season metrics often showed similar patterns between river plumes (Tables 3 through 5). For example, all river plumes in 2009 displayed $SSC_{\text{msm}}$ at or below 2000 to 2012 mean conditions, while during the warm 2010 year all plumes displayed $SSC_{\text{msm}}$ at or above 2000 to 2012 mean conditions.
Plume conditions also varied independently between rivers. All fjords studied with multiple rivers displayed complex SSC responses in certain years, even for plumes with nearby catchments that presumably experienced similar surface melt conditions. During the extreme melt year of 2012 (Nghiem et al., 2012; McGrath et al., 2013), the Watson River plume SSC$_{msm}$ was at or below its 2000 to 2012 mean, while the Umiiviit River plume with a catchment directly south of the Watson’s, experienced a SSC$_{msm}$ 43 ± 17% higher that its 13 yr mean. This asynchronicity is especially notable as the rivers discharge into the same fjord, and thus background oceanographic conditions (salinity and tidal conditions) are very similar. Differences in SSC for the core of the plume appear to drive this asynchronicity. Both the Watson and Umiiviit has $P_{>50}$ metrics elevated above there 13 yr mean, but Watson’s $P_{>250}$ metric was close to mean conditions, while the Umiiviit was 83 ± 19% above. Figure 11d shows that the distal plume was above mean values, while pixels near the river mouth were below mean values.

River plumes displayed asynchronicity in other years. During another reportedly high melt year, 2002 (Hanna et al., 2008), the Umiiviit River plume SSC$_{msm}$ was close to its 13 yr mean SSC, while the Watson SSC$_{msm}$ was 28 ± 14% below its 13 yr mean SSC. In 2005, the Umiiviit plume SSC$_{msm}$ was at or above its 13 yr mean SSC, while the Sarfartoq River SSC$_{msm}$, with a catchment abutting Umiiviit’s GrIS catchment, was 23 ± 15% below its 13 yr mean SSC.

Observed asynchronicity is indicative of the complex sediment dynamics expected in systems controlled by glaciers. Collins (1990) hypothesized that variability in subglacial hydrologic networks drove variability in SSC emerging from glaciers. Stott and Grove (2001) recorded, “transient flushes” of increased SSC without increases in discharge for a glaciated (non-GrIS) catchment in Greenland. As mentioned in Sect. 1, river sediment dynamics often show a relationship between discharge and SSC. However, the variable response found may indicate in certain years that ice sheet sediment dynamics, not river sediment dynamics have been the dominant control on suspended sediment delivery to the coastal ocean.
5 Conclusions

We developed a retrieval algorithm capable of using MODIS reflectance to measure SSCs in fjords around Greenland. The SSC retrieval algorithm utilized the largest, most geographically expansive suspended sediment concentration dataset for Greenland. Data were collected over four different melt seasons between 2008 through 2012, three different melt months, and four different fjords with highly variable oceanographic conditions. Most of the geologic terrain eroded by the GrIS that drain into study fjords is similar, consisting of Archean and Paleoproterozoic age gneiss (Escher and Pulvertaft, 1995) and the geologic terrain in southwest Greenland sourcing river plumes is generally similar. As a result we argue that the retrieval algorithm is applicable to a larger area than just the study fjords.

The retrieval utilized both the red visible (band one, 620 to 670 nm) and near-infrared (band two, 841 to 876 nm) 250 m resolution bands of MODIS. The combination of these bands provided accurate retrievals of SSC concentrations much higher than the 80 mg L$^{-1}$ threshold band one can detect by using just a single band (Chu et al., 2009). In addition, a rigorous cloud detection scheme was developed for differentiating between turbid water and clouds, which minimizes cloud contamination of the retrieval.

An analysis using this retrieval on 13 yr of MODIS imagery showed that SSC$_{msm}$ has increased for most plumes. Intra-annual and regional variability is large in Arctic regions though, and this makes it difficult to arrive at statistically significant trends over the 13 yr record.

Metrics describing the duration and the spatial extent of river plumes responded more pronouncedly. The high concentration plumes that can impair biological productivity expanded in most fjords and one-third of the river plumes expand its high concentration core zone in time and extent (95% confidence interval).

While the relationship between the SSC$_{msm}$ of a river plume and its volume of water discharged is not correlated inter-annually, there appears to be a relationship between

6117
yearly modeled runoff totals and \( \text{SSC}_{\text{msm}} \) when rivers are studied in regional aggregate. We found that larger rivers tended to have higher plume \( \text{SSC}_{\text{msm}} \).

The behavior of fjord sediment plumes over the MODIS record was complex, and variable from catchment to catchment, even in catchments that are extremely proximal to one another. This high spatial and temporal variability suggests that approaches that treat the GrIS as a whole, or even composed of distinct regions, may not be applicable to the study of sediment dynamics. Numerous and frequently non-linear processes operating in the glacio-fluvial-fjord systems preclude a simple relationship between these parameters, yet orbital remote sensing of sediment dynamics still offers a unique and important perspective for observing change in the GrIS system.

Appendix A

MODIS batch processing

Due to the high volume of MODIS swath imagery available scripts written in the Python 2.7 programming language requested, downloaded, and processed required MODIS products. Spatial subsets of MODIS MOD02, MOD03, and MOD35_L2 data for all three study fjords were requested and downloaded using the MODAPS Web Services Application Programming Interface (http://ladsweb.nascom.nasa.gov/data/api.html).

Multiple scientific datasets were processed to accurately extract SSC concentrations. The following steps were taken to process reflectance imagery. Only scenes imaged between 13:00 and 17:00 UTC were processed to control for solar illumination.

1. MOD02 band one and two scaled integers were converted into \( R_{rs} \) using the procedure outlined in the MODIS Level 1B Product User's Guide (Toller et al., 2009)

2. \( R_{rs} \) values were corrected for atmospheric effects using dark object subtraction by finding the lowest reflectance value in the scene and subtracting all scene reflectance values by it
3. $R_{rs}$ were corrected with the MOD03 solar zenith angle product

4. Pixels that had a solar zenith angle greater than 60° were masked out

5. A land mask derived from the DNSC coastline and Landsat 7 Global Land Survey data (http://gls.umd.edu/) was used to mask out land pixels, and pixels that had a mixture of land and water in them to avoid land contamination effects.

6. A supplementary mask omitted all pixels wherein band two reflectance was greater than 0.17 for zones near each river mouth, and 0.13 for the rest of each fjord.

For Ameralik Fjord, all scenes between day 135 and 295 were considered. For Kangerlussuaq and Pakitsuup fjords, scenes were considered as soon as ice-free conditions were observed for each unique year. For Kangerlussuaq Fjord this ranged between day 143 and day 160. For Pakitsuup Fjord it ranged between day 144 and day 166.

The MOD35 L2 cloud mask uses many of the 36 MODIS bands to detect and mask clouds. However it incorrectly masks as clouds the highly turbid core of river plumes, due to the faulty assumption that water in the near infrared should have very low reflectance. For turbid water, the MOD35 L2 visible reflectance test (bit 20) and reflectance ratio test (bit 21) incorrectly classify turbid plumes as cloud (Ackerman et al., 2006). Instead, a custom cloud mask was built from MOD35 L2 data using the following clouds tests, and a custom mask described in Hudson et al. (2013). The following tests were applied (for more detail on tests see Ackerman et al., 2006)

1. Non-cloud obstruction (i.e. smoke from fires, dust storms) (bit 8)

2. Thin cirrus test using 1380 nm light (bit 9)

3. Shadow test for high confidence clear pixels (bit 10)

4. Thin cirrus test independent of bit 9, (bit 11)
5. High cloud test using 6700 nm light (bit 15)

6. IR temperature difference test using 8600, 11 000, 12 000 nm light (bit 18)

7. Low level water clouds test using difference between 11 000 nm and 3900 nm light (bit 19)

8. Test described in Hudson et al., 2013.

If any test was failed, the pixel was excluded from analysis. All datasets were then grid-
ded to a common 250 m by 250 m UTM grid using MOD03 and MOD35_L2 geolocation data.

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sula Rick, Katy Barnhart, Albert Kettner, and Leif Anderson helped with field and oceanographic
work. We thank Alberto Reyes (Queen’s University Belfast) for collection of samples used from
Orpigsøq Fjord, Wendy Freeman for help in the processing of sediment samples, and Jonathan
Bamber (University of Bristol), Janneke Ettema (University of Twente), and Michiel van den
Broeke (Utrecht University) for sharing RACMO2 data.

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Oerlemans, J.: Twenty-one years of mass balance observations along the K-transect, West 
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Table 1. Various measures of Greenland Ice Sheet catchments in paper.

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<tr>
<th>Catchment</th>
<th>Maximum elevation (m)</th>
<th>Minimum elevation (m)</th>
<th>Area (km²)</th>
<th>Area under ELA (km²)</th>
<th>Area under ELA (%)</th>
<th>Area-weighted average elevation (m)</th>
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Table 2. Surface water samples used in retrieval algorithm.

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<th>Dates samples collected</th>
<th>Number of samples used</th>
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<td>Sarfartoq</td>
<td>66°29'29.6'' N, 52°1'29.5'' W</td>
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<td>Naujat Kuat</td>
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### Table 3. Melt-season mean suspended sediment concentration (SSC$_{msm}$) percent of 13 yr mean. Underlined values are statistically below the mean, bold values statistically above.

<table>
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<th>Year</th>
<th>Pakitsuup Fjord North</th>
<th>Pakitsuup Fjord South</th>
<th>Kangerlussuaq Fjord Watson</th>
<th>Kangerlussuaq Fjord Umiiviit</th>
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<td><strong>134</strong></td>
<td>107</td>
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Table 4. Percent available fjord pixels in a melt season above 50 mgL$^{-1}$ ($P_{>50}$) as a percent of statistics’ 13 yr mean. Underlined values are statistically below the mean, bold values statistically above.

<table>
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<th>Pakitsuup Fjord South</th>
<th>Kangerlussuaq Fjord Watson</th>
<th>Kangerlussuaq Fjord Umiiviit</th>
<th>Ameralik Fjord Sarfartoq</th>
<th>Ameralik Fjord Naujat Kuat</th>
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**Table 5.** Percent available fjord pixels in a melt season above 250 mg L⁻¹ ($P_{>250}$) as a percent of statistics' 13 yr mean. Underlined values are statistically below the mean, bold values statistically above.

<table>
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<th>Ameralik Fjord</th>
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<tr>
<td>2012</td>
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Fig. 1. Overview map of study fjords, with surface suspended sediment concentration (SSC) sample locations used in retrieval overlaid on 13 yr mean SSC values derived from MODIS algorithm. Background images are Landsat 7 ETM+ from (a) 7 July 2001 (b) 20 August 2002, and (c) 3 August 2001. Inset shows location of fjords relative to Greenland Ice Sheet. An asterisk shows the location of the Watson River gauge.
Fig. 2. Relationship between MODIS band one and two corrected remote sensing reflectance ($R_{rsc}$) and surface water sample SSC values used in retrieval algorithm.
Fig. 3. Example of how “truth” computational experiment allowed scene suspended sediment concentration (SSC) to vary during a simulated melt season. For each scene a scene intensity was randomly generated that controls the scene mean SSC (plotted). Three melt seasons are plotted to show a wider range of SSCs allowed.
Fig. 4. Example of (a) “truth” image and (b) cloud masked “truth” image masked with clouds for Watson River region of interest (ROI).
Fig. 5. Deviation of “truth” from MODIS cloud masked “truth” metrics used to constrain error budget. $P_{>50}$ values (not plotted) are similar to $P_{>250}$ values.
Fig. 6. Melt season mean suspended sediment concentration (SSC$_{msm}$) for yearly melt season with error bars for (a) Pakitsuup, (b) Kangerlussuaq, and (c) Ameralik fjord plume(s).
Fig. 7. $P_{>50}$ for yearly melt season with error bars for (a) Pakitsuup, (b) Kangerlussuaq, and (c) Ameralik fjord plume(s).
Fig. 8. $P_{>250}$ for yearly melt season with error bars for (a) Pakitsuup, (b) Kangerlussuaq, and (c) Ameralik fjord plume(s).
Fig. 9. Relationship between yearly RACMO2 modeled runoff and melt season mean suspended sediment concentration ($\text{SSC}_{\text{ msm}}$) for four river plumes. Watson River observed discharge and $\text{SSC}_{\text{ msm}}$ plotted for comparison, but these data were not included in relationship.
Fig. 10. Maps of 2010 melt season mean suspended sediment concentration (SSC_{msm}) and its anomaly from its 13 yr mean for plume regions of interest. Landsat 7 false color image from 7 July 2001 underlaid for reference.
Fig. 11. Maps of 2012 melt season mean suspended sediment concentration (SSC_{msm}) and its anomaly from its 2000 through 2012 mean for plume regions of interest. Landsat 7 false color image from 7 July 2001 underlaid for reference.