Improved Monitoring of Subglacial Lake Activity in Greenland.

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Abstract. Subglacial lakes form beneath ice sheets and ice caps if water is available, and if bedrock and surface topography are able to retain the water. On a regional scale, the lakes modulate the timing and rate of freshwater flow through the subglacial system to the ocean by acting as reservoirs. More than one hundred hydrologically active subglacial lakes, that drain and recharge periodically, have been documented under the Antarctic <u>ice sheetIce Sheet</u>, while only <u>a handful of approximately 20</u>

- 5 active lakes have been identified in Greenland. Active lakes may be identified by local changes in ice topography caused by drainage or recharge of the lake beneath the ice. The small size of the Greenlandic subglacial lakes puts additional demands on mapping capabilities aiming to resolve the evolving surface topography in sufficient detail to record their temporal behavior. Here, we explore the potential for combining data from using CryoSat-2, swath-processed data together with TanDEM-X, and AreticDEM to document the evolution of four active subglacial lake sites digital elevation models to improve the monitoring
- 10 capabilities of active subglacial lakes in Greenland. The inclusion of the new We focus on four subglacial lakes previously described in the literature, and combine the new data with ArcticDEMs to obtain improved measurements of the evolution of these four lakes. We find that with careful tuning of the swath-processor and filtering of the output data, the inclusion of these new data together with the TanDEM-X data sources provides important information on lake activity, documenting, for example, that the ice surface collapse basin on Flade Isblink ice cap Ice Cap was 50% (30 meters) deeper than previously
- 15 recorded. We also present evidence of a new active subglacial lake in Southwest Greenland, which shows signs of being hydrologically connected to another subglacial lake in that region. This is to our knowledge the first evidence indication of hydrologically connected subglacial lakes in Greenland, indicating that where water is transferred from one lake to another following a draining event. These findings show how improving the measurement capabilities of subglacial lakes, improves our current understanding and knowledge of the subglacial water system and its connection to surface hydrology.

1 Introduction 20

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A subglacial lake is a body of water stored beneath an ice sheet, glacier, or ice capice cap, or glacier. Subglacial lakes are part of the basal hydrology and drainage system and may act as buffers between the melt generated on and below the ice, and the flux to the ocean (Fricker et al., 2007; Siegert et al., 2016). The origin of the water contained by in a subglacial lake depends on its regional setting. The water that feeds subglacial lakes may be generated by ice melting caused by geothermal heat or by frictional

heat from ice flow, or surface water channeled channelled to the bed as is the case in some regions (Karlsson et al., 2021). The number of observed subglacial lakes is growing (Livingstone et al., 2013) (Livingstone et al., 2022; Fan et al., 2023) and while their total volume is not large, mapping their dynamics is important to better understand the movement of meltwater through the subglacial system. Currently, 64-less than 100 subglacial lakes have been discovered in Greenland, while 675 have been detected in Antarctica (Livingstone et al., 2022). The Greenland Ice Sheet (GrIS) is warmer, thinner, and generally char-

30 acterized by a larger surface slope steeper slope of the ice surface than the Antarctic Ice Sheet (AIS), and it is possible that past subglacial lakes in Greenland drained at the end of the last glacial period (Pattyn, 2008). AdditionallyFurthermore, subglacial lakes in Greenland are typically small and located close to the margin of the ice sheet (Bowling et al., 2019), where the rapidly evolving surface mass balance further hampers hinders the detection of subglacial lake activity.

In accordance with Livingstone et al. (2022), we define a subglacial lake as stable if its volume remains relatively constant over time, or as being active if it is observed to periodically drain and refill. Triggering of subglacial lake drainage events 35 can e.g. occur after a prolonged addition of surface meltwater (Livingstone et al., 2022). The lake will eventually drain when filled with enough water to resist the pressure exerted by the overlying glacial load (Chandler et al., 2013), hence a subglacial

lake drainage event can be triggered by a prolonged addition of surface meltwater (Livingstone et al., 2022). The sudden drainage and outburst flood of a subglacial lake might temporarily affect ice flow velocities downstream from the lake location

- (Palmer et al., 2015; Liang et al., 2022). as documented by Magnússon et al. (2007); Liang et al. (2022); Stearns et al. (2008). 40 This is not always the case, as shown by Smith et al. (2017), where a drainage event under Thwaites Glacier in West Antarctica had no apparent impact on the ice velocities. Therefore, the behavior of subglacial lakes is important to consider when discussing the response of the ice sheets to a warming climate (Willis et al., 2015). The behavior of active subglacial lakes is also an important indicator of hidden subglacial processes. Notably, by monitoring after a lake drainage event, In particular,
- 45 the period of lake recharge provides information about subglacial water production and conditions at the bed (Malczyk et al., 2020).

Active subglacial lakes may be identified by ice surface collapse basins (surface depressions) created when a lake drains, and localised surface uplift as the lake refills. The surface expressions of subglacial lake volume oscillations are controlled by viscous ice flow (Stubblefield et al., 2021). In contrast, stable subglacial lakes are typically identified using radio-echo

sounding as they cannot be identified from ice surface topography. Evans and Smith (1970) were the first to detect a subglacial 50 lake under the AIS by Radio-Echo-Sounding (RES). RES can penetrate the ice sheet and map the bedrock topography, where a strong, flat reflection indicates basal water presence (Bingham and Siegert, 2007; Tulaczyk and Foley, 2020). RES has been used to detect and map numerous subglacial lakes under the ice sheets (Wright and Siegert, 2012; Siegert et al., 2016; Bowling et al., 2019). Stable subglacial lakes cannot be identified from the characteristics of the overlying ice surface, except for large

55 lakes which may influence surface topography as seen at Lake Vostok, AIS where the surface relief is exceptionally flat (Ridley et al., 1993). Active subglacial lakes, may on the other hand be identified by ice surface collapse basins (surface depressions) created when the lake drains, and localised surface uplift as the lake refills. The surface expressions of subglacial lake volume oscillations are controlled by viscous ice flow (Stubblefield et al., 2021). Only-

In a recent study, Fan et al. (2023) identify 18 active subglacial lakes in Greenland from surface variability based on ICESat-2
data. Prior to the Fan et al. (2023) study only four of the known subglacial lakes in Greenland have had been identified by ice surface collapse basins, while the rest are were identified by the use of RES (Bowling et al., 2019).

The small size of the subglacial lakes in Greenland (<1 km²) makes them impossible to map from conventionally processed radar satellite altimetry (Meloni et al., 2020), whereas their large Antarctic counterparts (10-100 km²) have been monitored extensively (Livingstone et al., 2022). However, due to the novel Interferometric Synthetic Aperture Radar synthetic aperture

- 65 <u>radar interferometric</u> (SARIn) mode of the European Space Agency's (ESA) first Earth explorer mission CryoSat-2 (CS2) and recent advances in so-called swath-processing (Gray et al., 2013; Foresta et al., 2016; Gourmelen et al., 2018a; Andersen et al., 2021), we now can look at even smaller targets <u>can now be monitored</u> as suggested by Wingham et al. (2006). The swath processing enables us to generate a swath of elevation estimates across track, which increases the spatial data resolution and coverage. This processing method means that the ability to map topographic lows is improved compared to conventional
- 70 retracking, which preferably tracks the point of closest approach (topographic highs). This increases the chance thus increasing the possibility of acquiring data over small surface features. An additional source of high-resolution ice surface topographic information is provided by two sources of Digital Elevation Models (DEMs); TanDEM-X DEMs derived from the X-band TanDEM-X satellite mission interferometric synthetic aperture radar (InSAR) satellite mission TanDEM-X (Rizzoli et al., 2017), and ArcticDEM from the panchromatic band WorldView satellite mission (Porter et al., 2018). Here, we investigate the
- 75 capabilities and added value of the CS2 swath-processed altimetry data and high-resolution TanDEM-X DEMs, focusing on the four active subglacial lakes in Greenland that have previously been identified and described in the literature (Palmer et al., 2013, 2015; Howat et al., 2015; Bowling et al., 2019) since this allows us to evaluate our results. These four subglacial lakes are all characterized by the occurrence of collapse basins in the ice sheet/ice cap surface topography after a lake drainage event. Through analysis of swath-processed CS2 data, TanDEM-X DEM scenes, and ArcticDEMs we present time series of collapse
- 80 basin depths and volumes at an unprecedented temporal resolution, thus advancing our understanding of the subglacial lake draining and refilling timing and rates of the four active subglacial lakes.

2 Subglacial Lake Sites

The four active subglacial lakes in Greenland identified by Bowling et al. (2019) from that we will investigate here are all identified by observations of surface collapse basins, and are located as followingfollows; one in West Greenland, two in
Southwest Greenland, and one under the Flade Isblink ice cap Ice Cap in Northeast Greenland. Their locations are shown in Fig.1 on a background image of bedrock elevation from BedMachine v3 (Morlighem et al., 2017). Here, we summarize the

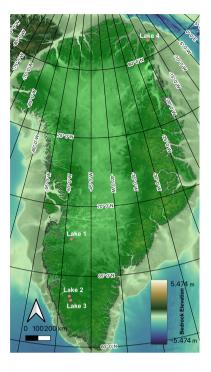


Figure 1. The locations of the four sites of active subglacial lakes investigated in this study.

present knowledge of the lakes following Palmer et al. (2015); Willis et al. (2015); Bowling et al. (2019). We have not included the three subglacial lakes located beneath the highly dynamic Isunnguata Sermia glacier due to the small size of the collapse basins and their location very close to the ice sheet margin (Livingstone et al., 2019).

90 2.1 Lake 1: West Greenland

Lake 1 is a subglacial lake located in the western part of the GrIS (67.611°N, 48.709°W) northwest of the Inuppaat Quuat glacier(Fig. 7a), and, just below the Equilibrium Line Altitude (ELA). Its location and regional setting are shown in Fig. 7(a) using a Landsat-8 image from 2019 as background. The temporal evolution of Lake 1 has previously been studied by Palmer et al. (2015) and Howat et al. (2015), using a variety of elevation data sets such as Search-space Minimization (SETSM) DEMs

- 95 derived from WorldView data, the Greenland Mapping Project (GIMP) DEM, ICESat along-track elevationsand, airborne LiDAR data, and optical imagery. The datasets constrained the spatio-temporal spatiotemporal evolution of the lake drainage and the associated ice surface collapse. These studies found find that Lake 1 drained both in 2004 (smaller event) and in 2011 (larger event). The 2011 drainage occurred at an unknown rate within two weeks (28 June, 2011 to 12 July, 2011), resulting in the formation of a collapse basin in the ice sheet surface. A SETSM DEM from October 28th, 2011, revealed a collapse basin
- 100 of about 1 km in width and 60-70 m in depth. The bottom of the collapse basin was flat, which suggested that the subglacial lake was still partially filled. According to both Palmer et al. (2015) and Howat et al. (2015) it is likely that Lake 1 receives meltwater from the surface, and that the drainage of Lake 1 in 2011 may have been triggered by the drainage of a nearby

supraglacial lake. The routing of water to the bedrock from the surface, e.g. through a moulin, has been known to trigger drainage of subglacial lakes due to overfilling by meltwater (Willis et al., 2015). Howat et al. (2015) also found indications of

105 a 2004 subsidence event above the lake. The collapse basin was observed to partially refill between 2011 and 2013, however, it could not be concluded whether the subglacial lake recharged or if the depression simply filled up with surface water or snow.

2.2 Lake 2 and 3: Southwest Greenland

Bowling et al. (2019) identified two collapse basins Two collapse basins are found in Southwest Greenland, located between the Sermeq and Sioqqap glaciers (Fig. 8a and 9a). They classified with very high confidence that and according to Bowling et al. (2019) these

- 110 surface depressions are (with very high confidence) associated with the drainage of subglacial lakes. We denote the northernmost lake as Lake 2 (63.542°N, 48.449°W), and the southern one as Lake 3 (63.261°N, 48.207°W). The respective locations of the two lakes are shown in Fig. 8(a) and Fig. 9(a). They are located about 35 km apart, and the collapse basin over Lake 2 was 15 m deep in August 2012, while the one over Lake 3 was 18 m deep in June 2012. Using ArcticDEM strips from 2015, they Bowling et al. (2019) also found that both collapse basins decreased in volume in the period 2012-2015, which suggests
- 115 a refilling of the subglacial lakes, and it was further estimated that the recharge of Lake 3 has been ongoing since 2001, while the <u>timing of the</u> drainage event for Lake 2 was not identified. Optical images show supraglacial lake drainage in the region, which could indicate that some recharge of the subglacial lakes to some extent are filled by surface water.

2.3 Lake 4: Flade Isblink ice capIce Cap

The collapse basin above a subglacial lake on the southern dome of Flade Isblink ice cap Ice Cap in the northern part of Greenland (see Fig. 10a(a)) has been described by Willis et al. (2015); Liang et al. (2022)Willis et al. (2015) and Liang et al. (2022), and we denote it Lake 4 (81.157°N, 16.613°W). Willis et al. (2015) base their analysis on DEMs from stereo satellite imagery together with airborne LiDAR observations, while Liang et al. (2022) investigate the subglacial lake using ArcticDEMs and ICESat-2 altimetry data between 2012 and 2021. From MODIS optical imagery Willis et al. (2015) find found that the ice surface above Lake 4 had subsided in the autumn of 2011, leaving a surface depression shaped like a mitten. The basin was estimated to have formed over a three-week period between August 16, 2011, and September 6, 2011, and it comprises two sub-basins. The first estimate of elevation measurements was from a WorldView-1 derived DEM from May 2012 showing a maximum depth of the collapse basin of about 70 m. The elevation of the collapse basin rapidly increased by 30 meters over the following two years due to inflow of surface water to the subglacial lake, and between August 2012 and April 2013 a topographic bulge appeared in the basin (Willis et al., 2015). Liang et al. (2022) find that the lake drained again in 2019, but that it released much less water than during the 2011 eventresulting in a 10 m elevation change.

3 Data Sets and Data Processing

In the following, we present the data sets and outline our processing steps for the remote-sensing data-.

3.1 CryoSat-2 Data

3.1 TanDEM-X

- 135 TanDEM-X is an interferometric synthetic aperture radar system consisting of two satellites, TerraSAR-X and TanDEM-X, launched in 2007 and 2010, respectively. Its primary mission objective was the generation of a global digital elevation model, which was completed in 2016 with a spatial resolution of 0.4 arc seconds (i.e., about 10 m 12 m) (Rizzoli et al., 2017). Here, we use TanDEM-X data acquired between the years 2010 and 2017. The InSAR data were requested via a TanDEM-X science proposal as co-registered single-look complex data, and the interferometric processing and calibration to produce
- 140 interferometric DEM scenes were done by the German Aerospace Center (DLR) using the Integrated TanDEM-X Processor (ITP) and the Mosaicking and Calibration Processor (MCP) (Rizzoli et al., 2017). The ITP processed the interferometric bistatic data to interferograms and then performed phase unwrapping and geocoding (Lachaise et al., 2018; Rossi et al., 2012). The absolute vertical calibration of the uncalibrated DEM scenes was performed by MCP block adjustment. For Greenland, this procedure relied on ICESat points over rocks as vertical reference and tie points to transfer the height level to the data located
- 145 further inland (Wessel et al., 2016). For each DEM scene, an individual height offset was estimated and applied (up to 10 m). The penetration of the X-band SAR signal into the snow and ice surface by several meters (Rott et al., 2021; Fischer et al., 2020; Wessel et a maintained by the block adjustment and may therefore complicate validation and comparison with other data. However, this issue was by-passed in this study by aligning different DEMs at stable anchor points, which is described in Section 4.1. Figure 2 shows examples of TanDEM-X DEMs covering Lake 1 at different time steps, with figures 2(a), 2(b), and 2(c) from 2011,
- 150 2012 and 2016, respectively. Fig. 2(d) shows the collapse basin in an optical image from July 2012. Figure 2(e) shows the elevation difference between February 2011 and April 2012, while Fig. 2(f) shows the elevation difference between April 2012 and December 2016.

3.2 ArcticDEM

The ArcticDEM input data comprise of timestamped 2 m spatial resolution DEM strips covering the period 2011-2017
(Porter et al., 2018). These DEMs have been generated from stereoscopic WorldView and GeoEye satellite imagery by the ArcticDEM Team, using the Surface Extraction with TIN-based SETSM algorithm (Noh and Howat, 2018). Following their generation, they are freely distributed as strip files to the community by the US Polar Geospatial Centre. The DEMs are then co-registered using lateral and vertical corrections provided within the ArcticDEM metadata (Porter et al., 2018). Examples of ArcticDEM strip data are shown in Fig. 7(c) for Lake 1 (October 2011), Fig. 8(c) for Lake 2 (July 2012), Fig. 9(c) for Lake 3

160 (May 2012), and in Fig. 10(c) for Lake 4 (May 2012).

3.3 CryoSat-2

CS2 measures in three different operational modes; Low Resolution Mode (LRM), Synthetic Aperture Radar (SAR) and Interferometric SAR (SARIn mode and SARIn mode, of which the latter can be used for swath processing (Wingham et al.,

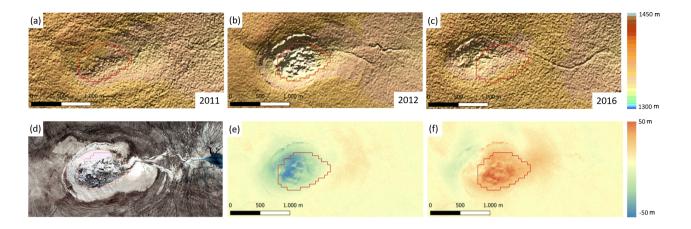


Figure 2. Subset of three color-shaded TanDEM-X DEMs for Lake 1: (a) from February 17, 2011 (before the collapse), (b) from November 05, 2012 (after the collapse) and (c) from December 10, 2016 (after some refilling). (d) Optical image from July 2012 (after collapse). (e) TanDEM-X elevation difference 2012-11-04 minus 2011-02-17. (f) TanDEM-X elevation difference 2016-12-10 minus 2012-11-04. The outline of the core collapse basin from 2012 is displayed in red in all subfigures.

2006). Here, we swath process the CS2 SARIn level 1b baseline D data (Meloni et al., 2020). The conventional SARIn Level 2

- 165 (L2) elevation product from CS2 consists of a single surface elevation measurement at the Point-Of-Closest Approach (POCA) along the satellite flight line. This L2 elevation product <u>exploits just only takes advantage of a fraction of the information contained in the</u> 1024 measurements contained <u>in within</u> a single CS2 waveform. By implementing swath processing of radar altimetry, it is possible to generate additional elevation measurements using part or all of the remaining echo (Gourmelen et al., 2018b; Andersen et al., 2021). The CS2 pulses scatter from Earth's surface from distinct locations in the satellite across track
- 170 <u>across-track</u> direction, and these scattering locations are recorded within the Level 1b waveforms. The principle of the L2 swath processing algorithm is to identify the <u>high coherence high-coherence</u> data points that scatter off the ice and to extract the corresponding elevation and location. The main steps of L2 swath processing for extracting elevations are: (1) Identifying high-quality records (e.g. based on selected coherence and power thresholds), (2) Unwrapping (removing phase jumps from the data), and (3) computing Computing elevation and geographic location (Gourmelen et al., 2018b; Andersen et al., 2021).
- 175 This leads to the generation of elevation measurements at ranges beyond the POCA location , and to an overall increase in the spatial density and coverage compared to the L2 product. This increase is needed in order to map the Depending on e.g. the chosen processing thresholds, the physical properties and the topography of the area, the L2 swath processing leads to a 10 to 100 folds increase in elevation measurement compared to conventional L2 processing.

Regional tuning of the swath processor is required to allow us to detect small-scale features of the ice surface collapse basins

180 above active such as the Greenlandic subglacial lakes in Greenland (Noël et al., 2015; Andersen et al., 2021). To increase the chance of detecting the small-scale signatures of subglacial lake activity in Greenland, the level 1b waveforms from CryoSat-2. To illustrate the nature of the signal, we show in Figure 3 two CryoSat-2 waveforms recorded over Lake 1. The first waveform

(Fig. 3(a)) is recorded on August 7, 2010, which is prior to the drainage of Lake 1, while the second waveform is recorded on August 10, 2011 (Fig. 3(b)), which is after the drainage event. The dark red parts of the waveforms indicate bins that are

- 185 accepted by the chosen power (0.001) and coherence (0.6) thresholds. The locations of the processed elevation estimates are shown on top of an ArcticDEM from October 28, 2011 in Fig. 3(c), where the blue points correspond to the blue part of the waveform in Fig. 3(b). It is seen that the part of the signal that originates from the collapse basin corresponds to a small but clear peak late in waveform #2, which is indeed not present in waveform #1, and that this peak has a much lower power than the principal peak in the waveform. The satellite nadir points are indicated by the triangles in Fig. 3(c).
- 190 To capture the peak from the drainage basin the swath processor was tuned to allow for the inclusion of elevation estimates associated with lower coherence and power thresholds than what is compared to those usually applied in the literature (Gourmelen et al., 2018b; Andersen et al., 2021). The appropriate thresholds for our study were determined by comparing CS2 swath elevation estimates derived using different threshold values, with the earliest ArcticDEM scene that contained the collapse basin. We note that the threshold requirement is very site-specific, depending on the local conditions. Factors such as
- 195 the geometry of the surface depression and the scattering properties of the underlying surface play a crucial role in determining the appropriate threshold values. Therefore, it is important to consider these local conditions and adapt the threshold values accordingly for accurate and meaningful analysis in different study areas.

4 Methods

4.1 Filtering of CryoSat-2 data

- 200 Figure 4(a) shows as an example the spatial data coverage of CS2 data over Lake 4, with POCA locations as black circles and swath-processed data in colours according to the elevation. It is clear that the swath-processed data offer an increase in data coverage, which is needed in order to map the small-scale features of the ice surface collapse basins above active subglacial lakes in Greenland, since no POCA points are located within the collapse basin (Noël et al., 2015; Andersen et al., 2021). Figures 4(b) and 4(c) show the output of the swath-processor when applying a coherence threshold of 0.8 (standard) and
- 205 0.5, respectively. In Figure 4(b) no swath-processed data points are obtained inside the lake outline, while in Figure 4(c) the number of available swath-processed data points increases due to the decreased coherence limit. Decreasing the normalised coherence limit increases the number of generated elevation estimates, but increases the noise and errors in the output. Andersen et al. (2021) found that decreasing the coherence limit from 0.8 to 0.6, Andersen et al. (2021) found that increased by 25%, but increased the standard deviation of intra-mission crossover elevation
- 210 difference increased by 35-65 %.

Lowering-

Since lowering the coherence and power thresholds increases the probability of phase unwrapping errors in the L2 elevation product (Gourmelen et al., 2018a), and filtering of the generated elevation point data is therefore required to remove erroneous data from the subsequent analysis. Here, filtering based on coherence, power and range bin numbers of each produced elevation

estimate is applied. We find that the swath-processed CS2 data from within the collapse basin and from the surface near the

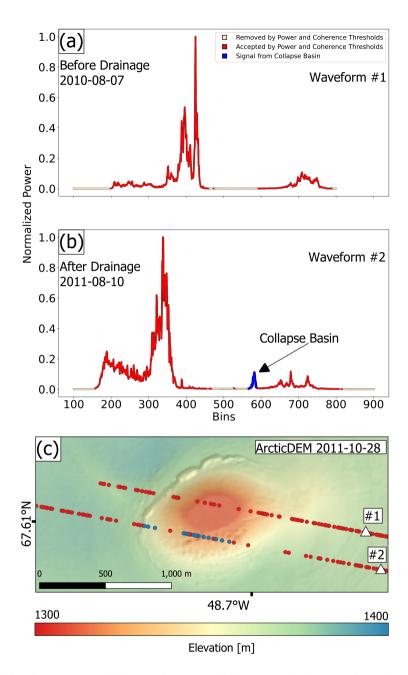


Figure 3. (a) CryoSat-211b data from August 7, 2010 (waveform #1) and (b) August 10, 2011 (waveform #2) over Lake 1. Dark red sections of the waveform are those accepted by the algorithm thresholds, the light red sections are not. The blue section in the waveform in (b) that located over the collapse basin after processing. (c) ArcticDEM scene (October 28, 2011) showing where the processed swath points from waveforms #1 and #2 are located. The satellite nadir position is shown with triangles.

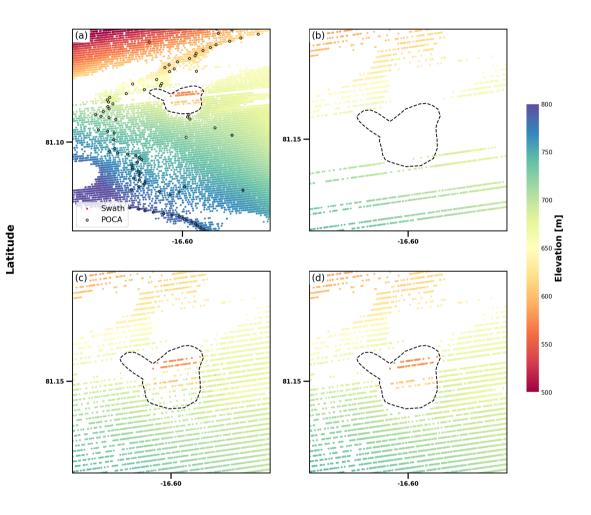




Figure 4. (a) POCA and swath processed elevation data from November 2011, in the area surrounding Lake 4. (b) The standard swath processed elevation data with a coherence threshold of 0.8. (c) The swath processed elevation data with a decreased coherence limit of 0.5. (d) The filtered swath processed elevation data with a decreased coherence limit of 0.5.

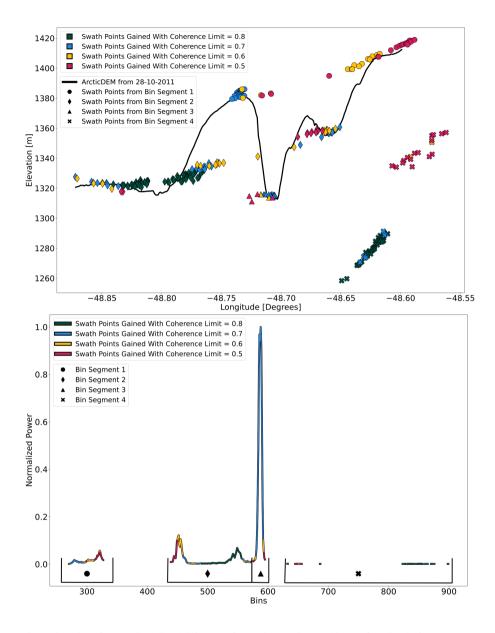


Figure 5. (a) Processed swath points from a CryoSat-2 11b Waveform (September 7, 2011) after the drainage event at Lake 1. The colours indicate what coherence limit was used to process each point and the symbols specify from where on the 11b waveform they are from. The black line is the topography along the swath from an ArcticDEM scene from October 28, 2011. (b) Corresponding CryoSat-2 11b waveform, with the coloured sections illustrating what waveform part is gained when changing the coherence limit. The waveform is divided into four bin segments labelled by four distinctive symbols.

rim, respectively, have distinct compositions of these three waveform parameters. This <u>clustering in the data</u> allows us to identify and remove <u>noisy erroneous</u> data points by choosing appropriate threshold values for these parameters. The thresholds were determined by comparing CS2 swath elevation estimates derived using different threshold values, with a temporally close ArcticDEM scene that contained the collapse basin. The optimal normalised coherence and power threshold values were found to be

220 to be

To illustrate these clusters in the data we show as an example in Fig. 5 some detailed information from a single waveform from September 7, 2011 over Lake 1. Fig. 5(b) shows the normalized power in the waveform with colours indicating their corresponding coherence range; The dark green sections of the waveform are those accepted at a coherence threshold of 0.8, light blue sections are those additionally accepted from lowering the threshold to 0.7. Likewise, the yellow and pink are those

225 for coherence thresholds of 0.6 and 0.001, respectively 0.5, respectively. The power threshold is fixed at 0.001 at this site. Also, we have divided the waveform bins into four segments and assigned each one a symbol in Fig. 5(b). The round points are from the early bin range at \sim 300, the diamond shapes are from the range at \sim 500, the triangles are from the large peak at \sim 600, and the crossed points are from the low power section after the large peak.

In Fig. 5(a) the elevation estimates obtained by the swath processor are shown together with an elevation profile from an
 ArcticDEM scene from October 28, 2011, shown as a black line. The elevation estimated is colour-coded and annotated with a symbol according to Fig. 5(b).

From Fig. 5(a) we can see that if applying a coherence threshold of 0.8 no measurements from the drainage basin (longitude -48.80 degrees) are obtained. When decreasing the threshold to 0.7, much of the large peak in bin segment 3 is processed, and all the swath points from that bin segment originate from the collapse basin. When setting the threshold to 0.6, we gain a

- 235 few more points at the basin from bin segment 3, which is beneficial, but we also see that two data points show elevations of \sim 1380 m even though they are located above the drainage basin (\sim 1320 m). These points originate from bin segment 1 and are therefore easy to separate from the other data points from within the drainage basin. Likewise, it is clear from the figure that all measurements from bin segment 4 (crossed points) deviate from the real surface, hence this cluster of data will be filtered out and will not be used in the analysis.
- Even after the removal of data based on this filtering, some erroneous data points are still found might persist over the collapse basins. This is believed to be caused by the highly dynamic surface at the collapse basins, which changes the scattering mechanisms over time, resulting in a larger incoherent component in the swath processing. This could be negated by increasing the thresholds, but it was not possible to find an ideal combination that removed every error and also keeping the data.

These apparent errors are therefore basin, which are removed in a second step of the filtering by applying a by applying lowest-level filtering to the elevation estimates within the outlined collapse basin, so that the cluster of data with the lowest elevation within the defined lake outline are assumed to is assumed to be representative of the bottom of the collapse basin. The removed estimates were deemed as errors after we compared them with ArcticDEMs that were close in time to the CryoSat-2 SARIn data. Across swath tracks close to the basin rim and the slanting walls would often give erroneous estimates compared to the ArcticDEMs and were removed. An example of results obtained by swath-processing and subsequent filtering over Lake 4 is shown in Fig. 4(d), which shows the final elevations estimates after the filtering of the data shown in Fig. 4(c). We find that the fraction of data, which is removed , removed data depends on the scattering mechanisms of the basin basin's scattering mechanisms and the distance to the CS2 nadir track.

4.2 TanDEM-X Data

- 255 TanDEM-X is an interferometric synthetic aperture radar (InSAR) system consisting of two satellites, TerraSAR-X and TanDEM-X, launched in 2007 Between Figure 4(c) and 2010, respectively. Its primary mission objective was the generation of a global digital elevation model, which was completed in 2016 with a spatial resolution of 0.4 are seconds (i.e., about 10 m 12 m)(Rizzoli et al., 2017). Here, we use TanDEM-X data acquired between the years 2010 and 2017. The InSAR data were requested via a TanDEM-X science proposal as co-registered single-look complex data, and the interferometric processing
- 260 and calibration to produce interferometric DEM scenes was done by the German Aerospace Center (DLR) using the Integrated TanDEM-X Processor (ITP) and the Mosaicking and Calibration Processor (MCP) (Rizzoli et al., 2017). The ITP processed the interferometric bistatic data to interferograms and then performed phase unwrapping and geocoding (Lachaise et al., 2018; Rossi et al., 201 The absolute vertical calibration of the uncalibrated DEM scenes was performed by MCP block adjustment. For Greenland, this procedure relied on ICESat points over rocks as vertical reference and tie points to transfer the height level to the data takes
- 265 located further inland (Wessel et al., 2016). For each DEM scene, an individual height offset was estimated and applied (up to 10 m). The penetration of the X-band SAR signal into the snow and ice surface by several meters (Rott et al., 2021; Fischer et al., 2020; Wes maintained by the block adjustment and may therefore complicate validation and comparison with other data. However, this issue was by-passed in this study by aligning different DEMs at stable anchor points, which is described in Section 4.1. 4(d), there is a reduction of 47% of data points inside the collapse basin outline.

270 4.2 ArcticDEM

The ArcticDEM input data comprise of timestamped 2 m spatial resolution DEMs covering the period 2011-2017 (Porter et al., 2018). These DEMs have been generated from stereoscopic WorldView and GeoEye satellite imagery by the ArcticDEM Team, using the Surface Extraction with TIN-based SETSM algorithm (Noh and Howat, 2018). Following their generation, they are freely distributed as strip files to the community by the US Polar Geospatial Centre. The DEMs are then co-registered using lateral and vertical corrections provided within the ArcticDEM metadata (Porter et al., 2018).

5 Methods

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4.1 Alignment of data sets

We vertically align the three elevation data sets (TanDEM-X, CS2, and ArcticDEM) to account for fluctuations in the surface elevation caused by regional surface mass balance (SMB), ice dynamics (similar to what was done in (Palmer et al., 2015))

- and for different penetration biases for the radar measurements, typically 0.5 m 1 m for CS2 and 3 m 4 m for TanDEM-X (Wessel et al., 2016; Abdullahi et al., 2019). The vertical alignment is done by defining anchor points close to the rim of each subglacial lake surface depression in the earliest available DEM. All consecutive DEMs are height corrected to align with the reference DEM at the location of the chosen anchor point. To vertically align the discrete CryoSat-2 swath data to the DEMs, the median of the difference between CS2 points close to the collapse basin and the reference DEM were used to correct for
- 285 the vertical the bias. This procedure ignores different penetration depths at the surface on the rim, but it preserves the relative heights from the top of the rim to the bottom of the depression for each sensor. This alignment allows us to focus on the local height change within the collapse basin likely caused by the dynamic hydrological processes. The maximum vertical alignment needed was found to be offset was 12 m, but for the vast majority of the data sets the alignment correction was less than 3 m.
- The data are also aligned horizontally to account for ice flow. Lakes 1-3 are located in the upper ablation zone in the southwestern part of the GrIS, where the ice flows in a westerly direction. Optical imagery as well as the three data sets provide evidence that the collapse basins advect with the ice flow. Figure 6 shows the drift of the collapse basin over Lake 1 in the period 2011-2021. ArcticDEMs were used to create the collapse basin outlines between 2011 and 2017, while Landsat-8 was used until 2021. We compared an ArcticDEM scene and a Landsat-8 image from the summer period of 2015, and found no considerable visual bias in the basin outline. The basin drifts in a westerly direction and decreases 95 % in surface area from 2011-2021 as the subglacial lake gradually recharges leading to and/or the collapse basin moving away from the subglacial
- 295 2011-2021 as the subglacial lake gradually recharges , leading to and/or the collapse basin moving away from the subglacial lake location is filled due to ice flow and deformation. This movement has created a surface depression about results in a horizontal offset of 100 meters downstream m between the surface depression in 2021, ten years after the drainage, and its initial location in 2011.

To consistently track the evolution of the depth of the collapse basins, even when they change location due to ice flow, we 300 horizontally align the data sets by correcting for the observed local ice flow using MEaSUREs Greenland Ice Sheet Velocity Map from InSAR Data (Joughin et al., 2010, 2015) and Greenland Ice Velocity from Sentinel-1 data, Edition 2 (Solgaard et al., 2021; Solgaard and Kusk, 2021). At the location of Lake 4, the ice flow velocity is found to be <17 m/yr but and we see no evidence in the elevation that this collapse basin has moved over time. One reason for this can be that the subglacial lake drains again in the observational period, and also the larger size of this lake makes the which could make the potential ice flow less

305 evident <u>No-since the collapse basin is re-formed over the stationary location of the subglacial lake. Therefore, no horizontal alignment was made applied at Lake 4.</u>

4.2 Time Series of Deepest Point

The temporal evolution of the depth and shape of the collapse basins is controlled by local factors such as including refilling of the subglacial lake, SMB, and ice dynamics. Here, we determine the location of the deepest point within a basin. Lake 4 is the

310 largest lake with a relatively flat 1000 m bottom diameter in contrast to the 100 m to 400 m ground diameter of Lake Lakes 1-3. We define the deepest point based on the first DEM in which the basin is detected , and use the temporal change in elevation of this location as a measure for the evolution of the subglacial lake. For each time-stamped ArcticDEM and TanDEM-X, all grid points within a distance of 50 meters from the initially deepest point were sampled as this ensures enough data to calculate a

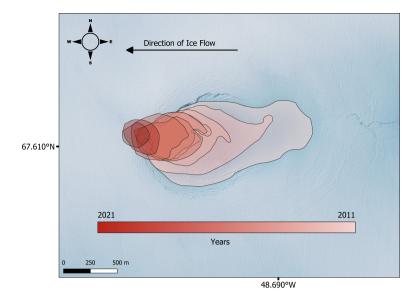


Figure 6. The horizontal drift of the collapse basin over Lake 1 during the period from 2011 to 2021. Its location is shown in Fig. 7. ArcticDEMs were used to create the collapse basin outline between 2011 and 2017, while Landsat-8 images were used until 2021. The basin drifts in a westerly direction and the surface area decreases substantially.

robust mean and standard deviation (σ), while still avoiding sampling of the slanting basin walls. Due to the scattered coverage
and the limited number of CS2 data points, the corresponding estimate of the basin depth at Lake 1 is derived from all CS2 point data within the outline of the basin. We only used CS2 track crossings from which 10 or more points are available within the basin, and points within 5σ of all data within the basin. The horizontal alignment and the filtering of the CS2 swath data, ensures that we do not need to change the basin shapefile manually delineated basin outline through time. At the substantially larger Lake 4, we sampled all CS2 points within 400 meters from the deepest point, thus not sampling the inclining basin floor.
We compute the mean of each of the sampled data sets and use this as the value for the elevation of the deepest point. Their standard deviation represents the spatial variability of these data, and we use 2σ as the error bar on the depth estimate , but note

4.3 Subglacial Lake Volumes

that this is not a measure of their true accuracy.

We estimate subglacial lake volumes by assuming that the subglacial lake volume changes are directly transferred to the surface topography. While this is an approximation, the study by Stubblefield et al. (2021) supports this, demonstrating that at an ice thickness of ~1 km (500 m - 1.2 km at our study sites) and draining time of less than half a year (we find it to be within weeks at study sites), the volume of the lake and the collapse basin will be almost the same. We derive the collapse basin volume by extracting surface elevations of the basin area using the TanDEM-X and ArcticDEM DEMs manually delineate the collapse basin outline from the earliest available DEM, which contains the collapse basin. The extracted DEM height anomalies are

330 derived by subtracting the median height at the basin rim as defined by the manually delineated outline. The volume of each

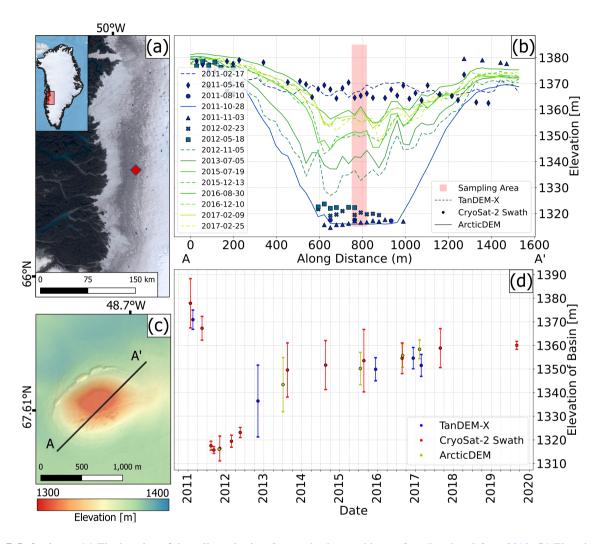


Figure 7. Lake 1 : (a) (a) The location of the collapse basin \cdot (b) on a background image from Landsat-8 from 2019. (b) Elevation profiles from the three aligned data sets, ; TanDEM-X is the dashed line, CryoSat-2 data is represented by dots symbols and ArcticDEM is the solid line. e) (c) The collapse basin as seen in an ArcticDEM, 28th of October 2011. Line A-A' is the profile used in Figure (b). (d) (d) Time series of the deepest point of the lake basin from TanDEM-X (blue), CS2 (red) and ArcticDEM (yellow). The depth is based on sampling within the area indicated by red in (b).

grid point within the outline is summed to provide a total basin volume. The horizontal and vertical DEM alignment (described in Sect. 4.1) ensures that the original outline also can be used to derive the temporal changes in basin volumes. The uncertainty of the DEM volumes are estimated by cubing the 2σ from the depth estimates.

At collapse basins where CryoSat-2 swath data is also available, we use a different approach: Based on each available DEM 335 over a given lake, we calculate the shape factor (R) between lake depth (H) and volume (V) at times t_i :

$$R(t_i) = \frac{V(t_i)}{H(t_i)}$$

Since the collapse basin shape and form changes over time, the shape factor R changes over time, and we fit a smoothed function, $\tilde{R}(t)$, through all available $R(t_i)$ values, taking their error bars into account. We the use $\tilde{R}(t)$ to estimate lake volumes from the CS2 depth estimates ($H_{CS2}(t)$) when available, through:

$V_{CS2}(t) = \widetilde{R}(t) \cdot H_{CS2}(t),$

to construct a time series of lake volumes constrained by both the available DEMs and CS2 data with their depth uncertainties. To account for the CS2 uncertainties in the volume estimation, the uncertainty at all $V_{CS2}(t)$ were computed by using the depths $H_{CS2}(t) \pm 2\sigma$, where σ is the standard deviation of the CS2 depth mean.

5 Results

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345 For each lake site, we present the temporal evolution of the ice surface collapse basins both by height profiles across the basins and as a time series of the elevation of the deepest point in each basin. Estimated lake volume changes are presented at the end of this section.

5.1 Subglacial Lake Depths

Our findings for Lake 1 are presented in Figure 7. Figure 7b (b) shows the ice sheet elevations of Lake 1 along the A-A' profile shown in Figure 7e(c). The location of A-A' is adjusted through time to account for ice flow. With CS2 swath processing, it was possible to extract point measurements from 12 satellite tracks passing over Lake 1. The CS2 data are point observations of elevations and do not provide a full surface coverage, as the TanDEM-X and ArcticDEM DEMs do. Hence, the CS2 observations plotted are those data points that are located closer than 180 m to the A-A' profile, as this <u>This</u> distance ensures that data from all satellite crossings are represented. The colours of the <u>line height profiles</u> indicate the time of the measurements, with darker colours for the start of the measurement period (2011) and lighter colours for the end of the measurement period (2017). We find that a TanDEM-X DEM from February 17, 2011, and CS2 data from May 18, 2011, show that that the surface is generally flat, indicating that the ice surface collapse basin has not yet been created. The first data set that shows measurements that

show signs of a collapse basin is a CS2 crossing from August 10th, 2011, which detects a depression of the surface of surface

depression with a depth of approximately 60 m, and which has with a flat surface in at the bottom. Over time, the collapse

- basin decreases in size (both in depth and area), but a depression is still evident in the most recent displayed data set, which is an ArcticDEM from September 2nd, 2017. For clarity, we have not shown all available data sets in Figure 7b(b). Figure 7d (d) shows the temporal evolution of the depth of the deepest part of the collapse basin (Sect. 4.2), and it shows that the basin depth did remain stable during the winter of 2011/2012, following the collapse. In the period from February 2012 to July 2013 TanDEM-X and ArcticDEM DEMs reveal a rapid decrease of the basin depth-rise of the depression floor over the 15-month
- 365 period from February 2012 to July 2013, during which the depth is reduced by 35 m to a depth of to about 25 m. The results show a slower decrease rate after 2015. At the time of the last measurement by CS2 in late 2019, the depth is approximately 15 m, which agrees within the error bars with the height from an ArticDEM from September 2017. The subglacial lake recharge collapse basin filling rate can be divided into a fast basin uplift of ~ 13 m/yr in the period 2011-2015, and a slow-uplift of ~ <1 m/yr in the period period with no substantial uplift from 2015-2019.
- Figures 8 and 9 show the data available for Lake 2 and Lake 3, respectively. While several TanDEM-X DEMs covering these sites were successfully produced, we were not successful in obtaining any useful CS2 swath-processed data over these subglacial lakes. The collapse basins over Lake Lakes 2 and 3 span a smaller area smaller areas than those over lakes Lakes 1 and 4, and they are also shallower, which could be the reason for the lack of CS2 data here. The early 2011 TanDEM-X observations of Lake 2 show no signs of a collapse basin being formed, however, the four subsequent DEMs (July 2012 to
- 375 April 2013) clearly show the imprint of a collapse basin (see Fig. 8(b)). The collapse basin had a depth of approximately 15 m, which did not change over this time span (from July 2012 to April 2013), but, while in the period from April 2013 to December 2013 it appears to fill up completely. After the 2013 melt season, one DEM (TanDEM-X December 26, 2016) shows a collapse-basin feature with a depth of 10 m, while all others (indicated with light red area in Fig. 8d(d)) show a relatively flat surface at pre-collapse elevations. These observed flat surfaces could be is the result of the collapse-basin filling with water, and
- 380 this was indeed surface water, as confirmed by optical images close in time to the DEMsin the red box. At Lake 3(Fig. 9b), the earliest TanDEM-X observations from January 20, 2011, show a surface depression with a maximum depth of approximately 20 m. As it was the case at (Fig. 9(b)). Similar to Lake 2 most DEMs available at Lake 3 show no surface depression after the 2013 melt season (Fig. 9b and d(b) and (d)), and only two near-coincident ArcticDEMs from the summer of 2015 show a small surface depression of approximately 10 m.
- Figure 10 shows our results over Lake 4 on the Flade Isblink ice cap Ice Cap in Northeast Greenland. For this collapse basin, the largest of those analysed, several CS2 crossings provide elevation measurements of the collapse basin, and also several TanDEM-X DEMs are available. A TanDEM-X DEM from January and CS2 data from February, 2011 show a flat ice surface in the region of interest. The first data set to observe the collapse basin is CS2 point swath data from late November, 2011, and it reveals showing a relatively flat bottom of a collapse basin with a depth of approximately 95 m. The following data set is
- 390 also from CS2 data (February, 2012), and it reveals an upward movement of the collapse basin floor of more than 5 m since the previous measurements, three months earlier. As the collapse basin is filled over time, we observe the development of a dome shape at the base of the collapse basin. By April 2017, the last available DEM data (a TanDEM-X), shows that the height of the top of the dome was 20 meters from reaching the pre-collapse surface elevation. The filling rate of the collapse basin

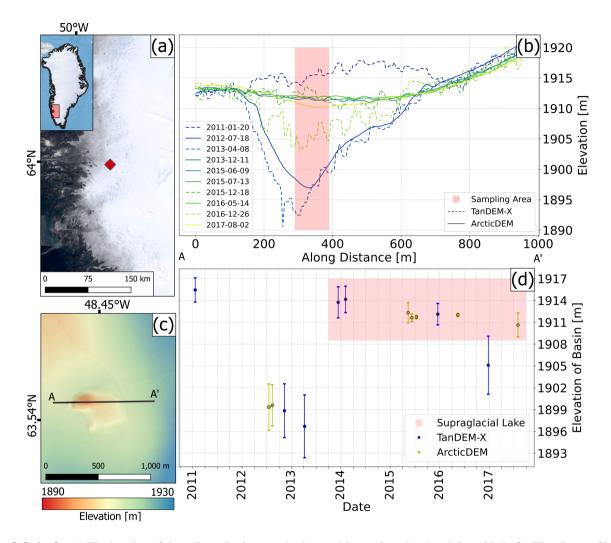


Figure 8. Lake 2 : (a) The location of the collapse basin <u>-on a background image from Landsat-8 from 2019</u>. (b) Elevation profiles from the two aligned data sets, TanDEM-X is the dashed line and ArcticDEM is the solid line. (c) The collapse basin as seen in an ArcticDEM, 18th of July 2012. Line A-A' is the profile shown in (b). (d) Time series of the deepest point of the collapse basin from TanDEM-X (blue) and ArcticDEM (yellow). The light red box marks those estimates where surface water was present.

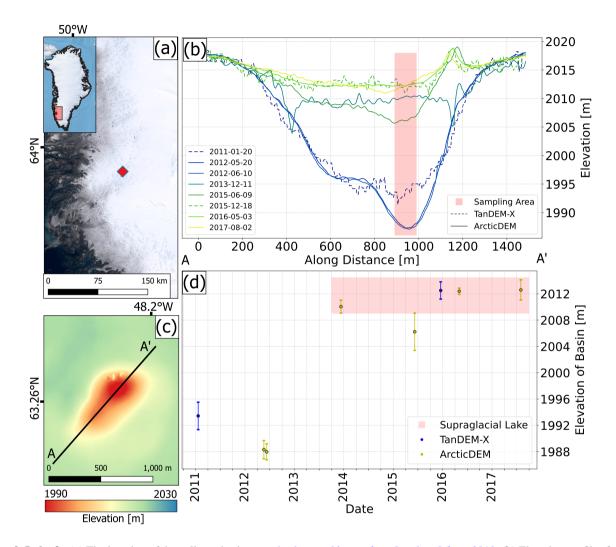


Figure 9. Lake 3 : (a) The location of the collapse basin – on a background image from Landsat-8 from 2019. (b) Elevation profiles from the two aligned data sets, TanDEM-X is the dashed line and ArcticDEM is the solid line. (c) The collapse basin as seen in an ArcticDEM, 20th of May –2012. Line A-A' is the profile shown in (b). (d) Time series of the deepest point of the collapse basin from TanDEM-X (blue) and ArcticDEM (yellow). The light red box marks those estimates where surface water was present.

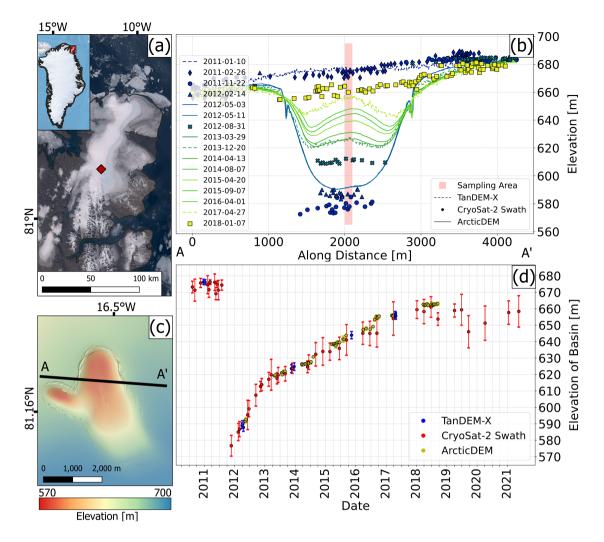


Figure 10. Lake 4 : (a) The location of the collapse basin -on a background image from Landsat-8 from 2019. (b) Elevation profiles from the three aligned data sets, ; TanDEM-X is the dashed line, CryoSat-2 is the dotted line data represented by symbols and ArcticDEM is the solid line. c) The collapse basin as seen in an ArcticDEM, 3rd of May 2012. Line A-A' is the profile used in Figure (b). (d) Time series of the deepest point of the collapse basin from TanDEM-X (blue), CS2 (red) and ArcticDEM (yellow). The depth is based on sampling within the area indicated by red in (b)

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changes over time, with a faster rate of elevation change (\sim 38 m/yr) immediately after the collapse (November, 2011 - March, 2013) than in the following years, where the rate of elevation change of the deepest point has decreased to \sim 11 m/yr (for the period after March 2013). To maintain a visually clear plot, not all data sets are shown in Figure 10b(b). Figure 10d-(d) shows the temporal evolution of the deepest point of the collapse basin until 2021. CS2 swath-processed data after 2017 indicate that Lake 4 drains again in the summer period of 2019, creating a negative elevation change of \sim 12 m. The lake seems to quickly recharge from this drainage during the melt season of 2020.

400 5.1 Subglacial Lake Volumes

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Following the methodology outlined in Sect. ?? we estimate the temporal changes in the subglacial lake storage. Figure ??a shows the estimated volume changes of Lake 1. We find that the lake had a volume of $0.013 \text{ km}^3 \pm 0.001 \text{ km}^3$ prior to the drainage event assuming that the lake completely emptied during the drainage (i.e. reaching a volume of 0 km^3). After the drainage event in summer 2011, a rapid recharge occurs in the 2012 melt season reaching a volume of $0.008 \text{ km}^3 \pm 0.004 \text{ km}^3$. The lake volume stagnates at ~ $0.010 \text{ km}^3 \pm 0.002 \text{ km}^3$ after 2014.

Figure ??b shows the volume change calculations for Lake 2. The lake had a maximum volume of $\sim 0.0006 \text{ km}^3 \pm 0.0005 \text{ km}^3$ in January, 2011. After the drainage in 2011 or 2012 we see a slow recharge, culminating with a lake volume of $\sim 0.0003 \text{ km}^3 \pm 0.0003 \text{ km}^3$ in December, 2016. Figure ??c shows the volume calculations from Lake 3. We do not have data before the drainage, and we therefore assume that the lake volume starts at 0 in January 2011. Our data show a slow recharge over

- 410 the next four years reaching a volume of $\sim 0.0028 \text{ km}^3 \pm 0.00005 \text{ km}^3$. Lake 4 has a maximum volume of $\sim 0.3 \text{ km}^3 \pm 0.02 \text{ km}^3$ before the lake drained in the summer period of 2011 (Fig. **??**d). The lake recharged to half its volume in 2 years reaching $\sim 0.15 \text{ km}^3 \pm 0.02 \text{ km}^3$ in 2013. The recharge then slowed down reaching $\sim 0.25 \text{ km}^3 \pm 0.02 \text{ km}^3$ in 2018. The lake partially drained again in the summer period of 2019, shrinking to a volume of $\sim 0.21 \text{ km}^3 \pm 0.015 \text{ km}^3$, before recharging to $\sim 0.25 \text{ km}^3 \pm 0.015 \text{ km}^3$, before recharging to $\sim 0.25 \text{ km}^3 \pm 0.015 \text{ km}^3$, before recharging to $\sim 0.25 \text{ km}^3 \pm 0.015 \text{ km}^3$.
- 415 Subglacial lake volume : (a) time series of Lake 1 volume estimates shown with the red line, and DEM estimates with blue points. (b) Lake 2 volume estimations based on DEMs, shown in the blue points c) Lake 2 volume estimations based on DEMs, shown in the blue points c) Lake 2 volume estimates with blue shown in the blue points. (d) time series of Lake 4 volume estimates shown with the red line, and DEM estimates with blue points. We assume that the lakes drain completely, and the lake volumes are therefore set to zero immediately after the first drainage event at each site.

420 6 Discussion

6.1 Lake 1

Over Lake 1 the swath-processed CS2 data and the TanDEM-X data provide new insight into the temporal evolution of the collapse basin. The CS2 data agree within the error margins with near-coincident ArcticDEM and TanDEM-X elevations, giving us confidence of in its validity. The availability of the CS2 and TanDEM-X data greatly increases the temporal resolution
of data. Compared to previous studies (Palmer et al., 2015; Howat et al., 2015) based on optical imagery with less than one observation per year we are now able to extract multiple observations for each year, and our results confirm the timing of the drainage event of the subglacial lake. Notably, the addition of CS2 observations during the winter 2011/2012 allows us to conclude that no significant substantial recharge of the subglacial lake occurred during this period, in the second half of 2011, while recharge is observed throughout 2012. The fact that the rate of recharge increased during the following summer
confirms the hypothesis proposed by Palmer et al. (2015) and Howat et al. (2015) that the subglacial lake is primarily driven by

surface meltwater drained to the bedrock bed through moulins during the melt season. We further hypothesize that the infilling

of the collapse basin after 2014/2015 is likely primarily caused by snowfall and ice flow , and not and deformation, and less by recharging of the subglacial lake . This is supported by the observed decrease in filling rate of the collapse basin after 2013, and the due to the fact that the center centre of the collapse basin moves away from the subglacial lake as a result of local location

435 of the subglacial lake due to ice flow. This also agrees with model estimates of basal melt rates and subglacial catchment size indicating that small volumes of subglacial water flows into the site of the subglacial lake(see Append ??) The uplift rate due to filling usually slows over time due to the geometry of the lake. This kind of glacier response was further modelled and investigated by Aðalgeirsdóttir et al. (2000), where the vatnajökull ice cap Vatnajökull Ice Cap in Iceland showed a similar modelled response after a subglacial drainage event created a surface depression.

440 6.2 Lakes 2 and 3

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It was not possible to successfully obtain any CS2 swath-processed data over Lake 2 and Lake 3. This is likely due to their small size (collapse basin area_areas less than 0.6 km²)that, which does not allow for adequate waveform signals with a coherent phase difference. However, with the availability of several TanDEM-X DEMs covering both lakes, we are able to augment existing observations from Bowling et al. (2019) and gain further insights into their temporal characteristics. Previous studies that relied exclusively on ArcticDEMs could not conclude on the timing of the drainage event over Lake 2 (Bowling et al.,

- 2019), but the addition of TanDEM-X scenes reveals that the drainage event did not occur earlier than January 2011. Optical imagery indicates that surface water is present in the collapse basins several times after the subglacial lakes drain. At lake Lake 2 Landsat-8 imagery from sprintspring/summer 2015 shows first a flat surface followed by water presence, indicating a frozen supraglacial lake that melts, which then drains in the summer period of 2016. This is clear in Figure 8(b) where a flat surface
- 450 is present in the elevation profiles in 2015, which then shows a collapse basin in the TanDEM-X profile from late 2016. At lake Lake 3 Landsat-8 images from autumn 2013 and spring 2014 show a flat surface, also suggesting a frozen supraglacial lake, which then melts and drains during summer 2014. This is also clear in the elevation profiles in Figure 8 where the acquisition from late 2013 shows a flat surface, but then in spring 2015 the collapse basin is visible again. Based on this finding, we suggest that the elevation profiles from the DEMs highlighted with a light red box in both Figure 8(d) and 9(d) map the height of a supraglacial lake forming in the surface depressions, rather than the actual surface depressions. Disregarding those estimates that we believe are associated with surface water, we conclude that the collapse basins over Lake 2 and Lake 3

have not completely filled up by the latest available measurements (late 2016 for Lake 2 and mid 2015 mid-2015 for Lake 3).

6.3 Lake 4

Willis et al. (2015) reported that the collapse basin over Lake 4 had a depth of approximately 70 m on May 3rd, 2012, and
TanDEM-X DEMs processed for this study indeed confirm this estimate (Fig. 10b). However, with (b)). With the inclusion of CS2 data from November 2011, we can furthermore conclude that the collapse basin has been at least 95 m deep prior to May 2012. In fact, we suggest 2012, suggesting that the collapse basin likely was more than 100 m deep at the time of the collapse. We arrive at this depth by using the observed average rate of infilling during the period November 2011 to March 2013 (~30)

 $\underline{m/yr}$), and then assuming that this infilling rate is representative for <u>of</u> the period from the collapse in August/September 2011 (Willis et al., 2015) to our first post-collapse measurement in November 2011.

Willis et al. (2015) also showed an uplift rate of ~ 9 m/yr based on three near coinciding ArcticDEM scenes in May 2012. With the inclusion of CS2 in the analysis, we find that the uplift rate appears to be significantly-larger, as we find a relatively constant uplift rate of ~ 32 m/yr in the period November 2011 - end 2012. This indicates that a significant substantial level of lake infilling happened within a year of the drainage event indicating that meltwater is readily available at this site. The lake appears to fill also-during winter months, indicating that at least some of the input is basal meltwater.

The, although the study by Liang et al. (2022) argues that the long-term recharge of the lake is dominated by a seasonal influx of surface meltwater. The high-resolution data sets gathered here also give insights into the physical processes driving the refilling of a subglacial lake collapse basin. At Flade Isblink Ice Cap (Lake 4), we observe a dome forming in the central part of the collapse basin as it refills. The CS2 data also observed a ~15 meters drop in elevation (Fig. 10(d)) between May

475 31 and August 24, 2019, indicating that Lake 4 partially drained a second time, but that the event is substantially smaller than that in 2011. The surface lowering in 2019 is also documented from ICESat-2 data by Liang et al. (2022), who identified it as a drainage eventwhich shortly affected the local horizontal ice velocity. They also found that this event caused the ice velocity downstream from the lake to abruptly but briefly increase. Further investigations into available melt water meltwater sources and local ice cap settings are necessary in order to understand why Lake 4 did not drain completely in 2019.

480 6.4 Data Limitations

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We are able to retrieve CS2 swath-processed elevation data over Lake 1 and Lake 4 prior to drainage, immediately after the drainage, and during the following recharge period. In Figure 7(d) and 10(d) the CS2 swath data provides better temporal coverage than TanDEM-X and ArcticDEM just after the collapse. When applying the CS2 swath processing, careful filtering based on coherence, power and bin number is <u>done. We speculate applied. We suggest</u> that when the surface depressions are

- filled over time, the signal from the bottom does not stand out as clearly in the waveform because it is located closer to the surface returns in the waveform. This could indicate that CS2 swath processing is useful to detect primarily the bottom of only the relatively deep surface depressions with a sizable area, and may explain why it was not possible to obtain data over Lakes 2 and 3, which are not as deep or large as Lake 1 and 4. The fact that the bottom of the collapse basins is are flat at Lake 1 and Lake 4 further aids the elevation retrievals immediately after the collapse, because the flat bottom makes the surface a better
- 490 reflector for the radar. Due to the need for careful filtering and tuning of the CS2 swath processor, the CS2 data is not ideal for finding and locating subglacial lake activity, since the analysis is dependent on available DEMs. We find that another limiting factor for the success of CS2 swath processing for subglacial lake mapping is that the satellite track must pass directly over the surface depression for bottom echoes to be retrieved. If the surface feature is located too far off-nadir, we do not obtain any valuable data. This limits the use of CS2 to very specific cases of subglacial lake activity, however as seen above, we do find
- 495 useful satellite crossings for well-developed collapse basins.

The interferometric X-band elevation data from TanDEM-X suffer from penetration of the X-band SAR signal into the snow and ice surface by several meters, which depends on the properties of the ice (e.g. density, grain size, and dielectric properties)

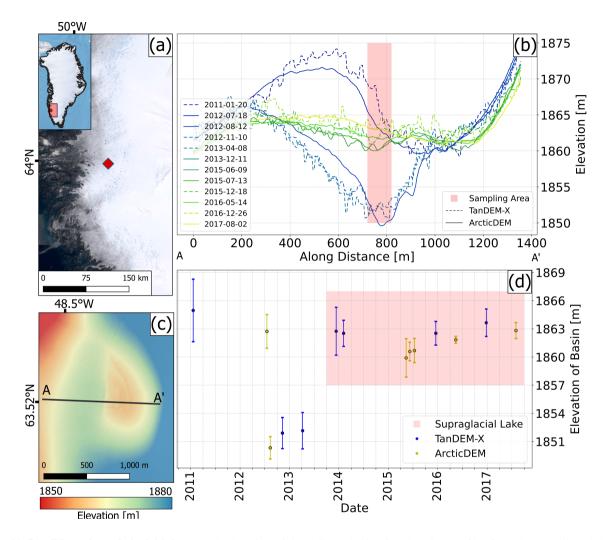


Figure 11. Possible active subglacial lake: (**a**) The location of the collapse basin. (**b**) Elevation profiles from the two aligned data sets, TanDEM-X is the dashed line and ArcticDEM is the solid line. (**c**) Possible collapse basin as seen in an ArcticDEM, 12th of August 2012. Line A-A' is the profile used in Figure (b). (**d**) Time series of the deepest point of the lake basin from TanDEM-X (blue) and ArcticDEM (yellow).

and on InSAR geometry parameters (Abdullahi et al., 2019). Here, we circumvent the penetration bias effect by examining solely height differences within TanDEM-X DEM scenes. Also the CS2 signal may penetrate into the snow, with a penetration

- 500 depth different from that of the X-band data. To minimize the elevation changes caused by different surface penetration as well as the elevation changes caused by surface mass balance, we vertically align all the data sets at the rim of the collapse basins (Sect. 4.1). This implies an assumption that the surface mass balance and the ice properties within the collapse basin are the same as at the rim. This is likely associated with an error since e.g. snow conditions in the depression may differ from those on the rimaround it. Additionally, the surface penetration depth of both CS2 and TanDEM-X can vary spatially, and
- 505 <u>could potentially be different in the collapse basin from the surrounding areas</u>. At present, we do not include this error in our estimates, since we do not have the means to quantify it.

The high-resolution data sets gathered here also give insights into the physical processes driving the refilling of a subglacial lake collapse basin. At Flade Isblink ice cap (Lake 4), we observe a dome forming in the central part of the collapse basin as it refills. This formation suggests a highly active hydrological system at the base of the ice cap, where the low pressure

510 of the central parts of the collapsed basin causes an influx of meltwater which exerts sufficient force to push the central part upwardssteep and deep basins could lead to phase unwrapping errors in side-looking InSAR. In this study, we checked the InSAR elevations individually to avoid the use of phase unwrapping error distortioned DEMs.

We do not include ICESat-2 data satellite laser altimetry in this study as the main goal has been to densify the time series covered by the CS2 mission, but we acknowledge that this sensor provides an obvious dataset for future monitoring of subglacial lake activity (Fan et al., 2023).

6.5 Basal Melt Flux

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The derived subglacial lake volume changes can be compared to calculations of basal melt to conclude whether the lake refilling is actually driven by surface water or basal melt. As an example, we calculated the theoretical basal melt across the predicted upstream catchment of Lake 1, following the method described by Karlsson et al. (2021). The result can bee seen in Fig. ??. The drainage basin for the lake is small and we find that the basal melt flux into the lake is too small (of the order of 10^5m^3 /year) to explain the rapid recharge of the lake that we observed. The calculation of the basal melt is uncertain, and the bedrock topography might not be known in sufficient detail to conclude on the shape of the catchment. Even so, the estimates suggest that the subglacial lake is more likely to originate from surface melt water.

6.5 Hydrologically Connected Subglacial Lakes

- 525 In our analysis of Lake Lakes 2 and 3, we identified an interesting signal on the ice sheet, which to our knowledge has not been described elsewhere. Located only 3 km previously. The signal is located southwest of Lake 2, we found and appears to show indications of a new active subglacial lake (here denoted Lake 5). The location, as well as the elevation profiles from ArcticDEMs and TanDEM-X DEMs from Lake 5, are shown in Figure 11. The elevation data show a dome-shaped feature on the surface in January 2011 and July 2012. 2012 (Fig. 11(b)). Between July 18, 2012, and August 12, 2012, a rapid surface
- 530 elevation change-lowering occurs and a feature resembling a collapse basin is formed. Landsat 7 imagery (Fig. A1) confirms

the existence of this feature and details its formation between two scenes taken on The low point of the surface depression is shifted horizontally compared to the dome. To more precisely determine the timing of the formation of the collapse basin, we also investigated optical imagery from the area and found that in a Landsat-7 image from July 25, 2012, there are no indications of a collapse basin (See Appendix, Fig. A1), suggesting a lake drainage between July 25 and August 26, 2012, revealing a likely

- 535 drainage event occurring over the span of maximum 18 days. 12, 2012. This drainage event coincides with the unprecedented melt event in July 2012 across the GrIS (Nghiem et al., 2012). The low point of the surface depression is shifted horizontally compared to the dome. As this newly discovered subglacial lake Lake 5 is located approximately 3-2 km downstream of Lake 2, we suggest that this lake (see Appendix A, Fig. A2), we hypothesise that the two could be hydrologically connected to Lake 2. The timing of the event further supports this, with the. For Lake 2 we know from a TanDEM-X DEM showing a flat
- 540 surface at Lake 2 that the collapse hasn't occurred in January 2011 (Fig. 8b(b)), and Landsat-7 imagery further confirms that no collapse basin was present in the summer period of 2011 or in early Juneof does not show a collapse basin either in early June, 2012 (Fig. A3), and although the image is not very clear possibly due to snow cover we see a body of surface water that partially intersects with the outline of the collapse basin as observed in an August ArcticDEM from 2012, indicating that a local depression had not yet formed in June 2012. The following ArcticDEM and Landsat-7 image from July 18, 2012,
- 545 both show a distinct collapse basin after the draining event (Fig. 8 and A3). Therefore, we hypothesize that can constrain Lake 2 to have likely drained between June 9 and July 18, 2012, and then prompted which is prior to but temporally close to the draining of the subglacial lake 3 km downstream Lake2 between July 25 August 12, 2012. This is to our knowledge the first evidence indication of hydrologically connected subglacial lakes in Greenland, indicating that water is transferred from one lake to another following a draining event. The small size of the investigated subglacial lakes makes it difficult to assess routing pathways at the bedrock considering the limited spatial resolution of available data sets like BedMachineV4. Hydrologically
- connected lakes have already been documented in Antarctica (Wingham et al., 2006; Smith et al., 2017), and their influence on the subglacial system has been investigated (Malczyk et al., 2020). It should be noted that neither Lake 2, Lake 3 nor Lake 5 are detected in the new study by (Fan et al., 2023), likely because they are not active in the period covered by ICESat-2, which hampers further investigations of the inter-connectivity of Lake 2 and 5.

555 7 Conclusions

The importance of subglacial lakes to the hydrology of the Greenland ice sheet Ice Sheet and surrounding glaciers and ice caps is not well understood, largely due to a lack of observations. In this study, we have investigated the elevation changes over four Greenlandic subglacial lakes, which that previous studies have identified by the presence of a collapse basin at the ice surface. We demonstrate how the inclusion of CS2 swath-processed data and TanDEM-X DEMs in addition to ArcticDEMs improve

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- 50 the mapping of subglacial lake activity in Greenland, demonstrating that these and similar data sets should be included in future analyses in order to increase the temporal resolution of the observational records. The small size of the subglacial lakes in Greenland provides a challenge for satellite radar altimetry, and we are only able to recover useful CS2 data over the two largest collapse basins. The TanDEM-X mission provides a valuable additional elevation data source for all lakes throughout

the entire investigated time span (2011-2018). Both TanDEM-X and CS2 data agree well with the ArcticDEM data when we

- 565 vertically align them at the rim of the collapse basins, and it gives us confidence that they are indeed reliable. The use of TanDEM-X DEMs and CS2 data significantly substantially increase the sampling frequency over the four subglacial lake sites. Over Lake 1, the addition of CS2 data during the winter 2011/2012 showed that no significant substantial recharge of the subglacial lake occurred during this period. Previous literature did not conclude on the timing of the drainage event over Lake 2, but TanDEM-X scenes revealed that it had not drained in January 2011. CS2 data show that the initial depth
- 570 of the collapse basin over Lake 4 is \sim 35 % deeper (approximately 95 m in late November 2011) than previously found based on ArticDEMs, and that the filling rate of the collapse basin changes significantly over time. Finally, we identified a signal which we argue is evidence of that could stem from a previously undetected active subglacial lake in the vicinity of Lake 2. The improved temporal resolution also indicates that this new lake may be hydrologically connected to its upstream neighbour Lake 2, making this the first discovery of hydrologically connected lakes in Greenland.

575 Appendix A: Modelled basal flux and melt rates

The local basal conditions for Lake 1 is presented in Fig. **??**. Here, we use the estimates from Karlsson et al. (2021) and ealculate the lake catchment based on local surface and bed topography, following the method from that study. We find that the eatchment (black dashed lines in Fig. **??**) is fairly small extending only 20 km upstream of the lake location. The flux of basal melt water into the catchment is approximately 5*10⁴ m³/year and local basal melt rates are of the order of 1-2 cm per year.

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Modelled basal conditions for Lake 1 (grey lines) and its catchment (dashed black lines). (a) The flux of subglacial meltwater. (b) Basal melt rates.

Appendix A: Optical observation Drainage Times of the fifth lake Lake 2 and Lake 5

In the area of Lake 2, we used Landsat-7 imagery to establish if the timing of the draining is connected to the lake which is observed approximately $\frac{32}{2}$ km downstream.

- 585 Author contributions. LSS and RB planned the study. LSS, RB and SBS developed methodology. NG, AH, NHA and RB carried out swath processing of CS2 data. NBK assisted with geophysical interpretations. AMS analyzed ice velocities. BW calibrated and delivered the TanDEM-X data. MM, AL, JB, MM and JM contributed to the analysis and discussion of ArcticDEMs. LSS and RB wrote the manuscript with input from all the co-authors, who all discussed and revised the manuscript, and all authors have read and agreed to the published version of the manuscript.
- 590 *Competing interests.* Some authors are members of the editorial board of The Cryosphere. The peer-review process was guided by an independent editor, and the authors have also no other competing interests to declare.

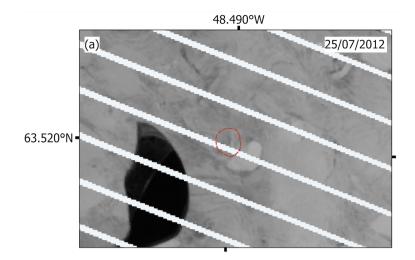


Figure A1. Landsat-7 images taken on July 25, 2012, June 9 and August 26, 2012 over Lake 5 showing no signs of the surface before and after the Lake 5 a collapse basinhad formed. The AreticDEM observed August 12, 2012, created the red outline that bounds indicated the collapse basin --outline as created from an ArcticDEM from August 12, 2012.

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https://www.pgc.umn.edu/data/arcticdem/

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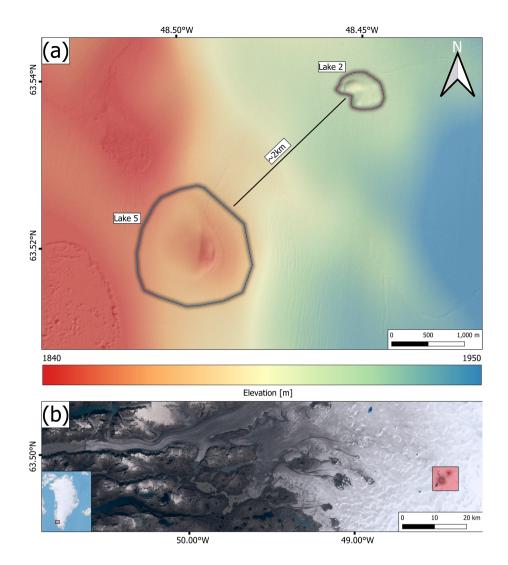


Figure A2. Landsat-7 images taken on August 18, 2011, June 9 and July 18, 2012 (a) Indication of the surface before and after the Lake 2 collapse basin had formed. The location of Lake 2 relative to Lake 5 shown on an ArcticDEMobserved July 18, 2012, created the red outline that bounds the collapse basin. On June 9, 2012, it is observed that a body (b) Overview of water partially intersects with the red outline , indicating that a local depression had not yet formed in June 2012. On July 18, 2012, a depression or a peak is observed inside the red outline, which is confirmed as the collapse basin in Figure 8two lakes' locations.

References

- Abdullahi, S., Wessel, B., Huber, M., Wendleder, A., Roth, A., and Künzer, C.: Estimating penetration-related X-band InSAR elevation bias - A study over the Greenland ice sheet, Remote Sensing, 11, 1–19, https://doi.org/10.3390/rs11242903, 2019.
- 605 Andersen, N. H., Simonsen, S. B., Winstrup, M., Nilsson, J., and Sørensen, L. S.: Regional assessments of surface ice elevations from swath-processed SARin data from CryoSat-2, Remote Sensing, 13, 1–15, https://doi.org/10.3390/rs13112213, 2021.

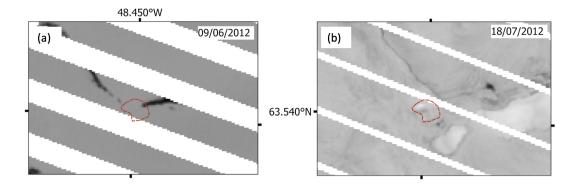


Figure A3. (a) Landsat-7 images taken on June 9, 2011 and (b) from July 18, 2012 of the surface before and after the Lake 2 collapse basin had formed. The red outline is describing the collapse basin outline as created from an ArcticDEM observed July 18, 2012

Aðalgeirsdóttir, G., Gudmundsson, G., and Björnsson, H.: The response of a glacier to a surface disturbance: a case study on Vatnajökull ice cap, Iceland, Annals of Glaciology, 31, 104–110, https://doi.org/10.3189/172756400781819914, 2000.

Bingham, R. G. and Siegert, M. J.: Radio-echo sounding over polar ice masses, Journal of Environmental and Engineering Geophysics, 12,

610 47–62, https://doi.org/10.2113/JEEG12.1.47, 2007.

- Bowling, J. S., Livingstone, S. J., Sole, A. J., and Chu, W.: Distribution and dynamics of Greenland subglacial lakes, Nature Communications, 10, 1–12, https://doi.org/10.1038/s41467-019-10821-w, 2019.
- Chandler, D. M., Wadham, J. L., Lis, G. P., Cowton, T., Sole, A., Bartholomew, I., Telling, J., Nienow, P., Bagshaw, E. B., Mair, D., Vinen, S., and Hubbard, A.: Evolution of the subglacial drainage system beneath the Greenland Ice Sheet revealed by tracers, Nature Geoscience,

615 6, 195–198, https://doi.org/10.1038/ngeo1737, 2013.

- Evans, S. and Smith, B. M. E.: Radio echo exploration of the Antarctic ice sheet, 1969–70, Polar Record, 15, 336–338, https://doi.org/10.1017/s0032247400061143, 1970.
- Fan, Y., Ke, C.-Q., Shen, X., Xiao, Y., Livingstone, S. J., and Sole, A. J.: Subglacial lake activity beneath the ablation zone of the Greenland Ice Sheet, The Cryosphere, 17, 1775–1786, https://doi.org/10.5194/tc-17-1775-2023, 2023.
- 620 Fischer, G., Papathanassiou, K. P., and Hajnsek, I.: Modeling and Compensation of the Penetration Bias in InSAR DEMs of Ice Sheets at Different Frequencies, IEEE J. Sel. Topics Appl. Earth Observ. Remote Sens., 13, 2698–2707, https://doi.org/10.1109/JSTARS.2020.2992530, 2020.
 - Foresta, L., Gourmelen, N., Pálsson, F., Nienow, P., Björnsson, H., and Shepherd, A.: Surface elevation change and mass balance of Icelandic ice caps derived from swath mode CryoSat-2 altimetry, Geophysical Research Letters, 43, 12,138–12,145,
- 625 https://doi.org/https://doi.org/10.1002/2016GL071485, 2016.
 - Fricker, H. A., Scambos, T., Bindschadler, R., and Padman, L.: An active subglacial water system in West Antarctica mapped from space, Science, 315, 1544–1548, 2007.
 - Gourmelen, N., Escorihuela, M. J., Shepherd, A., Foresta, L., Muir, A., Garcia-Mondéjar, A., Roca, M., Baker, S. G., and Drinkwater, M. R.: CryoSat-2 swath interferometric altimetry for mapping ice elevation and elevation change, Advances in Space Research, 62, 1226–1242,
- 630 https://doi.org/10.1016/j.asr.2017.11.014, 2018a.

- Gourmelen, N., Escorihuela, M. J., Shepherd, A., Foresta, L., Muir, A., Garcia-Mondéjar, A., Roca, M., Baker, S. G., and Drinkwater, M. R.: CryoSat-2 swath interferometric altimetry for mapping ice elevation and elevation change, Advances in Space Research, 62, 1226–1242, https://doi.org/10.1016/j.asr.2017.11.014, 2018b.
- Gray, L., Burgess, D., Copland, L., Cullen, R., Galin, N., Hawley, R., and Helm, V.: Interferometric swath processing of Cryosat data for 635 glacial ice topography, The Cryosphere, 7, 1857–1867, https://doi.org/10.5194/tc-7-1857-2013, 2013.
 - Howat, I. M., Porter, C., Noh, M. J., Smith, B. E., and Jeong, S.: Brief communication: Sudden drainage of a subglacial lake beneath the Greenland Ice Sheet, Cryosphere, 9, 103–108, https://doi.org/10.5194/tc-9-103-2015, 2015.
 - Joughin, I., Smith, B. E., Howat, I. M., Scambos, T., and Moon, T.: Greenland flow variability from ice-sheet-wide velocity mapping, Journal of Glaciology, 56, 415-430, https://doi.org/10.3189/002214310792447734, 2010.
- Joughin, L. Smith, B. E., Howat, I. M., and Scambos, T.: MEaSUREs Greenland Ice Sheet Velocity Map from InSAR Data, Version 2, 0478, 640 [20190807], https://doi.org/10.5067/OC7B04ZM9G6O, 2015.
 - Karlsson, N. B., Solgaard, A. M., Mankoff1, K. D., Gillet-Chaulet, F., MacGregor, J. A., Box, J. E., Citterio, M., Colgan, W. T., Larsen, S. H., Kjeldsen, K. K., Korsgaard, N. J., Benn, D. I., Hewitt, I. J., and Fausto, R. S.: A first constraint on basal melt-water production of the Greenland ice sheet, Nature Communications, 12, 1-10, https://doi.org/10.1038/s41467-021-23739-z, 2021.
- 645 Lachaise, M., Fritz, T., and Bamler, R.: The Dual-Baseline Phase Unwrapping Correction framework for the TanDEM-X Mission Part 1: Theoretical description and algorithms, IEEE T. Geosci. Remote, 56, 780-798, https://doi.org/10.1109/TGRS.2017.2754923, 2018.
 - Liang, O., Xiao, W., Howat, I., Cheng, X., Hui, F., Chen, Z., Jiang, M., and Zheng, L.: Filling and drainage of a subglacial lake beneath the Flade Isblink ice cap, northeast Greenland, The Cryosphere Discussions, pp. 1–17, 2022.
- Livingstone, S. J., Clark, C. D., Woodward, J., and Kingslake, J.: Potential subglacial lake locations and meltwater drainage pathways beneath 650 the Antarctic and Greenland ice sheets, Cryosphere, 7, 1721–1740, https://doi.org/10.5194/tc-7-1721-2013, 2013.
 - Livingstone, S. J., Sole, A. J., Storrar, R. D., Harrison, D., Ross, N., and Bowling, J.: Brief communication: Subglacial lake drainage beneath Isunguata Sermia, West Greenland: geomorphic and ice dynamic effects, The Cryosphere, 13, 2789–2796, 2019.
 - Livingstone, S. J., Li, Y., Rutishauser, A., Sanderson, R. J., Winter, K., Mikucki, J. A., Björnsson, H., Bowling, J. S., Chu, W., Dow, C. F., et al.: Subglacial lakes and their changing role in a warming climate, Nature Reviews Earth & Environment, pp. 1–19, 2022.
- 655 Magnússon, E., Rott, H., Björnsson, H., and Pálsson, F.: The impact of jökulhlaups on basal sliding observed by SAR interferometry on Vatnajökull, Iceland, Journal of Glaciology, 53, 232-240, 2007.
 - Malczyk, G., Gourmelen, N., Goldberg, D., Wuite, J., and Nagler, T.: Repeat Subglacial Lake Drainage and Filling Beneath Thwaites Glacier, Geophysical Research Letters, 47, https://doi.org/10.1029/2020GL089658, 2020.

Meloni, M., Bouffard, J., Parrinello, T., Dawson, G., Garnier, F., Helm, V., Di Bella, A., Hendricks, S., Ricker, R., Webb, E., Wright, B.,

- Nielsen, K., Lee, S., Passaro, M., Scagliola, M., Simonsen, S. B., Sandberg Sørensen, L., Brockley, D., Baker, S., Fleury, S., Bamber, J., 660 Maestri, L., Skourup, H., Forsberg, R., and Mizzi, L.: CryoSat Ice Baseline-D validation and evolutions, The Cryosphere, 14, 1889–1907, https://doi.org/10.5194/tc-14-1889-2020, 2020.
 - Morlighem, M., Williams, C., Rignot, E., An, L., Bamber, J., Chauche, N., Dowdeswell, J., Dorschel, B., Holland, D., Holland, D., et al.: BedMachine v3: Complete bed topography and ocean bathymetry mapping of Greenland from multi-beam radar sounding combined with mass conservation, 2017.

665

Nghiem, S. V., Hall, D. K., Mote, T. L., Tedesco, M., Albert, M. R., Keegan, K., Shuman, C. A., DiGirolamo, N. E., and Neumann, G.: The extreme melt across the Greenland ice sheet in 2012, Geophysical Research Letters, 39, https://doi.org/10.1029/2012GL053611, 2012.

Noël, B., Van De Berg, W. J., Van Meijgaard, E., Kuipers Munneke, P., Van De Wal, R. S. W., and Van Den Broeke, M. R.: Evaluation of the updated regional climate model RACMO2.3: Summer snowfall impact on the Greenland Ice Sheet, Cryosphere, 9, 1831–1844,

670 https://doi.org/10.5194/tc-9-1831-2015, 2015.

Noh, M. J. and Howat, I. M.: Automatic relative RPC image model bias compensation through hierarchical image matching for improving DEM quality, ISPRS Journal of Photogrammetry and Remote Sensing, 136, 120–133, https://doi.org/10.1016/j.isprsjprs.2017.12.008, 2018.

Palmer, S., Mcmillan, M., and Morlighem, M.: Subglacial lake drainage detected beneath the Greenland ice sheet, Nature Communications,

675 6, https://doi.org/10.1038/ncomms9408, 2015.

- Palmer, S. J., Dowdeswell, J. A., Christoffersen, P., Young, D. A., Blankenship, D. D., Greenbaum, J. S., Benham, T., Bamber, J., and Siegert, M. J.: Greenland subglacial lakes detected by radar, Geophysical Research Letters, 40, 6154–6159, https://doi.org/10.1002/2013GL058383, 2013.
- Pattyn, F.: Investigating the stability of subglacial lakes with a full Stokes ice-sheet model, Journal of Glaciology, 54, 353–361,
- 680 https://doi.org/10.3189/002214308784886171, 2008.
 - Porter, C., Morin, P., Howat, I., Noh, M.-J., Bates, B., Peterman, K., Keesey, S., Schlenk, M., Gardiner, J., Tomko, K., Willis, M., Kelleher, C., Cloutier, M., and Husby, E, M.: ArcticDEM [dataset accessed 03-01-2020], Tech. rep., Polar Geospatial Center (PGC), Harvard Dataverse, Minneapolis, Minnesota, USA, https://doi.org/10.7910/DVN/OHHUKH, 2018.

Ridley, J. K., Cudlip, W., and Laxon, S. W.: Identification of subglacial lakes using ERS-1 radar altimeter, Journal of Glaciology, 39, 625–634,

685 https://doi.org/10.1017/S002214300001652X, 1993.

Rizzoli, P., Martone, M., Gonzalez, C., Wecklich, C., Borla Tridon, D., Bräutigam, B., Bachmann, M., Schulze, D., Fritz, T., Huber, M., Wessel, B., Krieger, G., Zink, M., and Moreira, A.: Generation and performance assessment of the global TanDEM-X digital elevation model, ISPRS J. Photogramm. Remote Sens., 132, 119–139, https://doi.org/10.1016/j.isprsjprs.2017.08.008, 2017.

Rossi, C., Rodriguez Gonzalez, F., Fritz, T., Yague-Martinez, N., and Eineder, M.: TanDEM-X calibrated Raw DEM generation, ISPRS J.

690 Photogramm. Remote Sens., 73, 12–20, https://doi.org/10.1016/j.isprsjprs.2012.05.014, 2012.

- Rott, H., Scheiblauer, S., Wuite, J., Krieger, L., Floricioiu, D., Rizzoli, P., Libert, L., and Nagler, T.: Penetration of interferometric radar signals in Antarctic snow, The Cryosphere, 15, 4399–4419, https://doi.org/10.5194/tc-15-4399-2021, 2021.
- Siegert, M. J., Ross, N., and Le Brocq, A. M.: Recent advances in understanding Antarctic subglacial lakes and hydrology, Philosophical Transactions of the Royal Society A: Mathematical, Physical and Engineering Sciences, 374, 20140306, https://doi.org/10.1098/rsta.2014.0306, 2016.
 - Smith, B. E., Gourmelen, N., Huth, A., and Joughin, I.: Connected subglacial lake drainage beneath Thwaites Glacier, West Antarctica *f*, The Cryosphere, 11, 451–467, 2017.

- 700 Solgaard, A., Kusk, A., Merryman Boncori, J. P., Dall, J., Mankoff, K. D., Ahlstrøm, A. P., Andersen, S. B., Citterio, M., Karlsson, N. B., Kjeldsen, K. K., et al.: Greenland ice velocity maps from the PROMICE project, Earth System Science Data, 13, 3491–3512, 2021.
 - Stearns, L., Smith, B., and Hamilton, G.: Increased flow speed on a large East Antarctic outlet glacier caused by subglacial floods, Nature Geosci, pp. 827–831, https://doi.org/10.1038/ngeo356, 2008.

Stubblefield, A., Creyts, T., Kingslake, J., Siegfried, M., and Spiegelman, M.: Surface Expression and Apparent Timing of Subglacial Lake

705 Oscillations Controlled by Viscous Ice Flow, Geophysical Research Letters, 48, e2021GL094 658, 2021.

Solgaard, A. and Kusk, A.: Greenland Ice Velocity from Sentinel-1 Edition 2, https://doi.org/10.22008/promice/data/sentinel1icevelocity/greenlandiceshee 2021.

- Tulaczyk, S. M. and Foley, N. T.: The role of electrical conductivity in radar wave reflection from glacier beds, The Cryosphere, 14, 4495–4506, 2020.
- Wessel, B., Bertram, A., Gruber, A., Bemm, S., and Dech, S.: A new high-resolution elevation model of Greenland derived from TanDEM-X, ISPRS Ann. of Photogramm. Remote Sens. Spatial Inf. Sci., 3, 9–16, 2016.
- 710 Willis, M. J., Herried, B. G., Bevis, M. G., and Bell, R. E.: Recharge of a subglacial lake by surface meltwater in northeast Greenland, Nature, 518, 223–227, https://doi.org/10.1038/nature14116, 2015.
 - Wingham, D. J., Francis, C. R., Baker, S., Bouzinac, C., Brockley, D., Cullen, R., de Chateau-Thierry, P., Laxon, S. W., Mallow, U., Mavrocordatos, C., Phalippou, L., Ratier, G., Rey, L., Rostan, F., Viau, P., and Wallis, D. W.: CryoSat: A mission to determine the fluctuations in Earth's land and marine ice fields, Advances in Space Research, 37, 841–871, https://doi.org/10.1016/j.asr.2005.07.027, 2006.
- Wright, A. and Siegert, M.: A fourth inventory of Antarctic subglacial lakes, Antarctic Science, 24, 659–664, 2012.
 https://www.overleaf.com/project/61a892205046b7294c26f040