



1 Inter-annual variation of lake ice composition in European Arctic: observations based on high-

- 2 **resolution thermistor strings**
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15 Abstract

Climate change and global warming strongly impact the cryosphere. The rise of air temperature and 16 change of precipitation patterns lead to dramatic responses of snow and ice heat and mass balance. 17 18 Sustainable field observations on lake air-snow-ice-water temperature regime have been carried out in Lake Orajärvi in the vicinity of the Finnish Space Centre, a Flagship Supersite in Sodankylä in 19 20 Finnish Lapland since 2009. A thermistor string-based snow and ice mass balance buoy called "Snow and ice mass balance apparatus (SIMBA)" was deployed in the lake at the beginning of each ice 21 22 season. In this paper, we describe snow and ice temperature regimes, snow depth, ice thickness, and 23 ice compositions retrieved from SIMBA observations as well as meteorological variables based on 24 high-quality observations at the Finnish Space Centre. Ice thickness in Lake Orajärvi showed an 25 increasing trend. During the decade of data collection: 1) The November-May mean air temperature had an increasing trend of 0.16° C/year, and the interannual variations were highly correlated (r = 26 27 0.93) with the total seasonal accumulated precipitation; 2) The maximum granular ice thickness





ranged from 15 to 80% of the maximum total ice thickness; 3) The snow depth on lake ice was not correlated (r = 0.21) with the total precipitation. The data set can be applied to investigate the lake ice surface heat balance and the role of snow on lake ice mass balance, and to improve the parameterization of snow to ice transformation in snow/ice models. The data are archived at <u>https://zenodo.org/record/4559368#.YIKOOpAzZPZ</u> (Cheng et al., 2021)

33

34 1. Introduction

35

The rapid climate warming in the Arctic (Box et al., 2019; Przybylak and Wyszyński, 2020) has also affected lakes, in particular lake surface temperatures and lake ice phenology (Woolway, et al., 2019). In the Northern Hemisphere, the lake ice season has become shorter and lake ice has become thinner, and these trends are projected to continue throughout the 21st century (Sharma, et al., 2019). Lakes are important in the Earth system, as they can adjust local climate (Brown and Duguay, 2010), and affect the environment through interactions among physical, hydrological, biological, and chemical processes (Leppäranta, 2010).

Observations on snow depth and lake ice thickness are needed for (a) monitoring of climate variability 43 and trends (Filazzola et al., 2020), (b) practical applications, such as use of lake ice for winter fishing, 44 transport, and recreational activities (Leppäranta, 2015), and (c) to provide initial conditions for 45 operational forecasting (Anderson et al., 2018). Snow depth and lake ice thickness can be measured 46 manually. For example, in Finland, lake ice thickness is measured via manual drilling in a single 47 location in 45 lakes with ten-day intervals throughout the ice season. However, this requires a lot of 48 49 manpower, and accordingly does not allow collection of time series with a better spatial and temporal resolution. During recent decades, the number manual observations has strongly declined in many 50 countries (Duguay et al., 2006). Satellite remote sensing yields information on lake ice cover (Wu et 51 52 al., 2021) and snow/ice surface temperature (Cheng et al., 2014) with a sufficiently high spatial and 53 temporal resolution. Kang et al., (2014) introduced a method to derive lake ice thickness from coarse 54 resolution (~10 km) passive microwave data over large lakes in Canada. However, the transferability 55 of the method to sub-pixel scale lakes has not been investigated. SAR polarimetry has shown some promise in retrieving ice depth over rivers (Mermoz et al., 2013); as fully polarimetric data is not to 56



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58 lakes is currently lacking. The SIMBA data set is potentially highly relevant for the development of land applications for 59 60 planned and existing passive microwave satellite sensors, such as the Copernicus Imaging Microwave 61 Radiometer (CIMR), new Metop multichannel radiometer sensors of EUMETSAT, ESA SMOS, NASA SMAP and Chinese sensors. Due to the inherent coarse resolution of these sensors (tens of 62 kilometers), a key issue is to acquire combined simultaneous data representing various processes in 63 lakes, in addition to surrounding land areas. As such, the SIMBA forms an integral part of the FMI 64 65 sensor network in Sodankylä. Thermistor sting-based snow and ice mass balance apparatus (SIMBA) have been applied for more 66 67 than a decade to measure snow depth, ice thickness and temperature profile from air through snow and ice to water (Jackson et al., 2013). Most of SIMBA have so far been deployed in Polar sea ice 68 (Lei et al., 2018), but also lake ice has been studied (Cheng et al., 2014; Wei et al., 2016). In this 69 paper we describe SIMBA observations from an ongoing program that started in Lake Orajärvi in 70 71 northern Finland in 2009. Supporting meteorological observations from Finnish Meteorological Institute Arctic Research Centre (FMI-ARC) are also presented. The objectives of the SIMBA 72 73 program were to evaluate the cost-effectiveness of SIMBA buoys in a remote lake environment 74 -75 _ to monitor climate variability and change as reflected in snow depth as well as lake ice thickness and composition 76 77 to investigate (a) atmospheric forcing on lake ice growth and melt, (b) the role of snow on -78 lake ice mass balance via formation of superimposed ice due to refreezing of melt water 79 and rain and formation of snow ice due to flooding under a heavy snow load, and (c) the role of granular ice in lake ice phenology 80 81 _ to develop better parameterizations of snow-to-ice transformation in numerical snow/ice 82 models. 83 **2** Observation 84 2.1 Sodankylä supersite 85

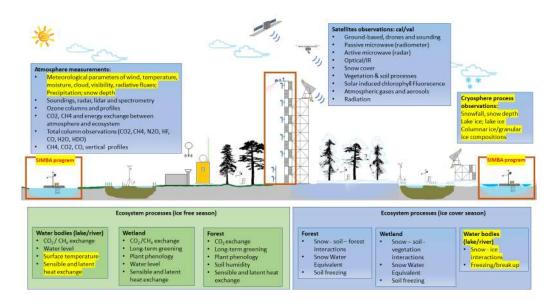
date widely available from existing SAR sensors extensive testing and application of the method for

86 The SIMBA program at Lake Orajärvi is a component of the FMI Sodankylä supersite. The Finnish





Meteorological Institute's Arctic Space Centre (FMI-ARC) in Sodankylä (67.367 °N, 26.629 °E), 87 Finland, is a super-observation site where various Earth observations (upper-air chemistry and 88 physics, atmospheric column measurements, snow and soil hydrology, biosphere-atmosphere 89 90 interaction) and ground truth measurements for satellite calibration-validation are carried out 91 continuously (Fig.1). The site is equipped with comprehensive in situ and remote sensing instrumentation placed in the forests, wetlands and freshwater bodies, which are the main landcover 92 types in the area. In this paper we focus on the cryospheric in situ observations of snow cover and 93 94 lake ice as well as meteorological parameters.



95

96 Figure 1. Schematic diagram of the FMI-ARC supersite observational systems at Sodankylä. The 97 original diagram is at <u>https://litdb.fmi.fi/</u>. The frames in red and text with yellow background 98 describes the measurements addressed in this paper.

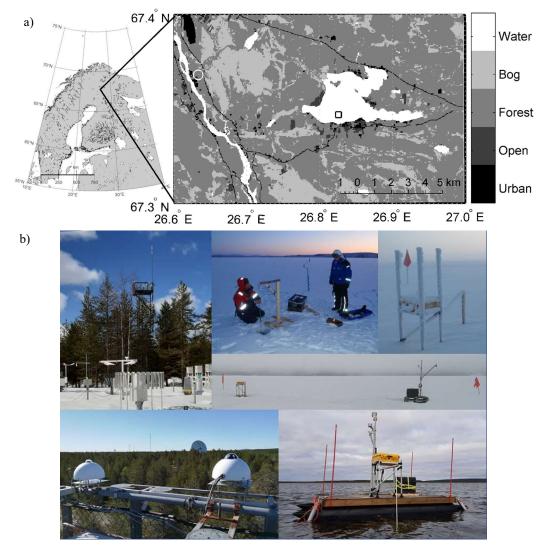
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The sub-Arctic climate and the geographic location between continental and marine climate zones result in a high inter-annual, seasonal and synoptic-scale variation in local weather conditions, enabling development of very different kinds of snowpack structures on land (Tikkanen, 2005) and snow/ice composition on lakes (Cheng et al., 2014). Lake Orajärvi is a boreal medium-sized lake located in Sodankylä municipality in the in the eastern Lapland. The lake has a surface area of about 11 km² with an average depth of 4.4 m and a maximum depth of 11 m close to the southern shore of





the lake (Fig. 2a). The estimated water volume in the lake is 0.0485 km³, and the shore length is 28 km. The lake surface elevation is 182 m above sea level. The ice season typically starts in November and lasts until May. The first snowfall typically occurs in late October, but the snow may melt during warmer autumn days. The seasonally permanent winter snow accumulation usually starts between mid-November and early-December. Snow is present on the lake ice surface every winter season.



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Figure 2. a) The location of Lak Orajärvi in Finnish Lapland and a map of lake Orajärvi and local catchment, where the open black square marks the SIMBA site and white circle is the Finnish Space





117 Centre. b) Snapshots of SIMBA deployment in Lake Orajärvi and a weather station at FMI-ARC
118 main camp. A raft was anchored in the lake in October 2019 aiming to extend the lake observations
119 beyond the ice season.

120

121 2.2 SIMBA

122 2.2.1 SIMBA program

SIMBA buoys have been deployed in Lake Orajärvi since 2009. The 2009 deployment was probably 123 124 the internationally first SIMBA application for a lake study. In each winter when ice was formed in Lake Orajärvi, one SIMBA was deployed around mid-December at the same site, 67.35° N, 26.83° E, 125 some 500 m from the shoreline. At the time of deployment, the snow depth, lake ice thickness and 126 ice freeboard were measured. A supporting frame made of fiberglass was constructed on lake ice, and 127 128 the SIMBA main control Peli case was placed on top of it (Fig. 2b). A separate wooden pole with 129 scale was standing vertically to hold the thermistor string. An ice borehole was drilled through the ice 130 layer, and the thermistor string was placed in it. The scene was left as it is, and then the thermistor string was frozen with surrounding water in the borehole. The SIMBA operated in the lake over the 131 winter and most of the spring melting season. The recovery of SIMBA took place usually in late April 132 133 but in some years as late as mid-May. Snow and ice conditions around the deployment site were 134 documented and measured before dismantling the SIMBA camp. The documentation on SIMBA 135 deployment and recovery is provided along with the SIMBA data as online files (see data availability). 136 Table 1 summarizes the SIMBA deployment and recovery status.

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Table 1. SIMBA deployment and recovery days and simultaneous manually observed snow depth (h_s) , total ice thickness (H_i) , and ice freeboard $(H_{fb}$, defined as negative if the lake water level was above the snow/ice interface) The seasonal mean values were derived from SIMBA-ET and SIMBA-HT observations. H_{gl} is the granular ice and H_{ci} is the congelation ice thickness.

Season	Deployment				Recovery					Seasonal mean \pm STD					
	Date	h_s H_i H_{fb}			Date	h_s	H _{ci}	Hi	H _{fb}	H_i	hs	H _{sfb}	H_{gi}	H _{ci}	
	DD/MM/YY	DD/MM/YY	(cm)												
2009/2010	16/12/09	07/04/10	31	54	64	5	NA								
2010/2011	SIMBA was not deployed; only manual observations every second week were available.														





2011/2012	19/12/2011	16	14	-4	12/04/2012	24	22	55	-3	38±16	22±6	5±2	15±10	23±7
2012/2013	12/12/2012	18	33	+1	25/04/2013	0	39	59	6	57±7	26±7	-5±3	4±4	53±4
2013/2014	12/12/2013	14	27	+1	30/04/2014	20	35	35	-3	49±7	17±4	-3±2	10±2	40±6
2014/2015	14/12/2014	19	30	-2.5	23/04/2015	2	35	69	4	54±11	24±7	-4±3	16±8	38±4
2015/2016	18/12/2015	18	27	-1	22/04/2016	5	30	71	6	60±16	19±7	-2±3	12±9	48±9
2016/2017	16/12/2016	8	31	-1	24/04/2017	10	38	72	4	58±13	19±6	-1±2	6±8	50±8
2017/2018	15/12/2017	25	23	-9	03/05/2018	0	28	55	6	48±15	24±6	-4±3	27±14	21±3
2018/2019	13/12/2018	15	19	-2	02/05/2019	1	20	55	6	51±17	21±7	-1±3	21±14	30±7
2019/2020	03/10/2019		-		12/05/2020	4	13	68	7	49±24	24±9	-1±3	32±20	20±5

The seasonal mean values of Hi, h_s , H_{gl} and H_{ci} were calculated by the SIMBA algorithm (Cheng et al., 2020). The seasonal mean value of ice freeboard (H_{sfb}) was calculated based on time series of snow depth (h_s), granular ice thickness (H_{gi}) and columnar ice thickness (H_i) according to the Archimedes' principle: $H_{sfb} = H_i + H_{gi} - (h_s \rho_s + H_{gi} \rho_{gi} + H_i \rho_i)/\rho_w$, where ρ_s , ρ_{gi} , ρ_i and ρ_w are seasonal mean densities of snow, granular-ice and columnar ice and lake water, assumed to be 320 kg/m³, 890 kg/m³ and 910 kg/m³ and 1000 kg/m³, respectively. The STD is the standard deviation.

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148 *2.2.2 SIMBA buoy*

SIMBA is a thermistor string-based Snow and Ice Mass Balance Apparatus. It has been developed 149 by the Scottish Association for Marine Science (SAMS) Research Services Ltd (SRSL) in UK. 150 SIMBA consists of a simple, robust thermistor string with 240 temperature sensors distributed evenly 151 152 (2 cm intervals) along a 4.8 m long heat-shrink PVC plastic sleeve coated flat white wire. White heat-153 shrink sleeve is used to minimize the possibility of solar heating of the sensors. The accuracy of the SIMBA thermistor sensor is ± 0.01 °C, which is comparable with other type of thermistor string based 154 155 IMBs (Jackson et al., 2013). Each sensor measures the environment temperature (SIMBA-ET). The 156 resolution of the thermistor sensor is 0.0625°C, i.e., smaller changes cannot be detected even if the 157 absolute accuracy of the sensor would allow it. In addition, the thermistor chain is equipped with 158 heaters, i.e. resistor components mounted next to the temperature-sensing elements. A weak voltage 159 (8 V) supply is connected to provide gentle identical heating of each sensor on the chain. After a short heating period, often 60 or 90 s, the heating temperatures (SIMBA-HT) are recorded. The SIMBA-160 HT reacts differently among the medium (air, snow, ice and water) which a sensor lies in. Therefore, 161 162 the heating temperature profiles can greatly enhance the detection of the interfaces between air, snow,





163 ice and water. The heating cycle is applied once per day. The SIMBA-HT is controlled not to disturb 164 the SIMBA-ET measurements which are carried out typically 4 times per day (Jackson et al., 2013). A SIMBA also includes a built-in GPS to record SIMBA drift positions (for sea ice applications), a 165 166 magnetometer for tilt and floe rotation, a barometer for surface air pressure, and an external sensor to 167 measure near-surface ambient air temperature. An iridium modem is applied for data transmission. 168 SIMBA has been used in various field campaigns targeting snow and ice mass balance in seasonally ice-covers in lakes (Cheng et al., 2014) and Polar Oceans (Hoppmann et al. 2015; Provost et al. 2017; 169 Lei et al. 2018, 2021). Table 1 presents a summary of SIMBA observations in Lake Orajärvi. 170 171 172 2.3 Weather station Meteorological data were collected at FMI-ARC station (67.3666°N, 26.6290°E, WMO code 02836) 173 11 km from Lake Orajärvi. The data sets include wind speed (Va), air temperature (Ta), relative 174 175 humidity (*RH*), cloudiness (*cn*), longwave (Ql) and shortwave (Qs) radiation, snow depth on land (*Hs*) and precipitation (Prec) (Table 2). The radiative fluxes were measured on a 10-m high tower above 176 177 treetops using Kipp&Zonen CM11 pyranometers (305-2800 nm) and Kipp&Zonen CG4 pyrgeometers (4500 - 42000 nm). Snow depth (Campbell Scientific SR50) and precipitation (OTT 178 Pluvio2) at ground level were also measured. All measurements were taken once a minute and 179 aggregated to 1-hour time intervals.

180 181

182 3 Data description

183 3.1 SIMBA data

The main output of a SIMBA buoy is the time series of environment (SIMBA-ET) and heating (SIMBA-HT) temperature measured at different depths from the lake water through ice and snow to air.

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188 *3.1.1 SIMBA-ET*

For each season, we have up to 241 time series of temperature (SIMBA-ET) at different depths. For those sensors located in the air, the temperature differences between the sensors are small, as the air in the lowermost 1.5 m layer mixes effectively and the sensors are close to each other. The temperatures inside snow reveal much larger vertical gradients because snow has a small thermal heat





193 conductivity. The temperature profile in ice has smaller vertical gradient compared to that in snow, 194 since the thermal heat conductivity of ice is larger than that of snow. At the ice bottom, temperature 195 is at the freezing point and gradually increases towards the lake bottom. Figure 8 shows an example 196 of seasonal SIMBA-ET. One can estimate the heat fluxes within snow and ice, and those at the air-197 snow, snow-ice, and ice-water interfaces.

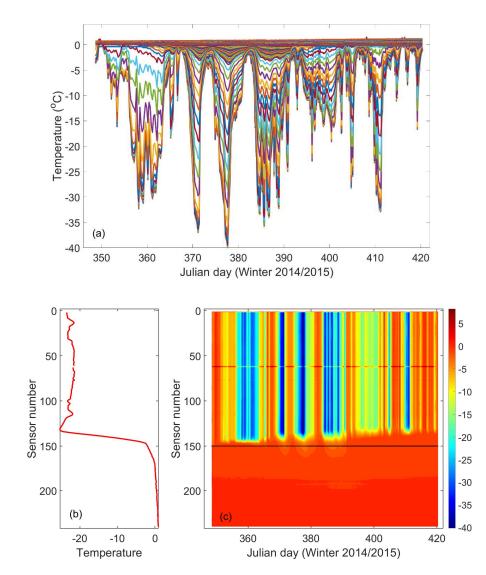




Figure 3. Illustrations of SIMBA-ET data: a) Time series of SIMBA-ET during observation period;
b) One snapshot (19 Jan 2014 08:00 UTC) of vertical SIMBA-ET profile through air-snow-lake ice-





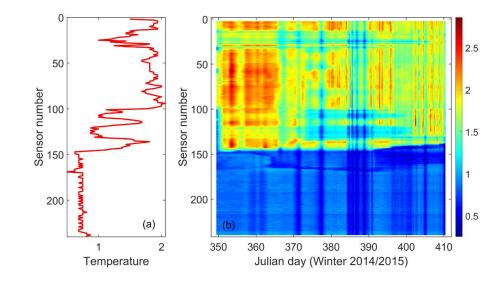
water; c) SIMBA-ET field observed by 240 sensors. Sensor 1 was placed in air and sensor 240 in

203 water.

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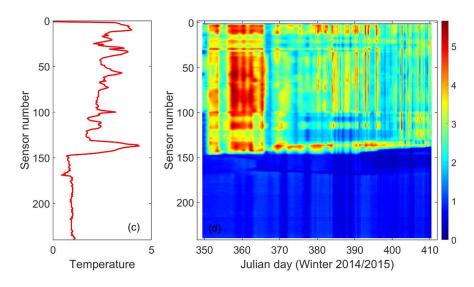
205 *3.1.2 SIMBA-HT*

SIMBA-HT shows the temperature increase in the medium each sensor was contacted during a short heating period of 60 s and 90 s. The temperature changes are largely dependent on the thermal diffusivity of the surrounding medium. Low heating power ensures that the temperature increasing will not be too high to melt snow and ice in contact with the sensor and guarantee a fast restore of environment temperature around the sensor before the next SIMBA-ET observation, and above all to minimize SIMBA power consumption. One example of SIMBA-HT is given in Figure 9.









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Figure 4. Illustrations of SIMBA-HT: a) a snapshot (25 January, 2015, 18:00 UTC) of vertical profile
of observed temperature increase after 60 s., b) SIMBA-HT (60s) field observed by 240 sensors; c)
Same as a) but after heating for 90 s., and d) SIMBA-HT (90s) field observed by 240 sensors;

217

218 3.1.3 SIMBA snow depth and ice thickness

Snow depth and ice thickness are derived from SIMBA-ET and SIMBA-HT data. A common 219 procedure is to look SIMBA-ET temperature profiles manually and identify sudden changes of 220 221 vertical temperature gradient to locate the air-snow, snow-ice and ice-water interfaces. The snow depth is then calculated as the distance between the air-snow and snow-ice interfaces, and the ice 222 thickness is the distance between the snow-ice and ice-water interfaces. However, a manual procedure 223 224 is a heavy task, especially if SIMBA operation covers long period or one would need real time SIMBA 225 results. Several studies have been carried out aiming development of an algorithm to obtain snow 226 depth and ice thickness automatically (Liao et al., 2019, Zuo, et al., 2019, Cheng et al., 2020).

Below we present an example of the application of the Cheng et al, (2020) algorithm to retrieve snow

depth and ice thickness from SIMBA data observed in Lake Orajärvi. When SIMBA was deployed,

the initial sensor position at snow-ice interface is known and we defined it as Z_{gi0} , i.e. zero reference

230 position for granular ice. During observation period, in case if initial snow-ice interface is moving

upward from Z_{gi0} , which is a common phenomenon in Arctic lakes, the distance between Z_{gi0} and

232 moving snow-ice interface is the new granular ice thickness formed by snow to ice transformation.





The depth difference between total ice thickness and granular ice thickness is the congelation ice formed at the ice bottom. Figure 5 shows the air-snow, snow-ice and ice-water interfaces with SIMBA-ET (a) and SIMBA-HT (b) as the background. For better clarity, 5-day running average can be produced as the final products.



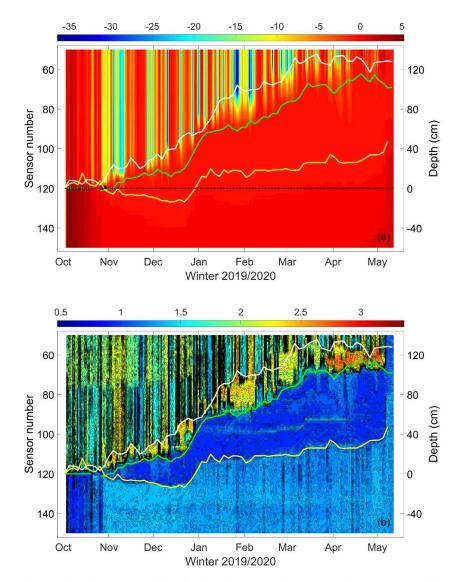




Figure 5. Time series of sensor position for the air-snow (white), snow-ice (green) and ice-water (yellow) interfaces, identified applying the SIMBA algorithm. The SIMBA-ET observation is

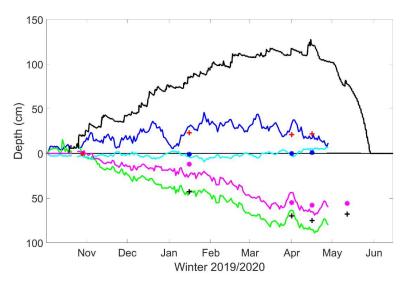




- illustrated as background in a), and SIMBA-HT ratio (HT60/HT90) in b). The black dash line shows the sensor number (120) at the initial ice surface (Z_{gi0}). For clarity, we only illustrate sensors 50 –
- 244 150.

245

- 246 Using snow/ice interface as the zero-reference level, time series can be calculated for the snow depth,
- snow-ice thickness, total ice thicknesses, and ice freeboard (Fig. 6). Figure 6 is an example of the
- 248 2019/2020 time series, indicating that the lake ice was mainly granular ice, which was related to
- 249 heavy snow fall during the ice season (the snow depth observed at FMI-ARC weather station on land
- 250 was highest in a decade).



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Figure 6. Products derived on the basis of SIMBA data: snow depth (blue), ice freeboard (cyan), granular ice thickness (magenta), and total ice thickness (green). The symbols represent manual observations of snow depth (+), ice freeboard (•), granular ice thickness (•) and total ice thickness (+). The black solid line denotes the snow depth on land.

256

257 3.2 Weather data

The observed daily mean values of meteorological parameters for all seasons are presented in Figure

259 7. The inter-annual mean, maximum and minimum air temperatures are -2.5 °C, -16.5 °C and -5.5 °C,

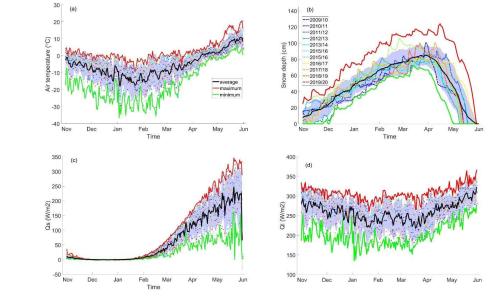
260 respectively. The air temperature reveals a constant decreasing pattern from November to January.

261 The coldest months are January and February. From March onward, the air temperature increased





- gradually due to increasing solar radiation (Fig. 7c). The inter-annual average, maximum and minimum downward longwave radiative fluxes are 259 W/m², 309 W/m², and 201 W/m², respectively.
- 264 The corresponding values for downward shortwave radiative fluxes are 64 W/m^2 , 97 W/m^2 , 26 W/m^2 .



265

Figure 7. The observed (dots) daily mean air temperature (a), snow depth (b), downward shortwave (c), and longwave (d) radiative fluxes for each ice season between 1 November and 31 May. The solid lines represent decadal daily maximum (red), minimum(green) and average (black) values. The shadow area represents the standard deviation (STD). For snow depth, daily mean values are given as thin color lines.

272

Table 2 Summary of various meteorological and physical observations between 1 November and 31
May. For meteorological parameters (*Va*, *Ta*, *RH*, *cn*, *Qs*, *Ql*) the values are seasonal mean ± standard
deviation.

Season	Va	Та	RH	сп	Qs	Ql	Tprec	Hsmax	FDD	TDD
	m/s	°C	(%)	(-)	W/m^2	W/m^2	(mm)	(cm)	°C	°C
2009/2010	2.2±0.3	-6.8±9.4	84±9	0.7±0.1	62.9±76.8	267±31	201	101	-1717	304
2010/2011	2.2±0.6	-8±9.5	83±9	0.6±0.1	64.1±76.5	259±27	157	72	-1955	286
2011/2012	2.4±0.4	-5.1±7.2	84±11	0.7±0.1	64.4±82.3	264±21	272	91	-1308	239
2012/2013	2.2±0.2	-6.3±8.5	80±13	0.6±0.2	67±85.6	250±31	192	82	-1683	346





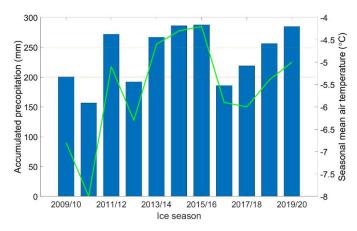
2013/2014	2.6±0.4	-4.6±6.4	81±10	0.7±0.1	61.6±81.8	261±19	267	81	-1214	243
2014/2015	2.7±0.6	-4.3±6.6	84±8	0.7±0.1	55.8±67.2	264±21	286	87	-1148	249
2015/2016	2.3±0.3	-4.2±8.4	84±10	0.7±0.1	61.5±81	265±25	287	106	-1261	354
2016/2017	2.8±0.3	-5.9±4.7	81±10	0.7±0.1	64.9±81.2	252±14	186	82	-1338	101
2017/2018	2.5±0.4	-6±8.7	80±12	0.7±0.2	66.5±82.7	256±27	219	101	-1615	362
2018/2019	2.7±0.4	-5.4±7.9	84±10	0.6±0.1	63.5±79.8	258±26	256	100	-1432	293
2019/2020	2.8±0.5	-5±5.1	84±13	0.6±0.1	70.1±92.9	258±13	285	124	-1242	188

276 Tprec: total accumulated precipitation in water equivalent (mm); Hsmax: the maximum observed snow depth on land.

FDD: The accumulated freezing degree day: the sum of daily mean air temperature below freezing point; TDD: The
accumulated thawing degree day: the sum of daily mean air temperature above freezing point.

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Figure 7b clearly indicates that snow depth for the 2019/2020 season represented an extreme condition in a decade. There is an increasing trend of total precipitation during the ice season (Fig. 8). The total seasonal accumulated total precipitation is highly correlated (correlation coefficient r =0.93) with the seasonal mean air temperature. The correlations between seasonal mean/maximum snow depth and corresponding air temperature are much lower r = 0.40 and r = 0.38, respectively. The correlation between total accumulated precipitation and maximum snow depth was 0.55. The difference is contributed by the snow drift and changes of snow metamorphism.



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Figure 8. The accumulated total precipitation and mean air temperature between 1 November and 31May.



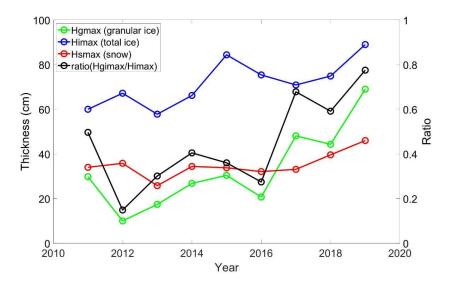


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292 **4. Discussions**

293 4.1 Inter-annual variation of SIMBA snow and ice products

Applying the SIMBA algorithm (Cheng et al., 2020), we obtained lake snow and ice products for all 294 295 seasons (see data availability). Figure 9 shows the observed seasonal maximum values for the snow 296 depth, maximum total ice thickness, and maximum granular ice thickness. During the observation period, both snow depth and ice thickness showed increasing trends. The increase of granular ice 297 298 thickness is the fastest among all the snow and ice components. It reached the maximum 80% of the total ice thickness in 2019/2020. In Lake Orajärvi, snow mass has contributed to the ice thickness 299 during every winter season. The maximum granular ice thickness was on average about 40% of the 300 301 maximum total ice thickness during the data period. For all seasons, the correlation coefficient 302 between the maximum granular ice thickness and the maximum ice thickness was 0.64. The 303 occurrence of maximum lake snow is, on the average, about one month prior to the maximum granular 304 ice formation (Fig. 10). Because of snow to ice transformation, the time series of snow depth in the 305 lake is not correlated with the snow depth on land. The snow depth on lake ice ranged from 25 to 43% of that on land. On the average the ratio was 0.33, some 11% less than observed for a lake in southern 306 307 Finland (Kärkäs, 2000). In several seasons, when SIMBA were recovered in late April or early May, 308 the entire snow layer on lake ice was transferred to granular ice. Granular ice reached its maximum 309 value when the ice surface was free of snow.







- 311 Figure 9. SIMBA observed seasonal maximum snow depth (red), maximum total ice thickness (blue),
- 312 maximum granular ice thickness (green) and the ratio between granular ice and total ice thickness
- 313 (black) during observation seasons.

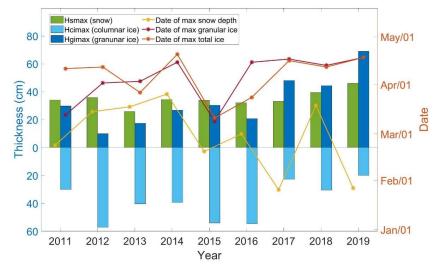


Figure 10. Seasonal maximum snow depth, granular ice thicknesses, congelation ice thicknesses,
and the date when those values were observed.

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318 **4.2 Inter-annual variation of temperature conditions**

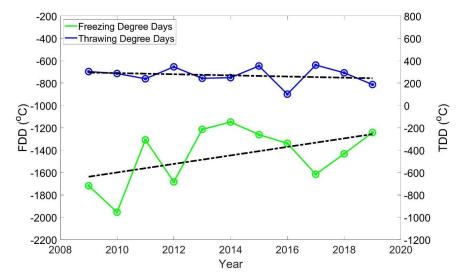
According to weather observations in Sodankylä, the air temperature increased by about 0.16 °C/year during the last deacde. For the period from 1980 – 2020, the air temperature has an increasing trend of about 0.06 °C/year. On the average, the increase of air temperature in last decade is about 3 times faster than the past 40 years in agreement with the findings of Przybylak and Wyszyński (2020) for the high Arctic. The accumulated precipitation correlated better to the maximum snow depth on land (r = 0.55) than the mean snow depth (r = 0.45). It is, however, not correlated (r = 0.21) with snow depth on the lake ice.

The seasonal integrated FDD and TDD for each winter season are shown in Figure 11. A negative decreasing of FDD was seen in response to the increase of air temperature. FDD is directly linked with thermodynamic ice formation. During a given period, a decrease of FDD is expected to result in less formation of columnar ice. However, during our observation period, the total ice thickness revealed an increasing trend. The increase of ice thickness is due to snow-ice formation. The trend of





331 TDD is very insignificant, suggesting that the melting of lake ice due to temperature increase has not



332 increased much during the observation decade.

Figure 11. The seasonal accumulated freezing degree day (FDD) and thaw degree day (TDD)
 during the observation period (2009/2010 – 2019/2020).

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333

337 4.3 Challenges of the SIMBA program

338 SIMBA observations in Lake Orajärvi represent a small but sustainable program, so far ran for a 339 decade. A few times we have encountered malfunction of SIMBA, especially in the early phases of the SIMBA program. In recent years, SIMBA has become more robust without need for heavy-duty 340 maintenances during field measurements, and the system has been remarkably improved with respect 341 342 to the quality of HT measurements. Several snow and ice products can be derived from SIMBA's two 343 type of temperature (SIMBA-ET and SIMBA-HT) measurements. The SIMBA program has largely 344 benefited from the Sodankylä supersite infrastructure, where the comprehensive and high standard 345 meteorological observations are available.

Challenges remain in further improvement of the SIMBA program. Due to safety issues, SIMBA must be deployed and recovered when ice is strong enough. Hence, the early freeze-up and late break-up cannot be monitored. In Autumn 2019, a wooden floating raft was deployed and anchored in Lake Orajärvi. SIMBA was, for the first time, deployed during ice-free season on 1 October. This kind of deployment will be carried out also in the future, allowing year-round SIMBA measurements.





Part of the thermistor chain exposed in the air above the snow surface may suffer from frost in winter or from solar heating in spring, and also the sensors in the upper layers of snow and ice may suffer from solar heating, resulting in large uncertainties in SIMBA-ET and SIMBA-HT readings. To compensate the effect of temperature errors on snow depth detection, one solution is to deploy Acoustic Rangefinder Sounders (ARS) to measure the evolution of snow surface. In fact, an ARS has been deployed in the past two winter seasons. These data sets can also be used to understand the effect of wind on snow drift and quantify snow surface sublimation in winter.

- 358 During the melting season, both SIMBA-ET and SIMBA-HT strongly raise in the upper part of the
- ice resulting an isothermal status of the entire ice column. In this condition, SIMBA snow depth and
- ice thickness values are liable to large errors. Combination of SIMBA observations and numerical
- 361 model experiments may yield more reliable results in such conditions.
- 362 SIMBA measurements have been taken automatically, but it is still important to carry out manual on 363 site observations, such as collecting ice core and snow samples, as such observations cannot be made 364 by automatic instruments.
- 365

366 4 Data availability

- The data are archived at <u>https://zenodo.org/record/4559368#.YEIYOtyxVPZ</u> (Cheng et al., 2021). The 4 zip-files should be unzipped in different file folders preferably using zip-file names as the folder names. A readme file exists in each folder. The manual in situ snow depth and ice thickness observation for 2009/2010 - 2012/2013 as well as a description of SIMBA deployment and recovery for each ice season (SIMBA_D&R_all_Years.docx) are provided.
- 372

373 5 Conclusions

A thermistor string-based snow and ice mass balance apparatus (SIMBA) has been deployed in an Arctic lake since 2009. The measurements covered most part of the ice season from mid-December to late April/early May. SIMBA-ET and SIMBA-HT temperature observations are described in this paper. The daily snow depth and ice thickness were derived from SIMBA temperature field applying a validated automatic algorithm (Cheng et al., 2020). The meteorological parameters for winter seasons (1 November - 31 May) are also collected and discussed. During the investigation decade, the air temperature in the ice season has had an increasing trend of 0.16 °C/year. The warming rate is





381 comparable to the result find for the high Arctic by Przybylak and Wyszyński (2020). The increase of 382 air temperature in winter season is highly correlated (0.93) with seasonal total accumulated precipitation. The precipitation in season 2019/2020 represented an extreme episode during the study 383 384 decade. Despite of the air temperature increase, the total maximum ice thickness in the lake has an 385 increasing trend. The increase of maximum ice thickness is due to the increase of granular ice. The 386 interannual variability of maximum granular ice thickness is large ranging from 15 to 80% of the total maximum ice thickness. The time series of the SIMBA ET and HT allow identification of moving 387 air-snow, snow-ice and ice-water interfaces. Because of the air temperature increase, the seasonal 388 389 accumulated FDD is reducing, suggesting reduced formation of columnar ice and, hence, a smaller 390 role of air temperature in controlling the ice thickness. On the contrary, the role of precipitation on 391 total ice formation is enhanced. The trend in TDD was negligible, suggesting that the effect of air temperature on ice melting has remained unchanged. 392

To our knowledge, this is the first decadal-scale SIMBA data set ever collected from an Arctic lake.
The data provides information on snow and ice mass balance and the controlling atmospheric factors.
The measurements will continue in the future.

The weather observations, e.g. decadal time series of daily maximum and minimum weather parameters, can be used to estimate snow and ice conditions in the lake applying a snow/ice model (e.g. Cheng et al., 2014). The SIMBA data are not only suitable for snow/ice surface heat and mass balance studies. The temperatures at the ice bottom and in the water below are valuable to understand the lake thermal structure and water-ice heat transfer (Huang et al., 2019b).

The SIMBA program, with Lake Orajärvi as a testbed, offers excellent opportunities for dissemination of cryospheric knowledge and related outreach, providing rich possibilities for community collaborations both nationally and internationally. The observed changes in snow depth and composition of lake ice contribute to better understanding of cryospheric aspects of climate change. For example, parameterizations of the discovered snow and ice processes can be improved in climate models.

Snow and ice measurements similar to those in Lake Orajärvi have been recently initiated in
Wulaingsuhai lake in an arid climate zone in Inner-Mongolia of China. The observations focused on
lake ice mass balance (Lu et al., 2020) and energy budget, in particular the solar radiation (Cao et al.,
2020). In a long run, the corresponding lake snow and ice measurements at both sites and possible





- similar observations in a thermokarst lake (e.g. Huang et al, 2019a, 2019b) at Qinghai-Tibet Plateau,
- 412 often referred to as the "Third Pole of the Earth" can be used together to carry out coordinated research.
- 413

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