

1 **Inter-annual variation of lake ice composition in European Arctic: observations based on high-**  
2 **resolution thermistor strings**

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14

15 Abstract

16 Climate change and global warming strongly impact the cryosphere. The rise of air temperature and  
17 change of precipitation patterns lead to dramatic responses of snow and ice heat and mass balance.  
18 Sustainable field observations on lake air-snow-ice-water temperature regime have been carried out  
19 in Lake Orajärvi in the vicinity of the Finnish Space Centre, a Flagship Supersite in Sodankylä in  
20 Finnish Lapland since 2009. A thermistor string-based snow and ice mass balance buoy called “Snow  
21 and ice mass balance apparatus (SIMBA)” was deployed in the lake at the beginning of each ice  
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  
26 had an increasing trend of 0.16° C/year, and the interannual variations were highly correlated ( $r =$   
27 0.93) with the total seasonal accumulated precipitation; 2) The maximum granular ice thickness

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

33

## 34 **1. Introduction**

35

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

43 Observations on snow depth and lake ice thickness are needed for (a) monitoring of climate variability  
44 and trends (Filazzola et al., 2020), (b) practical applications, such as use of lake ice for winter fishing,  
45 transport, and recreational activities (Leppäranta, 2015), and (c) to provide initial conditions for  
46 operational forecasting (Anderson et al., 2018). Snow depth and lake ice thickness can be measured  
47 manually. For example, in Finland, lake ice thickness is measured via manual drilling in a single  
48 location in 45 lakes with ten-day intervals throughout the ice season. However, this requires a lot of  
49 manpower, and accordingly does not allow collection of time series with a better spatial and temporal  
50 resolution. During recent decades, the number manual observations has strongly declined in many  
51 countries (Duguay et al., 2006). Satellite remote sensing yields information on lake ice cover (Wu et  
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 ( $\sim 10$  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  
56 promise in retrieving ice depth over rivers (Mermoz et al., 2013); as fully polarimetric data is not to

57 date widely available from existing SAR sensors extensive testing and application of the method for  
58 lakes is currently lacking.

59 The SIMBA data set is potentially highly relevant for the development of land applications for  
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,  
62 NASA SMAP and Chinese sensors. Due to the inherent coarse resolution of these sensors (tens of  
63 kilometers), a key issue is to acquire combined simultaneous data representing various processes in  
64 lakes, in addition to surrounding land areas. As such, the SIMBA forms an integral part of the FMI  
65 sensor network in Sodankylä.

66 Thermistor sting-based snow and ice mass balance apparatus (SIMBA) have been applied for more  
67 than a decade to measure snow depth, ice thickness and temperature profile from air through snow  
68 and ice to water (Jackson et al., 2013). Most of SIMBA have so far been deployed in Polar sea ice  
69 (Lei et al., 2018), but also lake ice has been studied (Cheng et al., 2014; Wei et al., 2016). In this  
70 paper we describe SIMBA observations from an ongoing program that started in Lake Orajärvi in  
71 northern Finland in 2009. Supporting meteorological observations from Finnish Meteorological  
72 Institute Arctic Research Centre (FMI-ARC) are also presented. The objectives of the SIMBA  
73 program were

- 74 - to evaluate the cost-effectiveness of SIMBA buoys in a remote lake environment
- 75 - to monitor climate variability and change as reflected in snow depth as well as lake ice  
76 thickness and composition
- 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  
80 role of granular ice in lake ice phenology
- 81 - to develop better parameterizations of snow-to-ice transformation in numerical snow/ice  
82 models.

83

## 84 **2 Observation**

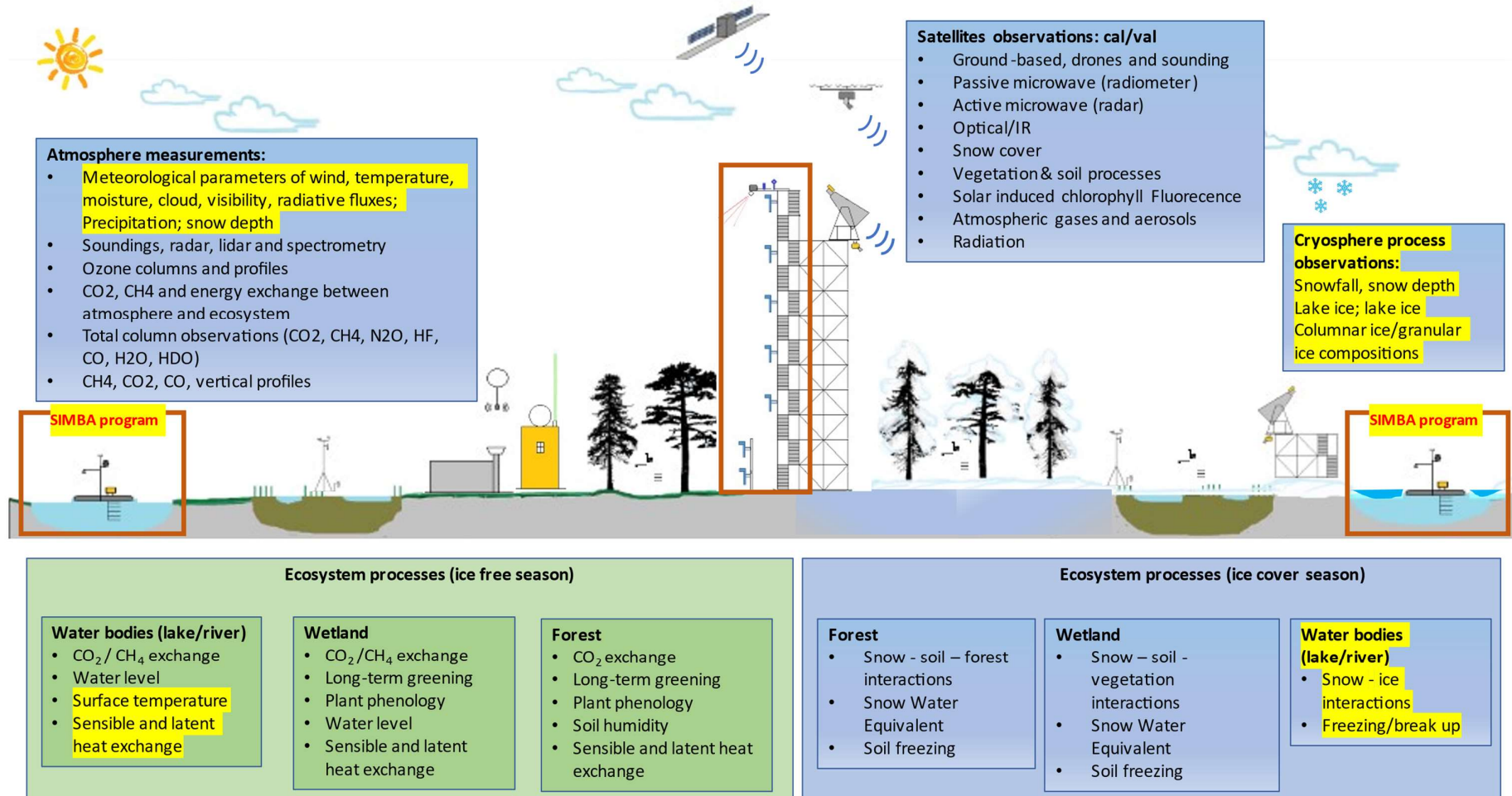
### 85 **2.1 Sodankylä supersite**

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

87 Meteorological Institute's Arctic Space Centre (FMI-ARC) in Sodankylä (67.367 °N, 26.629 °E),  
88 Finland, is a super-observation site where various Earth observations (upper-air chemistry and  
89 physics, atmospheric column measurements, snow and soil hydrology, biosphere-atmosphere  
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  
92 instrumentation placed in the forests, wetlands and freshwater bodies, which are the main landcover  
93 types in the area. In this paper we focus on the cryospheric *in situ* observations of snow cover and  
94 lake ice as well as meteorological parameters.

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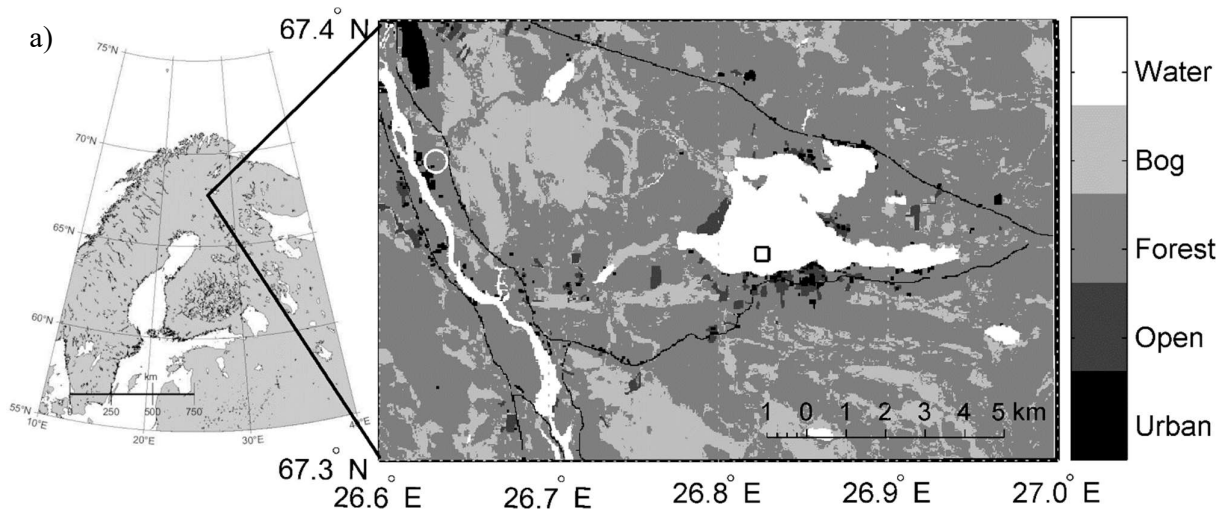
98 Figure 1. Schematic diagram of the FMI-ARC supersite observational systems at Sodankylä. The original diagram is at <https://litdb.fmi.fi/>. The frames

99 in red and text with yellow background describes the measurements addressed in this paper.

100

101 The sub-Arctic climate and the geographic location between continental and marine climate zones  
102 result in a high inter-annual, seasonal and synoptic-scale variation in local weather conditions,  
103 enabling development of very different kinds of snowpack structures on land (Tikkanen, 2005) and  
104 snow/ice composition on lakes (Cheng et al., 2014). Lake Orajärvi is a boreal medium-sized lake  
105 located in Sodankylä municipality in the in the eastern Lapland. The lake has a surface area of about  
106 11 km<sup>2</sup> with an average depth of 4.4 m and a maximum depth of 11 m close to the southern shore of  
107 the lake (Fig. 2a). The estimated water volume in the lake is 0.0485 km<sup>3</sup>, and the shore length is 28  
108 km. The lake surface elevation is 182 m above sea level. The ice season typically starts in November  
109 and lasts until May. The first snowfall typically occurs in late October, but the snow may melt during  
110 warmer autumn days. The seasonally permanent winter snow accumulation usually starts between  
111 mid-November and early-December. Snow is present on the lake ice surface every winter season.

112



113

b)



114

115

116 Figure 2. a) The location of Lak Orajärvi in Finnish Lapland and a map of lake Orajärvi and local  
117 catchment, where the open black square marks the SIMBA site and white circle is the Finnish Space  
118 Centre. b) Snapshots of SIMBA deployment in Lake Orajärvi and a weather station at FMI-ARC  
119 main camp. A raft was anchored in the lake in October 2019 aiming to extend the lake observations  
120 beyond the ice season.

121

## 122 2.2 SIMBA

### 123 2.2.1 SIMBA program

124 SIMBA buoys have been deployed in Lake Orajärvi since 2009. The 2009 deployment was probably  
125 the internationally first SIMBA application for a lake study. In each winter when ice was formed in  
126 Lake Orajärvi, one SIMBA was deployed around mid-December at the same site,  $67.35^{\circ}$  N,  $26.83^{\circ}$  E,  
127 some 500 m from the shoreline. At the time of deployment, the snow depth, lake ice thickness and  
128 ice freeboard were measured. A supporting frame made of fiberglass was constructed on lake ice, and  
129 the SIMBA main control Peli case was placed on top of it (Fig. 2b). A separate wooden pole with  
130 scale was standing vertically to hold the thermistor string. An ice borehole was drilled through the ice  
131 layer, and the thermistor string was placed in it. The scene was left as it is, and then the thermistor

132 string was frozen with surrounding water in the borehole. The SIMBA operated in the lake over the  
 133 winter and most of the spring melting season. The recovery of SIMBA took place usually in late April  
 134 but in some years as late as mid-May. Snow and ice conditions around the deployment site were  
 135 documented and measured before dismantling the SIMBA camp. The documentation on SIMBA  
 136 deployment and recovery is provided along with the SIMBA data as online files (see data availability).  
 137 Table 1 summarizes the SIMBA deployment and recovery status.

138

139 Table 1. SIMBA deployment and recovery days and simultaneous *in situ* observed snow depth ( $h_s$ ),  
 140 total ice thickness ( $H_i$ ), and ice freeboard ( $H_{fb}$ , defined as negative if the lake water level was above  
 141 the snow/ice interface) The seasonal mean values were derived from SIMBA-ET and SIMBA-HT  
 142 observations.  $H_{gi}$  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	$h_s$	$H_{ci}$	$H_i$	$H_{fb}$	$H_i$	$h_s$	$H_{sfb}$	$H_{gi}$	$H_{ci}$
	DD/MM/YY	(cm)			DD/MM/YY	(cm)								
2009/2010	16/12/09	5	27	0	07/04/10	31	54	64	5	NA				
2010/2011	SIMBA was not deployed; only <i>in situ</i> observations of $h_s, H_i$ and $H_{fb}$ every second week were available.													
2011/2012	19/12/2011	16	14	-4	12/04/2012	24	22	55	-3	38 $\pm$ 16	22 $\pm$ 6	5 $\pm$ 2	15 $\pm$ 10	23 $\pm$ 7
2012/2013	12/12/2012	18	33	+1	25/04/2013	0	39	59	6	57 $\pm$ 7	26 $\pm$ 7	-5 $\pm$ 3	4 $\pm$ 4	53 $\pm$ 4
2013/2014	12/12/2013	14	27	+1	30/04/2014	20	35	35	-3	49 $\pm$ 7	17 $\pm$ 4	-3 $\pm$ 2	10 $\pm$ 2	40 $\pm$ 6
2014/2015	14/12/2014	19	30	-2.5	23/04/2015	2	35	69	4	54 $\pm$ 11	24 $\pm$ 7	-4 $\pm$ 3	16 $\pm$ 8	38 $\pm$ 4
2015/2016	18/12/2015	18	27	-1	22/04/2016	5	30	71	6	60 $\pm$ 16	19 $\pm$ 7	-2 $\pm$ 3	12 $\pm$ 9	48 $\pm$ 9
2016/2017	16/12/2016	8	31	-1	24/04/2017	10	38	72	4	58 $\pm$ 13	19 $\pm$ 6	-1 $\pm$ 2	6 $\pm$ 8	50 $\pm$ 8
2017/2018	15/12/2017	25	23	-9	03/05/2018	0	28	55	6	48 $\pm$ 15	24 $\pm$ 6	-4 $\pm$ 3	27 $\pm$ 14	21 $\pm$ 3
2018/2019	13/12/2018	15	19	-2	02/05/2019	1	20	55	6	51 $\pm$ 17	21 $\pm$ 7	-1 $\pm$ 3	21 $\pm$ 14	30 $\pm$ 7
2019/2020	03/10/2019	-			12/05/2020	4	13	68	7	49 $\pm$ 24	24 $\pm$ 9	-1 $\pm$ 3	32 $\pm$ 20	20 $\pm$ 5

143 The seasonal mean values of  $H_i$ ,  $h_s$ ,  $H_{gi}$ , and  $H_{ci}$  were calculated by the SIMBA algorithm (Cheng et al., 2020). The  
 144 seasonal mean value of ice freeboard ( $H_{sfb}$ ) was calculated based on time series of snow depth ( $h_s$ ), granular ice thickness  
 145 ( $H_{gi}$ ) and columnar ice thickness ( $H_i$ ) according to the Archimedes' principle:  $H_{sfb} = H_i + H_{gi} - (h_s \rho_s + H_{gi} \rho_{gi} +$   
 146  $H_i \rho_i) / \rho_w$ , where  $\rho_s$ ,  $\rho_{gi}$ ,  $\rho_i$  and  $\rho_w$  are seasonal mean densities of snow, granular-ice and columnar ice and lake water,  
 147 assumed to be 320 kg/m<sup>3</sup>, 890 kg/m<sup>3</sup> and 910 kg/m<sup>3</sup> and 1000 kg/m<sup>3</sup>, respectively. The STD is the standard deviation.



148

### 149 2.2.2 SIMBA buoy

150 SIMBA is a thermistor string-based Snow and Ice Mass Balance Apparatus. It has been developed  
151 by the Scottish Association for Marine Science (SAMS) Research Services Ltd (SRSL) in UK.  
152 SIMBA consists of a simple, robust thermistor string with 240 temperature sensors distributed evenly  
153 (2 cm intervals) along a 4.8 m long heat-shrink PVC plastic sleeve coated flat white wire. White heat-  
154 shrink sleeve is used to minimize the possibility of solar heating of the sensors. The accuracy of the  
155 SIMBA thermistor sensor is  $\pm 0.1$  °C, which is comparable with other type of thermistor string based  
156 IMBs (Richter-Menge et al. 2006). Each sensor measures the environment temperature (SIMBA-ET).  
157 The resolution of the thermistor sensor is 0.0625°C, i.e., smaller changes cannot be detected even if  
158 the absolute accuracy of the sensor would allow it. In addition, the thermistor chain is equipped with  
159 heaters, i.e. resistor components mounted next to the temperature-sensing elements. A weak voltage  
160 (8 V) supply is connected to provide gentle identical heating of each sensor on the chain. The SIMBA  
161 heating cycle is usually long enough, often 60 or 90 s, for the temperature rise at the sensor to reach  
162 a steady state. Thermal conductivity determines how the heat is conducted away of the heated sensors  
163 placed in air, snow, ice and lake water. As a result, the SIMBA-HT profiles can greatly enhance the  
164 detection of the interfaces between air, snow, ice and water. The heating cycle is applied once per  
165 day. The SIMBA-HT is controlled not to disturb the SIMBA-ET measurements which are carried out  
166 typically 4 times per day (Jackson et al., 2013). A SIMBA also includes a built-in GPS to record  
167 SIMBA drift positions (for sea ice applications), a magnetometer for tilt and floe rotation, a barometer  
168 for surface air pressure, and an external sensor to measure near-surface ambient air temperature. An  
169 iridium modem is applied for data transmission. SIMBA has been used in various field campaigns  
170 targeting snow and ice mass balance in seasonally ice-covers in lakes (Cheng et al., 2014) and Polar  
171 Oceans (Hoppmann et al. 2015; Provost et al. 2017; Lei et al. 2018, 2021). Table 1 presents a summary  
172 of SIMBA observations in Lake Orajärvi.

173

### 174 2.3 Weather station

175 Meteorological data were collected at FMI-ARC station (67.3666°N, 26.6290°E, WMO code 02836)  
176 11 km from Lake Orajärvi. The data sets include wind speed ( $Va$ ), air temperature ( $Ta$ ), relative  
177 humidity ( $RH$ ), cloudiness ( $cn$ ), longwave ( $Ql$ ) and shortwave ( $Qs$ ) radiation, snow depth on land ( $Hs$ )

178 and precipitation (*Prec*) (Table 2). The radiative fluxes were measured on a 10-m high tower above  
179 treetops using Kipp&Zonen CM11 pyranometers (305-2800 nm) and Kipp&Zonen CG4  
180 pyrgeometers (4500 - 42000 nm). Snow depth (Campbell Scientific SR50) and precipitation (OTT  
181 Pluvio2) at ground level were also measured. All measurements were taken once a minute and  
182 aggregated to 1-hour time intervals.

183

### 184 **3 Data description**

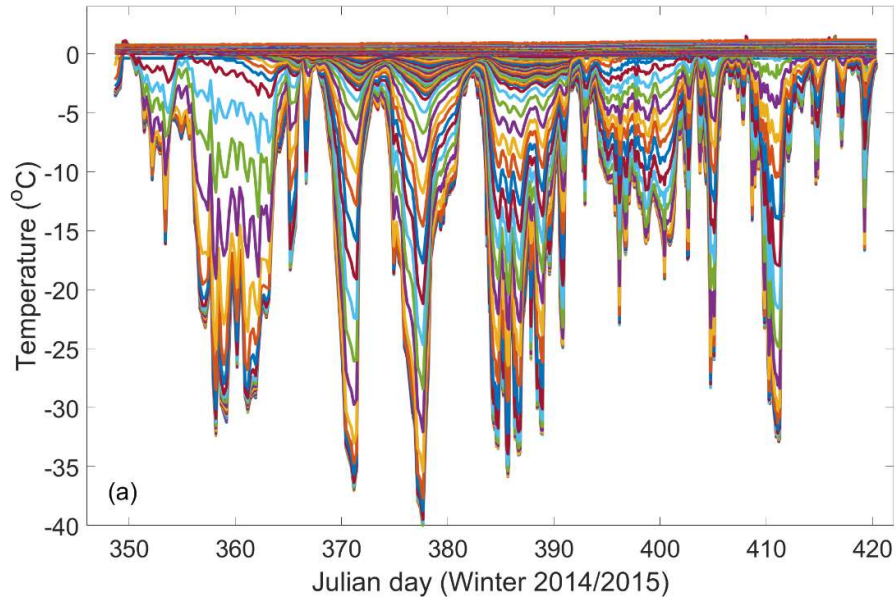
#### 185 **3.1 SIMBA data**

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

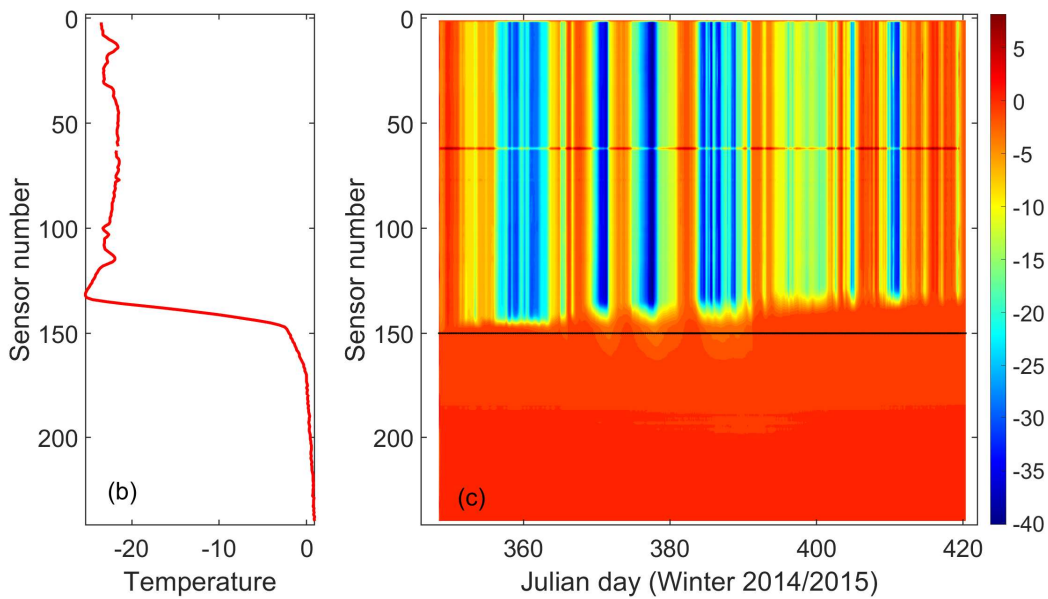
189

##### 190 *3.1.1 SIMBA-ET*

191 For each season, we have up to 241 time series of temperature (SIMBA-ET) at different depths. For  
192 those sensors located in the air, the temperature differences between the sensors are small, as the air  
193 in the lowermost 1.5 m layer mixes effectively and the sensors are close to each other. The  
194 temperatures inside snow reveal much larger vertical gradients because snow has a small thermal  
195 conductivity. The temperature profile in ice has smaller vertical gradient compared to that in snow,  
196 since the thermal conductivity of ice is larger than that of snow. At the ice bottom, temperature is at  
197 the freezing point and gradually increases towards the lake bottom. Figure 3 shows an example of  
198 seasonal SIMBA-ET. One can estimate the heat fluxes within snow and ice, and those at the air-snow,  
199 snow-ice, and ice-water interfaces.



200



201

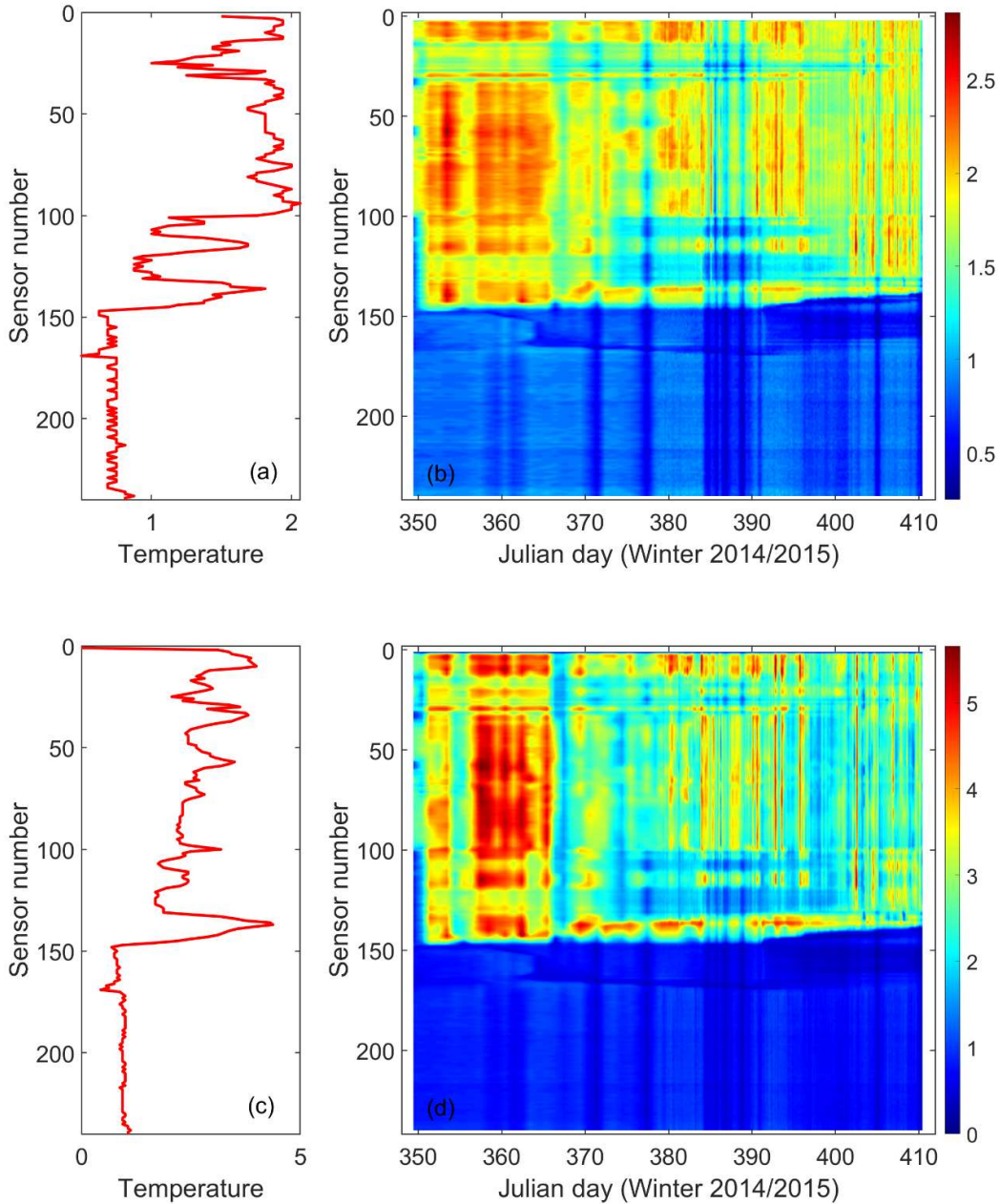
202 Figure 3. Illustrations of SIMBA-ET data: a) Time series of SIMBA-ET during observation period;  
 203 b) One snapshot (19 Jan 2014 08:00 UTC) of vertical SIMBA-ET profile through air-snow-lake ice-  
 204 water; c) SIMBA-ET field observed by 240 sensors. Sensor 1 was placed in air and sensor 240 in  
 205 water.

206

207 *3.1.2 SIMBA-HT*

208 SIMBA-HT shows the temperature increase in the medium each sensor was contacted during a short

209 heating period of 60 s and 90 s. The temperature changes are largely dependent on the thermal  
 210 diffusivity of the surrounding medium. Low heating power ensures that the temperature increasing  
 211 will not be too high to melt snow and ice in contact with the sensor and guarantee a fast restore of  
 212 environment temperature around the sensor before the next SIMBA-ET observation, and above all to  
 213 minimize SIMBA power consumption. One example of SIMBA-HT is given in Figure 4.



214

215

216 Figure 4. Illustrations of SIMBA-HT: a) a snapshot (25 January, 2015, 18:00 UTC) of vertical profile  
 217 of observed temperature increase after 60 s., b) SIMBA-HT (60s) field observed by 240 sensors; c)

218 Same as a) but after heating for 90 s., and d) SIMBA-HT (90s) field observed by 240 sensors;

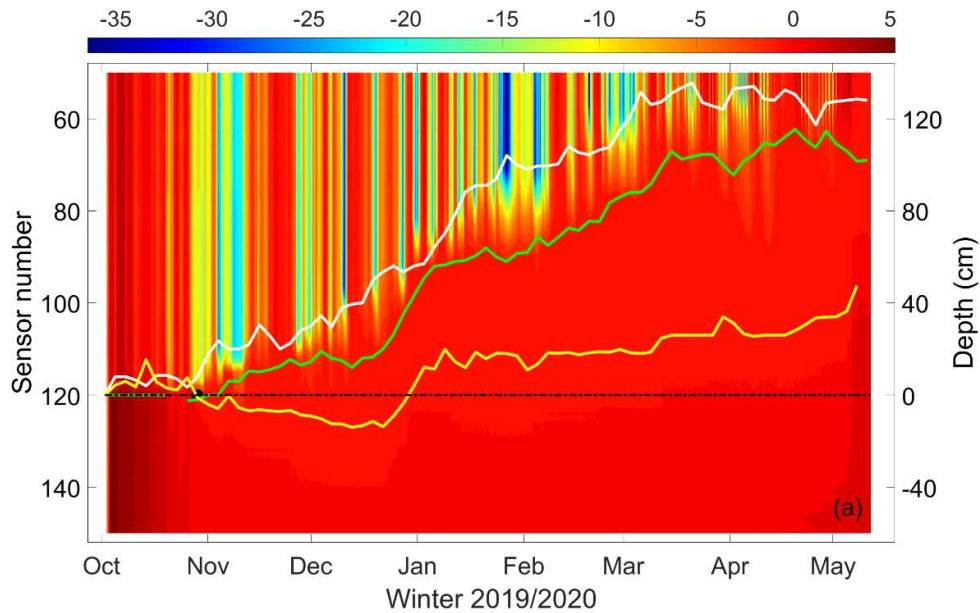
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### 220 *3.1.3 SIMBA snow depth and ice thickness*

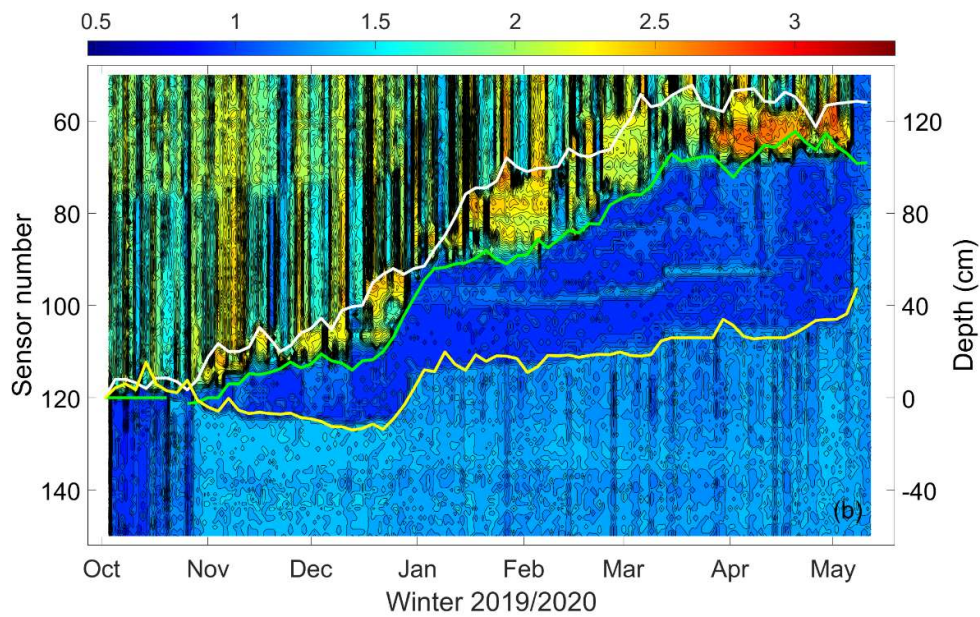
221 Snow depth and ice thickness are derived from SIMBA-ET and SIMBA-HT data. A common  
222 procedure is to look SIMBA-ET temperature profiles manually and identify sudden changes of  
223 vertical temperature gradient to locate the air-snow, snow-ice and ice-water interfaces. The snow  
224 depth is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

229 Below we present an example of the application of the Cheng et al, (2020) algorithm to retrieve snow  
230 depth and ice thickness from SIMBA data observed in Lake Orajärvi. When SIMBA was deployed,  
231 the initial sensor position at snow-ice interface is known and we defined it as  $Z_{gi0}$ , i.e. zero reference  
232 position for granular ice. During observation period, in case if initial snow-ice interface is moving  
233 upward from  $Z_{gi0}$ , which is a common phenomenon in Arctic lakes, the distance between  $Z_{gi0}$  and  
234 moving snow-ice interface is the new granular ice thickness formed by snow to ice transformation.  
235 The depth difference between total ice thickness and granular ice thickness is the congelation ice  
236 formed at the ice bottom. Figure 5 shows the air-snow, snow-ice and ice-water interfaces with  
237 SIMBA-ET (a) and SIMBA-HT (b) as the background. For better clarity, 5-day running average can  
238 be produced as the final products.

239



240



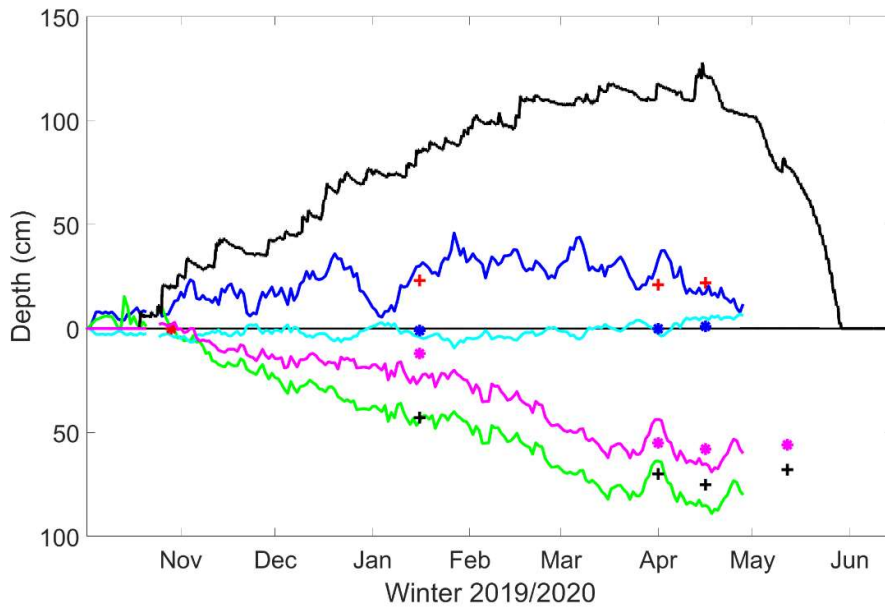
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242 Figure 5. Time series of sensor position for the air-snow (white), snow-ice (green)  
 243 (yellow) interfaces, identified applying the SIMBA algorithm. The SIMBA-ET observation is  
 244 illustrated as background in a), and SIMBA-HT ratio (HT60/HT90) in b). The black dash line shows  
 245 the sensor number (120) at the initial ice surface ( $Z_{gi0}$ ). For clarity, we only illustrate sensors 50 –  
 246 150.

247

248 Using snow/ice interface as the zero-reference level, time series can be calculated for the snow depth,

249 snow-ice thickness, total ice thicknesses, and ice freeboard. Figure 6 is an example of the 2019/2020  
 250 time series, indicating that the lake ice was mainly granular ice, which was related to heavy snow fall  
 251 during the ice season. The snow depth observed at FMI-ARC weather station on land was the highest  
 252 in a decade. A few *in situ* observations (symbols in Fig. 6) were made during ice season. Point  
 253 comparison between SIMBA algorithm detected and *in situ* observed values ranged from 2  
 254 centimeters up to 12 centimeters. The mean biases are 5, -1, 3 and 2 cm for snow depth, freeboard,  
 255 granular and total ice thickness, respectively. Small values were largely due to the compensation  
 256 effect. To validate algorithm, a lot more *in situ* observations are needed. Such analyses can be found  
 257 in Cheng et al., (2020).



258  
 259 Figure 6. Products derived on the basis of SIMBA data: snow depth (blue), ice freeboard (cyan),  
 260 granular ice thickness (magenta), and total ice thickness (green). The symbols represent *in situ*  
 261 observations of snow depth (+), ice freeboard (•), granular ice thickness (•) and total ice thickness  
 262 (+). The black solid line denotes the snow depth on land.

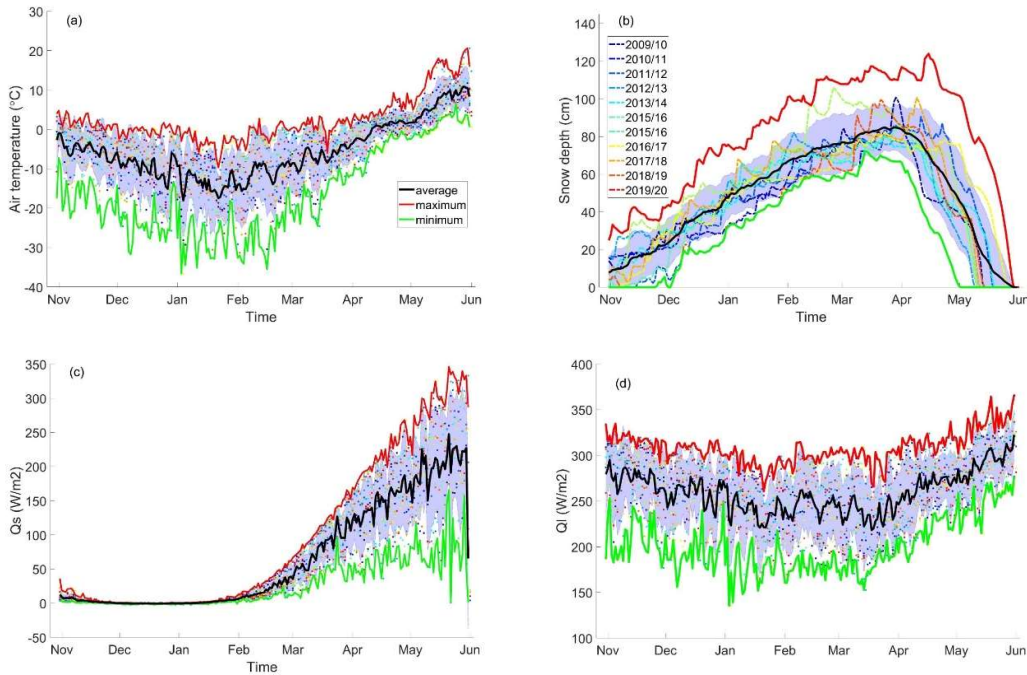
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### 264 3.2 Weather data

265 The observed daily mean values of meteorological parameters for all seasons are presented in Figure  
 266 7. The inter-annual mean, maximum and minimum air temperatures are -2.5 °C, -16.5 °C and -5.5 °C,  
 267 respectively. The air temperature reveals a constant decreasing pattern from November to January.  
 268 The coldest months are January and February. From March onward, the air temperature increased

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

272



273

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

279

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

Season	<i>Va</i> m/s	<i>Ta</i> °C	<i>RH</i> (%)	<i>cn</i> (-)	<i>Qs</i> W/m <sup>2</sup>	<i>Ql</i> W/m <sup>2</sup>	<i>Tprec</i> (mm)	<i>Hsmax</i> (cm)	<i>AFDD</i> °C	<i>ATDD</i> °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



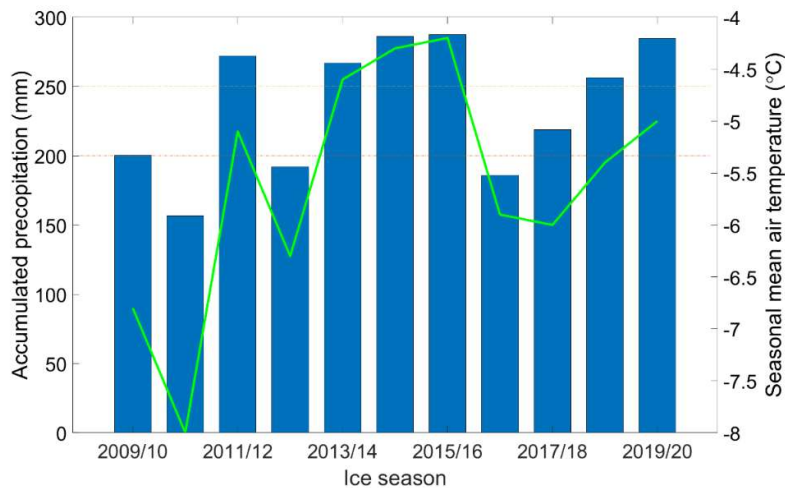
2013/2014	2.6±0.4	-4.6±6.4	81±10	0.7±0.1	61.6±81.8	261±19	267	81	-1214	243
2014/2015	2.7±0.6	-4.3±6.6	84±8	0.7±0.1	55.8±67.2	264±21	286	87	-1148	249
2015/2016	2.3±0.3	-4.2±8.4	84±10	0.7±0.1	61.5±81	265±25	287	106	-1261	354
2016/2017	2.8±0.3	-5.9±4.7	81±10	0.7±0.1	64.9±81.2	252±14	186	82	-1338	101
2017/2018	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

283 *T<sub>prec</sub>*: total accumulated precipitation in water equivalent (mm); *H<sub>smax</sub>*: the maximum observed snow depth on land.

284 AFDD: The accumulated freezing degree day: the sum of daily mean air temperature below freezing point; ATDD: The  
 285 accumulated thawing degree day: the sum of daily mean air temperature above freezing point.

286

287 Figure 7b clearly indicates that snow depth for the 2019/2020 season represented an extreme  
 288 condition in a decade. There is an increasing trend of total precipitation during the ice season (Fig.  
 289 8). The total seasonal accumulated total precipitation is highly correlated (correlation coefficient  $r =$   
 290 0.93) with the seasonal mean air temperature. The correlations between seasonal mean/maximum  
 291 snow depth and corresponding air temperature are much lower  $r = 0.40$  and  $r = 0.38$ , respectively.  
 292 The correlation between total accumulated precipitation and maximum snow depth was 0.55. The  
 293 difference is contributed by the snow drift and changes of snow metamorphism.



294

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

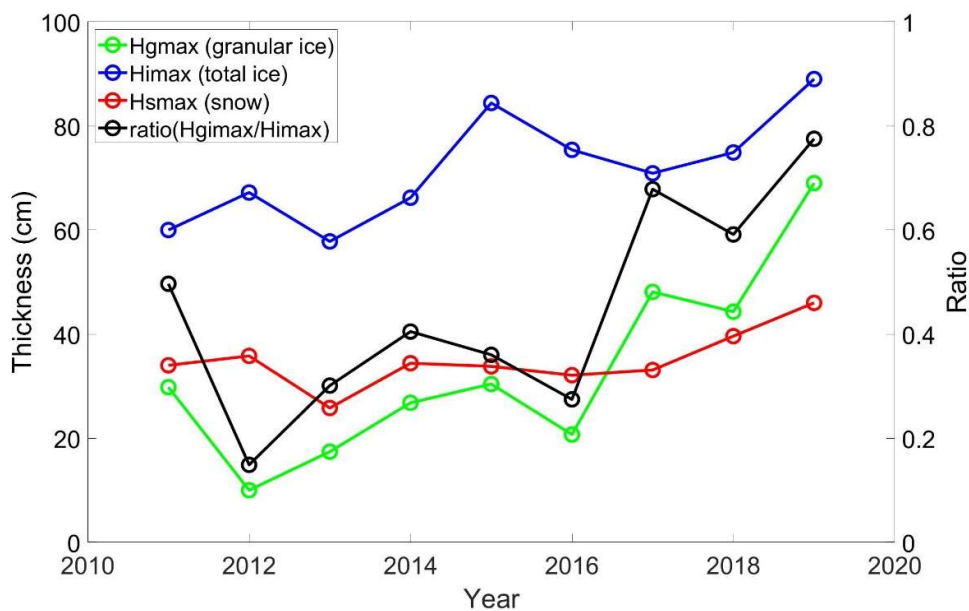
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298

## 299 4. Discussions

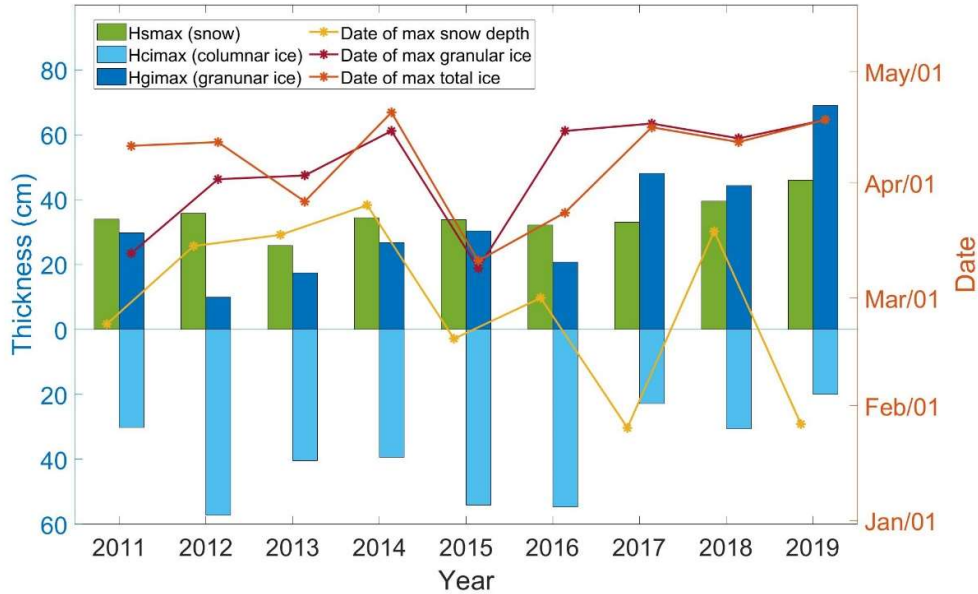
### 300 4.1 Inter-annual variation of SIMBA snow and ice products

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



317

318 Figure 9. SIMBA observed seasonal maximum snow depth (red), maximum total ice thickness (blue),  
 319 maximum granular ice thickness (green) and the ratio between granular ice and total ice thickness  
 320 (black) during observation seasons.



321  
 322 Figure 10. Seasonal maximum snow depth, granular ice thicknesses, congelation ice thicknesses,  
 323 and the date when those values were observed.

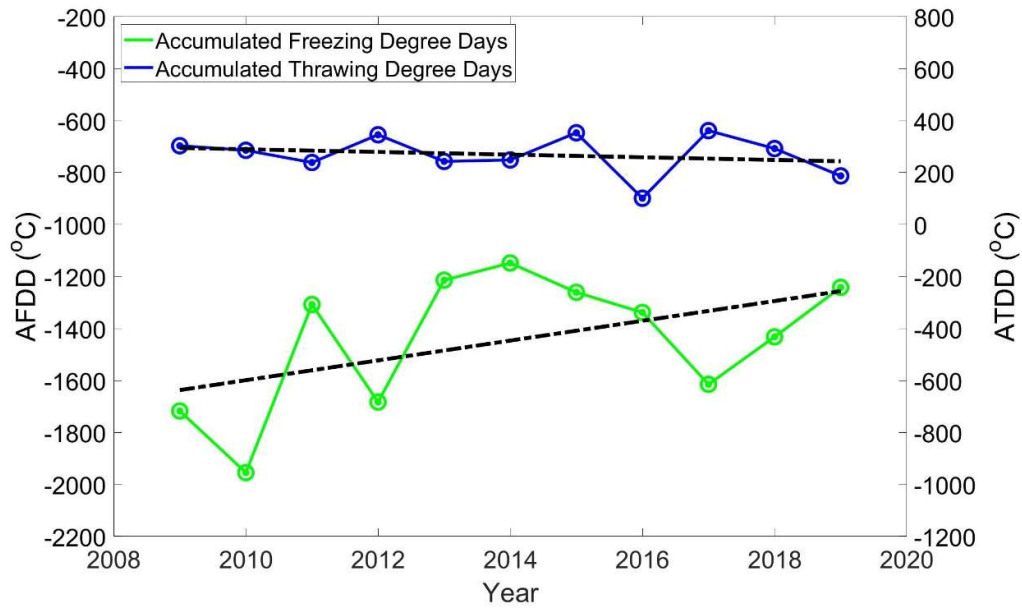
324

#### 325 4.2 Inter-annual variation of temperature conditions

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

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

338 ATDD is very insignificant, suggesting that the melting of lake ice due to temperature increase has  
339 not increased much during the observation decade.



340  
341 Figure 11. The seasonal accumulated freezing degree day (AFDD) and thaw degree day (ATDD)  
342 during the observation period (2009/2010 – 2019/2020).  
343

### 344 4.3 Challenges of the SIMBA program

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

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

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

365 During the melting season, both SIMBA-ET and SIMBA-HT strongly raise in the upper part of the  
366 ice resulting an isothermal status of the entire ice column. In this condition, SIMBA snow depth and  
367 ice thickness values are liable to large errors. Combination of SIMBA observations and numerical  
368 model experiments may yield more reliable results in such conditions.

369 SIMBA measurements have been taken automatically, but it is still important to carry out *in situ*  
370 observations, such as collecting ice core and snow samples, as such observations cannot be made by  
371 automatic instruments.

372

#### 373 **4 Data availability**

374 The data are archived at <https://zenodo.org/record/4559368#.YEIYOtyxVPZ> (Cheng et al., 2021).

375 The 4 zip-files should be unzipped in different file folders preferably using zip-file names as the  
376 folder names. A readme file exists in each folder. The *in situ* snow depth and ice thickness  
377 observations for 2009/2010 - 2012/2013 as well as a description file of SIMBA deployment and  
378 recovery for each ice season (SIMBA\_D&R\_all\_Years.docx) are provided.

379

#### 380 **5 Conclusions**

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

388 comparable to the result find for the high Arctic by Przybylak and Wyszyński (2020). The increase of  
389 air temperature in winter season is highly correlated (0.93) with seasonal total accumulated  
390 precipitation. This is because warm winters in the study region are also wet and characterized by a  
391 high cyclone activity. Transient cyclones are vital for the transport of warm, moist air masses to  
392 Northern Europe (Wickström et al., 2020). The precipitation in season 2019/2020 represented an  
393 extreme episode during the study decade. Despite of the air temperature increase, the total maximum  
394 ice thickness in the lake has an increasing trend. The increase of maximum ice thickness is due to the  
395 increase of granular ice. The interannual variability of maximum granular ice thickness is large  
396 ranging from 15 to 80% of the total maximum ice thickness. The time series of the SIMBA ET and  
397 HT allow identification of moving air-snow, snow-ice and ice-water interfaces. Because of the air  
398 temperature increase, the seasonal AFDD reduces. This results in a decreasing impact of below-zero  
399 air temperatures on lake ice growth during the freezing season, as the growth of columnar ice is  
400 reduced. Simultaneously, the role of precipitation on total ice formation is enhanced because snow-  
401 ice and superimposed ice contribute to an increasing fraction of the total ice thickness. The trend in  
402 ATDD was negligible, suggesting that the effect of air temperature on ice melting has remained  
403 unchanged.

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

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

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

418 Snow and ice measurements similar to those in Lake Orajärvi have been recently initiated in  
419 Wulaingsuhai lake in an arid climate zone in Inner-Mongolia of China. The observations focused on  
420 lake ice mass balance (Lu et al., 2020) and energy budget, in particular the solar radiation (Cao et al.,  
421 2020). In a long run, the corresponding lake snow and ice measurements at both sites and possible  
422 similar observations in a thermokarst lake (e.g., Huang et al, 2019a, 2019b) at Qinghai-Tibet Plateau,  
423 often referred to as the “Third Pole of the Earth” can be used together to carry out coordinated research.  
424

#### 425 **Author contributions**

426 BC and TV initiated the SIMBA program, draft the manuscript; YC, BC, FZ and YQ developed the  
427 SIMBA algorithm processed the SIMBA data and analyses the results; JP, AK and JL are responsible  
428 for supersite and various meteorological *in situ* observations and data collections. All authors  
429 contributed for writing the manuscript.

430

#### 431 **Competing interests**

432 The authors declare no competing interests.

433

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443

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