Automating long-term glacier dynamics monitoring using single-station seismological observations and fuzzy logic classification: a case study from Spitsbergen

W. GAJEK, J. TROJANOWSKI, M. MALINOWSKI

Institute of Geophysics, Polish Academy of Sciences, Warsaw, Poland
Correspondence: Wojciech Gajek <wgajek@igf.edu.pl>

ABSTRACT. Retreating glaciers are a consequence of a warming climate. Thus, numerous monitoring campaigns are being carried out to increase understanding of this on-going process. One phenomenon related to dynamic glacial changes is glacier-induced seismicity; however, weak seismic events are difficult to record due to the sparse seismological network in arctic areas. We have developed an automatic procedure capable of detecting glacier-induced seismic events using records from a single permanent seismological station. To distinguish between glacial and non-glacial signals, we developed a fuzzy logic algorithm based on the signal frequency and energy flow analysis. We studied the long-term changes in glacier-induced seismicity in Hornsund (southern Spitsbergen) and in Kongsfjorden (western Spitsbergen). We found that the number of detected glacial-origin events in the Hornsund dataset over the years 2013-14 has doubled. In the Kongsfjorden dataset, we observed a steady increase in the number of glacier-induced events with each year. We also observed that the seasonal event distribution correlates best with 1 month lagged temperatures, and that extreme rain events can intensify seismic emissions. Our study demonstrates the possibility of using long-term seismological observations from a single permanent station to automatically monitor the dynamic activity of nearby glaciers and retrieve its characteristic features.

KEYWORDS: calving, glacier geophysics, glacier monitoring, seismology

1. INTRODUCTION

Glaciers are dynamic systems that are in constant motion. Their movement, changes in their stress regime, cracks and crevasse expansion, iceberg calving and ice-basal friction all release energy, which can release detectable seismic signals across a broad frequency range (from seismic to seismo-acoustic). Therefore, glacial seismicity has been studied over many years with growing interest.

There are numerous kinds of seismic events, which have been identified in glaciated areas. For this paper the most important are: icequakes, iceberg calving, calving earthquakes, seiches, glacier tremors and ice-vibrations. Pioneering studies on icequakes were conducted in the 1960s and 1970s (Lewandowska and Teisseyre, 1964; Neave and Savage, 1970; Cichowicz, 1983). The authors characterised basic icequake parameters and related them to ice movement and internal ice stress changes. The second type of event, iceberg calving, was studied as another type of glacier-generated signal (Qamar, 1988; Amundson and others, 2000; O’Neel and others, 2010; Köhler and others, 2013; Bartholomaus and others, 2015). Temporal seismic and acoustic calving monitoring using receivers located in close proximity to calving fronts was later introduced by O’Neel and others (2007), Richardson and others (2010) and Glowacki and others (2015). Calving of a massive iceberg in a fjord generates very low-frequency seiche waves, which can be monitored from greater distances using broadband seismometers (Amundson and others, 2012; Walter and others, 2013). All these studies reveal the diversity of calving-related waveform characteristics depending on the type of calving event (Tsai and others, 2008; Köhler and others, 2012, 2015). Calving of extremely large iceberg can cause glacial earthquakes of considerable magnitude ($M_i \geq 5$) in seismological terms (Ekström and others, 2006), however their focal mechanisms are controversially debated. Different kind of signals, narrow-banded seismic tremors lasting from seconds to even several hours (Métaixian and others, 2003; O’Neel and Pfeffer, 2007; Röösli and others, 2014) were proposed to be caused by a resonance in fluid-filled cracks (Lipovsky and Dunham, 2015) and to be related to subglacial water discharge (Bartholomaus and others, 2015a). Another type of monochromatic event is a low-frequency ice-vibration, which is likely caused by large-scale dynamic processes within the glacier, for example the displacement of a considerable part of the glacier due to gravity or uplift forces (Górski and Teisseyre, 1991; Górski, 2004). The spectral characteristics of ice-vibrations vary between different glaciers, mainly due to varying glacier size and have been detected in tide-water, as well as in alpine glaciers. Górski (2004) determined the approximate location of these events on the Hambreen glacier at a distance from its calving front using an array of seismometers. For a more detailed insight into seismological research in the field of glaciology the interested reader is referred to a recent comprehensive review of cryoseismology by Podolskiy and Walter (2016).

The number of seismic events generated by glaciers, change dramatically during the year with a peak during...
late summer (Ekström and others, 2006; O’Neill and others, 2010, Köhler and others, 2015). Seasonal changes can be studied using dedicated local seismic networks targeting the detection and location of glacier-induced seismicity (e.g., Köhler and others, 2012; Bartholomaus and others, 2015b; Koubova, 2015); but, due to their temporary deployment, these networks cannot capture a full record of interannual changes. However, there is a wealth of the seismological data available from permanent stations located in the polar regions that have been recording continuous data streams over many years. Some of these stations are located near glaciers, making the detection of glacier-induced seismic events feasible. Recent studies by Köhler and others (2015, 2016) utilise seismological station records to assess interannual calving intensity at two locations from Spitsbergen. The key aspect related to the use of seismological observations for glaciers’ monitoring is the development of the robust detection and classification scheme that enables to distinguish a glacier-related event from other events like human origin disturbances or tectonic earthquakes.

In this paper, we describe an algorithm for single-station automatic seismic event detection and classification for continuous seismological records. The classification algorithm utilises fuzzy logic, while classification rules are based on theoretical knowledge and verified using a set of confirmed calving icequakes reported by Köhler and others (2015). We apply the developed algorithm to continuous long-term records from two permanent seismic stations located in Spitsbergen in the vicinity of glaciers. Obtained distributions of glacier-induced seismicity reveal several correlations with glaciological and meteorological data.

2. DATA

The analysed data were recorded at Spitsbergen (Fig. 1a), the largest island of the Svalbard Archipelago (Norway), which is a seismically active region located in the Arctic in the vicinity of the Knipovich and Gakkel Ridges, where oceanic spreading centres constantly generate tectonic earthquakes (Engen and others, 2003). The most active seismic zones in Svalbard are Heer Land, Nordaustlandet (Mitchell and others, 1990) and Storfjorden, the latter being recently affected by a sequence of earthquakes, with the largest reaching a local magnitude of $M_L = 6.0$ (Pirli and others, 2013). This relatively high seismic activity makes it challenging to distinguish between earthquakes and glacier-related events. Thus, this is a good testing area for the method’s performance. Spitsbergen is also the study area for many previous papers focussed on glacier-related seismicity. Recently, Köhler and others (2015) carried out a study of glacier seismicity in the Spitsbergen, providing a unique opportunity to verify our classification algorithm and compare results obtained using two independent methods. The proximity of seismic stations to the large glacial systems makes Spitsbergen’s conditions suitable for our research.

We used continuous seismic records from two seismological broadband stations, HSPB and KBS, which are located in Spitsbergen (Fig. 1). The HSPB station is located in the Hornsund fjord near the Polish Polar Station in southern Spitsbergen. This station belongs to the Polish Seismological Network (PL), which is operated by the Institute of Geophysics, Polish Academy of Sciences (IG PAS). In 2007, as part of the International Polar Year activities, the station was upgraded with the help of the NORSAR organisation, and a three-component (3C) broadband seismometer (STS-2) with sampling rate of 100 Hz was installed. This enabled the analysis of distant (teleseismic) earthquakes (e.g., Wilde-Piórko and others, 2009).

The HSPB station is located near the Hanskebreen glacier, the largest glacier in the area (Fig. 1c). The distance to its terminus is ~3 km. Hansbeene is a 16 km long polythermal tide-water glacier covering an area of 56 km$^2$. Its frontal cliff is ~1.5 km wide and 30 m a.s.l. (Blaszczyk and others, 2009; Grabiec and others, 2012). Another glaciers to the north from the station are Tuvbreen (tributary of Hansbeere) and Werenskioldbreen.

The KBS station is located in Kongsfjorden, Ny-Ålesund, in western Spitsbergen (Fig. 1b). The KBS station belongs to the Norwegian National Seismological Network. It is operated by the University of Bergen and equipped with the same type of broadband seismometer (STS-2) with sampling rate of 40 Hz.

The largest glaciers in the proximity of the station are Kroenbreen, Kongsvengage, Kongsbreen, Uvbergsbreen, Blomstrandbreen and Comfortbreesen (Fig. 1b). The distance to the combined terminus of Kroenbreen and Kongsvenge is ~15 km. The Kroenbreen’s surface is 445 km$^2$ (Trusel and others, 2010) and it contributes significantly to the overall area of the local glacial system.

Both datasets are freely available online at various seismological open data repositories, for example through the IRIS data management centre (http://ds.iris.edu/ds/nodes/dmc/). In case of the HSPB station, we used the data recorded between 2010 and 2014, available in the online Orfeus database (http://www.orfeus-eu.org/). Data recorded between 2008 and 2009 were taken from an in-house seismological database at the IG PAS. Due to the lack of data in the last quarter of 2009 (due to the station maintenance), we used the last quarter of 2007 instead to show year-to-year comparisons and overall statistics in order to maintain unbiased data.

In case of the KBS station, we processed the data between 2010 and 2014. The analysed period differs from the HSPB station due to a lack of continuous KBS raw data in the online IRIS database.

The meteorological data for the Hornsund region comes from the GLACIO-TOPOCLIM database (http://www.glaciopclim.org/) – currently unavailable online), while the dataset for Ny-Ålesund was downloaded from the eklima (http://www.eklima.met.no/) service.

3. METHODOLOGY

Seismic signals of various origins, referred to hereafter as ‘events’, were expected to occur in the analysed datasets. First, there are signals of glacial origin that we focus on in this study and records of local, regional and teleseismic earthquakes, which are the main objectives of seismological measurements. There are also signals of anthropogenic origin or those caused by natural phenomena other than glacier activity, such as ocean waves, hail, animals, etc., which we treat as disturbances. Our intent is to automate the processing of large volumes of continuous data for the detection and classification of glacier-induced seismic events. The sequence of processing procedures and their parameters were chosen to produce an autonomous processing sequence, which is easy to implement for any dataset. Each dataset was processed with strictly the same procedure and
parameters to provide results that would not be biased by data processing. A two-step data analysis procedure consisted of the following: (1) basic event detection; and (2) fuzzy logic event classification. The following subchapters contain a detailed description of our processing scheme.

3.1. Basic event detection workflow

At this stage, the following five processing steps were applied to the data:

1. Bandpass filtering (Butterworth filter with cut-off frequencies of 1 and 15 Hz);
2. Detection based on the ratio of short-term average to long-term average (STA/LTA) of the signal;
3. Multiple detection removal;
4. Elimination of the weakest events;
5. Estimation of the duration times of detected events and the elimination of events with duration times longer than 25 s.

A similar procedure is often used in earthquake seismology in a variety of approaches, and it is commonly implemented in more advanced acquisition systems to routinely pre-process the data stream from a seismic network. It was also used in the context of calving by Walter and others (2013), but their study was focussed on detecting seiches - normal modes of fjord waters induced by calving. Although the general processing scheme was the same as ours, seiches have very low frequency content and require another parameterisation.

The first step of our workflow – bandpass filtering – was chosen to preserve the frequency bands of expected glacier-generated signals (Górski, 2004; O’Neill and Pfeffer, 2007; O’Neill and others, 2007; Walter and others, 2010; Köhler and others, 2015) and to remove low-frequency microseisms generated by ocean waves, low frequency signals from teleseismic earthquakes (Stein and Wysession, 2003) and high-frequency noise.

For the 1–15 Hz filtered data, we ran a standard STA/LTA (Allen, 1978) detection algorithm on each component of the signal independently. Each time the ratio of running averages exceeded the chosen threshold, a detection was triggered. Then, multiple detections were removed by introducing a compulsory 5 s offset between subsequent detections. All of the remaining detections were saved as 50 s intervals of 3C records.

Then, the weakest events with low signal-to-noise ratios (SNR) within the saved interval were eliminated, because they were prone to be not reliably described by energy flow parameters. We calculated a mean power (energy per time) of the whole 50 s record and the mean temporal power over short time intervals. Events with no interval of signal power exceeding the whole record mean power value by at least 30% were discarded as unreliable and noisy.

For each remaining event, we calculated its duration time by applying a modified normalised energy density (NED) function (Sarma, 1971). The original NED function is given by

\[ NED(t) = \sum_{n=1}^{N} (U(n))^2, \]

where \( n \) is a signal sample index, \( U(n) \) is ground velocity at sample \( n \), \( t \) is a time index within a selected time window. NED is a cumulative sum of energy calculated from some arbitrary starting time. As we are not interested in nominal values of the energy, the obtained function is then normalised to one and plotted versus time (or sample index). The steeper its slope ascents the more energy is being recorded. The slope of the function in the earthquake occurrence time interval was originally used to calculate the energy flux and the duration of strong earthquakes. However, in our case, the signal was often not much stronger than the noise, which results in a significant contribution from noise energy to the NED function. Thus, to remove the background noise influence, which was negligible for strong earthquakes considered by Sarma (1971), we modified NED by subtracting a linear noise function (with slope \( N_f \)). The value \( N_f \) was calculated for each event using 4 s long signal interval (400

![Fig. 1. Study area: (a) general map of Svalbard with the KBS and HSPB seismological stations marked with red dots. (b) Enlarged KBS station area. (c) Enlarged HSPB station area. Modified from an online Map of Svalbard: http://toposvalbard.npolar.no/](https://www.cambridge.org/core/core/terms).
samples for 100 Hz sampled signal) 16 s prior to triggered
detection by:

\[ NF = \frac{\sum_{n=1}^{400} |U(n)|}{400} \]  

Then the modified \( mNED \) function was calculated for each

\[ mNED(t) = \sum_{n=1}^{t} |U(n)| - NF \cdot t. \]  

The character of \( mNED \) function and the effect of \( NF \) subtrac-
tion is well visible in Figure 2, where \( mNED \) function without
\( NF \) subtraction rises significantly in the pre-event time inter-
val. After \( NF \) subtraction, most of the \( mNED \) function value
change is related to the energy of the event, not the noise.

We defined the time needed for \( mNED \) to rise from a value
of 0.15–0.85 as the duration time of the event. Theoretically,
for small amounts of noise, this limit could be relaxed (i.e.,
the lower limit could be closer to 0 and the higher limit
could be closer to 1), but the chosen limits constrain the func-
tion to be more resistant to occasional rapid noise
fluctuations.

With the duration times of all of the events, we discarded
those with duration times exceeding 25 s, preserving typical
short-duration glacial-related events (Métxian and others,
2003; Bartholomäus and others, 2012; Köhler and others,
2012, 2015; Górska, 2014) and rejecting most of regional
earthquakes, which had longer duration times.

The detection procedure described above resulted in a
total amount of 8876 detections between the years 2008
and 2014 for the HSPB station. However, except the
glacier-triggered events, this dataset also contained tectonic
earthquakes and false detections, which were triggered by,
for example human activity near the Polish Polar Station,
which was too complex to be excluded by the described
workflow. Therefore, to automatically distinguish between
non-glacier- and glacier-induced seismic events, we further
developed a fuzzy logic classification algorithm.

### 3.2. Fuzzy logic event classification

The essence of fuzzy logic is to use variables, ranging
between 0 and 1, instead of using standard Boolean (two-
valued) algebra. It is useful when the definition of the deci-
sion criteria is subjective. For instance, it can be difficult to
decide whether a 1.8 m man is tall or not. Instead, using a
fuzzy logic approach, one defines membership functions
(e.g., linear, polynomial, windowed, etc.), which describe
to what extent a given height can be treated as tall or short.
Adding more parameters and associated membership func-
tions allows for the description of more complicated pro-
blems. The information obtained shows to what degree
each object, characterised by chosen variables, belongs to
each of the user-defined groups (Zadeh, 1965).

We defined four categories of events. Two categories repre-
senting glacier-induced events: ‘Low-frequency glacier-
related’ (shortly ‘LF glacier-related’) and ‘High-frequency
 glacier-related’ (‘HF glacier-related’) and two categories
accounting for events we wished to eliminate: ‘Tectonic earth-
quake’ and ‘False detection’. We aimed at analysing low
frequency glacier activity like ice-vibrations and short-duration moulin tremors (‘LF glacier-related’) induced by inter-glacial water flow (Górski, 2004; Röösli and others, 2014) as well as calving icequakes, which dominating frequencies can vary depending on authors from 2 up to 10 Hz (Bartholomaus and others, 2012; Köhler and others, 2015; Koubova, 2015) or even more (O’Neel and others, 2010).

To determine parameters of specific groups and associated membership functions we used HSP seismograms of confirmed earthquakes (from NORSAR catalogue (http://norsdata.no/NDC/bulletins/regional) containing reviewed regional earthquakes, mainly belonging to Storefjorden sequence (Pirli and other, 2013)), clearly noisy signals, which were formerly detected as icequakes and visually inspected events, which followed ice-vibration characteristics given by Görski (2004). Noise and earthquakes recognition was based on energy flow and duration time, while ‘LF glacier-related’ and ‘HF glacier-related’ on energy flow, duration time and frequency content criterion.

We selected four input parameters for each event based on signal power $P$ (energy per time):

1. $p_1$ – the number of time intervals at which temporal power exceeds mean power ($P_{temp} > P_{mean}$), despite of exceedance duration,
2. $p_2$ – the total length of time intervals (in samples) longer than 5 s at which temporal power exceeds mean power ($P_{temp} > P_{mean}$),
3. $p_3 = \frac{P_{max}^{1} - P_{mean}^{1}}{P_{max}^{1} - P_{mean}^{1}}$,
4. $p_4 = \frac{P_{max}^{1} - P_{mean}^{1}}{P_{max}^{1} - P_{mean}^{1}}$.

where $P$ is a smoothed signal power over time and $f_1$, $f_2$, $f_3$ are frequency bands: 1–5 Hz, 6–10 Hz and 11–15 Hz, respectively. Hence, for example $P_{mean}^{1}$ denotes the mean signal power in 6–10 Hz frequency band. All of the $P_{temp}$ values were smoothed (running average with 1 s window length) to obtain more stable values while calculating the $p_i$ parameters.

The parameter range for each category is given as follows:

1. Tectonic earthquake – strong and steady power flow with up to two onsets ($p_1, p_2 \leq 2$) that exceeds the mean value for at least 20 s ($p_2 \geq 20$).
2. False detection – short power bursts several times exceeding the mean value ($p_1 \geq 7$) or single impulse, spike-like signals ($p_2 \leq 1$).
3. Low-frequency glacier-related – signals with a dominant frequency band of 1–5 Hz ($p_1 \geq 1$, $p_4 \geq 1$), lasting from a few to more than a dozen seconds ($p_1 \leq 5$, $5 \leq p_2 \leq 20$)
4. High-frequency glacier-related – signals with a dominant frequency band of 6–15 Hz ($p_1 \leq 1$ or $p_4 \leq 1$), lasting from a few to more than a dozen seconds ($p_1 \leq 5$, $5 \leq p_2 \leq 20$).

The fuzzy logic algorithm was created in MATLAB using the Fuzzy Logic Toolbox and is based on the four input parameters $p_i$. The value of each parameter is evaluated by a membership function, which determines to what degree particular value fulfills class condition (e.g., Gaussian, which expected value is a desired one, while acceptance for deviation is expressed by standard deviation) (Fig. 3a). A sum of membership functions values for all criteria of a given class provides an assessment of the degree to which the event represents a given class (Fig. 3b). Then, the best-suited class was chosen (Fig. 3c).

Sample waveforms representing each event class are shown in Figure 4.

### 3.3. Verification of the classification algorithm

We tested the classification algorithm using a dataset kindly provided by Andreas Köhler, consisting of STA/LTA detections times at the KBS station that corresponded to visually observed calving events at Kronebreen. The data covered two periods: 14–26 August 2009 and 5–15 August 2010, and was part of the dataset that has been used in Köhler and others (2015). The latter period of the test dataset overlaps with dataset used in our work. Eleven out of 172 events were found in our catalogue and all of them were classified as ‘HF glacier-related’. Such a small number of detection is caused by low SNR of events in the test dataset: 83% of events has SNR < 2 (maximum amplitude divided by RMS of the whole signal filtered by 2–15 Hz bandpass filter) while SNR of all detected events was higher than 6. Detection threshold of STA/LTA algorithm used in our study was 3.

To further test the algorithm we performed our processing workflow, treating provided times as forced detections. We used test dataset consisting of 585 registrations from years 2009 and 2010. Such an approach resulted in rejection of 452 events, while 123 events were classified as follows: 1 ‘tectonic’, 8 ‘LF glacier-related’, 43 ‘HF glacier-related’ (in total 51 glacier-related), 71 ‘false’. Hence, the ratio of ‘false’ to ‘glacier-related’ was 0.72 (51/72). High number of rejected events is again caused by low SNR. Next, we decided to run only the classification algorithm using events with SNR > 2 only (102 events). The result was: 3 ‘tectonic’, 10 ‘LF glacier-related’, 38 ‘HF glacier-related’, 51 ‘false’. Hence, the ratio of ‘false’ to ‘glacier-related’ grew to 0.94 (48/51). Obtained results show that the detection threshold of the algorithm is high, because events which are detected and classified display high SNR. Lowering SNR threshold of processed events by forcing the detection resulted in increased number of events in ‘false’ group. When analysing events with high SNR, calving events from KBS tend to be classified as ‘HF glacier-related’. However, due to relatively wide frequency range of calving icequakes spectra, some of them were classified as ‘LF glacier-related’.

### 4. RESULTS

Applying our detection procedure to the continuous HSPB seismic data, we identified 8876 seismic events. Then, the fuzzy logic classification algorithm was applied. Its outcome is presented in Figure 5. There are two categories of events: ‘Tectonic’ and ‘False’, which are not glacier-induced and were therefore excluded from further analysis. Two groups remained: ‘LF glacier-related’ (1467 events) and ‘HF glacier-related’ (3533 events). Hence, 7020 detections were classified as glacier-induced (Fig. 6), while 1856 were not (grouped as ‘Tectonic’ and ‘False’).

We applied the same workflow to the KBS dataset, which was recorded between 2010 and 2014, acquiring 17 711 detections then the fuzzy logic classification procedure resulted in classifying 14 913 events as glacier-related.
(3275 ‘LF’ and 11 638 ‘HF’) and 2798 events as tectonic or false. Figure 7 presents various distributions of those events.

5. DISCUSSION

5.1. Classification

We detected several thousand events and classified them into four categories: ‘Tectonic Earthquakes’, ‘False Detections’, ‘LF glacier-related’ and ‘HF glacier-related’. The first two categories were clearly not related to a glacier and, therefore, rejected. The difference between ‘LF’ and ‘HF’ class was the domination of one of the higher frequency bands in the signal frequency spectrum, instead of the 1–5 Hz representing ‘LF’. A validation of classification criteria showed that the melt-season calving icequakes of considerable SNR are classified as ‘HF’ glacier-related much more likely than as ‘LF’, while noisy or weak calving icequakes are classified as ‘False’.

There is a significant disproportion between the obtained catalogues. Although HSPB station is situated closer to the nearest glacier, there is over two times more events detected at KSRe, regardless a shorter time span. This disproportion can be explained by many factors, which have much stronger cumulative effect than the distance alone. First of all, flow rates of the glaciers differ between the two sites. Kongsbreen and Kongsvegen are among the fastest glaciers in Svalbard (Schellenberger and others, 2015), with Kongsbreen velocity ca. 750 m a⁻¹ (Melvold and Hagen, 1998), which is over three times faster than Hansbreen (ca. 230 m a⁻¹) (Grabiec and others, 2012). Also, the width of the Kronebreen terminus is two times wider than Hansbreen’s, implying a higher calving rate at Kongfjorden. Furthermore, Kronebreen interacts with the Kongsvegen glacier (Trusel and others, 2010), generating seismic events that are localised in the interaction zone (Koubova, 2015) up to 5 km from its terminus. The disproportion can also be linked with the differences in the glacier’s geographical exposure to sea circulation. The ocean temperature on the western Spitsbergen coast can vary significantly, depending on the actual range of the West Spitsbergen Current (Walczowski and others, 2012), resulting in different calving rates along the coast. Finally, the background noise level, which determines the smallest detectable events, is higher for the HSPB than for the KBS dataset.

5.2. Seasonal distribution and correlation with meteorological data

The seasonal distributions of both ‘glacier-related’ groups follow the seasonal glacier seismic activity pattern (Jania and others, 1988; Ekström and others, 2006; Bartholomäus and others, 2015b) and illustrate the changes in the long-term seismic activity of glaciers in the vicinity of analysed stations. Figure 6a shows the periodicity of glacier-induced events and the year-to-year changes at the HSP station. The seasonal distribution follows the same pattern each year. During winter and spring when glacier is usually advancing activity remains at the base level; then, it intensifies during the melt season from June to November, with a peak in August and September. We found this scheme to be valid for all of the analysed years, except 2011. That year, the number of ‘HF’ events in July and August is significantly lower compared with May and September. This exceptionally low number of ‘HF’ events in 2011 melt season, showing no distinct peak in August nor in September is most probably caused by a presence of dense sea-ice floes drifting in Hornsund fjord, what had drastically decreased the Hansbreen’s calving intensity by cooling down surface water and suppressing waves (Petličkiová and others, 2015). However, the distribution of ‘LF’ events is not affected severely, what is consistent with results of classification.
algorithm validation and suggest that ‘LF’ events sources are mainly distinct from calving icequakes.

The monthly detection distributions summed over years are shown in Figure 6b along with the mean monthly temperatures and summed precipitation curves registered at Polish Polar Station in Hornsund. We observed a 1 month delay between the temperature and ‘glacier-related’ events peak. A Pearson’s correlation coefficient between the monthly ‘glacier-related’ events distribution and the mean monthly temperature equals 0.79, whereas the one calculated with a 1 month lag equals 0.85. In the case of the summed precipitation data, we obtained a correlation coefficient of 0.82 (no time lag observed).

Figure 6c presents the total number of ‘glacier-related’ events every year since 2008. It shows the mean temperature and summed precipitation during the period between June and November each year, which is the most active period of glacier seismic activity and for whole year. We noticed almost a doubling of the number of ‘HF’ events in 2013 compared with previous years, while ‘LF’ events distribution is kept stable. This high level of ‘HF’ activity was accompanied by a noticeably steady growth in the mean temperature in warm months (June–November) of 1.5°C over the analysed 7 year period. Although the correlation coefficients for the seasonal distribution were very high, they decreased severely for the year-to-year ‘glacier-related’ data suggesting that there is no correlation. Only correlation coefficient between ‘LFGR’ events and precipitation remains high: 0.78 for whole year and 0.84 for melt-season, confirming relationship between interglacial water flow and low-frequency seismic events.

Similar to the HSPB dataset, the monthly distribution of ‘glacier-related’ events near Ny-Ålesund correlates well with the number of days with mean monthly temperatures and summed precipitation measured in Ny-Ålesund (Fig. 7b), reaching the highest correlation coefficient of 0.85 (with a 1 month lag) and 0.79 (no lag) for temperatures and 0.82 for precipitation. Contrary to HSP dataset we observe high correlation coefficients for year-to-year all events distribution (Fig. 7c) with whole year summed precipitation (0.91), as well as for mean temperatures (0.88).

We observed a 1 month delay in the peak glacial seismic activity with respect to the air temperature in both datasets (Figs 6, 7). This phenomena, as well as its exceptional absence in 2012 in the Kongsfjorden region supports the theory that the sea temperature, not the air temperature, at
During the winter of 2011/12 temperatures higher than usual had been registered both in Hornsund and Ny-Ålesund meteorological stations. In January 2012 an extraordinary winter peak of ‘LF’ detections number was registered at KBS station, while there was no evidence of such peak at HSP station. However, the peak at KBS station clearly coincides with an extreme rain event in Kongsfjorden region (Fig. 2a). Moreover, the number of glacier-induced events is directly proportional to the precipitation (Fig. 8). Nearly all events registered between 28 and 31 January 2012 were classified as ‘LF glacier-related’ and have dominant frequencies of 1–2 Hz. There is no distinct peak in January 2012 in the study oriented to find calving events by Köhler and others (2015), most likely due to using higher frequency band (2–8 Hz). Schellenberger and other (2015) show that there was no observable increase of Kongsbreen flow in GPS velocity maps, hence the intensification of glacier seismic activity might not be connected to increased glacier flow. Registered waveforms match the characteristics of both ice-vibration (Górski and Teisseyre, 1991, Górski, 2004) and short-duration ‘moulin’ tremor (Métxian and others, 2003; Röösl and others, 2014), caused by resonance in fluid-filled cracks, in this case due to massive rainfall, rather than to calving icequakes, what indicates connection of the interglacial water flow with the low-frequency seismic emissions.

Observed increase in number of glacier-related events since 2011 registered at KBS station (Fig. 8a, c) agrees with observations of significantly increased ablation front in 2011–2013 of both Kronsbreen and Kongsbreen (Schellenberger and others, 2015).

5.4. Periodicity

Bartholomaus and others (2015b) claim that the periodicity of iceberg calving coincides with principal lunar semidiurnal tide. Following this work, we decided to analyse periodicity of detected ‘glacier-related’ events. We summed glacier-induced detection times in short-duration bins (3.83 h) and then computed the Lomb-Scargle periodogram, which is suitable for power spectra representation for irregularly spaced data (Press and others, 1992). For KBS dataset we recognised a distinct peak at 0.5179 d (1/1.931 d) coinciding with the principal lunar semidiurnal tide (Fig. 9), same as Bartholomaus and others (2015b) found for the Yahtse glacier. There was no evidence for this peak at HSP dataset, which we believe is caused by too sparse time coverage of events and lower calving rate. Recognised peak can support a connection between variations in sea level and increased calving showed already by only two authors (O’Neel and others, 2003; Bartholomaus and others, 2015b).

5.5. Algorithm’s performance

The developed algorithm was implemented on a PC desktop computer, which was able to process a few yearlong continuous datasets within 1 d. Thus, this approach can be easily implemented as a routine tool for real- or near-real-time glacier activity monitoring. The parameterisation of the grouping conditions of the fuzzy logic algorithm might differ for various datasets because of different factors such as the noise level, distance to the sources in a glacier or characteristics of background noise, however, after initial inspection of the detected signals, the chosen thresholds and parameters can be adjusted.

The performance of our method can be significantly improved by using more than one station, which would allow for the assessment of event locations, which are hardly possible to obtain using a single station (Asming and Fedorov, 2015). Without a location assessment, we cannot...
reliably affiliate the detected seismicity with any particular glacier. Consequently, we affiliate them with the station area, i.e., with the largest ice masses surrounding the station. On the other hand, presented method can be applied if there is only a single station available in the area.

6. CONCLUSIONS

We developed an automatic algorithm to detect glacier-related seismic events using the continuous seismic records of a single seismic station. Using fuzzy logic, our algorithm classifies detections as glacier-related or non-glacier-related. We used recordings from two broadband seismological stations, HSPB in southern Spitsbergen and KBS in western Spitsbergen, to analyse glacial dynamic activity near the Hansbreen and Kronebreen glaciers. We detected and classified more than 8000 events over a 7 year time span (2008–2014) in the HSPB station region, classifying 7020 of them as glacier induced. We also detected more than 17000 events throughout a 5 year time span (2010–2014) for the KBS dataset, with 14 913 of them classified as glacier-induced.

In recent years, the number of detected glacier-related events in the analysed regions of Spitsbergen increased significantly. In the HSPB dataset, they doubled over the last 2 years (2013/14), whereas in the KBS dataset, a steady increase from year to year was observed.

The cumulative seasonal event distribution correlates well with the seasonal temperature variations. The highest correlation coefficients (0.95 and 0.96) were observed between the glacier seismic activity and the number of days with positive mean temperatures delayed by 1 month for both datasets. Correlation coefficients with 1 month delayed mean temperatures and summed precipitation were also high. A year-to-year cumulative event distribution revealed much weaker correlations with meteorological data or no correlations at all.

Weather phenomena like, for example extreme rainfall can distinctly contribute to the seasonal distribution of glacier-induced seismic events, as observed in January 2012 in Kongsfjorden. Distribution of the rainfall-triggered events is dominated by the low-frequency events rather than typical calving events. Hence, the group of ‘LF glacial-related’ events should instead be affiliated with the internal processes within a glacier, like interglacial water flow.

Large cover of floating ice can efficiently decrease calving intensity throughout whole melt season, as happened in 2011 in Hornsund. Results from Kongsfjorden seem to...
confirm the idea that the principal lunar semidiurnal tide can contribute significantly to calving periodicity. Our results demonstrate the possibility of using long-term seismological observations from a single permanent seismic station to study the dynamic activity of glaciers located in their proximity. The time-efficient automatic event detection and classification algorithm based on fuzzy logic can be used to discriminate automatically and objectively between non-glacier and glacier-related events in real time, revealing temporal changes and long-term trends in glacier dynamics.

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Fig. 9. Power spectrum of ‘glacier-related’ seismic events detection times at KBS station calculated by applying the Lomb-Scargle algorithm to the demeaned, detrended detection times distribution in 3.83 h bins. The frequency peak of 1.931 d−1 corresponding to K1 principal lunar semidiurnal tide (1.932 d−1) is labelled.


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