Environmental forcings and micro-seismic monitoring in a rock wall prone to fall during the 2018 Buran winter storm

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Abstract
This study reports a comparative analysis of the environmental conditions and micro-seismicity recorded on a rock wall resulting from an intense meteorological event. The findings are focused on a quarry wall located in the Acuto Field Laboratory (Central Italy), where multi-parametric environmental monitoring is operating and an Artificial intelligence Camera Prototype has been placed to detect rock falls reaching a railway target. Six accelerometers were installed to detect micro-seismic events caused by the expected strong thermal transient caused by the Buran storm on February 2018. Within a few hours, a steep decrease in the average air and rock mass temperature down to 8 °C was recorded, and −4 °C and −8 °C were reached for the rock and air temperatures, respectively. A total of 103 micro-seismic events were analysed with respect to both rainfall and thermal forcing: while no correlation with rainfall was reported, the steep thermal transient was responsible for the strain effect that occurred during the heating phase of the rock mass following the Buran storm. An elastic deformation event with a maximum daily amplitude of 165 μ strain was recorded by the strain gages installed on the mm-joints due to the rock heating and cooling caused by the variation in temperature. The collected evidences show the relevance of short thermal transients in modifying stress conditions within rock masses and their relationship to a peculiar micro-seismic response. The main outcomes established the key role played by integrated monitoring systems to better understand the relationship between vibrational behaviour and environmental forcings in terms of understanding the precursors to rock failure.

Keywords Rock fall · Micro-seismic monitoring · Forcings analysis · Slope stability
1 Introduction

Rock slope stability is currently investigated through several approaches and techniques as rock falls contribute to landslide hazards due to their fast movement and the numerous factors that can trigger their occurrence (Varnes 1978; Menéndez Duarte and Marquinez 2002; Fischer et al. 2012; D’Amato et al. 2016). A contribution to risk mitigation strategies can be provided by monitoring and studying several preparatory factors (to both transient and continuous forcing) that when combined, can influence rock fall occurrence (Gunzburger et al. 2005). More specifically, thermal stresses (Gischig et al. 2011; Wang et al. 2016; Grämiger et al. 2018), rainfall (Krautblatter and Moser 2009; Delonca et al. 2014), and seismic inputs (Moore et al. 2012; Koukouvelas et al. 2015) combined with long-term deformation (Brückl and Parotidis 2005; Grøneng et al. 2010; Xu et al. 2014b) are among the main causes of rock mass damage. These phenomena can progressively lead to the development of weak sectors within the rock mass that have the potential to evolve into failures. In fact, small-size failures usually originate from highly jointed rock masses that have been stressed by recurrent forcing, such as rainfalls and daily thermal cycles that can induce freezing and thawing events (Matsuoka and Sakai 1999; Gunzburger et al. 2005; Mateos et al. 2012; Collins and Stock 2016; Dietze et al. 2017; Matsuoka 2019). All of these factors are responsible for the accumulation of inelastic deformations within the rock mass, leading to permanent damage, i.e., the non-negligible reduction of mechanical strength. Although the role of small-size failures in controlling rock slope stability seems to be self-evident, how these processes interact and progress over time is still far from completely understood.

Several monitoring strategies are usually adopted for managing the risk associated with rock falls by employing sensors specifically aimed at analysing the effects (i.e., elastic or inelastic) induced by natural and anthropic forcing that can negatively affect slope stability. While more traditional stress–strain sensors allow the direct investigation of specific portions of a rock mass by monitoring peculiar conditions across major joints (Loew et al. 2017), micro-seismic monitoring devices represent an opportunity for detecting incipient variations within a rock mass volume, providing information about the modification of its physical and/or mechanical properties (Papini et al. 2009; Xu et al. 2012; Colombero et al. 2016; Provost et al. 2018; Meyer et al. 2019) and quantitative indications of the vibrational and dynamic behaviour of prone to failure rock blocks. In particular, dynamic behaviour is assessed in terms of evaluating the seismic ambient noise (Bottelin et al. 2013; Kleinbrod et al. 2017; Valentin et al. 2017), increases in the recorded micro-seismicity (Senfaute et al. 2009; Levy et al. 2011) and the shifting of the main resonance frequency before a generalised collapse (Got et al. 2010; Lévy et al. 2010). Regarding environmental parameters, Helmstetter and Garambois (2010) studied seismic signals collected on the Séchilienne rockslide and observed a correlation between micro-seismic emissions and rainfall with the occurrence of rock falls. Similar outcomes were noted by Arosio et al. (2018), who analysed a three-year-long micro-seismic record collected on a limestone cliff. Regarding thermal forcing, the relevance of daily temperature variations and freeze–thaw cycles in the predisposition of rock falls has been widely documented (Mateos et al. 2012; Walbert et al. 2015; D’Amato et al. 2016). Additionally, the analysis of micro-seismic and acoustic emission datasets can help in monitoring ongoing damaging processes at a smaller scale (from metric to centimetric) (Xu et al. 2014a; Zhou et al. 2018; Colombero et al. 2018) since the inelastic strain associated with thermal oscillations is released as micro-seismic and acoustic emissions (Occhiena et al. 2012; Collins et al. 2014; Weber et al. 2018). Based
on the available literature, only a few studies have identified an increase in micro-seismic events during the rapid and intense lowering of air temperature, especially during the cooling phase, e.g. Amitrano et al. (2010) and Occhiena et al. (2012) at the Matterhorn site.

To study the complex interactions between meteorological forcings acting within rock masses and induced deformations, a permanent multi-sensor strain and environmental monitoring system was installed in a Field Laboratory located in an abandoned quarry and managed by the Department of Earth Sciences and the Research Centre for Geophysical Risks (CERI) of the University of Rome “Sapienza”. This monitoring system focuses on a 20 m³ protruding rock block that is kinematically predisposed to evolve towards failure and is sensitive to cyclical thermomechanical stress (Fiorucci et al. 2018). During an intense meteorological perturbation forecasted for the end of February 2018, during which a drastic drop of temperature associated with intense rain and snowfall was expected, an accelerometer array was installed to temporarily integrate with the permanent environmental monitoring system. In this study, the outputs from the monitoring system throughout the intense meteorological event are reported and discussed: micro-seismic data are compared with environmental (i.e., rainfalls, air, and rock mass temperatures) and deformational (i.e., extensometers and strain gauges) data to investigate the effects induced in the vibrational response of the rock mass to severe weather conditions.

2 The Acuto Field Laboratory

2.1 The multi-parametric monitoring system

To study the long-term rheological behaviour of rock masses exposed to environmental forcings (such as rainfalls, wind, daily and seasonal thermal cycles) and evaluate their role in inducing instability, a multi-sensor monitoring system has been installed since autumn 2015 in the abandoned quarry that was selected as a Field Laboratory (Fantini et al. 2017b). The quarry is located in the Acuto municipality 100 km SE of Rome and is characterised by a 500 m long and sub-vertical quarry front made up of Cretaceous limestone that reaches a height up to 50 m and was intensively pitted until the late 70 s. The monitored area is located in the NW sector of the quarry, where a 20 m³ rock block jutting out from the quarry wall is separated from the quarry wall itself by a decimetric-wide sub-vertical joint. The rock block has been instrumented with: (1) one thermometer for measuring the rock mass temperature installed at a depth of 8 cm from the free surface exposed eastward (sensitivity: 0.1 °C); (2) six strain gages (sensitivity: 1 μstrain) installed over the rock block on micro-joints; and (3) four extensometers (sensitivity: 0.01 mm) installed on the steeply dipping open joints that separate the rock block from the wall behind it (Fiorucci et al. 2020). Additionally, the monitoring system is completed by a fully equipped weather station placed at the top of the quarry wall that is composed of an air thermometer (measurement range −40 °C/+60 °C), a hygrometer, a rain gage (sensitivity: 0.2 mm H₂O, WMO standard) and an anemometer for determination of wind speed and wind direction (Fantini et al. 2017b). Meteorological, thermal, and strain data are sampled each minute and collected by a digital data logger set up to acquire data every minute. An Artificial intelligence Camera Prototype (AiCP) has also been installed in front of the quarry wall to detect rock falls impacting a potential target (i.e., a railway track) placed on the ground just below the monitored rock block (Fantini et al. 2017a). AiCP frame acquisition and processing are based on a change detection algorithm that continuously operates over a specified Region...
of Interest (RoI) focused on the railway track. Alerts are automatically sent if new volumes are detected within the RoI (Fantini et al. 2015, 2017a; Leone et al. 2016).

In addition to this permanent monitoring system, an array of seismic sensors was temporarily installed before the expected drop in temperature related to the incoming intense meteorological event. On 23 February 2018, 6 Bruel & Kjaer one-component accelerometers (type 8344: frequency range 0.2–3000 Hz, sensitivity 250 mV/ms\(^{-2}\)) were deployed on the sides of the protruding monitored rock block, spaced 15–20 cm apart along a horizontal alignment across major- and micro-joints over a space corresponding to the daily thermal active layer (Fiorucci et al. 2018; Marmoni et al. 2020). These sensors were tightened to studs previously fixed within the rock mass to ensure the best coupling with the rock mass matrix. Accelerometers 1, 2, and 3 were placed on the rock block; accelerometer 4 was fixed on a rock block portion delimited on both sides by two main joints; and accelerometers 5 and 6 were installed on the rock wall to monitor and weight the vibrational behaviour of both the rock wall and rock block (Fig. 1). In addition to the possibility of

![Fig. 1 View of the Acuto quarry wall where the Field Laboratory is hosted; in the foreground, the monitored rock block side is shown; the positioning and IDs of the accelerometers (from 1 to 6) and on-rock devices are marked a. The small frames provide details of some devices: fully equipped weather station (WS) b; thermocouple (TC) installed inside the rock c; joint gage (F01) on the main joint and strain gage (SG1 and SG2) on the -joint d; seismic data logger (SDL) e; Artificial intelligence Camera Prototype (AiCP) f]
investigating the vibrational behaviour and the effect of thermal cyclical forcings across major- and micro-joints, this set up also allowed the comparison of vibrational data with strain data collected by the closest extensometer and strain gages. In this arrangement, all the accelerometers were NS oriented or normally oriented with respect to the orientation of the rock block side. The accelerometers were connected via cables to an HBM digital acquisition system composed of a signal amplifier (SomatXR MX1601B-R) and a data processor (SomatXR CX23-R), which was set up to perform continuous mode acquisition with a sampling frequency of 2400 Hz. The power was first supplied by three 40 Ah lead batteries replaced every 3 days, and then, beginning 9 March 2018, by a 105 W solar panel equipped with a backup battery.

2.2 The Buran winter storm

Based on the environmental data collected since Autumn 2015 and the decennial weather dataset collected by Blasi (1994), the Acuto site belongs to a temperate sub-mountain climatic region characterised by dry summers and wet winters, with rainfall mainly distributed between late autumn and spring. Winters at the site are mild, with an average annual variable temperature of 12.0–13.6 °C, and the minimum average temperature of the coldest month is rarely lower than 0 °C. Annual rainfall exceeds 1400 mm (1430–1606 mm), with a few summer rains up to 173–200 mm each.

Although the weather conditions are not extreme, sudden temperature decreases, as well as snowfalls associated with meteorological perturbations, can occasionally strike this area with a recurrence of several years. During winter 2017–2018, Central and Eastern Europe were affected by an intense meteorological event characterised by the incoming cold air associated with the Buran, a strong and cold wind blowing from the northeast that originates in the steppes of the Sarmatic Plain. Between the end of February and March 2018, the air temperature reached a minimum of −40 °C in the Russian Plateau and tens of degrees below 0 °C in Italy, where significant snowfall covered northern and central Italy (Fig. 2).

At the Acuto Field Laboratory, a sharp temperature drop accompanied by snowfall occurred, with the latter contributing to keeping the air temperature below 0 °C for almost three subsequent days (Fig. 2). Marked temperature changes with amplitudes of up to 18 °C were registered for the air mass in one and a half days, during which minimum temperatures of -7.7 °C for the air and −3.8 °C for the rock mass were reached. This cooling scenario constituted a short thermal transient in the seasonal temperature trend, as indicated by variations higher than 8 °C recorded for the average air temperature during the Buran storm (26 February–1 March 2018) with respect to the period before and after it.

3 Materials and methods

Environmental and micro-seismic data were analysed throughout a time span of 18 days from 23 February to 13 March 2018 (i.e., the acquisition period of the micro-seismic monitoring system), during which a substantial difference between the air and rock mass temperature records was observed. Based on the air and rock mass average temperatures, two time windows were considered: (1) before and after the Buran storm (from 23/02/2018 00:00 to midnight of 26/02/2018 and from 02/03/2018 00:00 until the end of 13/03/2018, respectively), and (2) during the Buran storm period (from 26/02/2018 00:00 to 02/03/2018
Fig. 2 Weather forecast of cloud cover (a) and snow accumulation over 3 h, (b) released by CNR-ISAC for midnight of 26 February 2018. Resulting snow cover at the Acuto Field Laboratory after the Buran winter storm (c). Red stars indicate the location of the study area.

Table 1 Mean, maximum, and minimum temperatures recorded at Acuto Field Laboratory during and preceding/following the Buran storm

|                     | Buran                  | Pre—post-Buran          |
|---------------------|------------------------|-------------------------|
|                     | 26/02 00:00 ÷ 02/03 00:00 | 23/02 00:00 ÷ 26/02 00:00 ÷ 13/03 00:00 |
| Mean T rock (°C)    | 1.4                    | 7.1                     |
| Max T rock (°C)     | 9.7                    | 19.0                    |
| Min T rock (°C)     | −3.8                   | 2.5                     |
| Mean T air (°C)     | −1.2                   | 6.8                     |
| Max T air (°C)      | 9.1                    | 15.4                    |
| Min T air (°C)      | −7.7                   | −0.3                    |
00:00), which were characterised by mean temperatures of the rock matrix of 1.4 and 7.1 °C, respectively (Table 1). Approximately 10 days after the end of the Buran storm, during maintenance of the micro-seismic array, rock fall debris was directly surveyed in the area of the railway target. As the only rock fall that occurred in this quarry sector since the installation of the permanent monitoring system in 2015, the AiCP frames were also analysed to better constrain its temporal occurrence; moreover, additional analyses and surveys were planned with the aim of locating its source area and to estimate the detached volume (Figs. 3 and 4).

Two 3D models of the rock wall were compared: the first one obtained by a Leica MS60 Multistation before the Buran storm and the second one obtained following a photogrammetric survey performed in the days after the instability event. The pre-event 3D model was obtained using a sampling size of $5 \times 5$ cm, an angular precision of $1'$ and an accuracy of 1–2 mm for a distance of 50 m, while 158 images with a resolution of $3968 \times 2976$ pixels were considered for the post-event 3D model. Both models were scaled to the same resolution, and a volumetric change detection analysis was carried out using the M3C2 algorithm (Lague et al. 2013) managed by the CloudCompare open-source software.

### 3.1 Analysis of the micro-seismic dataset

The micro-seismic records collected during the 18 days of monitoring were analysed using the SAC software (Seismic Analysis Code; Goldstein et al. 2003; Goldstein and Snoke 2005) and customised scripts implemented in MATLAB. A 5 Hz high-pass filter was applied to all the datasets to adjust for the inaccurate low-frequency response of the installed accelerometers. A routine based on the STA/LTA algorithm (Allen 1982) was applied to automatically detect micro-seismic (MS) events over ambient noise variations to perform an objective analysis of the seismic records and to avoid uncertainties and possible loss of information in the recognition of the MS events. The mean of the amplitude in a short-time window (STA) was compared to the mean of the amplitude calculated over a longer time span (LTA) considering the absolute value of the seismic records. If the STA/LTA ratio exceeded a user-defined triggering threshold, the event was registered until the ratio decreased to a de-triggering threshold. In this case, STA was set to 0.01 s, LTA to 1 s
and STA/LTA to 4 s for both the triggering and de-triggering thresholds. The pre-and post-event times were set to 0.5 s to ensure the selection of the entire event. In the case of the repeated exceedance of the established STA/LTA ratio within a time span of 2 s, multiple events were recorded as a single event. The input parameters for the STA/LTA routine were selected following several iterations to obtain the best trade-off suitable for identifying events. Following Trnkoczy (2012), the STA duration selected was shorter than the shortest event expected, and the LTA length progressively varied until the obtained detections included reliable waveforms, validated by visual screening.

Furthermore, to reduce false positives due to signals not related to the rock mass deformation (i.e., single spikes, raindrops hitting surfaces, etc.), only quasi-simultaneous events were considered by selecting those detected on more than one sensor within a time span of 0.1 s. Signals recorded on at least two sensors returned background information about the intensity and time distribution of the events but were still affected by the presence of erroneously selected spurious signals. For this reason, the discussion only considered events detected by at least five sensors, further subsampled following a manual check of the waveforms and spectrograms to discard problematic MS signals, evaluating the coherence in the FFT and spectrograms for each signal using the maximum frequency from the spectrogram.

Fig. 4  a Picture of the main rock fall scenario with the detachment area (fallen rock volume of approximately 1 m$^3$) and accumulation surface highlighted. The monitored rock block, the AiCP, and the target RoI are also shown; b debris that reached the railway target; c background scenario by the AiCP with the RoI bounded by a blue dashed frame; d scenario resulting from change detection analysis carried out by the AiCP; white pixels inside the RoI represent the detected fallen blocks.
To weight the role of the recorded severe weather conditions in inducing micro-seismicity within the monitored rock mass, the micro-seismic dataset was compared to the rainfall intensity and the rock mass temperature datasets. Two subsets of MS data were considered according to the different temperature conditions that occurred in the defined pre/post-Buran and Buran periods; each subset was further dichotomised into rainy and dry periods to obtain four subsets of the MS events (Table 2). In particular, the time intervals were classified as “rainy” when the cumulative rainfalls were not null for at least 2 h. As a result, the MS events were counted by maintaining the environmental conditions, allowing a comparison of the occurrence of the MS events with the temperature trends in the absence of rainfall and the evaluation of the effect of rainfall during stationary temperature conditions.

To evaluate the possible correlation between micro-seismicity and rainfall, the validated MS events that occurred in the pre/post-Buran periods were clustered according to the classes of total cumulative rainfall in two hours (< 1 mm; 1 ≤ mm < 3; ≥ 3 mm). To exclude the possible interference due to rainfall, the MS events that occurred during dry periods (both during and pre/post the Buran storm) were selected and clustered with respect to the thermal gradient derived in two hours and to intervals of 2 °C in absolute temperature. The 2-h interval was considered to better observe temperature variations and was then used to accumulate rainfalls, with the aim of clustering MS events that occurred within the same time intervals. The classes considered for cumulative rainfalls were set arbitrarily.

Moreover, following Helmstetter and Garambois (2010) and Arosio et al. (2018), cross-correlation between the MS dataset, rainfall, and rock temperature gradient was performed. Cross-correlation is a mathematical operator that allows the similarity between two time series to be compared when provided, providing their offset as a temporal lag dependent on the sample step of the inputted time series, which must be the same for both. Rainfall and MS data were cumulated hourly, and the rock temperature gradient was accordingly derived. Periods with missing values from the MS data (due to a lack of power supply) were discarded from all the datasets to not affect the final results. Then, cross-correlation was performed on three time series, which considered the entire monitoring period, the Buran storm phase, and the pre/post-Buran period.

### Table 2

| Periods          | Buran (max/day) | Pre/post-Buran (max/day) | Total |
|------------------|-----------------|--------------------------|-------|
| Rainy            | 39 (20 on 28 February) | 20 (7 on 2 March) | 59    |
| Dry              | 40 (16 on 1 March)  | 4 (2 on 2 March)     | 44    |
| Total            | 79 (36)         | 24 (9)                  | 103   |

Differential analysis of the 3D models of the quarry wall enabled the examination of the detachment area of the rock fall that was released from the westernmost portion of the rock wall, approximately 10 m far from the monitored rock block (Figs. 3 and 4), which corresponded to a highly jointed portion of the rock mass. This comparison also enabled estimation of a rock fall volume of approximately 1 m³.

### 4 Results

Differential analysis of the 3D models of the quarry wall enabled the examination of the detachment area of the rock fall that was released from the westernmost portion of the rock wall, approximately 10 m far from the monitored rock block (Figs. 3 and 4), which corresponded to a highly jointed portion of the rock mass. This comparison also enabled estimation of a rock fall volume of approximately 1 m³.
Analysis of the frames collected by the AiCP facilitated the identification of 13 dm rock blocks in the debris within the RoI (Fig. 4), while another rock block was visible at the apex of the rock fall deposit (Fig. 3). Moreover, by considering the frames with and without the rock fall deposit, it was possible to establish that the time window in which the rock fall occurred ranged between 11 and 12 March. Unfortunately, at that time, the micro-seismic monitoring was not operating due to a lack of power supply; therefore, the vibrational signal of the rock fall impact was not recorded.

However, excluding short-time gaps due to a lack of power supply, the accelerometer array was operational for most of the 18-day monitoring period, as highlighted by the distribution of the 18,802 events detected by at least two sensors scattered throughout the whole period, particularly during the Buran storm phase (Fig. 5a).

To reduce false positives, only the events picked by 5 accelerometers were considered (1646 detections). From this dataset, a manual selection of the waveforms was carried out.

![Fig. 5](image-url)
to guarantee a strongly restrictive approach, allowing the identification of 103 MS events that had a high quality wave form and a consequently high certainty of the event type. These latter events were characterised by very short-time durations, i.e., in the order of hundredths of seconds and by a wide-varying frequency content ranging from 30 to 1200 Hz. The selected MS events were categorised according to their time of occurrence: 44 occurred during dry periods (40 in the Buran storm and 2 in the pre/post-Buran storm) and 59 occurred during rainy periods (39 in the Buran storm and 20 in the pre/post-Buran storm) (Table 2). The maximum daily rate of MS events during the Buran period occurred on 28 February, during which there were 20 and 11 detections for rainy and dry periods, respectively, and during the pre/post-Buran storm, when seven and two events were detected on 2 March during the rainy and dry periods, respectively. This distribution is clearly visible in the synoptic graph shown in Fig. 5a, which shows the maximum PGA recorded from the six sensors for each MS event vs. the rainfall cumulative and the temperatures recorded in the air and the rock, where the MS events that had the highest PGA mainly occurred during the Buran storm period. This graph also reports the non-selected events, which were recorded simultaneously by 5 sensors and were, therefore, considered to be non-random seismic events. By focusing on the period characterised by the coldest temperature, which ranged from 27 February to 3 March, it was possible to distinguish three rising temperature trends: while no MS data were available during the first period, in concomitance with the second one and particularly with the last one, a marked increase in micro-seismicity was registered (Fig. 6). The spectrograms computed for the two exemplary MS events showed frequency contents of approximately 140 and 65 Hz. Generally, the frequency maximum of the signals did not show peculiar features linked to the position of the sensor, resulting in random scattering among the accelerometers. In addition to the micro-seismic data, the extensometers and the strain gages only recorded the elastic deformation caused by variations in the rock temperature (i.e., heating dilation and cooling contraction), with a maximum daily amplitude of 165 μ strain (Fig. 5b).

5 Discussion

The automated STA/LTA routine allowed an initial analysis of the distribution of detected events across the monitoring period: 18,802 events, registered to at least two sensors, were scattered throughout the entire period but contained many false positives, which were strongly reduced by considering the 1646 events simultaneously recorded by five accelerometers during the first ten days. This latter dataset was further minimised through manual selection of the waveforms to perform a more restrictive analysis and to remove doubtful MS events, resulting in the identification of 103 MS events that appeared clustered during the Buran storm period and immediately after it (Figs. 5 and 6). Therefore, the reliability of the micro-seismic monitoring system for the entire period was appraised; however, reliable MS events only clustered within a few days while steep thermal transient occurred.

The correlations among rainfall, rock temperature, and time distribution of MS events were investigated. With respect to rainfall, the daily cumulative data showed that the monitoring period was characterised by several rainy days, with a maximum of 23.2 mm on 6 March during the pre/post-Buran time window and 6.6 mm on 27 February during the Buran storm phase (Fig. 7c). Nonetheless, with respect to the 59 MS events that occurred during rainy periods (2-h cumulative), 39 occurred during the Buran storm period and 20 in the pre/post-Buran storm phase (8 before 26 February and 12 after 2 March) but were
Fig. 6 Distribution of MS events during the Buran storm (see the time window reported in Fig. 5): a major cluster corresponds to the highest rock mass temperature rate and to the absolute minimum recorded in rock. The time history and the spectrogram from short-time Fourier transformations are also reported for events A and B identified in the upper panel.
scarcely correlated with the rainiest days. No events were identified after 4 March, while 20 events occurred on 28 February, and only 6 occurred on the previous day, which was characterised by a larger amount of rainfall. A major cluster of events occurred in concomitance with a small amount of cumulated rain (< 1 mm), particularly during the Buran storm phase, while few other events arose for larger accumulations (Fig. 7a). Therefore, a scarce correlation appears between rainfall and MS events across the period, as shown by the cross-correlation that reported unreliable values for the lag for both the entire period and the pre/post-Buran phase, while a lag of 20 h resulted for the Buran storm period. This lag appeared to be very extensive compared to the 3 h lag obtained by Helmstetter and Garambois (2010); thus, other forcings might have acted during this period.

It was determined that the rainy periods that occurred during the Buran storm should not be taken into account for two primary reasons: (1) this period was characterised by snowfall rather than rainfall, and (2) the pre/post-Buran period enables the consideration of rainfall in almost entirely stationary conditions of rock mass temperature. These hypotheses are confirmed by the maintenance of the micro-seismic array that was carried out on 27 February (Fig. 2) during which the quarry was covered by snow, and the rain gage consequently measured snow melt on the considered day. Based on this evidence, a correlation between rainfall and MS during the Buran phase likely does not exist, while the occurrence of the described sharp thermal transient played a key role in triggering the micro-seismicity that occurred during this time interval.

As a further test of the reliability of this hypothesis, among the MS events that occurred in the absence of rainfalls (i.e., dry periods), a high proportion was recorded during the Buran storm (40 MS events), while significantly fewer occurred during the pre/post-Buran storm periods (4 MS events). To correlate the occurrences of MS events to the rock mass temperature, excluding possible interferences due to rainfalls, micro-seismicity was analysed in connection to the thermal gradient derived in 2 h intervals within the dry periods both during and pre/post-Buran storm (Fig. 7b). This approach highlighted the fact that the MS events mainly occurred in response to positive thermal gradients, particularly focused at the end of the Buran storm, when the temperature rose above 0 °C (Fig. 6). By referring to the absolute rock temperature, the Buran storm was characterised by MS events that
mainly occurred when the temperature was higher than 3 °C and ranged between −1 and 1 °C, while only a few coherent events occurred in the pre/post-Buran storm. Moreover, by considering the recorded rock mass temperatures, it is worth noting the absence of a minimum peak in the time window ranging from 1 to 3 March 2018, implying a rising temperature trend for a longer period, during which most of the MS events occurred (Figs. 5 and 6). This evidence demonstrated that while the rock mass temperature always fluctuates according to air temperature with an associated variation of the elastic strain recorded by the extensometer and the strain gages, the MS data were focused only along the heating ramps after the Buran storm (Figs. 5 and 6). Given the sudden temperature decrease, it is possible to assume that the snowfalls were associated with the favoured cooling and subsequent freezing of the water-permeated joints, at least in the external thermal active layer (i.e., the first 8 cm from the surface; Fiorucci et al. 2018; Marmoni et al. 2020), as demonstrated by the negative temperature values recorded by the thermometer within the rock mass. However, no instrumental constraints on the wetness conditions of joints are available to date. These findings are in agreement with the maximum rate of daily events, which fell on 1–2 March, during which 18 MS events were detected throughout a dry period. Furthermore, based on the maximum daily value of the 2 h rate of rock temperature (Fig. 7c), a marked temperature variation similar to that experienced during the Buran storm period occurred between 7 and 10 March. However, even short lacks in power supply did occur, only 1 MS event characterised this period. This evidence is supported by the cross-correlation performed between MS data and the rock temperature gradient. While no correlation existed in the pre/post-Buran period, the highest similarity was found at zero lag during both the Buran phase and the entire period.

Based on the observations concerning the cluster of MS events after the thermal minimum phase, supported by screening the occurrences of the events with regard to the rock temperature rate, the following outputs were inferred: (1) the MS events were mainly related to the rock mass response to a transient thermal forcing that likely froze the rock mass with a consequent release of the accumulated inelastic strain; (2) a thermal threshold controlled the thermal stresses acting in the rock mass, as the micro-seismicity occurred only in concomitance with the lowest recorded temperatures (during the Buran storm) and no relevant MS events were detected in the following period. In fact, more intense MS events occurred after the sharp change in air and rock mass temperatures that fell below

Fig. 8 MS events distribution, rock temperature 2 h rate and one-day moving average
0 °C; on the contrary, no MS events were detected in the following days (8–10 March), when the temperature values increased to the seasonal average values. This trend is visible in Fig. 8, where the one-day moving average of the rock temperature, obtained to flatten the daily temperature oscillations to better observing its long-term variations, showed a concavity (i.e., colder temperature) during the Buran storm period, while this trend was not shown from 8 to 10 March.

Contrary to Amitrano et al. (2010) and Occhiena et al. (2012), the MS events were mainly distributed throughout the heating phase of the rock mass rather than during the cooling phase (Fig. 6), associating these events with the thermal expansion of the rock mass. Such differing results could be identified as the response of a complex system composed of intact rock and joints, where the strain accommodation does not necessarily reproduce the theoretical scheme of dilations and contractions assumed for an intact rock as well as a poorly jointed rock mass. Unfortunately, the micro-seismic dataset was not completed prior to the Buran storm occurrence, as the 6-accelerometers micro-seismic monitoring system was installed only a few days before the meteorological event; therefore, the reported considerations are mainly based on the MS data recorded after the Buran storm.

Contrary to Helmstetter and Garambois (2010) and Arosio et al. (2018), this study showed a scarce correlation between MS data and rainfalls, as reported by Colombero et al. (2018). The available recorded data allowed the observation of an intensification of the MS events due to the temperature drop experienced by the rock mass during the Buran storm. In agreement with Frayssines and Hantz (2006), ice jacking could be explored as the main factor responsible for MS events. Ice formation across the rock mass joints likely led to the accumulation of inelastic strain that was consequently released as micro-seismicity when the ice started melting.

Only elastic strains were recorded in the monitored joints, due to the punctual information provided by the installed sensors; however, the registered elastic cycles can be regarded as a reference for the strain suffered by the entire rock wall. Even if the deployed accelerometers covered only a limited sector of the rock face, a similar vibrational behaviour may have affected other rock mass portions, such as that where the rock fall originated. This assumption can be supported by the close distance and common exposure to thermal forcing of the instrumented block and the rock fall sector (Fig. 3). It is worth noting that no rock falls occurred since the installation of the AiCP device, and no similar low absolute temperature or cooling/heating rates were recorded at the Field Laboratory; therefore, supplementary analyses were performed to study the possible relationship between the occurrence of small-size rock fall events and the thermal transience experienced at the Field Laboratory. The instability phenomenon was likely triggered by high-angle and conjugate joint sets (evident from the pre-failure model of this sector – Fig. 4a) partially back filled by the brownish soil also observed in the deposit (Fig. 4b). The comparison of the 3D models of the quarry sector, realised before and after the slope failure, enabled the determination of the source area, while the frames acquired by the AiCP device permitted the detection of accumulation and constrained its occurrence between 11 and 12 March, approximately 10 days after the end of the Buran storm.

Although no micro-seismicity was recorded before the collapse, this time lag seems to be compatible with a freeze–thaw driven failure, in agreement with similar cases reported in the literature. In fact, Levy et al. (2010) studied the pre-failure behaviour of the Chamousset rock column, which collapsed two weeks after a decrease in air temperature that favoured freeze–thaw cycles. Frayssines and Hantz (2006) suggested that for limestone cliffs, such as those present in the Chamousset and the Acuto sites, frost can penetrate faster and deeper into opened joints than in a more contiguous rock mass due
to the efficient heat transfer associated with cold airflow in the opened joints. Moreover, the water seeping during the thaw phase may have contributed to decreasing the available rock mass strength in the sealed joints. Even if it is not possible to directly relate the rock fall to a specific triggering event, the above data demonstrate that rock mass damage occurs as a preparatory condition for rock failures and is related to the freeze–thaw effect linked to the short thermal transient associated with the Buran storm.

The here obtained findings highlights the relevance of preparatory conditions in the occurrence of rock failures, as a result of environmental forcings to highlight the jointed contribution of predisposing, preparatory, and triggering factors in causing rock failures and to individuate the precursory events of collapses relevant for developing a hazard mitigation strategy. The rock failure monitoring and event-to-trigger correlation primarily contribute to reducing vulnerability and the associated damage, with the potential to outline a structural resilience strategy.

6 Conclusions

This study presented and discussed the records of a multi-sensor monitoring system, installed at the Acuto quarry Field Laboratory (Central Italy), collected during the Buran storm, a transient weather event characterised by a sharp change in the seasonal temperature trend. According to rock temperature data, the Buran storm and the pre/post-Buran storm periods were distinguished. A micro-seismic monitoring system was installed to detect the vibrational emissions from a rock block partially detached from the rock wall: MS events over and through the 18-days monitoring period were compared with thermal and rainfall data to infer the role of intense thermal and pluviometric transient forcings in the vibrational and deformational responses of the monitored rock block. Many low-intensity MS events were recorded, but only 103 MS events manually selected from those detected by at least 5 sensors were included in the analysis: 44 events occurred during dry periods and 59 events occurred during rainy periods. No significant correlation between the MS events and rainfall events was found, while a major cluster of events corresponded to the periods that were characterised by a thermal minimum. In fact, the MS events detected during the Buran storm were of one order of magnitude higher than these that occurred in the pre/post-Buran storm, with 18 MS events occurring from 1 to 2 March during a rising temperature after rock joint freezing. These findings prove the key role of thermal factors in stressing the rock mass and confirm the importance of integrating seismic monitoring techniques with contact stress–strain techniques to study the deformative effects acting on rock block and rock mass scales by different environmental forcings.

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Author contributions D.D., L.L. and M.F. processed and analysed the seismometric data. G.M.M. and M.F. processed the thermo-pluviometric and strain data. D.D., M.F., R.I., L.L., G.M., and S.M. discussed and provided interpretation of seismometric and environmental datasets. A.F. managed the AiCP and performed the photogrammetric survey. S.M. coordinated the research activity and planned the experimental phases. All the authors wrote the manuscript and edited figures and tables.

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