

1 **Early warning signs for mining accidents: detecting crackling noise**

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6 **Abstract:** Acoustic emission signals carry much of the information about collapsing
7 cavities and progressing cracks in minerals. The acoustic noise signals are constituted by
8 individual events, so-called ‘jerks’, which form a spectrum of ‘crackling noise’ with well-
9 defined characteristics. The close connection between collapsing minerals under uniaxial
10 stress and the spectra of crackling noise has been systematically investigated over the last
11 5 years on small samples but it is only a new paper in *American Mineralogist* that Jiang
12 et al. (2017) have made a major breakthrough by using much larger samples and higher
13 stresses. They show that the crackling noise changes its spectral parameters from the non-
14 critical steady state to the all-important precursor regime before samples finally collapse.
15 They showed that the noise distribution always follows a power law but that the
16 exponents are different in the two regimes. Their work on coal and sandstone have hence
17 shown that early warning signs can indeed be extracted from acoustic signals and thereby
18 offer a new method to prevent mining accidents. **Keywords:** mining accidents, crackling
19 noise, acoustic emission, collapse predictability

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21 The prediction of mining accidents is one of the great outstanding problems for

22 mineralogists. Accidents in coal and hard rock mining are particular severe with an
23 alarming number of miners dying each year. Most of the deaths occur in developing
24 countries, in China, and in rural parts of developed countries. Historically, one of the
25 worst mining accidents occurred April 26, 1942 in the Chinese Benxihu (Honkeiko)
26 Colliery (coal mine), killing 1,549 miners. Several of these accidents are related to the
27 collapse of mine shafts. Such collapses are not fully preventable when large tectonic
28 movements occur but the death toll can massively be reduced if the collapse can be
29 predicted. A key research area for mineralogists is to identify any precursor events,
30 therefore, which can identify markers to predict the collapse event over relatively short
31 time spans.

32 This problem is closely related to the question of whether earth quakes have an early
33 warning sign, such as biological activities of mechanical pre-tremors. Systematic in-situ
34 investigations of these issues are difficult, however, because the events are mercifully
35 rare and few data will be collected during or after a disaster strikes. In an alternative
36 approach, Salje and his collaborators Jiang et al. (2017) transferred the investigation of
37 collapse events into the laboratory where ‘lab-quakes’ replace the larger geological
38 events (Baró et al. 2013).

39 This approach started in 2011 (Salje et al. 2011a) when Salje, together with his
40 collaborators in Barcelona investigated the collapse of porous SiO₂ materials. These
41 authors understood very quickly the close similarity between the collapse of mines and
42 earth quakes with the crackling noise emitted from the sudden collapse of a porous
43 cylinder under uniaxial stress. Similar work emerged from Ian Main’s group in
44 Edinburgh (Bell et al. 2011) and others by doing model simulations (Kun et al. 2014).

45 The first aim by the Salje team was then to identify which minerals or mineral assemblies
46 where prone to violent collapse. Furthermore, it was not clear what degree of porosity
47 would lead to precursor effects and which would not.

48 In a first series of investigations, small samples of SiO₂ (the ceramic ‘vycor’) (Salje et al.
49 2011a, Baro et al. 2013), goethite (Salje et al. 2013), corundum (Castillo-Villa et al.
50 2013), sandstone (Nataf et al. 2014a), and berlinite (Nataf et al. 2014b) were studied and
51 some universal behaviour was found: macroscopic observations show that the
52 compression of porous materials leads to sudden partial collapses rather than a smooth
53 elastic or plastic deformation. The partial collapse events are named ‘jerks’ and, within
54 the context of out-of-equilibrium dynamics, have been classified as avalanche
55 phenomena. Their statistical features are similar to crackling noise as reviewed by Sethna
56 et al. (2001). Jerks were also seen not only under compression, but also under shear
57 deformations where the microstructures of the sample change in sudden movements
58 rather than continuously.

59 One of the experimental techniques that had been most successful in yielding an
60 understanding of the statistical properties of jerks was the detection of the acoustic
61 emission (AE) associated with the microcracks occurring in the samples. The
62 compression of natural and synthetic porous mineral with 40% porosity have shown that
63 the event energies and the times of occurrence are virtually indistinguishable from those
64 of earthquake statistics, both concerning temporal correlations (aftershocks) and energy
65 distributions. In particular, the distribution of event energies shows a Gutenberg–Richter
66 behaviour, i.e. a lack of characteristic scales: the probability of a jerk with energy E
67 follows $P(E) \sim E^{-\epsilon}$ with $\epsilon = 1.40 \pm 0.05$ for SiO₂ related materials. Larger exponents were

68 found in corundum (up to 1.8) but the overall situation remained the same: in all samples
69 one observed the power law distribution of jerks and correlations between the events, in
70 particular near the end of the sequence when the final collapse was imminent. Salje and
71 Dahmen (2014) provided a full theoretical description of the jerk statistics in disordered
72 materials. The details of the acoustic emission mechanism are related to changes of
73 microstructures and were derived by Salje et al. (2014) based on a simple theoretical
74 model (Salje et al. 2011b).

75 These results pointed towards the possibility that correlation between the largest
76 avalanches and corresponding foreshocks may occur in the case of high porosity (>80%).
77 The experimental observations also lead to the first indication that large-scale major
78 collapses could be predicted by the detection of acoustic signals in the case of geological
79 materials constituted by crystalline (or amorphous) bulk phases with a highly porous
80 microstructure. On the other hand, in case of samples with lower porosity, no such
81 correlations were seen and avalanche prediction was deemed virtually impossible.

82 While these studies set out the pathway for collapse prediction, the breakthrough came
83 after the technology to measure collapse jerks was scaled up to more realistic sample
84 sizes. In the recent study by Jiang et al. (2017) these authors used a large scale hydraulic
85 press with cylindrical samples sizes with 50 mm diameter and 100 mm length. The stress
86 rate was as low as $d\sigma/dt = 8.5 \text{ kPa/s}$ (1 kN/min) with a maximum stress of 300 kN. The
87 experimental arrangement is such that even bigger samples under slower stress rates can
88 be investigated so that the way is now open to test ‘realistic’ collapses while measuring
89 the acoustic emission in the laboratory. Samples of coal and sandstone were measured

90 and a very clear signal for the precursor effect was found. The large sample volume leads
91 to a new phenomenon that was not possible to observe in small samples, which is the
92 evolution of the event centers during the experiment. The results indicate that all
93 avalanche characteristics can be divided into two groups, namely those at the early stages
94 and those closer to the main failure event. Using eight AE detectors the authors found
95 that the initial stages are characterized by randomly distributed event centers with large
96 distances between them. The events are hence approximately randomly distributed in
97 time and space. With increasing number of emission centres, clustering along the future
98 collapse planes occur and the events become highly correlated. The energy exponents for
99 the randomly distributed events are generally larger than those of the correlated events
100 (uncorrelated 1.7 versus correlated 1.5 for sandstone and uncorrelated 1.5 versus
101 correlated 1.3 for coal). The power law dependence is emphasized by the averaged slope
102 with Omori exponents $p = 0.84$ (sandstone) and $p = 0.95$ (coal). These values are typical
103 for Omori sequences in natural earthquakes. The Båth's law is equally satisfied with the
104 ratio of 1.2 between the energy magnitudes 4 of a main shock and an after shock.

105 The results of the research has two major implications. First, it shows that the prediction
106 of collapse events is possible. The two indicators for an impending disaster, such as a
107 collapse of a mining shaft or the collapse of a building, are the increase of acoustic
108 emission of crackling noise and the change of its energy exponent. Both indicators can be
109 measured easily by highly sensitive microphones attached to the material (sandstone
110 sculptures, houses, coals seams, etc.). The signals need to be processed, which can be
111 easily done using a simple computer device. The sensitivity of the method depends
112 largely on the microstructural properties of the minerals: uniform but porous sandstone is

113 much more sensitive to “early warning noise” than coal seams that contain a multitude of
114 local cracks.

115 Second, we now understand crackling noise much better because a much more detailed
116 measurement technique is available that allows the investigation of very large samples.
117 The analogy of crackling noise of porous minerals with earthquakes is firmly established
118 with the surprising additional feature: the energy exponent (or b-value in seismology)
119 depends on the collectiveness of the collapse event. Local cavity collapse displays higher
120 energy exponents than collective movements such in crack propagation and the formation
121 of microfaults.

122 Once these issues are now clarified it is crucial that further research is undertaken using
123 even larger samples and more acoustic detectors to clarify the collapse characteristics of
124 minerals. A procedure could be to extract minerals or geological samples from mines and
125 subject them to the same testing procedures. The statistical properties of the collapse such
126 as the change of the power law exponent and possible changes in the Oromi and the
127 Båth’s law can then be registered. Then detectors can be introduced into the mines and
128 permanently computer controlled. Whenever changes of the ‘crackling noise’ occur, it
129 can be compared with the collapse characteristics and, if appropriate, warnings of
130 impending collapse can be given.

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