SEISMICITY INDUCED BY MINING: TEN YEARS LATER

SLAWOMIR J. GIBOWICZ
Institute of Geophysics, Polish Academy of Sciences
ul. Ks. Janusza 64, 01-452 Warsaw, Poland

STANISLAW LASOCKI
Department of Geophysics, University of Mining and Metallurgy
al. Mickiewicza 30, 30-059 Krakow, Poland

1. INTRODUCTION

In the previous review of seismicity induced by mining (Gibowicz, 1990), the general description of the problems involved and the state-of-the-art of relevant research in this field at the end of 1980s was given. During the last decade, seismic monitoring has been expanded in several mining districts, a number of new techniques have been introduced, and new significant results have been obtained in studies of seismic events induced by mining. Dr. Renata Dmowska, coeditor of Advances in Geophysics, invited us therefore to contribute a new review, describing in some detail the latest achievements in the field.

A distinction between the adjectives induced and triggered, often used interchangeably to describe artificially stimulated seismicity, should be made as proposed by McGarr and Simpson (1997). They consider that "... 'induced seismicity' is that for which the causative activity can account for either most of the stress change or most of the energy required in order to produce the earthquakes. In contrast, 'triggered seismicity' is used if the causative activity accounts for only small fraction of the stress change or energy associated with the earthquakes..." In the first case, human activity and in the second case, tectonic loading play the primary roles. Thus, seismic events observed in mines are induced events, whereas seismic events associated with the impoundment of reservoirs are triggered events.

Another distinction, that between a seismic event in a mine and a rockburst, should be clarified as well. Earthquake seismologists tend to describe as rockbursts any seismic events which occur in a given mine. For mining engineers, however, rockbursts are violent failures of rock that result in visible damage to excavations. This distinction was also emphasized in the previous review (Gibowicz, 1990) as it is highly important in mining practice.

The first two International Symposia on Rockbursts and Seismicity in Mines, held in Johannesburg, South Africa, in 1982 and in Minneapolis, Minnesota, in
1988, were followed in the 1990s by the 3rd Symposium held in Kingston, Ontario, in 1993 and by the 4th Symposium held in Krakow, Poland, in 1997. The proceedings of these last two symposia (Young, 1993; Gibowicz and Lasocki, 1997) contain 12 keynote lectures and some 130 technical papers on the subject, written by experts from some 15 countries.

In the 1990s a number of books and special issues of well-known professional journals related to seismicity in mines have been published. The first ever textbook on this subject, *An Introduction to Mining Seismology*, was published in 1994 (Gibowicz and Kijko, 1994). The second book, *Seismic Monitoring in Mines*, written by Mendecki and his colleagues from the ISS International, South Africa, (Mendecki, 1997a) contains a description of a very specific methodology for assessing in real time the seismic hazard associated with mining. The other two books, *Induced Seismicity* (Knoll, 1992) and *Tectonophysics of Mining Areas* (Idziak, 1996a), contain a collection of papers related to the subjects specified by their titles. Five Special Issues of *Pure and Applied Geophysics* devoted to stimulated seismicity phenomena have been published between 1992 and 1998 and reprinted as separate books (McGarr, 1992a; Gupta and Chadha, 1995; Knoll and Kowalle, 1996; Talebi, 1997, 1998), and one Special Issue of *Tectonophysics* was published in 1998 (Trifu and Fehler, 1998).

This review is organized to some extent similarly to the previous one. New techniques in seismic monitoring in mines, geological and mining factors affecting seismicity, source parameters and their scaling relations, and shearing versus non-shearing source mechanisms are briefly described. Statistical techniques and methods, however, used extensively in recent years in studies of seismicity in mines, especially for seismic hazard assessment, have not been previously reported. They are discussed in some detail now. The other topic that has only recently become of considerable importance is the seismic discrimination between underground explosions and seismic events originating in deep mines. The problem is briefly discussed at the end of this review.

2. SEISMIC MONITORING

Seismicity induced by mining is commonly described as the occurrence of seismic events caused by rock failures. This is a result of changes in the stress field in the rock mass near mining excavations. The total state of stress around a mine excavation is the sum of the ambient stress in the rock mass and the stresses induced by mining works. The vertical component of ambient stress is due to the overburden, whereas applied tectonic forces can affect the horizontal ambient stresses. Mine tremors do not necessarily occur in all mining situations at a given depth where the lithostatic stress is of considerable value. The primary requirement for inducing seismicity appears to be human activity where the rocks are in a highly prestressed condition.
Grasso and Sornette (1998) examined the hypothesis, proposed in recent years (Bak and Tang, 1989), that the earth’s crust is in a self-organized critical (SOC) state to explain and understand the phenomenon of induced seismicity. They found that both pore pressure changes and mass transfers leading to incremental deviatoric stresses smaller than 1 MPa are sufficient to trigger seismic instabilities in the uppermost part of the crust in otherwise historically aseismic areas. Once triggered, stress variations of at least one order of magnitude less are enough to sustain seismicity. They argue that these observations are in agreement with the SOC hypothesis, as they show that a significant fraction of the crust is not far from instability. Although not all stress perturbations trigger seismicity, this fact is still compatible with the SOC hypothesis, which includes the presence of large heterogeneity in the stress field.

Sykes et al. (1999) disputed the claim that earthquakes represent a SOC phenomenon, however, implying a system maintained under near-failure conditions at all times; this is correct on global but not on regional scales. The critical point model for regional seismicity suggests that only during the periods of long wavelength correlations in the regional stress field, emphasized by increased seismic moment/energy release in moderate earthquakes, is a region of the earth’s crust truly in or near a SOC state (Jaumé and Sykes, 1999). The question of continuous versus discontinuous criticality is directly related to the question of earthquake predictions (e.g., Sammis and Smith, 1999).

McGarr et al. (1999) reviewed case histories of induced and triggered seismicity to illustrate a variety of causes, scales, and crustal environments. Mining-induced events are caused by increases in the shear stress or decreases in the normal stress acting on the fault planes. Pumping in mines to prevent flooding reduces the crustal pore pressure and strengthens the rock mass. In mines, therefore, seismic events are induced only in those regions where the ambient stress has been modified substantially by the mine excavations. The stress perturbations, although often quite small, frequently result in generation of seismicity because near-failure conditions are common in both active and stable tectonic environments.

Seismicity in underground mines is observed in numerous mining districts throughout the world. The most comprehensive studies are carried out in South Africa, Poland, Canada, the United States, and the Czech Republic. The literature is extensive and only the more interesting results published since 1990 in accessible journals and books are briefly reported here. References to works published before 1990 can be found elsewhere (Gibowicz, 1990; Johnston, 1992; Gibowicz and Kijko, 1994).

2.1. Seismicity in Underground Mines

Extensive studies of seismicity, mostly in deep gold mines, have been carried out for a long time in South Africa. The gold-bearing reefs of the Witwatersrand
system are mined by stoping at depths down to 3.5 km below the surface. This creates flat voids in the quartzitic strata extending horizontally up to several kilometers with an initial excavated thickness of a meter. Developments and methods used in seismicity studies have been reported in several publications in the 1990s. Seismicity associated with pillars at Western Deep Levels gold mine (Lenhardt, 1990; Leach and Lenhardt, 1990); seismic wave attenuation in deep-level gold and platinum mines (Spottiswoode, 1993); application of source parameters to detect changes in rock mass behavior in the Welkom gold mines (van Aswegen and Butler, 1993); and practical use of the instability concept of Mendekcki (1997b) in the Klerksdorp mining area (Glazer, 1997), are some of the topics recently studied in South Africa. Additional topics will be discussed in the following chapters. The seismic monitoring technique has been also used to study the caving processes above a longwall in a 160-m-deep coal seam within the Highveld coalfield in the province of Mpumalanga, South Africa (Minney et al., 1997).

Seismicity induced by underground mining is a well-known long-studied phenomenon in Poland, observed in the Upper Silesian Coal Basin and in the Lubin copper district in Lower Silesia. The bimodal character of seismicity in Upper Silesia has been confirmed by numerous studies. The low-energy events are associated directly with mining operations and occur around longwalls and openings. The high-energy events, often called regional events, occur as a rule in geologically disturbed areas and in their generation tectonic stresses play a significant role (e.g., Idziak, 1996b; Mutke and Stec, 1997). The spatial distribution of seismic events in the Upper Silesian Basin is not uniform. Despite mining operations throughout the Basin, the seismicity is concentrated in four distinct areas forming different geological units. The location of two major epicenter clusters and active mines there is shown in Fig. 1, reproduced from Lasocki and Idziak (1998). Nonuniform distributions of this type suggest a fractal character of seismicity in Upper Silesia (e.g., Idziak, 1996b). Similar phenomena are recognized in the Lubin copper district. A detailed study of the source mechanism and source parameters of seismic events at Polkowice copper mine has been completed by Król (1998). Their relationships with geological and mining factors are of special interest.

Seismic events and rockbursts are often observed in metalliferous, coal, and potash mines in Canada. Seismic activity in Canadian mines increased significantly during the 1980s, particularly in northern Ontario hard rock mines operating at depths down to 2 km. In response, the Canada-Ontario-Industry Rockburst Project was initiated in 1985. From 1990 to 1995 an expanded second phase of the Rockburst Project, the Canadian Rockburst Research Program, was in progress (e.g., Plouffe et al., 1993). Significant improvements were made in the past decade with the installation of the Sudbury Local Telemetered Network (SLTN) around the Sudbury mining district in Ontario and seismic stations in five other mining camps of northern Ontario and northwestern Quebec (Talebi et al., 1997). These improvements concerned the determination of source location and event magnitude,
FIG. 1. Location of two major clusters of seismicity in the Upper Silesian Coal Basin, Poland, displayed on the background of active mines. [Lasocki and Idziak (1998), Fig. 2; reprinted with kind permission from Birkhäuser Verlag AG.]
detection capabilities, quality of data, and the procedure for data transfer and analysis. The waveforms of seismic events recorded in a mine in the Sudbury district were processed to study attenuation effects, source parameters, and scaling laws (Talebi et al., 1994). SLTN operations were terminated in the late 1990s.

About half of the deep mines in the Sudbury Basin are monitored continuously by seismic systems. Not all deep mines in the area, however, are seismically active; several display little or no seismicity at all. There are several groups of two mines which share boundaries and which appear to have similar mining configurations and geological structures but display dramatically different levels of seismicity. A comparison of the characteristics of both seismic and aseismic mines could provide a valuable contribution to the study of mining-induced seismicity (Morrison, 1993).

Three-dimensional boundary-element modeling of stress effects, source mechanisms of seismic events, and underground observations have been used to understand mining effects at Ansil mine, located in the Noranda mining camp in the southern part of the Abitibi green-stone belt, and several conclusions have been drawn (McCreary et al., 1993). Noranda’s hard rock Brunswick No. 12 mine, located in Bathurst, New Brunswick, is affected by seismic activity and rockburst-related damage to underground excavations. Since 1994, the mine has employed several systems for seismic monitoring. A rockburst hazard assessment technique has been introduced, based on seismic energy, apparent stress, and seismic moment criteria, to identify those events that are relevant for assessment and decision-making processes (Alcott et al., 1998).

In the U.S.A., the Wasatch Plateau and Book Cliffs coal mining districts of east-central Utah, containing more than 30 underground mines, are notable as one of two areas in the western U.S. where mining-induced seismicity is significant (Wong, 1993). Mining tremors account for about one-fourth of the seismic events annually recorded in the seismically active Utah region (Arabasz et al., 1997). Total mining seismicity in the area sharply increased after 1992, posing a challenge for understanding the complex interaction between geology, mining, and observed seismicity. The largest mining-induced event in the region occurred in the Gentry Mountain area on May 14, 1981 and had an estimated local magnitude of 3.8 (Arabasz et al., 1997). The Coeur d’Alene mining district, located within a zone of minor natural seismicity in northern Idaho, is one of the few metal-mining areas in the United States where large rockbursts occur. Rockbursts have become increasingly common during recent years. During the 1960s and 1970s, rockbursts in the district were attributed to the failure of pillars ahead of advancing mine faces. During the 1980s, seismic activity at the Lucky Friday mine at Mullan increased significantly as mining went deeper, and observations of the in-mine damage suggested that the predominant mechanism of the rockbursts changed from pillar failure to fault slip (Sprenke et al., 1991). Three types of geological structures have been also identified that contribute to seismicity of this mine (Scott et al., 1993).
The highest level of mining-induced seismicity in the Czech Republic is observed in the Ostrava-Karvina Coal Basin. Seismic activity and the number of rockbursts there increased significantly in recent years as a result of mining at greater depth under complex geological conditions (e.g., Konečný, 1992). Observations from local seismic networks in mines and from a regional seismic network (in operation since 1988) are used to study the spatial and temporal distributions of seismic events in several mines and their relationships to mining and geological structures (e.g., Konečný, 1992; Kalenda, 1995; Vesela, 1995; Holub, 1996a). Investigation of seismic events in the Kladno coal district continues (e.g., Rudajev, 1995).

In India, seismic events and rockbursts induced by mining have been known for a long time in the Kolar Gold Fields situated in the Karnaka State. The deepest mining operations are carried out there at depths exceeding 3.2 km and the rockburst problems are almost as old as the mining activity itself (e.g., Behera, 1990; Srinivasan and Shringarputale, 1992). An unusual and probably unique seismic phenomenon is observed in these mines, called “area rockburst” (Jha et al., 1993). The area rockburst is a sequence of rockbursts that follow in quick succession with their sources concentrating in a small area of 100–200 m radius. The occurrence of a major rockburst triggers a chain of 20 to 200 rockbursts of similar or smaller size in the same area over a period of a few hours to a few days. They take place at all depth levels in both currently mined and old working areas. The spatial and temporal distributions of area rockbursts are highly irregular. They are similar to earthquake swarms and follow similar patterns with the number and magnitude of the events increasing gradually with time and then gradually decreasing.

The mining of the Khibiny Massif apatite deposits in the Kola Peninsula in Russia was started in 1929. The first seismic event with magnitude 3.1 was felt in 1948. A more intensive excavation began in the mid-1960s, and the first significant tremor occurred in 1981. At present more than 100 million tons of ore are extracted annually from three underground and three open-pit mines (Kremenetskaya and Tryapitsin, 1995). The strongest tremor with magnitude 4.1–4.3 occurred on April 16, 1989 causing extensive damage at Kirovsk mine (Syrnikov and Tryapitsin, 1990; Kremenetskaya and Tryapitsin, 1995). The maximum measured displacement was 20 cm and was traced along a fault on the surface for 1200 m and observed to a depth of at least 220 m. The main tremor was followed by several hundred aftershocks during the next 2 months. The event occurred simultaneously with a 240-ton explosion in one of the Kirovsk mines, implying that the blast triggered the tremor. Vigorous monitoring of the area is carried out, within a program developed by the Kola Complex of geodynamic measuring stations in the Khibiny and Lovozero rock masses. Monitoring includes acoustic and electromagnetic observations (Melnikov et al., 1996). There are several seismic stations in the area, and a Norwegian local array operates in the vicinity of Apatity town. The ARCESS seismic array in northern Norway records all larger events from the area.
In the northern Ural bauxite mines in Russia rockbursts first occurred at the beginning of 1970s at a depth of 350 m. Seismic observations there have been carried out from 1980. The largest events have magnitude ranging from 3.0 to 3.5 (Kolesov, 1993; Lomakin and Junusov, 1993), and some of them occur at depths distinctly greater than the depth of mining (Voinov et al., 1987). The largest tremors in the Tashtagol iron-ore deposits, Gornaya Shoria, have magnitudes ranging from 3.0 to 3.5 (Lomakin and Junusov, 1993).

The first recorded rockburst in China occurred in 1933 at the SL coal mine when the mining depth was about 200 m. The number of coal mines affected by seismicity increases steadily with increasing mining depth and mining intensity. At present, seismic events occur at 33 coal mines, and over 2000 rockbursts have taken place since 1949, with the largest event of magnitude 4.2 (Wu and Zhang, 1997). Rockbursts have occurred in almost all coal types, in various types of roof, and under all mining conditions. Seismic and acoustic monitoring systems are widely used, and a complex rockburst prevention system is claimed to be successfully applied. In recent years, rockbursts have also occurred in some metalliferous mines.

The Liaoning Province in northeastern China is a highly seismic region where tens of destructive tectonic earthquakes have occurred. During the last three decades, seismic events induced by mining have become quite frequent (Zhong et al., 1997). The seismic activity first appeared at Taiji coal mine, Beipiao County, where in 1970 the excavation of a shaft reached a depth of 500 m, and a series of small tremors occurred. Up to 1989 over 1300 events were recorded by local seismic stations, with the largest event of magnitude 4.3, which took place in April 1977. The observed seismicity is closely related to active faults. Furthermore, a correlation is observed between the number of seismic events in mines and the number of natural earthquakes in the area (Fig. 2). This observation suggests that many seismic events at Taiji coal mine are triggered by natural earthquakes (Zhong et al., 1997). Since the 1970s, a series of tremors have also occurred in other coal mines, such as Xilin, Haizhou, Wulong, and Aiyou. The Binggou coal mine, Jiachi County, has an excavation history exceeding 200 years, and there are many voids left by mining works. A number of seismic events caused by roof collapse have been observed, with the maximum magnitude of 3.9, particularly between 1976 and 1985 in which 56 large collapse events occurred (Zhong et al., 1997).

The Central Queensland University Regional Seismic Network in Australia, established in 1990, records seismic events generated by the rock mass failure and roof falls in longwall coal mines in the Bowen Basin. A relatively high compressive horizontal stress is dominant there, reaching three times the vertical stress value (McKavanagh et al., 1995). The concept of seismic monitoring from the surface of longwall caving characteristics was used at Capcoal’s Southern Colliery, Germain Creek, in late 1990. Over 150 longwall related seismic events were recorded, classified into strata failure events without a subsequent fall and events associated
with falls characterized by complex waveforms. The first detailed seismic monitoring study was undertaken at Gordonstone mine in Central Queensland in 1994 to investigate the extent of ground failure caused by longwall mining (Hatherly et al., 1997; Hatherly and Luo, 1999). Seismic monitoring systems are in use in coal mines under the massive strata at Newstan, West Wallsend, Moonee, and Clarence to monitor the development of dangerous caving situations in longwall mining.

Two experiments were carried out in 1988 and 1991 in the English Midlands at Littleton and Coventry collieries to image fracture development around the working face by monitoring seismic activity using three component down-hole geophones (Toon and Styles, 1993). A borehole seismic monitoring permitted delineation of the spatial and temporal development of the fractures around an active longwall face. A somewhat unusual application of seismic monitoring techniques was performed by the British Geological Survey around Rosslyn Chapel in the Midlothian Coalfield to monitor seismic activity following pit closure (Redmayne and Richards, 1998). Accurate locations were obtained for 247 events that occurred between November 1987 and January 1990. A rapid decay of seismicity following pit closure was demonstrated.

El Teniente Mine is an underground copper mine located in central Chile, which is strongly affected by rockbursts and where seismic activity is of considerable intensity. A digital full waveform system was installed in January 1992 to cover the whole mine. At present, the seismic network is composed of 29 stations with 4.5-Hz triaxial geophones (Dunlop and Gaete, 1997).
is the deepest productive sector at the mine. The ore production there began in 1989 but it was repeatedly stopped by large rockbursts between 1989 and 1992. The production was resumed in January 1994 and, following the observed relationships between mining and seismic parameters, was restricted to an area with a lower column of intact rock and controlled production rates to minimize the size of the ruptures. No rockbursts have been observed since.

In French coal mines, large seismic networks have been implemented to monitor the areas prone to rockbursts; and two of the seismic networks were used to study seismic activity in some detail (Ben Sliman and Revalor, 1992). The influence of wave propagation models and network size and geometry on the location accuracy of seismic events was assessed. At the Provence colliery, coal is mined by a longwall technique at a depth reaching 1100 m and the rate of production is steadily increased (Bigarre et al., 1993). The mine experiences some 20 seismic events daily with magnitude 1.5 and greater. Most of these events are attributed to the longwall operations and several of them cause serious damage at the advancing face and along haulage gateways. In the southern part of the colliery, several major tectonic faults are present and they most probably play a major role in dynamic loading of the coal seam by fault slips induced by mining. Numerical 3D modeling of fractured rock mass has been undertaken to quantify rockburst potential of fault slip along major preexisting geological structures.

Specific studies on seismicity in underground mines are occasionally reported in other countries. The Miike coal mine is an undersea coal mine in the Kyushu province in Japan. Rockbursts have repeatedly occurred there on the longwall face in a limited area in the deepest part of the mine (Kaneko et al., 1992). Seismic and acoustic emission monitoring systems, in situ stress measurements, in situ test drilling, and numerical analyses are used to clarify stress field conditions and to design destressing operations. It was confirmed that large seismic events occur in the area with a low seismic energy release background and high stress concentration.

In May 1993 more than 200 seismic events, with the largest event of magnitude 1.9, from an underground dolomite mine near Schwaz in Tyrol, Austria, were recorded by a local seismic network established in the late 1980s to monitor natural seismicity in Tyrol (Lenhardt and Pascher, 1996). Underground observations revealed a number of discontinuities along which the rock mass was able to move. Seismic records imply two different source mechanisms of events involved in the sequence. One seems to be associated with block sliding along several discontinuities and the other requires collapse as well.

It follows from studies of seismicity induced by underground mining that seismic events do not necessarily occur in all mining situations; their maximum size is different in different mining districts; and their depth is usually close to that of mining excavations.
2.2. Monitoring Systems

The first applications of seismic instruments to monitor seismicity in mines were introduced at the beginning of the twentieth century (see Gibowicz and Kijko, 1994), and routine seismic monitoring has been in use for over 30 years. Its main objective was immediate location of seismic events in mines, estimation of their strength, and prediction of large rock mass instabilities. The first two objectives were readily achieved, but the prediction of large seismic events in mines remains an open question, although the first positive results are reported from South Africa (e.g., Mendecki, 1997b,c; van Aswegen et al., 1997). Furthermore, modern statistical techniques, based on seismic catalogs as principal input data, provide indications on the present and future course of seismicity generation processes in mines (see Sect. 8). The introduction of modern digital seismic systems to mines, and progress in the theory and methods of quantitative seismology, have led to the implementation of real-time monitoring as a mine management tool for estimation of rock mass response to mining.

Seismic monitoring in mines should consist of sensors, data transmission and acquisition, signal processing, and data analysis and interpretation tools, leading to quantification of rock mass seismic response to mining and possibly to estimation of the potential for rock instability (see Mendecki, 1997a). The modern digital seismic system should provide, apart from the event time and location, seismic moment and radiated seismic energy, two independent source parameters. Other source parameters (source radius, average slip, stress drop, apparent stress) can be also estimated from the recorded high quality waveforms.

The transducer is the key element of any seismic monitoring system, transforming the ground motion into an electric signal. The type of transducer to be used depends on the range of amplitudes and frequencies to be recorded which, in turn, depend on the magnitude of seismic events and their distance from the sensor. The corner frequencies, corresponding to the maximum of the ground velocity spectra, define the range of frequencies that must be recorded for meaningful interpretation of seismic sources. For correct determination of seismic moment, frequencies down to an octave below the corner frequency of the largest expected event must be recorded, and to estimate correctly the radiated seismic energy, frequencies five times above the corner frequency of the smallest event are usually needed (Mendecki, 1993). A possible range of corner frequencies of shear waves as a function of seismic moment is shown in Fig. 3 for several values of stress drop and a range of rock types, reproduced from Mendecki (1997c). Ground motion amplitudes at the sensor depend also on the distance from the event and on the rock mass characteristics, and therefore an extremely wide dynamic range in amplitude is needed. For reasons of economy, the sensors used in mine monitoring systems are usually (for example in South African mines) either miniature geophones or
piezo-electric accelerometers (Mendecki, 1997c). For events with corner frequencies below 4 Hz the low-cost sensors are inadequate and 0.7–1.0 Hz seismometers, used for example in Polish mines, or force-balance accelerometers are required. More details about seismic transducers are given by Mountfort and Mendecki (1997a).

The data acquisition system must record the amplitude and timing of ground motion at sensors distributed throughout the selected volume of a mine and must assemble the records at a central site for processing. The main functions of a modern monitoring system and data flow are shown in Fig. 4, reproduced from Mountfort and Mendecki (1997b). This is a good illustration of the Integrated Seismic System (ISS), the most comprehensive automatic system widely used in South African mines and elsewhere, described in detail by Mendecki (1993) and Mountfort and Mendecki (1997b). Its dynamic range is greater than 120 dB with a resolution of 12 bits, and the sampling frequency is 2000 samples per second.

Generally, the records are digitized as close to the sensor as possible to reduce transmission errors. In mines, the results of seismic analysis must be available almost immediately, and various media are used for communication such as optical fibers, copper cables, and radio. Triggering is widely used to reduce the
Fig. 4. Basic functions of a modern seismic monitoring system used in mines. Dashed lines represent control signals, and vertical lines represent points at which data may be transmitted from multiple sites to a central site for common processing. [Mountfort and Mendecki 1997, Fig. 2.1, reprinted for Seismic Monitoring in Mines, © 1997 Chapman & Hall, with kind permission from Kluwer Academic Publishers.]
amount of data at the sensor site before transmission, and then the time must be maintained locally at each sensor. Triggering, validation, and data reduction processes can be used to select and preprocess the records for storage and transmission. The central site should have adequate resources for seismic processing and interpretation and for archival storage of event and waveform data (e.g., Mendecki, 1997c).

Various seismic monitoring systems are in use in different countries. In Canada, the MP250 system designed for automatic event detection and source location was used at the beginning of the 1990s by about 20 mines across the country. The system does not record waveforms; it records first arrival times only. A computer code, an automatic data analysis and source location system was developed by CANMET for a rapid and comprehensive analysis of MP250 data (Ge and Mottahed, 1993). The analysis consists of two parts: arrival time difference and residual analyses. The system has been adopted in two major Ontario mines, the Creighton mine at Sudbury and the Kidd Creek mine at Timmins, and is in operation in some mines in Quebec as well. Similarly, an MP250-based monitoring system, which uses simple voltage threshold picking of first wave arrivals, has been used at the Lucky Friday mine in northern Idaho since 1973 to provide source locations and energy estimates of seismic events (Dodge and Sprenke, 1992). A whole-waveform seismic monitoring system based on a computer has been developed to provide more accurate source locations and information on source characteristics. Comparison of source locations produced by the MP250 system to those produced by the whole-waveform system shows that significant timing errors are common in the MP250 system (Dodge and Sprenke, 1992).

Advances in signal processing technology have led to the development of high performance, low cost, whole-waveform seismic monitoring systems, such as the Queen's Microseismic System (QMS) designed at Queen's University, Kingston, Ontario (e.g., Young et al., 1992), replaced later by the Engineering Seismology Group (ESG) system from Kingston. Different versions of these systems have been installed in a number of Canadian underground mines as either an enhancement to an MP250 automatic source location system or as a stand-alone rock mass monitoring system.

A new digital seismic monitoring system was designed for Polish coal and copper mines and was tested at Wujek coal mine in Upper Silesia (Koza, 1997). The system transmits data in a continuous mode, the dynamic range is 90 dB (120 dB in a modified version), the sampling frequency is 500 Hz, and the maximum length of a communication line is 12 km. Transmission errors are automatically detected, centralized synchronization timing is available, and multiple signals can be multiplexed into a single channel. A modified version of the system was adapted for strain measurements at Rudna copper mine. The sampling frequency of 250 Hz is sufficient to separate dynamic and static strain signals.
2.3. Location of Seismic Events

Location of seismic events is the first step in studies and monitoring of seismicity in mines. Accuracy is one of the most important factors in such studies, and the requirements in this respect are demanding. In mining practice, the expected accuracy for hypocenter locations are a few tens of meters, or even a few meters in some cases. The location error might be considered to consist of two components: random location scatter and systematic bias (e.g., Gibowicz and Kijko, 1994). Random location scatter is caused by errors in arrival-time measurements, and systematic bias is generated by the differences between the rock mass structure at the source and receiver and the velocity model used in the location procedure.

The influence of systematic errors can be eliminated by a detailed analysis of travel-time anomalies, or by calibration explosions, or by the simultaneous location of a group of seismic events and velocity model determination. The values of random errors of hypocenter parameters can therefore be used as a quality criterion of the spatial distribution of seismic stations. Consequently, the problem of optimum event location is equivalent to an analysis of the spatial distribution of seismic stations ensuring the minimum values of random errors in the source location procedure (e.g., Gibowicz and Kijko, 1994; Kijko and Sciocatti, 1995a).

Although underground seismic networks are widely used in mines, surprisingly little attention is paid to the optimum design of such networks. Their performance is usually evaluated after installation and operation for a time, and the optimal configuration is predesigned and theoretically tested in a few cases only. A number of papers are available that are devoted to the subject of optimal network planning and event location techniques; see Gibowicz and Kijko (1994), Kijko and Sciocatti (1995a), and Mendecki and Sciocatti (1997) for comprehensive reviews and references. Two methods frequently used to design the optimum configuration of seismic networks in mines are based on either a statistical or a direct approach (Mendecki and Sciocatti, 1997).

With the statistical approach, the choice of one or another station configuration should depend upon the expected location error that would be found if a given configuration is used. Thus the choice should depend upon a certain value related to the given geometrical configuration, and the best network would be defined by its minimum. The value should depend upon a covariance matrix of the event unknown location parameters (Kijko and Sciocatti, 1995a). The covariance matrix can be graphically interpreted in terms of a confidence ellipsoid, where its eigenvalues form the lengths of the principal axes of the ellipsoid. Finding a configuration for which the volume of this ellipsoid is the smallest is referred to as D-optimum planning. The volume is proportional to the product of the eigenvalues, or equivalently, to the determinant of the covariance matrix. The D-optimization of the coordinates of seismic stations becomes much more complex when the travel times of seismic
waves are described by nonlinear equations and the vector containing the event coordinates is of random variable character. The location of seismic events, based on the travel times of seismic waves recorded at several stations, is of such a nature. The described approach to optimal planning of a seismic network can be used in practice only when a relatively small area of high seismicity can be defined. Otherwise, when a seismic area is large and divided into several subregions, the $D$-planning procedure must be modified (for details and applications, see Kijko and Sciocatti, 1995a).

In a direct approach to the optimal planning of seismic networks, the sensitivity of the solution of a system of nonlinear or linear equations to small changes in the data is measured by the condition number of the system. The optimal configuration of seismic stations is defined as that having the smallest condition number of the system of nonlinear equations (Mendecki and Sciocatti, 1997). This procedure is also known as C-optimal planning. For a small condition number, small changes in the errors of arrival times of seismic waves and velocity model should not considerably influence the location of seismic events. Otherwise, the location would be unstable and the condition number would be large. Minimization of the condition number can be taken as a deterministic criterion to design any experiment leading to solving a system of equations. The procedure is described in Mendecki and Sciocatti (1997), where examples of its practical application are also given.

Various approaches to the location of seismic events in mines are based as a rule on time residuals of seismic waves. For a given velocity model a time residual is defined as the difference between the observed and calculated arrival time of a given seismic wave. To constrain the location determination, the direction or the azimuth of the wave-front can be added; this is possible only if the straight line is a fair approximation of the ray path between the source and the station (Mendecki and Sciocatti, 1997).

Two conceptually different approaches to the location problem, based on time residuals, are used (Gibowicz and Kijko, 1994). In the first approach, no attempt is made to distinguish the time residuals caused by time reading errors and velocity model inaccuracy. Typical representatives of such an approach are the classic least-squares procedures and the Bayesian procedure. In the second approach, the time residuals are split into two components: random ones related to the reading errors of arrival times, and travel-time residuals caused by insufficient knowledge of the velocity structure.

Several location procedures are in use, depending on the assumptions made on the nature of velocity model uncertainty. The extension of the classic least-squares procedure to that with controllable variables (the velocity model parameters) subject to random errors is one of them. This approach is possible if the fluctuations of velocity model parameters, from an average model, can be expressed in statistical terms. In the mining application, knowledge of the rock mass seismic velocities and their variances is usually adequate for this technique. Another procedure
employs the concept of relative location of seismic events and is based on arrival-time differences between a reference (master) event and nearby events. The travel-time residuals caused by insufficient knowledge of the velocity model are of a systematic character and are removed implicitly. The method is commonly used in mines where large production blasts are fired during excavations. The most difficult location technique is the procedure of simultaneous inversion of velocity structure and location of a set of seismic events.

The classic way to solve a system of nonlinear equations, used in the location procedure, is to linearize the system by a Taylor expansion around a trial position and solve the resultant linear system, usually by a least-squares method. This means that the $L_2$ norm of the residual vector is minimized, which in turn requires an assumption that the distribution of residuals is of Gaussian nature. The time residuals do not follow the Gaussian distribution and occasionally large residuals strongly affect the location results. Then the $L_1$ norm is preferable: it minimizes the sum of the absolute values of time residuals and is less sensitive to outliers.

In general, the solution of nonlinear station equations can be found by an iterative approach; the Nelder–Mead simplex minimization procedure is often used. To find the global minimum, appropriately chosen starting points for the iterative procedure are needed. The weighting factors are also necessary, especially when automatic processing is performed, to quantify the measure of reliability of the arrival times. Centering and scaling techniques, standard methods used in statistical applications, lead to the simplification of inverse problems and to the improvement of numerical stability. For the location of seismic events in mines, the $L_1$ norm and centering were introduced by Prugger and Gendzwill (1988) and used at Saskatchewan potash mines, Canada; they were also implemented in the location procedures used in South African gold mines (Mendecki, 1993). Other nonlinear techniques which can be used to solve the location problems include genetic algorithms and simulated annealing (e.g., Sambridge and Gallagher, 1993).

Several field experiments have been conducted, each using four source location methods, to estimate the coordinates of a compact explosive source in an underground mine in Australia (Blair, 1993). The four location methods were (1) a least-squares solution, (2) an iterative improvement to the least-squares solution, and (3) a 5-point and (4) an 8-point simplex optimization scheme. The results showed that there was little difference between methods (2) and (4), and method (1) was inferior to all other methods. Synthetic models provided results consistent with the experimental data, provided that a small amount of travel-time scatter was assumed. When the scatter was increased, however, method (1) was found to be the most robust, implying that in real mining environments this method might well be the superior one for the location of seismic events under specific mining conditions.

In some cases, the location procedure can be significantly simplified. This can be achieved if a single velocity model is acceptable. After simple algebraic
transformations, the location parameters are least-square solutions of the set of linear equations. Linear methods used for event locations are very attractive. They are fast and free from all the problems related to iterative procedures. An obvious limitation to these simple methods follows from the fact that they ignore complex velocity structures and their applications may often lead to unreasonable solutions. At present, the linear methods are usually used for determination of a starting point for further iterative computation.

A constant isotropic velocity model was accepted for the location of seismic events at hard rock Galena mine at Wallace in northern Idaho (Swanson et al., 1992). Dense clusters of accelerometers are used there to achieve a high level of event detection. Various sources of error in the location process were examined, and the errors were used to investigate the accuracy of event locations for both analog and digital monitoring systems in the mine. Similarly, an isotropic velocity model was adopted for the event location in a mine in the Lorraine coal basin, France, where the rock mass structure is highly complex and could not be approximated by simple models (Ben Sliman and Revalor, 1992). The arrival time of shear waves was additionally included into the location procedure and the optimum velocity ratio of $P$ to $S$ waves was estimated. In another coal mine in the Arc basin, Provence, a four-layer non-horizontal model was used for the location of seismic events. The influence of the number and type of seismic waves used in the location procedure on the location accuracy was considered by Dębski (1995). The best results are provided by a combination of direct and reflected waves, which are often observed in underground mines. An automatic complex procedure for searching arrival times of seismic events and the optimal location of seismic events was described by Růžek (1995). The procedure is based on the Hilbert transform of input records (energy flux envelopes) and on the Simplex location algorithm.

One of the problems involved in the event location procedures is whether norm $L_1$ or $L_2$ is the best to handle arbitrarily erratic data. Finding an optimal value of $p$ for the norm $L_p$, where $p$ is not necessarily equal to 1 or 2, might lead to better results. Kijko (1994) introduced the concept and the application of the adaptive $L_p$ norm to the location of seismic events in mines. This concept allows selection of the most appropriate value of $p$ and provides more robust solutions than $L_1$ or $L_2$ procedures when outliers occur in the input data.

The performance of the three norms in the presence of one outlying time residual is shown in Fig. 5, adapted from Kijko (1994) by Mendek and Scioatti (1997). For each value of the outlying time residual present at one station, 1000 seismic events were generated. For each event, $P$-wave arrival times recorded by the other 11 stations were perturbed by normally distributed random numbers with a standard deviation equal to 5 ms. Figure 5 shows the average error of the hypocenter location calculated by the three norm procedures for outlying residuals in the range of 5 to 100 ms. The results indicate that up to a certain value of an outlier the use
of the adaptive norm provides a substantial improvement over the $L_1$ norm; for larger outliers both norms provide similar results. When the outliers are small, the least-squares technique and the adaptive norm are essentially the same.

In classic location algorithms the only input data are the arrival times of seismic waves recorded by a given network, and no additional, a priori, information is used. The areas of the most probable locations of seismic events in mines, on the other hand, are usually known. Seismic activity usually concentrates in the vicinity of mining works, and even a short practice at any seismic network in a mine is sufficient to find from visual inspection of seismic records that the event occurred in a particular area of the mine. This a priori information can be included into an algorithm based on the Bayesian estimation theory, and it has been formulated for the location of seismic events in mines (Gibowicz and Kijko, 1994). The algorithm permits the combination of a priori information with that contained in the arrival times of seismic waves. Several numerical simulations of its application in different situations typical for mining conditions have demonstrated the effectiveness of the Bayesian approach and its superiority over the conventional location procedure. The Bayesian procedure is used in several coal and copper mines in Poland, and some experiments have also been performed in gold mines in South Africa and at the Men-Tou-Gou coal mine near Beijing in China.
When the velocity model is not known adequately, the location of seismic events can be improved by the arrival time difference (ATD) method, also known as "master event" location or relative location (see Gibowicz and Kijko, 1994, or Mendecki and Sciocatti, 1997). This is an old and highly popular method in which the travel time anomalies resulting from velocity model uncertainties are removed implicitly during the formation of equations. The procedure requires an accurately located master event (e.g., blast) in the proximity of the event to be located, that has reliable arrival times at stations used in the location calculations; all locations performed are related to the master event. Thus it is inherently assumed that the velocities of seismic waves from the master event to the stations and those from the target event are the same. The location accuracy, therefore, is heavily dependent on the distance between the master and target events as well as the difference in average wave velocity along the respective ray paths. To demonstrate the efficiency of the ATD location procedure, comprehensive tests were performed at copper mines in the Lubin mining district (Król and Kijko, 1991). Over 350 blasts from eight mining sectors were located twice, by the ATD and conventional methods. The average errors of epicenter locations by the two procedures are shown in Fig. 6. The ATD approach provides evidently better results. This technique has been routinely used at potash mines in Germany and at coal and copper mines in Poland.

Spottiswoode and Milev (1998) proposed a novel approach to the location procedure of seismic events in mines. They assembled seismic events with similar waveforms (doublets) into multiplets, relocated them simultaneously using arrival-time differences based on cross-correlation between all event pairs, and identified the orientation of the most likely plane on which the events are located. In contrast to the conventional master event method, no single master event was used. Each
event was a master event for the other event. The method was tested on seismic data recorded at a site in Blyvooruitzicht gold mine, South Africa, where a stabilizing pillar has been mined with the aid of precondition blasting. Some 40% of events that were previously located in a diffuse cloud within the pillar were found to form three distinct multiplets.

Simultaneous location of a group of seismic events and determination of a velocity model are founded on the same assumptions as the ATD approach: the travel time anomalies from a set of close seismic events tend to be strongly correlated and have nearly constant values at the same stations. In contrast to the ATD approach, in which the hypocenter of only one event is located, in this procedure a set of hypocenters and velocity model parameters are jointly determined. The procedure is known as the simultaneous structure and hypocenter (SSH) determination method (e.g., Gibowicz and Kijko, 1994) or as the joint hypocenter and velocity determination (JHVD) method (e.g., Mendecki and Sciocatti, 1997). The method is effective when applied to the location of seismic events that are spatially clustered, which is often the case in mining. It is assumed that the average velocity of a seismic wave from one event in the cluster to a given station is the same for all the other events in the cluster. If two events in the cluster, therefore, are far apart the assumption that the wave travels through the same rock mass would be no longer valid, and conversely, if the cluster is too small, increased numerical instability may appear since the arrival time equations would become very similar. The procedure does not require velocity calibration blasts and is fast. Its application in an area of high seismicity for a specific period of time makes it also possible to detect spatial and temporal variations of the velocity model parameters caused by stress migration.

The JHVD procedure was applied by Mendecki (1987) to study seismicity associated with a pillar in a gold mine in the Klerksdorp district and by Mendecki and Sciocatti (1997) to improve the location accuracy of seismic events occurring in different clusters at the Elansrand gold mine in South Africa. An example of the distribution of seismic events located by the JHVD method and those located using fixed velocities of seismic waves at a Zambian copper mine is shown in Fig. 7, reproduced from Mendecki and Sciocatti (1997). Events located by the JHVD procedure are on the average 50 m closer to the mined ore body than the same events located using fixed velocities. In the case where seismic events are not spatially clustered, more complex techniques for velocity inversion are required.

2.4. Seismic Tomography

Seismic tomographic imaging is used to map the spatial variation of seismic parameters in rocks. Periodic repetition of tomographic imaging in a given area would provide monitoring of rapidly changing conditions in mines. Several seismic
Relocation of 82 seismic events using JHVD on Zambian Copper Mine data
August 1994

Ore body where damage occurred

Fig. 7. An example of the distribution of seismic events located by the JHVD method and those located using fixed seismic wave velocities at a Zambian copper mine. [Mendecki and Sciocatti, 1997, reprinted from Seismic Monitoring in Mines, Fig. 5.6, © 1997 Chapman & Hall, with kind permission from Kluwer Academic Publishers.]

quantities may be imaged, such as wave velocity or attenuation. The most common are P-wave velocity images because of the relative simplicity and accuracy of the measurement. Velocity imaging has been shown to be particularly useful for seismicity studies in mines (e.g., Young and Maxwell, 1992; Maxwell and Young, 1996). Velocity images can be calculated by using either controlled explosive sources or passively monitored seismicity. Passive source velocity images can be computed from seismicity data by a simultaneous inversion procedure for source location and velocity structure. The results show very good correlation with a velocity image calculated using data from active explosive sources (Young et al., 1992; Maxwell and Young, 1993a, 1997a).

Detailed velocity structures will obviously improve the accuracy of hypocenter location. This was shown by an experiment carried out at Strathcona mine in the Sudbury Basin, Ontario, where the controlled source blasts were used as simulated passively monitored sources to image the velocity structure by application of joint hypocenter location-velocity inversion techniques (Maxwell and Young, 1993a). Blast monitoring data provide an opportunity to compare passive and active source inversion results. Source location errors were determined by calculating hypocenter location of blasts and comparing them with true blast locations. Using the controlled source-imaged velocity structure, the average source location error was found to be 5.5 m, as shown in Fig. 8, reproduced from Maxwell and Young (1993a). For the simplified model of a constant velocity, the average source location error was determined to be 10.5 m.
Comparisons between velocity images and seismicity in Canadian mines (Young and Maxwell, 1992; Young et al., 1992) and in South African mines (Maxwell and Young, 1996, 1997a) have shown that the largest events are associated with regions of high velocity, whereas low velocity areas are generally not seismic. An example of such a pattern in an active stope at Strathcona mine is shown in Fig. 9, reproduced from Young et al. (1992). In this figure the results of a passive source P-wave velocity image are plotted on a vertical cutting plane, and the location of a large seismic event and associated seismicity are shown as well.

Controlled studies around an underground opening in homogeneous unfractured granite at the Underground Research Laboratory (URL), Manitoba, where a 3.5-m-diameter circular tunnel was excavated at a depth of 420 m, have shown a strong correlation between velocity and stress (Maxwell and Young, 1995, 1996). To determine the velocity structure at the URL and to assess the accuracy of the source locations, signals generated by a series of controlled seismic sources were recorded with seismic array. The travel times were inverted to image the 3D-velocity structure around the tunnel. The resulting velocity image plotted on a cutting plane perpendicular to the tunnel axis, superimposed with the location
FIG. 9. Perspective view of the geometry of an active stope at Strathcona mine (labeled X) superimposed with the results of a passive source $P$-wave velocity image plotted on a vertical cutting plane and an isosurface of event density (labeled Y). The cutting plane is also shown with the superimposition of a magnitude 2.6 event and associated seismicity. [Young et al. (1992), Fig. 11, reprinted with kind permission from Birkhäuser Verlag AG.]

of seismic events located in that section of the tunnel, is shown in Fig. 10, reproduced from Maxwell and Young (1996). Velocities are plotted as relative perturbations from an initial starting model of the average velocity around the tunnel. Figure 10 also shows contours of the magnitude of the maximum compressive
Fig. 10. Active-source image of spatial variation in P-wave velocity on a plane perpendicular to the tunnel axis in the Underground Research Laboratory, Manitoba, superimposed with the location of microseismic events in that section of the tunnel. Maximum compressive stress contours, computed by a boundary element method, are also shown. [Maxwell and Young, 1996; reprinted from International Journal of Rock Mechanics and Mining Sciences, © 1996 Elsevier Science Ltd, with permission from Elsevier Science.]
stress, computed using a boundary element program designed to calculate elastic stress changes around underground openings. *In situ* stress gauges and extensometers generally validated stress changes around the tunnel. The seismicity appears to be generated in a region of velocity gradient, corresponding to the area separating the plastically deformed or failed zone from the elastically stressed zone.

Thus, the relationship between velocity and seismicity can be interpreted by frictional stability; the largest events occur in regions of high normal stress where the rock mass is able to store significant strain energy. A holistic interpretation of velocity images and seismicity can therefore be used to understand the seismic response of the rock mass to mining induced stress changes (e.g., Maxwell and Young, 1997a). Sequential images of temporal velocity variations can be used to isolate stress changes, based, for example, on repeated explosions at the same location or from seismicity data using a doublet technique. Several practical applications of these techniques to alleviate rockburst hazard in Canadian and South African mines have been discussed (Maxwell and Young, 1992, 1993b,c, 1995, 1996, 1997b, 1998).

Watanabe and Sassa (1996) developed two methods of seismic attenuation tomography. One is amplitude attenuation tomography, based on the first arrivals of *P* waves. The effects of velocity structure on wave amplitude were removed, and the decrease of wave frequency during propagation was estimated in order to obtain directly the values of quality factor. The other method is pulse broadening tomography, based on broadening of rise time or pulse width of the first arrivals. In this method, attenuation information is obtained from time data only. Watanabe and Sassa (1996) integrated the seismic travel time and seismic attenuation tomography and applied it to field data at Kamioka mine in Gifu prefecture, central Japan. Faults and fractured zones were detected.

Seismic velocity tomographic techniques are widely used in Polish coal mines to assess rockburst hazard and the efficiency of prevention measures undertaken there (e.g., Dubiński, 1997). Tomographic high frequency measurements at the Merkers potash mine in Thuringen, Germany, were carried out to assess the zones of high permeability in a salt rock formation, associated with open microcracks leading to dilatancy and acoustic emission (Manthei, 1997). Both velocity and frequency-dependent attenuation, corresponding to travel time and amplitude tomography imaging, were obtained for *P* and *S* waves, indicating the areas of lower velocity and higher attenuation. The image reconstruction accuracy of active amplitude tomography was discussed by Dębski (1997) who considered the dispersion of amplitudes of seismic waves radiated by various explosive sources.

Large systems of equations generated by seismic tomography can be solved by a number of methods. One-row-action techniques, such as algebraic reconstruction technique (ART) and simultaneous iterative reconstruction technique (SIRT) are commonly used in seismic tomography. The conjugate gradient technique has also been widely used in recent years; details can be found in Gibowicz and
Kijko (1994) and Maxwell and Young (1997a). Practical problems connected with velocity inversions in mines were discussed by Maxwell and Young (1993a).

2.5. Spatial and Temporal Patterns

Seismic events induced by mining are not uniformly distributed either in space or in time. Extensive studies performed on the space–time–energy distributions of seismic events in mines show that the tendency to form nests, swarms, and clusters is commonly observed (e.g., Talebi et al., 1994; Čiž and Růžek, 1997; Dunlop and Gaete, 1997; Gibowicz, 1997a). This tendency is apparent not only on a regional scale (see, e.g., Fig. 1) but on a highly local scale as well. Figure 11, reproduced from Trifu et al. (1993), shows, for example, the spatial distribution of seismic activity in a 1-km³ volume within Strathcona mine, Ontario, below 600 m depth, observed between the middle of February and April 1991. Another example is presented in Fig. 12, reproduced from Lasocki et al. (1997), where the distribution of seismicity is shown on a whole mine scale at Wujek coal mine, Poland, observed between 1988 and 1993.

A variety of techniques have been applied to quantify the spatial and temporal properties of natural earthquakes. Similar methods have also been applied to study

---

Fig. 11. Spatial distribution of seismic events recorded during the monitoring period in 1991 from a small area below a depth of 600 m at Strathcona mine, Ontario, Canada. [Reprinted from Trifu et al. (1993), Fig. 3.]
anomalous patterns and clustering in mining-induced seismicity. In many respects, mining seismicity and tectonic seismicity are quite similar and the process involved in the generation of seismicity patterns should be scale invariant (e.g., Gibowicz and Kijko, 1994). Preliminary results from cluster analyses of seismicity in a deep South African gold mine were presented by Kijko and Funk (1996). In total over 20,000 seismic events were analyzed, covering the time period of 1 year. The ability of the space–time clustering analysis applied to detect anomalous clusters of seismicity is demonstrated in Fig. 13, reproduced from Kijko and Funk (1996), where the location of four out of seven identified clusters is shown. Furthermore, the results of the analysis support the notion that seismicity at given active faces is affected by seismicity at other mining faces, and that the degree of interaction is a function of distance between the faces. Mining engineers have suspected this for some time.

Xie and Pariseau (1993) studied the seismicity associated with a rockburst-prone pillar in a metalliferous ore vein in the Galena mine, Idaho, and found that the distributions of seismic events have a fractal clustering structure. The degree of clustering increased with an upcoming major event, corresponding to a decreasing fractal dimension. The fractal dimension is positively correlated with the coefficient $b$ in the frequency-magnitude/energy relations (Guo and Ogata, 1995). Holub (1996a) reported that lower $b$ values correspond to a higher level of
Seismicity; and the inverse holds true for a number of coal mines and for different time intervals in the Ostrava-Karvina Mining District, Czech Republic.

Seismicity patterns in space and time, based on a number of seismic parameters, were extensively studied in South Africa (e.g., van Aswegen and Butler, 1993; Mendecki, 1997b,c; van Aswegen et al., 1997). To monitor rock mass response to mining, Mendecki (1997b,c) assumed that rock could be analyzed as a fluid that tends to flow into the voids created by mining. A seismic event is considered to be the result of instability in the flow of rock. He proposed to describe seismicity quantitatively by at least four largely independent parameters: average time and distance between consecutive seismic events, including source size, sum of seismic moments, and sum of energy.

From these parameters, for a given volume of rock mass and time interval, several new quantities can be calculated, describing seismic flow of rock. These are the apparent volume, the seismic viscosity, the seismic relaxation time, the Deborah number, the seismic diffusivity, the Schmidt number, and numerous others, all based largely on various combinations of seismic moment and energy. Additionally, the energy index of a given event within a large suite of events is often calculated. The index is the difference between the seismic energy of the event and

**Fig. 13.** Four clusters of seismic events at a gold mine in South Africa identified by a space–time clustering analysis algorithm. [Reprinted from Kijko and Funk (1996), Fig. 1, with kind permission from Birkhäuser Verlag AG.]
that of the regression line between the seismic moment and the energy, on a logarithmic scale, fitted to the data from the selected set of events (e.g., van Aswegen and Butler, 1993). Mendecki (1997b,c) and his colleagues (van Aswegen et al., 1997) believe that plotting various combinations of these parameters as a function of time can provide indications of impending instabilities, corresponding to major seismic events.

A new efficient method to study the nonrandom character of the spatial distribution of seismic series in mines was proposed by Lasocki et al. (1997). The method is based on the deflection of straight lines connecting epicenters of every two consecutive events, measured from the NS direction. The deflection series are analyzed by a nonparametric kernel estimation of their probability density function. The method was applied to three sets of events from Wujek coal mine, Poland, which occurred between 1988 and 1993 in different areas of the mine. The dominant directions of the distribution were significantly different for the three selected data sets. Furthermore, the dominant trends in the same set of events became different when various energy thresholds were considered. This suggests that the generation of events is controlled by various factors and that mining-induced seismicity is a multimodal phenomenon. Some of these trends could be associated with known geological, tectonic, and mining features, and the others were not. The analysis of deflection was also used to study the directional patterns of seismic events of energy equal to or greater than $10^6$ J from two major event clusters (see Fig. 1) in the Upper Silesian Coal Basin, Poland (Lasocki and Idziak, 1998). The clusters are neither confined to particular mines nor cover evenly the whole area of mining works. Four dominant directions of epicenter migration were found. Two trends are connected with the events that belong to the same cluster and are related to its shape. The other two are linked with the subseries of events that alternate between the clusters, indicating that mutual positions of events in such series are not randomly distributed. The dominant directions identified from the in-cluster series support recent hypotheses about possible tectonic instability of the region (e.g., Zuberek et al., 1996).

Seismic anisotropy is a field undergoing rapid expansion. A new understanding of the behavior of fluid-saturated microcracked rock allows noncatastrophic deformation to be monitored by seismic shear waves. The evolution of such rock under changing conditions can be calculated and future behavior predicted by an anisotropic poroelasticity (APE) model, developed by Zatsepin and Crampin (1997) and Crampin and Zatsepin (1997), where the parameters controlling microscale deformation also control shear-wave splitting. Shear-wave splitting occurs when shear waves propagate through solids that are effectively anisotropic, such as rocks containing aligned cracks or ellipsoidal inclusions. The behavior of shear waves crossing an anisotropic region is illustrated schematically in Fig. 14, reproduced from Crampin (1997). The splitting writes characteristic signatures into three-component seismograms, which can readily be interpreted in terms of geophysical parameters. Such distributions of stress-aligned microcracks are known
as extensive-dilatancy anisotropy (EDA) and the individual inclusions are known as EDA-cracks (Crampin et al., 1984). Stress-aligned shear-wave splitting is now widely observed in almost all rocks in the uppermost half of the crust (Crampin, 1994). The importance of APE for investigation of seismic events in mines is that there is now an underlying mathematical basis for monitoring, modeling, and predicting the behavior of stress and seismic events and rockbursts as mining progresses (Crampin, 1997). It should be noted, however, that the stress-based interpretation of shear-wave splitting is controversial; preferred mineral orientation is an alternative interpretation.

This new technique for monitoring and modeling of rock mass deformation has not been widely used as yet in studies of seismicity in mines. Graham et al. (1991) reported the results of their analysis of ground acceleration and velocity seismograms recorded during 1987 by a GEOS (General Earthquake Observation System) station situated on the surface above an active mining face at Western Deep Levels gold mine in South Africa. The polarizations of shear waves have a nearly uniform alignment, consistent with shear waves propagating through the effective anisotropy of parallel vertical microcracks throughout the rock mass. The observed anisotropy appeared to be caused by the regional stress regime rather than local stress disturbances generated by mining operations. The only mine-by-experiment known to us was undertaken in the Underground Research Laboratory, Manitoba, to investigate the excavation damage around a 3.5-m-diameter tunnel at 420 m depth in the Lac du Bonnet granite batholith (Holmes et al., 1993;
Crampin, 1997). Four differently polarized shear-wave signals were generated by nylon rods, which were pulsed with a Schmidt hammer. These sources produced very consistent shear-wave signals with the four different radiated signals splitting in the expected polarizations for stress-oriented nearly horizontal EDA-cracks.

3. GEOLOGY, MINING, AND SEISMICITY

Recent studies of seismicity in mines, undertaken in various mining districts of the world, have confirmed that at least two broad types of seismic events are observed universally—those directly connected with mining operations and those associated with movement on major geological discontinuities. Obviously, seismicity in mines is strongly affected by local geology and tectonics, and interaction between mining and lithostatic and tectonic stresses on a local and regional scale (e.g., Gibowicz and Kijko, 1994). In situ stress measurements collected at the Underground Research Laboratory in Manitoba indicate that major geological discontinuities can act as boundaries for stress domains, and that the magnitude and direction of the stress field can change rapidly when these discontinuities are traversed (Martin, 1990).

Johnston (1992) published a worldwide survey of various parameters connected with mining-associated seismicity. Seismic events from the Republic of South Africa, Sweden, Poland, Czechoslovakia, Canada, and the United States were sorted according to type of geological conditions and preexisting regional stress field. The seismicity was found to be bimodal and was divided into two broad types: seismicity directly associated with mining activity and geometry, and seismicity occurring on preexisting faults and discontinuities in or near the mining district. Locations of the first-type events are determined consistently by the location of the mine face and by local geological structure. Their epicenters are located at a specific distance ahead and behind the active face and their depths usually correspond to geological and mechanical zones of weakness. The maximum magnitude of these events depends mostly on the strength properties of the rock mass, and the overall level of seismicity is believed to be a function of the extent of the excavation. The characteristics of the second-type events are much less clear. Their locations depend on the location of preexisting discontinuities in the area, but controlling factors of their maximum magnitude are not well understood.

Both types of seismicity were extensively studied in the mines of the Ostrava-Karvina Coal Basin, Czech Republic (Główacka et al., 1992; Holub, 1996b, 1997). The energy–frequency relation for seismic events at Doubrava coal mine from this district is shown, as an example, in Fig. 15, reproduced from Główacka et al. (1992). The observed values are approximated by a bimodal and Gutenberg-Richter truncated distributions. The bimodal distribution fits the data rather well. It was found that the events with energy greater than $10^6$ J were the result of stress accumulation in an area greater than that corresponding directly to mining excavations.
Similarly, Sprenke et al. (1991) believe that low-level seismicity in the immediate vicinity of the Lucky Friday mine in the Coeur d’Alene mining district, Idaho, is driven in the short term by mining rate. The district is situated within a minor zone of seismicity and in the longer term, cycles of tectonic stress build-up and release may be important.

3.1. Mining Operations and Seismicity

Seismic events of the first type are of low magnitude, and their number is generally a function of mining activity. The events usually occur within 100 m of the mining face or on some preexisting discontinuities and zones of weakness near the face. Intact rocks can be ruptured as well. The spatial distribution of seismic events of the first type is well illustrated in Figs. 12 and 16. Figure 16, reproduced from Hatherly et al. (1997), shows horizontal locations of 629 events recorded in 1994 at Gordonstone coal mine in Central Queensland, Australia. Several triaxial geophones were deployed there in three vertical boreholes and over a 6-week period more than 1200 events were recorded. The events were well located, with an accuracy of a few meters. The seismicity correlated with periods of longwall production and occurred mainly within the mining panel 250 m wide. The zone of activity extended from behind the face at the sides of the panel and up to 70 m ahead of the face in the middle. The zone extended to a height of about 120 m above the seam, and there was lesser activity to a depth of about 30 m into the floor. Hatherly and Luo (1999) proposed a model for longwall caving which involves...
two stages of failure. The first is the failure ahead of the longwall face and in the rock mass above and below the coal seam; seismicity with dominant shear failure is observed. When the longwall face mines through the damaged rock mass, the second stage commences. The seismic events can originate behind the face; tensile failure is postulated.

Senfaute et al. (1997) performed a detailed study of the space distribution of seismic events at a longwall working face at the Provence coal mine, France. The longwall was 720 m long and 200 m wide, and during its extraction over 2200 seismic events were recorded. Their distribution depends on the configuration of mining works (see Fig. 17). When the working face ran along the old mined out areas, the events were more numerous ahead of the face (Fig. 17a). But when the working face ran along a pillar, the events were more numerous behind the face (Fig. 17b).
Seismic monitoring of a test tunnel, at a depth of 420 m in the Underground Research Laboratory, provided interesting results (Talebi and Young, 1992; Martin and Young, 1993). During excavation of the tunnel, spalling in the roof and floor resulted in typical well-bore breakout geometry. The location of seismic events associated with the excavation generally fell within one radius of the tunnel wall and appeared as random events. Within the well-bore breakout geometry, however, the seismic events were clustered implying crack damage and coalescence. Laboratory experiments suggest that a small amount of crack damage can reduce intact rock strength by 50% or more.
The correlation between seismic activity and extracted deposit volume has long been known from observations. Such dependence is well illustrated in Fig. 18, reproduced from Arabasz et al. (1997), where cumulative seismic energy is shown as a function of the volume of coal mined from mine B in the Wasatch Plateau and Book Cliffs mining districts of east-central Utah. The sum of released seismic energy follows a power law with respect to the volume of mined rocks, and a power-law exponent of 2 to 3 provides the relation that fits the data quite well. Kremenetskaya and Tryapitsin (1995) obtained results to some extent similar for the cumulative seismic energy release and the amount of extracted ore in the mines of the Khibiny Massif, Kola Peninsula, Russia, between 1948 and 1991.

Stope convergence measurements are often used to study seismic processes in mines. Boler and Swanson (1992) investigated stope convergence behavior at Galena mine, Idaho. The mine experienced seismicity during extraction of near-vertical silver veins to a depth of 2 km. Three gauges for monitoring the convergence of a burst-prone stope were installed in June 1990 and monitored for several months. The stope was also monitored by an array of accelerometers, and seismic events with magnitude $-5$ and above were recorded with an automatic monitoring system. Coseismic convergence steps were observed for seismic events of magnitude $-2$ and greater. Figure 19 shows the cumulative convergence versus the cumulative number of seismic events for the study period. The tracking of event
Fig. 19. Cumulative convergence versus cumulative number of seismic events recorded in 1990 at Galena mine in the Coeur d'Alene district, Idaho. [Boler and Swanson (1992), Fig. 5, reprinted with kind permission from Birkhäuser Verlag AG.]

occurrence and stope convergence is remarkably steady, with some 1200 events recorded per millimeter of closure.

Malan and Spottiswoode (1997) found that time-dependent deformations around deep-level mines occur at rates well in excess of those expected from creep of a solid rock mass. They modeled the time-dependency by extension and mobilization of the fracture zone, using a boundary element technique. Viscoplastic behavior of existing cracks and the resulting formation of new fractures adequately described the rate of stope closure and seismicity over a time scale of hours. They introduced time-dependent weakening of the rock mass into the model, which provided a better fit to the observed stope closure during the weekly cycle of mining.

Detailed studies have shown an association between zones of high P-wave velocity and the hypocentral location of seismic events induced by mining (e.g., Maxwell and Young, 1993b). The mechanism of the association, however, is not well understood. Maxwell and Young (1995) investigated the relationship between stress, velocity, and induced seismicity around a tunnel in the Underground Research Laboratory. The rock mass around the tunnel was unfractured and essentially homogeneous. An image of the velocity structure around the tunnel indicated a distinct zone of high velocity corresponding to a zone of increased elastic stresses, where a breakout notch formed. An area of decreased velocity, perpendicular to this zone, was found to be associated with a region of tensile elastic stress. The stress changes are believed to have preferentially opened and closed preexisting
microfractures in the rock, decreasing and increasing the $P$-wave velocity. The seismicity induced by the excavation in the notch region was found to be located in the transition between the high-velocity/high-stress, intact rock and the low-velocity/destressed, fractured rock (Maxwell and Young, 1995, 1998). Scott et al. (1997) have reported similar results from investigation of a rockburst site at the Sunshine mine, Kellogg, Idaho. Seismic activity there appeared to be associated with areas where a velocity gradient was apparent, as documented by seismic tomography.

Mining operations are occasionally associated with exceptionally large seismic events as well. Irregular mine geometry, especially when “Chamber and Pillar” method is used, could lead to large magnitude events. The 1989 Volkershausen rockburst in the potash mining district along the river Werra, Germany, with magnitude 5.4 (Knoll, 1990) is a spectacular example of such an effect. The event caused the collapse of some 3200 pillars at a depth of 850–900 m within an area of about $3 \times 3$ km. Similarly, the 1995 Solvay Trona mine, Wyoming, event of magnitude 5.1 entailed the extensive collapse of pillars at a depth of 490 m over an area of about $2 \times 1$ km (Pechmann et al., 1995). Waveform modeling of seismic records indicated a considerable implosive component in the seismic source, which is consistent with observations of surface subsidence.

An interesting example of direct correlation between mining and seismicity comes from a copper mine in Poland (Wiejacz, 1993). In July 1992 a miner’s strike was declared for the whole Lubin Mining District, which lasted 32 days. At the Polkowice mine the labor stoppage resulted in a threefold drop in the number of observed seismic events with energy greater than $10^2$ J. After the mining was resumed, the number of seismic events reached its prestrike level within some 30 days. Later on, the number of events was a little larger than before the stoppage, possibly because of greater mining activity at that time. A similar decrease of seismicity was also observed by the GERESS (German Seismic System) array in Germany, at a distance of about 320 km (Jost and Jost, 1994). During the strike, the number of GERESS automatic locations of seismic events from the Lubin area was reduced by 57%, and the number of events in the NEIC bulletin decreased by 64%.

3.2. Geological Structures and Seismicity

Although lithostatic stresses and their redistribution around mine excavations are the dominant driving forces in the generation of most seismic events in mines, tectonic stresses also play a significant role. This is especially true in the case of the largest mine tremors, with magnitude close to 4 and above. Wong (1993) has reviewed the available focal mechanisms of large events from various mining districts to evaluate the effects of tectonic stresses. He found that, in general, seismic events in mines occur in regions subjected to horizontal compressive
tectonic stresses. The larger events are often the result of reverse and sometimes strike-slip faulting. This mode of deformation probably reflects the increase in shear stress along preexisting zones of weakness. Alternatively, tectonic stresses could act as a trigger of seismic events where the main sources of strain energy are the redistributed lithostatic stresses.

The seismicity induced in deep gold mines in South Africa may be an exception. The seismicity there is mainly characterized by normal faulting, believed to be caused by stope closure, and tectonic stresses there do not appear to be significant. Dennison and van Aswegen (1993) modeled stresses on the Tanton fault at President Steyn gold mine, taking into account its geometrical and frictional properties. They also monitored fault dynamic behavior during extensive mining of a tabular ore-body at a depth of 2400 m and found that tectonic stresses were not needed to explain the fault behavior, and in situ stress measurements showed that these stresses were not important. Gay (1993) evaluated the relationship between rockbursts and seismicity associated with the two main mining methods used in deep South African mines, namely scattered and longwall mining. He found that the frequency of rockbursts increases when mining takes place close to geological structures. The nature of the damage, however, is such that using better support systems could readily prevent it.

In the Coeur d'Alene mining district, Idaho, compressive tectonic stresses are important. Whyatt et al. (1993) described an unusual concentration of rockburst activity during development of the 5300 level of the Lucky Friday mine in 1990, strongly correlated with in situ stress and geological anomalies. The localized nature of the measured stress field has been established by comparing overcore stress measurements, raise bore breakouts, and geological indications of stress orientations collected from the whole Lucky Friday mine and the adjacent Star mine. The mining-induced stress at the crosscut stub was estimated with a homogeneous elastic model to be only 5% of the natural stress field. Swanson (1992) examined the relationship between local geology, vein/stope layout, and the location of large seismic events and rockbursts at Galena mine. He found two structures associated with large events and rockbursts: the largest vein in the mine and a near-vertical plane striking throughout the length of the mine that is subparallel to the major faults there. He used models of stress interaction to show that mining-induced deformation was mobilized along a 1.5-km length of this plane during the occurrence of a series of large seismic events in 1990.

Creighton mine is the deepest and most seismically active mine operating in the Sudbury Basin, Ontario (Morrison et al., 1993). Mining extends to a depth of 2400 m and seismicity is monitored below the 1800 m level. Morrison et al. (1993) have studied spatial distributions of seismic energy release in this mine, using a 2D numerical model involving a single frictional surface and a changing mining geometry. Their results imply that chaotic behavior is an intrinsic part of this simple representation of mining. The rock environment is naturally complex
but often appears to display unusual sensitivity to minor changes that point to self-organized criticality. Steacy et al. (1997) used a 3D self-organized critical model to simulate seismicity in Creighton mine. Almost 400 faults were included in the model, each represented by an individual 2D cellular automata. The model was loaded from point sources that simulated blasting, and was parameterized by loading it with an actual blast sequence and comparing the predicted and observed seismicity. The model successfully predicted the global statistics of seismicity observed in the mine, and the number and timing of large events as well.

Statistical analysis of spatial distribution of seismic events induced by mining in the Ostrava-Karvina Coal Basin, Czech Republic, shows that geological structures there act as concentrators of stresses generated by mining operations in their vicinity (Kaláb and Müller, 1996).

Recently, the most extensive studies on relationships between geological features and seismicity in mines have been carried out in the Upper Silesian Coal Basin, Poland (e.g., Idziak, 1996b; Idziak and Teper, 1996a; Teper, 1996; Zuberek et al., 1996; Idziak et al., 1997; Mutke and Stec, 1997; Wiejacz and Ługowski, 1997; Lasocki and Idziak, 1998). Figure 20, reproduced from Zuberek et al. (1996), shows the spatial distribution of seismic events induced by mining in the Basin on the complex tectonic background of the area. The seismic events concentrate in specific regions corresponding to different geological units. The studies indicate that tectonics plays a significant role in the generation of at least some of the largest events in the area. Some parts of the Upper Silesian Coal Basin (e.g., the zones of large latitudinal faults) are connected with large discontinuities in the deep crustal basement with active shear stresses. The spatial distribution of larger seismic events is not uniform on a mine scale either. Figure 21, reproduced from Mutke and Stec (1997), shows, as an example, the distribution of events with magnitude greater than 1.2 at Kleofas coal mine, observed from 1990 to 1995.

Nonuniform distribution of seismic events suggests the fractal character of seismicity in the Basin (Idziak and Zuberek, 1995; Idziak, 1996b), where the fault patterns also display their fractal nature (Teper and Idziak, 1995; Idziak and Teper, 1996a). The temporal variations of event coordinates indicate some kind of strange attractors, and imply involvement of dynamic processes on a regional scale in the generation of seismicity. Furthermore, migration of seismicity between its major clusters (see Fig. 1) cannot be explained on the ground of changes in mining works (Idziak, 1996b). The dominant directions of epicenter distribution within the clusters also support recent hypotheses addressing the internal structure and low tectonic instability of the deep crustal basement in Upper Silesia (Lasocki and Idziak, 1998).

Idziak et al. (1997) have investigated the focal mechanism of almost 400 seismic events, which occurred in the vicinity of the Klodnicki fault—the main regional dislocation in the central area of the Upper Silesian Coal Basin. For selected groups of events with similar orientations of nodal planes, local stress tensors were also
Fig. 20. Spatial distribution of seismic activity, marked by shaded zones, induced by mining in the Upper Silesian Coal Basin, Poland, displayed on the complex tectonic background of the region. [Reprinted from Zuberek et al. (1996), Fig. 1.]

determined. A relationship between the spatial distribution of slip vector and the pattern of neotectonic deformations of the Carboniferous rock mass was found. Wiejacz and Ługowski (1997) made another study of the effects of geological structures on the mechanism of seismic events at Wujek coal mine. Figure 22, reproduced from their work, shows focal mechanisms of events located near the Arkona fault—one of the principal faults in the mine. They found that such a fault, even if tectonically passive, could play an important role in the generation of seismicity in the mine. In the vicinity of faults the directions of excavations and longwall fronts seem to play a secondary role.
Fig. 21. Distribution of seismic events with magnitude greater than 1.2 recorded between 1990 and 1995 at Kleofas coal mine, Poland. Dashed lines mark tectonic faults. [Mutke and Stec (1997), Fig. 3, reprinted from Rockbursts and Seismicity in Mines, Proceedings of the 4th International Symposium, Krakow, Poland, 11-14 August 1997. 450 pp., EUR 137.50/US $162.00/GBP97.00, A. A. Balkema, P.O. Box 1675, Rotterdam, Netherlands.]

Fig. 22. Distribution of double-couple focal mechanisms of seismic events located near the Arkona fault (continuous line) at Wujek coal mine, Poland. [Wiejacz and Lugowski (1997), Fig. 4, reprinted from Rockbursts and Seismicity in Mines, Proceedings of the 4th International Symposium, Krakow, Poland, 11-14 August 1997. 450 pp., EUR 137.50/US $162.00/GBP97.00, A. A. Balkema, P.O. Box 1675, Rotterdam, Netherlands.]
4. Source Parameters and Scaling Relations

Determination of source parameters of seismic events in mines is based mostly on spectral analysis of $P$ and $S$ waves. The other approach, waveform modeling in the time domain, involves rather complex techniques and is not suitable for routine applications to large sets of data usually generated by seismic monitoring in mines. The approach to determine source parameters in the time domain from the source time function appears as a possibility for investigation of seismicity in mines; it is briefly described in Chapter 6.

Spectral analysis has become a standard technique used in studies of seismic events induced by mining. Early attempts have shown that simple source models in the form of a circular dislocation (Brune, 1970, 1971; Madariaga, 1976) can be successfully used for the interpretation of seismic spectra and for the determination of source parameters of seismic events in mines. Spectral analysis usually employs the fast Fourier transform algorithm, which is computationally efficient and produces reasonable results, provided that the effect of the windowing of time series is properly taken into account. The windowing problem manifests itself as "energy leakage" into the sidelobes, obscuring other spectral features and distorting source parameters. This effect is usually suppressed by application of various tapers. The use of adaptive multitapers instead of a single taper (e.g., Hamming window) provides more stable and better results (Mendecki and Niewiadomski, 1997).

4.1. Source Parameters

Seismic moment, corner frequency, and seismic energy flux are traditionally inverted from the spectra, corrected for the instrumental, distance, and attenuation effects of each waveform, and then averaged. Occasionally, the quality factor $Q$ is considered as an independent parameter during the inversion to improve the match between the observed and theoretical spectra (e.g., Spottiswoode, 1993; Mendecki, 1993). Mendecki and Niewiadomski (1997) proposed stacking the spectra of $P$, $SH$ or $SV$ waves over all the stations involved and determining the source parameters from the stacked spectra. This approach tends to average out incoherent site effects, source directivity, and random fluctuations of the high-frequency part of the spectrum. When $Q$ is not known, the best average $Q$ may be determined for all stations during the stacking procedure.

From the corner frequency, the source radius can be estimated either from the Brune (1970, 1971) source model or from the Madariaga (1976) model. Both models are used in studies of seismic events in mines. In general, Brune's model tends to overestimate the source size in comparison to the size of underground damage caused by rockbursts (see Gibowicz and Kijko, 1994). The source radii derived
from the Madariaga model, on the other hand, better fit documented evidence of underground damage (e.g., Trifu et al., 1995).

Imanishi and Takeo (1998) recently proposed a new method to estimate the source dimensions of small earthquakes based on stopping phases which are radiated at the termination of rupture propagation, without assuming a rupture velocity. They detected on accelerograms two high-frequency stopping phases, being the Hilbert transforms of each other for several small earthquakes at two local stations. Assuming a circular source, it was possible to determine relative locations of nearest and farthest edges of the circular fault using the difference of arrival times between these two stopping phases. It could be expected that the application of this method to seismic events in mines would provide better estimates of their source dimensions that would be independent of a rupture velocity.

Stress release estimates, the average stress drop and apparent stress, are readily calculated from seismic moment, radiated energy, and source radius. A multitude of other parameters has been also proposed to describe the behavior of seismic sources in mines (Mendecki, 1993; Mendecki, 1997b,c; van Aswegen et al., 1997); they all result from a combination of basically two independent parameters describing the seismic spectrum.

An automatic procedure for determining source parameters from spectra is needed for daily processing of large numbers of seismic events observed in mines. Many seismic networks record hundreds of events a day. A highly efficient algorithm for calculating source parameters is based on ground velocity and displacement power spectra, proposed by Andrews (1986), and is used by the ISS monitoring system (Mendecki and Niewiadomski, 1997).

Source parameters have been estimated for thousands of seismic events, in the moment magnitude range from −3.6 to over 5, in various mining districts all over the world. Figure 23, reproduced from Kaiser and Maloney (1997a), shows the range of seismic moment and seismic energy of seismic events most often studied. The relations between S- and P-wave corner frequencies are of practical interest, since they both are used to estimate source dimensions. In general, the corner frequencies of P waves are expected to be greater than those of S waves. Abercrombie and Leary (1993), who analyzed borehole records of small earthquakes in the 2- to 250-Hz frequency range, reported that the P-wave corner frequencies are approximately equal to those of S waves, but the scatter in their data is considerable. Gibowicz (1995) studied the corner frequencies of P and S waves generated by some 650 seismic events in various mines. He found that the relation between these corner frequencies is linear on a logarithmic scale for frequencies from 5 to 4010 Hz for P waves and from 4 to 3250 Hz for S waves. The relation is shown in Fig. 24, reproduced from Gibowicz (1995). The corner frequency of P waves is higher as a rule than that of S waves by about 25%.

Source parameters have been used for various practical purposes, such as assessment of rockburst hazard and damage potential, design of support in rockburst-prone areas, and rockburst prediction.
Urbancic and Young (1993) investigated space–time variations in source parameters of microseismic events, with moment magnitude less than 0, at Strathcona mine, Sudbury, Canada. The microseismicity was directly associated with excavation of ore at a depth of 710 m. Source parameters were calculated for 85 events located within the immediate excavation volume. Two-thirds of these events was characterized by low values of $S$ over $P$ wave energy ratio and low apparent stress, implying the existence of a non-shear component of failure. Spatially, the source parameters allowed definition of a pattern of systematic fracture behavior in the
rock mass, comparing favorably with mapped structure associated with ground problems. Temporally, the events were observed to locate either in close proximity to the excavation at depth or at shallower depths away from the excavation during the preexcavation and postexcavation periods. The source parameter levels generally increased from the preexcavation to the postexcavation periods, indicating that stress levels were increasing as a result of the excavation geometry and procedure. This was considered to be significant as rockbursts occurred above the area 6 months after the excavation.

Trifu et al. (1995) used source parameters for an evaluation of homogeneous and inhomogeneous faulting models to assess damage potential at Strathcona mine.
The analysis included an investigation of peak ground velocity and acceleration scaling relations with magnitude for near-source data. For the inhomogeneous source model, estimates of asperity radius were obtained. To validate the use of this source model for seismic hazard assessment, a comparison of its predicted ground motion level with underground geomechanical measurements was carried out. For distances close to the source, corresponding to relatively higher damage levels, a good agreement between these two kinds of measurements was found. Furthermore, a good agreement between the asperity radii and the size of the regions of the highest observed local damage provided additional support for use of the inhomogeneous source model in the assessment of seismic hazard in mines.

Alcott et al. (1998) discussed a methodology to assess potential rockburst hazards using seismic source parameters, which is designed to provide a simple means for including the most relevant source parameters into the daily ground control decision-making process. The ISS data and observed damage at a Brunswick mine, Canada, were used to calibrate and verify the proposed methodology. Seismic events were classified into four groups: those representing minor hazard, seismically triggered gravity-driven hazard, stress-adjustment-driven hazard, and deformation-driven hazard. Using the proposed technique, based on the variations of seismic energy, seismic moment, and apparent stress, it was possible to delineate areas of elevated rockburst hazard potential and to identify precursory and decay trends indicating progressive worsening and improving conditions in seismically active areas of the mine.

Glazer (1998) applied seismic source parameters to combat rockburst hazard in the Vaal Reefs gold mine, South Africa. He introduced the concept of apparent stress index to normalize apparent stress values for events within a wide range of seismic moment values. The index is calculated as the ratio of measured over mean values of apparent stress. Thus, values of the index above one indicate higher stress levels than the average, and values below one indicate stress levels lower than the average. It was found that large events are associated with areas of high gradients of the apparent stress index. Damage is often observed in areas characterized by high values of the index, even when a seismic event is located outside of these areas. Changes in the values of the index before large events are occasionally observed. Glazer (1998) believes that contour mapping of the apparent stress index could help to optimize the design of support.

Król (1998) investigated the focal mechanism and source parameters of 246 seismic events, which occurred in 1997 at Polkowice copper mine, Poland. He found a number of relationships between the average values of selected source parameters and the quantities describing geological structure, mining conditions, seismic activity, and rockburst hazard in the mine.

The dynamic loads imposed by rockbursts are directly related to the resultant ground motions. Thus, the first step in design of systems for the support of openings
affected by such loading is to assess the ground motion characteristics. Kaiser and Maloney (1997a,b) presented scaling laws, supported by theoretical considerations and based on world-wide data, for the design of support in rockburst-prone areas. The design scaling laws relate seismic source intensity (event magnitude, radiated energy, or a combination of seismic moment and stress drop) to the peak ground motion parameters in the design area, some distance from the source. These scaling laws reflect the most relevant critical conditions for engineering design.

Cichowicz (1991) investigated the time variation of source parameters of microseismic events before and after large seismic events in a deep gold mine in South Africa. The small-scale stope fracturing in the close vicinity of a dyke was studied. The magnitude of microseismic events ranged from -3 to -1. An automatic monitoring system was used, consisting of triaxial accelerometers installed in a borehole 150 m ahead of the advancing face. Seismic moment, radiated energy, corner frequency, dynamic stress drop, Q-coda of S waves, and the degree of polarization of P waves were estimated. The first four parameters describe the seismic source and the last two parameters characterize the rock mass. It was found that the degree of polarization of P waves decreased between 0.5 and 2 days before each larger event with magnitude greater than 0.5. Series of foreshocks typically had higher stress drops than the average corresponding microevents, and they were associated with gradual stress concentration within the dyke, which formed an asperity of greater strength than the surrounding rock mass.

4.2. Scaling Relations

A source scaling relation describes the manner in which the source duration or the source dimension increases with increasing seismic moment. In studies of large earthquakes it has been found that stress drop is roughly independent of the seismic moment, which means that seismic moment is proportional to the third power of the rupture length. The constant stress drop model implies a self-similar rupture process regardless of the scale of the seismic events.

This scaling relation has been confirmed by innumerable studies and has become an accepted model for large and moderate earthquakes. The constant stress drop pattern has also been reported for small volcanic events and for volcanic explosion earthquakes (Nishimura and Hamaguchi, 1993). The source parameters of small earthquakes recorded at 2.5-km depth in southern California are compatible with earthquake scaling at constant stress drop down to magnitude -2 (Abercrombie and Leary, 1993). The constancy of stress drop was found long ago for some seismic events in gold mines in South Africa and in coal mines in Poland (for a review, see Gibowicz and Kijko, 1994). In contrast to these results, several studies of small earthquakes have been interpreted as evidence of a breakdown in the similarity between large and small earthquakes (e.g., Glassmoyer and Borcherdt,
1990). The breakdown in scaling relations for small seismic events induced by mining has been also reported.

Feignier (1991) investigated the influence of geological structure on the scaling relations for seismic events with local magnitudes ranging from 1.0 to 4.2 from the Lacq gas field in southwestern France. He demonstrated that many factors influence source parameters and scaling relations, but the overall result for the whole data set implies a self-similar behavior. This indicates that the tectonic stress field and the depth of events are the main factors affecting the scaling relations. A detailed analysis of the seismic response of different geological layers, however, has shown that stress drop depends on the strength of the layer. In contrast to large and moderate earthquakes, this feature could be observed on small events because of their limited spatial extension; and this is the second-order factor influencing scaling relations. Furthermore, Feignier (1991) also found that self-similarity is not always verified in each geological layer. Areas with a high degree of fracturing induce low stress drops whereas a more intact area leads to higher stress drop events; and this could be a third-order factor affecting scaling relations.

Gibowicz et al. (1991) studied source parameters and scaling relations of seismic events at the Underground Research Laboratory in the magnitude range from \(-3.6\) to \(-1.9\). They found that the scaling relation between seismic moment and source radius indicated a decrease in stress drop with decreasing seismic moment. Characteristic tremors with a fault length of about 1 m seem to be generated in the area, enforced by the excavation geometry and local microtectonics.

Urbancic et al. (1992a) examined excavation induced microseismic events, with moment magnitudes less than 0, at the Underground Research Laboratory and at Strathcona mine in an attempt to determine the role of geology, excavation geometry, and stress on scaling relations between seismic moment and source radius. They observed that the interaction of stresses with preexisting fractures, fracture complexity, and depth of events are the main factors influencing source parameters and scaling relations of very small events. Self-similar relations were found for events at similar depths or for weakly structured rock mass with reduced clamping stresses. Nonsimilar behavior, on the other hand, was found for events with increasing depth or for heavily fractured areas under stress confinement. Overall, depth and fracture complexities appear to favor nonsimilar generation processes of seismic events in mines.

Urbancic et al. (1993b) presented the results of their analysis of 68 microseismic events, with moment magnitude from \(-2.4\) to \(-1.1\), associated with the excavation of ore in a highly stressed rock mass at about 620 m depth at Strathcona mine. A dependence of static and dynamic stress drops on seismic moment was found. The observed scaling dependence does not appear to be related to the source inhomogeneity, but is thought to be related to scale length effects. In contrast, Talebi et al. (1994) concluded that the scaling relation between seismic moment and source radius for events at Creighton mine strongly indicates a self-similar
behavior of their sources. They analyzed some 150 events in the magnitude range from 0.4 to 2.6.

Urbancic and Trifu (1996) expanded their investigation into scaling behavior by examining the role of source complexity, rupture velocity, and deviatoric stresses. They analyzed microseismic events, with moment magnitude ranging from $-2.4$ to $-0.3$, which occurred within both highly and moderately stressed and fractured rock masses at Strathcona mine. Their results indicate a relatively strong dependence between stress release, over two orders of magnitude, and seismic moment, over three orders of magnitude.

McGarr (1994) attempted to explain the apparent dependence of stress release parameters on seismic moment in his comparison between mining-induced and laboratory seismic events. He considered apparent stress as a function of seismic moment and found that over a huge range of seismic moment the apparent stress falls in the range from about 0.01 to 10 MPa. Thus, the apparent stress globally shows no systematic dependence on seismic moment, supporting the idea that earthquakes generally are constant stress drop phenomena. For the individual sets of events, however, induced by either mining or laboratory, the apparent stress exhibits strong and systematic dependence on seismic moment. This strong dependence is understandable for the laboratory events because the fault area in this case is fixed. For each set of induced events, the fault area also appears to be approximately fixed but the slip is inhomogeneous, presumably as a result of barriers distributed over the fault plane. Thus, for individual sequences the principal difference between large and small events involves average slip and not source dimension (McGarr, 1994). The trade-off between stress drop and rupture velocity inherent in all kinematic source models, on the other hand, implies that the dependence on seismic moment can be attributed just as well to systematic variations of rupture velocity (Deichmann, 1997).

Gibowicz (1995) reached some conclusions similar to those of McGarr (1994) in his study of scaling relations for 1575 seismic events induced in South African, Canadian, Polish, and German underground mines, with moment magnitude ranging from $-3.6$ to 4.1 (seismic moment from $5 \times 10^5$ to 2 $\times 10^{15}$ N·m). The constant stress drop scaling means that seismic moment is inversely proportional to the third power of corner frequency. Such behavior was confirmed for most of the considered data. The only evidence for a breakdown in the similarity between large and small events seems to be apparent for the smallest seismic events with moment magnitude below $-2.5$. Figure 25, reproduced from Gibowicz (1995), shows the distribution of seismic moment versus S-wave corner frequency for all the investigated events. It should be noted that the average values of seismic moment that referred to the same range of corner frequency are vastly different in various mining areas. For the same corner frequency, the seismic moment of the events from hard-rock gold mines in South Africa is higher than that of the events from Polish coal mines, on the average by a factor of about 150. Even at a single mine, such as Strathcona mine, the differences in seismic moment between various data sets
SEISMICITY INDUCED BY MINING

SOUTH AFRICA, CANADA, POLAND

Fig. 25. Seismic moment versus S-wave corner frequency for 1575 seismic events from South African (SAF, WDL), Canadian (CAN, SDB), and Polish (POL) mines. [Gibowicz (1995), Fig. 12, reprinted with kind permission from Birkhäuser Verlag AG.]

are characterized by a factor as high as 7. Similar conclusions were drawn in an additional study of scaling relations for seismic events induced at Polish copper mines (Gibowicz, 1997b). A comparison of source parameters of seismic events at Polish coal and copper mines was also made by Domafiski (1997).

5. SOURCE MECHANISM

The most important progress made in source mechanism studies of seismic events induced by mining during the last decade is the seismic moment tensor inversion technique, used on a massive scale. Results from earthquake mechanism
studies indicate growing evidence that alternative mechanisms other than shear failure are possible (e.g., Dziewonski et al., 1997). The most prominent cases of what appear to be anomalous focal mechanisms are reported from seismicity studies in mines. The moment tensor inversion technique has been used in numerous studies of seismic events in deep mines in South Africa, Poland, Canada, and Japan. A brief overview of the obtained results follows.

5.1. Seismic Moment Tensor

In general, seismic moment tensor is used to describe the source mechanism of an earthquake under the assumption that the point-source approximation is valid. This approximation is valid for the long period displacement—the fault-plane dimensions are assumed to be shorter than the wavelength of the seismic waves used in the analysis. To the first order, moment tensors describe completely the equivalent forces of general seismic point sources. The conservation of angular momentum for the equivalent forces leads to the symmetry of the moment tensor. The eigenvalues of a symmetric second-order tensor are all real and its eigenvectors are mutually orthogonal. For low frequencies the moment tensor can be treated as having the same source time function for all its components. Given a set of amplitudes at a given time, the moment tensor can be obtained by inversion of the linear system of equations containing the six components of the moment tensor, the observed amplitudes, and the transfer and excitation functions for seismic waves. A moment tensor then can be decomposed into isotropic and deviatoric parts. The isotropic part corresponds to a volume change in the source. The deviatoric part can be decomposed in turn into a Compensated Linear Vector Dipole (CLVD), corresponding to a uniaxial compression or tension, and a Double Couple (DC), describing shearing processes in the source. Such decomposition seems to be the most interesting one for source studies of seismic events induced by mining, although a multitude of different decompositions are possible (e.g., Gibowicz, 1993; Gibowicz and Kijko, 1994). During moment tensor inversions based on observations in close proximity to the source, the seismic radiation in the near field should be taken into account (Niewiadomski, 1997).

McGarr (1992b) made a detailed study of a seismic event recorded in early 1988 on the surface and at the underground GEOS stations in the Vaal Reefs gold mine in South Africa. The seismograms indicated seismic moment tensors having substantial implosive components. The moment tensor was decomposed into isotropic and deviatoric components from which both the coseismic volumetric closure and the total shear deformation could be estimated. He found that they are proportional to each other, consistent with earlier analyses of the tabular mine stope interaction with the surrounding rock mass. McGarr (1992c) in another study of 10 mine tremors recorded by GEOS surface and underground stations in the Western
Deep Levels and Vaal Reefs gold mines found that the moment tensors fall into two separate categories. Seven of the events involved substantial coseismic volumetric reduction together with normal faulting, and for the remaining three events the volumetric closure was not significantly different from zero. They were largely double-couple sources involving normal faulting.

Inversions of a standard second-order moment tensor are routinely performed by the ISS system (e.g., Mendezki, 1993), although the results are not readily available in the professional literature. Occasionally, when volumetric changes in the source area are significant, the different time dependencies of different moment tensor components need to be considered. Niewiadomski (1997) described a procedure for the time-dependent moment tensor inversion for P waves. The procedure allows evaluation of the source duration and provides insight into the source complexity by calculating the spatial distribution of pressure and tension axes in time. It seems that this procedure is available within the ISS system.

A relative moment tensor inversion method, based on the theory described by Dahm (1996), was used for the first time by Andersen (1999) to study seismic events induced by mining. The same technique was used by Jost et al. (1998) to study the source mechanism of injection-induced events at 9-km depth at the KTB deep drilling site in Germany. The relative inversion method can be applied to clusters of seismic events from a small source area, provided that the spacing between the events is smaller than the dominant wavelength of the waves used in the inversion. This approach is based on the concept of a common ray path between the cluster and a given receiver. Most relative methods require knowledge of the radiation pattern of at least one reference event, which is used to estimate the wave propagation for events from the same source area. Dahm (1996) extended this method by including the estimation of several sources without a reference event; the path effects described by Green's functions are eliminated analytically. Andersen (1999) applied the relative method to three clusters of seismic events recorded at a depth of about 1900 m at a site at Blyvooruitzicht Gold mine in South Africa. Her input data were the peak displacement amplitudes of the P, SH, and SV waves. A total of 58 events with magnitude from −1.4 to 0 were studied. The dominant source mechanism for most of the events was found to be of the shear type. The fault-plane solutions correlated well with the fracture mapping of the site. Some larger events, however, showed a significant deviation from the double-couple mechanism, thought to be caused by the opening of shear cracks. All events displayed a small positive isotropic component between 7 and 16% of the overall mechanism, representing volumetric expansion.

The application of moment tensor inversion to study the mode of failure of small seismic events induced by drilling a tunnel in the Underground Research Laboratory was reported by Feignier and Young (1992). They studied 33 microseismic events in the moment magnitude range from −4 to −2. From the moment tensor inversion they obtained the ratio of isotropic to deviatoric components and found
a number of explosional and implosional sources. Furthermore, the location of events displaying extensional components corresponded to breakouts observed in the roof of the tunnel. In another study, Feignier and Young (1993) analyzed 37 microseismic events associated with the same tunnel excavation. Their hypocenters correlated with a zone of damage observed underground, which developed in the minimum principal stress direction. They found that as much as 75% of the mechanisms contained a significant negative isotropic component. These events were interpreted as closure of cracks that opened during earlier excavation stages.

Baker and Young (1997) applied a time-dependent moment tensor inversion to 20 seismic events recorded ahead of an advancing tunnel excavation at the Underground Research Laboratory. They found three characteristic types of seismic events there. The first group showed implosion/shear failure source mechanisms, associated with the observed collapse of excavation holes drilled at the side of the tunnel. The second group occurred in the region ahead of the face and showed tensile/shear failure mechanisms, interpreted in terms of crack initiation ahead of the face. The third group of events indicated predominant shear failure mechanisms throughout their rupture history, interpreted in terms of failure on preexisting crack surfaces. Šílený and Baker (1997) used data from the Underground Research Laboratory to test the performance of a method for fast estimation of focal mechanisms by inverting P and S wave amplitudes into moment tensors. The procedure provides an estimate of the confidence ellipses of the source mechanism, which show the errors in resolution of the individual moment tensor components and the errors in orientation of the deviatoric elements.

In Japan, the moment tensor inversion performed by Fujii and Sato (1990) on the observations from two small seismic events at Horonai coal mine has shown that they were non-double-couple events. The events were associated with longwall mining and were located in the vicinity of the longwall face. Fujii et al. (1997) proposed a new method for predicting rockbursts and seismicity in deep longwall coal mining. Fracturing intensity and its variation with respect to face advance were represented by the maximum shear seismic moment release rate, called by the authors the Seismic Moment Method. This method was successfully applied to three cases at Horonai and Miike coal mines, where mining works were carried out at depths from 900 to 1100 m and from 600 to 650 m, respectively.

The most extensive studies of the source mechanism of seismic events induced by mining, based on standard second-order moment tensor inversions, were undertaken in Poland, in both the coal and the copper mining districts. The procedure used is elaborated by Wiejacz (1991), based on the moment tensor inversion in the time domain, using as input data the first motion displacement amplitudes and signs of P waves and the amplitudes of SV waves recorded by the underground networks.

Wiejacz (1991, 1992) has studied the source mechanism of 60 small seismic events that occurred in 1990 and 1991 at Rudna copper mine in the Lubin mining
district. In general, the solutions that were well constrained by observations (good coverage of the focal sphere) had dominant shear components, although occasionally the isotropic component (showing both extension and contraction) could be as large as 25% of the overall mechanism. The CLVD component corresponded to uniaxial compression in all cases and was usually larger than the isotropic component. In another study of 77 events from Rudna and 38 events from Polkowice copper mines recorded in 1995, Wiejacz and Gibowicz (1997) confirmed that the moment tensor solutions are generally of double-couple type. The exceptions to this were 4 tensile events at Polkowice and 3 events at Polkowice and 1 event at Rudna with a pillar burst type component. These were the largest seismic events observed during the period of study. The mechanisms in a given area were seldom alike and they tended to form groups of a few similar events. This may be explained by the chamber and pillar mining technique used in Polish copper mines, which does not favor any specific direction in the absence of any important tectonic structures.

The most detailed study of the source mechanism of seismic events at Polkowice copper mine was carried out by Król (1998). He performed moment tensor inversion, based on observations from 25 seismic sensors, for 246 events with moment magnitude from 1.2 to 3.1, which occurred in 1997. To test the procedure adopted from Wiejacz (1991), Król (1998) additionally computed moment tensor of 20 calibration shots fired periodically in the mine. The solutions show in all cases significant isotropic explosive components in the range from 16 to 48% of the total mechanism and small double-couple components. Two types of seismic events are well known in the Polkowice mine: those radiating dilatational first arrivals only on all the sensors and those with both dilatational and compressional first arrivals. In consequence, the general moment tensor solutions of studied events show in some cases large non-double-couple components with a dominant CLVD component, although these solutions are often not stable. The deviatoric solutions are well constrained and are similar to double-couple solutions. These solutions in turn show a good correlation with geological and mining structures in the area. It is interesting to note that the double-couple component is the highest in the source mechanism of the events located ahead of the mining face and the lowest behind the working face. Similarly, the constrained double-couple solutions corresponding to reverse faulting are relevant to about half of the events situated ahead of mining, whereas they characterize over 80% of the events located behind the face.

The source mechanisms of 55 seismic events from Wujek and 83 events from Ziemowit coal mines in the Upper Silesian Coal Basin, Poland, were studied by Gibowicz and Wiejacz (1994). The events occurred in 1993 and had moment magnitudes from 1.1 to 2.4. The general, deviatoric, and double-couple moment tensor solutions for 20 events from Wujek mine are shown, as an example, in Fig. 26. In most cases the double-couple component was dominant, forming above 60% of the general solution. The double-couple component in the deviatoric
Fig. 26. General, deviatoric, and double-couple moment tensor solutions displayed on lower hemisphere equal-area projections, for 20 selected events recorded in 1993 at Wujek coal mine, Poland. The areas of compression are shaded, and the event numbers are given at the left-hand side of the tensor solutions. [Reprinted from Gibowicz and Wiejacz (1994), Fig. 3.]
solution was as large as 80% of the mechanism. Further studies of the source mechanism of seismic events at Wujek and Ziemowit mines and their relationships to geological and mining structures were undertaken by Wiejacz (1995a,b) and Wiejacz and Lugowski (1997). Wiejacz (1995b) studied 58 events from Ziemowit mine, which occurred in 1994 and were associated with a longwall. He found that the effect of mining structures on source mechanisms is not very clear. The events split into two groups of different nature, one connected with the passing of the working face underneath the transport tunnel and the other with approaching the boundary of the extracted area in another seam. The mechanism of the first group of events was almost pure double-couple dip slips on near vertical fault planes, but with various orientations. The mechanisms of the second group had a larger non-shear component, and were often similar to the cavity collapse type mechanism.

In contrast, at Wujek mine the source mechanism correlates quite well with mining and geological structures. Wiejacz and Lugowski (1997) made a comparative study of seismic events induced in three selected regions of the mine. Study of 58 events associated with a longwall in an area distant from the existing faults has shown that the mechanisms there have near vertical nodal planes parallel to the side boundaries of the longwall area. In a region of a longwall close to the major fault, where 68 events were analyzed, the mechanisms were found to reflect the fault direction. The direction of mining and longwall front seem to play a secondary role. Similarly, the study of 25 events located in the vicinity of another large fault indicates that its direction is reflected in the mechanism of these events (see Fig. 22).

Sagan et al. (1996) performed moment tensor inversion of some 200 events with magnitude greater than 1.2, which occurred in 1993 at Wujek coal mine. A comparison of source mechanisms with geological features was the main purpose of the study; the Wujek mine area is situated in the region of one of the main fault discontinuities in the Upper Silesian Coal Basin. Once again, the double-couple component was dominant in most events, although a set of events with a large isotropic component was also observed. The spatial orientation of the nodal planes indicates a group of seismic events that might have their origin in unstable energy release on fault surfaces forming a regional structural pattern. Thus, stresses of tectonic origin cannot be ignored as an important factor in the generation of seismic events in the Upper Silesian Coal Basin. The results of another study of the source mechanism of 372 events from the same area (Idziak et al., 1997) led to similar conclusions. For selected sets of events with similar orientations of nodal planes, located in a limited area, the resultant stress tensor for composite mechanism was determined. The results of analysis point to a relationship between the spatial distribution of slip vectors in the seismic sources and the pattern of neotectonic deformation of the Carboniferous rock mass in the Upper Silesian Coal Basin.
5.2. Typical Source Mechanism and Complex Rock Failures

From numerous studies, based on moment tensor inversion, of the source mechanism of seismic events in mines it follows that often, but not always, shearing processes are dominant in the source and its mechanism can be well represented by a double couple, conventionally called the focal mechanism. Mining of a highly stressed remnant in a deep South African gold mine was accompanied by considerable seismic activity, and several shear ruptures corresponding to the larger seismic events were encountered during mining operations. Ortlepp (1992) performed a study of microcataclastic particles from underground shear ruptures and found that certain striking features revealed by a scanning microscopic study of the fresh cataclastic "rock-flour" forming part of the filling of these ruptures provide strong evidence of violent "shock-rebound" phenomena in the faulting process. He concluded that the formation of a fault by shearing through massive strong rock is a more violent and a less homogeneous process than was thought possible.

McGarr and Bicknell (1990) have demonstrated a very efficient technique for investigating the focal mechanism by matching synthetic to real seismograms, recorded by a broadband, wide-dynamic range digital system. They investigated the focal mechanisms of two tremors located in the Vaal Reefs gold mine, near Klerksdorp, South Africa, and used them to relate these events to the geological and mining conditions in the area.

First-motion analyses of seismic records and an assessment of damage provided conclusive evidence that a fault-slip source mechanism was responsible for a magnitude 2 rockburst that occurred at the Lucky Friday mine, Idaho, in 1990 (Williams et al., 1992). Movement along an argillite bed was observed, and the physical evidence of a shear-slip type failure confirmed the validity of a double-couple source model for many situations in mines.

A detailed study of focal mechanisms of seismic events in relation to results from numerical stress modeling and underground observations at Ansil mine, Canada, was performed by McCreary et al. (1993). They found that events with a purely shear mode of failure were prevalent in the footwall along geological contacts and the backs of development drifts. The events without double-couple solutions were scattered throughout the footwall, and were not associated with geological features. Tensile activity was associated with the greatest damage observed underground.

Urbancic et al. (1993a) derived fault-plane solutions for microseismic events, with magnitude less than zero, at two geologically different mine sites in Canada. They made a comparison of the focal mechanisms and the results of stress inversion and principal component analysis of microseismicity with in situ stress measurements and structural mapping. The obtained results suggest that microseismicity generally occurs along the most significant mapped fractures at the sites; the fault-plane solutions are consistent with in situ measurements of stresses; and fault planes lie in close proximity to the most significant mapped features at the sites.
In another study, Trifu and Urbancic (1996) used microseismic events, recorded within a volume at depth at Strathcona mine, to evaluate failure conditions associated with a magnitude 2.9 rockburst. A good correlation was found between stress inversion and microseismic failure planes and structural mapping, and between the in situ principal stress orientations and those obtained by stress inversion and numerical modeling.

Dubifiski and Stec (1996) calculated the regional stress tensor from fault-plane solutions of seismic events observed at Szombierki and Wujek coal mines in Upper Silesia, Poland, and found a distinct correlation of the principal stresses with local mining conditions at these mines. Similarly, Vesela (1995) calculated the double-couple focal mechanism for over 100 seismic events at Lazy mine in the Ostrava-Karvina coal mining district, Czech Republic, and found different types of the mechanism caused by changes in the stress state of rock mass, connected with coal face advances.

Although the double-couple focal mechanism seems to be dominant in underground mines, more complex rock failures associated with seismic events are also reported from various mining districts. Stickney and Sprenke (1993) studied 21 events in the Coeur d'Alene mining district of northern Idaho, where rockbursts are a major hazard to mining. Eleven of the events generated all-dilatational first motions. Seven events were found consistent with an implosional source and two others were consistent with a shear-implosional source. These results provide observational evidence that dilatational first motions are predominant in this deep metal-mining district.

The magnitude 3.5 seismic event of 1981 in the Gentry Mountain coal mining district, Utah, was studied by Taylor (1994). He modeled the event, using a simple source representation, as a tabular excavation collapse that occurred as a result of mining activity. In contrast, the magnitude 3.6 seismic event of 1993 in the Book Cliffs coal mining district, Utah, which coincided with the crushing out of 24 pillars encompassing an area of nearly 15,000 m², was found by Boler et al. (1997) to be of a different source mechanism. They examined six scenarios for the event mechanism for consistency with the seismic analysis and on-site damage observations, and concluded that this catastrophic rockburst was precipitated by a subsidence-related fault slip, with normal slip on a fault in close proximity to the mining works.

A controlled collapse experiment conducted at 320 m depth at the White Pine copper mine near Lake Superior, Michigan, was described by Yang et al. (1998). A rectangular panel with an area of about 20,000 m² and a room height of 3 m was designed for the experiment. A total of 59,000 kg of explosives were placed in the 72 supporting pillars and detonated. The pillars were destroyed and the roof collapsed almost immediately. The source characteristics were investigated by waveform forward modeling and time-dependent moment tensor inversion. The results indicate that the source mechanism of the collapse can be represented by a
horizontal opening and closing crack. Thus, a unique source characteristic of the
induced collapse is that, unlike spontaneous collapses, the induced collapse was
initiated as a tensile crack. The collapse magnitude was estimated to be 3.1 from
coda duration and 2.8 from surface waves recorded at regional distances (Phillips
et al., 1999). Over 4000 aftershocks were recorded. The rate of occurrence followed
the modified Omori law with an exponential of 1.3. Phillips et al. (1999) identified
41 shear slip, 18 implosional, and 2 tensile events, based on P-wave polarity
patterns. They also found that stress drops for implosional events were lower than
those for shear slip events.

Four mechanisms of pillar failure have been observed in strip pillars at Western
Deep Levels gold mine, South Africa (Lenhardt and Hagan, 1992). Relatively
narrow pillars crush in some cases, whereas the solid core of wide pillars, if
sufficiently stressed, slips or punches along one or two shear planes below the up-
and down-dip edges of the pillar. Occasionally, shearing occurs simultaneously on
two adjacent pillars. Closure occurs in the surrounding stopes, and serious damage
to footwall development close to the pillars might result.

The violent failure of a peninsular remnant at a depth of 2300 m in a mine in the
Carletonville Goldfield of South Africa, which occurred in 1994, was described
by Durrheim et al. (1998). A seismic event with magnitude of 2.1 resulted from
the failure of the remnant with movement into the workings. The event could not
be explained by a single shear slip. Two different rockburst damage mechanisms
were identified. Firstly, the face and footwall on the eastern side of the remnant
were violently ejected into the void between the original face and first line of timber
packs, following failure and dilation of the remnant and its foundation. Secondly,
the hangingwall on the southern side of the remnant fragmented and collapsed
when subjected to intense seismic shaking.

The mechanism of a partial collapse of dolomite-mine workings near Schwaz,
Austria, which occurred in May 1993, was described by Lenhardt and Pascher
(1996). The collapse was associated with a sequence of more than 200 seismic
events with the largest event of magnitude 1.9. Underground observations revealed
a number of discontinuities along which the rock mass was able to move. Seismic
records of this series indicate two types of focal mechanism involved during the
collapse. Some events seem to have been associated with pure block sliding along
several discontinuities and the others were collapse events. Several phases of the
collapse were also recognized from seismic observations.

The interaction of plane seismic waves with underground openings was studied
by Yi and Kaiser (1993), who applied the dynamics and continuum mechanics an-
alytical approaches. They found that low-frequency waves from a seismic source
hundreds of meters away could cause both rock ejection and dynamic stress con-
centration at both walls of an opening in underground mines. These dynamic
stress concentrations, superimposed on high static stress concentrations, may lead
to different types of rock failure, including secondary rockbursts.
A study of source parameters of seismic events at a coal mine in the Ruhr basin, Germany, and those induced by the excavation of a shaft in granite at the Underground Research Laboratory, provided evidence for non-shearing events (Gibowicz et al., 1990, 1991). It has been found that the ratio of $S$- to $P$-wave energy is often unusually low, and the high $P$-wave energy events are thought to be the most likely candidates for non-double-couple events. A direct comparison between the isotropic, compensated linear vector dipole and double-couple parts of the moment tensor and the energy ratio, determined for over 200 seismic events from Polish coal and copper mines, has shown that these quantities are correlated (Gibowicz, 1996). Though the correlation coefficients were small in all cases, not exceeding 0.5, they were significant at the 99% confidence level.

For small microseismic events, the source size provided by shear models is often unrealistically large when compared to visual observations of rock fractures near underground openings. A depletion of $S$-wave energy for events close to the excavations, on the other hand, implies that tensile cracking might be the dominant mechanism for these events. Cai et al. (1998) proposed a method to estimate the fracture size from seismic measurements. The method accepts tensile cracking as the dominant fracture mechanism for brittle rocks under compressive loads and relates the fracture size to the measured seismic energy, leading to more realistic values of the size when compared to field observations.

5.3. Stress Release Mechanism

In the well-known simple fault model of Orowan (1960) with constant friction, the final stress after the earthquake rupture is equal to the dynamic frictional stress. Two other possibilities have been considered. In the first case, the final stress is greater than the frictional stress and a partial stress drop mechanism is involved (Brune, 1970, 1976). In the second case, the final stress is smaller than the frictional stress and a frictional overshoot mechanism is considered (Savage and Wood, 1971). Several arguments in favor of both cases have been proposed (e.g., Savage and Wood, 1971; Brune, 1976; Madariaga, 1976; Smith et al., 1991; Zúñiga, 1993).

Gibowicz (1998) recently examined the stress drop mechanism of seismic events induced by mining. He collected the values of uniformly estimated apparent stress and stress drop of Brune (1970, 1971), taken as a measure of static stress drop, from 850 events with moment magnitude ranging from $-3.6$ to 3.6, induced at the Underground Research Laboratory (URL), Western Deep Levels (WDL) gold mine in South Africa, and two coal and copper mines in Poland. The quantity epsilon $\varepsilon$, defined as the ratio of the static stress drop $\Delta \sigma$ over the sum of apparent stress $\sigma_a$ and half the static stress drop, $\varepsilon = \Delta \sigma / (\sigma_a + \Delta \sigma / 2)$, proposed by Zúñiga (1993) as an indicator of stress-drop mechanism, was used to study the stress release mode
in various mining environments. This ratio is bounded by the values from 0 to 2. The Orowan condition is met when $\varepsilon = 1$ or $\sigma_a = \Delta \sigma / 2$, the partial stress drop mechanism corresponds to $\varepsilon < 1$, and the frictional overshoot case appears when $\varepsilon > 1$.

The values of apparent stress versus stress drop for various seismic events are presented in Fig. 27, reproduced from Gibowicz (1998). The straight line indicates Orowan’s condition corresponding to apparent stress equal to half the static stress drop. From this figure it follows that all the events from the WDL clearly display a frictional overshoot mechanism; the events from the URL indicate a partial stress drop mechanism; and all the events from Polish mines fulfill Orowan’s condition. The values of epsilon (Zúñiga, 1993) versus moment magnitude for seismic events from six selected mines are presented in Fig. 28 (Gibowicz, 1998), where Orowan’s condition ($\varepsilon = 1$) is marked by a continuous line.
The apparent stress defined as the ratio of the radiated energy over the seismic moment, multiplied by the medium rigidity, is independent of any source model assumptions. In contrast, the static stress drop determined from the seismic moment and corner frequency is heavily model dependent through the source radius–corner frequency relation. For the events from the URL, Orowan’s condition would be fulfilled if the constant in the source dimension–corner frequency relation is equal to 1.82, which is the middle value between the average value of 1.32 corresponding to $S$ waves in the source model of Madariaga (1976) and that of Brune (1970). The constant in the Madariaga model, however, depends on the rupture velocity and the azimuth of observation, and if the obtained value of 1.82 could be explained in these terms, the partial stress drop mechanism there would be irrelevant. For the events from the WDL, on the other hand, the obtained value of 3.92 of the same constant is too high for a reasonable explanation in terms of fault and observation geometry or rupture velocity, implying that the appearance of frictional overshoot mechanism there could be real (Gibowicz, 1998). It should be noted, however, that
comparisons between apparent stress and stress drop are somewhat controversial because of the stress drop dependence on the choice of source model.

The apparent stress is also defined as \( \sigma_a = \eta \sigma \), where \( \sigma \) is the average shear stress loading the fault plane leading to slip, and \( \eta \) is the seismic efficiency defined as \( E_a/W \) in which \( E_a \) is the seismic energy and \( W \) is the total energy released by the earthquake (e.g., McGarr, 1999). McGarr (1994) compared apparent stresses of mining-induced seismic events in South Africa and Canada with those measured for laboratory stick-slip friction events. He found that both mining-induced and laboratory events have very low seismic efficiency; nearly all of the energy released by faulting is consumed in overcoming friction. The comparison led McGarr (1994) to the important hypothesis that the seismic efficiency \( \eta = \sigma_a/\sigma \leq 0.06 \).

In his most recent study McGarr (1999) tested this hypothesis against a substantially augmented data set of earthquakes for which \( \sigma \) can be estimated, mostly from in situ stress measurements. The expanded data set covered a broad range of depths, rock types, pore pressures, and tectonic settings. He found that over about 14 orders of magnitude in seismic moment, apparent stresses display distinct upper bounds determined by a maximum seismic efficiency of about 0.06, consistent with his hypothesis. The relationship between the maximum apparent stress and the average shear stress is shown on a logarithmic scale in Fig. 29, reproduced from McGarr (1999), for several laboratory stick-slip friction events as well as for induced, triggered, and natural earthquakes. McGarr (1999) also found that the behavior of \( \sigma_a \) and \( \eta \) can be expressed in terms of the ratio of the static to the dynamic coefficient of friction and the fault slip overshoot, measured for stick-slip friction events in the laboratory. Typical values of these two parameters correspond to seismic efficiencies of about 0.06. Thus upper bounds on \( \sigma_a/\sigma \) appear to be controlled by just a few fundamental aspects of frictional stick-slip behavior that are common to shallow earthquakes everywhere, and the values of apparent stress can be used to deduce the absolute level of deviatoric stress at the hypocenter (McGarr, 1994, 1999).

6. Source Time Function

The source processes of seismic events induced by mining have been extensively studied by spectral analyses of the records of seismic waves and by the moment tensor inversion technique. In spite of undertaken efforts, a number of problems remain unsolved and new approaches to study the seismic sources in mines are needed for a better understanding of their properties and their mechanism of generation. Determination of the source time function is a new approach in studies of seismicity in mines, although this approach has been in use for a long time in studies of natural earthquakes.
The Source Time Function (STF) of a seismic event describes the energy release and the rupture evolution in the source. It could provide important information on rupture velocity and rupture direction when directivity effects are observed. Source parameters can be estimated in the time domain and compared with those estimated in the frequency domain. Furthermore, source tomography, based on Relative Source Time Functions (RSTF) observed at a number of stations, would provide a detailed description of the rupture processes in the source of a given event.

The extraction of STF from seismograms requires, however, separation of the source effects from those of the path, site, and instrumental response. For this, a deconvolution technique based on the Empirical Green's Function (EGF) can be used. Knowledge of the earth's crust structure, attenuation effects, and instrumental
response are then not required (e.g., Mueller, 1985; Li and Thurber, 1988; Mori
and Frankel, 1990; Mori, 1993; Li et al., 1995; Bour and Cara, 1997; Fletcher
and Spudich, 1998; Jost et al., 1998). For two seismic events of different strength,
located close to each other and having similar focal mechanisms, the record of the
smaller event can be considered as the EGF (Hartzell, 1978). It can be deconvolved
from the record of the larger event to obtain a RSTF at a given station for the larger
event (Mueller, 1985). Such deconvolution should result in a RSTF that is corrected
for the path, site, and instrumental effects.

This technique is especially convenient to study seismic sources in mines, where
underground seismic networks are situated in the source area and are often com-
posed of large number of sensors. Furthermore, seismic events are located with
high accuracy with standard errors not exceeding a few tens of meters. Domafiński
and Gibowicz (1999) and Domafiński et al. (2000) described the first attempts to
use the EGF technique to recover the STF of large seismic events observed at
Polkowice and Rudna copper mines in Poland. They applied the classical spectral
division deconvolution technique, developed by Mueller (1985), to separate the
STF from the records. The observed seismogram of a given larger event can be
expressed as the convolution of its STF, the impulse response of the path, recording
site, and recording instrument. The seismogram of a smaller event, whose STF is
assumed to be a delta function, is considered as an EGF. The deconvolution is
performed by spectral division in the frequency domain. Then, after taking the
inverse Fourier transform, the STF of the larger event is obtained.

The two events need careful selection. They should be closely located to each
other. The difference in distance between the two events in a mine should not exceed
some 100–200 m (Domafiński and Gibowicz, 1999). Their focal mechanisms should
be similar and the difference between seismic moment tensor axes should not
exceed 20 degrees. Finally, the magnitude of the main event should be considerably
greater than that of the smaller event. The difference of about one magnitude unit
provides reasonable results, although smaller differences are often acceptable.

The notion of the empirical Green’s function deconvolution approach may be
rather simple, but its practical applications are complex. There are numerical in-
stabilities resulting from spectral division and several other problems inherent in
the method. An example of the RSTF deconvoluted from P waves records of the
moment magnitude 2.7 event of October 27, 1996, from Rudna copper mine, using
the records of a magnitude 1.9 event as a Green’s function, is shown in Fig. 30,
reproduced from Domafiński et al. (2000).

After retrieving the RSTF pulses from the records of several stations, it is possi-
able to estimate whether they are symmetric or asymmetric, and in the latter case to
consider their dependence on the station azimuth. If the pulse widths of the RSTF
depend on the station azimuth, then it is an indication that the rupture propagated
unilaterally. For a unilaterally propagating rupture, the source duration is narrow-
est in the direction of rupture propagation and widest in the opposite direction.
Similarly, the maximum amplitude of the source pulse would be largest in the
direction of rupture propagation and smallest in the opposite direction. The spatial
distribution of pulse widths of the RSTF derived at various seismic stations for the
event of October 27, 1996, from the Rudna mine is shown in Fig. 31, reproduced
from Domaniński et al. (2000).

The rupture direction, rupture velocity, and source size can be determined si-
multaneously by a least-squares method either from the distribution of the RSTF
pulse widths (e.g., Li and Thurber, 1988) or from the distribution of their maximum
amplitude (Li et al., 1995). For a finite moving source, the relation between the
pulse width and the azimuth angle measured to the propagation direction is well
known (e.g., Ben-Menahem, 1962). The rupture direction is estimated by fitting
the pulse duration at stations with different azimuths to a best straight line. Then
the slope and intercept of this straight line are used to estimate the rupture velocity
and the fault length. An example of such a procedure is shown in Fig. 32, repro-
duced from Domaniński et al. (2000), for the same event of October 27, 1996, from
the Rudna mine. The surface of the RSTF pulse is a measure of seismic moment
in the time domain. More precisely, this surface is a ratio of seismic moment of
the main event over seismic moment of the smaller Green’s event.

Recently, new methods for determination of source time function by the em-
pirical Green’s function approach in the time domain were proposed. They are
based on the search by successive iterations for the solution of integral equations
describing the observed seismograms as the convolution of the source time func-
tion with the empirical Green’s function. An interesting and effective technique,
in this respect, seems to be the projected Landweber method which is an iterative
nonlinear technique allowing the introduction of physical constraints (positivity,
causality, etc.) on the final STF (Bertero et al., 1997; Piana and Bertero, 1997). A
nontrivial modification of an iterative blind-deconvolution method used for image
identification was also proposed as a novel technique to retrieve the STF (Bertero
et al., 1998). The main feature of this method, which is also based on the projected
Landweber method, is the use of different constraints on the STF and the EGF.
The convergence of the method is very fast and the results obtained by Bertero
et al. (1997) for synthetic and real data were quite satisfactory.

7. Fractals

Fractal geometry is a mathematical tool (Mandelbrot, 1982) to characterize self-
similar sets, which are essentially the same regardless of the scale on which the
set is examined. A fractal approach to natural earthquakes occurrence has been
widely applied in recent studies and it has been shown that seismicity exhibits
approximate fractal structure with respect to space, time, and size (e.g., Barton
and La Pointe, 1995).
Fig. 30. Example of the relative source time function derived from P waves, recorded by the underground seismic network at Rudna copper mine, Poland, from the moment magnitude 2.7 event of October 27, 1996, using the record of a magnitude 1.9 event as a Green's function. P-wave records (a) and selected windows (b) for larger and smaller events are shown at the top. The original, smoothed, and windowed source time functions (c) and the corresponding results of the convolution of source time functions with the record of Green's event (d) are presented at the bottom of the figure. [Domarski et al. (2000), Fig. 5, reprinted with kind permission from Birkhäuser Verlag AG.]
While the self-similarity of energy of seismic events in mines described by the Gutenberg-Richter relation has been known for a long time, Coughlin and Kranz (1991) showed for the first time that distributions in space and time of these events have also a fractal character. They analyzed three microseismicity data sets from Galena mine, USA, from the time periods encompassing three damaging rockbursts. The spatial and temporal fractal dimensions were estimated by the correlation integral method (Grassberger and Procaccia, 1983). For a set of \( N \) points parameterized with a variable \( \bar{x} \), the correlation integral is given by

\[
C^{(2)}(r) = \frac{1}{N(N - 1)} \sum_{i=1}^{N} \sum_{k \neq i} \Theta \left( r - \| \bar{x}_i - \bar{x}_k \| \right),
\]

where \( \Theta \) is the Heaviside step function.
where $\| \tilde{x}_i - \tilde{x}_k \|$ is the distance between points $i$ and $k$ in the domain of $\tilde{x}$, $r$ is the reference distance, and $\Theta(\cdot)$ is the Heaviside function, which in this case counts how many pairs of points ($\tilde{x}_i, \tilde{x}_k$) fall within the interevent distance $r$. If the set exhibits simple invariance, then the correlation integral is proportional to the power function of $r$

$$C^{(2)}(r) \propto r^{D_2}. \quad (2)$$

Constant $D_2$, called the correlation exponent dimension or the correlation dimension, is estimated from a linear part of $\log C^{(2)}(r)$ vs $\log r$ empirical relation.

Coughlin and Kranz (1991) considered the time of event occurrence and the location of event as separate parameterizations. In every case they found a reasonable range of $r$ over which relation (2) approximately held. The spatial correlation dimension was significantly less than 3.0—an expected result for points uniformly
distributed in a three-dimensional space. A time-variability of spatial $D_2$ in some concordance with the occurrence of stronger events suggested a possible use of this parameter for monitoring of fracturing processes in mining stopes and, in particular, its late stages before an impending main failure.

Since this first study, many works explored fractal properties of seismic events from mines in space, time, and energy. The fractal dimension is a one-parameter characterization of a degree of clustering of objects building a fractal set, thus it can be used to study geometrical similarities and differences of seismically active areas as well as development of a seismogenic process in time and space. The latter possibility, in particular, gave rise to various attempts to use fractal characteristics of seismic data from mines as precursors to strong damaging events. The works concerning this subject, specific for the seismicity induced by mining, are reviewed in the next chapter together with other methods designed to evaluate the time-dependent seismic hazard.

Xie and Pariseau (1992) and Xie (1993) studied the spatial fractality of microseismic activity of a rockburst prone pillar at Galena mine. They applied the so-called number-radius or mass-radius method introduced by Xie (1993). In this method a fractal set of geometrical points (e.g., hypocenters or epicenters) should obey the law of proportionality:

$$M(r) \propto r^{D_{cl}},$$

where $M(r)$ is the number of points inside the hypersphere of radius $r$, centered at the center of the mass of the set. The constant $D_{cl}$ is an estimate of the fractal dimension of the set and is called the clustering dimension. Since the position of the center of mass can be significantly biased by mislocations of outlying events, and this bias will propagate to a clustering dimension estimate, it is recommended to use an average $D_{cl}$, obtained from several independent estimations centered at points randomly distributed around the evaluated center of mass.

A good linear correlation between $\log M(r)$ and $\log r$ obtained by Xie and Pariseau (1992) evidenced statistical self-similarity of the studied data. Xie (1993) interpreted an observed increase of the degree of clustering before a rockburst, represented by a decrease of $D_{cl}$, in terms of clustering of local cracks in the rock mass. During a late preparatory stage the crack density within a zone of the future main failure becomes significantly greater than the background crack density. This clustering implies the decrease of fractal dimension of spatial distribution of events. Xie (1993) concluded that it could be possible to predict the occurrence of main rockbursts by observing a decrease in $D_{cl}$ for spatial distribution of microseismicity.

Fractal character of spatial distribution of seismic events in mines has been confirmed for practically every scale. Trifu et al. (1993) studied distribution of some 1500 microseismic events recorded within a selected volume at Strathcona mine in Canada. The magnitude range of the data was $[-2, 1]$. The spatial correlation
dimension, evaluated for two perpendicular directions, had an anomalous decrease before the occurrence of $M = 2.9$ rockburst and by the end of the aftershock sequence.

Mortimer and Lasocki (1996a, b) investigated local seismicity connected directly with longwall mining of single, isolated stopes. Such seismicity forms a cloud surrounding the longwall front and move together with advances of the front (Gibowicz and Kijko, 1994). For every considered seismic series the epicenters of events were related to a local moving system of coordinates centered at the center of the mining front. The series comprised all events located within the $400 \times 400$ m square around the center of the system of coordinates. In order to estimate the fractal dimension of epicenter distribution, Mortimer and Lasocki (1996a) used the box-counting method (e.g., Mandelbrot, 1982; Turcotte, 1992). In this method a volume containing studied objects (points in the case of spatial distribution of seismic sources) is covered by hypercubes (boxes) with a side length $r$. If the set of objects is a fractal, then the number, $N(r)$, of boxes covering the objects scales with $r$

$$N(r) \propto r^{-D_b},$$

and a constant $D_b$ is the box-dimension. The box-dimension, obtained for a finite sample, estimates the Kolmogorov capacity dimension of the fractal, which is the limit of $D_b$ for $r \to 0$.

All nine analyzed series from five coal mines in Poland statistically obeyed relation (4) for the range of $r$ starting from 20 m—the order of source location errors. The values of $D_b$ varied from series to series between 1.0 and 1.7. An interesting observation was made that the fractal character was preserved in sub-series comprising only smaller and larger events, but the fractal dimension values were different in such sub-series, and generally lower for strong events. This could suggest that, contrary to the conclusions inferred from natural seismicity (e.g., Kagan, 1994), larger events in this particular set-up of moving mining front were more strongly clustered than the smaller ones. However, due to a strong bias of the box-dimension, this result could not be considered as fully reliable. Mortimer and Lasocki (1996a) found that reliable estimates of $D_b$ could be obtained for series comprising no less than 1500 events. It was concluded, therefore, that the box-counting method is of limited usefulness for studies of seismic series in mines.

Mortimer and Lasocki (1996b) compared the results of three methods of fractal analysis: the box-counting, the number-radius, and the correlation integral, applied to local seismicity data from one longwall at Katowice coal mine in Poland. Regardless of the method used, the spatial distribution of events exhibited fractality within the distance range from 20 to 100 m. The analysis, repeated on successive samples of a constant number of events, showed that fractal dimension estimates significantly varied in time. The exponential correlation dimension was the most stable, whereas the other two dimensions frequently exposed saw-like variations.
The authors concluded that the correlation integral method is the most useful when the number of data is limited. Furthermore, it was also shown that the distribution of event origin times is fractal.

Domafiski (1997) compared the fractal correlation dimension of spatial distributions of stronger events from coal and copper mines in Poland. Magnitudes of seismic events from coal mines were in the range of 0.65 to 2.2. The events from copper mines were on the average larger, with the magnitude range of 1.1 to 3.1. The correlation dimensions were different for the data sets from different kinds of mines and similar for the sets from the same kind of mines; from 1.5 to 1.75 for coal and from 2.0 to 2.3 for copper mines. The author gave no suggestion for a possible cause of the significant difference of spatial clustering of events in coal and copper mines, although a potential of fractal analysis to identify differences between seismically active zones in mines was admitted.

Gibowicz (1997a) estimated the spatial correlation dimension to describe an unusual swarm-like seismic sequence, which occurred during 12 days in April 1993 at the Western Deep Levels gold mine in South Africa. The sequence consisted of four main shock–aftershocks sequences following one another within a volume of rock of $670 \times 630 \times 390 \text{ m}^3$. The identified main shock–aftershocks sequences had different correlation dimensions starting from 1.75 for the first and the most clustered one up to 2.4 for the second and the least clustered. The last two sequences had nearly the same correlation dimension of about 1.9. The author explained the differences by a limited time for the development of sequences. Only the first sequence had enough time to develop fully and to reach its “natural” clustering. The developments of the second and the third were interrupted by the successive main events.

Idziak and Teper (1996b) studied a spatial distribution of epicenters of the largest events observed in the Bytom syncline, one of the main structural units forming the Upper Silesian Coal Basin in Poland. The syncline of approximate size of $16 \times 8 \text{ km}^2$ is intensively mined and is highly seismic. Most of the seismic events are directly connected with mining operations, but the largest events seem to be correlated with local geological discontinuities rather than with mine openings (Gibowicz and Kijko, 1994). The analyzed data comprised 1651 events of local magnitude above 2.2, recorded from 1977 to 1991. Using the box-counting method, Idziak and Teper (1996b) found that the distribution of epicenters of studied events was fractal within the distance range of 0.5 to 8 km. The value of the box-dimension was 1.52. The same box counting technique was applied to investigate the geometry of the fault system of the Bytom syncline. The fractal dimension in this case was estimated as 1.60. The authors suggested that the close proximity of these two values might imply the engagement of the whole fault system in the generation of large events in the area.

A one-parameter characterization of fractal properties of seismicity in mines seems to be insufficient to represent complex features of seismic event distributions.
When a fractal set is not homogeneously scale invariant, it should be described by an infinite number of dimensions rather than by one fractal dimension alone (e.g., Mandelbrot, 1989).

The most widely used method to investigate multifractal properties of natural earthquakes employs estimation of the generalized correlation integral (e.g., Hirata and Imoto, 1991; Hirabayashi et al., 1992; Godano et al., 1996; Wang and Lee, 1996). Given a set of \( N \) points, parametrized by \( \mathbf{x} \), estimate of the generalized correlation integral is given by

\[
C^{(q)}(r) = \left\{ \frac{1}{N} \sum_{i=1}^{N} \left( \frac{1}{N-1} \sum_{k \neq i} \Theta \left( r - \| \mathbf{x}_i - \mathbf{x}_k \| \right) \right)^{q-1} \right\}^{\frac{1}{q-1}}, \tag{5}
\]

where the parameter \( q \) can take any value from \((-\infty, 1) \cup (1, +\infty)\) For complex self-similar sets the correlation integral scales with \( r^{D_q} \),

\[
C^{(q)}(r) \propto r^{D_q}, \tag{6}
\]

where \( D_q \) for various \( q \) is the multifractal spectrum of the set (e.g., Grassberger, 1983; Hentschel and Procaccia, 1983; Kurths and Herzel, 1987). In a region of low \( q \) the dimensions follow the relation \( D_{q+2} = D_1 - \mu(q + 1) \), which can be used to estimate \( D_1 \) (Paladin and Vulpiani, 1987). For a homogeneous fractal set all \( D_q \) are the same. For a heterogeneous fractal set, when \( q \) changes from positive to negative values, \( D_q \) characterizes clustering of the set from the most to the least intensive, respectively. The limits of a multifractal spectrum, \( D_{\infty} \) and \( D_{-\infty} \) are finite and characteristic for the studied set. In practice \( D_q \) does not change significantly for \( q \) above 10 to 20. It is clear from relation (5) that an estimation of the fractal dimensions for \( q < 1 \) is unstable, which often precludes a reliable interpretation of its results.

Shivakumar et al. (1996) applied multifractal analysis to study the spatial distribution of three major area rockbursts at Champion reef mines in the Kolar Gold Fields in India. Three data sets consisted of 125 seismic events that occurred for 14 days after a \( M = 2.48 \) main event; 64 events recorded during 1 week after a \( M = 2.44 \) rockbursts; and 66 events for 15 days following a \( M = 3.09 \) main event, respectively. For \( q \) greater than 1, a linear part of the logarithm of relation (6) was easy to identify. Figure 33, reproduced from this work, shows linear fits to the actual data, leading to the evaluation of generalized correlation dimensions in the range of \( q \) from 2 to 10 for the first data set. Shivakumar et al. (1996) analyzed \( D_2 \), considered \( D_{10} \) as \( D_\infty \), and analyzed the difference between these two values, which was supposed to characterize fractal heterogeneity of the studied data sets. The differences of \( D_2 - D_{10} \) were 0.52, 0.37, and 0.41 for successive data sets, respectively. The authors related these clustering features to the state of stress.
distribution, mining, and regional geological heterogeneity of the area, concluding that multifractal analysis can successfully distinguish individual features of factors governing the generation of induced seismicity.

The other application of multifractal analysis to seismic data from mines was presented by Mortimer et al. (1999) who used the generalized correlation dimension of the distribution of event energy to describe rock fracturing at different scales. The analyzed data sets comprised larger mining-induced events ($M > 2.2$), regionally distributed over the Upper Silesian Coal Basin in Poland, seismic events from a selected part of a gold mine in South Africa, local seismicity connected with mining of individual stopes in Polish coal mines, and acoustic emission of rock samples, recorded during laboratory tests. In all cases, the distribution of energy was fractal, but the multifractal spectra did not seem to be related to the spatial scale of the fracturing process. Figures 34 and 35, reproduced from Mortimer et al. (1999), show the spectra of regional seismic data from two active areas and of acoustic emission induced in coal samples by gas sorption and generated in a dolomite sample subjected to uniaxial compression, respectively. The multifractal
characteristics of the acoustic emission data are highly different, whereas the spectrum of the regional seismicity nearly matches the spectrum obtained for the second coal sample. The other interesting observation is a close correspondence between the multifractal spectra of seismic energy and the spectra of epicenter distribution.
FIG. 35. Generalized fractal dimension of the energy rate of acoustic emission (AE) from rock samples during laboratory experiments. Coal 1 and Coal 2 refer to the AE generated by gas sorption in the samples of coal prone and not prone to coal and gas outbursts, respectively. Dolomite refers to the AE recorded during uniaxial compression test of a dolomite sample. [Reprinted from Mortimer et al. (1999), Fig. 4.]

for local seismic events connected with particular active stopes. In both cases, they are very similar in shapes and differ by about 1.0.

An increasing interest in quantitative fractal analysis of seismic data from mines raised the question of the reliability of particular estimating procedures and the certainty of their results. By estimating the fractal dimensions for Monte Carlo simulated fractals, Lasocki and De Luca (1998) tried to determine how useful the box-counting, the number-radius, and the correlation integral methods are for identifying the features of geometry of epicenter distributions. The analysis showed that the dependence of dimension estimates on the epicenter distribution is complex and variable. It is clear from relation (3) that the clustering dimension, $D_{cl}$, is not sensitive to the variation of azimuth distributions of the polar system of coordinates as long as the position of the center of mass of a given set remains unchanged. Thus, in some cases this measure can be inappropriate to monitor the tendencies of collinear ordering of epicenters. The relation between the correlation dimension $D_2$ and the distribution of radial coordinates was complex. It was concluded that for the same distribution of azimuth, the direction of the change of $D_2$ does not determine the direction of the change of radius distribution. The box-dimension, in turn, is largely independent of the radius distribution variation only. Conversely, both the correlation and box dimensions respond monotonically to the variation of azimuth distributions. In their final conclusion, the authors stated that neither the values nor the inequalities among dimension estimates are preserved when different methods of fractal analysis are used.
An interval estimation of fractal dimensions is the next topic of importance for a quantitative fractal analysis. The fractal dimension is usually estimated by the least-squares method applied to the logarithmized form of selected relation (2), (3), (4), or (6). For the actual finite data sets, however, the logarithmic linear range of these relations is placed between the so-called depopulation and saturation regions of $r$. The extent of these regions depends upon the number of points in the set and the linear size of the hypercube encompassing the set (e.g., Nerenberg and Essex, 1990) and, in general, cannot be readily estimated. In practice, the range of linearity is selected visually, and the errors and biases in fractal dimension estimates are generated in most cases by the choice of this range. In consequence, the variance of the regression parameter estimator, which can be readily evaluated, does not represent a true variance of the estimator of fractal dimension and cannot be used to determine its confidence intervals (e.g., Pickering et al., 1995).

Cosentino et al. (1997) proposed assessment of the variance of clustering dimension $D_{cl}$ through the random split of a fractal set into subsamples, estimating the dimension for every subsample and evaluating the variance of the sample comprising these dimension estimates. The results of this procedure applied to simulated fractal sets were encouraging, although the requirement of at least 600 events in the initial data set can be difficult to satisfy by actual data. De Luca et al. (1999) tested the same procedure for the correlation dimension on large multicluster sets with different hierarchical levels. The tests proved that the distribution of dimension estimates for subsamples obtained by random split is approximately normal. Thus the confidence intervals of the correlation dimension could be readily evaluated. The authors also provided approximate empirical formulas for the calculation of bias in dimension estimates. In order to avoid unrealistic requirements for the size of the data, De Luca et al. (1999) proposed use of a resampling procedure, similar to the bootstrap (e.g., Hall, 1992; Shao and Tu, 1995) in place of the random split into nonoverlapping subsamples. This new procedure was successfully applied to natural seismicity of Sicily and to seismic data from Polish coal mines.

A comprehensive study of bias and other effects incurred on the results of fractal analysis by limitation of actual data was presented by Eneva (1996). Her actual data set consisted of 14,338 seismic events recorded for 3 months within a $400 \times 400 \times 180$ m$^3$ volume at Creighton mine in Canada. As the second actual set, she considered a subset of the first one, which comprised 1654 larger events. The spatial distribution of these data looked apparently like multifractal. Eneva (1996) repeated the fractal analysis of spatial distribution for four other sets of points, randomly generated within the parallelepiped encompassing the real data. The simulations assumed a uniform distribution of points and the monofractal distribution with a preset fractal correlation dimension. In both cases of the assumed distribution, either all generated points constructed test sets (regular simulation) or only those points that fell within a location error range from the hypocenters of actual data
Fig. 36. Generalized correlation integrals of hypocenter distribution estimated from (a) actual seismic data from Creighton mine in Canada and (b)–(e) randomly generated point sets, for $q$ from $-25$ to $25$. The simulations followed either uniform ((b), (c)) or monofractal ((d), (e)) distributions within the irregular ((b), (d)) or regular ((c), (e)) volumes outlined by the real data. [Eneva (1996), Fig. 3, reprinted with kind permission from Blackwell Science Ltd.]

(irregular simulation). Figure 36, reproduced from Eneva (1996), shows empirical log $C(q)(r)$ vs log $r$ relations for the actual data set (Fig. 36a) and for the generated sets (Figs. 36b–e). The spatial distribution of points in every simulated set could be interpreted as multifractal. The relations obtained for monofractal irregular simulations (Fig. 36d) distinctly resemble those achieved for real data (Fig. 36a). The author concluded that strong multifractality of real data may be, in many cases, an artifact created by limitations of the data. She indicated that comparing fractal dimension estimates of data having different limitations or attributing to their values a particular physical meaning could lead to false conclusions. If data, however, come from the same environment, it is reasonable to assume that their limitations remain the same. In this connection Eneva (1996) concluded that a careful fractal analysis applied to seismic data from the same region could be used
to monitor time variations of the source distribution geometry. She also suggested the use of random simulations or bootstrap techniques in order to evaluate a bias implied by actual data limitations.

In addition to the studies to the geometry of event distributions in various domains, there is a number of works related to seismicity in mines, which try to infer a possible wider physical context of the observed fractality. It is generally accepted that earthquakes result from a nonlinear dynamic process (e.g., Kagan, 1994; Newman et al., 1994). In particular, a description of natural seismicity through a self-organized criticality (SOC) of the crust (e.g., Bak and Tang, 1989; Carlson and Langer, 1989; Sornette and Sornette, 1989; Ito and Matsuzaki, 1990) is widely explored. The term SOC "refers to the spontaneous organization of a system driven from outside in a dynamical statistical stationary state, which is characterized by self-similar distributions of event size and fractal geometrical properties" (Grasso and Sornette, 1998). Thus although the fractality of distributions on its own does not imply that a system is a SOC state (Main et al., 1994; Grasso and Sornette, 1998 and the references therein), self-similarity is a property strongly supporting this hypothesis and/or possible chaotic behavior of the system (e.g., Turcotte, 1989; Bak et al., 1994).

The idea that seismically active rock mass of a mine might also be in a state of SOC was formulated by Grasso (1993), within a wider concept of induced SOC systems as the subsets of SOC proposed for tectonic earthquakes. He considered self-similarity in the source scaling and recurrence relations as the evidence for the SOC of seismicity in mines. Breaking of the linearity of recurrence relations was explained in terms of a finite size of local seismogenic volumes, e.g., mine pillars. A lifetime of an induced SOC system in mines was estimated from the time of activity decay after the stoppage of mining to a few tens of hours. Grasso (1993) also suggested that an observed disagreement between the spatial fractal dimension of distribution of seismic events and the fractal dimension of fault system of the active region could be attributed to the presence of aseismic instabilities. This induced SOC concept was explored later by Grasso and Sornette (1998) as evidence for a general self-organized criticality of the crust.

The state of SOC in the Sudbury Basin in Canada, encompassing 16 nickel-copper mines, was considered by Morrison et al. (1993) from spatial and temporal distributions of larger seismic events. They argued that a weak correlation between the origin times and locations of these events and the blasting times and locations of prominent geological structures points to chaotic behavior within a mine system.

Idziak (1999) studied fractal geometry in time and space of four spatial concentrations of larger, regional type, events from the Upper Silesian Coal Basin in Poland. He suggested, considering the observed fractality and an unexpected migration of seismic sources among active areas, that the whole region should be in a SOC state.
According to Mortimer (1997), the fractal character of temporal and spatial distributions of local seismic events, directly connected with mining of isolated longwalls, may also indicate that the mine rock mass is in a state of SOC.

Steacy et al. (1997) attempted to validate the idea of SOC in mines by simulating seismicity at Creighton mine in Canada by a 3D self-organized critical model. A network of 2D cellular automata distributed within the volume represented the fault structure of the mine, and the model was loaded from point sources following the actual blasting pattern. The simulated activity was then compared with the actual activity in the mine. Global characteristics of seismicity comprised a number of parameters including spatial and temporal fractal correlation dimensions. A chi-square testing procedure estimated the resemblance between these global characteristics of the actual and modeled seismicity to be about 95%.

A slightly different opinion on the mechanism of seismic event generation in mines was expressed by McGarr and Simpson (1997). They included seismic events from mines to a class of induced events, which are caused by large stress changes that result from mining or high-pressure liquid injection alone.

If a process is chaotic, due either to the SOC or other reasons, it could still be predictable, although in a limited sense. In the long term the process is not predictable but a short-term interval predictability is possible, provided the system is characterized by a low-dimensional strange attractor (e.g., Hilborn, 1994; Abarbanel, 1996). This possibility is of special interest for seismicity in mines where even short-term reliable forecasts could considerably increase the safety of miners. According to the theorem of Takens (1981), the phase-space of a dynamic system can be reconstructed from a time series of one of its observables, $s(t)$, using as a trajectory in a $d$-dimensional space, $\mathcal{R}^d$ the following vector $\mathbf{x}(t) = \{s(t), s(t + \tau), \ldots, s(t + (d - 1)\tau)\}$, where $\tau$ is a constant time lag and $d$ is an embedding dimension. The strangeness, i.e., the fractal dimension of an attractor, can be evaluated by using successive estimates of correlation integrals $C^{(2)}(r)$ (relation (1)) from $\mathbf{x}(t)$ for the increasing embedding dimension $d$ (Grassberger and Procaccia, 1983). For every $d$ there is a range of $r$ where $\log C^{(2)}(r)$ scales linearly with $\log r$, and the slope $D_2(d)$ is the correlation dimension of the embedded phase-space. With the increasing embedding dimension, $d$, the correlation dimension $D_2(d)$ approaches a limit, $D_2^l$, which is the dimension of the underlying attractor. A minimal value of $d$, $d_{\text{min}}$, from which $D_2(d)$ stabilizes, is considered as the dimension of the reconstructed phase-space. If the limit $D_2^l$ does not exist, the series is likely to be random (e.g., Beltrami and Mareschal, 1993).

The dynamics of a $d$-dimensional phase space is characterized by $d$ Lyapunov exponents, which are the average exponential rates of divergence or convergence of nearby trajectories (e.g., Wolf et al., 1985). These exponents are all zero for integrable systems, at least one is positive for a chaotic system, and all are negative for periodic states. In the case of chaotic behavior, the value of the largest Lyapunov exponent $\lambda_+$ determines the predictability of the system.
Radu et al. (1997) used the algorithm of Grassberger and Procaccia (1983) to determine the strange attractor and the dimension of phase space of a seismic flow in mines. According to these authors, the seismic flow of rocks in a mined rock mass is complex, with turbulence that, in turn, is associated with chaotic behavior. The algorithm was applied to various parameters of seismic events from three areas in the Welkom goldfield in South Africa. These were the logarithm of seismic energy and seismic moment, the hypocenter coordinates, and the interevent distance and time of consecutive events. The time lag, \( \tau \), was chosen as the first zero of the linear autocorrelation function (Abarbanel, 1996):

\[
C(m) = \frac{\sum_{k=1}^{N-m} (s_i - \bar{s})(s_{i+m} - \bar{s})}{\sum_{i=1}^{N} (s_i - \bar{s})^2},
\]

where \( m = 1, 2, \ldots, N \) is the number of events, \( s \) is the event parameterization and, \( \bar{s} = \frac{1}{N} \sum_{i=1}^{N} s_i \).

Figure 37, reproduced from Radu et al. (1997), presents the obtained relations between the embedding dimension \( d \) and the correlation dimension of the embedded phase-space \( D_2(d) \) for one of the studied seismic series. The authors concluded that all but one relation indicate that the phase-space is eight-dimensional. The dimension of the attractor was determined to be about 6.0. In one exceptional case (Fig. 37h), when the seismic series was parameterized by the interevent time, the phase-space looked four-dimensional and the dimension of the attractor was 3.3. Similar results from the seismic data from the other two areas led the authors to conclude that the process is chaotic with a low-dimensional strange attractor, and eight physical parameters are responsible for its dynamics. Furthermore, Radu et al. (1997) used the algorithm of Wolf et al. (1985) to estimate the greatest Lyapunov exponent, \( \lambda_+ \), which turned out to be positive and ranged from 0.9 to 2.9. The average limits of predictability for considered cases, therefore, were estimated to be from a dozen hours to 10 days. The authors have not, however, presented the number of samples in the studied series.

Mendecki (1997c) described an application of the same procedure in studies of acceleration waveforms of two different types of seismic events. The fractal dimension of attractor was from 3.3 to 3.9 for the very small seismic event and 5.2 for the complex event associated with blasting. The largest Lyapunov exponent for the first case was 0.12–0.46, and for the second case it was 0.12. The author concluded that in both cases the propagation was chaotic and deterministic.

Cichy and Mortimer (1999) applied the algorithms for determining the dimension of strange attractor and Lyapunov exponents (Wolf et al., 1985) to seismic data from the Belchatow open cast mine and from one stope at Katowice underground coal mine in Poland. They used the time series of the logarithms of event energy and times between the consecutive events. In both cases, the dimensionality of the
Fig. 37. Variation of the correlation dimension of embedded phase-space with embedding dimension $d$ for a seismic series from a selected area of Welkom goldfield, South Africa, parameterized by (a) Logarithm of seismic energy, (b) logarithm of seismic moment, (c) $x$ coordinate of hypocenter, (d) $y$ coordinate of hypocenter, (e) $z$ coordinate of hypocenter, (f) distance between hypocenters of the consecutive events, (g) interevent distance, including the source size, between the consecutive events and (h) interevent time between the consecutive events. $d_A$ is the estimated fractal dimension of the attractor and $d_{\text{min}}$ is the dimension of the reconstructed phase space. [Radu et al., 1997, Fig. 9.8, reprinted from Seismic Monitoring in Mines, ©1997 Chapman & Hall, with kind permission from Kluwer Academic Publishers.]
phase space was determined to be between 7 and 9 (attractor dimension 3–3.5). The largest Lyapunov exponent was positive although very small.

Evidence for low-dimensional deterministic chaos governing the generation of seismic events, obtained from the algorithm of Grassberger and Procaccia (1983) of the phase space reconstruction or from the Lyapunov exponent evaluation, can be, however, strongly challenged because the size of analyzed samples is often insufficient. A conservative condition for the number of points, \( N \), of a time series, required for a reliable estimation of the dimension of attractor \( D_2 \) is, at least, \( 42^{D_2} \) (Smith, 1988). More optimistic bounds proposed by Nerenberg and Essex (1990) and Eckman and Ruelle (1992) depend upon the diameter of the reconstructed attractor and other factors and range from \( 10^{D_2/2} \) to more than \( 10^{D_2} \). The number of points \( N \) needed to estimate Lyapunov exponents is about the square of that needed to estimate the dimension. Insufficient sample size could well be the reason of ambiguity of the results described above (phase space dimensionality depending on parameterization of seismic series, low Lyapunov exponents), and conclusions drawn in the described studies still need careful verification.

8. ASSESSMENT AND PREDICTION OF TIME-DEPENDENT SEISMIC HAZARD: STATISTICAL APPROACH

Here we review techniques based on seismic catalogs as principal input data, which provide indications on the present and future course of seismicity generation processes. Obviously, particular emphasis is put on estimating chances of the occurrence of seismic events that are hazardous for mining, usually strong events close to mining stopes. Recalling, however, the distinction between a rockburst and seismic event we note that such an approach narrows the problem of rockburst hazard assessment, which is a key interest of mine management, to the prediction of seismicity which can but does not necessarily have to result in rockbursts.

The term prediction, in turn, is used in a broader sense than just determination of time, location, and magnitude of a future event. Following the definition of earthquake prediction formulated by Kagan (1997a) we include into prediction or, more generally, into time-dependent seismic hazard assessment all techniques which provide time-dependent information relevant for undertaking decisions related to seismic risk. All these techniques accept in principle the variability in time of seismicity. Most of them deduce the future hazard from past-to-present time variations of a parameter or parameters estimated from seismic series.

We do not consider techniques for which seismic data are not the main input stream. Conversely, the methods, which make use of complementary, other than seismic, data are only occasionally mentioned.

The statistical sense of the time-dependent hazard assessment and prediction consists of the processing of series of seismic data. A rationale of prediction is an
expectation that such a series contains information that allows estimation of the further evolution of seismicity generation. In order to meet this condition a number of specific requirements must be met.

First of all, neither the generation of seismicity nor its parameterization can be governed by a process random in time, i.e., both stationary and memoryless. Given the seismic series parameterization $\xi$, the stationarity condition means that for every time moment $t_i$ and $t_j$, the probability density function of $\xi$ is $f(\xi|t_i) = f(\xi|t_j) = f(\xi)$. The process representation $\xi$ is memoryless if its probability distribution at time $t_i$ is independent of its values at previous times $t_{i-1}, t_{i-2}, \ldots$. The values $\xi(t_i)$ for different $t_i$ are then statistically independent, which implies that they are not correlated. In such a case the past does not have any influence on the present hazard which varies randomly in a way determined by the constant probability distribution.

From the practical point of view, however, we are not as much interested in the actual stochastic nature of the process as in its features reflected by the observations. The non-stationarity of the process can only be fruitful for prediction if this feature is represented in the recorded seismic series. In simple terms, this means that the data come from a time period covering significant process variations and that the average rate of event occurrence is greater than the average rate of these variations. Similarly, it is not enough that seismic signals generated by rock mass fracturing are correlated in general; this feature must also be detectable for the events from the observed magnitude range.

Seismicity in mines is under the dominant influence of mining (e.g., Gibowicz and Kijko, 1994). Mining activity itself varies in time and these variations take place during seismicity monitoring periods. Thus the non-stationarity of the event generation process can be expected, and its prints in seismic catalogs are commonly observed.

Statistical investigations of the stochastic nature of seismic series recorded close to mining stopes in underground coal mines in Poland were carried out by Lasocki (1992a) who tested the hypothesis of Poissonian structure of such series. If the event generation process were Poissonian, then it would be fully random, stationary, and memoryless, and the number of events $N$ in a constant time interval $\Delta t$ would follow the Poisson distribution

$$\Pr\{N = n; \Delta t\} = \frac{(\lambda \Delta t)^n}{n!} \exp(-\lambda \Delta t),$$  \hspace{1cm} (8)

where $\lambda$ is the constant mean rate of event occurrence. The interevent time, $\tau$, would have an exponential distribution

$$f(\tau) = \lambda \exp(-\lambda \tau).$$ \hspace{1cm} (9)

Four series from different mines were used to test the consistency between the sample distributions and distributions (8) and (9) implied by the Poisson process.
For long complete series, the hypothesis of consistency was rejected with very high probability. The Poissonian character of seismicity generation process, however, could not be excluded when the sampling period was shortened to 50 days. This result suggests that the recognized non-Poissonian structure of seismic series is more likely caused by time variations of the process itself rather than by a correlation between the events, which otherwise should be significant also in short samples. It was found that the series consisting of only large events tended to be Poissonian, which points to prediction of large events by the information carried by small events.

Lasocki (1992b) tested the consistency between the sample distribution of interevent time and the Weibull distribution for 14 seismic series from five Polish mines. The Weibull distribution, extensively used in survival analysis, describes the time distribution between the events in not fully random point processes. It represents a positive aging process (increasing occurrence rate process) when its shape parameter is greater than one, and a negative aging process (decreasing occurrence rate process) when the opposite is true. The Weibull distribution has occasionally been used for analysis of earthquake series (Hagiwara, 1974; Kiremidjian and Anagnos, 1984). Only the positive aging possibility was considered, which means that the longer the time has elapsed since the last event the more probable is the next event occurrence. The reason for this becomes obvious when a constant driving mechanism for earthquakes is accepted.

Surprisingly, the sample distribution of interevent time of mining events fitted the Weibull model quite well, but revealed the negative aging process. The probability of the next event occurrence seemed to decrease with the time elapsed since the last event. This result reflects a complex influence of irregular mining operations on seismic activity. In the utmost case, after stoppage or interruption of mining the negative aging is apparent: the event rate gradually decreases.

Kijko (1997) performed similar studies of interevent times from the 1-year data collected at one of deep gold mines in South Africa. The empirical cumulative distribution function of interevent time $F_{\text{OBS}}(\tau)$ was compared with the cumulative distribution function of the exponential distribution $F(\tau) = 1 - \exp(-\lambda \tau)$. A significant difference between $F_{\text{OBS}}(\tau)$ and $F(\tau)$ proved interrelations between the successive events. This time, however, the differences were mostly positive, which suggests the positive aging property. This positive aging extended to some 20 days.

Different results of these two studies are probably connected with different sources of nonrandomness in the studied samples. The mean rate of event occurrence, $\lambda$, in the series from the Polish mines was low, seldom reaching 10 events/day, and the nonrandom features found in the series from longer than 100-day periods were probably caused by time variability of the process. The South African data had $\lambda \approx 55$ events/day, and the positive interrelation was ascertained for interevent times shorter than 0.04 day. It is likely that this feature was due to the short time
correlation among the events, which cannot be found in sparse series. Thus, it
seems that the nonrandomness of event occurrence in mines results predominantly
from the time variation of event generation processes for long time scales and
from event interrelations for short time scales. Moreover, there are intermediate-
scale time periods from which the seismic series can be regarded as the out-
come of Poisson process, a very important conclusion from practical point of
view.

Most of the techniques for the assessment of time-dependent hazard introduce
and analyze certain parameterizations of the observed series. The values of param-
eters, assigned to the present time, \( t \), are usually estimated from the seismic data
from a time widow just preceding \( t \). It is then implicitly assumed that the data from
the considered window form a simple statistical sample, i.e., they are stationary
and statistically independent. This requirement is, in general, in contradiction to
the nonrandomness of seismicity. The last result from the described studies sug-
gests, however, that there exists a window length for which such a condition is
approximately fulfilled.

Fractal character of seismicity in mines in time, space, and energy/magnitude
domains has been confirmed by the data from various regions (e.g., Coughlin
and Kranz, 1991; Xie and Pariseau, 1992; Enev and Young, 1993; Idziak and
Zuberek, 1995; Mortimer and Lasocki, 1996a; Shivakumar et al., 1996). It has been
concluded, therefore, that the process of seismicity generation may be chaotic (e.g.,
Grasso, 1993; Mendecki, 1997c; Radu et al., 1997; Grasso and Sornette, 1998).
Such evidence for deterministic chaos in the generation of seismicity in mines is
another argument against the randomness of this process.

The other evidence of interrelations between the time of occurrence of seismic
events was given by Kijko (1997) who studied the Matsumara coefficient of ran-
domness for the seismic catalog from a gold mine in South Africa. The coefficient,
declared for interevent times, \( \tau \), as \( C_r = \frac{\sum (\tau_i)^2}{\sum \tau_i^2} \), is equal to \( \pi/4 \) for the
Poisson occurrence, greater than \( \pi/4 \) for periodic events and less than \( \pi/4 \) for
events clustered in time (Matsumara, 1984). For the studied series from mines,
\( C_r \) was definitely less than \( \pi/4 \), although it depended on a lower cut-off of the
apparent stress applied to the data. The interaction seemed to be the largest for
small events.

The next important condition for a successful prediction is that the input data
are more or less homogeneous, in that most of them result from the same pro-
cess. Furthermore, the future seismicity that is a target of prediction must also
result from the process, from which the analyzed data are collected. Regrettably,
this condition is frequently violated. It is well known that seismicity induced by
mining is generated by stress redistribution at different scales. Studies of the ori-
gin of seismic events in mines allows us to distinguish at least two classes of
seismic events: local events directly connected with mining operations and re-
gional events induced or triggered by mining (e.g., Gibowicz and Kijko, 1994).
The mixing of components from different processes in seismic series has been also confirmed by analyses of modality of magnitude, energy, or seismic moment distributions.

Kijko et al. (1987) studied the distribution of energy of seismic events from Polish coal mines using the extreme value theory of Gumbel (1962). If $X$ is an outcome from a certain process (e.g., seismic energy or seismic moment release) of values $x_i, i = 1, \ldots, n$, then its maximum values in equal time intervals, $y_k, k = 1, \ldots, m$, form the outcome $Y$ from the process of maximum generation. Gumbel has shown that when $n$ tends towards infinity the distribution of $Y$, if it exists, can take only one form from the three possible. The first form is created by unbounded distributions like normal or exponential ones, the second by the distributions bounded from below, and the third by those bounded from above. The three Gumbel distributions are smooth and monotonic functions. In contrast, the empirical cumulative distributions of the largest seismic energies observed in mines are not smooth and distinct inflexions are observed. Figure 38, reproduced from Kijko et al. (1987), shows that the point of inflexion splits the empirical distribution into two branches, which can be modeled by separate curves of the third Gumbel distribution type. Such results led the authors to the conclusion that at least two seismic energy release modes, the high- and low-energy mode, are present in the studied seismic series. Similar results were obtained by Idziak et al.
(1991), who studied seismic series from an area of three geological units in the Upper Silesian Coal Basin in Poland.

The multimodality of magnitude distribution from seismic events in mines was also tested statistically (Lasocki and Węglarczyk, 1998). A nonparametric kernel density estimator was used to represent the sample distribution of magnitudes from a 1-month seismic series observed in a South African gold mine. It was evaluated by the critical smoothing test (Silverman, 1986) that the unimodality of the actual distribution was highly improbable.

The degree of sample homogeneity also strongly depends on the sampling constraints in terms of time, space, and event size. The problem of optimal selection of seismic data for hazard assessment is site dependent, since hazard depends on local geology, geometry of mine layouts, and characteristics of mining activity.

The events in seismic catalogs are described by a limited number of parameters, which usually are the time of occurrence, location coordinates, and at least one parameter describing the event strength: magnitude, seismic moment, or seismic energy. The series of values of these parameters form the input data for prediction. The information from these series is condensed, with some exceptions, in their parameterization and their anomalies should indicate the changes in seismic hazard. Such a “second-level” parameterization can use one parameter of the seismic event, becoming a one-parameter technique, or it can use more than one parameter, and then it is called a multiparameter technique. The prediction, on the other hand, can be based on analyses of one or more parameterizations. Then we have either univariate or a multivariate prediction. Within the multivariate approach we should distinguish between the quasi-multivariate and true multivariate methods. In the first case, the only general recommendation is to observe simultaneously more than one parameterization. In the second case, specific algorithms are needed to combine either parameterizations or their indications.

Validation is the next hot topic in statistical prediction. By the proper validation we mean any objective method which evaluates profits from a given prediction technique in comparison to the random guess. Such a method must evaluate a statistically significant number of applications of the prediction technique so that the actual probabilistic effectiveness of this technique can be estimated with high confidence.

As long as a prediction technique is not statistically validated, its effectiveness can always be questioned, no matter how strong its physical basis, how complex its algorithms, or how many so-called typical examples have been shown. This obvious statement is often overlooked and we note relatively few reports on the validation in comparison to the number of proposed prediction techniques. There are several reasons for such a situation. Most introduced techniques have adjustable parameters, for example, the length of time windows for estimating parameters, the dimensions of the volume from which seismic events are considered as input data, some weights, and others. Determination of the rules for optimal
selection of these parameters requires time-consuming studies of different seismic series. The validation can be done only after any such rules (though not necessarily optimal) are established, and it needs quite a bit of data. If access to the data is limited, then the whole procedure could become unfeasible.

The other difficulty is created by the validation procedure itself. There are no ready rules for this procedure; neither is there an agreement on the best measures of prediction efficiency. Similar problems encountered in the attempts to predict natural earthquakes have been discussed by Kagan (1997a).

We should also note a certain negative social conditioning regarding prediction of seismic hazard in mines. An understandable demand for accurate prediction and impatience of mining management often result in hastily publicizing procedures, when neither all their properties are known, nor their real efficiency has been fully tested.

8.1. Recurrence Relationships: Source Models for Stationary Seismic Hazard Analysis

Although the techniques reviewed herein do not necessarily deal with parameters used in traditional seismic hazard analyses, many of them have been developed from the well-known methodology in earthquake seismology (e.g., Cornell, 1968; Reiter, 1991). It is not strange, therefore, that the recurrence relationships describing the percentage of seismic events from a particular magnitude range in a given population are similar to those used to study natural earthquakes.

The applicability, even if approximate, of the well-known Gutenberg-Richter relation to seismic events in mines provides a handy and popular parameterization of the magnitude and energy distributions. Above a certain limiting value, $M_{\text{min}}$, determined by the completeness of a seismic catalog, the logarithm of the number $N$ of events of magnitude not smaller than $M$ is linearly related to $M$: \( \log N = a - bM \). A similar relationship can be written for the logarithm of seismic energy and for the logarithm of seismic moment. The constant $a$ is determined by the event rate, $\lambda$, whereas $b$ depends upon the ratio of large to small events. This classical relation can be readily converted into the probabilistic distribution of magnitude or energy. For magnitude it is the unbounded from above exponential distribution (Aki, 1965)

\[
F(M) = \begin{cases} 
0 & \text{for } M < M_{\text{min}} \\
1 - \exp[-\beta(M - M_{\text{min}})] & \text{for } M \geq M_{\text{min}}
\end{cases},
\]  

and for the seismic energy (or seismic moment) it is the Pareto distribution (e.g., Lasocki, 1989; Gibowicz and Kijko, 1994)

\[
F(E) = \begin{cases} 
0 & \text{for } E < E_{\text{min}} \\
1 - (E_{\text{min}}/E)^{b'} & \text{for } E \geq E_{\text{min}}
\end{cases},
\]
where $F(\bullet)$ is the cumulative distribution function, $\beta = b \ln 10$, and $b'$ is the slope parameter of log $N$ vs log $E$ relation.

To ensure a finite seismic energy release (Knopoff and Kagan, 1977), the Gutenberg–Richter relation is combined with an assumption on the existence of a physical upper limit of the magnitude range, $M_{\text{max}}$, or the energy range, $E_{\text{max}}$. The magnitude distribution takes the form of the bounded from above exponential distribution (Page, 1968; Cosentino et al., 1977)

$$F(M) = \begin{cases} 
0, & \text{for } M < M_{\text{min}} \\
1 - \exp[-\beta(M - M_{\text{min}})], & \text{for } M_{\text{min}} \leq M \leq M_{\text{max}}, \\
1, & \text{for } M > M_{\text{max}} 
\end{cases} \quad (12)$$

and the energy distribution is the truncated Pareto distribution (Lasocki, 1992c, 1993a)

$$F(E) = \begin{cases} 
0, & \text{for } E < E_{\text{min}} \\
1 - \left(\frac{E_{\text{min}}}{E}\right)^{b'}, & \text{for } E_{\text{min}} \leq E \leq E_{\text{max}}, \\
1, & \text{for } E > E_{\text{max}} 
\end{cases} \quad (13)$$

where $M_{\text{max}}$ and $E_{\text{max}}$ are characteristic for specific conditions of rock mass fracturing and must be estimated.

The shape parameter $\beta$ of distributions (10)–(13) can be estimated by means of the maximum likelihood method (e.g., Aki, 1965; Utsu, 1965, 1966; Page, 1968; Kijko, 1982; Lasocki, 1989, 1992c; Gibowicz and Kijko, 1994), but the likelihood functions for relations (12) and (13) are monotonic with respect to $M_{\text{max}}$ and $E_{\text{max}}$, and the statistical estimation of these upper bounds is not straightforward.

Kijko and Funk (1994) proposed three alternative methods of $M_{\text{max}}$ estimation for the seismic hazard assessment in mines. The simplest method estimates the upper bound of the magnitude distribution from the largest $M_{\text{max}}$ and the second largest $M_{\text{max}-1}^{obs}$ observed magnitudes (Robson and Whitlock, 1964)

$$\hat{M}_{\text{max}} = 2M_{\text{max}}^{obs} - M_{\text{max}-1}^{obs}, \quad (14)$$

which can be considered as a rough but always available approximation of $M_{\text{max}}$. The second method makes use of some properties of the end-point estimator of a uniform distribution and of the fact that any cumulative distribution function follows a uniform distribution within the interval (0,1). If it is assumed that the event magnitude follows the bounded from above exponential model (12) the
resultant estimator of $M_{\text{max}}$ is

$$\hat{M}_{\text{max}} = -\frac{1}{\beta} \ln \left\{ \exp(-\beta M_{\text{min}}) - \left[ \exp(-\beta M_{\text{min}}) - \exp(-\beta M_{\text{max}}) \right]\frac{n + 1}{n} \right\},$$

(15)

where $n$ is the size of seismic catalog. The estimator depends upon the value of $\beta$, hence a recursive procedure must be used to obtain both $\beta$ and $M_{\text{max}}$ simultaneously. Moreover, on some occasions the system of equations comprising Eq. (15) and the likelihood equation for $\beta$, $\partial l/\partial \beta = 0$, where $l$ is the logarithmic likelihood function, has no solutions. The third method, used earlier in earthquake hazard studies (Kijko and Sellevoll, 1989), is based on the condition that the expected largest magnitude in the catalog is equal to the largest observed magnitude. In the case of magnitude model (12), the estimator for $M_{\text{max}}$ is

$$\hat{M}_{\text{max}} = M_{\text{obs}}^{\max} + \frac{E_1(nz_2) - E_1(nz_1)}{\beta \exp(-nz_2)} + M_{\text{min}} \exp(-n),$$

(16)

where $z_i = -A_i/(A_2 - A_1), A_1 = \exp(-\beta M_{\text{min}}), A_2 = \exp(-\beta M_{\text{max}})$, and $E_1(\bullet)$ is the exponential integral function. This method also requires a recursive procedure to estimate the distribution parameters.

Kijko and Graham (1998) recently introduced a generic formula for the evaluation of maximum regional magnitude, which can be used for any distribution of magnitudes

$$\hat{M}_{\text{max}} = M_{\text{max}}^{\text{obs}} + \int_{M_{\text{min}}}^{M_{\text{max}}^{\text{obs}}} [F(M)]^n \, dm.$$  

(17)

It was shown that when the cumulative distribution function of magnitude is given by relation (12), then estimator (16) is the approximate form of estimator (17). Kijko and Graham (1998) also considered the possibility of random variation of the Gutenberg-Richter parameter $b$ and provided an alternative estimator to deal with such a case. A comparison of the performance of different estimators revealed better estimator efficiency, either with constant or with variable $b$, based on generic formula (17).

As in earthquake seismology, an increasing number of cases is reported from mines where the observed recurrence patterns differ significantly from the Gutenberg-Richter relation (e.g., Dessokey, 1984; Kijko et al., 1987; Lasocki, 1987, 1993b; Subbaramu et al., 1989; Johnston and Einstein, 1990; Urbancic et al., 1992b; Trifu et al., 1993; Finnie, 1994; Feustel, 1997; Lasocki and Weglarczyk, 1998). Evidence of the nonlinear structure of empirical log-frequency-magnitude distributions stimulated the search for nonlinear models of recurrence relationships that could better fit the observations. Lasocki (1987) tested the
goodness of fit of three theoretical models for magnitude distributions, the gamma, the Weibull, and a second-order power law derived from a parabolic form of the log-frequency-magnitude relation, to the seismic data from a selected stope of a coal mine. Out of these three models, the Weibull distribution of the cumulative distribution function

\[
F(x) = \begin{cases} 
0 & \text{for } x \leq x_0 \\
1 - \exp\left[-\alpha(x - x_0)\gamma\right] & \text{for } x > x_0
\end{cases},
\]

where \(x\) is either magnitude or the logarithm of seismic energy, \(x_0\) is the lower limit of the model applicability, and \(\alpha\) and \(\gamma\) are parameters, turned out to be the only good candidate for a model of magnitude or energy distributions. Unlike linear models (12) and (13), the Weibull distribution belongs to the class of so-called “soft” maximum magnitude models which do not have a rigorous upper limit of the magnitude range (e.g., Main and Burton, 1984; Kagan, 1997b). Detailed studies of this distribution, carried out on multiple well-populated seismic series from four mines, showed that the distribution can be used to model some unimodal and smooth nonlinear frequency–magnitude distributions (Lasocki, 1993b). This was also successfully used as a source model for seismic hazard analysis by Kijko et al. (1993) and Finnie (1994).

Regardless of their linearity or nonlinearity, all these parametric models of magnitude/energy distributions are unimodal at best. In many cases, however, the observed frequency–magnitude relations for seismic data from mines exhibit a multicomponent structure (e.g., Kijko et al., 1987; Urbancic et al., 1992b; Trifu et al., 1993; Lasocki and Weglarczyk, 1998). In such cases the unimodal theoretical distributions may not fit the data. A recently proposed alternative to the parametric models, based on nonparametric, model-free estimators of the magnitude density, is expected to be suitable for linear as well as all kinds of nonlinear sample distributions of magnitude (Kijko et al., 2000; Lasocki et al., 2000). Given the sample data \(x_i, i = 1, \ldots, n\), the kernel estimator \(\hat{f}(x)\) of an actual probability density function \(f(x)\) is

\[
\hat{f}(x) = \frac{1}{nh} \sum_{i=1}^{n} K\left(\frac{x - x_i}{h}\right),
\]

where \(h\) is a positive smoothing factor and the kernel function \(K(*)\) is a probability density function symmetric around zero (e.g., Silverman, 1986). For the sample comprising event magnitudes, Kijko et al. (2000) used the Gaussian kernel function \(K(\xi) = 1/\sqrt{2\pi} \exp(-\xi^2/2)\) and the least-squares cross-validation to estimate \(h\) (e.g., Bowman et al., 1984), modified for the Gaussian kernel function. Assuming a sharp cut-off magnitude as a maximum magnitude \(M_{\text{max}}\), the nonparametric,
kernel estimator of the cumulative distribution function of magnitude is

\[
\hat{F}(M) = \begin{cases} 
0, & \text{for } M < M_{\text{min}} \\
\frac{\sum_{i=1}^{n} \left[ \Phi \left( \frac{M_i - M_{\text{min}}}{h} \right) - \Phi \left( \frac{M_{\text{max}} - M_i}{h} \right) \right]}{\sum_{i=1}^{n} \left[ \Phi \left( \frac{M_{\text{max}} - M_{\text{min}}}{h} \right) - \Phi \left( \frac{M_{\text{min}} - M_{\text{min}}}{h} \right) \right]}, & \text{for } M_{\text{min}} \leq M \leq M_{\text{max}}, \\
1, & \text{for } M > M_{\text{max}}
\end{cases}
\]

where \( \Phi(\xi) \) denotes the standard Gaussian cumulative distribution function. From the simulation tests of efficiency, Kijko et al. (2000) concluded that estimator (20) provides results with tolerable and limited errors regardless of whether the sample distribution of magnitude follows the Gutenberg–Richter relation, is nonlinear unimodal, or is nonlinear multimodal.

8.2. Time Variations of Energy/Magnitude Distributions

The most popular technique for assessing the variation of seismic hazard in mines is probably the use of time variations of energy or magnitude distributions of the observed events. This approach has been used in Poland (e.g., Lasocki, 1982, 1993a), in South Africa (e.g., Kijko et al., 1993; Stewart and Spottswood, 1993; Toper et al., 1997), in the Czech Republic (e.g., Rudajev, 1993; Holub, 1995; Kalenda, 1995), and in Canada (e.g., Feustel, 1997; Trifu et al., 1997).

Significant time variations of the Gutenberg–Richter \( b \)-value in mining induced seismicity and their potential for prediction were noted more than 20 years ago (e.g., Brady, 1977; Gibowicz, 1979). Since then, numerous confirmations of these variations have come from different countries. Opposite opinions, occasionally expressed, were either not substantiated empirically (e.g., Mendecki, 1997b, p. 194) or likely due to a data analysis not sufficiently precise. Urbancic et al. (1992b) and Young et al. (1992) concluded that \( b \) significantly changes in space but not in time. The studied data comprised seismic observations from a 2-month period within a specific region of Strathcona mine in Canada. Because of nonlinearity of the frequency–magnitude curve, the authors limited the magnitude range that was used to estimate the value of \( b \). Excluding stronger events from the samples and using very narrow magnitude range (0.7 of a magnitude unit) resulted in most probably losing information on the time-variation of magnitude distribution.

The application of time changes of \( b \) for the time-dependent seismic hazard assessment in mines also has a relatively long history. In the early attempts (Lasocki, 1982), \( b \) values were estimated by the least-squares method (linear regression). Later the more robust and accurate maximum likelihood method was applied (e.g., Lasocki, 1989), although the least-squares (e.g., Stewart and Spottswood, 1993;
Holub, 1995; Kalenda, 1995, 1996) or weighted least-squares procedures (Trifu and Shumila, 1996; Trifu et al., 1997) seem to be still in use. The time-dependent $b$ values provide at best a one-parameter technique for the hazard assessment, since $b$ represents the distribution of magnitudes only. This distribution quantifies the probability that an oncoming event will be large, with no regard to the chance that the event will occur at all. The indications provided by the analysis of $b$ value, therefore, should be integrated with information on the event rate.

In order to overcome this problem, Trifu et al. (1997) tried to analyze concurrent variations of $a$ and $b$ estimates of the classical unbounded frequency–magnitude relation. Their catalog comprised 1502 microseismic events ($M < 1$) that occurred during a 2-month period within a $200 \times 200 \times 200$ m volume at Strathcona mine in Canada. On the 28th day of the analyzed time interval a rockburst of $M = 2.9$ was recorded. The authors estimated $b$ parameter and occurrence rates for a space–time grid. Figure 39, taken from Trifu et al. (1997), presents the $b$ versus occurrence rate patterns for the intervals prior to and after the rockburst. Every $b$ estimate is related to two occurrence rate estimates, connected with relatively smaller (filled circles) and larger (open circles) events, respectively. Trifu et al. (1997) interpreted the decrease in $b$ prior to the rockburst (Fig. 39c) as due to a steady increase in the occurrence rate of the larger events, whereas the activity of the smaller events remained more or less constant.

Feustel (1997) presented similar retrospective studies of the data from an open-stope mine in eastern Canada. He analyzed time–space distributions of $b$ and the number of events in order to identify the relations between the time-dependent hazard and the occurrence of ground falls. On all occasions, the falls were preceded by an increase in $b$ values. It was suggested that this effect was associated with fracture development due to gravity.

An attempt to supplement the analysis of $b$ time changes with another parameter of energy distribution was undertaken by Slavik et al. (1992) and Kalenda (1995). They used as the second parameter the energy value corresponding to zero of the left-hand side of the Gutenberg–Richter relation $\log N = a' - b' \log E$. This quantity, equal to $10^{(a'/b')}$, is related to $a'$. Thus, it can replace $a'$ or the event rate in a 2D parameterization of the seismic energy flow, although its use as a measure of the upper limit of seismic energy results in a rather considerable over-interpretation.

A two-parameter approach, which fully integrates the time variations of the event rate and the distribution of magnitudes, was formulated by Lasocki (1993a). It was assumed that the series of seismic events connected with one mining stope results from the generalized Poisson process. The process is quasi-stationary in a time window suitable for a reliable estimation of its parameters, and the distribution of event energy is known. Thus the events are regarded as uncorrelated, which means that the probability distribution of the number of events in a constant time interval is given by the Poisson distribution (Eq. 8), but with the time-dependent $\lambda$. The
seismic hazard is quantified by the probability that within a short time interval $[t, t + \Delta t]$ the events will occur and the first of them will exceed a prescribed value of the energy $E_p$

$$R(E_p, t) = (1 - \exp[-\lambda(t)\Delta t])(1 - F(E; t)),$$  \hspace{1cm} (21)

where $F(E; t)$ is the cumulative distribution function of energy, whose parameters vary in time. Three optional models of the distribution of energy were considered: the one-parameter Pareto distribution (Eq. 11) coming from the simple Gutenberg–Richter relation, the truncated (two-parameter) Pareto distribution (Eq. 13), and the Weibull distribution (Eq. 18).

The values of model parameters and event rate, $\lambda$, for a given time, $t$, are estimated from seismic data from a time window preceding $t$. The length of the
widow, selected according to the level of seismicity and the time variability of seismic activity, either is kept constant or is variable to maintain the same advance of mining front in every time window. Consecutive estimates of \( R(E_p, t) \) for a moving time window represent variations of seismic hazard. Strong events are expected when values of \( R \) systematically increase or are high.

Lasocki (1993a) provided a comprehensive description of this technique, the maximum likelihood estimators of parameters, statistical tests of the energy distribution models, a discussion on adjustable parameter selection (e.g., time window length), and practical examples. Figure 40, reproduced from this work, presents the results of \( R \) estimation for seismicity recorded in an area of one longwall at Katowice coal mine in Poland during a 1-year period. The threshold, \( E_p \), defining the strong event, was \( 10^6 \) J. The constant longwall advance of 30 m determined the time window, which was moved consecutively by 1 day. The time of occurrence of

![Graph](image-url)

**Fig. 40.** Time-dependent probability of the occurrence of seismic events with energy \( E \geq 10^6 \) J in a longwall area at Katowice mine, Poland, shown for the sake of clarity on three separate graphs. The arrows mark the time of the occurrence of observed stronger events; their energy is also marked. [Reprinted from Lasocki (1993a), Fig. 6.]
strong events, marked by arrows, visibly correlates with local maxima of $R$. After numerous retrospective analyses of seismic series from various sites, the practical usefulness of this method has been confirmed and the method is used at some Polish mines for a routine day-by-day analysis of seismic hazard.

As a result of a relatively low event rate at Polish mines and consequently poorly populated time windows, the Polish applications of the described method are mostly based on a simple one-parameter Pareto model of energy distribution (Eq. 11). In some cases, however, the curvature of the frequency–magnitude relation may be significant and the classical Gutenberg–Richter relation may not be adequate. Kijko et al. (1993) used this method to analyze the spatially clustered seismic events from the Western Deep Levels gold mine, South Africa, recorded during a 1-year period. Seismic hazard was quantified by the probability of occurrence of a seismic event with a value of $M$ greater than or equal to $M_p$

$$\Pr [M \geq M_p, t] = 1 - \exp\{\lambda(t)\Delta t[1 - F(M_p; t)]\},$$

(22)

rather than by $R(M_p, t)$. A very high event rate (altogether, more than 14,300 seismic events) made it possible to apply the Weibull distribution (Eq. 18) as a model for event magnitude. In Fig. 41, taken from Kijko et al. (1993), a comparison is shown of the time-dependent estimates of the probability of event occurrence (Eq. 22), calculated using the classical Gutenberg–Richter exponential form (Eq. 10) and the Weibull model for magnitude. The correlation of occurrence of strong events, indicated by arrows, with local maxima of the Weibull-model-based estimates is distinctly better than of those provided by the exponential model. It was concluded that the Weibull model should be applied for small volume clusters of seismicity. The clusters of larger volumes show a more linear frequency–magnitude relation.

Finnie (1994) presented another retrospective application of this method based on the Weibull model for magnitude distribution. His catalog consisted of 5431 events that occurred in a selected area of the Far West Rand mining district in South Africa, observed during an 11-year period between 1982 and 1993. Figure 42, reproduced from Finnie (1994), shows the time-dependent probability of the occurrence of seismic events with magnitude $M_L \geq 3.8$. The analysis was done with a constant 1-year-long time window, moving every 3 months. The correlation between the actual occurrence of seismic events, indicated by bars, and the increasing branches or local maxima of the probability curve is rather good.

Lasocki (1993a) also proposed a new version of this time-dependent hazard analysis, based only on the largest energies from nonoverlapping, equal time intervals covering the time window. Assuming a Poissonian structure for seismic series in a given time window, the exact distributions of maximum energy for all three options of underlying distributions (11), (13), and (18) and the maximum likelihood estimators of their parameters were introduced. It was suggested that this version of the analysis can be used when the data due to the process under study are mixed with other seismic data, unimportant for the undertaken analysis,
or when the input is somehow contaminated (e.g., the level of completeness changed during the period of observation).

A three-parameter time-dependent seismic hazard assessment method, which integrates information supplied by the time of event occurrence, its seismic energy, and seismic moment, and, furthermore, accounts for a possible correlation among the events, was presented by Kijko and Sciocatti (Kijko, 1996, 1997 after Kijko and Sciocatti, 1995b). The method was fostered by the well-known fact that the events in a foreshocks–main shock–aftershocks sequence are strongly correlated and, as such, are not Poissonian. Thus, if such sequences are present in the seismic

---

**Fig. 41.** Comparison of probabilities of the occurrence of larger events in a selected volume in Western Deep Levels gold mine, South Africa, estimated using the classical Gutenberg-Richter exponential form (top) and the Weibull model for magnitude distributions (bottom). The time of the occurrence of observed events is marked by arrows. [Kijko et al. (1993), Fig. 4, reprinted from *Rockbursts and Seismicity in Mines*, Proceedings of the 3rd International Symposium, Kingston, Ontario, 16–18 August 1993. 462 pp., EUR 154.00/US $181.00/GBP108.00, A. A. Balkema, P.O. Box 1675, Rotterdam, Netherlands.]
series from a mine, the assumption about the Poissonian structure of the series in the time window may not be valid. Kijko and Sciocatti (1995b) introduced an alternative assumption that the seismic series consists of statistically independent (uncorrelated) clusters of events, but that the events within individual clusters are correlated. It was also assumed that the event size is expressed independently by seismic moment and seismic energy, whose bivariate distributions are the same for all the events, and that the cluster of events always starts from the strongest (main) event. The conditional intensity function, i.e., the probability of the occurrence of an event with energy $E$ and seismic moment $M_0$

$$\Lambda(E, M_0) = \lambda \cdot f(E, M_0) + \sum_i v(E_i, M_{0i}) \cdot f(\tau) \cdot f(E, M_0)$$

comprises two terms. The first term describes the Poissonian occurrence of the main events, in which $\lambda$ is the mean activity rate of the main events, and $f(E, M_0)$ is the probability density of the energy-seismic moment. The second term is responsible for the occurrence of dependent events following the main event. In this term $v(E_i, M_{0i}) f(\tau_{ij}) f(E_{ij}, M_{0ij})$ is the conditional distribution of the $j$-th event dependent on $i$-th main event, where $M_{0i} > M_{0ij}$ and $E_i > E_{ij}$, $\tau_{ij}$ is the time between the occurrence of the main and the $j$-th events and its probability density is $f(\tau)$.

Two models for the energy-seismic moment distribution were proposed. The
The unbounded model assumes that the 2D random variable \([E, M_0]\) follows a bivariate power-law distribution

\[
f(E, M_0) = \begin{cases} \frac{\beta(\beta + 1)}{E_C M_{0C}}(E/E_C + M_0/M_{0C} - 1)^{-\beta-2} & \text{if } E \geq E_{\text{min}} \\
0 & \text{otherwise}
\end{cases},
\]

where \(\beta, E_C, \text{ and } M_{0C}\) are model parameters, and \(E_{\text{min}}\) and \(M_{0\text{min}}\) are the levels of completeness of the data in energy and seismic moment domains, respectively. The assumption of finiteness of the upper boundary of both the energy and seismic moment domains modifies the density function (24) to

\[
f(E, M_0) = \begin{cases} K \cdot (E/E_C + M_0/M_{0C} - 1)^{-\beta-2} & \text{if } E_{\text{max}} \geq E \geq E_{\text{min}} \\
0 & \text{otherwise}
\end{cases},
\]

where \(E_{\text{max}}\) and \(M_{0\text{max}}\) are the respective boundaries and

\[
K = \frac{\beta(\beta + 1)}{E_C M_{0C}} \left[ 1 - \left( \frac{E_{\text{max}}}{E_C} \right)^{-\beta} - \left( \frac{M_{0\text{max}}}{M_{0C}} \right)^{-\beta} + \left( \frac{E_{\text{max}}}{E_C} + \frac{M_{0\text{max}}}{M_{0C}} - 1 \right)^{-\beta} \right]^{-1}.
\]

The model for the event interdependency is similar to that formulated by Kagan (1991). The function \(\nu(E_i, M_{0i})\) takes the form

\[
\nu(E_i, M_{0i}) = k M_{0i}^\gamma,
\]

where \(k\) and \(\gamma\) are parameters, while the time-probability density is

\[
f(\tau) = \frac{\alpha \tau_0^\alpha}{\tau^{\alpha+1}}.
\]

As in previous methods, it is assumed that time variations of the conditional intensity function (23) stem from the time variations of all the nine parameters \(\alpha, \beta, \gamma, \lambda, k, E_C, M_{0C}, E_{\text{max}}, \text{ and } M_{0\text{max}}\), whose values for time \(t\) can be estimated from seismic events from the time window preceding \(t\). The details on estimation of these parameters can be found in the work of Kijko (1996). Having determined the parameters, the probability that an event will occur within a time interval \([t + \Delta t]\) after the last event, with the energy within \([E_C, E]\) and the seismic moment within \([M_{0C}, M_0]\), is

\[
Pr(E, M_0, t) = 1 - \exp \left\{ - \int_t^{t+\Delta t} \int_{E_C}^{E} \int_{M_{0C}}^{M_0} \Lambda(\xi, \zeta, \nu) d\xi d\zeta dv \right\}.
\]

Kijko and Sciocatti (1995b) used this approach to analyze a series of 292 events recorded for 153 days in an area of a stope at a gold mine in South Africa. They applied the time window of a constant length of 20 days, moved every 12 hours,
SEISMICITY INDUCED BY MINING

141

1.1
1.0
0.9
0.8
0.7
0.6
0.5
0.4
0.3
0.2
0.1

0 20 30 40 50 60 70 80 90 100 110 120 130 140 150
Time (days)

FIG. 43. Estimates of probability of the occurrence of seismic events with energy \( E \geq 5.5 \times 10^5 \) J in a stope area at a gold mine in South Africa, based on the three-parameter time dependent hazard assessment method. The time of the occurrence of observed events is marked by arrows. [Reprinted from Kijko (1996), Fig. 5.9.]

to estimate the probability of occurrence of an event with energy \( E \geq 5.5 \times 10^6 \) J in the next \( \Delta t = 10 \) days. It was also assumed that if the event was closer than 1 hour to the preceding event, it belonged to the same cluster. The results of the analysis are shown in Fig. 43, reproduced from Kijko (1996).

A simple model-free approach to the time variations of one-parameter representation of seismic energy distributions has gained some popularity in the Czech Republic and in Slovakia (e.g., Slavik et al., 1992; Rudajev, 1993; Holub, 1995; Kalenda and Pompura, 1996). A cumulative curve of the square root of the total energy released in consecutive time windows, called the Benioff graph, represents the seismic energy flow in the rock mass. Nonlinearity of this curve, in particular a faster increase followed by a certain slow-down, is considered as a precursor to a larger event. A typical example of such a graph is shown in Fig. 44, reproduced from Kalenda and Pompura (1996). The analysis was carried out for microseismic data from an area of one chamber at Jalsava magnesite mine in Slovakia. The nonlinear features of the graph, marked by rectangles, were correlated with the occurrence of rockbursts.

Not many papers are available that are devoted to testing the efficiency of prediction techniques based on a distribution of magnitude or energy. Lasocki (1993a) studied the correlation between the seismic hazard, estimated using linear (Eq. 11) and nonlinear (Eq. 18) models of energy distribution, and the actual hazard. The
study comprised five seismic series from different coal mines. The periods of observation were split into consecutive nonoverlapping time intervals. In every interval the expected number of strong events was estimated from the considered models. The significance of Spearman’s rank correlation between the series of estimated numbers and the series of actual numbers of strong seismic events was then evaluated. In all five cases, one case of nonlinear model and four cases of linear model, the correlation turned out to be highly significant. The significance of the overall correlation of results from all mines studied was greater than 0.99999. It was concluded that all the models considered for the energy distribution have high potential to estimate the actual seismic hazard.

In another work (Lasocki, 1994), correlation analysis was applied to investigate the potential for prediction of the technique of time-dependent seismic hazard assessment. The studies comprised eight local seismic series. The significance of correlation between the number estimated from relation (21) and the actual number of strong events was found to range from 0.956 to 0.99993.
SEISMICITY INDUCED BY MINING

Among the four parameters introduced by Stewart and Spottswoode (1993) for seismic hazard assessment, one was the estimate of the number of events corresponding to a given critical magnitude, $M_P$, obtained from the Gutenberg–Richter relation determined from 40 events. As a measure of the performance of every parameter the authors used the gain, which was defined as the ratio of the percentage of the total number of larger events occurring during the periods of the parameter anomaly to the percentage of the total time of the anomaly. This gain can be useful as an indicator of the prediction efficiency only if both the total time of anomaly and the number of events associated with the anomaly are considerable. Otherwise it can reach very large values by chance. The authors did not provide a global summary of the performance of the parameters but presented only some of the results. The parameter connected with the Gutenberg–Richter relation displayed a variable gain, dependent upon the definition of the anomaly, but in most cases the gain was greater than one—the level of randomness.

Kalenda (1996) considered the efficiency of prediction based on time-variations of $b$-value, the slope of Benioff graph, and the anomalous increase of event rate. He concluded from unequivocally defined rules of the game that 21 out of 25 strong events, identified in the studied data set as directly linked to mining, could have been predicted. The other strong events, identified as “regional,” were determined as unpredictable.

8.3. Parameterization of Spatial Distributions of Seismic Events

It is well known that seismic sources in mines are not distributed randomly in space but tend to concentrate in certain areas, forming clusters of various shapes. It is also known that this spatial distribution is not constant in time. Some of the time-dependent features of source distributions are clearly linked to the variable geometry of mining stope. The others seem to respond to internal changes during the process of rock mass fracturing.

A preparatory process leading to a rockburst may comprise different phases. At the early stage an external geometrical factor like a local geological discontinuity or mining remnants may start to affect the local stress field connected with the stope. This process may yield new stress concentration zones, which can be more seismically active. Such zones may be either isolated or connected with the zone of background activity surrounding the stope. As the preparatory process develops these zones may change both the shape and the location. In later stages of the process a certain clustering of events around a place of the future main event is expected (e.g., Ohnaka, 1992; Xie, 1993). This effect implies, in general, some ordering of the distribution of seismic sources, such as alignments, coplanar tendencies, etc.
Attempts to use the temporal changes of space distribution of mining seismicity in prediction of strong events have been undertaken for a long time. At first, they were limited to visual inspections of hypocentral maps and cross sections (e.g., Blake, 1980; Brink and O'Connor, 1984; Sato and Fujii, 1988). A more sophisticated quantitative approach was introduced during the 1990s. The reasons for growing interest in the studies of spatial distributions of seismic sources are obvious. The mechanism linking the distribution changes with increasing hazard seems to be straightforward. The expected changes also seem to be readily identifiable. Eneva and Ben-Zion (1997) even suggested that spatial distribution could be as informative as magnitude distribution. Finally, since source location is an independent parameterization of the seismic series, the analysis of its distribution could provide an independent hazard estimate, which could be eventually used, together with information coming from other parameterizations, in a multivariate prediction.

It is well documented that spatial distribution of seismic sources has a statistical fractal character. The fractal dimension quantifies a level of clustering of objects that build the fractal. The evolution of the fracturing process prior to the main event is expected to modify the geometry of source distributions. We hope, therefore, that these modifications are reflected in changes of the fractal dimension or multifractal spectrum. Since it is speculated that the distribution changes tend to increase its orderliness, then a drop in the fractal dimension may indicate an impending large event.

Coughlin and Kranz (1991) were probably the first who used fractal analysis to study temporal variations of the spatial distribution of microseismic events induced by mining. Using the data from two stopes at Galena mine, USA, from the time periods comprising damaging rockbursts they showed qualitatively that the fractal correlation dimension (relations 1 and 2) varies in time. On at least two occasions, a decrease in the dimension prior to a rockburst was noted. It was also observed that the periods of intense blasting or just following the main bursts could be associated with a decrease of fractal dimension.

Xie and Pariseau (1992) and Xie (1993) reported a significant correlation between the decrease of the fractal clustering dimension (relation 3) of spatial distribution of small events and the occurrence of significant failures of mined rock mass. Figure 45, reproduced from Xie (1993), shows the clustering dimension for five consecutive days preceding a rockburst in a pillar at Galena mine. Prior to the rockburst the spatial distribution of seismic sources reached the lowest fractal dimension. The other study of microseismicity data from a United States deep coal mine indicated a close correspondence between the low fractal dimension and the occurrence of rockbursts or bumps, whereas the dimension maintained high values in the peaceful periods of mining (Xie, 1993).

These encouraging results have evoked interest in using the spatial fractal dimension of sources as a precursor to strong events. The fractal dimension was considered as the only precursory parameter (e.g., Mortimer and Lasocki, 1996a,b;
FIG. 45. Time variations of clustering dimension of spatial distribution of microseismic events for a pillar at Galena mine in USA for 5 days before the occurrence of a rockburst. [Xie (1993), Fig. 14.42, reprinted from Fractals in Rock Mechanics, Geomechanics Research Series I, 464 pp., EUR 93.50/US $110.00/GBP66, A. A. Balkema, P.O. Box 1675, Rotterdam, Netherlands.]

Mortimer, 1997; Mortimer and De Luca, 1999) or as combined with other parameters describing the distribution of events (e.g., Eneva and Young, 1993; Stewart and Spottiswoode, 1993; Eneva and Villeneuve, 1997; Eneva, 1998; Lasocki and Mortimer, 1998).

Stewart and Spottiswoode (1993) introduced four parameters for seismic hazard assessment. One was related to the distribution of magnitude and another was the median seismic stress drop. The remaining two were connected with the spatial distribution of seismic sources. These are the reciprocal of the spatial fractal dimension estimate and the reciprocal of the median distance between event pairs. They were considered to represent different features of source geometry, namely the orderliness and scale of clustering of events. Daily values of these parameters were estimated from the seismic series recorded at an experimental site of a preconditioning experiment at the Blyvooruitzicht gold mine in South Africa (Adams et al., 1993). Every single value was determined from 40 seismic events. All parameters exhibited considerable variations in time. Stewart and Spottiswoode (1993) used their gain over the randomness to estimate the efficiency of introduced parameters. In the examples of the results they demonstrated that if the percentage of the total time of anomaly was kept constant at 20%, the gain was 1.95 for the reciprocal of the fractal dimension and 1.65 for the reciprocal of the median distance between the events. The other two parameters were more efficient. Their respective gain was 2.55 for the parameter quantifying the distribution of magnitude and 2.8 for the median stress drop. It was concluded that the introduced parameters, including the parameters related to the spatial distribution of events, have certain potential to assess time-dependent hazard. Although these parameters
represent different features of rock mass fracturing, they behave in a complementary way, and they should be used therefore in combination rather than individually.

Eneva and Young (1993) studied 1654 seismic events recorded during a 2-month period in a volume of $400 \times 400 \times 180 \text{ m}^3$ at the Creighton nickel-copper mine in Canada. Six events of magnitude not smaller than 2.0 occurred within the studied volume. Two others were located within a margin of 150 m outside of this volume. The analysis comprised estimation of time changes of the degree of spatial nonrandomness and spatial fractal correlation dimension. The degree of spatial nonrandomness measures the discrepancy between the distribution of actual interevent distances and the distribution of distances between points simulated uniformly within the volume occupied by the actual events. For any $N$ points in a volume this degree of nonrandomness for the distance range $[a, b]$ is evaluated as

$$c_{[a,b]} = \pm 100\% \sqrt{\sum_{i=a}^{b} \frac{\Delta P_i}{P}}, \quad (29)$$

where $\Delta P_i$ is the excess (plus) or deficiency (minus) of event pairs with respect to the reference random distribution in the distance interval $i$, and $P = N(N-1)/2$ is the total number of pairs for $N$ events. The summation extends over all intervals from the one to which $a$ falls ($i_a$) to the last one to which $b$ falls ($i_b$). The authors evaluated the degree of nonrandomness over the range $[0, 76 \text{ m}]$ and the correlation dimension from overlapping groups of 100 events moved with a step of 20 events. The values obtained were considered to be associated with the times of occurrence of the last event in each group. Figures 46 and 47, reproduced from Eneva and Young (1993), present the results of the analysis, where solid lines and filled

![Figure 46](image-url)
Fig. 47. Temporal variations of the correlation exponent dimension of seismicity recorded for 2 months in a selected volume at Creighton nickel-copper mine in Canada. The solid and dashed lines show the results for induced events and blasts, respectively. The arrows mark larger events located within the volume (large arrowheads) and in the marginal zone outside the volume (small arrowheads); their magnitudes are also marked. [Eneva and Young (1993), Fig. 8, reprinted from Rockbursts and Seismicity in Mines, Proceedings of the 3rd International Symposium, Kingston, Ontario, 16–18 August 1993. 462 pp., EUR 154.00/US $181.00/GBP108, A. A. Balkema, P.O. Box 1675, Rotterdam, Netherlands.]

Squares represent the values of the degree of spatial nonrandomness and the fractal correlation dimension, respectively. Dashed lines and open squares represent the same quantities for the events identified as blasts. The vertical arrows mark the days of occurrence of strong events, either located within the volume under the study (large arrowheads) or located in the marginal zone outside of the volume (small arrowheads). The authors concluded that an increased degree of nonrandomness preceded most of the larger events. They also identified two decreases of the fractal correlation dimension prior to larger events, two increases after the events, and one pair of larger events not associated with any change of this parameter.

Using the same technique to process data from the same volume of Creighton mine but from a different time interval, Eneva (1998) applied an original procedure to determine the prediction efficiency of the analyzed parameters. The procedure called the association in time (Eneva and Ben-Zion, 1997) compares the time of appearance of local extremes of the tested parameter with the time of occurrence of strong events. A separate count of local maxima and local minima in consecutive days preceding and following every strong event, respectively, provides the association frequency for the parameter in four categories: preceding minima, preceding maxima, following minima, and following maxima, respectively. If an extreme occurred between two events, it is considered to be associated with the event closer in time. In this way the classification of the extremes into “preceding” and “following” is equivocal.

In addition to the degree of spatial nonrandomness and spatial correlation dimension, Eneva (1998) studied the time interval of occurrence of a constant number of events, as a third candidate for the precursor to strong events. Moreover, since the results of previous studies were ambiguous with regard to the sign of anomaly
of the tested parameters, a possibility was tested of local maxima as well as local minima preceding strong events. The results were not unique and were strongly dependent on the data taken for the analysis. From 20 to 80% of the larger events were preceded by a minimum degree of spatial nonrandomness over the range [0, 75 m]. From 40 to 80% of such events were preceded by the maximum of the spatial correlation dimension, and from 60 to 80% by the maximum of the time interval. These results suggested, contrary to the earlier expectations, that the nonrandomness and orderliness of spatial distributions decrease prior to a main failure. The observed drop of event rate could support the well-known hypothesis of a preceding quiescence. Because the analysis is based, however, on only nine strong events that occurred during the period of study, these suggestions are not conclusive.

Eneva and Villeneuve (1997) analyzed extended databases from the same area, covering 3.5 years of observations. Apart from the already described parameters related to the spatial distributions and event rate, the study considers two other quantities connected with blasting. The potential for the prediction of all parameters was determined with the use of the association in time procedure. Although, in selected time intervals, the particular parameter performance was very good (e.g., precursory maxima within 6 days before 90% of the larger events), no simple and stable precursory patterns could be identified over long periods of time. It was therefore suggested using for prediction a combination set of parameters rather than relying on any single parameter. More than 70% of the larger events were found to be preceded, with 10 days lead-time, by the local extremes of at least two out of the considered six parameters.

The use of spatial and temporal fractal dimension estimates to assess the local time-dependent seismic hazard in mines was investigated by Mortimer and her colleagues (e.g., Mortimer and Lasocki, 1996a,b; Mortimer, 1997; Lasocki and Mortimer, 1998; Mortimer and De Luca, 1999). The studies were performed on seismic series recorded in the direct vicinity of mining longwalls. The source locations were related to a local system of coordinates connected with the advancing front. Three estimators of spatial fractal dimension were analyzed: the box dimension (Eq. 4), the clustering dimension (Eq. 3), and the correlation dimension (Eqs. 1 and 2). The correlation exponent dimension represented the temporal distribution of seismic events. The studies carried out on numerous seismic series from various mines showed that fractal dimensions significantly and systematically vary in time. It was also shown that the results strongly depend on sample size and the applied estimator. Figure 48, reproduced from Lasocki and Mortimer (1998), shows time variations of the box fractal dimension ($D_b$) and the correlation dimension ($D_2$) of spatial distribution of epicenters. The estimates are based on data samples comprising 100 (left-hand side graph) and 50 (right-hand side graph) consecutive events. Although they show some degree of correlation, $D_b$ and $D_2$ curves are rather different. A smoothing effect imposed by the larger sample size is also visible.
FIG. 48. Temporal variations of the correlation ($D_2$) and box ($D_b$) dimensions of the distribution of epicenters for a seismic series from a longwall area at Katowice coal mine, Poland. The estimates were obtained from 100-event (left) and 50-event (right) samples, respectively. [Lasocki and Mortimer (1998), Fig. 3, reprinted with kind permission from Trans Tech Publications.]
A correlation analysis was used to test relationship between the changes of fractal dimension and the changes of seismic activity and seismic hazard. The analysis took into account the spatial and temporal dimensions, on one hand, and the total number of events, number of strong events, total seismic energy, seismic energy released by strong events, maximum energy of one event, and the b-value, on the other. The results related to the spatial correlation dimension were ambiguous and erratic. The only significant positive correlation dimension were ambiguous and erratic. The only significant positive correlation was found, in most cases, between the fractal temporal dimension and the number of strong events and between the fractal temporal dimension and the sum of energy of strong events (Mortimer, 1997; Mortimer and De Luca, 1999).

In general, all the results strongly indicate that the variability in time of the spatial fractal dimension of seismicity is directly connected with mining. Nevertheless, it is still an open question as to what is behind these time variations and how much they are related to changes in seismic hazard. It seems that some ambiguity of described results is connected with properties still not fully recognized relating to the fractal dimension estimation procedures.

Investigation of two other simple statistical parameters of epicenter distributions was initiated by Lasocki (1996) and continued by Laskownicka (1998, 1999). These parameters are the direction of the best linear fit to a constant number of epicenters and the root-mean-square dispersion of epicenters around the fitted line. Statistical differences between the values of parameters prior to the occurrence of strong events and the values evaluated during quiet periods were tested. Altogether, 10 seismic series from different stopes at two Polish coal mines, 2 series from two Polish copper mines, and 2 series from a South African gold mine were analyzed. The significance of the differences between mean values taken from five consecutive days of the respective quiet and strong event preceding periods was tested by the Mann–Whitney and Wilcoxon tests. Of 14 examined series, the differences of the directions of a straight-line fit were significant in 12 cases and the differences of the root-mean-square dispersion were significant in 11 cases. This highly positive result points to the prediction potential of parameterizations of the spatial distributions of seismic events, although a specific working technique based on such parameterizations is a matter of further studies.

8.4. Rock Mass Instability Concept

Seismic energy and seismic moment are independent, although strongly correlated, parameters of a seismic event. Events in mines of the same seismic moment could differ in seismic energy by three orders of magnitude (e.g., Mendecki, 1993). Consequently, it has been attempted to ground the time-dependent hazard assessment on both parameters simultaneously. The technique introduced by Kijko and
Sciocatti (1995b), described in Section 8.2, which integrates the information provided by probabilistic distribution of seismic moment and energy and accounts for possible correlation among the events, represents such an approach. Mendecki and his colleagues (e.g., Mendecki, 1997b,c; van Aswegen et al., 1997) developed the same initial idea into a complex quasi-multivariate methodology of seismic data analysis. A brief presentation of general concepts and applied parameterization of seismicity is given. For full details see the book edited by Mendecki (1997a).

The apparent stress

$$\sigma_A = \mu \frac{E}{M_0}$$

(30)

where \(\mu\) is the modulus of rigidity, is a parameter that unites the seismic energy \(E\) and seismic moment \(M_0\). By definition, it is sensitive to variations of the energy-to-moment ratio rather than to individual changes of these parameters. Following Gibowicz et al. (1990), apparent stress is used as a model independent indicator of stress release or, in general, of the stress state within the rock mass (e.g., Mendecki, 1993). Stress release estimates for low magnitude seismicity to determine the zones of an increased rockburst potential was also used by Urbancic et al. (1992b) and Urbancic and Young (1993).

A strong linear correlation between the log \(E\) and log \(M_0\) results in the correlation between log \(\sigma_A\) and log \(M_0\). The regression lines of log \(E\) versus log \(M_0\) and log \(\sigma_A\) versus log \(M_0\), estimated for a given seismic series, represent the expected value of seismic energy and apparent stress, respectively, given the seismic moment of an event. An anomalous apparent stress for an individual event becomes more distinct when \(\sigma_A\) of this event is compared with its expected value. For this, a parameter called stress index, \(I_s\), (Glazer, 1998) or, alternatively, energy index, \(EI\), (van Aswegen and Butler, 1993) was introduced for practical application in deep gold mines in South Africa. \(I_s\) is defined as the ratio of the apparent stress of an event to its expected value evaluated from a log \(\sigma_A\) versus log \(M_0\) regression line. \(EI\) is the ratio of seismic energy of that event to its expected value for a given seismic moment of the event. Glazer (1998) presented the examples of positive anomalies of stress index, routinely used to delineate the areas of higher than average rockburst potential.

The apparent volume, \(V_A\), introduced by Mendecki (1993), is another parameter expressing the relation between the energy and the moment of an individual event

$$V_A = \frac{M_0}{2\sigma_A} = \frac{M_0^2}{2\mu E}.$$  

(31)

Apparent volume is supposed to be a measure of the rock volume with inelastic strain, associated with the seismic event.

The concept of prediction of Mendecki and his group (e.g., Mendecki, 1997b,c) relies on monitoring the process of rock mass fracturing by a set of parameters of
observed seismicity in order to recognize when the fracturing becomes unstable. Such an unstable fracturing eventually leads to a large event or a swarm of smaller events. To ascertain the instability, the following criterion is used

\[ \int_{AV} \sigma \varepsilon_e dV + \int_{AV} \sigma \varepsilon_{in} dV < 0, \]  

(32)

where \( \sigma, \varepsilon_e, \) and \( \varepsilon_{in} \) are the stress, elastic strain, and inelastic strain rates, respectively, and the integration extends over the volume of interest \( AV \). Since the term with elastic strain must be positive, the fracturing becomes unstable when the second term with inelastic strain is negative and balances the positive term. It is assumed that both the stress and inelastic strain changes can be inferred from parameters related to the seismic strain rate and the seismic stress, which are defined as

\[ \dot{E} = \frac{\sum_{t_i}^{t_2} M_0}{2 \mu \Delta V \Delta t} \]  

and

\[ \sigma_s(\Delta V, \Delta t) = \frac{\sum_{t_i}^{t_2} E}{\dot{E} \frac{\Delta V}{\Delta t}} = 2 \mu \frac{\sum_{t_i}^{t_2} E}{\sum_{t_i}^{t_2} M_0} \]  

(34)

respectively, where the summation extends over all events recorded within the volume \( \Delta V \) during the period \( \Delta t = t_2 - t_1 \). For instance, the stress rate in relation (32) is estimated from the changes in an average value of the energy index \( EI \) and the strain rate is inferred from the changes in the cumulative apparent volume \( \sum V_A \). A large event is expected when the average \( EI \) drops down together with an anomalous increase of the cumulative apparent volume.

There are several other parameters proposed by Mendecki and his group, such as seismic viscosity and relaxation time, seismic softening and seismic diffusion, and seismic Schmidt and seismic Deborah numbers, to monitor the instability of the rock mass (Mendecki, 1997a). These parameters, however, are strongly redundant since they are mostly related to only two independent quantities: the sums of seismic energy and seismic moment. It seems that the authors multiplied nearly the same parameterizations of seismic series in order to provide different options for a pair of quantities, which could be used for practical monitoring of the instability criterion (Eq. 32). In most of their examples (e.g., Mendecki 1997b,c; van Aswegen et al., 1997), the cumulative apparent volume is used as the one parameter in the pair while the other parameter is different in different cases. This procedure suggests that the authors ascribe a particular significance to the variations of apparent volume. The use of cumulative quantities, on the other hand, obscures fine variations of this parameter and the interpretation can be ambiguous.

This briefly described approach to prediction, based on the rockmass instability concept, is widely used in real-time applications in some South African mines.
SEISMICITY INDUCED BY MINING

(e.g., Butler, 1997; Glazer, 1997, 1998; Minney et al., 1997; van Aswegen et al., 1997; Alcott et al., 1998) and is considered a valuable tool in monitoring variable seismic hazard. Unfortunately, the proposed parameters have not been objectively validated so far and their actual prediction efficiency is not known.

8.5. Other Studies

The techniques described in the previous section are based to some extent on a possible precursory potential of the apparent stress. Such expectations related to stress release premonitory changes led to other studies of this and other stress release estimates. Cichowicz et al. (1990) investigated various parameters of microevents recorded ahead of a stope at Western Deep Levels gold mine in South Africa, including the static and dynamic stress drops. They noted that static stress drops of individual microevents were higher before than after the main event.

Stewart and Spottiswoode (1993) tried to use the seismic stress drop as an indicator of the time-dependent seismic hazard. For this purpose they introduced the parameter $\tau = (Rv)^3/E_s$, where $Rv$ is the peak ground-velocity scaled by hypocentral distance and $E_s$ is the seismic energy, which is proportional to the stress drop. The authors speculated that an increase of the stress drop from events located ahead of an advancing mining front could indicate fracturing of previously solid rock in which a stronger event is likely to occur. The tests performed on seismic data from the Blyvooruitzicht gold mine in South Africa, in which time variations of the median value of $\tau$ for 40 consecutive events were analyzed, have shown in some cases reasonable correlation between the periods of high stress drop and the times of occurrence of larger events.

Gibowicz (1998) studied the stress drop mechanism of seismic events from the Underground Research Laboratory in Canada, Western Deep Levels gold mine in South Africa, and two coal and two copper mines in Poland. He used the parameter $\varepsilon$ of Zúñiga (1993) defined as the ratio of the static stress drop over the sum of apparent stress and half the static stress drop. For the partial stress drop mechanism $\varepsilon < 1$ while the frictional overshoot mechanism corresponds to $\varepsilon > 1$. It was found that strong events occur when the partial stress drop mechanism is dominant (low values of $\varepsilon$), and then they are followed by the events for which the frictional overshoot mechanism begins to dominate (high values of $\varepsilon$).

It is well known that seismicity in mines depends strongly on mining; the dependence is complex and is not fully understood quantitatively. Several attempts, however, have been made to relate seismic parameters with some parameters describing mining operations, and then to come up with quantities sensitive to the variations of seismic hazard, which would depend upon these mining parameters. One such attempt estimates seismic hazard parameters from the total seismic energy released within a time window and the volume of rock mined out during this
time (e.g., Głowacka et al., 1988, 1992; Głowacka and Kijko, 1989; Głowacka, 1993). The technique has its origin in McGarr’s (1976) relation between the sum of seismic moments and the volume of convergence accompanying elastic deformation around a stope, and is based on Kijko’s (1985) reformulated relation showing that

\[ \sum E = C \cdot V_e^B, \]  

(35)

where \( \sum E \) is the sum of seismic energy, \( V_e \) is the volume of excavated rock, and \( C \) and \( B \) are parameters characterizing the state of the rock mass. Relation (35) is used to evaluate the anticipated value of seismic energy released within the time interval \([t - \Delta t, t]\). According to Głowacka (1993) this anticipated value is

\[ (\sum E(t - \Delta t, t)) = \left\{ C[V_e(t)^B - V_e(t - \Delta t)^B](1 - e^{-\frac{\Delta t}{\tau_e}}) + \Delta E(t - \Delta t)\right\}(1 - e^{-\frac{\Delta t}{\tau_e}}), \]  

(36)

where \( V_e(t) \) is the volume of excavated rock up to time \( t \), \( \Delta E(t - \Delta t) \) is the energy excess, which was not released up to time \( t - \Delta t \), the exponential term represents time effects of energy loading and relaxation, and \( \tau_e \) is a constant. It is assumed that the actual sum of seismic energy is a random variable of either normal distribution, truncated from the left by zero (e.g., Głowacka et al., 1988, 1992; Głowacka and Kijko, 1989), or lognormal distribution (e.g., Głowacka and Lasocki, 1992; Głowacka, 1993). In every case the parameter of location of the distribution is expressed by relation (36) or a similar one, and the method provides the probability, \( \text{Pr}[\sum E(t - \Delta t, t) > E_p] \), that the sum of seismic energy will exceed a specified value \( E_p \) within \([t - \Delta t, t]\). For a short time interval \( \Delta t \), this probability can be understood as the probability that an event of energy greater than \( E_p \) will occur. The probability increase and an appearance of the excess of energy are supposed to precede the occurrence of a strong event. This approach was successfully tested in some coal mines in Poland (e.g., Głowacka and Kijko, 1989), in the Czech Republic (e.g., Głowacka et al., 1992), and in the USA (Arabasz et al., 1997).

Although it is obvious that integrating various precursory parameterizations of seismic series into a true algorithmic multivariate prediction would considerably increase the efficiency of prediction, only a few attempts in this regard have been reported. Głowacka (1992, 1993) and Głowacka and Lasocki (1992) presented one such multivariate approach. A probabilistic synthesis of precursors, proposed by Aki (1981), was used to integrate the results from the method based on the volume of excavated rock with the results from the time-dependent seismic hazard analysis based on the distribution of energy (see Sect. 8.2). Let \( \text{Pr}[\sum E(t - \Delta t, t) > E_p] \) be the probability provided by the former method and \( R(E_p, t) \) be the probabilistic outcome of the time-dependent seismic hazard analysis (21). The probability of occurrence of an event with energy greater than \( E_p \), estimated by both methods
taken together is given by

$$\Pr[E > E_p] = p_0 \frac{\Pr[\sum E(t - \Delta t, t) > E_p]}{p_0} \frac{R(E_p, t)}{p_0}.$$  \hspace{1cm} (37)

where $p_0$ is the a priori probability of occurrence of the event, estimated by the ratio $n(E_p)/N$, where $n(E_p)$ is the number of time intervals $\Delta t$ in which such an event occurred and $N$ is the number of time intervals in the whole studied period. Relation (37) can be readily used if the component methods of hazard assessment are statistically independent and the component probabilities are low. An example of the use of this technique in assessing time-dependent seismic hazard is shown in Fig. 49, reproduced from Glowacka and Lasocki (1992). The

Fig. 49. Probabilistic synthesis of the results of two time dependent seismic hazard assessment methods for a longwall area at Bobrek coal mine, Poland. (a) Independent estimates of probability of the occurrence of seismic events with energy $E \geq 3 \times 10^5$ J; dotted line (1), a priori probability of the event occurrence, continuous line (2), probability estimates based on the volume of extracted rock, dashed line (3), estimates based on the distribution of energy. (b) Synthesized probability of the occurrence of seismic events. The time of the occurrence of observed events is indicated by arrows. [Glowacka and Lasocki (1992), Fig. 1, reprinted with kind permission from the Institute of Rock Structure and Mechanics, the Academy of Sciences of the Czech Republic.]
analyzed data were associated with a longwall at Bobrek coal mine in Poland. The estimates of $\Pr\{\sum E(t - \Delta t, t) > 3.10^5 J\}$ marked by a continuous line and of $R(3.10^5 J, t)$ marked by a dashed line in the upper graph of the figure can be compared with the synthesized probability $\Pr\{E > 3.10^5 J, t\}$ shown in the lower graph. The arrows mark the occurrence of events with energy $> 3.10^5 J$, and a horizontal dashed line in the upper graph represents the a priori probability $p_0$. It can be seen that the synthesis caused, in some cases, a significant increase in the probability estimates.

Marcak (1993) presented another example of multivariate prediction. He applied the pattern recognition algorithm, well known in earthquake seismology (e.g., Keilis-Borok and Kossobokov, 1990; Keilis-Borok and Rotwein, 1990), to seismic data from Halemba coal mine in Poland. In his example, seven out of nine stronger events occurred during the time of increased probability, whose total duration was 10.4% of the total time of analysis.

8.6. Automatic Prediction: Extrapolation of Parameterizations of Seismicity in Mines

All the techniques discussed estimate certain parameterizations of seismic series, which are supposed to indicate the potential to generate events hazardous for mining. Past-to-present time variations of these estimates are used to evaluate the future potential. The inference is usually made by a qualitative extrapolation of the most recent trends of estimates. From the formal point of view all such methods assess the present seismic hazard and do not predict a future one, since no formal algorithms of extrapolation into the future are provided.

The problem of automatic prediction of mining seismicity has been formulated in terms of either linear (e.g., Buben and Rudajev, 1974, 1991; Rudajev and Fučík, 1982) or nonlinear (e.g., Čič and Rudajev, 1996, 1999; Rudajev and Čič, 1999) statistical extrapolations. The methods of such an extrapolation are usually multi-channel, i.e., they can accept more than one identically timed time series as input data. Thus such methods can conveniently integrate different parameterizations of seismic series, as well as other, nonseismological, data. The results of their application are the estimates of future values of desired parameterization(s).

A linear extrapolation requires that the input time series result from a stationary Markov process. The future values are modeled by a linear combination of the past values. Buben and Rudajev (1991) used Wiener extrapolation to various combinations of input data comprising daily number of events, daily energy release, daily values of convergence in investigated stopes, and the number of seismoacoustic pulses recorded daily in the coal seam and in the overlaying rocks. If $\tilde{x}_k = [x_k^{(1)}, x_k^{(2)}, \ldots, x_k^{(m)}]^T$ represents the input values for time $k$, where the superscript denotes the specific input time series, the predicted values $\hat{x}_t$ for time $t$, ...
according to the linear extrapolation, are

\[ \hat{x}_t = [\hat{x}_{t-1}, \hat{x}_{t-2}, \ldots, \hat{x}_{t-L}] \cdot \bar{g}, \]  

(38)

where \( \bar{g}_k = [g_1, g_2, \ldots, g_L]^T \) is the vector of filter coefficients and \( L \) is its length. The filter coefficients are determined by minimization of the sum of squared errors of prediction

\[ \sum_t (\hat{x}_t - \hat{x}_t)^T (\hat{x}_t - \hat{x}_t) = \text{min}. \]  

(39)

Buben and Rudajev (1991) attempted to predict the daily number of events and the sum of their energy using various versions of the method and filters of a length up to \( L = 5 \). They noted that prediction of the number of events was more accurate than prediction of energy released daily, but the overall result was not encouraging.

The main drawback of linear extrapolation rules is the stationarity condition, which in general is not fulfilled in the series generated by seismicity in mines. Nonlinear extrapolation methods make use of artificial neural networks. Formally they do not require stationarity of input time series and therefore seem to be more appropriate for the data from mines. Rudajev and Čiž (1999) applied two separate neural networks with three layers—input, hidden, and output—to predict the number of events and the sum of maximum amplitudes of their signals for consecutive 1-day periods. The latter quantity represents to some extent the daily release of seismic energy. The response \( \{y_j\}, j = 1, \ldots, M \), of neurons of the hidden layer to the input pattern \( \{x_i\}, i = 1, \ldots, N \), was modeled by a nonlinear sigmoid function

\[ S(\bullet) = \tanh(\bullet), \]  

(40)

where \( N \) is the dimension of the input vector, equal to the number of input neurons, \( M \) is the number of neurons in the hidden layer, \( w_{ij} \) are weights among neurons from the input and hidden layers, and \( b_j \) are the thresholds for hidden neurons. Since Rudajev and Čiž (1999) used the networks with only one output neuron, the net response of the input pattern was

\[ z = \sum_{j=1}^{M} (w_{j*} y_j - b), \]  

(41)

where \( w_{j*} \) are the weights between the hidden and output layers and \( b \) is a threshold for the output neuron. Here the input pattern denotes all values supplied simultaneously to the network, which could represent any combination of data from the same or different sources (various seismicity parameterizations, nonseismic data, etc.). The study was performed on seismic data recorded for 1100 days at Mayrau mine in the Czech Republic. The seismic input data—the daily number
of events and the sum of maximum amplitudes—were supplemented by the total daily length of active mining. The whole data series was split into training and testing parts; a usual procedure for neural networks. The training part was used to evaluate an optimal set of weights minimizing the sum of squares of differences between the obtained and desired output from the network. The testing part was used to evaluate the effectiveness of the technique. Figure 50, reproduced from Rudajev and Čič (1999), shows a comparison between the predicted and actual numbers of events for consecutive days of the testing part. The method turned out to be robust with regard to the apparent nonstationarity of the input series. Its effectiveness was measured on the testing part by a relative mean-squared error

$$RMSE = \frac{\sum_i (PRED_i - OBS_i)^2}{\sum_i [MEAN(OBS) - OBS_i]^2},$$

(42)
where $PRED$ are the predicted values, $OBS$ are the observed values, and $MEAN$ ($OBS$) is the mean of all the observed values. Prediction is efficient when the $RMSE$ is significantly less than one and for perfect prediction this quantity would be equal to zero. $RMSE$ was about 0.55 for prediction of the number of events and 0.79 for prediction of the sum of maximum amplitudes. This difference in the accuracy of prediction has been explained by a lower accuracy of the amplitude data.

In another work, Čiž and Rudajev (1999) compared the prediction of the numbers of events per day and the daily sum of maximum amplitudes, based on a linear extrapolation and three different neural network architectures. For the tested data, the linear Wienner extrapolation and nonlinear prediction were equally efficient, although some differences in efficiency for different neural networks were noted. The statistical extrapolation of the number of events was 40–46% more efficient than the prediction by the average value and 27–42% more efficient than the prediction by the last value. The increase of efficiency in the case of the sum of maximum amplitudes was 20–26 and 33–38% respectively.

9. **Seismic Discrimination**

Large seismic events in mines are of interest to seismologists monitoring compliance with the Comprehensive Test Ban Treaty (CTBT) prohibiting nuclear explosion testing. Mining-induced events often appear explosion-like for many discrimination criteria between earthquakes and underground nuclear explosions (e.g., Bennett and McLaughlin, 1997; Bowers, 1997; Bowers and Douglas, 1997). Mine tremors occur at depths similar to underground nuclear explosions. The ratio of the body wave magnitude and surface wave magnitude from large seismic events in mines is more like that observed from an explosion than an earthquake. Short period $P$ waves recorded at teleseismic distances appear simple as those observed from many underground explosions (Bowers and Douglas, 1997). Such results could lead to accusations of evasion under the CTBT (Blandford, 1996).

For CTBT research, much attention is paid to the study of chemical explosions (e.g., Denny et al., 1996; Murphy, 1996). The large number of mining blasts can cause problems for any monitoring system. In addition, large mining blasts could be used to hide a simultaneous decoupled nuclear explosion (e.g., Smith, 1993; Taylor, 1994). Some mining experts believe that a rockburst could be triggered by suddenly removing a few last supports in a large excavation. This leads to the possibility of hiding a decoupled explosion in a rockburst as well (Blandford, 1996).

McGarr et al. (1990) investigated seismic records of three mine tremors and three explosions in deep gold mines in South Africa to test the methods of discriminating between the two types of events. All three explosions generated $S$ waves that appeared to be a consequence of deviatoric stress release in the source area. The
three tremors generated $P$ and $S$ waves whose spectra and source parameters agree well with standard earthquake source models. For events of fixed low-frequency spectral asymptotes of $P$ waves, the explosions typically have higher corner frequencies than natural earthquakes or mine tremors.

Taylor (1994) raised the problem of false alarms associated with mining activities. He studied, as an example, an anomalous event from the Gentry Mountain coal mining district in the eastern Wasatch Plateau, Utah. The magnitude 3.5 event occurred on May 14, 1981, and was consistently classified as an explosion based on various regional discriminants. Using a simple source representation, Taylor (1994) modeled the event as a tabular excavation collapse that occurred as a result of normal mining activities and had the signal characteristics of a vertical point force. His study raises the importance of having good catalogs of seismic data and mining information from various areas to prevent potential false alarms that may be generated from seismicity in mines.

Many spontaneously occurring shallow mine collapses have implosional source mechanisms that might provide a physical basis for discriminating them from explosions. Another kind of underground mine collapse, induced by explosively demolishing the natural pillars that support underground mine opening (Yang et al., 1998), is described in Sect. 5.2.

Occasionally, large mining events are misinterpreted as natural earthquakes. Fan and Wallace (1995) studied the October 30, 1994, seismic event of body wave magnitude 5.6 and surface wave magnitude 4.7 from South Africa, when reports from the mining region were not yet available. They suggested that the event was unlikely to be a mine tremor because of its large magnitude and midcrustal depth. They determined the source mechanism and interpreted it either as representing plateau uplift or as evidence of a southward continuation of deformation related to the East African Rift system. Bennett et al. (1996), on the other hand, identified the 1994 event as a mining event, using observations of regional phase spectral ratios and high-frequency surface Rayleigh waves. The event in fact was caused by slip on the Stuirmanspan fault at a depth between 2 and 3 km in the President Brand mine in the Orange Free State gold mining district. Bowers (1997) has shown that surface waves from the 1994 event, recorded by broadband instruments at regional distances, are inconsistent with collapse mechanisms recently proposed for some large mining events. He found that modeling of teleseismic broadband $P$ waves and regional distance surface waves indicates that the source depth is shallow and the observed source radiation is consistent with a normal dip-slip fault. The explosion-like ratio of body wave over surface wave magnitudes is attributable to the effect of the mechanism and the shallow depth of the source.

Bowers and Douglas (1997) used short-period and broadband $P$ waves recorded at teleseismic distances to characterize large seismic events in different mining regions, such as South Africa, Eastern Europe, and North America. They found that mining events from South African gold and Polish copper mining districts could be
characterized by simple teleseismic $P$ waveforms that have dilatational first motions and short duration. The dilatational first motion, observed from many mining events, is a potential criterion for discriminating between mine tremors and underground explosions, as generally only compressional first motions are observed from explosions. These waveforms can be explained by a shallow double-couple source. But there are exceptions as well. Both the March 13, 1989, Volkershausen, Germany, and the February 3, 1995, Wyoming events generated surface waves that indicated a collapse mechanism and had broadband $P$ waveforms that provided some evidence of source complexity.

Bennett and McLaughlin (1997) investigated seismic characteristics and source mechanism of mining events for use in seismic discrimination. They used regional-distance observations from several large mine tremors in a variety of source areas, such as western and eastern USA, Poland, Germany, South Africa, and Russia. The signal analyses indicate that most mine tremors have $L_g/P$ ratios which are not explosion-like but tend to mix with the earthquake population. They also tried to find theoretical mechanisms corresponding to the observations and have shown that cavity collapse can be represented as a closing horizontal crack. Such a mechanism is a relatively inefficient source for long-period Rayleigh waves, possibly explaining the low values of surface wave magnitude often found for mining events, and shows relatively strong short-period surface waves consistent with observations. Bennett and McLaughlin (1997) concluded that although the tension crack model seems to explain observations from mine tremors with large collapse features, it might be expected that in some mining areas the source mechanisms would be better represented by more complex combinations of the closing crack with tectonic release or faulting components.

Thus, the occurrence of large seismic events in mining areas all over the world is one of the many problems facing seismic monitoring of nuclear explosions under the Comprehensive Test Ban Treaty. Some traditional seismic discriminants may not work for many mining events, but some others, especially those based on seismic signals at regional distances, seem to provide the means for discrimination of seismic events in mines from nuclear explosions (e.g., Bennett and McLaughlin, 1997). Further studies on the source mechanism are needed to identify those mechanisms of mine tremors that could be problematic for various seismic discriminants.

10. SUMMARY

Rock failure and seismic activity are often unavoidable phenomena in extensive deep mining. If the earth's crust is in a self-organized critical state, then a significant fraction of the crust is not far from instability. Pore pressure changes and mass transfers leading to small incremental deviatoric stresses, therefore, are
sufficient to trigger seismic instabilities in the uppermost crust in otherwise historically aseismic areas (Grasso and Sornette (1998)). Increases in the shear stress or decreases in the normal stress acting on the fault planes cause mining-induced events. The stress perturbations, often quite small, frequently result in generation of seismicity where near-failure conditions are present, which are believed to be common in both active and stable tectonic environments (McGarr et al., 1999).

Seismic monitoring provides a powerful means for the detection and evaluation of seismic events occurring around underground openings in numerous mining districts throughout the world. The most comprehensive studies of seismicity in mines are carried out in South Africa, Canada, the United States, and Poland. Routine seismic monitoring has been in use for over 30 years. Its main objective was immediate location of seismic events in mines, estimation of their strength, and prediction of large rock mass instabilities. The first two objectives were readily achieved, but prediction of large seismic events in mines is not as yet resolved. The introduction of modern digital seismic systems to mines, however, and progress in the theory and methods of quantitative seismology have led to the implementation of real-time monitoring as a mine management tool for estimation of the rock mass response to mining.

Location of seismic events is the first step in seismic monitoring in mines. In mining practice, the expected accuracy for hypocenter locations is a few tens of meters, or even a few meters in some cases. This demanding requirement could be met with the present-day technology when the optimum design of underground seismic networks and new location methods, now available, are applied in mining practice.

Seismic tomographic imaging is used to map the spatial variations of seismic and elastic parameters in rocks. Seismic wave velocity imaging is particularly useful for seismicity studies in mines. Velocity images can be calculated using either controlled explosive sources or passively monitored seismicity. Passive source velocity images can be computed by a simultaneous inversion procedure for source location and velocity structure. Detailed velocity structures obviously improve the accuracy of hypocenter location. Comparisons between velocity images and seismicity have shown that the largest events are associated with regions of high velocity of seismic waves, whereas low velocity areas are generally not seismic. Controlled studies have shown a strong correlation between velocity and stress.

Seismic events induced by mining are not uniformly distributed in either space or time. Extensive studies performed on the space–time–energy distributions of seismic events in mines show that the tendency to form nests, swarms, and clusters is commonly observed. Obviously, seismicity in mines is strongly affected by local geology and tectonics, and by interaction between mining and crustal state of stress on a local and regional scale. This is especially true in the case of the largest mine tremors, with magnitude close to 4 and above. Extensive studies on the relationship between geological features and seismicity show that seismic events concentrate
in specific regions corresponding to different geologic units, in spite of mining works being carried out in other regions as well.

Source parameters have been estimated for thousands of seismic events, in the moment magnitude range from -3.6 to over 5, in mining districts worldwide. They are used for various practical purposes, such as assessment of rockburst hazard and damage potential, combating rockburst hazard, design of support in rockburst-prone areas, and attempts at rockburst prediction.

Extensive studies of scaling relations of seismic events in mines, describing the manner in which the source duration or the source dimension increases with increasing seismic moment, indicate that stress drop shows globally no systematic dependence on seismic moment. The constant stress drop model implies a self-similar rupture process regardless of the strength of the seismic event. The average values of seismic moment referring to the same range of corner frequency, however, are vastly different in various mining areas.

The most important progress made in source mechanism studies of seismic events in mines during the last decade is the use on a massive scale of seismic moment tensor inversion technique. From numerous studies it follows that shearing processes are often dominant in the source and its mechanism can be well represented by a double-couple model. More complex rock failures associated with seismic events, especially with large ones, are also reported from various mining districts. Seismic events in mines are characterized by very low seismic efficiency; nearly all of the energy released by faulting is consumed in overcoming friction. The apparent stress seems to display distinct upper bounds determined by a maximum seismic efficiency of about 0.06 (McGarr, 1994, 1999).

Seismicity induced by mining exhibits a fractal structure with respect to the space, time, and event size. Fractal geometry seems to characterize distributions of various parameterizations of seismic events on every scale. Larger regional events, distributed for example over the Upper Silesian Coal Basin in Poland; all events from a given mine; seismicity accompanying mining of single isolated stopes; as well as microseismicity observed in one pillar or acoustic emission generated during laboratory experiments, are all statistical fractals in the space, time, and energy domains. Though not conclusive, such self-similarity of seismicity strongly suggests that the process of rock mass fracturing is governed by deterministic chaos. There are also indications that this process might be characterized by a low-dimensional attractor, and its short-term predictability could be feasible.

Statistical analyses of seismic series from mines provide another argument for predictability, in a sense, of rock mass response to mining. Seismicity associated with mining of individual stopes is non-Poissonian due to the time variability of generating processes in the long term and correlation among the events on the short time scale. Processing of seismic series in real time, although it cannot at present result in a precise reliable prediction, can provide useful general characteristics of the past, present, and future states of rock mass fracturing. This is accomplished
by a variety of approaches to the assessment of time-dependent hazard. Recent proposals in this respect are based on the time variability of recurrence relations, variations of spatial distribution of sources, rock mass instability criteria, changes of the apparent stress and stress drop, the probabilistic relation between the seismic energy release and the volume of extracted rock, among others. Attempts are also made to integrate particular methods in a multivariate forecasting system, including the application of neural networks for this purpose. Though this research is still in an early stage and indications provided by the present methods of hazard analysis do not comply with demands, the practical usefulness of these methods for combating rockburst hazard in mines has been widely acknowledged.

Large seismic events in mines are of interest to seismologists monitoring compliance with the Comprehensive Test Ban Treaty which prohibits nuclear explosion testing. These events often appear explosion-like for many discrimination criteria between earthquakes and underground nuclear explosions. Many mine collapses, on the other hand, have an implosional source mechanism that might provide a physical basis to discriminate them from explosions. Thus, some traditional seismic discriminants may not work for many mining events, but some others, especially those based on seismic signals recorded at regional distances, seem to provide the means for discriminating seismic events in mines from nuclear explosions.

ACKNOWLEDGMENTS

We are grateful to Art McGarr and Shahriar Talebi for their most helpful comments and suggestions, which considerably improved the original manuscript.

REFERENCES


