



Application of Power-Law Frequency Fractal Model for Recognition of Vertical Cu Distribution in Milloieh Porphyry Deposit, SE Iran

N.R. Ahmadi, M.R. Jafari*, M.R. Espahbod and N. Kohansal Ghadimvand

Department of Geology, North Tehran Branch, Islamic Azad University, Tehran, Iran

Received 10 February 2020; received in revised form 18 March 2020; accepted 19 March 2020

Keywords

Power-law frequency fractal model

Vertical distribution

Milloieh

Abstract

Identification of the vertical and horizontal distributions for elemental grades is of an important sign in different mineral exploration stages. The main aim of this work is to determine the vertical distribution directional properties of Cu values in the Milloieh Cu porphyry deposit, Kerman (SE Iran) using the power-law frequency fractal model. This work is carried out based on four mineralized boreholes. The Cu grade vertical distribution in mineralized boreholes indicates a positively skewed distribution in the former and multi-modal distribution in the latter types. The power-law frequency analysis in nature represents that the Cu values in four mineralized boreholes are bifractal. The two sections of these graphs outline a break point of about 0.5-1% for Cu values and fractal dimension range of 1.70-4.97 in the mineralized boreholes.

1. Introduction

Recognition of the distribution of ore elements in different axes (X, Y and Z), specifically at exploratory boreholes, is a significant study, which is essential to evaluate the quality and interpret the quantity of mineral resources/reserves in the mine planning and extraction method selection. In the past decades, the distribution nature of the ore and trace elements in various rock types has been studied in different ore deposits (Ahrens, 1966; Monecke et al., 2001; Zou et al., 2009; Afzal et al., 2011; Khalili and Afzal, 2018). The proposed law states that the frequency distribution of most ore elements in rock units and ore deposits is positively skewed and that the distribution of ores is lognormal. The distribution of geochemical elements in boreholes also shows power-law relationships, which can be fitted into the fractal models (Sanderson et al., 1994; Monecke et al., 2001; Li et al., 2003; Rashidnejad Omran et al., 2011; Zuo and Wang, 2016; Soltani et al., 2019; Mirzaei et al., 2020).

The fractal theory has been developed by Mandelbrot (1983) and widely applied since 1980s, and also applying it to geosciences based on the

fractal geometry has proved to be useful especially in geochemical explorations. This methodology has been mainly used to characterize the geological structures, some other features, and also for identification of the geochemical and geophysical anomalies and mineralized zones from background values (Cheng et al., 1994; Cheng, 1995; Carranza, 2009; Zuo, 2011; Zuo and Wang, 2016; Daneshvar Saein and Afzal, 2017; Afzal et al., 2018; 2019; Yasrebi et al., 2014; 2017). Furthermore, the fractal models are useful in mineral resource/reserve studies for detection of quantitative details of mineralization and mineral deposit parameters (e.g. Agterberg, 1993; Wang et al., 2006; Afzal et al., 2012; Karami and Afzal, 2015; Daneshvar Saein, 2017; Yasrebi and Hezarkhani, 2019). Zou et al. (2009) have applied three fractal models consisting of the vertical distribution characterization of Cu concentration in the Qulong copper deposit in Tibet, western China. Rashidnejad Omran et al. (2011); they have used this method in the eastern part of the Kahang porphyry deposit (Central Iran). On the other hand, the fractal methodologies based on the power law relationships have been used for

surface studies or in block models. However, investigation of the exploratory boreholes by this method is rare. In this work, the power-law frequency model is utilized to characterize the vertical distribution of Cu to evaluate the mineralization continuity based on the borehole datasets in the Milloieh copper porphyry deposit located in SE Iran.

2. Power-law frequency fractal model

In this work, the power-law frequency fractal technique is utilized to calculate the frequency distribution of ore grades. This model has been demonstrated by many researchers including Turcotte, 1996; Zuo et al. 2009; Rashidnejad Omran et al., 2011, and has the common form of:

$$(N \geq c) \propto c^{-D} \tag{1}$$

$$\log(N \geq c) = C - D \log(c) \tag{2}$$

where $N(\geq c)$ is the sample number with elemental grade greater than c , C is a constant, and D is the fractal dimension. This methodology is a pretreatment technique without any estimation.

3. Geological setting and sampling

The Milloieh copper deposit is situated 80 km NE of Sirjan (SE Iran) and the Cenozoic Urumia-Dokhtar magmatic belt (Figure 1). The Urumia-Dokhtar magmatic belt is one of the main regions of copper mineralization in the world, and is a part of the Himalaya-Alp orogenic belt. The most important feature of this magmatic belt is the presence of calc-alkaline intrusive masses that have penetrated from Oligocene to Miocene of volcanic units. Many of these intrusive masses with porphyry texture have copper-molybdenum

mineralization. This magmatic belt extends from NW to SE Iran. The Iranian large porphyry copper deposits within this belt such as Sarcheshmeh, Sungun, Meiduk, and Darehzar are shown in Figure 1 (Shahabpour, 1994; Richards, 2015). According to the exploratory studies, three major mineralization zones for this area have been considered: Milloieh (1), Milloieh (2), and Milloieh (3), which are located in the western part of the area. The rock facies of Milloieh (1) and Milloieh (3) include dark porphyry andesite, while the rock facies in Milloieh (2) are composed of light-colored tuff crystals. There are volcanic, volcano-plutonic, and pyroclastic rocks. The pyroclastic rocks are crystalline tuffs, and green and grays tuffs. There are intrusive units with dioritic and gnenodioritic compositions. In addition, the volcanic rocks include andesitic and andesitic tuffs. The ore elements are chalcopyrites, covelites, chalcocites, and pyrites with minor ores including malachites, azorites, and magnetites. The alteration zones are located as a ring such as classic model for porphyry deposits that are propylitic, argillic, and phyllic from the marginal to the central parts of the studied deposit. Few index minerals for potassic alteration were detected in depth cores. Four boreholes were drilled in the crystal and green tuffs, diorites, and andesitic tuffs. These four Cu mineralized datasets were obtained by continuous sampling every 2 m along and within these borehole in the mineralized zones, respectively. The samples obtained were analyzed by the ICP-OES method in the Zarazma Co., Tehran, Iran. The rock ore properties and the statistical results of analysis of the boreholes datasets are summarized in Table 1.

Table 1. Basic information and statistical properties of the four mineralized borehole data obtained from the Milloieh copper deposit.

Borehole name	Depth (m)	Number of samples	Cu Max (ppm)	Cu Min (ppm)	Cu Mean (ppm)	Std Dev.	CV
BH02	48	19	14500	440	4100	0.49	1.20
BH04	51	10	8500	460	2800	0.11	1.25
BH05	27	4	30300	900	10800	1.38	1.28
BH07	36	11	9000	120	1600	0.32	1.98

4. Results and discussion

4.1. Frequency distribution of Cu

The property of frequency distributions of element concentrations are usually evaluated by the histograms that are displayed as arbitrarily chosen, linearly scaled concentration intervals on the frequency of individual analyses whose results fall

in a particular class within the interval on the ordinates. The histograms are depicted in Figure 2 and show that distribution of the Cu values in the mineralized boreholes is positively skewed. On the other hand, there are L shapes in the histograms, which indicate that low grade data is more than high grade samples (Figure 2).

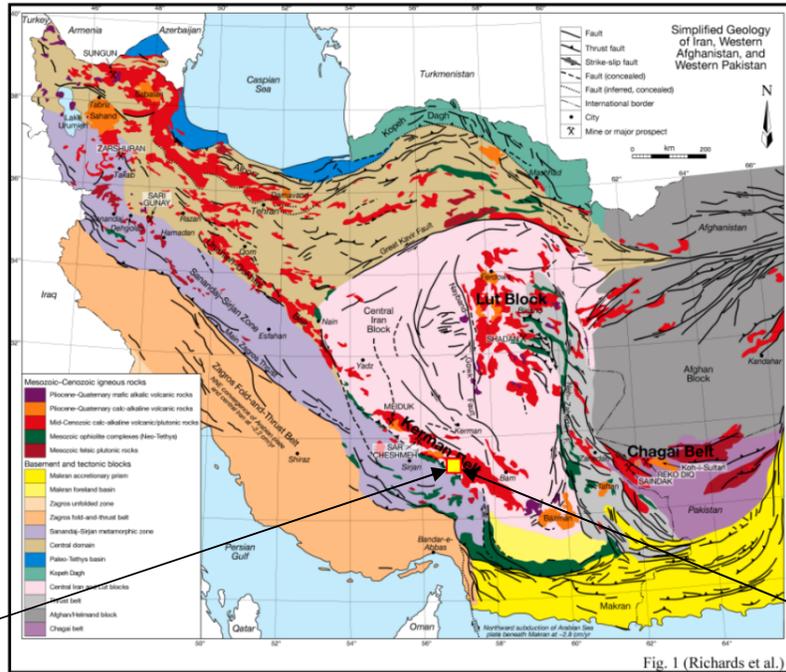


Fig. 1 (Richards et al.)

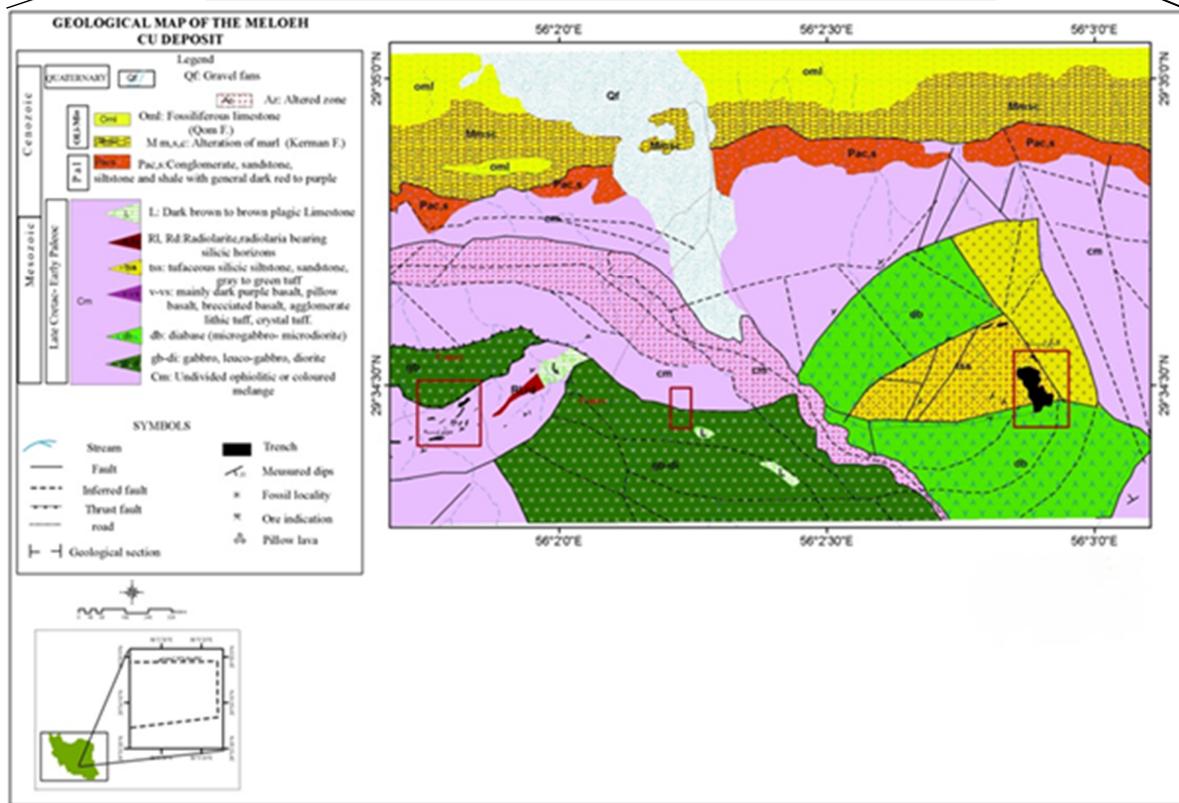


Figure 1. Location of the Molloieh copper deposit (Yellow Square) on the map of the Iran structural zones (Richards et al. 2006) and geological map of the Molloieh area in scale of 1:10000.

4.2. Power-law frequency distribution of Cu values

The results derived via the power-law frequency model for the mineralized boreholes are depicted in Figure 3 and are listed in Table 2. The log-log plots

of cumulative number versus Cu values reveal that the Cu concentration variation distribution in the mineralized boreholes satisfy a bifractal model. The two sections of the plot identify a cross-over point between the 0.56% and 1.09%, values, as

shown in Table 2. for Cu values less than and greater than these values, fractal dimension 1, and

range from 0.36 to 3.66, in the mineralized rocks, i.e. fractal dimension 2 in this deposit.

Table 2. Fractal dimensions of Cu in the Molloieh copper porphyry deposit calculated by the power-law frequency model.

Borehole name	Fractal dimension 1	R ²	Break point	Fractal dimension 2	R ²
BH02	0.18	0.92	0.63	2.51	0.92
BH04	0.42	0.95	0.56	3.66	0.97
BH05	0.89	0.98	1.09	0.36	0.99
BH07	0.34	0.97	0.63	0.89	0.99

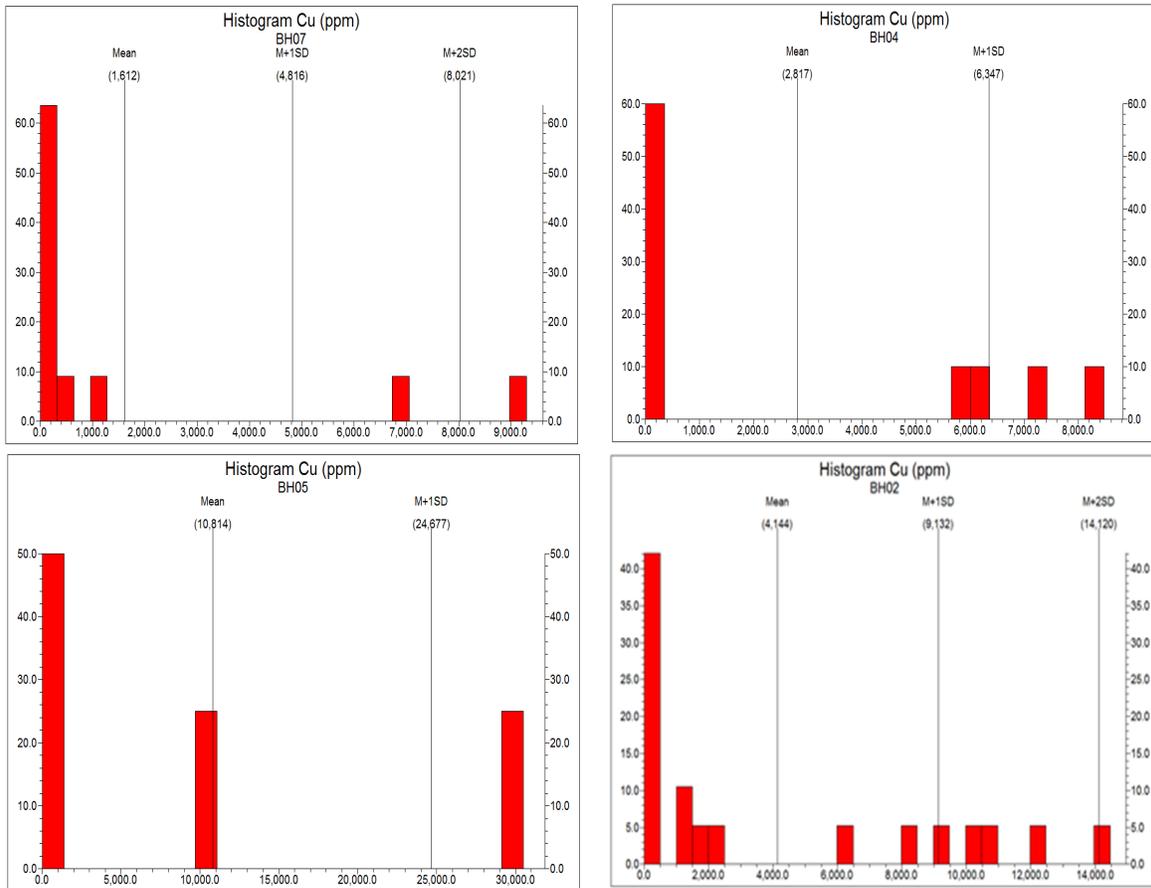


Figure 2. Histograms of Cu values for four mineralized boreholes.

A bigger fractal dimension of mineralization indicate a more homogeneous mineralization. The fractal dimensions are calculated from the frequency of the Cu values, and can reflect the Cu concentration value proportion. A larger fractal dimension also reveals lower Cu values greater than certain specific Cu grades. It means that the Cu concentrations change slowly along the depth of borehole, indicating a more homogeneous mineralization. The Coefficient of Variation (CV), which is the standard deviation divided by the

average, identifies the degree or range of change. In the case of an approximate average of Cu value, a low CV implies the elemental grades changing slightly along the borehole depth. In other words, it means a more homogeneous mineralization. Therefore, the fractal dimensions of mineralization are inversely related to the CV values. The regression line between the fractal dimensions and the CV values is depicted in Figure 4. The squared correlation coefficient "R²" is 0.29, which reveals a weak correlation between the two variables.

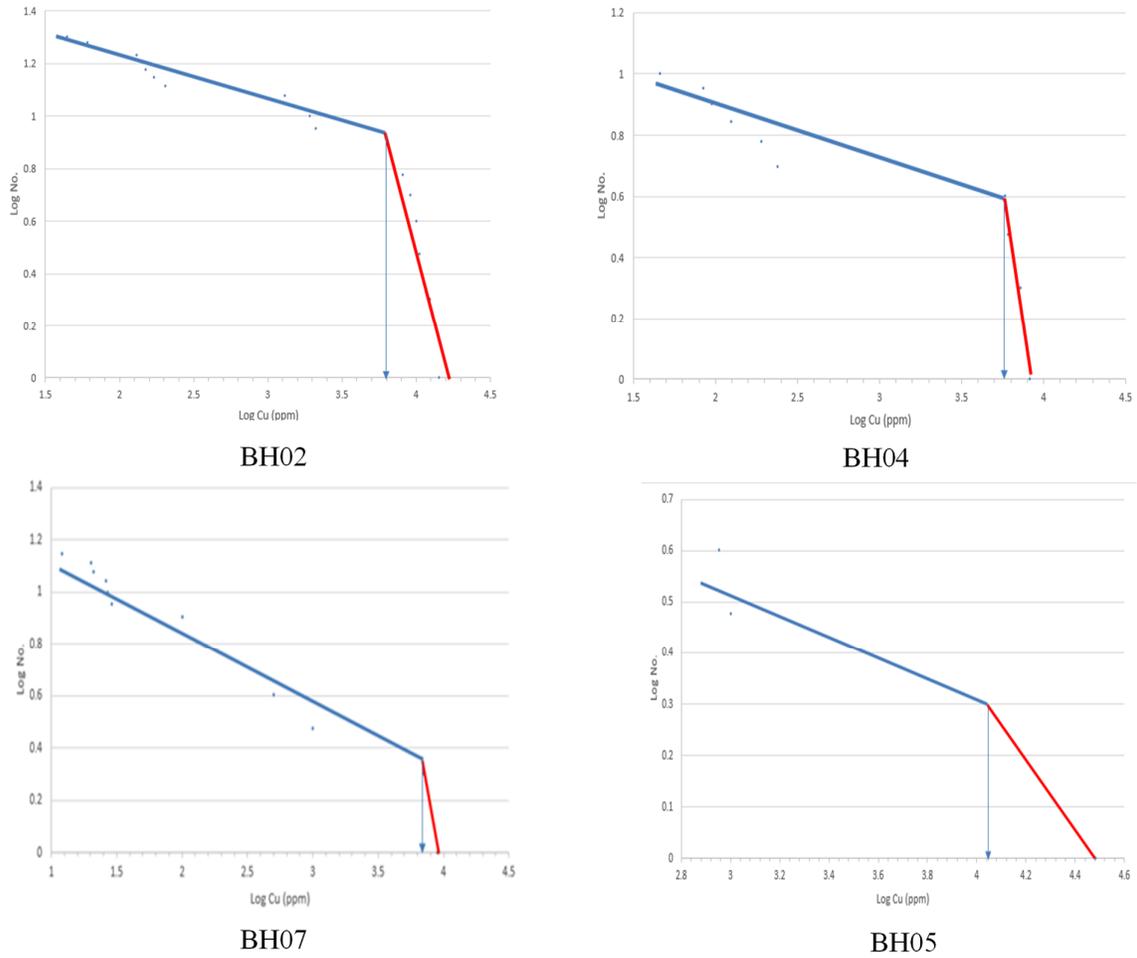


Figure 3. Log–log plots of cumulative numbers versus Cu grade in four mineralized boreholes.

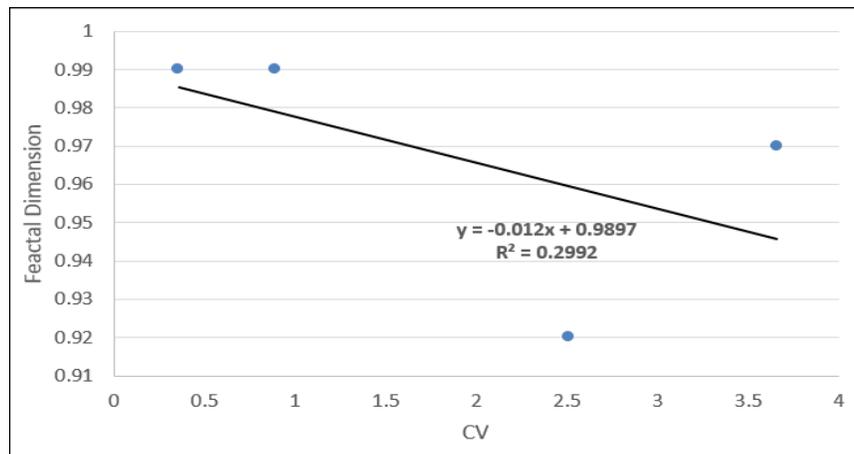


Figure 4. Plot of fractal dimensions for Cu mineralization versus CV along with the illustrated regression line.

5. Conclusions

The continuity of ores is a key role in the interpretation of potential mineralization assessment in a target deposit. The mineralization potential in depth can be identified by characterizing the vertical distribution of elemental

concentrations of the mineralized zones in borehole datasets. In this work, the fractal dimensions obtained by a power-law frequency model were utilized to delineate and interpret the irregularities and the continuity of Cu mineralization in the Milloieh deposit. The results

obtained exhibit that the vertical distribution of Cu grades in mineralized boreholes has different characteristics: the former has a positive skewed distribution character and exhibits a bifractal model, while the latter satisfies a multimodal distribution. It can be shown that there is a low grade mineralization in depth based on statistical parameters, frequency of ore grades, and fractal modeling. This methodology can be improved for detection of potential of mineralization in depth for different ore deposits.

References

- [1]. Afzal, P., Adib, A. and Ebadati, N. 2018. Delineation of seismic zonation using fractal modeling in West Yazd province, Central Iran. *Journal of Seismology* 22 (6): 1377-1393.
- [2]. Afzal, A., Fadakar, Y., Khakzad, A., Moarefvand, P. and Rashidnejad, N. 2011. Geochemical anomaly separation by multifractal modeling in Kahang (Gor Gor) porphyry system, Central Iran. *J. Geochem. Explor.* 108:220-232.
- [3]. Afzal, P., Fadakar Alghalandis, Y., Moarefvand, P., Rashidnejad Omran, N. and Asadi Haroni, H. 2012. Application of power-spectrum-volume fractal method for detecting hypogene, supergene enrichment, leached and barren zones in Kahang Cu porphyry deposit, Central Iran, *Journal of Geochemical Exploration* 112, 131–138.
- [4]. Afzal, P., Yousefi, M., Mirzaei, M., Ghadiri-Sufi, E., Ghasemzadeh, S. and Daneshvar Saein, L. 2019. Delineation of podiform-type chromite mineralization using Geochemical Mineralization Prospectivity Index (GMPI) and staged factor analysis in Balvard area (southern Iran). *Journal of Mining and Environment* 10: 705-715.
- [5]. Agterberg, F.P. 1993. Calculation of the variance of mean values for blocks in regional resource evaluation studies. *Nonrenewable Resources* 2: 312–324.
- [6]. Ahrens, L.H. 1966. Element distributions in specific igneous rocks-VIII. *Geochim. Cosmochim. Acta* 30: 109–122.
- [7]. Carranza, E.J.M. 2009. Geochemical anomaly and mineral prospectivity mapping in GIS. *Handbook of Exploration and Environmental Geochemistry*, 11. Elsevier, 351 pp.
- [8]. Cheng, Q. 1995. The perimeter-area fractal model and its application to geology. *Math. Geol.* 27: 69–82.
- [9]. Cheng, Q., Agterberg, F.P. and Ballantyne, S.B. 1994. The separation of geochemical anomalies from background by fractal methods. *J. Geochem. Explor.* 51: 109–130.
- [10]. Daneshvar Saein, L., 2017. Delineation of enriched zones of Mo, Cu and Re by concentration-volume fractal model in Nowchun Mo-Cu porphyry deposit, SE Iran. *Iranian Journal of Earth Sciences* 9: 64-72.
- [11]. Daneshvar Saein, L. and Afzal, P. 2017. Correlation between Mo mineralization and faults using geostatistical and fractal modeling in porphyry deposits of Kerman Magmatic Belt, SE Iran. *Journal of Geochemical Exploration* 181: 33-343.
- [12]. Karami, K. and Afzal, P. 2015. Application of multifractal modeling for separation of sulfidic mineralized zones based on induced polarization and resistivity data in the Ghare-Tappeh Cu deposit, NW Iran. *Iranian Journal of Earth Sciences* 77: 134-141.
- [13]. Khalili, H. and Afzal, P. 2018. Application of spectrum-volume fractal modeling for detection of mineralized zones. *Journal of Mining and Environment* 9: 371-378.
- [14]. Mandelbrot, B.B., 1983. *The fractal geometry of nature*. New York: W.H. Freeman and Company. 468.
- [15]. Mirzaei, M., Afzal, P., Adib, A., Rahimi, E. and Mohammadi, Gh. 2020. Detection of zones based on ore and gangue using fractal and multivariate analysis in Chah Gaz iron ore deposit, Central Iran. *Journal of Mining and Environment (In press)*.
- [16]. Monecke, T., Gemmell, J.B. and Monecke, J. 2001. Fractal distributions of veins in drill core from the Hellyer VHMS deposit, Australia: constraints on the origin and evolution of the mineralising system. *Miner. Depos.* 36: 406–415.
- [17]. Rashidnejad Omran, N., Afzal, P., Harati, H., Moarefvand, P., Asadi Haroni, H. and Daneshvar Saein, L. 2011. Application of power-law frequency fractal model in determination of vertical geochemical distribution of Cu in Kahang porphyry deposit, Central Iran. *Journal of Mining and Metallurgy* 47 A: 1–8.
- [18]. Richards, J.P. 2006. *Exploration and Mining Geology now under new management*. *CIM MAGAZINE* 1 (7): 62.
- [19]. Richards, J.P. 2015. Tectonic, magmatic, and metallogenic evolution of the Tethyan orogen: From subduction to collision. *Ore Geology Reviews* 70: 323-345.
- [20]. Sanderson, D.J., Roberts, S. and Gumiel, P. 1994. A Fractal relationship between vein thickness and gold grade in drill core from La Codosera, Spain. *Econ. Geol.* 89: 168–173.
- [21]. Shahabpour, J. 1994. Post-mineral breccia dyke from the SarCheshmeh porphyry copper deposit, Kerman, Iran. *Exploration and Mining Geology*, 3: 39-43.
- [22]. Soltani, F., Moarefvand, P., Alinia, F. and Afzal, P. 2019. Characterizing Rare Earth Elements by coupling multivariate analysis, factor analysis and geostatistical simulation; case-study of Gazestan

deposit, central Iran. *Journal of Mining and Environment* 10: 929-945.

[23]. Turcotte, D.L. 1996. *Fractals and Chaos in Geophysics*, 2nd ed. Cambridge University Press, Cambridge UK, 81–99.

[24]. Wang, Z., Cheng, Q. and Cao, L. 2006. Fractal modeling of the microstructure property of quartz mylonite during deformation process. *Math. Geol.* 39: 53–68.

[25]. Yasrebi, A.B., Hezarkhani, A. and Afzal, P. 2017. Application of Present Value-Volume (PV-V) and NPV-Cumulative Total Ore (NPV-CTO) fractal modelling for mining strategy selection. *Resources Policy* 53: 384-393.

[26]. Yasrebi, A.B. and Hezarkhani, A. 2019. Resources classification using fractal modelling in Eastern Kahang Cu-Mo porphyry deposit, Central Iran. *Iranian Journal of Earth Sciences* 11: 56-67.

[27]. Yasrebi, A.B., Wetherelt, A., Foster, P., Coggan, J., Afzal, P., Agterberg, F.P. and Kaveh Ahangaran, D. 2014. Application of a density-volume (D-V) fractal model for rock characterisation and correlation of RQD and lithological units with density model in the Kahang porphyry deposit, Central Iran. *International Journal of Rock Mechanics and Mining Sciences* 66: 188–193.

[28]. Zuo, R. 2011. Decomposing of mixed pattern of arsenic using fractal model in Gangdese belt, Tibet, China. *Appl. Geochem* 26:S271-S273.

[29]. Zuo, R., Cheng, Q. and Xia, Q. 2009. Application of fractal models to characterization of vertical distribution of geochemical element concentration, *Journal of Geochemical Exploration*, 102 (1): 37-43.

[30]. Zuo, R. and Wang, J. 2016. Fractal/multifractal modeling of geochemical data: A review. *Journal of Geochemical Exploration*, 164: 33-41.

کاربرد روش فرکتالی فراوانی توان جهت تعیین توزیع عمودی عیار مس در کانسار مس پورفیری میلویه، جنوب شرق ایران

نویدرضا احمدی، محمدرضا جعفری*، محمدرضا اسپهبد، نادر کهنسال قدیم‌وند

گروه زمین‌شناسی، واحد تهران شمال، دانشگاه آزاد اسلامی، تهران، ایران

ارسال 2020/2/10، پذیرش 2020/3/19

* نویسنده مسئول مکاتبات: mr.jafari_1348@yahoo.com

چکیده:

شناخت نحوه توزیع عمودی و افقی عیار عناصر نقش مهمی در فازهای گوناگون اکتشاف مواد معدنی دارد. هدف اصلی این مقاله، بررسی و تعیین مشخصات توزیع عمودی عیار مس با استفاده از مدل فرکتالی فراوانی توان در کانسار مس پورفیری میلویه واقع در استان کرمان و جنوب شرقی ایران است. این پژوهش براساس داده های چهار گمانه اکتشافی در زون کانه‌دار است. توزیع عمودی عیار مس در گمانه‌های حفرشده در زون کانه‌دار نشانگر چولگی مثبت با توزیع فراوانی چندمدی می‌باشد. انجام این مدل فرکتالی بر روی این چهار گمانه نشانگر توزیع تک فرکتالی و دو جامعه‌ای در هر یک از این گمانه‌ها است. در این گمانه‌ها نقطه شکست نمودارهای فرکتالی در عیارهای بین 0/5 تا یک درصد و نیز تغییرات بعد فرکتال بین 1/7 تا 4/97 است.

کلمات کلیدی: روش فرکتالی فراوانی توان، توزیع عمودی، میلویه.
