A new model for predicting the advance rate of a Tunnel Boring Machine (TBM) in hard rock conditions

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Abstract
The prediction of the advance rate of a Tunnel Boring Machine (TBM) in hard rock conditions is one of the most important concerns for estimating the time and costs of a tunnel project. In this paper, in the first step, a model based on Rock Engineering Systems (RES) is proposed to predict geotechnical risks (representing media characteristics) in rock TBM tunnelling. Fifteen main parameters that influence the geotechnical hazards were used in the modelling. In establishing an interaction matrix and also a parameter rating, the views of five experts were taken into account. The Vulnerability Index (VI) (geotechnical risk levels) for 2058 datasets out of 2168 sets of data from 53 geological zones in 11 km of the Zagros long tunnel was obtained. In the second step, based on the machine operating parameters such as torque, cutter head rotation per minute, cutter normal force and media characteristics (represented by VIs), which were used as input parameters and advance rate was used as an output parameter, while using 2058 datasets, linear and non-linear multiple regression analyses were carried out. 110 datasets (out of 2168 datasets), which were not used in the modelling, were applied to evaluate the performance of regression models and other models in literature and the results were compared. The obtained results showed that the new linear model proposed with $R^2=0.83$ and RMSE=0.12 has a better performance than the other models.

Keywords:
Advance rate, Regression models, TBM geotechnical risk, Rock engineering systems, Hard rock TBM, Zagros long tunnel

1. Introduction
In tunnel construction by TBM, the estimation of time and cost of a project are one of the most important parameters, which has always been a challenge between contractors and owners. Penetration and advance rates are key parameters in the performance estimation of a TBM, which results in estimating the time and cost of a project. In recent years, many researchers devoted their work to predict the performance of a TBM. There are two categories of research works while considering modelling of predicting performances of hard rock TBMs. One category is theoretical models and the other empirical models (Rostami, 2015). In principle, theoretical models are developed, by use of tests of indentation or full-scale cutting tests in a laboratory, which provide an estimation of cutting forces based on the type of cutter, geometry of cutting, spacing, and the cut penetration (Crow, 1975; Roxborough and Phillips, 1975; Ozdemir, 1978; Snowdon et al., 1982; Sanio, 1985; Sato et al., 1991; Rostami and Ozdemir, 1993; Rostami, 1997; Balci et al. 2009; Bilgin et al. 2005, 2007, 2012; Tumac et al., 2012). The main weakness of these models is that the tests used do not entirely describe the facts regarding rock mass characteristics which the disc cutters of TBM face in the field. This type of equipment for tests may not be easily accessible in all centers for research around the world (Khademi Hamidi et al. 2010). The first models were developed, using intact rock parameters such as tensile strength and Uniaxial Compressive Strength (UCS) (Graham, 1977; Farmer and Glossop, 1980). Cassinelli et al. the Penetration Rate (PR) of TBM by the use of Rock Structure Rating (RSR) in 1982. A good relationship between RSR, UCS, and the penetration rate of TBM was introduced by Innaurato et al. in 1991. Snowdon et al. and Sanio disclosed relationships between the UCS of rock and the specific energy in 1983 and 1985 respectively (Farrokh, 2018). Different laboratory tests such as Taber abrasion, Shore hardness, Schmidt hammer, point load index, and drilling rate index tests were applied for the prediction of PR (Howarth et al. 1986; McFeat-Smith, 1977; Ozdemir, 1977; Pang et al., 1989; Yagiz, 2002; Yagiz and Ozdemir, 2001; Fattahi and Moradi, 2017). Other prediction models have been improved in the last three decades to consider a wide range of rock properties and rock mass conditions. Palmstrom (1995) introduced a forecast model based on the Rock Mass index (RMI) and the $Q_{TBM}$ model was developed by Barton (1999, 2000).
based on the developed Q system. The NTNU model was introduced by the Norwegian University of Science and Technology (Bruland, 2012; Blindheim, 1979). Rock Mass Rating (RMR) parameters were employed for predicting TBM performance by Sapigni et al. (2002); Bieniawski et al. (2007); Khademi Hamidi et al. (2010). In 2008, Yagiz developed a relationship between rock mass properties and PR. Hassanpour et al. (2009, 2010, 2011, 2016) investigated TBM performance based on the Field Penetration Index (FPI). Gong and Zhao (2007, 2009) and Gong et al. (2007) introduced the index of boreability and rock brittleness to investigate the rock mass characteristics that affect the performance of TBM. Bieniawski et al. (2007, 2008), examined TBM boreability using rock mass classification systems (Namli and Bilgin, 2017). The difficulties of TBM excavation in the blocky area were examined by Delisio and Zhao (2013). Neuro-fuzzy techniques were used by Alvarez Grima et al. (2000) for modelling of TBM performance. A study was carried out by Acaroglu et al. (2008) for predicting the specific energy requirement for the performance of TBM by applying fuzzy logic. Zhao et al. (2007) used a neural network predictor for TBM performance prediction. Yagiz (2009) applied artificial neural networks and non-linear multiple regression to estimate the TBM performance. The hard rock TBM penetration rate was predicted by Yagiz and Karahan (2011), using the Particle Swarm Optimization (PSO) method. In an attempt, Farrokh et al. (2012) investigated different models used to predict the penetration rate of hard rock TBMs.

In the ABROCK Research Project, different research works are being carried out for the analysis and prediction of cutter wear and penetration rate of TBM tunneling in hard rock conditions (Moradi and Ebrahimi, 2014). Ghasemi et al. (2014) and Fattahi and Babanouri (2017) employed soft computing methods for TBM performance prediction.

### Table 1: The TBM specifications, section two, the Zagros long tunnel Salimi et al. (2016)

| Parameter                      | Value | Unit |
|-------------------------------|-------|------|
| Diameter of machine           | 6.73  | m    |
| Diameter of cutters           | 4318  | mm   |
| Number of disc cutters        | 42    |      |
| Disc cutters nominal spacing  | 90    | mm   |
| Maximum of cutterhead thrust  | 28.134| kN   |
| Power of cutterhead           | 2100  | kW   |
| Speed of cutterhead           | 0-11  | rpm  |
| Torque (nominal) of cutterhead| 4450  | kN.m |
| Stroke of thrust cylinder     | 1700  | mm   |
| Capacity of conveyer          | 650   | t/h  |
| Total weight of TBM           | 573   | ton  |

### Table 2: A descriptive statistics of datasets for section two of the Zagros long tunnel

| No  | Parameter                                      | Unit  | Symbol | Min  | Max  | Mean  | St.d  |
|-----|------------------------------------------------|-------|--------|------|------|-------|-------|
| 1   | Cutter head rotation per minute                | RPM   | RPM    | 2.65 | 8    | 5.42  | 0.91  |
| 2   | Cutter normal force                            | kN    | Fs     | 52.73| 231.04| 156.62| 34.84 |
| 3   | Torque                                        | kN.m  | Ts     | 292  | 8790 | 814.45| 330.91|
| 4   | Advance rate                                   | m/h   | AR     | 1.68 | 3.96 | 2.87  | 0.45  |
| 5   | Uniaxial compressive strength                  | MPa   | UCS    | 15   | 125  | 35.16 | 21.69 |
| 6   | Elastic modulus                                | GPa   | E      | 5    | 22.5 | 5.73  | 2.66  |
| 7   | Quartz content                                 | %     | QC     | 0.5  | 4    | 1.76  | 1.01  |
| 8   | Dry density                                    | g/cm² | Dd     | 2.25 | 2.55 | 2.41  | 0.06  |
| 9   | Porosity                                       | %     | P      | 3.5  | 12.5 | 8.02  | 2.28  |
| 10  | Tensile strength (Brazilian test)              | MPa   | TS     | 1    | 6    | 2.50  | 0.95  |
| 11  | Over burden                                   | m     | Ob     | 40   | 1000 | 211.76| 96.09 |
| 12  | Joint spacing                                  | mm    | JS     | 15   | 300  | 136.27| 84.50 |
| 13  | Rock quality designation                       | %     | RQD    | 22.5 | 85   | 65.98 | 12.64 |
| 14  | Rock mass rating (1989)                        | RMR<sub>89</sub> | 34  | 65 | 47.54 | 6.70 |
| 15  | Gas emission (H,S)                             | ppm   | Ge     | 0    | 30   | 7.05  | 4.47  |
|     |                                                | CH4   |        | 0    | 7    | 0.5   | 0.7   |
| 16  | Water table                                   | m     | Wt     | 50   | 400  | 262.68| 138.66|
| 17  | Water inflow                                  | lit/min in 10 m | Wi | 0 | 300 | 12.51 | 27.64 |
| 18  | Clay contents in mass rock                    | %     | Cc     | 5    | 40   | 2.80  | 5.48  |
| 19  | Width of crushed zone                         | m     | CZ     | 0    | 200  | 2.18  | 15.11 |

Maximum: Max; Minimum: Min; Standard deviation: St.d
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Figure 1: Geological profile of section two of the Zagros long tunnel alignment (Lar consulting engineers, 2003, 2004).
The objective of the present piece of work is to introduce a new model to improve the accuracy of advance rate prediction in different geotechnical conditions in rock TBM tunnelling, considering both operational parameters of TBM (machine parameters) as well as media characteristics. Media characteristics are represented by the geotechnical risk levels in the modelling. The geotechnical risk levels are obtained, using an RES based model. In the new model, machine parameters and geotechnical risk levels are input parameters and the advance rate is an output parameter. The field data used for the modelling was collected from section two of the Zagros long tunnel in Iran.

2. Description of the site and data collection

The Zagros long tunnel with a circular cross-section, an excavation diameter of 6.73 m, a final diameter of 6 m, and a length of 52 km is a water transfer tunnel which consists of two sections. This tunnel is located in the northwest of Iran. It was planned to transfer about 70 m$^3$/s of water through this tunnel. The data was collected from the second section of the tunnel, which has a length of 26 km. This section of the tunnel was excavated by a double-shield TBM. The machine specifics are shown in Table 1.

The tunnel passes through three formations. The Pabdeh (PEPd) formation includes a mix of greenish-grey limestone argillaceous and dark grey limy shale. The Gurpi (KGu) formation combines argillaceous limestone and limy shale. Furthermore, the Ilam (Ki) formation consists of brownish-grey limestone as illustrated in Figure 1 (Lar consulting engineers, 2003, 2004).

Rock mass conditions vary from weak to good (RMR: 17 to75). According to geological studies, 53 zones were recognized, which mainly are mainly consist of layers of shale, limestone, and marl. The dominant lithological characteristics were used to name the zones (Marl (MA), Shale (SH), Limestone (LI), Limy Shale (LI-SH), Marly-Limestone (LI-MA), Fractured Zone (FZ), and Crushed Zone (CZ)). Twenty-three different types of faults intersect with the tunnel. Overburdens vary between 40 m and 1000 m. The water table above the tunnel varies from 50 m to 400 m (Lar consulting engineers, 2003, 2004).

The database for this research work (2168 sets of data) was obtained by combining the data in the pre-construction and construction phases. During the construction phase, the geological and geomechanical data, which were obtained before construction, were examined (Khademi Hamidi et al. 2010; Salimi et al. 2016). During construction, the operational data of TBM such as Revolutions per Minute (RPM), torque, thrust, and advance rate were recorded. The basic descriptive statistics of the sets of data are revealed in Table 2.

3. Rock engineering system

Hudson (1992) introduced the Rock Engineering Systems (RES) approach in which interactions between rock mechanics parameters in a rock system can be obtained. This approach was applied to different rock engineering fields such as an investigation of underground excavation stability (Lu and Hudson, 1993), with rockfall risk assessment (Cancelli and Crosta, 1993), and characteristics of a rock mass to indicate natural slope stability (Mazzoccola and Hudson, 1996). (Latham and Lu, 1999) introduced an assessment system to determine the rock mass’s blastability, (Benardos and Kaliampakos, 2004) used an assessment system for geotechnical hazards in TBM tunnelling (Shin et al., 2009) studied a hazard assessment of tunnel collapses, a quantitative approach, (Faramarzi et al., 2013) predicted...
rock fragmentation by blasting and predicted the advance rate of face, while determination of the operation efficiency in retreat longwall mining panel was studied by (Aghababaei et al., 2019). In the approach of RES, establishing the matrix of interaction plays an important role. By generating the interaction matrix, the weighting of the parameters of the rock mass system can be evaluated. In this matrix, the principal parameters, which affect the system, are placed on the leading diagonal of the matrix and the interactions (effects of each parameter on any other parameter) are located on the off-diagonal cells. The assignment of values to off-diagonal cells is called coding of the interaction matrix. The simplest of the interaction matrix, a two parameter system, is shown in Figure 2(a). Beside that, a view of interaction matrix coding is shown in Figure 2(b). The row passing through Pi denotes the impact of Pi on all the other parameters in the system, while the column through Pi shows the impacts of other parameters, or the remaining of the system, on the Pi. In principle, there is no limit to the number of parameters that may be included in an interaction matrix.

For numerically coding the interaction matrix, different procedures such as the 0-1 binary, Expert Semi-Quantitative (ESQ) (Hudson, 1992), and the Continuous Quantitative Coding (CQC) (Lu and Latham, 1994) were applied. Among the suggested coding procedures, the most commonly used is the ESQ coding. Based on this coding procedure, the interaction intensity is presented by the values from 0 for no interaction to 4 for critical interaction as revealed in Table 3.

Table 3: The values for coding of interaction matrix in the ESQ procedure (Hudson, 1992)

| Coding | Description               |
|--------|---------------------------|
| 0      | No interaction            |
| 1      | Weak interaction          |
| 2      | Medium interaction        |
| 3      | Strong interaction        |
| 4      | Critical interaction      |

In the matrix of interaction, the summation of a row is called the “cause” value ($C_i = \sum_j I_{ij}$) and the summation of a column is the “effect” value ($E_i = \sum_j I_{ij}$), which are the coordinates (C, E) for a parameter in the system. In the cause and effect space, the coordinate values for each parameter in the system can be plotted, which is called the C-E plot. The interactive intensity value of each parameter in the system is indicated as the summation of the C and E values ($C + E$) and it can be applied as an indication of the significance of a parameter in the system. Also, a parameter’s weighting factor ($a_i$) is defined using the percentage value of ($C + E$) as shown in Equation 1.

$$a_i = \frac{(C_i + E_i)}{(\sum C_i + \sum E_i)} \times 100$$  \hspace{1cm} (1)

Where:

- $C_i$ – Cause of $i^{th}$ parameter,
- $E_i$ – Effect of $i^{th}$ parameter,
- $a_i$ – Weighting of the $i^{th}$ parameter.

4. Prediction of the level of geotechnical risks, using an RES based model

The level of geotechnical risks was used to reflect the effects of media characteristics in the modelling of TBM performance prediction. In this research, the principles of the work carried out by Benardos and Kaliampakos, (2004) were adopted to define a model for predicting the level of geotechnical risks in rock TBM tunnelling. The merit of this modelling is that all of the important and easily obtainable parameters affecting geotechnical risks in rock TBM were taken into account.

There are two main phases for defining the RES based model. In the first phase, the geotechnical parameters (hazards), which cause the occurrence of risk in the case of rock TBM tunnelling, are defined. Also, their behaviors are analyzed and the impact (weight) that each one has in the whole risk conditions is evaluated. Further, the RES principles can be applied to evaluate the weighting of the parameters that influence the risk system.

In phase two, the Index of Vulnerability (VI), representing the level of risk, can be determined, using Equation 2 (Hudson, 1992):

$$VI = \left(\frac{100 - \sum_{i=1}^{n} a_i \frac{P_i}{P_{\text{max}}}}{P_D}\right)$$  \hspace{1cm} (2)

Where:

- VI – Index of Vulnerability,
- $a_i$ – Weighting of the $i^{th}$ parameter,
- $P_i$ – Value (rating) of the $i^{th}$ parameter,
- $P_{\text{max}}$ – Maximum value assigned the $i^{th}$ parameter (normalization factor).

In this equation, the level of total risk can be calculated based on the weight and individual influence of each parameter on the total risk level (the parameter’s value rating). The parameter value ratings can be carried out based on the effects of these parameters on the vulnerability conditions.

Based on the VI obtained, (using Equation 2) and the classification of VI, the level of total risk in each zone of the tunnel can be identified. In the classification for VI, it is divided into three main classes with a normalized scale of 0 to 100, (see Table 4). Class I is including, small-scale problems. These problems cannot significantly affect tunnel driving. In class II, a challenging region might occur, which must be taken into consideration.
In class III, it is expected that in certain individual regions, several difficulties during tunnel excavation, are to be observed.

4.1. Parameters influencing geotechnical risks of TBM

In reviewing the TBM geotechnical risk assessment in hard ground conditions and the pieces of literature published (Anagnostou and Kovári, 1996; Barla and Pelizza, 2000; Barla, 2001; Dalg, 2003; Shang et al., 2004; Farrokh, 2006; Ramoni and Anagnostou, 2010; Jung et al., 2011; Shaterpour Mamaghani et al., 2015 Arbabsiar et al., 2020) many parameters can influence the TBM geotechnical risks encountered. The parameters in Table 5 are the most important ones from available parameters in the database and can be obtained easily. These parameters cover the most important hazards which might be encountered during rock TBM tunnelling such as instability, fault zone, squeezing, sticking due to muddy condition, and water ingress. In the selected parameters, the components of RMR such as UCS, RQD, joint spacing, joint orientation, joint condition, and water inflow are preferred instead of RMR as they describe other issues of risk raised as well as instability as a hazard.

| Parameter | Symbol | Parameter | Symbol |
|-----------|--------|-----------|--------|
| RQD | P1 | Gas emission | P9 |
| UCS | P2 | Squeezing | P10 |
| Joint condition | P3 | Mixed face | P11 |
| Joint orientation | P4 | Karstic condition | P12 |
| Joint spacing | P5 | Abrasive rocks (quartz contents) | P13 |
| Water table | P6 | Width of crushed zone | P14 |
| Permeability | P7 | Clay contents in mass rock | P15 |
| Water inflow | P8 | - | - |

RQD describes the fracturing degree of a rock mass. It is one of the most important components of RMR, which has a significant role in the excavation’s stand-up time, indicated by RMR (Goodman, 1989). Also, rocks with low RQD cause difficulties through TBM gripping, putting excavation at risk and causing a reduction in the advance rate.

UCS shows the compressive strength of intact rocks and also has significant effects on RMR and the associated stability (Goodman, 1989). Moreover, rocks with very high UCS cause wear and tear of cutters as well as a considerable reduction in the advance rate. Also, in rocks with low UCS, gripping problems might occur. The condition of the joints significantly influences an excavation’s stability. It also affects the gripping of TBM. The total performance of TBM in joints with very rough surfaces of a limited extent and a hard wall is much better than open joints filled with more than 5 mm of gouge, or open more than 5 mm, or when joints extend more than several meters.

The orientation of the joints comparative to the tunnel axis has an impact on the chipping behavior of rocks during the excavation by TBM. Based on Bieniawski (1984, 1989), the orientation of the joints perpendicular to the tunnel axis is the most favorable for TBM tunnelling.

The spacing of joints affects the rating in RMR and therefore the stability of the tunnel. On the other hand, based on the recommendations by DAUB (German Tunnelling Committee) (DAUB, 1997), in the rocks with joint spacing of 0.2 m-0.6 m, gripping problems might be encountered, for which shield-TBM suggests. In rocks with joint spacing of 0.06 m-0.2 m, the possible falling rocks might trap the shield-TBM causing serious delays.

Groundwater can strongly influence rock mass behavior (Goodman, 1989). Also, the pore pressure in rocks depends on the level of groundwater. Higher groundwater level means more pore pressure, resulting in decreasing stability, and in the case of high permeability and the presence of joints, an increase in water inflow. In return, the high inflow of water causes delays and possible damage.

Gas emission such as Methane (CH₄), Hydrogen Sulfide (H₂S), Carbon Monoxide (CO) and Hydrogen Cyanide (HCN) into a tunnel might cause safety concerns such as human loss and damage due to an explosion, respiratory problems as well as corrosion damage, resulting in delays in tunnelling. In most cases, the main concern is CH₄ and H₂S.

CH₄ is a flammable gas, lighter than air and odorless. The Lower Explosive Limit (LEL) and Upper Explosive Limit (UEL) for CH₄ are 5% and 15% respectively (Industrial Training Branch of the National Coal Board, 1981).

H₂S is a colorless and toxic gas with LEL and UEL of 4% and 44% respectively. It is slightly heavier than air with a strong smell of rotten eggs. H₂S dissolves in water, making a weakly acidic solution. The main dangers are related to its toxic effect and corrosive property to metals. It can be recognized by smelling below 1 ppm. The 8 hour occupational exposure limit is 10 ppm, for short times (15 min), 15 ppm is acceptable. Above 15 ppm, full face masks with filters are used. This restricts the ability to work to a great extent. Above 100 ppm, H₂S deadens a person’s sensitivity to smell it. Also,
above 100 ppm it is immediately dangerous to life or health. Therefore, filters should only be used up to 100 ppm. At higher concentrations, masks with positive pressure self-supplying respirators shall be used.

Concentrations between 50 and 100 ppm have to be negotiated for longer periods. H2S also causes corrosion to metals, in particular to the electrical installations on the TBM. To reduce the effects, some electrical cabinets, the operator cabin, and a rest cabin have been connected to a fresh air supply (Industrial Training Branch of the National Coal Board, 1981).

Squeezing ground may lead to inadmissible deformations of the tunnel, damage to the support or in the case of mechanized excavation, to the immobilization of TBM due to sticking cutter heads or jamming of the shield. Depending on the number and the length of the critical stretches, squeezing conditions may even put the feasibility of the TBM drive into question (Ramoni and Anagnostou, 2010). Factors such as high overburden, the presence of water, poor ground condition (rock quality), high stress and properties of rocks, all affect the occurrence of ground squeezing. Several approaches have been proposed by various authors such as Jethwa et al. (1984), Singh et al. (1992), Aydan et al. (1996), Goel et al. (1995), Hoek and Marinos (2000), based on practical experience and documented case histories, to identify squeezing rock conditions and potential tunnel squeezing problems. In this research, Singh et al. (1992) and Jethwa et al. (1984) approaches were used to determine the squeezing potential in different zones.

The occurrence of karstic zones through tunnel alignment results in decreasing rock strength, which in return reduces the load bearing of rock, while gripping is required. On the other hand, in the case of having gases such as CH4 and H2S, karstic zones facilitate the leakage of such gases into the tunnel, which raises the TBM tunnelling risks (Marinos, 2001). The requirements, which cause karstic zones include carbonate rocks, joints, and flow of water. In the Zagros long tunnel, in limited zones, the needs for the occurrence of karstic zones are met.

Encountering the alternatives of hard and soft rocks in the tunnel face, the so-called mixed face condition might cause serious damage to the cutters, which results in a delay in tunnelling. Also, in the case of facing abrasive rocks, which are identified by their quartz contents, cutter wear might occur, which decreases the penetration rate, causing significant delays to tunnelling and economic loss.

The presence of water and rocks containing clay minerals such as shale and marl might cause swelling, which in return increase the stress on the support system used. Also, in the presence of water and such rocks, a sticky condition might be encountered, causing delays and sometimes trapping of the cutterhead.

The width of crushed zones, relevant water ingress and RQD affect the risk associated with the fault zones. With a wider crushed zone, lower RQD and in the presence of high water inflow, a higher degree of risk is expected, which in return causes jamming of the TBM, delays and economic loss.

### 4.2. Interaction matrix and rating of parameters
#### 4.2.1. Interaction matrix

The 15 principal parameters affecting the geotechnical risk in TBM tunnelling are located along the leading diagonal of the matrix and the effects of each parameter on any other parameter (interactions) are placed on the off-diagonal cells. The assigning values to off-diagonal cells, coding the matrix, were carried out by five experts, using the ESQ coding as proposed by Hudson (1992).

| Table 6: The interaction matrix for geotechnical hazards in TBM tunnelling |
|---|---|---|---|---|---|---|---|---|---|---|---|---|---|---|---|
| P1 | 1 | 0 | 0 | 0 | 1 | 3 | 3 | 3 | 3 | 0 | 2 | 0 | 1 | 2 |
| P2 | 3 | 0 | 0 | 2 | 0 | 1 | 0 | 1 | 4 | 0 | 3 | 0 | 2 | 0 |
| P3 | 2 | 0 | 1 | 0 | 1 | 3 | 4 | 3 | 4 | 2 | 2 | 3 | 1 | 2 |
| P4 | 3 | 3 | 0 | 0 | 3 | 3 | 2 | 2 | 3 | 2 | 1 | 2 | 0 | 0 |
| P5 | 4 | 2 | 1 | 0 | 3 | 4 | 3 | 3 | 1 | 2 | 2 | 2 | 2 | 2 |
| P6 | 2 | 3 | 2 | 0 | 1 | 3 | 3 | 2 | 2 | 0 | 3 | 0 | 0 | 0 |
| P7 | 0 | 1 | 1 | 0 | 0 | 3 | P1 | 4 | 4 | 2 | 0 | 3 | 1 | 1 |
| P8 | 1 | 1 | 1 | 0 | 0 | 1 | 1 | 3 | 1 | 0 | 3 | 1 | 0 | 1 |
| P9 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| P10 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 1 | 1 | 2 | 0 | P10 | 4 | 0 | 0 |
| P11 | 2 | 0 | 0 | 0 | 0 | 0 | 0 | 1 | 1 | 0 | P11 | 1 | 0 | 0 | 0 |
| P12 | 2 | 1 | 1 | 0 | 1 | 3 | 2 | 2 | 2 | 1 | 0 | P12 | 1 | 1 | 1 |
| P13 | 1 | 1 | 0 | 0 | 0 | 1 | 1 | 1 | 0 | 0 | P13 | 2 | 0 |
| P14 | 2 | 0 | 1 | 0 | 0 | 1 | 2 | 2 | 2 | 2 | 0 | 1 | 0 | P14 |
| P15 | 2 | 3 | 2 | 2 | 2 | 1 | 3 | 2 | 2 | 2 | 0 | 1 | 1 | 2 | P15 |
The interaction matrix for geotechnical hazards in TBM tunnelling is presented in Table 6.

Table 7 gives the cause (C), effect (E), interactive intensity (C+E), dominance (C-E), and weight of each parameter (a_1) calculated by Equation 1 for the parameters affecting the hazards in TBM tunnelling. Also, the weight for each parameter is shown in Figure 3. As can be seen in Table 7 and Figure 3, the most interactive parameters are permeability, RQD, joints condition, karstic zones, whereas the least interactive parameter is the mixed face condition.

### Table 7: The principal parameters weighting in the TBM tunnelling geotechnical risk

| No. | Parameter                                      | C   | E   | C+E | C-E | a_i (%) |
|-----|-----------------------------------------------|-----|-----|-----|-----|---------|
| 1   | RQD                                          | 19  | 25  | 44  | -6  | 8.6     |
| 2   | UCS                                          | 17  | 18  | 35  | -1  | 6.9     |
| 3   | Joint condition                              | 32  | 11  | 43  | 21  | 8.4     |
| 4   | Joint orientation                            | 22  | 2   | 24  | 20  | 4.7     |
| 5   | Joint spacing                                | 32  | 7   | 39  | 25  | 7.6     |
| 6   | Water table                                  | 21  | 19  | 40  | 2   | 7.8     |
| 7   | Permeability                                 | 21  | 28  | 49  | -7  | 9.6     |
| 8   | Water inflow                                  | 14  | 27  | 41  | -13 | 8.0     |
| 9   | Gas emission                                 | 0   | 32  | 32  | -32 | 6.3     |
| 10  | Squeezing                                    | 3   | 25  | 28  | -22 | 5.5     |
| 11  | Mixed face                                   | 6   | 4   | 10  | 2   | 2.0     |
| 12  | Karstic condition                            | 18  | 26  | 44  | -8  | 8.6     |
| 13  | Abrasive rocks (quartz contents)             | 10  | 7   | 17  | 3   | 3.3     |
| 14  | Width of crushed zone                       | 15  | 13  | 28  | 2   | 5.5     |
| 15  | Clay contents in mass rock                   | 25  | 11  | 36  | 14  | 7.0     |
| Total|                                              | 255 | 255 | 510 | 0   | 100     |

Figure 3: The principal parameters weighting in the TBM tunnelling geotechnical risk

4.2.2. Rating of parameters

The rating of the parameters’ values was carried out based on their effect on the vulnerability conditions. Totally 4 classes (except UCS, which has five classes), from 0 to 3 were considered, where 0 denotes the worst case (most unfavorable) and 3 the best (most favorable condition). In the case of TBM tunnelling, the rating of each parameter is presented in Table 8. The ranges of parameters in Table 8 were proposed based on empirical results, practical limits and the experiences of different experts (five experts in this research work).
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4.2.3 Geotechnical risk estimation of the Zagros long tunnel

Geotechnical risk estimation of 2058 sets of data from 11 km of section two of the Zagros long tunnel was carried out using the vulnerability index (Equation 2). To have a clear understanding, the parameter values and the corresponding VI obtained for one set of data are shown in Table 10. In this table, Q\textsubscript{i} and Q\textsubscript{Max} for each parameter can be obtained based on its value or description through Table 8.

| Parameter                     | Values or description and ratings |
|-------------------------------|-----------------------------------|
| 1 RQD (%)                     | Value 0-25 25-50 50-75 75-100     |
| Rating                        | 0 1 2 3                                    |
| 2 UCS (MPa)                   | Value <5 5-25 25-50 50-150 >150          |
| Rating                        | 4 3 2 1 0                                 |
| 3 Joint condition             | Description A\textsuperscript{1} A\textsuperscript{2} A\textsuperscript{3} A\textsuperscript{4} |
| Rating                        | 3 2 1 0                                 |
| 4 Joint orientation (According to Table 9) | Description Very favorable and favorable Fair Unfavorable Very unfavorable |
| Rating                        | 3 2 1 0                                 |
| 5 Joint spacing (m)           | Value <0.06 0.06-0.2 0.2-0.6 >0.6       |
| Rating                        | 0 1 2 3                                 |
| 6 Water table (m)             | Value <0 0-50 50-150 >150              |
| Rating                        | 3 2 1 0                                 |
| 7 Permeability                | Value <10\textsuperscript{-4} 10\textsuperscript{-4}-10\textsuperscript{-4} 10\textsuperscript{-4}-10\textsuperscript{-4} >10\textsuperscript{-4} |
| Rating                        | 3 2 1 0                                 |
| 8 Water inflow (lit/min in 10 m length of the tunnel) | Value <10 10-25 25-125 >125 |
| Rating                        | 3 2 1 0                                 |
| 9 Gas emission                | CH\textsubscript{4} (%) Value <5 5-15 15-25 25-50 >50 |
| Rating                        | 3 0 2 1 0                                      |
| H\textsubscript{2}S (ppm)     | Value <10 10-20 20-100 >100            |
| Rating                        | 3 2 1 0                                  |
| 10 Squeezing*                 | Value <0.4 0.4-0.8 0.8-2 >2           |
| Rating                        | 0 1 2 3                                  |
| 11 Mixed face                 | Description Non-mixed Mixed             |
| Rating                        | 3 0                                     |
| 12 Karstic condition          | Description Non-karstic Mildly karstic Moderately karstic Highly karstic |
| Rating                        | 3 2 1 0                                  |
| 13 Abrasive rocks (quartz contents) % | Value <5 5-10 10-20 >20 |
| Rating                        | 3 2 1 0                                  |
| 14 The width of the fractured or crushed zone | Value None <10 10-25 >25 |
| Rating                        | 3 2 1 0                                  |
| 15 Clay contents in mass rock (muddy condition) (%) | Value <10 10-30 30-70 >70 |
| Rating                        | 3 2 1 0                                  |

* Based on Jethwa et al., (1984)

A\textsuperscript{1}: Very coarse surfaces of limited extent, hard wall rock; A\textsuperscript{2}: Slightly coarse surfaces, opening less than 1 mm, hard wall rock; A\textsuperscript{3}: Slightly coarse surfaces, opening less than 1 mm, soft wall rock; A\textsuperscript{4}: Open joints with smooth surfaces, 1 to more than 5mm aperture, filled with gouge filling or 1 to more than 5mm aperture, joints extended more than several meters

4.3 Multiple regression modelling for the prediction of the advance rate of TBM in hard rock condition

In this paper, based on 2058 datasets collected from section two of the Zagros long tunnel (out of 2168 datasets), linear and non-linear regression analyses were carried out for developing a new model to predict the advance rate of TBM in hard rock conditions. Torque (T\textsubscript{q}), cutter head rotation per minute (RPM), cutter normal force (F\textsubscript{n}) as machine parameters, and vulnerability in-
Table 9: The discontinuity strike and dip orientation effect in tunnelling (Goodman, 1989)

| Strike perpendicular to the tunnel axis | Strike parallel to the tunnel axis |
|---------------------------------------|-----------------------------------|
| Drive with dip                         | Drive against dip                 |
| Dip 45°-90°                           | Dip 20°-45°                       |
| Very favorable                         | Favorable                         |
| Fair                                  | Fair                              |
| Strike parallel to the tunnel axis    | Irrespective of strike            |
| Dip 20°-45°                           | Dip 45°-90°                       |
| Fair                                  | Very unfavorable                   |

Table 10: The corresponding VI to parameters’ values for one set of data of section two of the Zagros long tunnel

| Parameter | RQD (%) | UCS (MPa) | Jc   | Jo   | Js (cm) | Wt (m) | Per. (m/s) | Wi (lit/min) | Ge (%) | Sq | Mf | Kc | Qc (%) | CZ (m) | Cc (%) |
|-----------|---------|-----------|------|------|---------|--------|------------|--------------|--------|----|----|----|--------|--------|--------|
| Value     | 72-85   | 10-20     | A1   | Un-  | 300     | 50     | 2*10^-8    | 22           | <5 (%) | 6.1| Mixed | None | 0-5 | None | 5     |
| Value rating (Pi) | 2   | 3   | 3   | 1   | 2   | 2   | 2   | 3   | 3   | 0   | 3   | 3   | 3   | 3   |
| Weighting (ai) (%) | 8.6 | 6.9 | 8.4 | 4.7 | 7.6 | 7.8 | 9.6 | 8   | 6.3 | 5.5 | 2   | 8.6 | 3.3 | 5.5 | 7   |
| VI        | 21      |           |      |      |        |        |            |              |        |    |    |    |        |        |        |

Jc: Joint condition; Jo: Joint orientation; Js: Joint spacing; Wt: Water table; Per.: Permeability; Wi: Water inflow; Ge: Gas emission; Sq.: Squeezing; Mf: Mixed face; Kc: Karstic condition; Qc: Abrasive rocks (quartz contents); CZ: Width of a crushed zone or fractured zone; Cc: Clay contents in mass rock (muddy condition)

Figure 4: Histogram of (a) Tq, (b) RPM, (c) Fp, (d) VI

dex (VI) as the effect of media characteristics were used as input parameters and Advance Rate (AR) was the output parameter. Data distributions for these parameters are shown in Figure 4. Also, by using the database, a simple regression analysis was performed to identify the relationship between each parameter with AR as shown
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To carry out a multiple linear regression analysis. According to the statistical analysis, the predictive model is shown in Equation 3:

\[
AR = -0.00004T_q + 0.3662RPM + 0.002F_n - 0.0088VI + 0.9039
\]

Where:
- \(AR\) – Advance rate (m/h),
- \(T_q\) – Torque (nominal) of the cutterhead (kN.m),
- \(RPM\) – Cutter head rotation per minute (rotation/minute),
- \(F_n\) – Cutter normal force (kN),
- \(VI\) – Index of Vulnerability.

### Table 11: Model summary for Equation 3

| Model | R  | \(R^2\) | Adjusted \(R^2\) | Std. Error of the Estimate |
|-------|----|---------|------------------|---------------------------|
| 1     | 0.935 | 0.874 | 0.874 | 0.12560 |

### Table 12: Coefficients of multiple linear regression and collinearity statistics for Equation 3

| Model | Unstandardized Coefficients | Standardized Coefficients | t    | Sig. | Collinearity Statistics |
|-------|-----------------------------|---------------------------|------|------|-------------------------|
|       | B   | Std. Error | Beta |      | Tolerance | VIF |
| (Constant) | 0.904 | 0.028 |        | 32.450 | 0.000 |
| \(T_q\) | -4.954E-5 | 0.000 | -0.019 | -2.345 | 0.019 | 0.976 | 1.025 |
| \(F_n\) | 0.002 | 0.000 | 0.197 | 24.483 | 0.000 | 0.946 | 1.057 |
| RPM | 0.366 | 0.003 | 0.858 | 107.655 | 0.000 | 0.964 | 1.038 |
| VI | -0.009 | 0.000 | -0.259 | -33.068 | 0.000 | 0.996 | 1.004 |
The model summary for Equation 3 is shown in Table 11. Also, to check the degree of correlation between input variables, a multicollinearity analysis was carried out. The Variance Inflation Factor (VIF), as one of the most useful tools for finding the level of multicollinearity, was used in this research. Its range varies from one to infinity. In general, VI greater than 10 shows a possibility of multicollinearity (Montgomery and Peck, 1992).

The VIF values for independent variables in Equation 3 were determined and can be observed in Table 12. As can be seen from Table 12, the VIF for each independent variable in Equation 3 is less than 10. It shows that there is no high correlation between input independent variables. Furthermore, analysis of variance (ANOVA) for Equation 3 is illustrated in Table 13. The model statistic

| Model          | Sum of Squares | df | Mean Square | F     | Sig. |
|----------------|----------------|----|-------------|-------|------|
| Regression     | 225.086        | 4  | 56.272      | 3.567E3 | .000 |
| Residual       | 32.404         | 2054 | 0.016       |       |      |
| Total          | 257.490        | 2058 |             |       |      |

| Table 13: Variance (ANOVA) analysis for Equation 3 |

Variance Inflation Factor (VIF), as one of the most useful tools for finding the level of multicollinearity, was used in this research. Its range varies from one to infinity. In general, VI greater than 10 shows a possibility of multicollinearity (Montgomery and Peck, 1992).

The VIF values for independent variables in Equation 3 were determined and can be observed in Table 12. As can be seen from Table 12, the VIF for each independent variable in Equation 3 is less than 10. It shows that there is no high correlation between input independent variables. Furthermore, analysis of variance (ANOVA) for Equation 3 is illustrated in Table 13. The model statistic

| Table 14: TBM performance prediction empirical models |

| Model                                | Relation                                   |
|--------------------------------------|--------------------------------------------|
| Graham (1977)                        | \[ PR_{cr} = 3940 \times \frac{F_n}{TS} \] |
| Farmer and Glossop (1980)            | \[ PR_{cr} = 624 \times \frac{F_n}{TS} \] |
| Cassinelli et al. (1982)             | \[ PR = -0.0059 \times RSR + 1.5 \]       |
| Innaurato et al. (1991)              | \[ PR = \sigma_{un}^{0.437} - 0.047 \times RSR + 3.15 \] |
| Barton, (1999)                       | \[ PR = 5 \times Q_{TBM}^{0.2} \]         |
| Hassanzpour et al. (2009, 2010)      | \[ FPI = 0.425 \times RMCI + 11.28 \]     |
|                                      | \[ RMCI = 0.01 \times UCS \times RQD^2 \] |
| Khademi et al. (2010)                | \[ FPI = 4.161 + 0.091 \times UCS + 0.077 \times J_c - 1.077 \times \log(\alpha) \] |

PR: penetration rate, PRev: penetration per revolution, FPI: field penetration index, \( F_n \): cutter normal force, UCS: uniaxial compressive strength of intact rock, TS: tensile strength, RQD: rock quality designation, RMCI: rock mass cuttability index, \( \sigma_{un} \): uniaxial compressive strength of rock, \( \alpha \): the angle between the tunnel axis and the planes of weakness, Jc: RMR joint condition partial rating, RSR: rock structure rating, RMR: rock mass rating, \( Q_{TBM} \): Barton rock mass quality rating for TBM driven tunnels

4.3.2. Multiple non-linear regression modelling

For multiple non-linear regressions modelling, polynomial, power, exponential, and logarithmic models with the same datasets were used. For the polynomial model with \( R^2=0.86 \) obtained, the mathematical equation is as shown in Equation 4:

\[
AR = \frac{1.9016}{J_{q}^{0.0738}} + 1.0445 \times RPM^{0.6195} +
+ 1.8003 \times \frac{J_{q}^{0.1099}}{\pi} + \frac{7.6556}{\sqrt{\frac{0.5595}{\pi}}} - 10.5979
\] (4)
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5. Evaluation performance of the models

The performance of regression models were compared with the models from literature such as Khademi et al. model, Hassanpour et al. model, NTNU model, Farmer and Glossop model, Cassinelli et al. model, Innaurato et al. model, and Graham model.

Using the power model, the relation with $R^2=0.87$ is as shown in Equation 5:

$$AR = 0.8578 \cdot \left(\frac{RPM^{0.7209} F_{n}^{0.1113}}{T_{p}^{0.0187} V_{I}^{0.1316}}\right)$$

(5)

Furthermore, the mathematical equation for logarithmic model with $R^2=0.85$ is as shown in Equation 6:

$$AR = -0.0504 \cdot \ln(T_{p}) + 1.9207 \cdot \ln(RPM) + 0.2931 \cdot \ln(F_{n}) - 0.3435 \cdot \ln(V_{I}) - 0.2936$$

(6)

Finally, for the exponential model with $R^2=0.87$, the relation is as shown in Equation 7:

$$AR = \exp(-0.000027T_{p} + 0.1320 \cdot RPM + 0.0076F_{n} - 0.0033V_{I} + 0.3431)$$

(7)

Using the power model, the relation with $R^2=0.87$ is as shown in Equation 5:
et al. (2010), Hassanpour et al. (2009, 2010), $Q_{\text{TBM}}$ (Barton, 1999), NTNU (Bruland, 2012), Farmer and Glossop (1980), Cassinelli et al. (1982), Innaurato et al. (1991), and Graham (1977) (Table 14), using 110 randomly selected datasets (out of 2168 datasets) from the Zagros long tunnel, which were not used in the modelling. Two signs, the determination coefficient ($R^2$) and the Root Mean Square Error (RMSE) (Equations 8 and 9) were used to do the models performance evaluation and the results obtained are illustrated in Table 15. The predicted AR from all models for 110 datasets was compared with the recorded AR as shown in Figure 6. The predicted AR from these models for five datasets was compared with the recorded AR as illustrated in Figure 7. As it can be seen from Figure 7 and Table 15 the linear regression model with $R^2=0.83$ and RMSE=0.12 shows the best performance among the models.

\begin{equation}
R^2 = \frac{\left(\sum_{i=1}^{n} (x_{\text{predicted}} - \bar{x}_{\text{predicted}})(x_{\text{actual}} - \bar{x}_{\text{actual}})\right)^2}{\sum_{i=1}^{n} (x_{\text{predicted}} - \bar{x}_{\text{predicted}})^2 \sum_{i=1}^{n} (x_{\text{actual}} - \bar{x}_{\text{actual}})^2}
\end{equation}

Where:

- $R^2$ – Coefficient of determination,
- $x_{\text{actual}}$ – The $i_{\text{th}}$ measured element,
- $x_{\text{predicted}}$ – The $i_{\text{th}}$ predicted element,
- $n$ – The number of datasets.

\begin{equation}
\text{RMSE}(x) = \sqrt{\frac{1}{n} \sum_{i=1}^{n} (x_{\text{actual}} - x_{\text{predicted}})^2}
\end{equation}

Where:

- RMSE – Root Mean Square Error,
- $x_{\text{actual}}$ – The $i_{\text{th}}$ measured element,

6. Conclusions

In this research work, using parameters of machine and media characteristics, different models were developed based on linear and non-linear multiple regression analysis to determine the advance rate in rock TBM tunnelling. In the modelling, torque, cutter head rotation per minute, and cutter normal force are machine parameters and the level of geotechnical risks (VI), representing media characteristics, which were used as input parameters and AR was the output parameter. For predicting VI, an RES based model was proposed. The merit of this modelling is that all of the important and easily obtainable parameters affecting geotechnical risks in rock TBM were taken into account.

Validation was carried out, using 110 randomly selected datasets (out of 2168 datasets) from the Zagros long tunnel, which were not used in the modelling. For 110 datasets, recorded AR were compared with the AR estimated through regression models and also with the AR obtained from the models such as Khademi et al., Hassanpour et al., QTBM, NTNU, Farmer and Glossop, Cassinelli et al., Innaurato et al., and Graham. The obtained results showed that the multiple linear regression model with $R^2$ of 0.83 and RMSE of 0.12 has a better performance among the different models.

As it is evident, the linear regression model was constructed based on site-specific parameters and limited datasets from the Zagros long tunnel. This model can be used to estimate the advance rate in the Zagros long tunnel or tunnels with almost the same ground conditions. However, it cannot be generalized for other tunnel pro-
jects. For tunnelling projects with other geological environments, a similar approach with other parameters can be developed.

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Novi model za predviđanje napredovanja tunelskoga bušačeg stroja u tvrdim stijenama

Predviđanje brzine napredovanja tunelskoga bušačeg stroja u tvrdim stijenama jedna je od najvažnijih stvari kod procjene vremena i troškova izrade tunela. Prikazana je priprema stijenskog inženjerskog sustava, modela kojim se predviđa geotehnički rizik (ovisan o svojstvima stijena) kod izrade tunela bušačim strojevima. Uporabljeno je 15 varijabli koje utječu na spomenuti rizik. Na temelju 5 ekspertnih mišljenja postavljena je interakcijska matrica i rangiranje varijabli. Indeks ranjivosti, kao izvedena geotehničkoga rizika, izračunan je za 2058 (od ukupno 2168) skupova podataka, uzorkovanih u 55 geoloških zona kroz 11 km dugačak tunel Zagros. U sljedećem koraku proučeni su operativni parametri strojeva poput okretnoga momenta, rotacije glave sjekača po minuti, sile na glavi te svojstva stijena (prikazanih indeksom ranjivosti). Oni su bili ulazne varijable, a iznos napredovanja izlazna. Na temelju 2058 podataka izračunane su linearni i nelinearni višestruke regresije. Dodatnih 110 podataka (nisu korišteni za modeliranje) uporabljeni su za provjeru regresijskih (i drugih) modela. Rezultati su pokazali kako novi linearni model, s vrijednostima $R^2 = 0,83$ i RMSE = 0,12, ima bolja svojstva nego drugi razmatrani modeli.

Ključne riječi: iznos napredovanja, regresijski model, geotehnički rizik, inženjerski sustav stijena, bušenje u tvrdim stijenama, tunel Zagros

Authors contribution
Mohammad Hossein Arbabsiar (Ph.D. candidate) initialized the idea, completed literature review and participated in all work stages, such as data analysis and fieldwork and executed data analysis and tested its accuracy. Mohammad Ali Ebrahimi Farsangi (Associate Professor) and Hamid Mansouri (Associate Professor) managed the whole process and supervised it from the beginning to the end.

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