REAL-TIME CORRELATION ANALYSIS MONITORING SLOPE MOVEMENT USING SLOPE STABILITY RADAR (SSR605-XT) WITH ROCK MASS RATING AND ALTERATION TYPE DATA

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Abstrak. Lokasi Pit North Main Ridge adalah area penambangan emas terbuka dengan beberapa jenjang dan lereng yang curam, yang memiliki potensi menyebabkan longsor. Oleh karena itu, pemantauan pergerakan lereng sangat penting untuk mengurangi risiko longsor dan memastikan kegiatan penambangan berjalan dengan aman. Pemantauan ini menggunakan metode real-time secara langsung di dalam lubang tambang dengan menggunakan alat Slope Stability Radar (SSR605-XT). Data yang diperoleh dari pemantauan radar sekitar 2 bulan dianalisis menggunakan perangkat lunak IQ Monitor untuk mendapatkan informasi mengenai deformasi dan kecepatan pergerakan lereng. Pemetaan di dalam lubang tambang dilakukan dengan mengukur jarak 5 meter untuk setiap segmen, dengan total panjang 575 meter yang dibagi menjadi 115 segmen. Kondisi lokasi pemantauan di lereng tambang, yang ditandai dengan pergerakan lereng yang signifikan, sangat dipengaruhi oleh data estimasi penilaian massa batuan dan jenis alterasi. Semakin lemah kelas massa batuan (3 dan 4) dan mempunyai tipe alterasi argillic, maka berpotensi terjadinya pergerakan lereng. Sebaliknya semakin kuat kelas massa batuan (1 dan 2) dan mempunyai alterasi silicic dan advanced argillic, maka pergerakan lereng semakin berkurang.

Kata Kunci: alteration; monitoring; Rock Mass Rating; slope stability radar

Abstract. The North Main Ridge Pit location is an open gold mining area with several benches and steep slopes, which have the potential to cause landslides. Therefore, monitoring slope movements is crucial to reducing landslide risks and ensuring safe mining activities. This monitoring utilizes real-time methods directly within the mine pit using the Slope Stability Radar (SSR605-XT) tool. Data obtained from radar monitoring is analyzed using the IQ Monitor software to acquire information on slope deformation and movement speed. Mapping within the mine pit is conducted by measuring a distance of 5 meters for each segment, with a total length of 575 meters divided into 115 segments. The condition of the monitoring site at the mining level, characterized by significant slope movements, is heavily influenced by data estimates of rock mass assessment and alteration types. The weaker the rock mass class (3 and 4) and the argillic alteration type, the potential for slope movement to occur. On the other hand, the stronger the rock mass class (1 and 2) and the silicic and advanced argillic alteration, the less slope movement.

Keywords: monitoring; Rock Mass Rating; slope stability radar; alteration

INTRODUCTION

In recent years, numerous researchers have delved extensively into the intricacies of rock mechanisms (Chen et al., 2023). The analysis of slope stability has consistently remained a complex and pivotal concern within the realm of earth geotechnical engineering (Huang et al., 2024). PT. J Resources Bolaang Mongondow is a company engaged in gold mining operations, employing the openpit method, which involves excavating minerals along a relatively horizontal surface descending towards the mineral deposits, employing a tiered mining approach. Slope stability stands as a critical subject in engineering geology, boasting a rich history spanning centuries. In geotechnical engineering, accurately gauging the safety factor of a slope to avert failure is of paramount importance (Kumar et al., 2023). Slope stability analysis is a critical step in many mining and geotechnical engineering projects, including open-pit mining, embankments, earth dams, landfills, and highways (Zheng et al., 2024). Undertaking such analysis prior to slope alteration is indispensable for devising and executing mitigation strategies aimed at preventing further slope deterioration (Al-E'Bayat et al., 2024). Characterizing rock mass and analyzing rocky slope stability are essential for ensuring user safety (Delgado et al., 2023). Analyzing slope stability in rock masses is complex due to inherent uncertainties and risks (Jia et al., 2023). Assessing landslide susceptibility and hazards in mountainous regions is crucial for managing regional landslide risk. Slopes subjected to gravity can collapse when faced with unfavorable morphology, lithology, structure, or various triggering factors (Kundu et al., 2023).

Slope stability is one of the important topics in engineering geology that has been studied for more than 300 years. Various stability assessment methods have been developed, including limit equilibrium analysis, numerical methods, as well as probabilistic and statistical approaches to determine the factor of safety (F.S) and potential failure mechanisms (Azarafza et al., 2021). Slope stability is a vital concern for safety and production in open-pit mines. Slope stability monitoring techniques track minor initial movements before a landslide occurs. These systems can detect wall deformation with sub-millimeter accuracy and high spatial resolution (Reeves et al., 2001). Adhitama et al. (2021) provide suggestions for geotechnical monitoring and analysis activities by carrying out routine measurements with convergence tools in the research area and monitoring. This slope-monitoring system can use robotic devices and geotechnical sensors (Vasilev & Toshkov, 2016), and slope stability monitoring also considers geological and topographic conditions (Boyle et al., 2014; Yang et al., 2024). Making mining slopes on rocks that are manageable can cause landslide problems. This can result in accidents for mining workers and damage to heavy equipment at the mining front. Companies must improve security to reduce landslide potential (Sulistio & Wijaya, 2022). Slope stability is considered a significant problem. Also, slope failure can cause psychological damage, including loss of property and human lives worldwide (Moayedi et al., 2021; Kumar et al., 2023).

The safety of mining operations depends on effective mine design. This design is influenced by the quality of the rock mass, which varies between different mining sites based on geological conditions (Wijaya et al., 2014). To categorize rocks, the rock mass classification system is used, which has been applied in various engineering projects and stability studies. This system focuses on rock mass parameters and their applications in engineering, such as tunnels, slopes, and foundations. Rock mass classification is particularly beneficial in locations where sample collection and observation are challenging (Qazi & Singh, 2023).

Natural rock masses form in specific geological contexts and are made up of many rocks and structural planes with variable properties. When these rock masses are integrated into the foundations of dams, slopes, and underground structures, they become engineered rock masses with both natural and engineered features. Due to their complexity, engineered rock masses are classified into different categories based on rock mass quality and stability, guided by rock engineering characteristics and practical experience (Wu et al., 2023a).

The Modified Slope Mass Rating (M-SMR) system has been successfully utilized in re-proposing slope levels (Abd Rahim et al., 2023). The occurrence of unforeseen slope failures in surface mines prompted the creation of various monitoring instruments for pit slopes. Among these tools utilized for continuous monitoring of open-pit mine slopes is the slope stability radar (SSR) (Gong et al., 2021). SSR is a technology that can monitor strata movements effectively (Kumar, 2020; Wang et al., 2024). SSR is a technology capable of effectively monitoring movements within rock strata (Kumar, 2020; Wang et al., 2024). The Slope Stability Radar (SSR) system provides continuous surveillance of pit walls and has the capability to identify slope deformations at a sub-millimeter scale (Shellam & Coggan, 2020). Utilizing electromagnetic waves, slope stability radar (SSR) serves as a technological tool for monitoring slope stability. Swift and accurate predictions of slope stability are vital for ensuring safe operations and cost-efficient maintenance of slopes (Zheng et al., 2024). Slope Stability Radar is adept at detecting and interpreting movements within steep and unstable pits (Saunders et al., 2016). Monitoring findings are presented in the form of trends or graphs and are subsequently analyzed. Key elements for geotechnical monitoring include slope control points determined through GPS or total

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station, visual inspections, and geotechnical instruments. These factors play a crucial role in guiding subsequent actions based on standard operating procedures to mitigate potential landslides. This type of alteration influences slope stability through changes in the mechanical properties of rocks, the formation of weak zones, and increased water interaction with the rock mass. Rock alteration in a geotechnical context is significant for designing stable slopes, especially in open-pit mining areas (Kumar & Rathee, 2017).

Slope stability radar is used in open-pit mines to monitor slope movements effectively. This technology employs sensors and computerized analysis techniques in a variety of geotechnical sectors to reduce negative impacts and prevent early mine design failures, guaranteeing that mining activities may continue safely (Elmouttie & Dean, 2020). A theoretical framework is crucial for efficiently and cost-effectively monitoring slope stability. According to simulation data, the horizontal movement of slopes peaks around 12 meters from the slope's toe.

RESEARCH METHOD

This research method uses quantitative methods. This research carried out direct observations and measurements in the field. Observations include North Main Ridge pit slope conditions, alteration type, rock mass classification, potential areas prone to landslides at the mine level, and direct slope monitoring. This study examines rock mass classification based on Wijaya and Isnawan's (2015) approach, which assesses landslide potential on mining slope walls by evaluating rock mass strength. The quality of rock masses in open-pit mines can progressively degrade over time due to vibrations from repeated blasting and other mining activities (Saunders et al., 2020). The necessary slope condition data includes slope orientation, rock strength, geological structures, groundwater conditions, and the type of rock forming the slope. Khajehzadeh et al. (2022), who applied ANN and adaptive algorithms, provide a strong foundation for developing a more accurate predictive model to identify landslide potential using real-time data from SSR605-XT. The SSR605-XT radar monitoring captures key data on slope movement, including coordinates, deformation trends, and velocity, offering valuable insights for slope stability assessment. Sampling, observing, and determining the type of rock alteration is vital in geological exploration and mineralization evaluation. Rock samples were taken using drill core samples in several representative locations. Then, the location points are recorded using GPS, followed by a visual description of the rocks and geological structures. Observations were carried out using a megascopic method to determine the physical and mineralogical characteristics of rocks. Then, microscopic analysis, XRD, and chemical analysis will be performed to determine the elemental content and type of mineral. Determining the type of rock alteration is observing the color and process of change due to interaction with hydrothermal fluids and petrographic analysis for alteration minerals.

Data processing is carried out after all the data is complete. Radar monitoring analysis is carried out using Monitor IQ software, which is the basis for creating maps of monitoring location results that display areas that are safe and prone to landslides (Indriastuty et al., 2021).

RESULTS AND DISCUSSION

Research sites

The study is conducted at the PT's North Main Ridge gold mining pit site, which is situated within J Resources Bolaang Mongondow in Bakan Village, Lolayan District, Bolaang Mongondow Regency, North Sulawesi Province.

Characteristics of Rock Masses and Alteration Zones

Over the years, several surveys have been conducted to evaluate the geomechanical characteristics of rock masses within the mining area. Two notable assessments stand out: the first, dating back to 2011, was conducted as part of the mine expansion plan, covering the entire extension and containing extensive data spread across a wide area. The second survey, initiated since 2020, followed the most

severe instability events, focusing on enhanced accuracy and attention to identifying zones of weakness (Martinelli et al., 2023). Geotechnical mapping and laboratory testing of rock physical and mechanical properties were employed to characterize the rock masses (Wu et al., 2023b).

The classification of rock masses on mining slopes utilizes the Rock Mass Rating System (RMR) (Bieniawski, 1989; Ganesan & Mishra, 2024). RMR is employed to evaluate rock mass quality, predesign excavations, and conduct procedures within this framework (Jaiswal et al., 2024). It offers recommendations for underground tunnel support primarily based on empirical dimensional measurements, including those by Terzaghi, Bienawski, and Nick Barton (Wardana & Wijaya, 2021). RMR provides data on numerous rock mass parameters, which are empirically estimated to yield an elemental rock mass RMR value for the first assessment of rock mass quality. Fundamental RMR information is useful in the early stages of slope stability assessment (Singh & Kumar, 2020). RMR is widely used in a variety of engineering disciplines dealing with rocks, including mining, civil tunnel building, hilly terrain highways, bridges, dams, and hydroelectric power projects (Jaiswal et al., 2024).

The research location was observed at several points, each with several segments. Each segment was observed for the characteristics of the RMR rock mass and the type of alteration. The results of the RMR analysis are: Class 2 (29.57%), Class 3 (20%) and Class 4 (50.43%). At the observation location, there are three types of alteration, namely: silicic type (7.94%), advanced argillic (21.43%), and argillic (70.63%). The results of the Rock Mass Rating System (RMR) estimation and alteration type for each observation segment at the North Main Ridge site are shown in (Table 1).

Slope Stability Radar Monitoring Results

Slope stability is a key attribute in geotechnical engineering systems that can be analyzed through the calculation of the factor of safety (Singh et al., 2023). Rock slope stability poses a significant challenge in the field of rock engineering. The mechanical properties of rock masses and discontinuities are difficult to ascertain directly due to scale effects, leading to uncertainties. Additionally, the presence of various failure mechanisms, particularly those involving complex failures, further complicates the process of obtaining viable solutions (Oliveira & Lana, 2023). By utilizing SSR605-XT radar technology, which can detect slope deformation with sub-millimeter accuracy and high spatial resolution, this study integrates real-time data on deformation and slope movement speed with rock mass classification and alteration types. This approach enables the identification of areas with high instability potential, such as expanding plastic zones and significant horizontal displacement, thereby improving the accuracy of landslide prediction and prevention (Liu et al., 2021). In this study, the data collected by the SSR605-XT includes time, maximum deformation, and maximum speed over a specified period. Each pixel selected for analysis represents a division of segments on benches 760, 745, 730, 715, and 705. By monitoring using radar, slope movement stages can be observed to see whether they fall into linear, regressive, and progressive stages. To find out the detailed hourly speed and daily speed, use the VCP60 and VCP1440 settings. Knowing the maximum speed with the VCP60 and VCP1440 settings correlates with the threshold data to determine the risk value. CP60 and **VCP1440** represent different configurations that dictate the operational speeds of a system or process. The maximum speeds allowed by these settings can significantly influence performance and safety metrics. Risk values can be calculated by analyzing historical data related to speed settings and their outcomes. For instance, if data shows that operating at or above the maximum speed of VCP1440 correlates with a high incidence of failures, this can be quantified into a risk value that reflects the potential for loss. Thresholds are predefined limits that indicate acceptable levels of performance or risk. They help in identifying when a system operates within safe parameters versus when it may be at risk of failure or inefficiency. When the maximum speed exceeds established thresholds, it raises the likelihood of adverse events. This correlation allows for the assessment of risk values based on how often and by how much these speeds are exceeded. The results of monitoring deformation and speed of slope movement are presented in (Table 2).

No	Location	Segment	RMR Basic	RMR Class	Alteration Type
1	760	A' – B'	54	3	Argilic
2		B' - C'	54	3	Argilic
3		C' – D'	54	3	Argilic
4		D' - E'	63	2	Adv. Argilic
5		E' - F'	63	2	Adv. Argilic
6		F' - G'	65	2	Adv. Argilic
7		G' – H'	54	3	Argilic
8		H' – I'	54	3	Argilic
9		I' – J'	53	3	Argilic
10	745	A' – B'	32	4	Argilic
11		B' - C'	54	3	Argilic
12		C' – D'	50	3	Argilic
13		D' - E'	38	4	Argilic
14		E' – F'	38	4	Argilic
15		F'-G'	38	4	Argilic
16		G' – H'	38	4	Argilic
17		H' – I'	38	4	Argilic
18		I' - J'	38	4	Argilic
19		J' – K'	38	4	Argilic
20		K' - L'	38	4	Argilic
20		K = L L' – M'	38	4	Argilic
21		M' - N'	38	4	Argilic
22		N' - N' N' - O'	38	4	Argilic
23 24		$\mathbf{N} = \mathbf{O}$ $\mathbf{O}' - \mathbf{P}'$	38	4	Argilic
24 25		O = P P' - Q'	38		
23 26		P = Q Q' - R'	38	4	Argilic
		Q = R R' - S'		4	Argilic
27			38	4	Argilic
28 20		S' - T'	38	4	Argilic
29 20		T' - U'	38	4	Argilic
30		U' - V'	38	4	Argilic
31		V' – W'	38	4	Argilic
32		W' – X'	38	4	Argilic
33		X' – Y'	38	4	Argilic
34		Y' - Z'	38	4	Argilic
35		Z'-AA'	38	4	Argilic
36		AA' – BB'	38	4	Argilic
37		BB' - CC'	38	4	Argilic
38		CC' - DD'	38	4	Argilic
39		DD' - EE'	38	4	Argilic
40	730	A' – B'	80	2	Silicic
41		B' - C'	74	2	Silicic
42		C' – D'	74	2	Silicic
43		D'-E'	67	2	Adv. Argilic
44		E'-F'	64	2	Adv. Argilic
45		F' - G'	69	2	Adv. Argilic
46		G' - H'	69	2	Adv. Argilic
47		H' - I'	70	2	Adv. Argilic
48		I' – J'	67	2	Adv. Argilic

Table 1. Estimated RMR and Alteration Type at the North Main Ridge Site

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49	J' - K'	69	2	Adv. Argilic
50	K' – L'	54	3	Argilic
51	L' - M'	39	4	Argilic
52	M' - N'	39	4	Argilic
53	N' - O'	39	4	Argilic
54	O' - P'	39	4	Argilic
55	P' - Q'	39	4	Argilic
56	Q' - R'	39	4	Argilic
57	$\mathbf{R}^{\prime} - \mathbf{S}^{\prime}$	39	4	Argilic
58	S' - T'	39	4	Argilic
59	Σ 1 Τ'-U'	39	4	Argilic
60	U' – V'	39	4	Argilic
61	V' – W'	39	4	Argilic
62	W' – X'	54	3	Argilic
63	X' - Y'	54	3	Argilic
64	Y' - Z'	69	2	Adv.Argilic
65	Z' - AA'	54	3	Argilic
66	AA' - BB'	53	3	Argilic
67	BB' - CC'	67	2	Adv.Argilic
68	CC' – DD'	58	3	Adv.Argilic
69	DD' – EE'	33	4	Argilic
70	EE' – FF'	29	4	Argilic
70	FF' – GG'	38	4	Argilic
72 715	A' – B'	74	2	Silicic
72 713	$\mathbf{B}^{\prime} - \mathbf{C}^{\prime}$	74	2	Silicic
73 74	$\mathbf{D} = \mathbf{C}$ $\mathbf{C}' - \mathbf{D}'$	72	2	Adv. Argilic
74	C = D D' - E'	72	2	Adv. Argilic
75 76	E' - E' E' - F'	72	2	Adv. Argilic
70	F' - G'	52	3	Adv. Argilic
78	G' – H'	64	2	Adv. Argilic
70 79	H' – I'	64	2	Adv. Argilic
80	II - I I' - J'	66	2	Adv. Argilic
81	J' – K'	64	2	Adv. Argilic
82	у - К К' – L'	64	2	Adv. Argilic
83	L' - M'	64	2	Adv. Argilic
84	M' - N'	69	2	Adv. Argilic
85	N' - O'	54	3	Argilic
86	O' - P'	54 54	3	Argilic
87	P' - Q'	52	3	Argilic
88	Q' - R'	53	3	Argilic
89	R' - S'	50	3	Argilic
90	$\mathbf{S}' - \mathbf{T}'$	40	4	Argilic
91	T' – U'	40	4	Argilic
92	U' – V'	40	4	Argilic
92 93	$\mathbf{U} = \mathbf{V}$ $\mathbf{V}' - \mathbf{W}'$	40	4	Argilic
94	$\mathbf{W}^{\prime} - \mathbf{W}^{\prime}$	40	4	Argilic
95 705	$\frac{W - X}{A' - B'}$	77	2	Silicic
95 705 96	A = B B' - C'	77	2	Silicic
90 97	$\mathbf{D} = \mathbf{C}$ $\mathbf{C}' = \mathbf{D}'$	71	2	Adv. Argilic
97 98	C = D D' - E'	75	2	Adv. Argilic
98 99	$\mathbf{D}' = \mathbf{E}'$ $\mathbf{E}' - \mathbf{F}'$	73 58	3	Silicic
))	$\Gamma = L$	50	5	Juleic

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100	F' - G'	50	3	Silicic
101	G' - H'	40	4	Adv. Argilic
102	H'- I'	28	4	Silicic
103	I' – J'	40	4	Adv. Argilic
104	J' - K'	40	4	Adv. Argilic
105	K' – L'	40	4	Argilic
106	L' - M'	40	4	Argilic
107	M' - N'	40	4	Argilic
108	N' - O'	40	4	Argilic
109	O' – P'	40	4	Argilic
110	P' – Q'	40	4	Argilic
111	Q' – R'	40	4	Argilic
112	R' - S'	40	4	Argilic
113	S' - T'	40	4	Argilic
114	T' - U'	40	4	Argilic
115	U' - V'	40	4	Argilic

Table 2. Estimated Monitoring Results of Deformation and Slope Movement Speed

No	Location	Segment	Alteration Type	Pixel Radar (X,Y)	Deformati on Max (MM)	VCP 60 (MM/hou r)	VCP 60 (MM/hou r)	Risk
1	760	A' – B'	Argilic	-	-	-	-	Lowrisk
2		B' - C'	Argilic	-	-	-	-	Lowrisk
3		C' – D'	Argilic	118.69	6.3	3.4	2.5	Mediumrisk
4		D' – E'	Adv. Argilic	119.68	2.4	2.8	3.8	Mediumrisk
5		E' - F'	Adv. Argilic	-	-	-	-	Lowrisk
6		F' - G'	Adv. Argilic	-	-	-	-	Lowrisk
7		G' – H'	Argilic	-	-	-	-	Lowrisk
8		H' - I'	Argilic	-	-	-	-	Lowrisk
9		I' – J'	Argilic	-	-	-	-	Lowrisk
10	745	A' - B'	Argilic	-	-	-	-	Lowrisk
11		B' - C'	Argilic	-	-	-	-	Lowrisk
12		C' – D'	Argilic	-	-	-	-	Lowrisk
13		D' – E'	Argilic	126.65	7.3	2.9	3.9	Mediumrisk
14		E' - F'	Argilic	128.65	6.7	2.6	3	Mediumrisk
15		F' - G'	Argilic	-	-	-	-	Lowrisk
16		G' – H'	Argilic	-	-	-	-	Lowrisk
17		H' - I'	Argilic	-	-	-	-	Lowrisk
18		I' – J'	Argilic	-	-	-	-	Lowrisk
19		J' - K'	Argilic	-	-	-	-	Lowrisk
20		K' – L'	Argilic	-	-	-	-	Lowrisk
21		L' - M'	Argilic	-	-	-	-	Lowrisk
22		M' - N'	Argilic	-	-	-	-	Lowrisk
23		N' – O'	Argilic	-	-	-	-	Lowrisk
24		O' – P'	Argilic	-	-	-	-	Lowrisk
25		P' - Q'	Argilic	-	-	-	-	Lowrisk
26		Q' – R'	Argilic	-	-	-	-	Lowrisk
27		R' - S'	Argilic	-	-	-	-	Lowrisk
28		S' - T'	Argilic	-	-	-	-	Lowrisk

29		T' - U'	Argilic	-	-	-	-	Lowrisk
30		U' - V'	Argilic	-	-	-	-	Lowrisk
31		V' - W'	Argilic	-	_	_	_	Lowrisk
32		W' – X'	Argilic	-	_	-	-	Lowrisk
33		X' - Y'	Argilic	-	_	-	_	Lowrisk
34		X - Z'	Argilic	-	_	_	_	Lowrisk
35		1 - L Z' – AA'	Argilic	-	_	-	_	Lowrisk
36			Argilic	-	-	-	-	Lowrisk
30		AA' – BB'	Arguie	-	-	-	-	LOWIISK
37			Ancilia					Louwick
57			Argilic	-	-	-	-	Lowrisk
20		CC'	4 .7.					T L
38		CC' –	Argilic	-	-	-	-	Lowrisk
20		DD'						
39		DD' –	Argilic	-	-	-	-	Lowrisk
		EE'						
40	730	A' – B'	Silicic	-	-	-	-	Lowrisk
41		B' – C'	Silicic	-	-	-	-	Lowrisk
42		C' – D'	Silicic	-	-	-	-	Lowrisk
43		D' -E'	Adv. Argilic	-	-	-	-	Lowrisk
44		E' - F'	Adv. Argilic	-	-	-	-	Lowrisk
45		F' - G'	Adv. Argilic	-	-	-	-	Lowrisk
46		G' – H'	Adv. Argilic	-	-	-	-	Lowrisk
47		H' – I'	Adv. Argilic	-	-	-	-	Lowrisk
48		I' – J'	Adv. Argilic	-	-	-	-	Lowrisk
49		J' - K'	Adv. Argilic	-	-	-	-	Lowrisk
50		K' – L'	Argilic	-	-	-	-	Lowrisk
51		L' - M'	Argilic	-	-	-	-	Lowrisk
52		M' - N'	Argilic	-	-	-	-	Lowrisk
53		N' - O'	Argilic	-	-	-	-	Lowrisk
54		O' – P'	Argilic	-	-	-	-	Lowrisk
55		P' – Q'	Argilic	-	-	-	-	Lowrisk
56		Q' – Ř'	Argilic	-	-	-	-	Lowrisk
57		R' – S'	Argilic	145.62	167	16	50	Highrisk
58		S' - T'	Argilic	146.62	122	12	18	Highrisk
59		T' - U'	Argilic	147.62	122.7	12	18.6	Highrisk
60		U' – V'	Argilic	151.62	90	3	7.6	Highrisk
61		U' – W'	Argilic	-	-	-	-	Lowrisk
62		W' – X'	Argilic	-	_	-	-	Lowrisk
63		X' - Y'	Argilic	152.62	260	4.4	25	Highrisk
64		Y' - Z'	Adv.Argilic	153.6	335	3.4	3	Mediumrisk
65		Z' - AA'	Argilic	155.61	121	4	17.9	Mediumrisk
66		AA'-BB'	Argilic	-	-	-	-	Lowrisk
67		BB' –	Arguic Adv.Argilic	_	-	_	_	Lowrisk
07		DD – CC'						Lowrisk
68		CC' –	Adv.Argilic	-	_	_	_	Lowrisk
00		DD'	Auv.Arguic	-	-	-	-	LOWIISK
69		DDI	Arailia	156.61	361	26	206	Highrisk
09		DD' – EE'	Argilic	130.01	301	20	200	Tignitsk
70			Anailia	157 66	207	10.5	61 0	Highwigh
70		EE' - FF'	Argilic Argilic	157.66	207	10.5	61.8	Highrisk Historial
71	715	FF' - GG'	Argilic	158.62	100.21	11.49	56	Highrisk
72	715	A' – B'	Silicic	-	-	-	-	Lowrisk

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73	B' - C'	Silicic	-	-	-	-	Lowrisk
74	C' – D'	Adv. Argilic	-	-	-	-	Lowrisk
75	D' - E'	Adv. Argilic	-	-	-	-	Lowrisk
76	E' - F'	Adv. Argilic	-	-	-	-	Lowrisk
77	F' - G'	Adv. Argilic	-	-	-	-	Lowrisk
78	G' - H'	Adv. Argilic	-	-	-	-	Lowrisk
79	H' – I'	Adv. Argilic	-	-	-	-	Lowrisk
80	I' – J'	Adv. Argilic	-	-	-	-	Lowrisk
81	J' - K'	Adv. Argilic	-	-	-	-	Lowrisk
82	K' – L'	Adv. Argilic	-	-	-	-	Lowrisk
83	L' - M'	Adv. Argilic	-	-	-	-	Lowrisk
84	M' - N'	Adv. Argilic	-	-	-	-	Lowrisk
85	N' - O'	Argilic	-	-	-	-	Lowrisk
86	O' – P'	Argilic	-	-	-	-	Lowrisk
87	P' – Q'	Argilic	131.66	14.	2.9	2.4	Mediumrisk
88	Q'-R'	Argilic	132.66	5.6	3.3	2.6	Mediumrisk
89	R' - S'	Argilic	135.65	10.9	4.4	4.3	Mediumrisk
90	S' - T'	Argilic	136.65	6.9	3.9	3.6	Mediumrisk
91	T' - U'	Argilic	-	-	-	-	Lowrisk
92	U' - V'	Argilic	-	-	-	-	Lowrisk
93	V' - W'	Argilic	-	-	-	-	Lowrisk
94	W' - X'	Argilic	-	-	-	-	Lowrisk
95 705	A' – B'	Silicic	-	-	-	-	Lowrisk
96	B' - C'	Silicic	-	-	-	-	Lowrisk
97	C' - D'	Adv. Argilic	-	-	-	-	Lowrisk
98	D' - E'	Adv. Argilic	-	-	-	-	Lowrisk
99	E' - F'	Silicic	-	-	-	-	Lowrisk
100	F' - G'	Silicic	-	-	-	-	Lowrisk
101	G' - H'	Adv. Argilic	-	-	-	-	Lowrisk
102	H' – I'	Silicic	-				I
103				-	-	-	Lowrisk
	I' – J'	Adv. Argilic	-	-	-	-	Lowrisk Lowrisk
104	I' – J' J' – K'	Adv. Argilic Adv. Argilic	-	-	- -	- -	
			-	-	- - -	- - -	Lowrisk
104	J' - K'	Adv. Argilic	-	-	- - - -	- - -	Lowrisk Lowrisk
104 105	J' – K' K' – L'	Adv. Argilic Argilic	-				Lowrisk Lowrisk Lowrisk
104 105 106	J' – K' K' – L' L' – M'	Adv. Argilic Argilic Argilic	- - -		- - - -		Lowrisk Lowrisk Lowrisk Lowrisk
104 105 106 107	J' – K' K' – L' L' – M' M' – N'	Adv. Argilic Argilic Argilic Argilic Argilic	- - -				Lowrisk Lowrisk Lowrisk Lowrisk Lowrisk
104 105 106 107 108	J' – K' K' – L' L' – M' M' – N' N' – O'	Adv. Argilic Argilic Argilic Argilic Argilic Argilic	- - -	- - - - - - 9.7	- - - - - 3.5	- - - - - 3.9	Lowrisk Lowrisk Lowrisk Lowrisk Lowrisk Lowrisk
104 105 106 107 108 109	J' – K' K' – L' L' – M' M' – N' N' – O' O' – P'	Adv. Argilic Argilic Argilic Argilic Argilic Argilic Argilic	- - - -	- - - - - 9.7 6.6	- - - - 3.5 3.1	- - - - 3.9 3.6	Lowrisk Lowrisk Lowrisk Lowrisk Lowrisk Lowrisk Lowrisk
104 105 106 107 108 109 110	J' – K' K' – L' L' – M' M' – N' N' – O' O' – P' P' – Q'	Adv. Argilic Argilic Argilic Argilic Argilic Argilic Argilic Argilic	- - - - 124.62				Lowrisk Lowrisk Lowrisk Lowrisk Lowrisk Lowrisk Lowrisk Mediumrisk
104 105 106 107 108 109 110 111	J' - K' K' - L' L' - M' M' - N' N' - O' O' - P' P' - Q' Q' - R'	Adv. Argilic Argilic Argilic Argilic Argilic Argilic Argilic Argilic Argilic	- - - - 124.62 126.62				Lowrisk Lowrisk Lowrisk Lowrisk Lowrisk Lowrisk Lowrisk Mediumrisk Mediumrisk
104 105 106 107 108 109 110 111 112	J' - K' K' - L' L' - M' M' - N' N' - O' O' - P' P' - Q' Q' - R' R' - S'	Adv. Argilic Argilic Argilic Argilic Argilic Argilic Argilic Argilic Argilic Argilic	- - - - 124.62 126.62 -				Lowrisk Lowrisk Lowrisk Lowrisk Lowrisk Lowrisk Mediumrisk Mediumrisk Lowrisk
104 105 106 107 108 109 110 111 112 113	J' - K' K' - L' L' - M' M' - N' N' - O' O' - P' P' - Q' Q' - R' R' - S' S' - T'	Adv. Argilic Argilic Argilic Argilic Argilic Argilic Argilic Argilic Argilic Argilic Argilic	- - - - 124.62 126.62 - -				Lowrisk Lowrisk Lowrisk Lowrisk Lowrisk Lowrisk Mediumrisk Mediumrisk Lowrisk

Meanwhile, observation photos and slope movement monitoring analysis results are presented in (Figure 1), and Slope Stability Radar pixel photos are presented in (Figure 2).

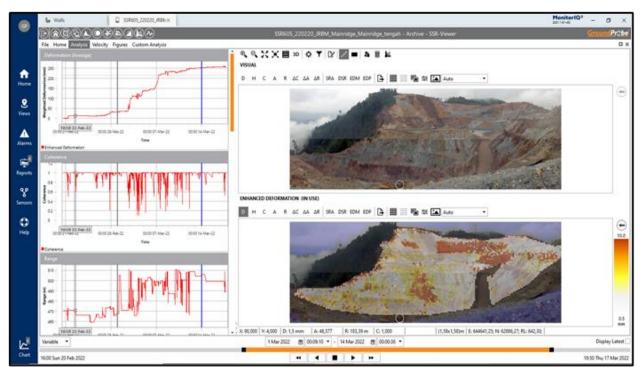


Figure 1. Observation photos and results of slope movement monitoring analysis (View MonitorIQ)

The default software from Groundprobe, MonitorIQ, reads the radar results algorithm. This software can visualize and facilitate real-time slope monitoring analysis and make back analyses of landslide events. There are several parameters of the main graphs for reading radar monitoring results, namely; Deformation: describes the movement of readings based on the wave phase difference compared to the previous reading. Amplitude: the strength of the radar signal reflected by the slope, based on the density of the reflecting surface. Range: The distance between the SSR and the slope based on the wave travel time. Coherence: comparison of the combination of several ranges and amplitudes that have changed in the previous scan. Velocity: the speed of movement of the mine slope towards or away from the radar. Inverse Velocity: an analysis technique used to predict landslide time whose value is the inverse of velocity.



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Artikel Masuk : 25-07-2024

Artikel Diulas : 13-11-2024



Figure 2. Pixel photo (a) and Slope Stability Radar (SSR605-XT) (b)

Correlation Analysis of Radar Monitoring Results with RMR and Alteration Type Data

The research location on bench RL705-RL760 Pit Main Ridge North, as the location of the research object in this analysis, has various maximum deformations and velocities. Of the total 115 segments, seven segments are categorized as high risk (6%), 13 segments are categorized as medium risk (11.30%), and 95 segments are categorized as low risk (82.61%) (Figure 3).

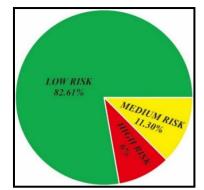


Figure 3. Landslide Risk Distribution Diagram

The classification results for each segment are correlated with the RMR zoning, so it can be seen that the high-risk category is in RMR class 4, while for RMR class 3 and Class 2, it is in moderate-low risk. An example of monitoring analysis in the 730DDEE segment was a linear-progressive-regressive deformation pattern on March 2- 6, 2022. The trend transition from linear-progressive is called the Onset of Failure time, precisely March 3, 2022, at 14.04 WITA with a maximum speed of 21mm/h (using VCP60) and 206mm/day (using VCP1440), is categorized as high-risk velocity. The deformation during the progressive phase is 361mm (Figure 4).

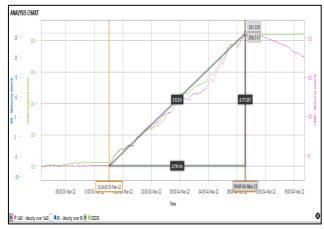


Figure 4. Deformation-Velocity-Time Graph on the 730DDEE Segment

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Landslides are geological disasters that often occur and cause significant losses globally. Evaluating slope stability somewhat is very important (Chen & Dai, 2021) so mitigation measures can be designed and implemented appropriately to avoid worsening slope damage (Al-E'Bayat et al., 2024). Based on the results of monitoring analysis and RMR estimation for each type of alteration in the research area, the landslide-prone zone is divided into 2: the safe zone and the landslide-prone zone. The safe zone includes the silicic and advanced argillic alteration zone, with an RMR value of more than 60 (Class 2) and a low-risk velocity value. Meanwhile, the argillic alteration zone includes the landslide-prone zone includes the formation for each 2) and a low-risk velocity value. Meanwhile, the argillic alteration zone includes the landslide-prone zone. The criteria for each zone are shown in (Table 3).

Criteria	Alteration Type	RMR	Velocity
Safe Zone	Silicic, Advanced Argilic	>60 Class 2	Low Moderate
Landslide Prone Zone	Argilic	< 60 Class 3,4	High Risk

Table 3. Criteria for Landslide Prone Areas

Following is data from 115 segments analyzed (figure 5), there are 16 segments that are prone to landslides, namely:

a). 760 CD	e). 730 ST	i). 730 EEFF	m). 715 RS
b). 760 DE	f). 730 TU	j). 730 FFGG	n). 715 ST
c). 760 EF	g). 730 XY	k). 715PQ	o). 705 PQ
d). 730 RS	h). 730 DDEE	l). 715QR	p). 705 QR

The location map of 16 segments that are prone to landslides (high risk) in the red zone.

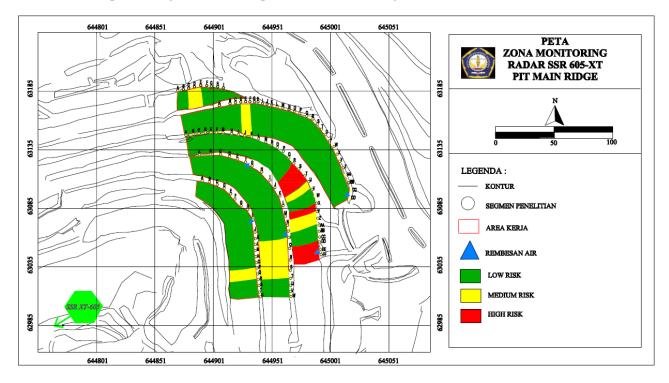


Figure 5. The Location Map of 115 Segments of Radar SSR 605-XT Monitoring, including 16 Segments that are Prone to Landslides (High Risk)

CONCLUSION

The findings from monitoring slope movement using slope stability radar indicate that the stability level of the mine slope is influenced by the characteristics of the rock mass and the type of alteration. In the study area, the safety zone for potential landslides in the North Main Ridge Pit corresponds to RMR class 2, characterized by an RMR rating value exceeding 60, and exhibits low-risk velocities in advanced argillic and silicic alteration types. Conversely, the landslide-prone zone in the North Main Ridge Pit falls within RMR classes 3–4, with an RMR value below 60, and demonstrates high-risk velocities in the argillic alteration type. Future investigations could leverage technologies like artificial intelligence (AI) to assess slope stability, as AI has demonstrated its ability to forecast safety factors for land slopes under both static and seismic loads (Khajehzadeh et al., 2022).

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