

# Mapping of Land Subsidence Vulnerability: Case Study at the Tarkwa-Prestea Mining Areas of Ghana

Edward A. A. KWESI, Kwame N. ASAMOAH, Kwame TENADU, Benjamin E. METEKU, Alexander O. MENSAH, and Gershon PIEDU; Ghana

**Key words:** Land subsidence; Vulnerability Mapping, Mining Areas, Land use, DRASTIC and AHP

## SUMMARY

Landslides and subsidence with disastrous consequences are serious geo-environmental events with high probability of occurrence in mining areas. To minimize their occurrence and negative impacts on humans and the environment, appropriate preventive and mitigation measures must be put in place and these require knowledge and understanding of the risk factors involved and the vulnerable areas within a given geographical region. This paper discusses the combined use of the 'DRASTIC' ground vulnerability modeling technique, the analytical hierarchy process (AHP), Geographic Information System (GIS), GPS and Remote Sensing to collect, process, analyze and evaluate the relative and combined influences of the risk factors involved and to map the susceptible areas of land subsidence in the Tarkwa-Prestea Mining Areas (TPMA) of Ghana. The relevant risk factors identified in the study area include high rainfall, drainage density, elevation and slope, soil, land use/land cover (LULC), depth to ground water, proximity to mine sites, geology and hydrogeology. The relative influence of each of the factors were estimated and combined to generate land subsidence vulnerability maps for the study area. The results indicate that about 11% of TPMA lies within high vulnerability zones and these occur mainly at north-western, western and central parts of the study area. It is recommended that the high vulnerability zones should not be used for siting residential buildings, landfills and other projects that can compound the negative effects of landslides or subsidence in TPMA and similar areas unless adequate interventions measures can be ensured. These areas include highlands and slopes close to active mining sites where blasting and heavy rainfall increase the risk of their occurrence. The results of the current studies may be used as preliminary references or criteria to check the suitability of proposed land uses or development projects in terms of subsidence risk in the study area.

# Mapping of Land Subsidence Vulnerability: Case Study at the Tarkwa-Prestea Mining Areas of Ghana

Edward A. A. KWESI, Kwame N. ASAMOAH, Kwame TENADU, Benjamin E. METEKU, Alexander O. MENSAH, and Gershon PIEDU; Ghana

## 1. INTRODUCTION

Land subsidence may be caused by multiple factors of both natural and anthropogenic sources. Inappropriate and uncontrolled location and development of land-based projects and activities (such as mining, quarries, sand winning, lumber and sawmill operations, residential and commercial developments and waste disposal) can disrupt natural topographic, hydrogeological, biological and climatic settings and increase the frequency or potential of environmental disasters like flooding, fires, earth tremors, landslides and subsidence, and collapse of structures with serious consequences on humans and the environment (Kwesi *et al.*, 2020; Kim *et al.*, 2006; Sun *et al.*, 1999). There is a strong perception that the natural ability of the topography and underlying geology of most mining areas to withstand land subsidence is reduced or compromised due to the long and widespread operations of mining activities like blasting, excavations and piling of loose waste materials on the land surface (Ghorbanzadeh *et al.*, 2020; Kim *et al.*, 2006; Sun *et al.*, 1999). This is especially relevant in mining areas like Tarkwa, Ghana, where underground and surface mining activities have been in operations for over a century, with little or no reliable reclamation of the sites, and rising urbanisation has necessitated the need for other major land uses and developments within mining enclaves (Kwesi *et al.*, 2020; Anon., 2014). It is thus necessary to evaluate the subsidence risk at various sites and apply them to assess the suitability of proposed land uses as a way to reduce the potential occurrence and consequences of land subsidence in mining areas. The focus of this paper is thus to discuss and demonstrate the production and use of land subsidence vulnerability maps for assessing the suitability of proposed land uses and locations of developmental projects in mining areas like Ghana. A case study approach is adopted, using the Tarkwa-Prestea Mining Areas of Ghana as the study location.

## 2. BACKGROUND OF STUDY AREA

### 2.1 Geographical and Socio-economic Setting

The study area is the Tarkwa-Prestea Mining Areas (TPMA) of Ghana which is located generally between latitudes 5° 10' N and 5° 35' N and longitudes 1° 52' W and 2° 14' W (Fig. 1). It lies across two administrative districts in Ghana, namely, the Tarkwa-Nsuaem Municipal Area (TNMA) and the Prestea Hunivalley Municipal Area (PHMA). Tarkwa, Bogoso, Prestea,



About 20 % of the total Tarkwaian rocks within the study area is made up of intrusive igneous rocks, which form conformable to slightly transgressive sills with small number of dykes. The Tarkwaian is underlain by the Birimian Supergroup (Kesse, 1985). The study area is faulted and jointed with the most prominent joints trending in WNW to ESE direction (Hirdes and Nunoo, 1994). The Tarkwaian and Birimian rocks of the area do not have adequate primary porosity. They are largely crystalline and inherently impermeable, unless fractured or weathered (Ewusi *et al.*, 2017). Groundwater occurrence is thus associated with the development of secondary porosity and permeability. The zones of secondary permeability are often discrete and irregular and occur as fractures, faults, lithological contacts and zones of deep weathering (Kortatsi, 2002). Groundwater in the Tarkwa area occurs in two distinct hydraulically connected aquifer systems; an upper weathered zone aquifer and a deeper unweathered aquifer or fractured zones and dyke contacts (Junner *et al.*, 1942). The weathered zone aquifer is generally phreatic and the principal groundwater flow occurs where relic's quartz veins are more abundant. The regolith is generally dominated by clay and silt rendering the aquifer highly porous, with high storage but low permeability. Thus, the aquifers are either unconfined or semiconfined depending on the clay and silt proportion. Aquifers are recharged by direct infiltration of precipitation through brecciated zones and the weathered outcrop has estimated groundwater recharge and evapotranspiration values averaging about 14 % and 54 % respectively (Kuma, 2007; Kortatsi, 2002).

### 3. RESOURCES AND METHODS USED

#### 3.1 Data Sources

Secondary data was used to carry out this research work. The hydrogeological parameters were obtained from previous publications. The Digital Elevation Model (DEM) for the slope analysis was obtained from ASTER Global DEM (GDEM). ASTER GDEM is a product of METI and NASA. The Soil media data was obtained from soil map of Ghana published by FAO ISRIC. Proximity to mining site data was estimated from google earth. For the Land Use/Land Cover (LULC) model, Landsat 8 Image (March 29, 2020 scene; path: 194, row: 56) was downloaded from US Geological Survey's website ([earthexplorer.usgs.gov](http://earthexplorer.usgs.gov)). It was downloaded from the Landsat Level 1 Collection. The data in geotiff format was projected onto UTM zone 30 N and then extracted by mask to the study area. It was then converted from digital numbers (DN) to Top of Atmosphere (TOA) Planetary Spectral Reflectance. The TOA Reflectance data (bands 2, 3, 4, 5, 6 and 7) was composited and classified using the unsupervised classification technique in ESRI ArcMap 10.3 software. The steps and other methods adopted for this study are summarized in the flow chart in Fig. 3 and discussed in the next sections.

### 3.2 Land Subsidence Vulnerability Analysis

Similar to groundwater vulnerability assessment, a number of approaches have been developed for assessing and mapping land subsidence vulnerability (Ghorbanzadeh *et al.*, 2020; Bui *et al.*, 2018; Dehghani *et al.*, 2014; Kim *et al.*, 2006; Saraf and Choudhury, 1998; Kalani *et al.*, 2017;). These may be classified into three main groups—overly and index methods; methods employing process-based simulation models, and statistical methods. In overly and index methods, the main contributing factors to vulnerability are mapped based on available primary and/or derived data. Subjective numerical values (ratings and/or weights) are then assigned to the factors based on their relative contributions towards ground vulnerability. The rated and/or weighted maps are then combined by linear functions to produce resultant vulnerability maps of the study area. The ground vulnerabilities evaluated by such methods are qualitative and relative. The main advantage of such methods is that some of the controlling factors (e.g., lithology and depth to groundwater table) can be evaluated over large areas, making them suitable for regional scale assessment (Kwesi *et al.*, 2020; Jaseela *et al.*, 2016).

With the advent of remote sensing (RS), global positioning system (GPS) and GIS, adoption of such methods for creating vulnerability databases, maps and assessments are no longer so difficult. Several overly and index methods have been developed for groundwater vulnerability and these may be applied in land subsidence analysis. The common ones include ‘DRASTIC’, ‘AVI’, ‘SINTACS’ and ‘EPIK’ ((Aller *et al.*, 1987; Civita, 1993; Van Stempvoort *et al.*, 1993). The DRASTIC method which is the most popular overlay and index method was adopted, modified and integrated with the analytical hierarchy process (AHP) for this study. Details of the DRASTIC method and its application are well discussed in earlier publications (Kwesi *et al.*, 2020; Al-Abadi *et al.*, 2014; Rundquist *et al.*, 1991) and thus not presented in this paper. It is an overlay and index method designed to produce vulnerability scores by combining several thematic maps. Its principles of using the most intrinsic influential factors of ground vulnerability within a given geographical setting was applied in selecting the factors for this study which include depth to water table, elevation, slope, lithology, LULC, soil and proximity to mining sites (Table 1). These factors were not limited in number and in substance to the original seven (7) DRASTIC factors. Also, the weighting method in DRASTIC was replaced by the AHP method to improve its reliability (Kalani *et al.*, 2017; Al-shabeeb, 2016; Saaty and Vargas, 2012). Fig. 3 shows the method, data and processing flow chart used.

### 3.3 Weighting by AHP Method

AHP has been applied in a number of site selection and suitability studies (Kalani *et al.*, 2017; Al-shabeeb, 2016; Saaty and Vargas, 2012; Saaty, 2000; Saaty, 1980). It employs Pairwise Comparison Matrices (PCMs) to compare the relative importance among a set of criteria and then determine their relative weights in a consistent manner. Saaty (1980) suggests a scale from 1 to 9 (Table 1) for PCM elements, where the value of 1 indicates that the criteria are equally important and a value of 9 indicates that the criterion under consideration is extremely

important compared to the other criteria. PCM includes a consistency check where judgement errors are identified and a consistency ratio is calculated by the following formulae:

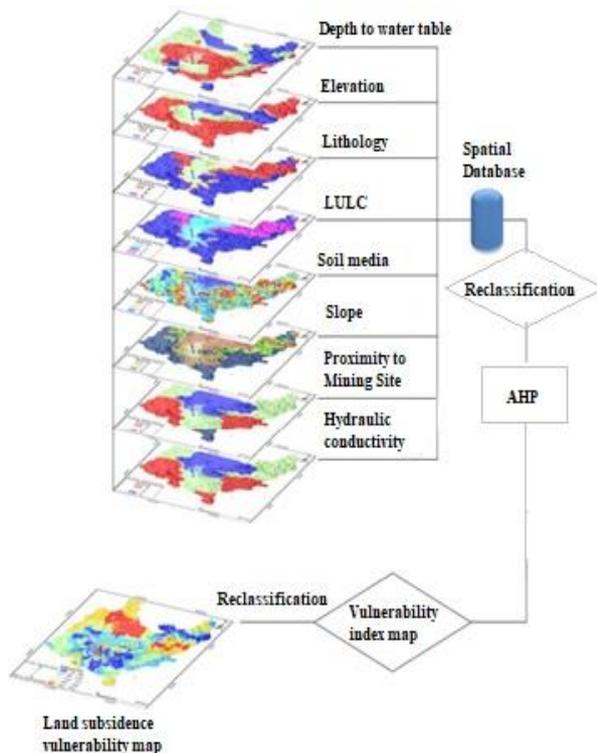
$$\lambda_{max} = \frac{1}{n} \left[ \sum_i^n (Aw_i) / w_i \right] \quad \dots 1$$

$$C.I = (\lambda_{max} - n) / (n - 1) \quad \dots 2$$

$$CR = \frac{CI}{RI} \quad \dots 3$$

where,

$\lambda_{max}$  is the eigenvalue vector;  $n$  is the total number of factors being compared;  $CI$  is the consistency index;  $CR$  is the consistency ratio and  $RI$  is the random consistency index.



**Fig. 3 Flow Chart of the Method**

**Table 1 Ratings and Weights for the Subsidence Vulnerability Parameters**

S. N	Parameter	Range	Rating	AHP Weight
1	Depth to Water table (m)	3.50 - 5.34	9	0.317
		5.35 - 8.00	7	
2	Elevation	4 - 62	10	0.182
		63 - 92	8	
		93 - 125	6	
		126 - 171	4	
3	Lithology	Volcanic rocks	2	0.044
		Quartzite/ Conglomerate	4	
		Phyllite	5	
		Sandstone	6	
4	LULC	Forest	3	0.032
		Sparse vegetation /Farmland	5	
		Built-up /Mine Sites	9	
5	Soil Media	Silt	5	0.207
		Laterite	3	
6	Slope (%)	0 - 7	10	0.124
		8 - 13	9	
		14 - 20	5	
		21 - 31	3	
7	Proximity to Mining Site	32 - 73	1	0.024
		0 - 400	8	
		401 - 800	6	
		801 - 1200	4	
8	Hydraulic Conductivity (m/day)	1201 - 1600	2	0.071
		1601 - 2000	1	
		0.06 - 0.30	1	
		0.31 - 0.43	2	

There are tables that show values of **RI** against **n** (Saaty, 1980). From such tables, **RI** = 0.41 for using 8 factors in this study. The consistency index rule of thumb is that a Consistency Ratio (**CR**) less than or equal to 0.1 indicates an acceptable reciprocal matrix, while a value over 0.1 indicates that the matrix should be revised (Saaty 1980). The weight for each parameter was computed using the AHP method described above and the results (Table 1) were assigned to their corresponding data layers. The various consistency checks were done using equations 1, 2, and 3, and a **CR** value of 0.09 was obtained to establish acceptable consistency for the pairwise value judgements and weight estimation for the subsidence factors.

## 2.4 Subsidence Vulnerability Index

The adopted DRASTIC method has a numerical ranking system that contains three major parts— weights, ranges and ratings. For this study, the main parameters were assigned weights computed from AHP method to reflect their relative influence on land subsidence. The significant variations or classes within each parameter or data layer were rated from 1 to 10 based on their relative effect on subsidence vulnerability (Table 1). The method employs a numerical subsidence index that is derived from the ratings and weights assigned to the main parameters. This index is computed by a linear combination of all the rated and weighted factors using functions like equation (4):

$$\text{Subsidence Index} = \sum_{i=1}^n w_i r_i \quad \dots 4$$

where;

**w** and **r** are the weight and rating of a given parameter, **i**, at a given cell within each data layer of the study area (and  $i = 1, 2, 3, \dots n$ ).

## 3 RESULTS AND DISCUSSION

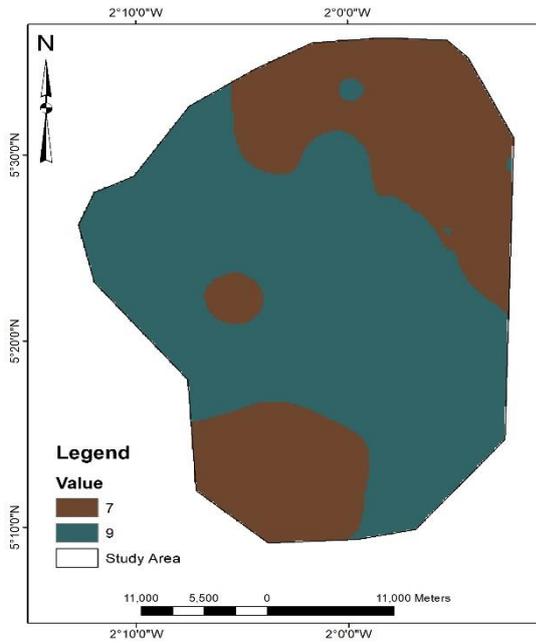
### 3.1 Depth to Groundwater

The groundwater or water table depth is the estimated vertical distance or height from the ground surface to the water table in unconfined aquifer and to the bottom of the confining layer in confined aquifer. The smaller the water table depth, the more vulnerable the land is to subsidence and vice versa. The depth to groundwater data was interpolated across the study area using the Inverse Distance Weighting (IDW) method. The raster output result was reclassified and rated as shown at Table 1 and Fig. 4.

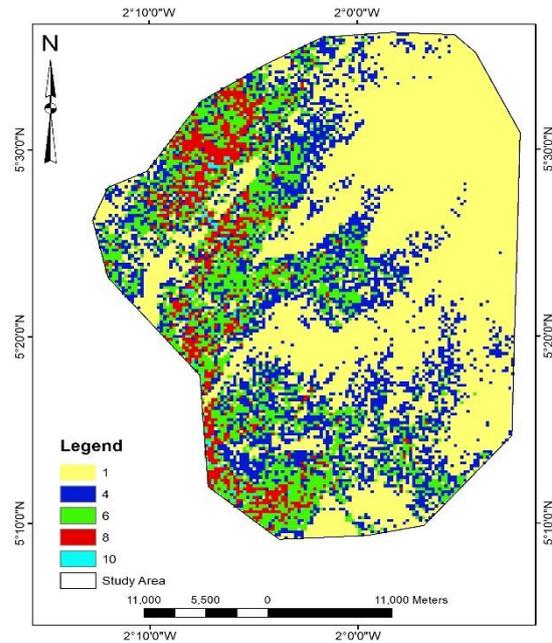
### 3.3 Elevation

Higher elevations generally have lower moisture while lower elevations generally have higher soil moisture. This is because water flows downhill due to gravity. Thus, chemical weathering is higher at lower elevations than at higher elevations. As a result, land subsidence is higher at

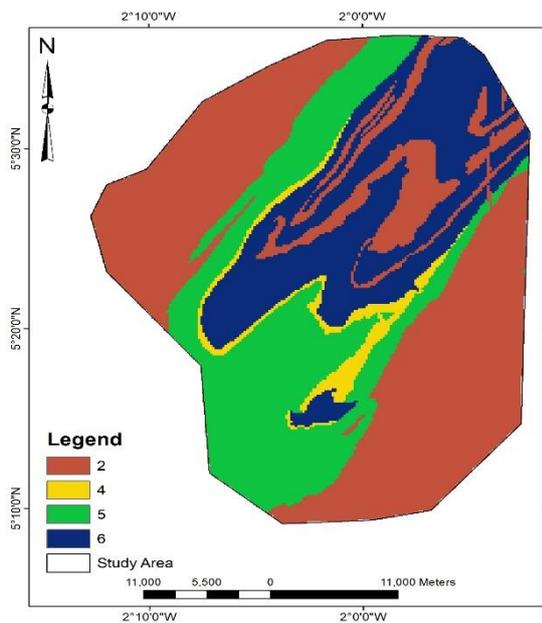
lower elevations and vice versa. Therefore, higher ratings were given to lower elevations than higher elevations. Fig. 5 is a map showing the elevation ratings map.



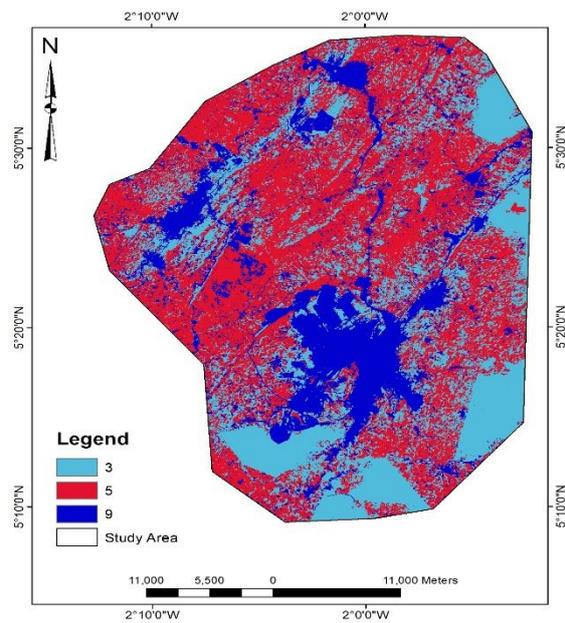
**Fig. 4 Groundwater Depth Ratings Map**



**Fig. 5 Elevation Ratings Map**



**Fig. 6 Map Showing the Lithology Ratings**



**Fig. 7 LULC Ratings Map**

### 3.2 Geology/Lithology

Lithology refers to the composition and type of rocks in the study area. Based on the geological description of the study area (Kesse 1985), the underlying rocks in the area include volcanic rocks, Phyllites, quartzites, sandstones and conglomerates. The harder the rock, the more resistant it is to chemical weathering and the less susceptible it is to land subsidence. The hardest rock type in the area are the volcanic rocks, followed by Quartzite/Conglomerate, Phyllite and Sandstone. Fig. 6 is a map showing the ratings of for the classes of rock types or lithology (Table 1).

### 3.3 Landuse/Landcover (LULC) Model

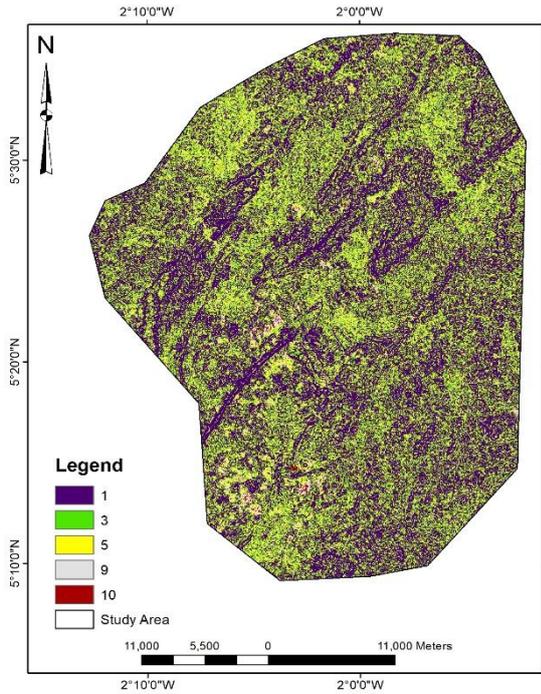
The LULC model for the study area is shown in Fig. 7. It is categorised into Forest, Sparse vegetation/farmland and built-up/mining sites. Rates of aquifer drawdown at built-up/mining site is high, as people draw water for wells on daily basis for domestic and mining purposes. Water is also drawn from wells for irrigation purposes. Drawdown is known to be one of the major factors for land subsidence. Hence the higher the drawdown, the higher the risk of land subsidence. Thus built-up/mining site category was assigned the highest rate followed by farmland and forest, as shown in Fig. 7.

### 3.4 Soil Media

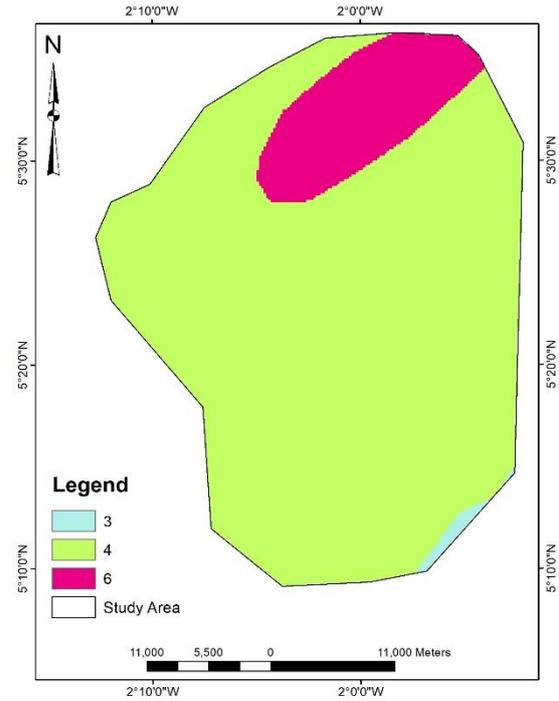
Soil media is the upper weathered zone of the earth, which averages a depth of six feet or less from the ground surface (Alwathaf and Mansouri, 2011). The predominant soil types in the area are laterite and silt. Laterites have larger grain sizes than silt, hence high draining capability than silt. The higher the draining capability, the lower the risk of land subsidence. In addition, cohesive soils such as silt are more susceptible to land subsidence since they shrink and swell depending on their moisture content. Consequently, the silt was assigned a rate of 5 whereas the laterite, a rate of 3 (Table 3). The vector layer of the soil map was first converted to a raster grid and reclassified by the rating factors (Table 3) to produce the map presented in Fig. 8.

### 3.5 Slope

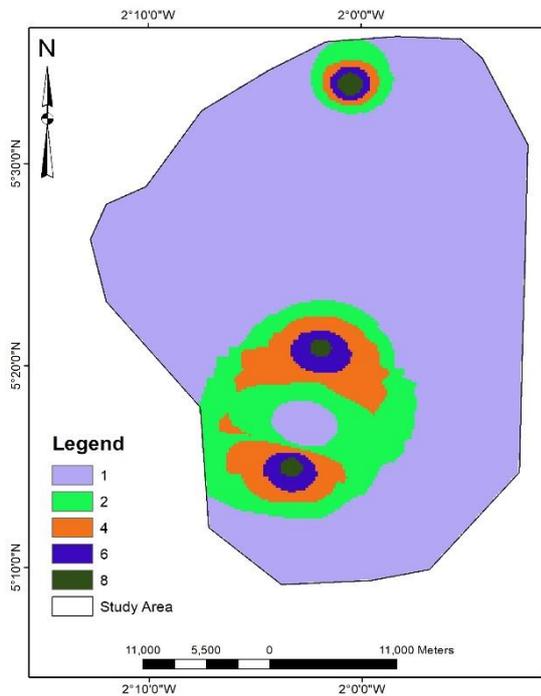
Slope refers to the rise or fall of the land surface. Where slopes are low, there is little runoff, and the potential for water to seep through the ground to cause chemical weathering is high. The higher the rate of chemical weathering, the higher the risk of land subsidence. Digital elevation model (DEM) was used to calculate slope percentages. The resulting slope map was reclassified according to Table 1, to generate the slope ratings map (Fig.9).



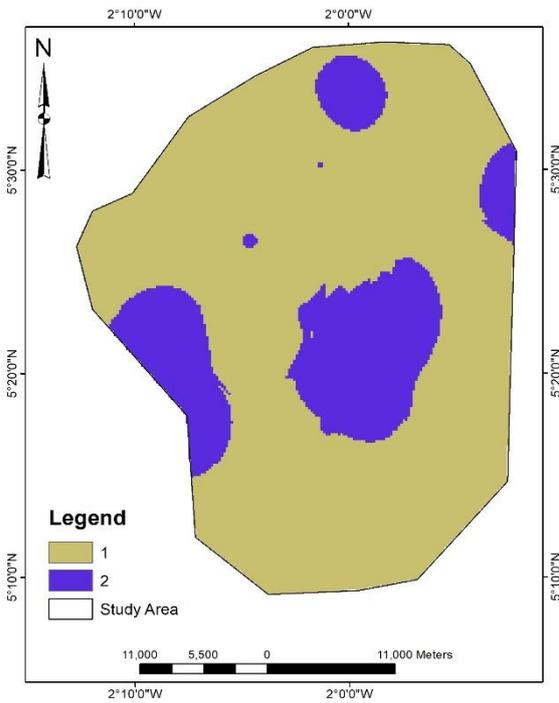
**Fig. 8 Soil Media Ratings**



**Fig. 9 Slope Ratings**



**Fig. 10 Proximity to Mining Sites Ratings Map**



**Fig. 11 Hydraulic Conductivity Ratings Map**

### 3.6 Proximity to Mining Sites

Areas closer to mining site are usually subject to ground vibrations due to the blasting activities of the mines. Land subsidence can occur during ground vibrations due to offset along fault lines. It also occurs because of settling and compaction of unconsolidated sediments from the vibration of the ground. Consequently, areas closer to the mines were given higher ratings and vice versa. The ratings map for the proximity to mining sites layer is shown in Fig. 10.

### 3.7 Hydraulic Conductivity

Hydraulic conductivity is a measure of how easy the water can flow through the soil or rock. The higher the hydraulic conductivity, the less susceptible the land is to subsidence. The hydraulic conductivity within the study area ranges between 0.06 to 0.5 m/day. The hydraulic conductivities of the shallow aquifers within the study area were reclassified and rated as shown at Table 1 and Fig. 11.

### 3.8 Land Subsidence Vulnerability Map

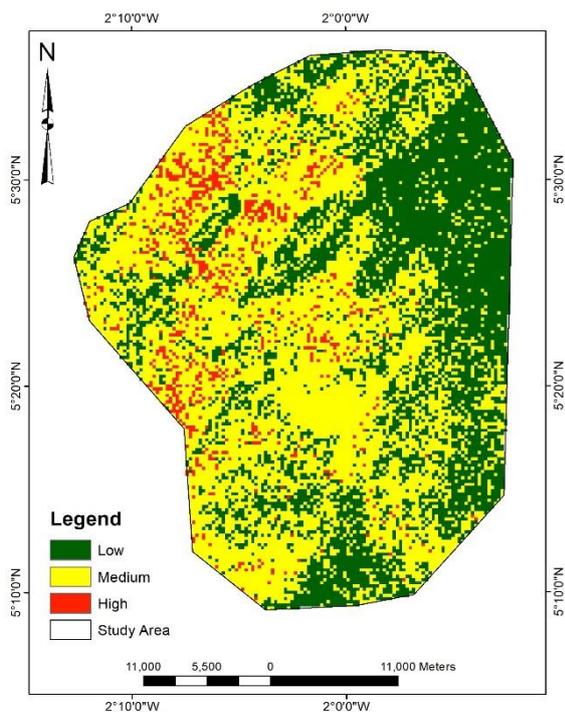
Land subsidence vulnerability index map was generated using the raster calculator in spatial analyst tool in ArcMap 10.3. Equation 4 was used to generate the subsidence vulnerability index (SVI) for each cell within each data layer and aggregated. The SVI range was from 3.6 to 7.4 and its related output index map was reclassified based on Table 2 to produce the final land subsidence vulnerability map (Fig. 12). The land subsidence susceptibility map shows three main classes within the current study area, namely; low, moderate and high. The low risk zones occur mainly at the north-eastern, southern and northern parts of the study area, and occupy about 14% of the study area. They mainly occur within the volcanic rocks with few cases occurring in the quartzites and sandstones and these areas have relatively higher elevations (125 – 332 m).

The moderate vulnerability zones generally cover the entire area with concentrations at the central, south-eastern and northern parts and occupies about 75 % of the study area. It thus appears to occur in all the rock types in the area, namely the conglomerate, phyllite, volcano and sandstone formations. The high vulnerability zones occur at the north-western, western and central parts of the study area, with few isolated cases at the southern and northern parts. It constitute about 11 % of the area. They are located predominantly within the quartzites, conglomerates and sandstones with relatively lower water tables (3.5 – 4.4 m) and ground elevations (4 – 92 m). Consequently, the zones are susceptible to chemical weathering activities, which increase the risk of land subsidence. Few of the high risk zones can however be found in the volcanic rock formation.

The land subsidence vulnerability maps such as shown at Fig 12 may be used as references or criteria to check the suitability of proposed land uses or the locations of development projects

in terms of subsidence risk or potentials. Appropriate decisions can thus be arrived at such as rejecting or disapproving the proposal or requesting more stringent mitigating measures against subsidence potentials and their impacts before allowing or approving the use of such sites.

The reliability of the methods and results presented in this paper depends on the quality of the data sets used. In the current work, some of the data sets used were generalised regional and district data (for example the geological and soil data) while some were site-specific ones (for example the water levels) but did not cover the entire study area and hence interpolations were applied. The results presented in this paper are therefore useful for the initial screening of proposed developments in which land subsidence prevention is a key factor to account for. Detailed site-specific land subsidence vulnerability investigation will still be necessary in the final site selection process before development begins.



**Table 2 Criteria for Vulnerability Classification**

Class	Vulnerability Potential
< 2.0	Very Low
2.0 – 4.0	Low
4.0 – 6.0	Moderate
6.0 – 8.0	High
> 8.0	Very High

**Fig. 12 Land Subsidence Vulnerability Map**

#### 4 CONCLUSIONS AND RECOMMENDATIONS

This study has demonstrated the combined use of the ‘DRASTIC’ ground vulnerability modelling technique, the analytical hierarchy process (AHP), Geographic Information System (GIS), GPS and Remote Sensing to collect, process, analyse and evaluate the relative and

combined influences of the risk factors involved and to map the susceptible areas of land subsidence in the Tarkwa-Prestea Mining Areas (TPMA) of Ghana. The resulting vulnerability map from this integrated approach indicates areas which must have high priority in terms of protection or monitoring against subsidence vulnerability. The computed land subsidence vulnerability index (SVI) values range from 3.5 to 7.5 leading to 3 main vulnerability classes for the area.

The high subsidence vulnerability zones constitute about 11% of the study area and occur mainly at the north-western, western and central parts of the study area. They are located predominantly within the quartzites, conglomerates and sandstones with relatively lower water table depths and ground elevations. These zones were observed to be susceptible to chemical weathering activities which increase the risk of subsidence. The percentage of high subsidence vulnerability observed emphasizes the need to pay more attention to the phenomenon in the study area. The high risk zones could also be linked to close proximity to mining activity zones. It is recommended that the method and results of the current study may be used as preliminary references or criteria to check the suitability of proposed land uses or developments in terms of land subsidence risk in the study area and that land subsidence vulnerability analysis should be integrated in existing land use and resource development planning and approval processes in TPMA and similar mining areas.

## REFERENCES

- Al-badi, A. M., Al-Shamma, A. M. and Aljabbari, H. M., (2014), “A GIS-Based DRASTIC Model for Assessing Intrinsic Groundwater Vulnerability in North-eastern Missan Governorate, Southern Iraq”, *Springer*, No. 7, pp. 89 - 101.
- Aller, L., Bennett, T., Lehr, J. H., Petty, R. H. and Hackett, G. (1987), “DRASTIC: A Standardized System for Evaluating Groundwater Pollution Potential Using Hydrogeologic Setting”, *USEPA Report 600/2-87/035*, Robert S. Kerr Environmental Research Laboratory, Ada, 252 pp.
- Al-shabeeb, A. R., (2016), “The Use of AHP within GIS in Selecting Potential Sites for Water Harvesting Sites in the Azraq Basin—Jordan”, *Journal of Geographic Information System*, Scientific Research Publishing, 8, pp. 73-88.
- Alwathaf, Y. and El Mansouri, B. (2011), “Assessment of Aquifer Vulnerability Based on GIS and ArcGIS Methods: A Case Study of the Sanaa Basin (Yemen)”, *Journal of Water Resource and Protection*, Vol. 3 (2011), pp. 845-855.
- Al-Zabet, T. (2002) “Evaluation of Aquifer Vulnerability to Contamination Potential Using the DRASTIC Method,” *Environmental Geology*, Vol. 43, No. 1-2, pp. 203-208.
- Anon. (2014), ‘2010 Population & Housing Census’, *District Analytical Report for the Tarkwa-Nsuaem Municipality*, Ghana Statistical Service (GSS), Ghana, pp. 1-67.

- Bui, D. T., Shahabi, H., Shirzadi, A., Chapi, A., Pradhan, B., Chen, W., Khosravi, K., Panahi, M., Ahmad, B. B., Saro, L. (2018), “Land Subsidence Susceptibility Mapping in South Korea Using Machine Learning Algorithms”, *Sensors*, Vol. 18 (2464), pp. 1-20
- Civita M. (1993), “Ground Water Vulnerability Maps: A Review”, *Proceedings., IX Symposium on Pesticide Chemistry, Mobility and Degradation of Xenobiotics*, Placenza, Lucca, Italy, October 11-13, pp. 587-631.
- Dehghani, M., Rastegarfar, M., Ashrafi, R. A., Ghazipour, N. and Khorramrooz, H. R. (2014), “Interferometric SAR and Geospatial Techniques Used for Subsidence Study in the Rafsanjan Plain”, *American Journal of Environmental Engineering* Vol. 4(2), pp. 32-40.
- Ewusi, A., Ahenkorah, I. and Kuma, J. S. Y. (2017), “Groundwater Vulnerability Assessment of the Tarkwa Mining Area Using SINTACS Approach and GIS”, *Ghana Mining Journal*, Vol. 17, No. 1, pp. 18 - 30.
- Ghorbanzadeh, O., Blaschke, T., Aryal, J., Gholaminia, K. (2020), “A new GIS-based technique using an adaptive neuro-fuzzy inference system for land subsidence susceptibility mapping”, *Journal of Spatial Science*, Vol. 65:3, pp. 401-418.
- Hirdes W. and Nunoo B. (1994), “The Proterozoic Paleo Placers at Tarkwa Gold Mine, Southwest Ghana”, *Geological Journal* , Vol. 1, pp. 22-24.
- Jaseela, C., Prabhakar, K., Sadasivan, P. and Harikumar, P. (2016), “Application of GIS and DRASTIC Modeling for Evaluation of Groundwater Vulnerability near a Solid Waste Disposal Site”, *International Journal of Geosciences*, Vol.7, pp. 558-571.
- Junner, N. R., Hirst, T. and Service, H. (1942), “The Tarkwa Goldfield”, *Gold Coast Geological Survey, Memoir*, No. 6, pp. 48-55.
- Kalani, E., Ka-zem-Zadeh, R. B. and Kamrani, E. (2017), “The Pathology of the Hindrance Factors Impeding the Application of Value Engineering in the Construction Industry in Iran and Ranking them by Use of Analytical Hierarchy Process”, *Journal of Human Resource and Sustainability Studies*, Vol. 5, pp. 57-67.
- Kesse, G. O. (1985), *The Mineral and Rock Resources of Ghana*, A. A. Balkema Publishers, Rotterdam, 610 pp.
- Kortatsi, B. K. (2002), “Hydrochemistry of Groundwater in the Mining Area of Tarkwa-Prestea, Ghana”, *Ph.D. Thesis*, University of Ghana, pp. 70-85.
- Kuma, J. S. (2007), “Hydrogeological Studies in the Tarkwa Gold Mining District, Ghana”, *Bulletin of Engineering Geology and the Environment* Vol. 66, pp. 89 - 99.
- Kim, K., Lee, S., Oh, H. (2006), “Assessment of Ground Subsidence Hazard Near an Abandoned Underground Coalmine Using GIS”, *Environ Geol* Vol. 50, pp. 1183–1191.
- Kwesi, E. A. A, Asamoah, K. N., Arthur, F. A. and Kwofie, J. A. (2020), “Mapping of Ground Water Vulnerability for Landfill Site Selection Assessment at the District Level – A Case Study at the Tarkwa Nsuaem Municipality of Ghana”, *Ghana Journal of Technology*, Vol. 4, No. 2, pp. 57 - 65.
- Rolland, A. and Rangarajan, R. (2013), “Runoff Estimation and Potential Recharge Site Delineation Using Analytic Hierarchy Process”, *Geocarto International*, 28, pp. 159-170.

- Rundquist, D. C, Rodekohr, D. A, Peters, A. J., Ehrman, L. D. and Murray, G. (1991), “State-wide Groundwater-vulnerability Assessment in Nebraska Using the DRASTIC/ GIS Model”. *Geocarto Int*, Vol. 2, pp. 51–58.
- Saaty, T. L. (2000), *Fundamentals of Decision Making and Priority Theory with the Analytic Hierarchy Process* (Analytic Hierarchy Process Series, Vol. 6), RWS Publications, Pittsburgh, pp. 10-35.
- Saaty, T. L. and Vargas, L.G. (2012), *Models, Methods, Concepts and Applications of the Analytic Hierarchy Process*, Springer Science & Business Media, New York, pp. 15-25.
- Saraf, A. and Choudhury, P. (1998) “Integrated Remote Sensing and GIS for Groundwater Exploration and Identification of Artificial Recharge Sites”, *International Journal of Remote Sensing*, 19, pp. 1825-1841.
- Sun, H., Grand staff, D., Shagam R. (1999), “Land Subsidence due to Groundwater Withdrawal: Potential Damage of Subsidence and Sea Level Rise in Southern New Jersey, USA”, *Environmental Geology* Vol. 37 (4), pp. 290-296
- Van Stempvoort, D., Ewert, D. and Wassenaar, L. (1993), “Aquifer Vulnerability Index: a GIS Compatible Method for Groundwater Vulnerability Mapping”, *Water Resources Journal*, Vol. 18 (1), pp. 25–37.

## BIOGRAPHICAL NOTES

### Authors

**E. A. A. Kwesi** is currently a Lecturer at the Geomatic Engineering Department of the University of Mines and Technology (UMaT), Tarkwa, Ghana and a member of GhIS, GhIG, FIG and GPL. His research and consultancy works cover Surveying and Mapping, Community Involvement and Multicriteria Decision Analysis and their applications in Sustainable Management of Land, Agriculture, Waste, and Communities in Mining Areas.

**K. N. Asamoah** currently working as a Graduate Teacher and holds MPhil. Degree in Geology at the University of Mines and Technology (UMaT), Tarkwa, Ghana. His research areas include geophysics, geostatistics, GIS and remote sensing and their applications in Mining, Waste Management and related fields.

**Kwame Tenadu Snr.** holds a Master of Arts Degree in Environmental Management and Policy and University Diploma in Geodetic Engineering. He is a member of FIG (currently Vice President), GhIS, the Commonwealth Association of Surveyors and Land Economists, and has had over 30 years experience in land administration.

**Dr Benjamin Edem Meteku** is currently a lecturer at UMaT and holds PhD degree in Petroleum Engineering

**Alexander O. MENSAH** is currently a Postgraduate Assistant at UMaT School of Petroleum Studies, Ghana.

**Gershon Piedu** is a Lecturer at the Institute of Distant and e-Learning, University of Education, Winneba, Ghana. He holds both Bachelors and Masters Degrees and Certificates in Sociology and Education, and General Drilling from UMaT and KNUST, Ghana; and Norwegian University of Science and Technology, Norway.

**CONTACTS: E. A. A. KWESI, University of Mines and Technology, Tarkwa, GHANA; Email: eaakwesi@umat.edu.gh**