



Hydrogeochemical Characterization and Quality Assessment of Groundwater Based on Water Quality Index in Imo state, South Eastern Nigeria

*¹OSISANYA, OW; ²AGHO, IF; ³SALEH, AS; ⁴THOMPSON, E.

¹Department of Physics, University of Benin, Benin City, Edo state, Nigeria

²Department of Microbiology, University of Benin, Benin City, Edo state, Nigeria

^{3,4}Department of Petroleum Engineering and Geosciences, Petroleum Training Institute, Effurun, Nigeria.

*Corresponding Author Email: wasiu.osisanya@uniben.edu; Tel: +2347036555765, +2348150914664

ABSTRACT: Water Quality Index (WQI), Principal Component Analysis (PCA), Correlation matrix, Metal Pollution Index (MPI), Contamination Factor (CF), Pollution Load Index (PLI), Geoaccumulation Index (Igeo), Health Risk Assessment, and Hydrogeochemical facies were used to analyze statistical indexes and hydrogeochemical facies in groundwater resources in Imo state, Nigeria. All across the study area, twenty (20) groundwater samples were collected in a systematic manner. The samples were examined in accordance with the American Public Health Association standard (APHA) method. Findings from the study revealed that WQI, is of poor quality and should only be used for irrigation. Weathering and redox reactions are important in groundwater geochemistry, according to PCA results. TDS and Cl, HCO₃ and Zn, Cl; Mg and Ca, Ca and Na were all found to have a positive correlation in the correlation matrix while PH and K, HCO₃ and Fe, Cl and SO₄ are found to have a negative correlation in the correlation matrix. The findings show that the items have a weak correlation and that there is no relationship between the two variables. Further MPI, CF, and PLI findings revealed that groundwater is pure, the main source of pollution is geological and anthropogenic processes, and there is no pollution in sampled groundwater. Hydrogeochemical trend revealed that groundwater is Na⁺+K⁺ > HCO₃⁻+CO₃ > Mg⁺ > SO₄ > Cl⁻ > Ca⁺. Based on the finding, pre-use treatment of water resources is strongly advised.

DOI: <https://dx.doi.org/10.4314/jasem.v26i12.13>

Open Access Policy: All articles published by **JASEM** are open access articles under **PKP** powered by **AJOL**. The articles are made immediately available worldwide after publication. No special permission is required to reuse all or part of the article published by **JASEM**, including plates, figures and tables.

Copyright Policy: © 2022 by the Authors. This article is an open access article distributed under the terms and conditions of the **Creative Commons Attribution 4.0 International (CC-BY- 4.0)** license. Any part of the article may be reused without permission provided that the original article is clearly cited.

Cite this paper as: OSISANYA, O. W; AGHO, I. F; SALEH, A. S; THOMPSON, E. (2022). Hydrogeochemical Characterization and Quality Assessment of Groundwater Based on Water Quality Index in Imo state, South Eastern Nigeria. *J. Appl. Sci. Environ. Manage.* 26 (12) 1989-2003

Keywords: Groundwater; geochemical properties; water quality index; Hydrogeochemical Characterization

Water for human consumption and agriculture is stored as groundwater, which is the world's largest liquid freshwater supply (Eyankware, 2018; Eyankware, et al., 2020a). Despite the fact that climate change has caused significant spatial-temporal precipitation variability and has affected reservoir storage volume, groundwater remains an important and reliable source of water; however, pollutants do occasionally reach the aquifer as a result of natural or human factors. Inadequate waste disposal sites, which have an impact on all natural resources, are one of the major threats to the quality of groundwater designated

for human use (Akakuru, *et al.*, 2021; Agidi, et al., 2022). In unregulated and poorly built municipal landfills, leachate migration from landfills to groundwater is seen as a severe environmental issue (Obasi, et al., 2015; Ezech, et al., 2015; Omo-Irabor, et al., 2019). Impact on the environment. In this way, it's critical to estimate the risk of pollution of groundwater. Landfills are avoided. The quality of groundwater is currently a major environmental concern. For a variety of reasons, groundwater resource studies have become more important across different local, state, and nations (Sundraiah *et al.*

*Corresponding Author Email: wasiu.osisanya@uniben.edu

2014; Subba Rao 2017; Faten et al. 2016). However, as the quality of the groundwater improves as a result of the influences of geogenic and non-geogenic sources, it becomes an inferior type that will not be acceptable for any future purposes (Subba Rao *et al.* 2021c). Geochemical processes govern groundwater quality, which is influenced by a number of factors. Mineral weathering is linked to lithology and rock weathering (*ref*). Assessment of ground water quality using chemical approach is a very vital tool in classification of groundwater. The determining factors of groundwater quality area dependent on the rate of chemical weathering of the different geologic formation.geology, recharge water quality and rock-water interaction. (Domenico 1972, Satya et al., 2015). More so, sea water invasion which results from intensive absorption of freshwater in coastal regions is another determining factor of groundwater quality. (Sophiya and Syed 2013). Due to a lack of surface water sources, the residents of the study area rely primarily on groundwater resources for drinking and irrigation. The current study region is experiencing deterioration in groundwater quality in some areas due to a lack of proper awareness on the planning of household waste disposal, septic tank leakages, animal wastes, uncontrolled use of chemical composts, leakage from oil spillage, and so on. As a result, the groundwater quality may not be suitable for drinking or irrigation. However, no research has been conducted in the study region so far, hence, the objective of this study is to evaluate the hydrogeochemical characteristics and groundwater quality assessment based on water quality index in Imo state, South Eastern Nigeria.

MATERIALS AND METHODS

Description of the study area: The study area is 42.74 square kilometers in size and is located between latitudes 07° 10'N and 07° 18'N, as well as longitudes 5° 09'E and 05° 16'E in Imo state. The research is being conducted in Nigeria's rain forest zone, which has a wet season from April to October and a dry season from November to March. From February to March, the town experiences the hottest weather, as well as a dry wind (harmattan) from December to early February. The region receives moderate annual rainfall of approximately 1,300 to 1,700 mm, with high humidity during the wet season and low humidity

during the dry season, and year-round temperatures ranging from 22°C to 31°C (Iloeje, 1981).

Geology: The research area lies in the eastern portion of the Niger Delta Sedimentary Basin, as shown in Fig. 1 by the Niger Delta lithographic units. The units are: Akata Formation, Agbada Formation, and Benin Formation, in order of decreasing age (Opara et al. 2021). The research area, which comprises alluvium, Meander belts, fresh water marshes, and deltaic plain sands, is depicted in Fig. 1. All boreholes in the research region plunge into the younger Benin Formation, and the tertiary residue has a thickness of about 10,000 meters. The Miocene to late development is made up of sand, which is often medium to coarse grained, and stone, which is respectably organized with local focal points of inadequately cemented sand and muds. The Niger Delta Basin, also known as the Niger Delta area, is a right-hand basin in the Niger Delta and the Gulf of Guinea, near to Nigeria's western bank, with thought or demonstrated access to Cameroon, Equatorial Guinea, and Sao Tome and Principe. The bowl is perplexing, and it's worth a lot of money because it houses a quite useful oil framework. With 75,000 km², 300,000 km² of all out areas, and 500,000 km² of silt fill, the Niger Delta is one of the largest subareas (Urom et al. 2021). The Niger Delta Basin was formed during the break of the South American plate and the entrance of the African plate of the south Atlantic during the blasted fracture junction. This basin formed in the late Jurassic and laste.

Method: As shown in Fig. 1, twenty (20) water samples were collected at random from hand-dug wells/motorized boreholes. The sampling was done during the dry season, when the water level was lower and the concentrations of cations, anions, and trace elements were more stable. Before collecting water samples from the site, the bottles were washed with clean water, cleaned with cleaning reagents, and thoroughly rinsed with distilled, de-ionized water. pH, electrical conductivity, total dissolved solids, total hardness, magnesium, calcium, chloride, nitrate, sulphate, potassium, sodium, bicarbonate, iron, cobalt, lead, zinc, manganese, mercury, cadmium, and silver were all measured using the APHA (2012) standard (Table 1).

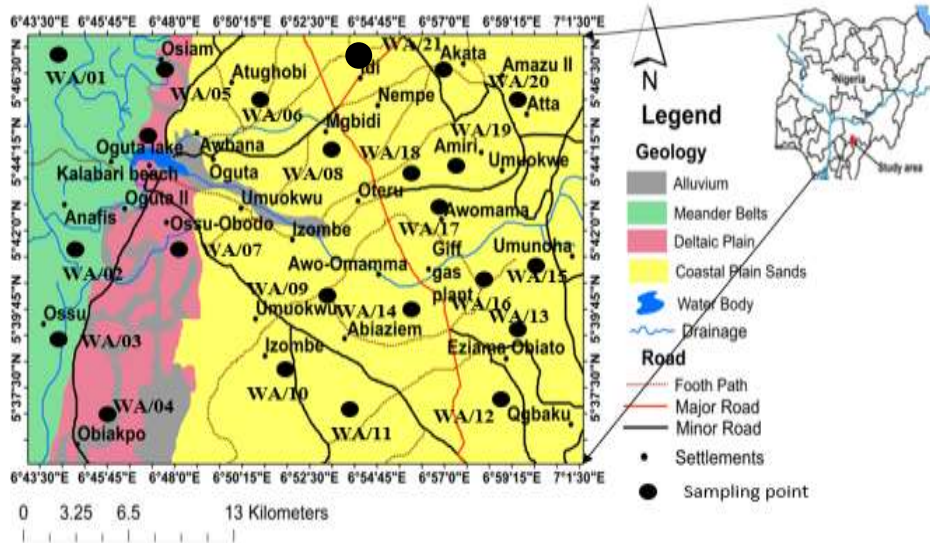


Fig.1. The study area's geology map, showing groundwater sample points

Table 1. Methods used to analyze physicochemical parameters

S/No	Parameters	Analytical Method
1	pH	pH meter Hach sensION + PH1 portable pH meter and Hach sensION + 5050 T Portable Combination pH Electrode
2	Total dissolved solids (TDS)	TDS meters (model HQ14D53000000, USA).
3	Magnesium (Mg ²⁺)	EDTA titrimetric method
4	Calcium (Ca ²⁺)	Titrimetric method
5	Chloride (Cl ⁻)	Titrimetric method
6	Sulphate (SO ₄ ²⁻)	Turbidimetric method using a UV-Vis spectrometer
7	Potassium (K ⁺)	Jenway clinical flame photometer (PFP7 model)
8	Sodium (Na ⁺)	Jenway clinical flame photometer (PFP7 model)
9	Bicarbonate (HCO ₃ ⁻)	Titrimetric method
10	Iron (Fe)	(Flame Atomic Absorption Spectrophotometer) FAAS
11	Manganese (Mn)	FAAS
12	Arsenic (As)	FAAS
13	Zinc (Zn)	FAAS
14	Copper (Cu)	FAAS

*Water quality index

The Water Quality Index (WQI) was used to assess the quality of surface and groundwater for human consumption in the study area. Brown et al. (1970), following Horton (1965), developed this index to provide a comprehensive overview of the state of surface and groundwater quality for residential use. WQI, on the other hand, has a few flaws. WQI, for example, may not provide sufficient information on the true state of water quality. Furthermore, many users of water quality data may not find the index adequate. Finally, because the assignment of weights to parameters is left to the user's discretion (Ukah et al., 2020), there may be bias in the calculation. The estimation of WQI in equation 1 was done in two steps:

$$WQI = \frac{\sum_{i=1}^n W_i \times Q_i}{W_i} \quad (1)$$

Where W_i denotes the parameter's relative weight and Q_i denotes the parameter's quality index.

Phase I: Each parameter tested in the water samples was assigned a relative weight (W_i) based on its value in total water quality. The unit weights of each of the ten parameters studied (pH, Cl⁻, SO₄²⁻, HCO₃, NO₃, Fe²⁺, Zn²⁺, Ni²⁺, Cr³⁺, and Pb²⁺) were calculated using the following formula (Eq. 2):

$$W_i = \frac{k}{S_i} \tag{2}$$

With S_i representing the acceptable standard value of the water quality parameter and k representing the proportionality constant calculated using the formula (Eq. 3):

$$k = \frac{1}{\left(\frac{1}{S_i} = 1, 2, \dots \dots \dots i\right)} \tag{3}$$

Phase 2: Using Eq. (4), we determined the quality index for each parameter:

$$Q_i = \frac{(V_a - V_i)}{(V_s - V_i)} \times 100 \tag{4}$$

Where V_a represents the measured value of parameter I , V_i represents the ideal value of parameter I (0 for all parameters except pH, which is 7.0), and V_s represents the WHO-recommended standard for parameter i . When the quality rating is zero, there are no pollutants. A quality rating of 0 Q_i 100 indicates that the pollutants have exceeded the allowable limits (McClelland, 1974). Based on the absolute value of the index discovered through computations, the estimated WQI values are divided into four categories: "excellent," "good," "poor," "very poor," and "unfit for drinking."

Principal Component Analysis (PCA): Component loading in PCA entailed dividing a large dataset with many variables into a smaller number of linear combinations in the component that accounted for an appropriate fraction of the total data variance and easily associating the variables to the sources or processes via Equation (5).

$$\sigma^2 = \frac{1}{N} \sum_{i=1}^N (X_{N-} - N)^2 \tag{5}$$

When X_i is the statistical mean, I is the range of variables, and N is the number of outcomes.

Metal Pollution Index (MPI): Caeiro et al., (2005) used Equation 6 to evaluate MPI, which calculated the aggregate impact of individual heavy metals on water quality (Horton, 1965; Mohan et al., 1996).

$$MPI = \sqrt[n]{M_1 \times M_2 \times M_3 \dots \dots \dots M_n} \tag{6}$$

Where M_n = concentration of the metal.

Pollution Load Index (PLI): The PLI was calculated using Hakanson's (1980); Eyankware and Akakuru, (2022) formula.

$$PLI = \sqrt[n]{CF_1 \times CF_2 \times CF_3 \times \dots \times CF_n} \tag{7}$$

Where CF represents the contamination factor and n represents the element number.

Geo-accumulation index (Igeo): Igeo could be used to assess the presence of toxic materials (Igeo). Eyankware and Akakuru compute it as follows (2022)

$$I_{geo} = \log_2 \frac{C_n}{1.5 B_n} \tag{8}$$

Where C_n represents the toxic metal n 's content and B_n represents the toxic metal n 's background, and 1.5 represents the factor of possibility.

Health risk assessment: The criteria of the US Environmental Protection Agency were used to determine the non-carcinogenic risk associated with ingesting contaminated river water (USEPA, 1989; see Table 2). The risk was calculated in equation 9 for both children and adults.

$$CDI = \frac{C_w \times IRW \times EF \times ED}{BW \times AT} \tag{9}$$

where CDI stands for Chronic Daily Intake (mg/kg/day), C_w stands for contaminant concentration in water (mg/L), IRW stands for ingestion rate (for adults, IRW is 2 L/day, while for children, IRW is 1 L/day), EF stands for Exposure Frequency (365 days/year), and ED stands for exposure duration (for adults and children, ED is 70 and 6 hours, respectively) (Bortey-Sam et al., 2015; Duggal et al., 2017; Barzegar et al., 2018).

The hazard quotient (HQ), as defined by Li et al. (2016) and Zhang et al. (2018), was used to assess the non-carcinogenic risk posed by the following elements in equation 10:

$$HQ = \frac{CDI}{RfD} \tag{10}$$

RfD denotes the reference dose (mg/kg/day) for a specific element. According to Duggal et al. (2017) and Barzegar et al. (2017), the RfD for various elements is approximately 0.7 (Fe), 0.3 (Zn), 0.0001 (Hg), 0.03 (Co), and 0.0035 (Co) (2018). (Pb). Finally, using equation 6, the total HQ values for the elements are calculated to determine the hazard index of the water samples (HI).

$$HI = \sum HQ \tag{11}$$

Values greater than one indicate that a specific element's non-carcinogenic risk exceeds the limit of acceptance (HI = 1), while values less than one indicate that the non-carcinogenic risk is within acceptable bounds (USEPA 1989; Su et al., 2017). According to Bortey-Sam et al. (2015) and Barzegar et al. (2015), Table 3 depicts the USEPA (1989) classification of non-carcinogenic risk (2015).

Table 2: In 1989, the United States Environmental Protection Agency (USEPA) classified non-carcinogenic risk.

Risk Level	Hazard Index (HI)	Chronic risk
1	<0.1	Negligible
2	0.1-1	Low
3	1-4	Medium
4	>4	High

RESULTS AND DISCUSSION

Water quality index: The WQI provides a thorough picture of the quality of surface and ground water for most residential purposes. The Water Quality Index (WQI) is a metric that evaluates the impact of a variety of water quality elements when used together (Eyankware, et al., 2021a). The appropriateness of surface and/or groundwater for human consumption determines the WQI. WQI is a ranking tool that provides a more accurate and convenient way to categorize water. The suitability of groundwater is critical because it influences its availability for a variety of uses such as household, drinking water, irrigation, industry, and so on (Eyankware et al., 2020; Subba Rao et al., 2012; Gopinath et al., 2019). The WQI results for all samples in the study area (Table 4) show that 70% of all samples are of very low quality and should only be used for irrigation. Similarly, 30% of the samples had values above 300. This indicates that the sample is not suitable for drinking and should be processed before it can be used for any purpose. This cross-section demonstrated the country's ability

to estimate the water quality of drinking water using WQI, which provides a statement on the combined effect of material boundaries on water. Poor water quality in the area may be due to human activity, which constantly exposes available water sources to unwanted pollution. This result is incomparable to studies conducted in India (Brindha et al., 2020; Gopinath et al., 2019); and Nigeria (Akakuru et al., 2021; Eyankware et al., 2020; Agidi et al., 2022).

Principal Component Analysis (PCA): PCA is a pattern recognition technique that attempts to explain the variation of multiple connected variables (Akakuru et al., 2021a&b). It demonstrates the relationship between variables, which simplifies the dataset. PCA derives eigenvalues and eigenvectors from the original data's covariance matrix. The uncorrelated (orthogonal) variables that result from multiplying the original correlated variables by the eigenvectors are known as principal components (PCs) (loadings). The PCs' eigenvalues are used to calculate the variance associated with them, the loadings are used to assess the participation of the original variables in the PCs, and the converted observations are referred to as scores (Helena et al., 2000; Wunderlin et al. 2001; Singh et al. 2004). The results of the PCA in Table 5, it shows that there is loading between elements. For PC1, 50% of the elements had loadings, for PC2, 35.7% of the elements had loading, for PC3, 28.6% of the element had loading while for PC4, 21.4% of elements had loadings. As indicated each part is more prominent, as evidenced by Kaiser's model (Akakuru et al., 2021b), also known as the eigenvalue 1 measure, which is one of the most common rules for describing parts set in the PCA. Must be simply maintained with a unique eigenvalue. From 1.00 decoding. This is because each of the above factors provides a unit of variability in the overall diversity of information gathered. Therefore, parts with eigenvalues greater than 1.00 are expected to explain a more important measure of variance than the variables provide. In this sense, such characteristic parts are the cause of many changes and should be retained, while the parts with an intrinsic value of less than 1.00 are the cause of diversity more than one variable brings.

Nonetheless, the basic goal of PCA is to reduce the set of perceived elements to a more conservative number of parts without jeopardizing the true translation of the referenced information.

Table 3 shows the study's findings, which include physicochemical and heavy metal levels in groundwater in the study area

Sample code	TDS (mg/L)	pH	Mg ²⁺ (mg/L)	Ca ²⁺ (mg/L)	HCO ₃ ⁻ (mg/L)	Na+ (mg/L)	Cl- (mg/L)	SO ₄ ²⁻ (mg/L)	K (mg/L)	Fe (mg/L)	Mn (mg/L)	As (mg/L)	Zn (mg/L)	Cu (mg/L)
WA/01	60.2	6.6	19.3	5.8	45.3	57.8	31	40.1	56.3	0.003	0.0001	0.0001	0.005	0.0012
WA/02	98.8	7.1	26	7.9	13.2	79	20.6	24.9	10.6	0.001	0.0013	0.0001	0.002	0.0031
WA/03	183.9	7	17.3	10.2	11.7	69.6	15.9	32.5	15.3	0.001	0.0001	0.0001	0.0004	0.02
WA/04	493	6.9	20.9	8.7	56.3	297.4	19.6	69.3	38.5	0.002	0.0001	0.0011	0.003	0.002
WA/05	686.1	6.8	8.4	10.2	26.1	167.8	38.2	42.7	60.3	0.001		0.0001	0.0001	0.0001
WA/06	811.2	7	7.7	4.3	67.4	68.7	39.6	68.7	21.1	0.002	0.0021	0.0001	0.0002	0.0002
WA/07	208.2	7.1	10.2	6.5	36.2	26.9	10	99.1	56.1	0.001	0.00011	0.0011	0.0006	0.0003
WA/08	503.9	6.8	13.6	8.6	585.3	59.3	50	58.3	35.2	0.0001	0.0001	0.0001	0.0071	0.0021
WA/09	1072.1	6.9	15.2	5.3	103.2	20.2	68.5	7.3	16.2	0.0011	0.0001	0.0011	0.00011	0.0013
WA/10	293.1	7.1	6.1	3.7	56.4	19.3	36.9	13.2	59.3	0.0013	0.00014	0.0001	0.0002	0.008
WA/11	503.2	6.9	20.1	6.6	85.2	77.1	28.7	36.8	38.4	0.0011	0.0001	0.00014	0.0001	0.0003
WA/12	755.3	7	12.2	8.1	45.7	92.4	57.8	50.9	13.2	0.0001	0.0012	0.00012	0.0006	0.00001
WA/13	1002.1	6.9	10.1	7.6	67.4	34.2	40.5	16.7	22.6	0.0011	0.00015	0.00016	0.0003	0.004
WA/14	507.6	7	11.7	5.4	143.2	64.1	18.9	48.9	47.5	0.001	0.0011	0.00015	0.0001	0.0002
WA/15	464.1	6.8	15.8	6.6	243.3	11.4	47.1	49.3	68.2	0.00011	0.0014	0.00014	0.0002	0.0001
WA/16	897.4	6.5	19.2	7.5	56.4	9.3	35.9	68.5	59.2	0.0001	0.002	0.0002	0.0004	0.0005
WA/17	749.1	6.8	14.8	5.3	532.2	37.3	59.1	6.87	48.2	0.0001	0.0001	0.0001	0.0005	0.0001
WA/18	583.4	7	13.1	3.5	47.6	14.5	33.7	30.1	18.2	0.0021	0.0011	0.0011	0.00001	0.0006
WA/20	393.9	7.1	16.2	2.4	89.76	17.7	13.9	69.2	20.3	0.0011	0.0002	0.0001	0.0002	0.0021
WA/21	307.6	6.9	18.1	6.6	103.5	27.9	5.3	75.4	16.5	0.0001	0.0005	0.00013	0.0004	0.0004
Min	60.2	6.5	6.1	2.4	11.7	9.3	5.3	6.87	10.6	0.0001	0.0001	0.0001	0.00001	0.00001
Max.	1072.1	7.1	26	10.2	585.3	297.4	68.5	99.1	68.2	0.003	0.0021	0.0011	0.0071	0.02
Aver.	528.71	6.91	14.8	6.54	120.768	62.595	33.56	45.4385	36.06	0.00102	0.00063	0.00031	0.00107	0.002331
Stand Dev.	291.86	0.16189	5.03472	2.13280	158.582	66.9161	17.3161	25.0774	19.3973	0.0008	0.00069	0.00040	0.00187	0.004577

As a result, retaining parts that are less volatile than those caused by individual factors overcomes the points of PCA. As a result, parts with eigenvalues less than 1.00 are considered defective and are not retained (Subba Rao, 2007). This result may provide insights due to geological processes such as weathering and redox reactions (Akakuru et al., 2021b; Egbueri, 2019; Yahaya et al., 2021; Eyankware and Akakuru, 2022).

Correlation matrix: Correlation matrices are a useful tool for determining the relationship between two variables. Normally, the correlation coefficient ranges between -1 and +1. The relationship has a negative slope or anti-correlation if

the r-value is close to -1. If r is close to +1, the relationship is said to have a positive slope or to be correlated. The points are said to be uncorrelated if the value is equal to zero (Srivastava et al., 2014). TDS and Cl, HCO₃ and Zn, Cl; Mg and Ca, Ca and Na were all found to have a positive correlation in the correlation matrix while PH and K, HCO₃ and Fe, Cl and SO₄ are found to have a negative correlation in the correlation matrix (Table 7). The results show that there is a weak correlation between the items and that there is no relationship between the two variables. That is, if one variable moves in one direction, the other variable moves in a completely unrelated direction. This also suggests that heavy metals in groundwater are primarily sourced from anthropogenic sources (Eyankware et al. 2021; Agidi et al. 2022; Akakuru et al. 2022).

Table 4: Results of the Water quality index

Sample code	qi*wi												WQI
	Mg	Ca	HCO ₃	Na	Cl	SO ₄ ²⁻	K	Fe	Mn	As	Zn	Cu	
WA/01	0.772	0.0145	0.11325	0.1445	0.0496	4.45511	0.14075	333.333	0.0625	100	0.02	0.03	439.1352
WA/02	1.04	0.01975	0.033	0.1975	0.03296	2.76639	0.0265	111.111	0.8125	100	0.008	0.0775	216.1251
WA/03	0.692	0.0255	0.02925	0.174	0.02544	3.61075	0.03825	111.111	0.0625	100	0.0016	0.5	216.2703
WA/04	0.836	0.02175	0.14075	0.7435	0.03136	7.69923	0.09625	222.222	0.0625	1100	0.012	0.05	1331.915
WA/05	0.336	0.0255	0.06525	0.4195	0.06112	4.74397	0.15075	111.111	0	100	0.0004	0.0025	216.916
WA/06	0.308	0.01075	0.1685	0.17175	0.06336	7.63257	0.05275	222.222	1.3125	100	0.0008	0.005	331.948
WA/07	0.408	0.01625	0.0905	0.06725	0.016	11.01001	0.14025	111.111	0.06875	1100	0.0024	0.0075	1222.938
WA/08	0.544	0.0215	1.46325	0.14825	0.08	6.47713	0.088	11.1111	0.0625	100	0.0284	0.0525	120.0766
WA/09	0.608	0.01325	0.258	0.0505	0.1096	0.81103	0.0405	122.2221	0.0625	1100	0.00044	0.0325	1224.208
WA/10	0.244	0.00925	0.141	0.04825	0.05904	1.46652	0.14825	144.4443	0.0875	100	0.0008	0.2	246.8489
WA/11	0.804	0.0165	0.213	0.19275	0.04592	4.08848	0.096	122.2221	0.0625	140	0.0004	0.0075	267.7492
WA/12	0.488	0.02025	0.11425	0.231	0.09248	5.65499	0.033	11.1111	0.75	120	0.0024	0.00025	138.4977
WA/13	0.404	0.019	0.1685	0.0855	0.0648	1.85537	0.0565	122.2221	0.09375	160	0.0012	0.1	285.0707
WA/14	0.468	0.0135	0.358	0.16025	0.03024	5.43279	0.11875	111.111	0.6875	150	0.0004	0.005	268.3854
WA/15	0.632	0.0165	0.60825	0.0285	0.07536	5.47723	0.1705	12.22221	0.875	140	0.0008	0.0025	160.1089
WA/16	0.768	0.01875	0.141	0.02325	0.05744	7.61035	0.148	11.1111	1.25	200	0.0016	0.0125	221.142
WA/17	0.592	0.01325	1.3305	0.09325	0.09456	0.763257	0.1205	11.1111	0.0625	100	0.002	0.0025	114.1854
WA/18	0.524	0.00875	0.119	0.03625	0.05392	3.34411	0.0455	233.3331	0.6875	1100	0.00004	0.015	1338.167
WA/20	0.648	0.006	0.2244	0.04425	0.02224	7.68812	0.05075	122.2221	0.125	100	0.0008	0.0525	231.0842
WA/21	0.724	0.0165	0.25875	0.06975	0.00848	8.37694	0.04125	11.1111	0.3125	130	0.0016	0.01	150.9309

Table 5: PCA of the samples

	Communalities	Components			
		1	2	3	4
TDS	0.679756	-0.72276	-0.16845	-0.04099	0.356818
pH	0.639903	0.375018	-0.64396	-0.28462	-0.05986
Mg	0.425936	0.40761	0.507778	0.038106	-0.02234
Ca	0.578907	0.277041	0.64836	-0.27265	-0.08615
HCO ₃	0.650051	-0.56113	0.547631	-0.12264	-0.01548
Na	0.617388	0.513712	0.340714	0.051171	0.484544
Cl	0.876893	-0.80877	0.125519	-0.26687	0.368516
SO ₄	0.694021	0.331156	0.079699	0.724036	-0.2319
K	0.353713	-0.22513	0.287997	0.454961	-0.11444
Fe	0.593849	0.466475	-0.25476	0.132874	0.541933
Mn	0.415313	-0.26565	-0.19923	0.463304	-0.30067
As	0.642838	0.194242	-0.23927	0.213644	0.708672
Zn	0.677581	0.159289	0.771308	0.039468	0.236081
Cu	0.737556	0.421234	-0.01211	-0.70712	-0.24486
Eigen Value		0.562332	1.791268	0.422674	1.620868
Variance (%)		21.57902	16.69256	13.42568	10.69897
Cumm. Variance (%)		21.58	38.27	51.7	62.4

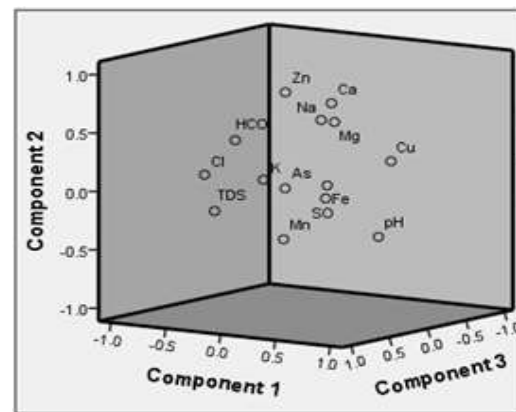


Fig 2: Component plot showing the loadings of elements

Table 6: Correlation Matrix

	TDS	pH	Mg	Ca	HCO ₃	Na	Cl	SO	K	Fe	Mn	As	Zn	Cu
TDS	1													
pH	-0.252	1												
Mg	-0.323	-0.311	1											
Ca	-0.089	-0.244	0.404	1										
HCO	0.173	-0.292	-0.1	0.044	1									
Na	-0.099	0.056	0.331	0.442	-0.113	1								
Cl	0.674	-0.313	-0.271	-0.048	0.459	-0.177	1							
SO ₄	-0.234	-0.002	0.01	0.092	-0.135	0.184	-0.54	1						
K	-0.155	-0.451	-0.142	-0.103	0.238	-0.109	0.04	0.142	1					
Fe	-0.242	0.092	-0.011	-0.311	-0.478	0.297	-0.222	-0.074	-0.036	1				
Mn	0.252	-0.131	-0.021	-0.076	-0.197	-0.144	0.09	0.234	-0.01	-0.127	1			
As	0.153	0.16	-0.026	-0.08	-0.225	0.262	-0.011	0.131	-0.052	0.32	-0.188	1		
Zn	-0.309	-0.362	0.255	0.356	0.434	0.328	0.068	0.129	0.116	0.162	-0.274	-0.076	1	
Cu	-0.32	0.237	0	0.375	-0.212	0.04	-0.225	-0.279	-0.207	0.066	-0.304	-0.177	-0.045	1

Metal Pollution Index (MPI): MPI has proven to be a useful technique in the classification of groundwater. The MPI of very pure water is 0.3 or lower, whereas the MPI of pure water is 0.3 to 1.0. Those with an MPI of 1.0 to 2.0 are considered minimally affected in Class III, while those with an MPI of 2.0 or higher are considered severely affected in Class IV. Water samples with MPIs ranging from 4.0 to 6.0 are classified as severely affected in Class V, and those with MPIs greater than 6.0 are classified as gravely affected in Class VI. The results in Table 7 show that the entire sample is less than 0.3, implying that the samples are very pure.

Contamination Factor (CF): In groundwater studies, CF was used to calculate the concentration ratio of heavy metals to background levels. The pollution factor values are explained using the following criteria: $CF < 1$, low pollution; $1 \leq CF \leq 3$, unhealthy; $3 \leq CF \leq 6$, severe contamination; $CF > 6$, Very high load. (Bhutian et al., 2017; Akakuru et al., 2021). The CF of this study shows that the concentrations of all parameters are well below 1 (unhealthy), as shown in Table 7. This result indicates that geological processes are the primary source of pollution. This also implies that anthropogenic activities in the area have had little effect on the available water resources. This result is comparable to that of Bhutian et al. (2017) for India and Nigeria (Yayaha et al., 2021).

Pollution Load Index (PLI): PLI is an efficient method for determining the toxicity of heavy metals in representative samples (Akakuru, et al., 2021b; Yang et al., 2011). The most common PLI categories are unpolluted ($PLI < 1$), unhealthy ($1 < PLI < 2$), severely polluted ($2 < PLI < 3$), and very heavily polluted ($3 < PLI < 4$).

According to this, the groundwater concentration value in the survey area was less than 1 (Table 7). That means there is no pollution, which contradicts the study conducted in India (Gopinath et al., 2019; Bhutian et al., 2017), but it is consistent with the study conducted in Nigeria by Yahyaya et al. (2021). This result is also consistent with the CF result in this study.

Geo-accumulation index (Igeo): Muller (1979) developed the geoaccumulation index (Igeo) to examine the amount of pollution of vital concentrations in waste, water, biomass, and soil, and it has been widely used in researching their contamination status all over the world (Eyankware and Akakuru, 2022).

Igeo ≤ 0 (essentially unpolluted), $0 < Igeo \leq 1$ (unpolluted), $1 < Igeo \leq 2$ (respectably contaminated), $2 < Igeo \leq 3$ (modestly to emphatically dirty), $3 < Igeo \leq 4$ (firmly contaminated), $4 < Igeo \leq 5$ (unequivocally to incredibly contaminated), and $Igeo \geq 5$ (very contaminated) are the characterizations of (Igeo) and their separate translations (Urom et al. 2021; Agidi et al. 2022; Usman et al. 2022). From the results in Table 7, it shows that 100% of the samples are unpolluted. This result is in conformity with the earlier results of CF, and PLI in this study which showed that there is relatively no direct effect of the anthropogenic activities in the area on the groundwater resources. This study is in line with the study done in Nigeria (Agidi et al, 2022; Eyankware and Akakuru 2022)

Table 7: CF, PLI, Igeo and MPI in the study area

Sample code	CONTAMINATION FACTOR					PLI	Igeo	MPI
	Fe	Mn	As	Zn	Cu			
WA/01	0.075	1.18E-06	0.00002	3.57E-05	3.33E-05	4.58E-11	0.01507	1.34E-08
WA/02	0.025	1.53E-05	0.00002	1.43E-05	8.61E-05	9.70E-11	0.005044	2.84E-08
WA/03	0.025	1.18E-06	0.00002	2.86E-06	0.000556	3.06E-11	0.005133	8.94E-09
WA/04	0.05	1.18E-06	0.00022	2.14E-05	5.56E-05	1.24E-10	0.010094	3.63E-08
WA/05	0.025	0	0.00002	7.14E-07	2.78E-06	0	0.005022	0
WA/06	0.05	2.47E-05	0.00002	1.43E-06	5.56E-06	1.40E-11	0.010045	4.10E-09
WA/07	0.025	1.29E-06	0.00022	4.29E-06	8.33E-06	1.59E-11	0.005064	4.67E-09
WA/08	0.0025	1.18E-06	0.00002	5.07E-05	5.83E-05	1.32E-11	0.000528	3.86E-09
WA/09	0.0275	1.18E-06	0.00022	7.86E-07	3.61E-05	1.42E-11	0.005571	4.16E-09
WA/10	0.0325	1.65E-06	0.00002	1.43E-06	0.000222	1.84E-11	0.006572	5.40E-09
WA/11	0.0275	1.18E-06	0.000028	7.14E-07	8.33E-06	2.32E-12	0.005527	6.80E-10
WA/12	0.0025	1.41E-05	0.000024	4.29E-06	2.78E-07	1.00E-12	0.00051	2.94E-10
WA/13	0.0275	1.76E-06	0.000032	2.14E-06	0.000111	1.92E-11	0.005548	5.63E-09
WA/14	0.025	1.29E-05	0.00003	7.14E-07	5.56E-06	6.21E-12	0.005027	1.82E-09
WA/15	0.00275	1.65E-05	0.000028	1.43E-06	2.78E-06	2.24E-12	0.000562	6.57E-10
WA /16	0.0025	2.35E-05	0.00004	2.86E-06	1.39E-05	9.66E-12	0.000518	2.83E-09
WA /17	0.0025	1.18E-06	0.00002	3.57E-06	2.78E-06	7.64E-13	0.000507	2.24E-10
WA /18	0.0525	1.29E-05	0.00022	7.14E-08	1.67E-05	1.33E-11	0.010586	3.90E-09
WA /20	0.0275	2.35E-06	0.00002	1.43E-06	5.83E-05	1.04E-11	0.005535	3.04E-09
WA /21	0.0025	5.88E-06	0.000026	2.86E-06	1.11E-05	3.48E-12	0.000511	1.02E-09
Min	0.0025	0	0.00002	7.14E-08	2.78E-07	0	0.000507	0
Max.	0.075	2.47E-05	0.00022	5.07E-05	0.000556	1.24E-10	0.01507	3.63E-08
Aver.	0.025513	7.06E-06	6.34E-05	7.69E-06	6.47E-05	2.21E-11	0.005149	6.47E-09
Stand Dev.	0.020011	8.16E-06	8.05E-05	1.34E-05	0.000127	3.25E-11	0.004023	9.51E-09

Health risk assessment: Drinking water contaminated with heavy metals poses many health risks. Most metals are highly polluting, making locals more vulnerable to the health risks associated with metal consumption. Heavy metals in water can cause stains on objects and linen, sticky coatings, and deposits in water pipes, in addition to an unpleasant taste in beverages (WHO, 2011; Mgbenu and Egbueri, 2019). Despite the lack of health-based temperature guidelines, hot water has been found to promote the growth of bacteria harmful to the human system, causing taste, odor, color and corrosion problems (WHO, 2011; Mgbenu and Egbueri, 2019).

Human health risk assessment is the method used in this study to determine the magnitude of the adverse health effects associated with human exposure to environmental hazards. The four major stages of risk assessment are hazard identification, exposure assessment, dose-response assessment, and risk characterization (Kollunu, *et al.*, 1996; Paustenbach, 2002). Obiri *et al.* (2006) developed a method for identifying hazards.

Asante-Duah (2002) includes a review of key literature to identify potential health concerns associated with metals and metalloids. The extent, duration, and order of heavy metal exposure are determined as part of the exposure assessment. The amount of metal and metalloids required to cause

various degrees of health effects that can lead to illness is determined by dose-response analysis. Finally, risk characterization entails determining the likelihood of heavy metals causing cancer or other diseases in the target population (Eyankware, *et al.*, 2022d). The HI results in Table 9 show that children have a higher HI for all parameters than adults. Anthropogenic sources may be the primary source of the study area's increased HI.

Hydrogeochemical Facies: The hydrochemical evolution, grouping, and areal distribution of groundwater's major dissolved constituents (major cations and major anions) can be graphically depicted (Akakuru *et al.*, 2015, Akakuru *et al.*, 2017; Eyankware *et al.*, 2021b). This study examined hydrochemical facies variation using Piper trilinear diagrams, Scholler plots, and Durov plots.

Piper Trilinear and Durov plots: The Piper Trilinear plot (Piper, 1944) is one of the most successful graphical representations in groundwater quality investigations; it aids in understanding shallow groundwater geochemistry and highlights chemical interactions more precisely than other plotting methodologies (Akakuru *et al.*, 2013; Eyankware *et al.*, 2020).

Table 8: CDI in the study area

Sample code	Fe(A)	Fe (C)	Mn (A)	Mn (C)	As (A)	As (C)	Zn (A)	Zn (C)	Cu (A)	Cu (C)
WA/01	8.57143E-05	0.0002	2.85714E-06	6.66667E-06	2.85714E-06	6.66667E-06	0.000143	0.000333	3.42857E-05	0.00008
WA/02	2.85714E-05	6.66667E-05	3.71429E-05	8.66667E-05	2.85714E-06	6.66667E-06	5.71E-05	0.000133	8.85714E-05	0.000207
WA/03	2.85714E-05	6.66667E-05	2.85714E-06	6.66667E-06	2.85714E-06	6.66667E-06	1.14E-05	2.67E-05	0.000571429	0.001333
WA/04	5.71429E-05	0.000133333	2.85714E-06	6.66667E-06	3.14286E-05	7.33333E-05	8.57E-05	0.0002	5.71429E-05	0.000133
WA/05	2.85714E-05	6.66667E-05	0	0	2.85714E-06	6.66667E-06	2.86E-06	6.67E-06	2.85714E-06	6.67E-06
WA/06	5.71429E-05	0.000133333	0.00006	0.00014	2.85714E-06	6.66667E-06	5.71E-06	1.33E-05	5.71429E-06	1.33E-05
WA/07	2.85714E-05	6.66667E-05	3.14286E-06	7.33333E-06	3.14286E-05	7.33333E-05	1.71E-05	0.00004	8.57143E-06	0.00002
WA/08	2.85714E-06	6.66667E-06	2.85714E-06	6.66667E-06	2.85714E-06	6.66667E-06	0.000203	0.000473	0.00006	0.00014
WA/09	3.14286E-05	7.33333E-05	2.85714E-06	6.66667E-06	3.14286E-05	7.33333E-05	3.14E-06	7.33E-06	3.71429E-05	8.67E-05
WA/10	3.71429E-05	8.66667E-05	0.000004	9.33333E-06	2.85714E-06	6.66667E-06	5.71E-06	1.33E-05	0.000228571	0.000533
WA/11	3.14286E-05	7.33333E-05	2.85714E-06	6.66667E-06	0.000004	9.33333E-06	2.86E-06	6.67E-06	8.57143E-06	0.00002
WA/12	2.85714E-06	6.66667E-06	3.42857E-05	0.00008	3.42857E-06	0.000008	1.71E-05	0.00004	2.85714E-07	6.67E-07
WA/13	3.14286E-05	7.33333E-05	4.28571E-06	0.00001	4.57143E-06	1.06667E-05	8.57E-06	0.00002	0.000114286	0.000267
WA/14	2.85714E-05	6.66667E-05	3.14286E-05	7.33333E-05	4.28571E-06	0.00001	2.86E-06	6.67E-06	5.71429E-06	1.33E-05
WA/15	3.14286E-06	7.33333E-06	0.00004	9.33333E-05	0.000004	9.33333E-06	5.71E-06	1.33E-05	2.85714E-06	6.67E-06
WA/16	2.85714E-06	6.66667E-06	5.71429E-05	0.000133333	5.71429E-06	1.33333E-05	1.14E-05	2.67E-05	1.42857E-05	3.33E-05
WA/17	2.85714E-06	6.66667E-06	2.85714E-06	6.66667E-06	2.85714E-06	6.66667E-06	1.43E-05	3.33E-05	2.85714E-06	6.67E-06
WA/18	0.00006	0.00014	3.14286E-05	7.33333E-05	3.14286E-05	7.33333E-05	2.86E-07	6.67E-07	1.71429E-05	0.00004
WA/20	3.14286E-05	7.33333E-05	5.71429E-06	1.33333E-05	2.85714E-06	6.66667E-06	5.71E-06	1.33E-05	0.00006	0.00014
WA/21	2.85714E-06	6.66667E-06	1.42857E-05	3.33333E-05	3.71429E-06	8.66667E-06	1.14E-05	2.67E-05	1.14286E-05	2.67E-05

Table 9: H.Q and HI in the study area

Sample code	Fe(A)	Fe ©	Mn (A)	Mn ©	As (A)	As (C)	Zn (A)	Zn ©	Cu (A)	Cu (C)
WA/01	0.000122449	0.000286	6.21118E-05	0.000145	0.009524	0.022222	0.000476	0.001111	0.000857	0.002
WA/02	4.08163E-05	9.52E-05	0.000807453	0.001884	0.009524	0.022222	0.00019	0.000444	0.002214	0.005167
WA/03	4.08163E-05	9.52E-05	6.21118E-05	0.000145	0.009524	0.022222	3.81E-05	8.89E-05	0.014286	0.033333
WA/04	8.16327E-05	0.00019	6.21118E-05	0.000145	0.104762	0.244444	0.000286	0.000667	0.001429	0.003333
WA/05	4.08163E-05	9.52E-05	0	0	0.009524	0.022222	9.52E-06	2.22E-05	7.14E-05	0.000167
WA/06	8.16327E-05	0.00019	0.001304348	0.003043	0.009524	0.022222	1.9E-05	4.44E-05	0.000143	0.000333
WA/07	4.08163E-05	9.52E-05	6.8323E-05	0.000159	0.104762	0.244444	5.71E-05	0.000133	0.000214	0.0005
WA/08	4.08163E-06	9.52E-06	6.21118E-05	0.000145	0.009524	0.022222	0.000676	0.001578	0.0015	0.0035
WA/09	4.4898E-05	0.000105	6.21118E-05	0.000145	0.104762	0.244444	1.05E-05	2.44E-05	0.000929	0.002167
WA/10	5.30612E-05	0.000124	8.69565E-05	0.000203	0.009524	0.022222	1.9E-05	4.44E-05	0.005714	0.013333
WA/11	4.4898E-05	0.000105	6.21118E-05	0.000145	0.013333	0.031111	9.52E-06	2.22E-05	0.000214	0.0005
WA/12	4.08163E-06	9.52E-06	0.000745342	0.001739	0.011429	0.026667	5.71E-05	0.000133	7.14E-06	1.67E-05
WA/13	4.4898E-05	0.000105	9.31677E-05	0.000217	0.015238	0.035556	2.86E-05	6.67E-05	0.002857	0.006667
WA/14	4.08163E-05	9.52E-05	0.00068323	0.001594	0.014286	0.033333	9.52E-06	2.22E-05	0.000143	0.000333
WA/15	4.4898E-06	1.05E-05	0.000869565	0.002029	0.013333	0.031111	1.9E-05	4.44E-05	7.14E-05	0.000167
WA/16	4.08163E-06	9.52E-06	0.001242236	0.002899	0.019048	0.044444	3.81E-05	8.89E-05	0.000357	0.000833
WA/17	4.08163E-06	9.52E-06	6.21118E-05	0.000145	0.009524	0.022222	4.76E-05	0.000111	7.14E-05	0.000167
WA/18	8.57143E-05	0.0002	0.00068323	0.001594	0.104762	0.244444	9.52E-07	2.22E-06	0.000429	0.001
WA/20	4.4898E-05	0.000105	0.000124224	0.00029	0.009524	0.022222	1.9E-05	4.44E-05	0.0015	0.0035
WA/21	4.08163E-06	9.52E-06	0.000310559	0.000725	0.012381	0.028889	3.81E-05	8.89E-05	0.000286	0.000667
HI	0.000833061	0.001944	0.007453416	0.017391	0.60381	1.408889	0.00205	0.004782	0.033293	0.077683

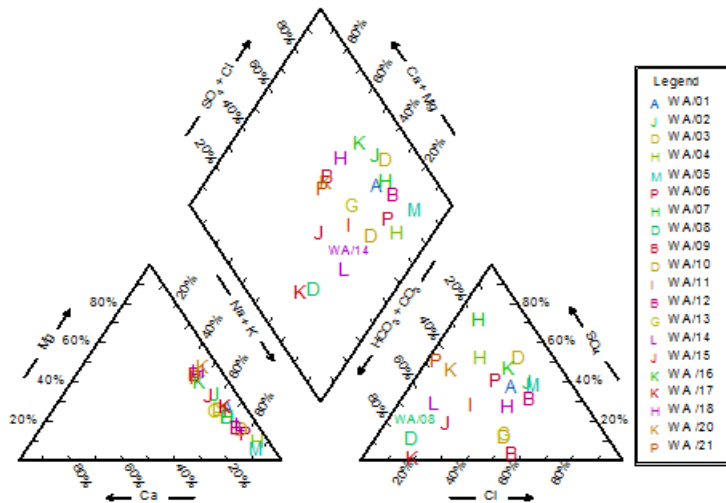


Fig 3: Piper diagram

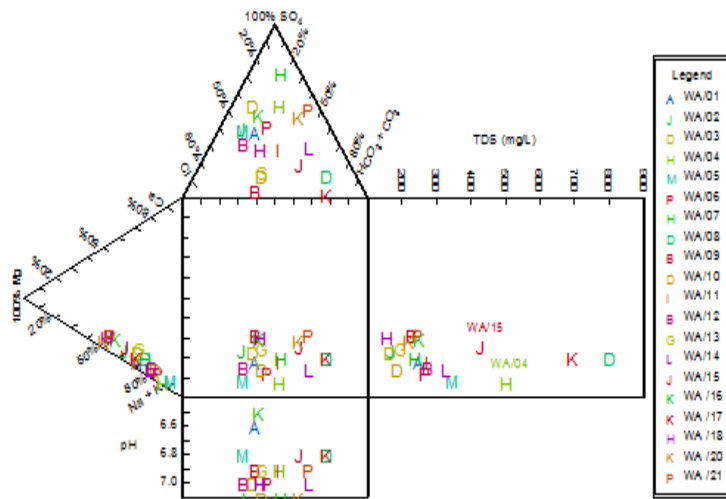


Fig. 4: Durov diagram

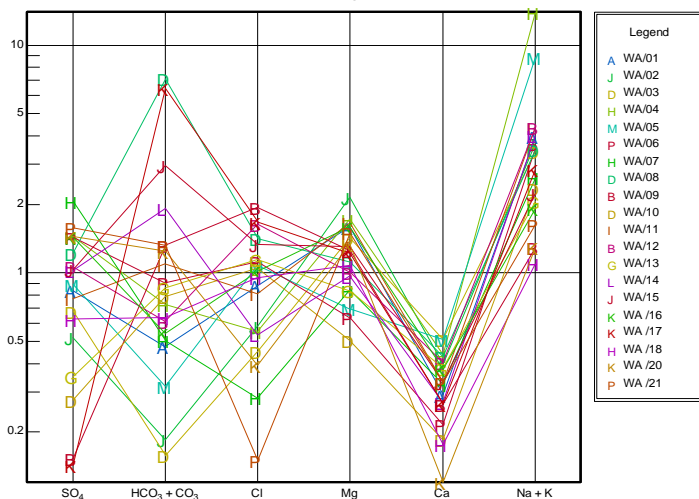


Fig. 5: Schoeller Semi-logarithm plot

From the Piper and Durov plots (Figs. 3 and 4), it shows that 79.2% of the water sample had Na +K had as the major ionic specie while 20.8% had no dominant ionic specie within the cation area, while within the anion area, it has 25%, 20% and 5% of ionic dominances of HCO_3+CO_3 , SO_4 , and Cl respectively, with 50% of its samples having no dominant ionic specie. In the same vein, the geochemical zone of the samples are 7 (which is the sodium chloride type (60%), and the 9 (mixed type (40%)). Sodium chloride and most other chloride salts have high solubility. Depending on the temperature, the dissolution limit for NaCl is about 35.5 weight percent. As shown in the analysis of the table, many wells show significant chloride and alkali metal concentrations. When the concentration reaches a level comparable to seawater (3.5% by weight or more), they are called brine. A medium-concentration source that contains enough NaCl to produce a salty taste is brackish water. The following mechanism can produce high salt springs: a salt water spring from the evaporative bed in the catchment area of the spring; sources contaminated by anthropogenic sources, especially salt used in de-icing roads; coastal fountain affected by inrush current; and well to drain oil field brine.

Schoeller Semi-logarithm chart: Another relationship technique that uses direct diagrams is the Schoeller Semi-logarithm chart. The most popular charts for communicating water quality use math or logarithmic scales (Sakram et al., 2013).

The chart proposed by (Schoeller, 1977) depicts a collection of investigations on equidistant verticals, the number of which depends on the number of constituents being communicated. This diagram is especially useful for examining waters with low focus and waters with a wide range of fixation (Sakram et al., 2013; Saha et al., 2019; Olofinlade et al., 2018). The Schoeller graph of the review region (Figure 5) uncovers a hydrogeochemical pattern of $\text{Na}^+ + \text{K}^+ > \text{HCO}_3^- + \text{CO}_3^{2-} > \text{Mg}^{2+} > \text{SO}_4^{2-} > \text{Cl}^- > \text{Ca}^{2+}$ in the request for the most noteworthy to the least constituent. The information's Schoeller semi-logarithmic plots confirmed the previous plot's water type. The pinnacles represent the most common particles in the water tests, while the box represents the less common particles. This plot corresponds to the Piper and Durov plots in this study. The similarity of the hydrogeochemical movement trend, on the other hand, implies that the groundwater must have originated from the same source (Akakuru et al. 2017; 2021a)

Conclusion: Groundwater quality assessments have been conducted to assess the impact of human activities on groundwater quality. The hydrogeochemical facies results revealed that 79.2 percent of the water samples had Na +K as the major ionic specie, while 20.8 percent had no dominant ionic specie within the cation area, and 25 percent, 20 percent, and 5 percent of its samples had no dominant ionic specie within the anion area, with 25 percent, 20 percent, and 5 percent of its samples having no dominant ionic specie. Similarly, the geochemical zones of the samples are 7 (which is the sodium chloride type (60 percent)) and 9 (mixed type) (40 percent). Sodium chloride and most other chloride salts are highly soluble. The temperature-dependent dissolution limit of NaCl is 35.5 weight percent. In order of most significant to least significant constituent, the Schoeller graph of the review region reveals a hydrogeochemical pattern of $\text{Na}^+ + \text{K}^+ > \text{HCO}_3^- + \text{CO}_3^{2-} > \text{Mg}^{2+} > \text{SO}_4^{2-} > \text{Cl}^- > \text{Ca}^{2+}$. The Schoeller semi-logarithmic plots of the data confirmed the previous plot's water type. Nevertheless, appropriate remedial treatment for groundwater is recommended in areas where trace elements exceed the WHO recommended upper limit. As a result, it is possible to conclude that the groundwater in the study area is "safe" for drinking and domestic use.

REFERENCE

- Agidi, B.M., Akakuru, O.C., Aigbadon, G.O. Schoeneich K., Isreal H, Ofoh I, Njoku J, Esomonu I (2022). Water quality index, hydrogeochemical facies and pollution index of groundwater around Middle Benue Trough, Nigeria. *Int J Energ Water Res* (2022). <https://doi.org/10.1007/s42108-022-00187-z>
- Akakuru, O.C., Maduka, E.C., Akakuru, O.U. (2013). Hydrogeochemical characterization of surface water sources in Owerri Capital Territory, Southeastern Nigeria. *IOSR J. Appl. Geol. Geophysics. 1* (2). 32-38
- Akakuru, O., Akudinobi, B., Okoroafor, P., & Maduka, E. (2017a). Application of geographic information system in the hydrochemical evaluation of groundwater in parts of Eastern Niger Delta Nigeria. *American J. Environ. Policy and Manage.* 3(6), 39–45
- Akakuru, O.C., Akudinobi, B., Opara, A.I., Onyekuru, S.O. & Akakuru, O.U. (2021a). Hydrogeochemical facies and pollution status of groundwater resources of Owerri and environs, Southeastern Nigeria. *Environ. Monit. Assess.* 193: 623.
- Akakuru, O.C., Akudinobi, B.E., Nwankwoala, H.O., Akakuru, O.U. & Onyekuru, S.O (2021b). Compendious evaluation of groundwater in parts of Asaba, Nigeria for agricultural sustainability. *Geosciences J.* <https://doi.org/10.1007/s12303-021-0010-x>
- Akakuru, O.C, Eze, C.U., Okeke, O.C, Opara A.I., Usman, A.O., IHEME O.K, Ibeneme, S.I, & Iwuoha, P.O (2022). Hydrogeochemical evolution, water quality indices, irrigation suitability and pollution index of groundwater (PIG) around Eastern Niger Delta, Nigeria. *Inter. J. Ener. Wat. Res.* <https://doi.org/10.1007/s42108-021-00162-0>
- Bhutian R, Dipali Bhaskar Kulkarni, D. R. Khanna, Ashutosh Gautam (2017) Geochemical distribution and environmental risk assessment of heavy metals in groundwater of an industrial area and its

- surroundings, Haridwar, India. *Energy. Ecol. Environ.* 2(2):155–167
- Bortey-Sam, N., Nakayama, S.M., Ikenaka, Y., Akoto, O., Baidoo, E., Yohannes, Y.B; Mizukawa, H., Ishizuka, M. (2014). Human health risks from metals and metalloid via consumption of food animals near gold mines in Tarkwa, Ghana: Estimation of the daily intakes and target hazard quotients (THQs). *Ecotoxicol. Environ. Safety.* 111, 160
- Barzegar R, Moghaddam AA, Nazemi AH, Adamowski J (2018) Evidence for the occurrence of hydrogeo-chemical processes in the groundwater of Khoy plain, northwestern Iran, using ionic ratios and geochemical modeling. *Environ. Earth Sci.* 77:597.
- Caeiro, S., Costa, M. H., Ramos, T. B., Fernandes, F., Silveira, N., Coimbra, A., Medeiros, G., Painho, M. (2015). Assessing heavy metal contamination in Sado estuary sediment: an index nnuivsis approach. *Ecol Indic* 5:151–169
- Domenico PA (1972) Concepts and models in groundwater hydrology. McGraw-Hill, New York
- Duggal, V., Sharma, S., Saini, K. et al. (2017) Assessment of carcinogenic and non-carcinogenic risk from exposure to uranium in groundwater from western Haryana, India. *J Geol Soc India* 89, 663–668.
- Eyankware, M.O., Nnabo, P.N., Ogwah, C. (2020a). Impact of past mining activity on water resources around active and abandoned mines in Ebonyi State, South-Eastern Nigeria- A mini review. *Hydro Science and Marine Engineering*, 2(2): 29-35.
- Eyankware, M. O., Obasi, P.N., Omo-Irabor, O. O., Akakuru. O. C. (2020b). Hydrochemical characterization of abandoned quarry and mine water for domestic and irrigation uses in Abakaliki, southeast Nigeria. *Modeling Earth Systems and Environment*. <https://doi.org/10.1007/s40808-020-00827-5>
- Eyankware, M. O., Igwe, E. O., Onwe, M. O. (2021a). Geochemical study using geochemical modelling approach in Ojekwe region of southern Benue Trough, Nigeria. *International Journal of Energy and Water Resources*. <https://doi.org/10.1007/s42108-021-00163-z>.
- Eyankware, M.O., Ulakpa, R.O.E, Ike, J. (2021b). Re-evaluating coastal aquifer using graphical and geochemical approach; a case study of Niger Delta Region, Nigeria. *World Scientific News*, 154; 133-151.
- Eyankware, M. O., Ogwah, C., Umayah, O. S. (2021c). Integrated geophysical and hydrogeochemical characterization and assessment of groundwater studies in Adum West Area of Benue State, Nigeria. *J. Geological Res.* 65, 65881289
- Eyankware, M. O., Ephraim, B. E. (2021). A comprehensive review of water quality monitoring and assessment in Delta State, Southern Part of Nigeria. *J. Environ. Earth Sci* 3(1); 16-28.
- Eyankware, M.O., Akakuru, C. O. (2022). Appraisal of groundwater to risk contamination near an abandoned limestone quarry pit in Nkalagu, Nigeria, using enrichment factor and statistical approaches. *Journal of Energy and Water Resources*, <https://doi.org/10.1007/s42108-022-00186-0>
- Eyankware, M.O., Akakuru, C. O., Eyankware, E.O. (2022a). Interpretation of hydrochemical data using various geochemical models: a case study of Enyigba mining district of Abakaliki, Ebonyi state, SE. Nigeria. *Sustainable Water Resources Management*, <https://doi.org/10.1007/s40899-022-00613-4>
- Eyankware, M.O., Akakuru, C. O., Ulakpa, R. O. E., Eyankware, E.O. (2022c). Hydrogeochemical approach in the assessment of coastal aquifer for domestic, industrial, and agricultural utilities in Port Harcourt urban, southern Nigeria. *International Journal of Energy and Water*

- Resources. <https://doi.org/10.1007/s42108-022-00184-2>
- Eyankware, M.O., Akakuru, C. O., Eyankware, E.O. (2022d). Hydrogeophysical delineation of aquifer vulnerability in parts of Nkalagu areas of Abakaliki, SE. Nigeria. *Sustainable Water Resources Management*, <https://doi.org/10.1007/s40899-022-00603-6>
- Gopinath, S, K., Srinivasamoorthy, K., Saravanan, R., Prakash, D., Karunanidhi, D. (2019). Characterizing groundwater quality and seawater intrusion in coastal aquifers of Nagapattinam and Karaikal, South India using hydrogeochemistry and modeling techniques, Human and Ecological Risk Assessment: *An Inter. J.* 1-22.
- Horton, R. K. (1965). An index-number system for rating water quality. *J. Water Pollute Control Federation*, 37:300–306
- Li, P., Lin, C., Cheng, H., Duan, X., Lei, K. (2015). Contamination and health risks of soil heavy metals around a lead/zinc smelter in southwestern China. *Ecotoxicology Environ. Safety*, 113: 391–399.
- Mgbenu C. N., Egbueri, J. C. (2019). The hydrogeochemical signatures, quality indices, and health risk assessment of water resources in Umunya district, southeast Nigeria. *Appl. Water Sci.* 9:22
- Muller G (1979) Index of Geoaccumulation in sediments of Rhine River. *Geojournal* 2:108–118
- Obiri, S. (2007). Determination of heavy metals in water from boreholes in Dumasi in the Wassa West District of western region of Republic of Ghana. *Environ. Monit. Assess.* 130(1–3):455–463
- Olofinlade, W. S., Daramola, S. O., & Olabode, O. F. (2018). Hydrochemical and statistical modeling of groundwater quality in two contrasting geological terrains of southwestern Nigeria. *Modeling Earth Systems and Environment*.
- Piper, A. M. (1944). A graphic procedure in the geochemical interpretation of water analysis. *Transactions of the American Geophysical Union*, 25, 914–923.
- Saha S, Reza A. H. M. S and Roy MK (2019) Hydrochemical evaluation of groundwater quality of the Tista foodplain, Rangpur, Bangladesh. *Appl. Water Sci.* 9:198
- Satya NS, Tajdarul HS, Alok K, Debanjan S (2015): Hydrogeochemical characterization and quality assessment of groundwater in parts of Southern Gangetic Plain. *Environ Earth Sci.* (2016) 75:232
- Singovszka E, Balintova, M., Demcak, S., & Pavlikova, P. (2017). Metal Pollution Indices of Bottom Sediment and Surface Water Affected by Acid Mine Drainage. *Metals*, 7, 284
- Singh, V. K., Singh, K. P., Mohan, D. (2005). Status of heavy metals in water and bed sediments of River Gomti—a tributary of the Ganga River, India. *Environ. Monit. Assess.* 105:43–67
- Sophiya MS, Syed TH (2013) Assessment of vulnerability to seawater intrusion and potential remediation measures for coastal aquifers: a case study from eastern India. *Environ Earth Sci.* 70(3):1197–1209.
- Subba Rao, N. (2012). PIG: a numerical index for dissemination of groundwater contamination zones. *Hydrology Process*, 26 (22), 3344–3350.
- Subba Rao, N., Subrahmanyam, A., Ravi Kumar, S., Srinivasulu, N., Babu Rao, G., Rao, P. S., Reddy, G. V. (2012) Geochemistry and quality of groundwater of Gummanampadu sub-basin, Guntur District, Andhra Pradesh, India *Environ. Earth Sci.* 67:1451-1471
- Usman, A.O., Iheme, K.O, Chinwuko, A.I, Azuoko, G. & Akakuru, O.C (2022). Hydro-geophysical investigation of groundwater resources within Abakaliki, Lower Benue Trough, Nigeria. *COOU J. Phys. Sci.* 5 (1), 473-491
- Urom, O.O., Opara, A.I., Usen, O.S., Akiang, F.B., Isreal, H.O., Ibezim, J.O. & Akakuru, O.C.(2021). Electro-geohydraulic estimation of shallow aquifers of Owerri and environs,

- Southeastern Nigeria using multiple empirical resistivity equations. *Inter. J. Energy and Water Resources*. 1-22.
- US Environmental Protection Agency (USEPA). (1989). Risk assessment guidance for superfund, vol 1, human health evaluation manual (Part A). Ofce of Emergency and Remedial Response, Washington, DC
- USEPA (1994) Drinking water: maximum contaminant level goal and national primary drinking water regulation for lead and copper. *Fed Regist* 59(125):33860–33864
- World Health Organization. (2011). *Guidelines for Drinking Water Quality*, 4th ed. (World Health Organization, Geneva).
- Yang, Q., Lun, W., Fang, Y. (2011). Numerical modeling of three dimension groundwater flow in Tongliao (China). *Procedia Engineering*. 24, 638-642
- Yahaya, S. M., Fatima, A., Nafiu, A. (2021). Ecological risk assessment of heavy metal contaminated soils of selected villages in Zamfara State, Nigeria. *SN Appl. Sci*. 3:168