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Fluoride contamination in groundwater sources in Southwestern Nigeria: Assessment using multivariate statistical approach and human health risk

Order of Authors

1. **Emenike Chidozie PraiseGod* (***praisegod.emenike@covenantuniversity.edu.ng***)**
[1] Department of Civil Engineering,
Covenant University, Ota, Ogun State, Nigeria

[2] Cranfield Water Science Institute,
School of Water, Energy and Environment,
Cranfield University, Bedfordshire, United Kingdom.
***(Corresponding author)**

2. **Tenebe Imokhai Theophilus (***imokhai.tenebe@covenantuniversity.edu.ng***)**
[1] Department of Civil Engineering,
Covenant University, Ota, Ogun State, Nigeria.

3. **Peter Jarvis (***p.jarvis@cranfield.ac.uk***)**
Professor of Water Science & Technology
[2] Cranfield Water Science Institute,
School of Water, Energy and Environment,
Cranfield University, Bedfordshire, United Kingdom.

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Fluoride contamination in groundwater sources in Southwestern Nigeria: Assessment using multivariate statistical approach and human health risk

Abstract

The present study investigated the ionic and fluoride concentrations in tap water and its associated health risk to local dwellers of Ogun State (Abeokuta south), Nigeria. 63 samples were collected from twenty-one different locations. Results obtained revealed the mean concentration of fluoride (F^-) as 1.23 mg/L. Other water quality parameters such as total dissolved solids (TDS), electrical conductivity (EC), $Fe^{2+,}$ and SO_4^{2-} surpassed the WHO limits for drinking water quality. Strong positive correlation was observed between F^- and TDS; F^- and pH; TDS and EC; TDS and Mg^{2+}; TDS and SO_4^{2-}; TDS and HCO_3^-; EC and HCO_3^-; EC and SO_4^{2-}; Na^+ and Cl^-; SO_4^{2-} and Cl^-. In addition, Empirical Bayesian Kriging (EBK) model was employed to spatially distribute the concentration of the analyzed elements within the study region. The chronic daily dose (*CDD*) and hazard quotient (*HQ*) were also used to evaluate the health risk associated with F^-, considering dermal and ingestion as pathways. The results revealed that the associated HQ for infants between the age range of 6–12 months within about 91% of the study region surpassed the accepted *HQ* limit. However, the *HQ* for age categories 11–16 years; > 65years; 18–21years; ≥ 21years; 16–18years within 95.2%, 90.5%, 80.95% and 100% of the study location were less than 1. Conclusively, the *HQ* values obtained in this study should serve as a baseline information for water management authorities, policymakers and the society at large towards addressing these pollution issues.

Keywords: Risk assessment; Abeokuta; Pollution; Groundwater; Dispersion; Southwestern Nigeria.

1. Introduction

The presence of fluoride at elevated concentrations in drinking water has caused severe health effects in humans in some parts of the world (Bhatnagar et al., 2011; Shen and Schäfer, 2015). Fluorine exists in the environment through combination with other elements to form highly soluble fluoride compounds. The main source of fluorine in water is from natural deposition from geogenic sources in aquifers (EPA, 2010; Sun et al., 2013). The primary way by which humans ingest fluoride is through consumption of contaminated groundwater (Singh et al., 2013; Singh and Mukherjee, 2015; Subba Rao et al., 2013). The World Health Organization (WHO) recommended the concentration of fluoride that can cause minimal health risk to be 1.5 mg/L. However, more than 200 million people living across 20 developing and developed countries regularly consume water with elevated fluoride concentrations above the standard guidelines set by the WHO (Amini et al., 2008; Fawell et al., 2006; Shen and Schäfer, 2015). Some countries such as Tanzania in the East of Africa have a drinking water standard for fluoride of 4 mg/l, well above the WHO recommended value. This can cause a number of possible health problems, including dental and skeletal fluorosis. In the Rift Valley, East Africa, more than 80 million people display a range of symptoms consistent with dental fluorosis (Shen and Schäfer, 2015; Smedley et al., 2002). However, this should also be placed in the context of water scarcity, population growth and access to clean water in the region.

High concentrations of fluoride in humans can lead to various health problems such as nervous system damage (Kaoud and Kalifa, 2010), reduced fertility (Izquierdo-Vega et al., 2008), intellectual impairment in children (Ding et al., 2011; Shivaprakash et al., 2011), urinary tract disease (Jha et al., 2011), as well as dental and skeletal fluorosis in children and adults (Maguire,

2014). In turn, this can lead to significant lower back pains (Namkaew and Wiwatanadate, 2012). It has been proposed that regularly consuming water with fluoride concentrations of at least 0.9 mg/l is the cause of at least 37% of dental fluorosis cases (McGrady *et al.,* 2012). Levy and Leclerc, (2012) also correlated fluoride in drinking water with bone diseases (Osteosarcoma) in adolescents and children. Sun *et al.,* (2013) discovered that cases of hypertension in adults could be linked to fluoride present in drinking water. They further emphasized that fluoride exposure could cause an increase in plasma Endothelin-1 (ET–1) levels. Liu *et al.,* (2014) also identified a link between fluoride exposure from drinking water and carotid artery atherosclerosis in adults. In a recent study conducted by Irigoyen–Camacho *et al.* (2016), reports showed that increased morbidity and mortality rate could be correlated with nutritional deficiencies propagated by fluoride intake. It should be noted that at low concentrations, ingesting fluoride from drinking water can hinder dental caries and some health authorities deliberately add fluoride to drinking water to reduce the incidence of enamel decay (Freire et al., 2016).

In the urban regions of Nigeria, the wholescale provision of reliable access to drinking water has been made difficult due to increasing population and habitation spread within cities (Emenike et al., 2016; Odjegba et al., 2015, 2014), a similar case as seen in Tanzania. Many localities in southwestern Nigeria and Tanzania, in these regions, are supplied with surface water for their source of drinking water (Adekunle et al., 2013). However, in many cases, these surface waters are heavily contaminated and polluted, particularly with respect to the microbial quality of the water, and are subsequently poorly treated making this water unfit for human consumption (John-Dewole, 2012; Tenebe et al., 2017, 2016). For this reason, many people have accessed groundwater for their drinking water given its perceived higher quality (Adekunle et al., 2013). However, given the noted pollution of such groundwater systems with fluoride and the evidenced

adverse effect on human health, a stronger link between fluoride concentration and its health impacts is needed to enable more thorough risk assessment and mitigation measures to be applied.

Recently, researchers have adopted the human health risk assessment (HHRA) model to determine the adverse effect of absorbing or ingesting chemicals at a range of concentrations (Yang et al., 2012; Zhai et al., 2017). Therefore, this study seeks to consider two common pathways (ingestion and dermal) by which human populations can be exposed to high fluoride concentrations. In the same vein, the study capitalizes on the risk assessment indicators endorsed by the US EPA (2011) to assess the extent of fluoride contamination in groundwater and evaluate in detail, the health risk associated with fluoride pollution in groundwater. In furtherance of this, reports of elevated fluoride concentration in drinking water and its potential health risk has been investigated in Tunisia (Guissouma et al., 2017), Iran (Battaleb-Looie et al., 2013; Dehbandi et al., 2018; Yousefi et al., 2018), Ghana (Craig et al., 2015; Salifu et al., 2012), China (Zhang et al., 2017) and United Arab Emirates (Walia et al., 2017). But, to the best of our knowledge, there are a dearth of literatures in Nigeria with no previous investigation in the studied region in this regard. Therefore, this study focuses on the following objectives. First, to determine the spatial distribution of fluoride and other groundwater quality parameters in Abeokuta. Secondly, to analyze the interrelationship between fluoride and other water quality parameters. Thirdly, adopt geostatistical modeling in which semivariogram graphs will be used to describe and validate the extent of contamination via the interpolation of known concentration from sampled locations. Finally, assess the health risks associated with fluoride concentration in groundwater. To establish a more realistic base for judgement, ingestion and dermal pathways were investigated. Also, the receptors (individuals) were classified into seven age groups (6 – 12 months, 6 – 11 years, 11 – 16 years, 16 – 18 years, 18 – 21 years, $\geq$ 21 years and > 65 years). This will assist in adding relevant data to local rural

93 water management, policy and decision makers to take adequate measures in safeguarding the
94 lives of residents in endemic fluoride regions.

95 ## 2. Materials and Methods

96 ***Study area***

97 Abeokuta, the capital of Ogun state, occupies 40.6 km^2 (latitude 7.17–7.25^{o}N; longitudes 3.28–
98 3.43^{o}E). The population of Abeokuta is estimated to be 451,600 with an annual growth rate of
99 3.5% (National Population Commission, 2010). The area sits on a complex geological system,
100 composed of a mix of rock from the Precambrian period. The complex spreads through the
101 southwestern region of Nigeria, sharing a part with the Dahomey basin composed of sedimentary
102 rock (Rahaman, 1976). The region is also linked to Lagos by river/canal (130 km) or by railway
103 (77 km) and shares common borders with Ibadan, Shagamu, Ilaro, Ketuo, and Iseyin. The sample
104 locations were targeted at regions with high population density.

105

106 **"Fig. 1 is about here."**

107

108 ***Sample collection***

109 The water samples analyzed in this study were obtained from taps that were connected directly to
110 the underlying groundwater aquifer. In total, 63 water samples were collected from twenty-one
111 locations (Fig. 1) in the study area (R1 – R21). The sampled taps were regularly used by
112 householders mainly for consumption and domestic activities. At the point of collection, the taps
113 were allowed to run for about 15 min before representative samples were collected to reflect the

114 status of the aquifer and the water consumed by the householder. Sample bottles (made of
115 polyethylene) were washed with distilled water (mixed with 20% nitric acid) prior to sampling.
116 The sample bottles were further rinsed three times with distilled water to remove any trace of acid
117 and then air-dried before transporting to the sampling site. At the sampling location, the bottles
118 were rinsed three times with tap water prior to collection to ensure no acid interference. After
119 sample collection, the containers were sealed with screw corks, labeled appropriately, placed in a
120 container filled with ice and immediately transferred to a refrigerator regulated at 4 ^{o}C.

121 As soon as each sample was collected, unstable and sensitive water quality parameters such as
122 total dissolved solids (TDS), electrical conductivity (EC), temperature (Temp.), alkalinity (Alka.)
123 and pH were measured in situ with a calibrated multiparameter probe (HANNA – HI2030
124 Salinity/TDS/EC meter and HANNA – HI98130). Standard analytical procedures (APHA, 2005)
125 were followed to measure the ionic concentration of major ions. Sodium (Na+), potassium (K^{+}),
126 magnesium (Mg^{2+}), calcium (Ca^{2+}), iron (Fe^{2+}) and manganese (Mn) were measured using flame
127 photometric method (Flame Atomic Absorption Spectrophotometry), dissolved silica (SiO_2) by
128 molybdosilicate method with UV–Visible spectrometer and Nitrates (NO_3^{-}) concentration by UV–
129 Visible spectrometer. Sulfate (SO_4^{2-}) concentration was determined by the turbidimetric method,
130 bicarbonate (HCO_3^{-}), chloride (Cl^{-}), and carbonate (CO_3^{2-}) concentration were measured by
131 volumetric method. A calibrated potentiometric ion-selection electrode (HANNA–HI5315)
132 attached to a water-resistant portable ORP/pH/ISE meter (HANNA–HI98191) was used to
133 measure the concentration of fluoride ion (F^{-}).

134 ***Quality control***

135 Quality assurance was achieved through the implementation of standard laboratory procedures and
136 quality control techniques which included standardized calibration, replication, use of analytical

grade reagent blanks and spikes, and following standard operating measures. Calibration of multiparameter instruments was performed after collecting samples from a location. The samples were analyzed in triplicate, and the mean value were recorded. During metal analysis, constant monitoring of reagent blanks was administered, and standard detection limits were maintained.

Geostatistical modeling

Water quality parameters were modelled spatially using the Empirical Bayesian Kriging (EBK) model. The EBK model automatically transforms the complicated aspects of the Kriging model by forming a restricted neighbor between a mapped property via interpolation. The interpolation relies on the subpopulation of the available dataset to produce independent trends (Samsonova et al., 2017). One essential element that differentiates the EBK model from other Kriging models is the ability to consider the uncertainties in semivariogram computation (Magesh et al., 2017). Thus, the operational module of the EBK estimation is based on the exploitation of original data to predict the semivariogram model (SVM) where a new set of data is generated at the original data point. Furthermore, the freshly simulated data is used to predict the SVM. During the estimation of the new SVM, predicted standard errors are processed at locations that had no observation, producing several spectrums of SVM due to repeated operations from original points. This study adopted the K-Bessel SVM which uses empirical transformations in simple remodeling. A study by Krivoruchko, (2011) noted that the method mentioned above is the most applicable, explicit and reliable technique but with a limitation of prolonged processing time.

2.1 Health risk assessment

Health risk assessment is a verified method that has been adopted extensively for the evaluation of potential hazards on human health after being exposed to certain chemicals over a period (US

159 EPA, 1989). In this study, the potential health risk of fluoride concentration on the population was
160 evaluated. Due to the behavioral and physiological attributes of different age groups, the
161 population was divided into seven age categories (6–12 months, 6–11 years, 11–16 years, 16–18
162 years, 18–21 years, ≥ 21 years and > 65years) assessed from the US EPA exposure factor handbook
163 (US EPA, 2011). Several pathways have been identified through which humans can be exposed to
164 chemical risk. They include inhalation, dermal and ingestion pathways. However, this study
165 considered dermal and ingestion pathways only because of the unavailability of the transfer
166 efficiency of fluoride from water to air and its inhalation reference dose. Furthermore, inhalation
167 was not considered as a major exposure route for fluoride (National Academy of Science, 2006)
168 The chronic daily dose of fluoride via ingestion and dermal pathways were estimated using Eqn.
169 (1) and Eqn. (2) based on (US EPA, 2011) recommendation.

170 $$CDD_{IN} = \frac{C_{fw} \times IR_w \times EF_r \times ED}{BW \times AT_r} \quad (1.1)$$

171 $$CDD_{DE} = \frac{C_{fw} \times SA \times K_p \times EF_r \times ED \times ET \times CF}{BW \times AT_r} \quad (1.2)$$

172 Where CDD_{IN} and CDD_{DE} are the estimated chronic daily dose of fluoride via ingestion and
173 dermal exposure route respectively $(\mu g / kg \bullet day)$; C_{fw} is the concentration of fluoride in drinking
174 water $(\mu g / L)$; IR_w is the ingestion rate (L / day); SA is the exposed skin area (cm^2); K_p is the
175 dermal permeability coefficient for water $(unitless)$; EF_r is the resident exposure frequency
176 $(days / year)$; ED is the exposure duration $(year)$; ET is the water exposure time $(hours / day)$;
177 BW is body weight (kg); AT_r is the averaging resident time $(days / year)$ and CF is the unit
178 conversion factor (L / cm^3).

179 The hazard quotient (*HQ*) of fluoride exposure via ingestion and dermal pathways was calculated
180 using Eqn. (3) and Eqn. (4)

181 $$HQ_{IN} = \frac{CDD_{IN}}{RfD} \quad (3)$$

182 $$HQ_{DE} = \frac{CDD_{DE}}{RfD} \quad (4)$$

183 Where HQ_{IN} and HQ_{DE} are the ingestion hazard quotient and dermal hazard quotient respectively;
184 RfD is the reference dose of fluoride (equals 0.06 mg/kg-day according to the Integrated Risk
185 Information System (IRIS) database of the US EPA) in a particular route. The reference values of
186 each parameter used for calculating the CDD_{IN}, CDD_{DE}, HQ_{IN} and HQ_{DE} are compiled in Table
187 S1 (in the supplementary material).

188 **2.2 Data analysis**

189 After the analyses of the water samples, the results were subject to descriptive statistical analyses.
190 For each water quality parameter, the mean, minimum, maximum, and quartiles were calculated
191 using GraphPad Prism 6 for Windows (GraphPad Software Inc.). The geospatial map of fluoride
192 contamination and water quality parameters was achieved using ArcMap 10.3.1. The principal
193 component analysis (influence plot, scree plot, score plot and loading plot) and correlation matrix
194 were executed using Unscrambler X (CAMO software AS, version 10.4). The representation of
195 the different water types on a Piper diagram was achieved using Rockworks17 64bit. The EBK
196 model and semivariogram plots were calculated using geostatistical analyst tool in ArcMap 10.3.1.

197

198

3. Results and discussion

Within the study region, the TDS and EC values ranged from 498.33 to 2126.00 mg/L and 665.33 to 3314.67 µS/cm respectively (Table S2). According to the results, 4.76% of the TDS samples were in a range considered desirable for drinking, 42.86% were in the permissible for drinking category and, 52.85% were useful for irrigation purposes as defined by WHO standards (WHO, 2011). Concerning EC, 23.80% were within the permissible limits for drinking, and the remaining 76.2% exceeded WHO stipulated standards. The cause of the increased TDS and EC concentrations could be due to heavy application of agro-chemicals, rainwater percolation, ion exchange and sediment dissolution (Chabukdhara et al., 2017). Relating the results of other water quality parameters with WHO guidance for drinking water (WHO, 2011), Cl^- values varied from 32.60 mg/L to 546.20 mg/L in all samples with 19% of samples exceeding the WHO threshold. The presence of chloride originates from dissolved chloride salts found in minerals as well as animal and human waste. Moreover, the heightened chloride concentration may be as a result of industrial and commercial activities within these zones.

Na^+ concentration varied from 55.97 mg/L to 514.73 mg/L. Out of the twenty-one locations sampled, only eight locations revealed Na^+ concentration exceeding the WHO threshold of 200 mg/L. It was evident that all locations had a high degree of variability in Na^+ and Cl^- concentrations with high standard deviation values of ±112.88 and ±140.88 respectively. It should be noted that Na^+ can be caused by cation exchange processes that occur in the aquifer. Furthermore, wastewater pollution from anthropogenic sources and intrusion from septic tanks may also increase sodium levels (Wayland et al., 2003).

HCO_3^- and CO_3^{2-} concentrations were observed in the range from 278.30 to 666.27 mg/L and 0–25.67 mg/L respectively. The concentration of HCO_3^-in the region exceeded WHO recommended

levels of 500 mg/L in 14.30% of the samples. Meanwhile, no permissible limit has been provided by the WHO standards for CO_3^{2-} in drinking water. Elevated levels of bicarbonate were likely to be from natural dissolution from rocks, as well as runoff, irrigation, and infiltration processes having contact with the groundwater systems during recharge (Rasool et al., 2016; Singh et al., 2013).

From the results displayed in Table S2, it was observed that SO_4^{2-} values varied from 35.73 mg/L to 978.60 mg/L. When the SO_4^{2-} values were compared with WHO guidelines for drinking water, 42.9% of the samples exceeded these limits. It is important to note here that elevated SO_4^{2-} concentrations in drinking water may result in respiratory illnesses (Subba Rao, 1993). NO_3^- concentrations ranged from 0.00 mg/L to 25.67 mg/L. These NO_3^- concentrations were low and within permissible limits, although its presence in water is attributed to anthropogenic activities resulting from fertilizer use (Tirkey et al., 2017).

Ca^{2+} and Mg^{2+} are important indicators of water hardness. Within the study area, Ca^{2+} values were within the permissible limits (200 mg/L) set by WHO except for sample point R17 (227.67 mg/L) that exceeded the limit. Some 33.33% of the water samples exceeded the WHO most desirable limits (75 mg/L) for Ca^{2+} while 52.38% of the analyzed water samples were above the WHO most desirable limit (50 mg/L) for Mg^{2+} concentration in drinking water. The reason for the variability of Mg^{2+} concentration may be associated with ion exchange linked with dissolved rock and soil minerals in the water. K^+ and Fe^{2+} values ranged from 2.07 to 9.30 mg/L and 0.02 to 2.96 mg/L respectively. All samples showed K^+ concentration lower than the WHO threshold but the Fe^{2+} concentration surpassed the WHO allowable limits in 38.10% of the samples. Although iron is not regarded as hazardous to human health, its presence could be seen as a nuisance or aesthetic pollutant. Fe^{2+} often coexists with Mn, which was found to be lower than the allowable limits

recommended by WHO guidelines for drinking water (WHO, 2011) in all samples. The co-existence of Mn and Fe^{2+} is proposed to occur from the natural dissolution from gneiss and biotite rock formation. Furthermore, a study by Magesh et al. (2017) showed that a deep groundwater aquifer could be laden with Fe^{2+} if ionic activities in aquifers were high.

The concentration of F^- ranged from 0.48 to 1.84 mg/L with a mean value of 1.23 mg/L. It was observed that 33% of the samples surpassed the WHO allowable limits for fluoride ion concentration in drinking water (1.5 mg/L). This result suggests that inhabitants may be faced with the risk of fluorosis, from consumption of this water. On the other hand, dental caries may be a risk when the fluoride content in drinking water falls below 0.5 mg/L. In line with this, it was also discovered that 14% of the samples had fluoride concentrations less than 0.5 mg/L, while 52.4% had fluoride concentration ranging from 0.5 to 1.5 mg/L which is the adequate dose for the development of healthy bones and teeth.

3.1 Water Quality Assessment

The hydrogeochemical facies of the region was characterized using Piper diagrams. The Piper plot consists of two triangles and a diamond-shaped diagram (Fig. 2a). The two triangles represent the plot of anion and cations, while the diamond-shaped plot shows the combination of the anion and cation fields. The concept of hydrogeochemical facies is derived from regions that contain identifiable characteristics of different anion and cation concentrations within the diamond-shape diagram.

From the Piper plot, it was observed that Na^+ and K^+ were the dominant cations. Nevertheless, Ca^{2+} and Mg^{2+} also contributed to the overall classification. In the anion region, HCO_3^- and SO_4^{2-} dominated the facies. With Na^+ dominating over other cations, it suggests the occurrence of ion

exchange activities as a result of rock weathering. However, the dominance of HCO_3^- and SO_4^{2-} also reveals ion contributions from silica weathering (Achary et al., 2016; Giridharan et al., 2009). In corroboration, Aghazadeh et al., (2016) and Barzegar et al., (2017) remarked that if the ratio of Ca^+/Mg^{2+} ranges from 0.6 to any value greater than 2, it signifies the dissolution of dolomite and silicate rock materials in groundwater. After estimating the ratio of Ca^+/Mg^{2+}, the majority of the water samples (71.43%) produced ratios ranging from 0.65 to 5.24, indicating the interference of silicate and dolomitic constituents in the groundwater system. However, the general classification of all samples suggests 9% Ca–Mg–Cl–SO_4, 24% Ca–Mg–HCO_3, 29% Na–K–Cl–SO_4, and 38% Na–K–HCO_3 water types.

3.2 Multivariate statistical investigation

Pearson's correlation analysis calculations have been deployed by different researchers in water quality evaluation to obtain the controlling ion in the hydrochemical process (P. C. Emenike et al., 2017; Giridharan et al., 2009; Li et al., 2013; Rasool et al., 2016). The data obtained from the CA served as a proportional measure to represent the association of one variable with the other. This study examined the interrelationships between the water quality parameters. In cases where the correlation coefficient (*r*) was less than 0.3, the relationship is regarded as a weak correlation. If *r* value ranges from 0.3 to 0.7, the relationship was considered moderate, and when the value of *r* was greater than 0.7 it is considered strong (Salifu et al., 2012; Xiao et al., 2015)

From the physico–chemical correlationships (Table S3 in the supplementary material), fluoride was seen to have a strong positive relationship with TDS (r = 0.79), SO_4^{2-} (r = 0.35), EC (r = 0.40), and Mg^{2+} (r = 0.34). Similarly, fluoride showed a weak but positive relationship with Ca^{2+} (r = 0.23). Other relationships that can be drawn from the analysis show that there was a moderate positive correlation between pH and TDS (r = 0.41), EC (r = 0.38), Alka (r = 0.46), Fe^{2+} (r =

290 0.50) and Cl^- (r = 0.34). Similarly, a moderate negative relationship existed between pH – K^+ (r
291 = –0.45) and pH – SiO_2 (r = –0.50).

292 In addition, the statistical analysis showed a strong positive inter–relationship between TDS – EC
293 (r = 0.91), TDS – Mg^{2+} (r = 0.78), TDS – SO_4^{2-} (r = 0.78). Moderate positive inter-relationships
294 exist between TDS – Ca^{2+} (r = 0.44), TDS – Na^+ (r = 0.49), TDS – HCO_3^- (r = 0.66) and TDS –
295 Cl^- (r = 0.59). However, a moderate negative relationship existed between TDS – SiO_2 (r = –0.56)

296 The statistical study identified moderate positive correlations for EC – K^+ (r = 0.34), EC – Ca^+ (r
297 = 0.64), EC – Mg^{2+} (r = 0.57), EC – Na^+ (r = 0.34), EC – HCO_3^- (r = 0.65), EC – Cl^- (r = 0.54)
298 and a strong positive correlation for EC – SO_4^{2-} (r = 0.79). Other moderate positive relationships
299 were seen in the results between Ca^{2+}– SO_4^{2-} (r = 0.66), Mg^{2+}– SO_4^{2-} (r = 0.66), SO_4^{2-} – HCO_3^-
300 (r = 0.65) while a strong positive relationship existed between Na^+ – Cl^- (r = 0.84) and SO_4^{2-} –
301 Cl^- (r = 0.76).

302 In comparison, the average concentration of F^- obtained from this study were higher than the mean
303 values obtained from the Birbhum district (Das and Nag, 2017), the Northern region of Ghana
304 (Salifu et al., 2012), the Varahi river basin (Ravikumar and Somashekar, 2017) and less than that
305 seen in the South of India (Viswanathan et al., 2009), Northern Tanzania (Shen and Schäfer, 2015),
306 and Umarkot sub-district, Pakistan (Rafique et al., 2015). It is important to note that the
307 concentration of fluoride in water tends to increase when the TDS value is high (Rafique et al.,
308 2009). Also looking at fluoride concentration with respect to pH, Guo *et al.,* (2012) noted that at
309 pH 5.0 – 6.5, the solubility of fluoride is at its lowest. Saxena and Ahmed (2001) observed that the
310 dissolution of fluoride occurs between pH 6 to pH 9. In this study, the pH of the water was 6.76,
311 a condition where the solubility of fluoride was expected to be high. The presence of fluoride in
312 water may be as a result of dissolution from quartzite and paleosols near the underlying

groundwater table (Amanambu and Egbinola, 2015; Chuah et al., 2016; Xiao et al., 2015) and it could be the reason for the moderate positive relationship existing between F^- and pH (r = 0.53). Also, in the correlation matrix **(Table S3** in the supplementary material**)**, fluoride ions had a strong positive relationship with TDS (r = 0.79) and a moderate relationship with EC (r = 0.40), Mg^{2+} (r = 0.34), SO_4^{2-} (r = 0.35), and Alkalinity (r = 0.40). The moderate positive correlation between F^- and alkalinity confirms the penetration of fluoride enriched minerals from an alkaline environment into the groundwater systems.

3.3 Principal component analysis (PCA) of water parameters

PCA is a widely used technique that explains the variation in a large dataset where the variables are interconnected. It extracts information from the data by reducing dimensionality through variable reduction as well as qualitative structure–activity relationship. The interpretation of principal components (PC) can be linked to the hydrochemical characteristics of water samples. The loadings on each PC explains the water-rock interactions in the geochemical process (Fig. 2b). The most important parameters controlling the hydrochemical processes are mainly found in PC1, having the highest Eigenvalue. In this study, PCA was performed, and from the scree plot and explained variance **(Fig. 2c)**, the PCA produced two components PC1 and PC2. PC1 contributed 90.09% of the total variance. Within the total variance explained by PC1, strong positive loadings were recorded in EC (r = 0.9889), TDS (r = 0.9599), SO_4^{2-} (r = 0.8335), HCO_3^- (r = 0.6780), Mg^{2+} (r = 0.6597), Ca^{2+} (r = 0.5999), Cl^- (r = 0.6003) and high negative loadings in SiO_2 (r = –0.5243). Considering the loadings mentioned above, PC1 can be regarded as a salinity index. PC2 explains 4.61% of the total variance with high negative loadings from Na^+ (r = –0.7052), Cl^- (r = –0.6128) and Mg^{2+} (r = –0.4519).

The positive loading of pH on PC1 suggests an interaction between silicate and carbonate materials. It also suggests dissolution processes involving gneiss and granite rock formations, which may be the reason for the dominance of Ca^{2+}, HCO_3^- and Mg^{2+}. These results support earlier works of Oke and Tijani (2012) and Ufoegbune et al. (2009), where it was observed that the geological formation of Abeokuta is mainly sedimentary and basement rocks of Paleozoic age enriched with MgO, SiO_2, K_2O, and CaO. Concurrently, an increase in pH was seen when protons were consumed during the dissolution process. This may also be the reason why the pH had a positive loading on PC1. The varying pH values observed in these results is an indication that the groundwater system must have been recharged with water that passed through rock formations with a significant amount of salt.

Concerning the dominating parameters in PC1, the high positive loadings may be as a result of dissolved porphyritic biotite granite and porphyroblastic gneiss. This result supports the findings of Oke and Tijani, (2012) and confirms the presence of these rock formations contributing to the increase in TDS, EC, HCO_3^-, Mg^{2+}, Ca^{2+}, and Cl^- loadings recorded in PC1. The presence of these components represents a combined process comprising chemical weathering and mineralogical dissolution of rock constituents.

The score plot (Fig. 2d) accounts for the information and similarities within the samples analyzed and helps to understand the spatial distribution of the samples. From Fig. 2d, samples R14, R19, R12, R18, R13, and R21, were observed to have characteristics of similar water quality. Interestingly, within the samples, the total variance explained by PC1 (90.09%) received a high loading from R12 (r = –737.455), R13 (r = – 827.740), R14 (r = –1146.254), R18 (r = –705.517), R21 (r = –805.961), R19 (r = –962.665) and R4 (r = –1053.184). Also from the results, the contribution from PC2 to the score plot emanates from R11 (r = –697.740). Similar contributions

359 can be seen from R6 and R7 with scores r = –258.166 and r = –310.154 respectively. The influence
360 plot showing how the different samples contribute to the components is presented in Fig. 2 (e).

361

362 **"Fig. 2(a–e) is about here."**

363 ### 3.4 Geostatistical modeling

364 After analyzing the water quality parameters with the EBK model, the spatial distribution maps of
365 all examined parameters are presented in Fig. 3(a–p). Four important factors revealed by the
366 prediction statistics of the EBK model explained the authenticity of the model. They include, root
367 mean squared predicted (RMSP), average standard error (ASE), mean standardized (MS), and root
368 mean square standardized (RMSS) statistics. The results obtained established the potency of the
369 different distribution of the EBK prediction model, one of which is associated with the closely
370 related values of the ASE and RMSP values. From the results presented in Table 1, it could be
371 seen that the ASE and RMSP are near, signifying the validity of the prediction. It is also important
372 to note that RMSS should be close to 1 to explain the extent of estimation of the EBK model
373 further. If the RMSS is > 1, the prediction is said to be overestimated and if < 1, the prediction is
374 underestimated. Therefore, the results revealed the RMSS value of pH as 0.968; TDS, 0.967; EC,
375 0.966; F^{-}, 0.979; Fe^{2+}, 0.999; Mn, 0.998; Ca^{2+}, 0.966, K^{+}, 0.993; Mg^{2+}, 0.970; Na^{+}, 0.975; SO_4^{2-},
376 0.910, SiO_2, 0.981, HCO_3^{-}, 0.986; Cl^{-}, 0.972; NO_3^{-}, 0.959 and CO_3^{2-}, 0.986—authenticating the
377 variability prediction since all values are close to 1. Furthermore, the MS value should be close to
378 0 for the prediction to be valid. This is confirmed, as the MS values in this study ranged from 0.072
379 to 0.092.

380

"Fig. 3 is about here."

In the semivariograms of the examined water parameters (Fig 4a–4p), the blue crosses represent the semivariance obtained from empirical transformation, the bold red lines represent the intermediate distribution, the faint blue plots describe the different semivariograms distribution, the faded red plots (dotted) displays the 25^{th} and 75^{th} percentile respectively, and the densely-packed blue lines indicate the series of semivariogram passing through the zone. To examine the stability of the model, the majority of the blue crosses should remain within the confines of the faint blue spectrum (Magesh et al., 2017; Tenebe et al., 2018, 2017). This was the case observed for all of the water parameters studied.

"Fig. 4 is about here."

3.5 Fluoride risk assessment

Ingestion risk

The hazard quotient (HQ) associated with fluoride concentration on different age classification (6–12 months, 6–11 years, 11–16 years, 16–18 years, 18–21 years, ≥ 21 years and > 65 years) was evaluated, integrating the indicators obtained from the US EPA Exposure Factor Handbook (US EPA, 2011) as well as the fluoride concentration in the groundwater from the sampled locations. The non-carcinogenic risks were computed, and the age classification with high values of HQ_{IN} are presented in Table 2 and Table S4 **(**in the supplementary material). The mean CDD_{IN} values ranged from 53.846–201.832 μg/(kg day) for age 6–12 months. The mean CDD_{IN} values for age 6 – 11years varied from 22.075 to 82.744 μg/(kg day). The mean CDD_{IN} values for age 11–16 years varied from 16.454 to 61.674 μg/(kg day). Similarly, the mean CDDs for age classifications 16–18 years, 18–21 years, ≥ 21 years and > 65 years varied from 12.902 to 48.362 μg/(kg day), 16.962

to 63.577 µg/(kg day), 18.282 to 68.525 µg/(kg day) and 16.721 to 62.672 µg/(kg day) respectively, in relation to water ingestion (Table S4 in the supplementary material). After computing the HQ_{IN}, further observation revealed that 90.5% of the sampled locations were above the permissible level because for age classification 6–12 months, because the HQ_{IN} values were greater than 1. The implication is that the children in the 6 – 12 months age range are more likely to suffer from health complications associated with consumption of water laden with a high concentration of fluoride (Ding et al., 2011). Also, children within the age range of 6 – 11 years were also expected to be affected as 42.9% of the sampled region exceeded the HQ_{IN} limits.

Furthermore, 4.8% of the modelled exposures surpassed the acceptable HQ_{IN} limit for age classes 11–16 years and > 65 years, 9.5% of the sampled region exceeded the permissible HQ_{IN} level for age 18–21years, 19.05% of the study location was beyond the allowable HQ_{IN} limits for age ≥ 21 years. No risk was modelled for age class 16 – 18 years in the region. Despite the fact that age classifications 11–16 years, 18–21 years, ≥ 21 years and > 65 years had overall low percentage risks, it was observed that at specific locations (R16 and R17), much higher risks were seen across all age classifications noted above. This identifies explicitly the need for more extensive water treatment at locations R16 and R17 since the region is prone to high fluoride concentrations in the tap water.

Dermal risk

Table S5 (in the supplementary material) and Table 3 shows the dermal CDD_{DE} and HQ_{DE} values respectively. The dermal mean CDD_{DE} varied from 0.125 µg/(kg day) to 0.470 µg/(kg day) and 0.111 µg/(kg day) to 0.418 µg/(kg day) for age classification 6–12months and 6–11years

respectively. Correspondingly, the CDD_{DE} for age class 11–16years and 16–18years varied from 0.073 μg/(kg day) to 0.275 μg/(kg day) and 0.068 μg/(kg day) to 0.253 μg/(kg day). After computing the chronic daily dose as a result of dermal exposure (CDD_{DE}) for age class 18–21years, ≥ 21years and > 65years, the values ranged from 0.096 μg/(kg day) to 0.362 μg/(kg day), 0.083 μg/(kg day) to 0.310 μg/(kg day) and 0.081 μg/(kg day) to 0.303 μg/(kg day) respectively. The HQ_{DE} obtained in this study suggests low risk via dermal exposure for all age classification (HQ_{DE} – 1). However, the HQ_{DE} values presented in the results establish the need for attention, especially now that the aftermath of climate change poses significant change to groundwater resources. Significant health impediments, in addition to dental and skeletal fluorosis, such as neurotoxicological diseases (Choi et al., 2012), skeletal cancer (Bassin et al., 2006), hypertension (Sun et al., 2013), carotid artery atherosclerosis (Liu et al., 2014) and cardiac dysfunction (Varol et al., 2010) or possibly death may occur due to long-term exposure of fluoride. In other words, if the current trend of fluoride contamination occurs continuously in the study area, the devastating effect on lives will be enormous since survival possibilities rely on the availability of clean (non-contaminated) drinking water (Emenike et al., 2017).

4. Conclusion

The different ionic concentration of groundwater, fluoride concentration and its associated health risk were investigated in Abeokuta, Nigeria. Considering the water quality of the sixty-three taps analyzed, only about 5% were found to be fit for consumption (based on the percentage violation of TDS from WHO standards). According to the WHO guidelines for drinking water quality (WHO, 2011), parameters such as TDS, EC, F^- and SO_4^{2-} exceeded allowable limits. The findings in this study revealed that chemical weathering and dissolved rock compounds were responsible for the presence of F^-, HCO_3^-, Mg^{2+}, high TDS and EC in the groundwater. The CA and PCA

techniques validate the interrelationship of elements and its potential source as it reveals a statistical relationship between F^- and TDS, EC, Alkalinity, $Mg^{2+,}$ and SO_4^{2-}. The classification of the water types could be ranked as Na–K–HCO_3– Na–K–Cl–SO_4–Ca–Mg–HCO_3–Ca–Mg–Cl–SO_4 type. The information obtained from the EBK model validates the prediction of the spatial distribution as the standardized RMS values were close to 1 Also, the values of ASE and RMSP were almost equal which validates the authenticity of the spatial model. Further observation revealed that 90.5% of the modelled exposure was above the standard level for age classification 6–12 months. 4.8% of the modelled exposures exceeded the acceptable HQ_{IN} limit for age classes 11–16 years and > 65 years, 9.5% of the sampled region exceeded the permissible HQ_{IN} level for age 18–21years and 19.05% of the study location was beyond the allowable HQ_{IN} limits for age ≥ 21 years.

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References

Achary, M.S., Panigrahi, S., Satpathy, K.K., Prabhu, R.K., Panigrahy, R.C., 2016. Health risk assessment and seasonal distribution of dissolved trace metals in surface waters of Kalpakkam, southwest coast of Bay of Bengal. Reg. Stud. Mar. Sci. 6, 96–108. https://doi.org/10.1016/j.rsma.2016.03.017

Adekunle, A.A., Adedayo O. Badejo, Abiola O. Oyerinde, 2013. Pollution Studies on Ground Water Contamination : Water Quality of Abeokuta , Ogun State , South West Nigeria. J. Environ. Earth Sci. 3, 161–166.

Aghazadeh, N., Chitsazan, M., Golestan, Y., 2016. Hydrochemistry and quality assessment of groundwater in the Ardabil area , Iran. Appl. Water Sci. 7, 3599–3616. https://doi.org/10.1007/s13201-016-0498-9

Amanambu, A.C., Egbinola, C.N., 2015. Geogenic contamination of groundwater in shallow aquifers in Ibadan, south-west Nigeria. Manag. Environ. Qual. An Int. J. 26, 327–341. https://doi.org/10.1108/MEQ-12-2013-0135

Amini, M., Abbaspour, K.C., Berg, M., Winkel, L., Hug, S.J., Hoehn, E., Yang, H., Johnson, C.A., 2008. Statistical modeling of global geogenic arsenic contamination in groundwater. Env. Sci Technol 42, 3669–3675.

Barzegar, R., Asghari Moghaddam, A., Tziritis, E., 2017. Hydrogeochemical features of groundwater resources in Tabriz plain, northwest of Iran. Appl. Water Sci. https://doi.org/10.1007/s13201-017-0550-4

Bassin, E.B., Wypij, D., Davis, R.B., Mittleman, M.A., 2006. Age-specific fluoride exposure in drinking water and osteosarcoma (United States). Cancer Causes Control 17, 421–428. https://doi.org/10.1007/s10552-005-0500-6

Battaleb-Looie, S., Moore, F., Malde, M.K., Jacks, G., 2013. Fluoride in groundwater, dates and wheat: Estimated exposure dose in the population of Bushehr, Iran. J. Food Compos. Anal. 29, 94–99. https://doi.org/10.1016/j.jfca.2012.08.001

Bhatnagar, A., Kumar, E., Sillanpää, M., 2011. Fluoride removal from water by adsorption-A review. Chem. Eng. J. 171, 811–840. https://doi.org/10.1016/j.cej.2011.05.028

Chabukdhara, M., Gupta, S.K., Kotecha, Y., Nema, A.K., 2017. Groundwater quality in Ghaziabad district, Uttar Pradesh, India: Multivariate and health risk assessment. Chemosphere 179, 167–178. https://doi.org/10.1016/j.chemosphere.2017.03.086

Choi, A.L., Sun, G., Zhang, Y., Grandjean, P., 2012. Developmental Fluoride Neurotoxicity: A Systematic Review and Meta-Analysis. Environ. Health Perspect. 1362, 1362–1368.

Chuah, C.J., Lye, H.R., Ziegler, A.D., Wood, S.H., Kongpun, C., Rajchagool, S., 2016. Fluoride: A naturally-occurring health hazard in drinking-water resources of Northern Thailand. Sci. Total Environ. 545–546, 266–279. https://doi.org/10.1016/j.scitotenv.2015.12.069

Craig, L., Lutz, A., Berry, K.A., Yang, W., 2015. Recommendations for fluoride limits in drinking water based on estimated daily fluoride intake in the Upper East Region, Ghana. Sci. Total

503 Environ. 532, 127–137. https://doi.org/10.1016/j.scitotenv.2015.05.126

504 Das, S., Nag, S.K., 2017. Application of multivariate statistical analysis concepts for assessment
505 of hydrogeochemistry of groundwater—a study in Suri I and II blocks of Birbhum District,
506 West Bengal, India. Appl. Water Sci. 7, 873–888. https://doi.org/10.1007/s13201-015-0299-
507 6

508 Dehbandi, R., Moore, F., Keshavarzi, B., 2018. Geochemical sources, hydrogeochemical behavior,
509 and health risk assessment of fluoride in an endemic fluorosis area, central Iran. Chemosphere
510 193, 763–776. https://doi.org/10.1016/j.chemosphere.2017.11.021

511 Ding, Y., YanhuiGao, Sun, H., Han, H., Wang, W., Ji, X., Liu, X., Sun, D., 2011. The relationships
512 between low levels of urine fluoride on children's intelligence, dental fluorosis in endemic
513 fluorosis areas in Hulunbuir, Inner Mongolia, China. J. Hazard. Mater. 186, 1942–1946.
514 https://doi.org/10.1016/j.jhazmat.2010.12.097

515 Emenike, C.P., Tenebe, I.T., Omole, D.O., Ngene, B.U., Oniemayin, B.I., Maxwell, O., Onoka,
516 B.I., 2017. Accessing safe drinking water in sub-Saharan Africa: Issues and challenges in
517 South–West Nigeria. Sustain. Cities Soc. 30, 263–272.
518 https://doi.org/10.1016/j.scs.2017.01.005

519 Emenike, P.C., Omole, D.O., Ngene, B.U., Tenebe, I.T., 2016. Potentiality of agricultural
520 adsorbent for the sequestering of metal ions from wastewater. Glob. J. Environ. Sci. Manag.
521 2, 411–442. https://doi.org/10.22034/gjesm.2016.02.04.010

522 Emenike, P.C., Tenebe, T.I., Omeje, M., Osinubi, D.S., 2017. Health risk assessment of heavy
523 metal variability in sachet water sold in Ado-Odo Ota, South-Western Nigeria. Environ.
524 Monit. Assess. 189, 1–16. https://doi.org/10.1007/s10661-017-6180-3

525 EPA, 2010. Fluoride: Exposure and Relative Source Contribution Analysis. Washington, D.C.

526 Fawell, J., Bailey, K., Chilton, J., Dahi, E., Fewtrell, L., Magara, Y., 2006. Fluoride in Drinking-
527 water, World Health Organization. https://doi.org/10.1007/BF01783490

528 Freire, I.R., Pessan, J.P., Amaral, J.G., Martinhon, C.C.R., Cunha, R.F., Delbem, A.C.B., 2016.
529 Anticaries effect of low-fluoride dentifrices with phosphates in children: A randomized,
530 controlled trial. J. Dent. 50, 37–42. https://doi.org/10.1016/j.jdent.2016.04.013

531 Giridharan, L., Venugopal, T., Jayaprakash, M., 2009. Assessment of water quality using
532 chemometric tools: A case study of river cooum, South India. Arch. Environ. Contam.
533 Toxicol. 56, 654–669. https://doi.org/10.1007/s00244-009-9310-2

534 Guissouma, W., Hakami, O., Al-Rajab, A.J., Tarhouni, J., 2017. Risk assessment of fluoride
535 exposure in drinking water of Tunisia. Chemosphere 177, 102–108.
536 https://doi.org/10.1016/j.chemosphere.2017.03.011

537 Guo, H., Zhang, Y., Xing, L., Jia, Y., 2012. Spatial variation in arsenic and fluoride concentrations
538 of shallow groundwater from the town of Shahai in the Hetao basin, Inner Mongolia. Appl.
539 Geochemistry 27, 2187–2196. https://doi.org/10.1016/j.apgeochem.2012.01.016

540 Irigoyen-Camacho, M.E., García Pérez, A., Mejía González, A., Huizar Alvarez, R., 2016.
541 Nutritional status and dental fluorosis among schoolchildren in communities with different

drinking water fluoride concentrations in a central region in Mexico. Sci. Total Environ. 541, 512–519. https://doi.org/10.1016/j.scitotenv.2015.09.085

Izquierdo-Vega, J.A., Sánchez-Gutiérrez, M., Del Razo, L.M., 2008. Decreased in vitro fertility in male rats exposed to fluoride-induced oxidative stress damage and mitochondrial transmembrane potential loss. Toxicol. Appl. Pharmacol. 230, 352–357. https://doi.org/10.1016/j.taap.2008.03.008

Jha, S.K., Mishra, V.K., Sharma, D.K., Damodaran, T., 2011. Fluoride in the environment and its metabolism in humans. Rev. Environ. Contam. Toxicol. 211, 121–142. https://doi.org/10.1007/978-1-4419-8011-3_4

John-Dewole, O.., 2012. Adverse Effects of Inadequate Water Supply on Human Health : a case study of Kajola Local Government in Oyo State , Nigeria By. Greener J. Med. Sci. 2, 115–119.

Kaoud, H., Kalifa, B., 2010. Effect of fluoride, cadmium and arsenic intoxication on brain and learning–memory ability in rats. Toxicol. Lett. 196, S53. https://doi.org/10.1016/j.toxlet.2010.03.212

Krivoruchko, K., 2011. Spatial Statistical Data Analysis for GIS Users Spatial Statistical Data Analysis for GIS Users, Esri Press, Rehlands, CA. Esri Press, Rehlands, CA.

Levy, M., Leclerc, B.S., 2012. Fluoride in drinking water and osteosarcoma incidence rates in the continental United States among children and adolescents. Cancer Epidemiol. 36, 83–88. https://doi.org/10.1016/j.canep.2011.11.008

Li, P., Qian, H., Wu, J., Zhang, Y., Zhang, H., 2013. Major Ion Chemistry of Shallow Groundwater in the Dongsheng Coalfield, Ordos Basin, China. Mine Water Environ. 32, 195–206. https://doi.org/10.1007/s10230-013-0234-8

Liu, H., Gao, Y., Sun, L., Li, M., Li, B., Sun, D., 2014. Assessment of relationship on excess fluoride intake from drinking water and carotid atherosclerosis development in adults in fluoride endemic areas, China. Int. J. Hyg. Environ. Health 217, 413–420. https://doi.org/10.1016/j.ijheh.2013.08.001

Magesh, N.S., Chandrasekar, N., Elango, L., 2017. Trace element concentrations in the groundwater of the Tamiraparani river basin, South India: Insights from human health risk and multivariate statistical techniques. Chemosphere 185, 468–479. https://doi.org/10.1016/j.chemosphere.2017.07.044

Maguire, A., 2014. ADA clinical recommendations on topical fluoride for caries prevention. Evid. Based. Dent. https://doi.org/10.1038/sj.ebd.6401019

McGrady, M.G., Ellwood, R.P., Srisilapanan, P., Korwanich, N., Worthington, H. V., Pretty, I.A., 2012. Dental fluorosis in populations from Chiang Mai, Thailand with different fluoride exposures - Paper 1: assessing fluorosis risk, predictors of fluorosis and the potential role of food preparation. BMC Oral Health 12, 1. https://doi.org/10.1186/1472-6831-12-16

Namkaew, M., Wiwatanadate, P., 2012. Association of fluoride in water for consumption and chronic pain of body parts in residents of San Kamphaeng district, Chiang Mai, Thailand. Trop. Med. Int. Heal. 17, 1171–1176. https://doi.org/10.1111/j.1365-3156.2012.03061.x

National Academy of Science, 2006. Fluoride in Drinking Water: A Scientific Review of EPA's Standards. The National Academies Press. Washington, D.C. https://doi.org/10.17226/11571

National Population Commission, 2010. Population Distribution by Sex,State,LGA&Senatorial District, 2006 Population and Housing Cencus.

Odjegba, E., Idowu, O., Ikenweiwe, N., Martins, O., Sadeeq, A., 2015. Public Perception of Potable Water Supply in Abeokuta. J. Appl. Sci. Environ. Manag. 19, 5–9.

Odjegba, E.E., Idowu, O.A., Oluwasanya, G.O., Ikenweiwe, N.B., Martins, O., 2014. Assessment of Water Demand and Seasonal Variation of Bacteriological Content of Public Water Systems in Abeokuta ... Int. J. Inst. Ecol. Environ. Stud. 2, 81–92.

Oke, S.A., Tijani, M.N., 2012. Impact of chemical weathering on groundwater chemistry of Abeokuta area ,. Elixir Pollut. 46, 8498–8503.

Rafique, T., Naseem, S., Ozsvath, D., Hussain, R., Bhanger, M.I., Usmani, T.H., 2015. Geochemical controls of high fluoride groundwater in Umarkot Sub-District, Thar Desert, Pakistan. Sci. Total Environ. 530–531, 271–278. https://doi.org/10.1016/j.scitotenv.2015.05.038

Rafique, T., Naseem, S., Usmani, T.H., Bashir, E., Khan, F.A., Bhanger, M.I., 2009. Geochemical factors controlling the occurrence of high fluoride groundwater in the Nagar Parkar area, Sindh, Pakistan. J. Hazard. Mater. 171, 424–430. https://doi.org/10.1016/j.jhazmat.2009.06.018

Rahaman, M.., 1976. Review of the basement geology of South-Western Nigeria. In: C.A. Kogbe (Ed.) Geology of Nigeria. Elizabethan Publishing Co;, Lagos, Nigeria.

Rasool, A., Xiao, T., Farooqi, A., Shafeeque, M., Masood, S., Ali, S., Fahad, S., Nasim, W., 2016. Arsenic and heavy metal contaminations in the tube well water of Punjab, Pakistan and risk assessment: A case study. Ecol. Eng. 95, 90–100. https://doi.org/10.1016/j.ecoleng.2016.06.034

Ravikumar, P., Somashekar, R.K., 2017. Principal component analysis and hydrochemical facies characterization to evaluate groundwater quality in Varahi river basin, Karnataka state, India. Appl. Water Sci. 7, 745–755. https://doi.org/10.1007/s13201-015-0287-x

Salifu, A., Petrusevski, B., Ghebremichael, K., Buamah, R., Amy, G., 2012. Multivariate statistical analysis for fluoride occurrence in groundwater in the Northern region of Ghana. J. Contam. Hydrol. 140–141, 34–44. https://doi.org/10.1016/j.jconhyd.2012.08.002

Samsonova, V.P., Blagoveshchenskii, Y.N., Meshalkina, Y.L., 2017. Use of empirical Bayesian kriging for revealing heterogeneities in the distribution of organic carbon on agricultural lands. Eurasian Soil Sci. 50, 305–311. https://doi.org/10.1134/S1064229317030103

Saxena, V.K., Ahmed, S., 2001. Dissolution of fluoride in groundwater: A water-rock interaction study. Environ. Geol. 40, 1084–1087. https://doi.org/10.1007/s002540100290

Shen, J., Schäfer, A.I., 2015. Factors affecting fluoride and natural organic matter (NOM) removal from natural waters in Tanzania by nanofiltration/reverse osmosis. Sci. Total Environ. 527–528, 520–529. https://doi.org/10.1016/j.scitotenv.2015.04.037

Shivaprakash, P.K., Ohri, K., Noorani, H., 2011. Relation between dental fluorosis and intelligence quotient in school children of Bagalkot district. J. Indian Soc. Pedod. Prev. Dent. 29, 117–120. https://doi.org/10.4103/0970-4388.84683

Singh, C.K., Kumari, R., Singh, N., Mallick, J., Mukherjee, S., 2013. Fluoride enrichment in aquifers of the Thar desert: Controlling factors and its geochemical modelling. Hydrol. Process. 27, 2462–2474. https://doi.org/10.1002/hyp.9247

Singh, C.K., Mukherjee, S., 2015. Aqueous geochemistry of fluoride enriched groundwater in arid part of Western India. Environ. Sci. Pollut. Res. 22, 2668–2678. https://doi.org/10.1007/s11356-014-3504-5

Smedley, P., Nkotagu, H., Pelig-Ba, K., MacDonald, A., Tyler-Whittle, R., Whitehead, E., Kinniburgh, D., 2002. Fluoride in groundwater from high-fluoride areas in Ghana and Tanzania. "Minimising fluoride in drinking water in problem aquifers." Br. Geol. Surv. Comm. Report, CR/02/316 77.

Subba Rao, N., 1993. Environmental impact of industrial effluents in groundwater regions of Visakhapatnam Industrial Complex. Indian J. Geol. 65, 35–43.

Subba Rao, N., Subrahmanyam, A., Babu Rao, G., 2013. Fluoride-bearing groundwater in Gummanampadu Sub-basin, Guntur District, Andhra Pradesh, India. Environ. Earth Sci. 70, 575–586. https://doi.org/10.1007/s12665-012-2142-9

Sun, L., Gao, Y., Liu, H., Zhang, W., Ding, Y., Li, B., Li, M., Sun, D., 2013. An assessment of the relationship between excess fluoride intake from drinking water and essential hypertension in adults residing in fluoride endemic areas. Sci. Total Environ. 443, 864–869. https://doi.org/10.1016/j.scitotenv.2012.11.021

Tenebe, I.T., Emenike, C.P., Ogbiye, A.S., Ngene, B.U., Omeje, M., Olatunji, O.O., 2018. A laboratory assessment of the effect of varying roughness on dissolved oxygen using error correction method. Cogent Eng. 0, 1–11. https://doi.org/10.1080/23311916.2018.1427191

Tenebe, I.T., Ogbiye, A., Omole, D.O., Emenike, P.C., 2016. Estimation of longitudinal dispersion co-efficient: A review. Cogent Eng. 3. https://doi.org/10.1080/23311916.2016.1216244

Tenebe, I.T., Ogbiye, A.S., Omole, D.O., Emenike, P.C., 2017. Modelling and sensitivity analysis of varying roughness effect on dispersion coefficient : a laboratory study 21298, 1–7. https://doi.org/10.5004/dwt.2017.21298

Tirkey, P., Bhattacharya, T., Chakraborty, S., Baraik, S., 2017. Assessment of groundwater quality and associated health risks: A case study of Ranchi city, Jharkhand, India. Groundw. Sustain. Dev. 5, 85–100. https://doi.org/10.1016/j.gsd.2017.05.002

Ufoegbune, G.., Lamidi, K.., Awomeso, J.., Eruola, A.., Idowu, O.., 2009. Hydro-geological characteristics and groundwater quality assessment in some selected communities of Abeokuta , Southwest Nigeria. J. Environ. Chem. Ecotoxicol. 1, 010–022.

US EPA, 2011. Exposure Factors Handbook: 2011 Edition, 2011 Editi. ed, U.S. Environmental Protection Agency. https://doi.org/EPA/600/R-090/052F

US EPA, 1989. Risk Assessment Guidance for Superfund Volume I Human Health Evaluation

660 Manual (Part A), US EPA. https://doi.org/EPA/540/1-89/002

661 Varol, E., Akcay, S., Ersoy, I.H., Ozaydin, M., Koroglu, B.K., Varol, S., 2010. Aortic elasticity is
662 impaired in patients with endemic fluorosis. Biol. Trace Elem. Res. 133, 121–127.
663 https://doi.org/10.1007/s12011-009-8578-4

664 Viswanathan, G., Jaswanth, A., Gopalakrishnan, S., Siva ilango, S., Aditya, G., 2009. Determining
665 the optimal fluoride concentration in drinking water for fluoride endemic regions in South
666 India. Sci. Total Environ. 407, 5298–5307. https://doi.org/10.1016/j.scitotenv.2009.06.028

667 Walia, T., Abu Fanas, S., Akbar, M., Eddin, J., Adnan, M., 2017. Estimation of fluoride
668 concentration in drinking water and common beverages in United Arab Emirates (UAE).
669 Saudi Dent. J. 29, 117–122. https://doi.org/10.1016/j.sdentj.2017.04.002

670 Wayland, K.G., Long, D.T., Hyndman, D.W., Pijanowski, B.C., Woodhams, S.M., Haack, S.K.,
671 2003. with Synoptic Sampling and R-Mode Factor Analysis. J. Environ. Qual. 32, 180–190.

672 WHO, 2011. WHO Guidelines for Drinking-water Quality., 4th Editio. ed, World Health
673 Organization. https://doi.org/10.1016/S1462-0758(00)00006-6

674 Xiao, J., Jin, Z., Zhang, F., 2015. Geochemical controls on fluoride concentrations in natural
675 waters from the middle Loess Plateau, China. J. Geochemical Explor. 159, 252–261.
676 https://doi.org/10.1016/j.gexplo.2015.09.018

677 Yang, M., Fei, Y., Ju, Y., Ma, Z., Li, H., 2012. Health risk assessment of groundwater pollution-
678 A case study of typical city in North China plain. J. Earth Sci. 23, 335–348.
679 https://doi.org/10.1007/s12583-012-0260-7

680 Yousefi, M., Ghoochani, M., Hossein Mahvi, A., 2018. Health risk assessment to fluoride in
681 drinking water of rural residents living in the Poldasht city, Northwest of Iran. Ecotoxicol.
682 Environ. Saf. 148, 426–430. https://doi.org/10.1016/j.ecoenv.2017.10.057

683 Zhai, Y., Zhao, X., Teng, Y., Li, X., Zhang, J., Wu, J., Zuo, R., 2017. Groundwater nitrate pollution
684 and human health risk assessment by using HHRA model in an agricultural area, NE China.
685 Ecotoxicol. Environ. Saf. 137, 130–142. https://doi.org/10.1016/j.ecoenv.2016.11.010

686 Zhang, L., Huang, D., Yang, J., Wei, X., Qin, J., Ou, S., Zhang, Z., Zou, Y., 2017. Probabilistic
687 risk assessment of Chinese residents' exposure to fluoride in improved drinking water in
688 endemic fluorosis areas. Environ. Pollut. 222, 118–125.
689 https://doi.org/10.1016/j.envpol.2016.12.074

690

691

692

Table 1: Prediction error information obtained from the empirical Bayesian kriging (EBK) model

	pH	TDS	EC	F^-	Fe^{2+}	Mn	Ca^{2+}	K+	Mg^{2+}	Na^+	SO_4^{2-}	SiO_2	HCO_3^-	Cl^-	NO_3^-	CO_3^{2-}
Mean	0.104	0.832	12.048	0.004	0.007	0.002	0.746	0.086	-4.519	1.348	3.378	-0.162	3.550	3.727	0.111	0.229
RMS Pred.	0.391	451.368	679.930	0.406	0.707	0.023	40.291	2.114	85.810	104.470	192.912	2.009	110.187	135.657	1.859	7.985
MS	0.022	0.005	0.022	0.009	0.019	0.092	0.017	0.041	-0.054	0.009	0.015	-0.072	0.030	0.021	0.555	0.029
RMSS	0.968	0.967	0.966	0.979	0.999	0.998	0.966	0.993	0.970	0.975	0.910	0.981	0.986	0.972	0.959	0.968
ASE	0.402	464.203	716.913	0.416	0.697	0.023	41.830	2.128	88.373	107.705	215.608	2.037	109.195	141.679	1.970	8.276

Table 2: Hazard quotients via ingestion pathway (**HQ_{IN}**) for different age classification

	HQ_{IN} VALUES						
REGION	**(6 – 12months)**	**(6 – 11years)**	**(11 – 16years)**	**(16 – 18years)**	**(18 – 21years)**	**(≥ 21years)**	**(> 65years)**
R1	*1.877	0.770	0.574	0.450	0.591	0.637	0.583
R2	*2.405	**0.986	0.735	0.576	0.758	0.817	0.747
R3	*2.833	*1.161	0.866	0.679	0.892	**0.962	0.880
R4	*3.016	*1.236	0.922	0.723	**0.950	*1.024	0.937
R5	*2.021	0.828	0.617	0.484	0.637	0.686	0.628
R6	*2.906	*1.191	0.888	0.696	0.915	*0.987	0.902
R7	*2.369	**0.971	0.724	0.568	0.746	0.804	0.736
R8	*1.215	0.498	0.371	0.291	0.383	0.412	0.377
R9	0.913	0.374	0.279	0.219	0.288	0.310	0.283
R10	*2.650	*1.086	0.810	0.635	0.835	0.900	0.823
R11	*1.484	0.608	0.453	0.355	0.467	0.504	0.461
R12	0.910	0.373	0.278	0.218	0.287	0.309	0.282
R13	*2.473	*1.014	0.756	0.592	0.779	0.839	0.768
R14	*2.259	0.926	0.690	0.541	0.712	0.767	0.701
R15	*3.034	*1.244	0.927	0.727	**0.956	*1.030	0.942
R16	*3.364	*1.379	*1.028	0.806	*1.060	*1.142	*1.045
R17	*3.217	*1.319	**0.983	0.771	*1.013	*1.092	**0.999
R18	*2.430	**0.996	0.742	0.582	0.765	0.825	0.755
R19	*2.295	0.941	0.701	0.550	0.723	0.779	0.713
R20	*2.814	*1.154	0.860	0.674	0.887	**0.956	0.874
R21	0.897	0.368	0.274	0.215	0.283	0.305	0.279

* HQ values exceeding the US EPA and WHO standards

** HQ values exceeding 0.95

Table 3: Hazard quotients via dermal pathway (**HQ_{DE}**) for different age classification

	HQ_{DE} VALUES						
REGION	**(6–12months)**	**(6–11years)**	**(11–16years)**	**(16–18years)**	**(18–21years)**	**(≥ 21years)**	**(> 65years)**
R1	0.004	0.004	0.003	0.002	0.003	0.003	0.003
R2	0.006	0.005	0.003	0.003	0.004	0.004	0.004
R3	0.007	0.006	0.004	0.004	0.005	0.004	0.004
R4	0.007	0.006	0.004	0.004	0.005	0.005	0.005
R5	0.005	0.004	0.003	0.003	0.004	0.003	0.003
R6	0.007	0.006	0.004	0.004	0.005	0.004	0.004
R7	0.006	0.005	0.003	0.003	0.004	0.004	0.004
R8	0.003	0.003	0.002	0.002	0.002	0.002	0.002
R9	0.002	0.002	0.001	0.001	0.002	0.001	0.001
R10	0.006	0.005	0.004	0.003	0.005	0.004	0.004
R11	0.003	0.003	0.002	0.002	0.003	0.002	0.002
R12	0.002	0.002	0.001	0.001	0.002	0.001	0.001
R13	0.006	0.005	0.003	0.003	0.004	0.004	0.004
R14	0.005	0.005	0.003	0.003	0.004	0.003	0.003
R15	0.007	0.006	0.004	0.004	0.005	0.005	0.005
R16	0.008	0.007	0.005	0.004	0.006	0.005	0.005
R17	0.007	0.007	0.004	0.004	0.006	0.005	0.005
R18	0.006	0.005	0.003	0.003	0.004	0.004	0.004
R19	0.005	0.005	0.003	0.003	0.004	0.004	0.003
R20	0.007	0.006	0.004	0.004	0.005	0.004	0.004
R21	0.002	0.002	0.001	0.001	0.002	0.001	0.001

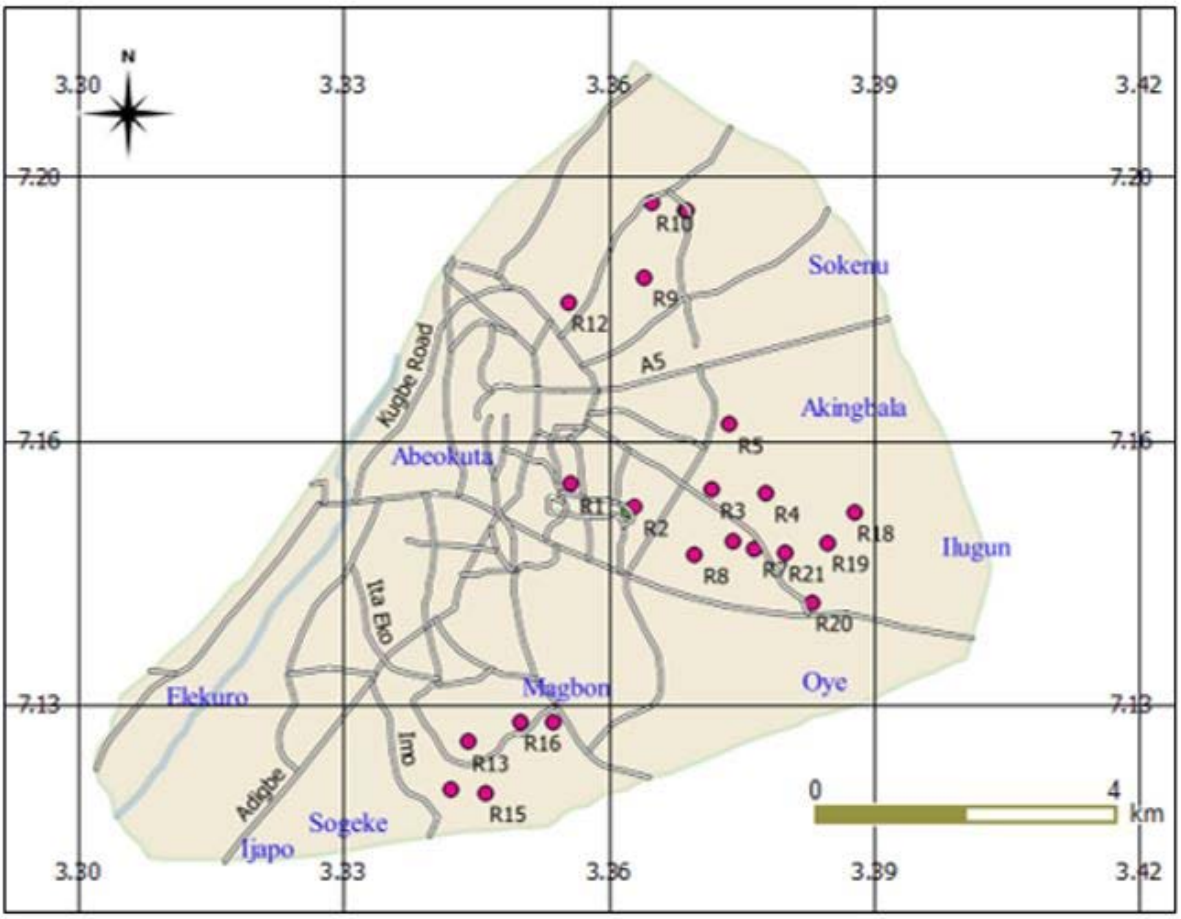
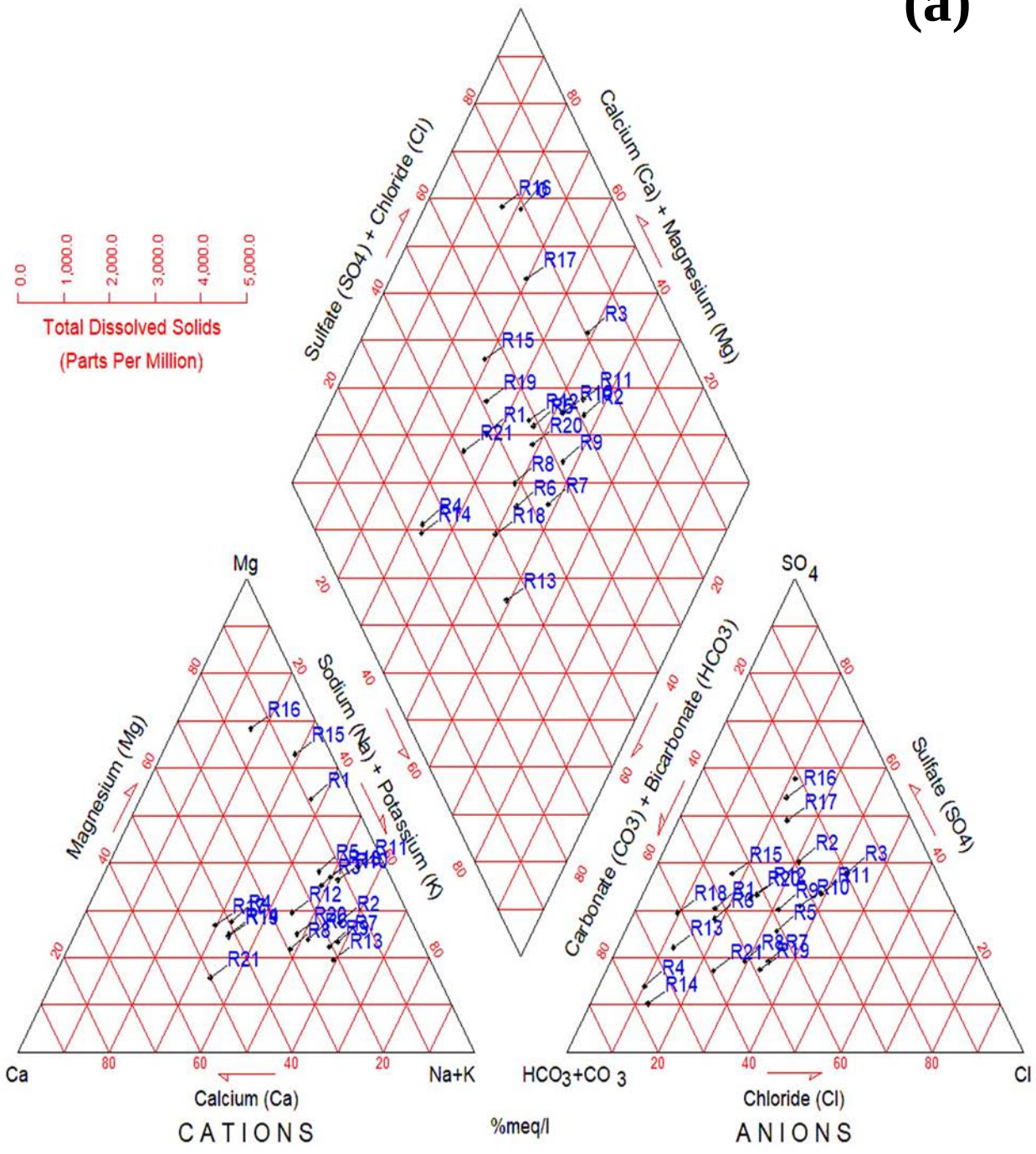
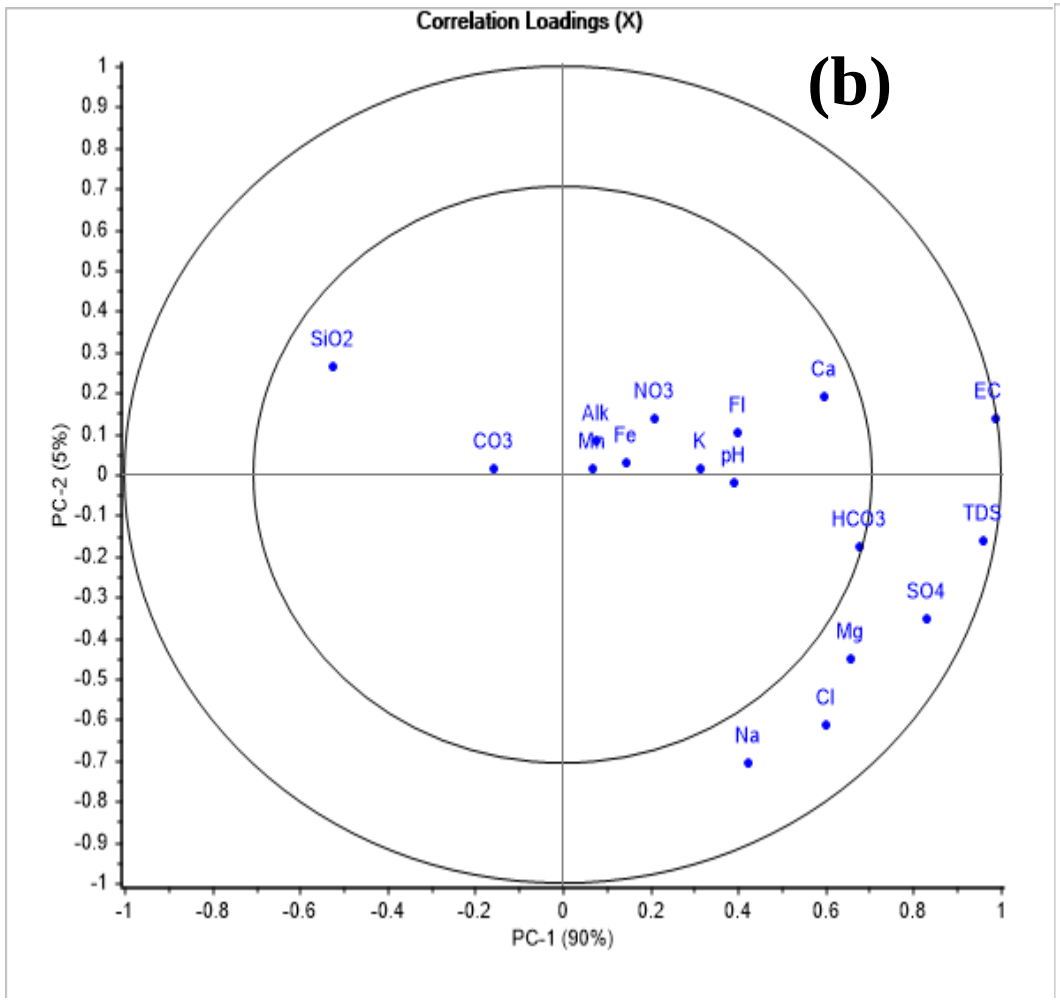


Fig. 1: Map of study area showing sampling locations

(a)


0.0
1,000.0
2,000.0
3,000.0
4,000.0
5,000.0
Total Dissolved Solids
(Parts Per Million)
Sulfate (SO4) + Chloride (Cl)
Calcium (Ca) + Magnesium (Mg)
Sodium (Na) + Potassium (K)
Carbonate (CO3) + Bicarbonate (HCO3)
Mg
Magnesium (Mg)
Ca
Calcium (Ca)
Na+K
CATIONS
%meq/l
SO4
Sulfate (SO4)
HCO3+CO3
Chloride (Cl)
Cl
ANIONS
R1
R2
R3
R4
R5
R6
R7
R8
R9
R10
R11
R12
R13
R14
R15
R16
R17
R18
R19
R20
R21



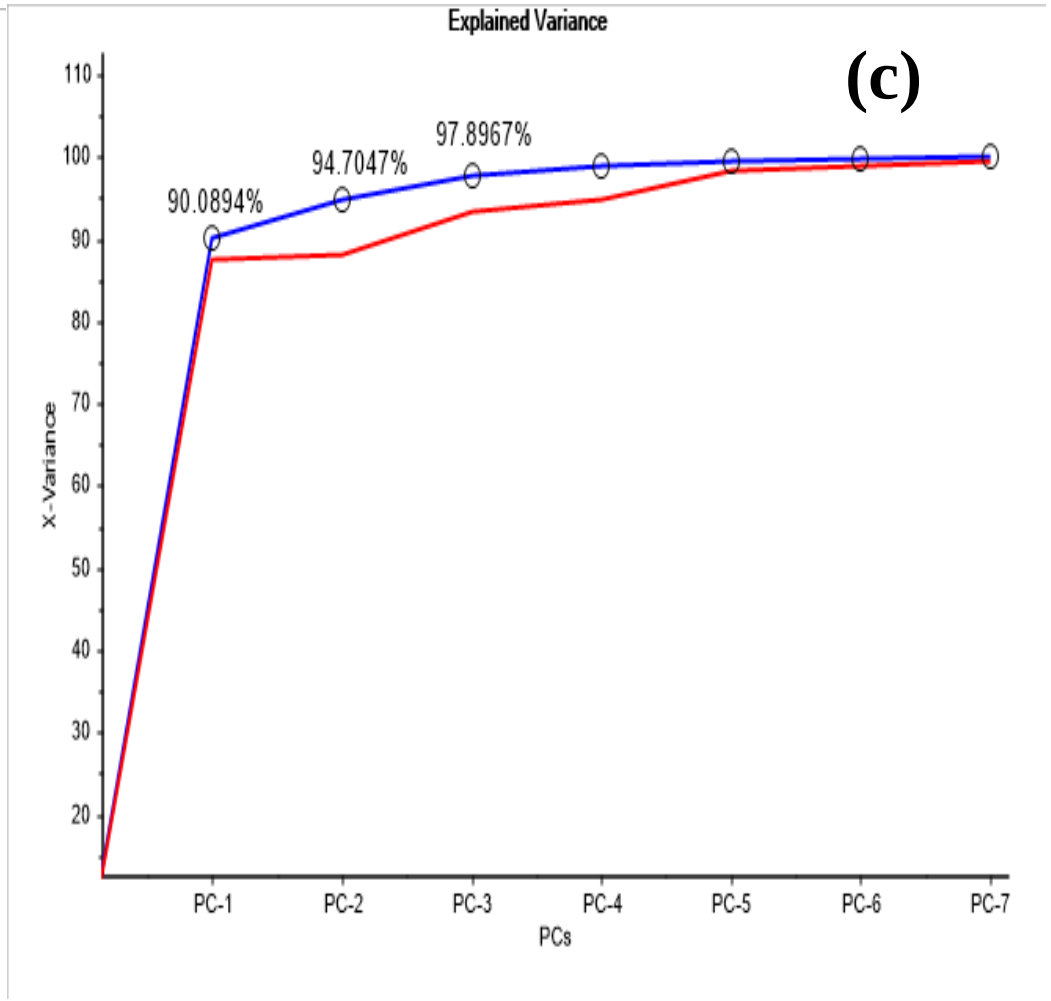


Fig. 2(a–c): (a) Characterization of hydrochemical facies with piper plot (b) PCA plot showing the loadings of each water quality parameter (c) Scree plot showing the percentage contribution of each principal component

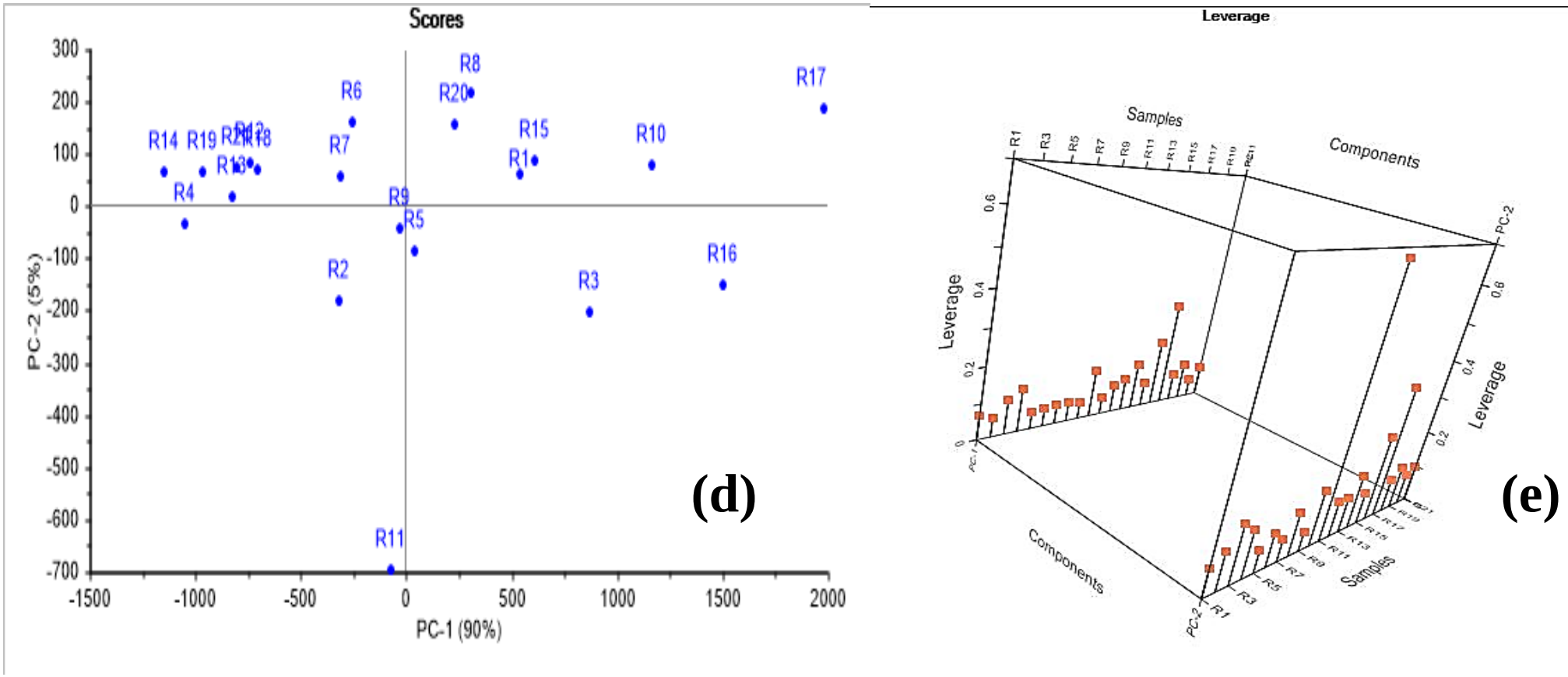
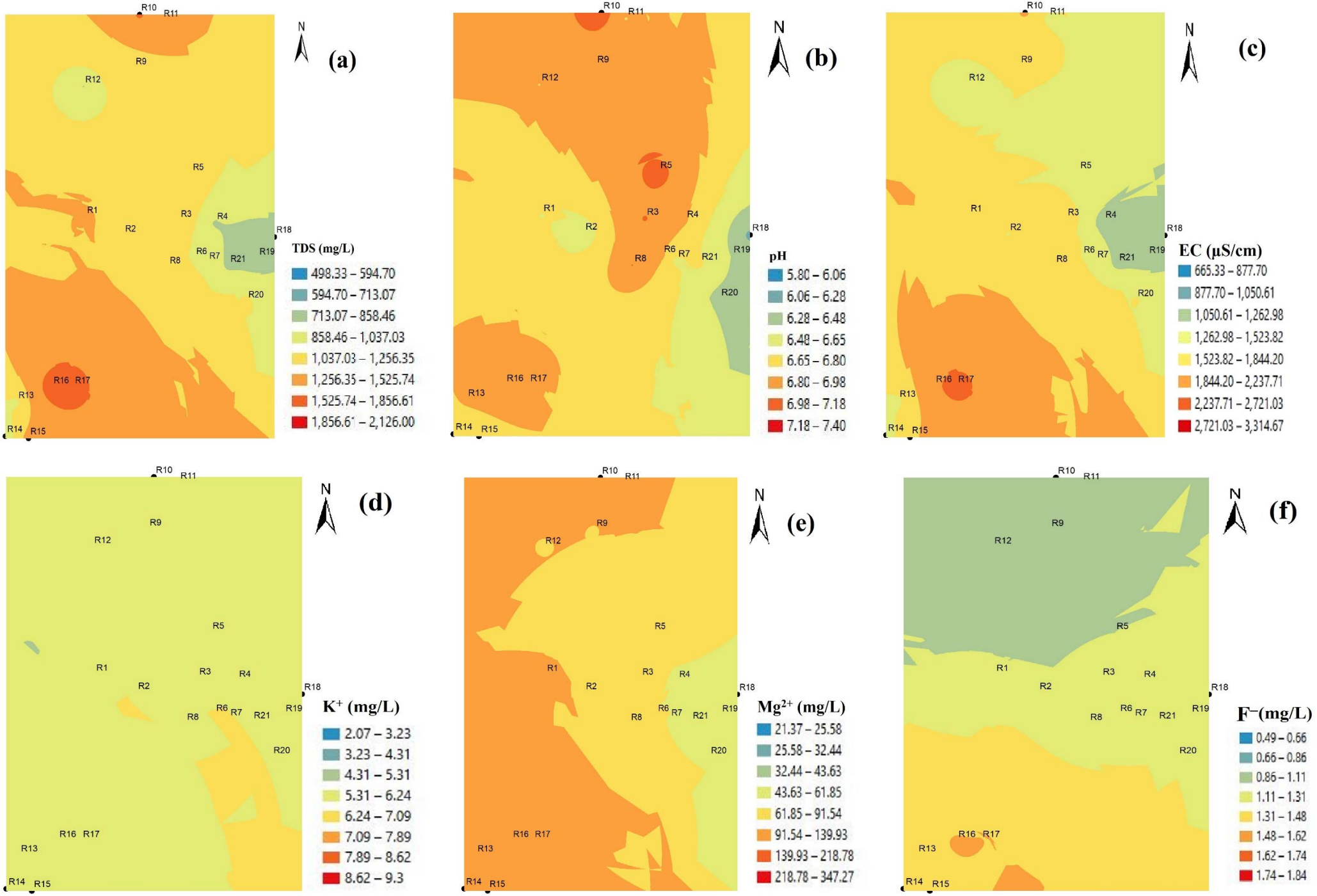
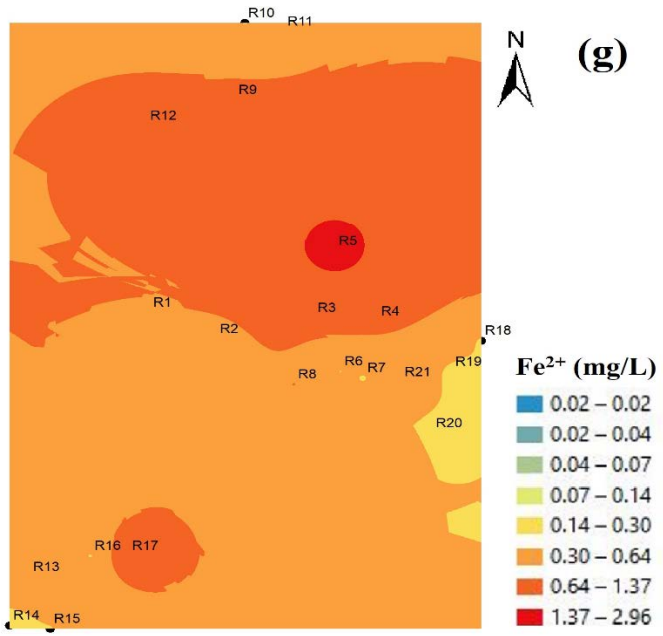
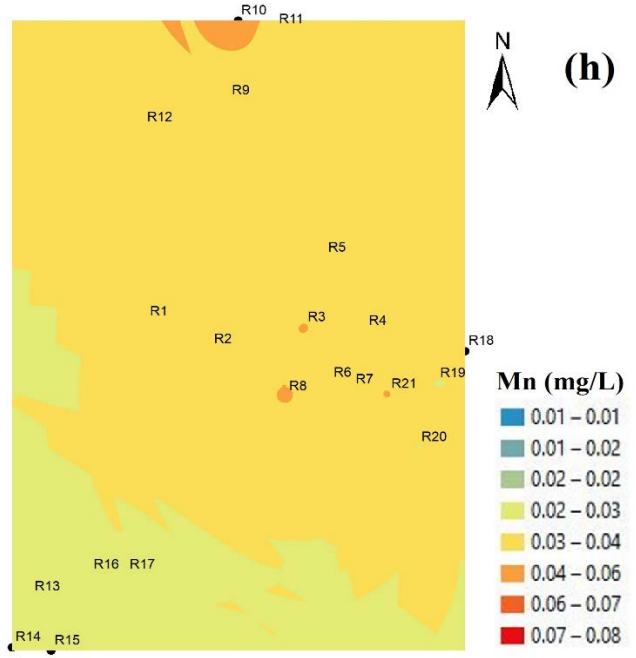
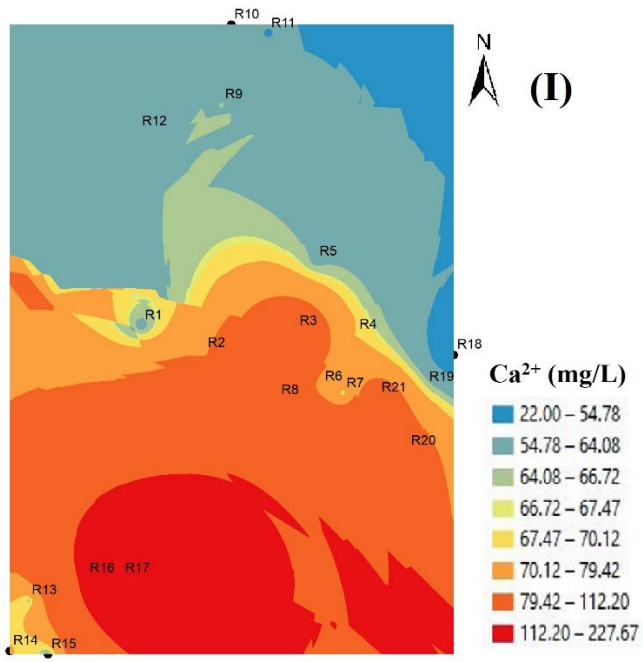


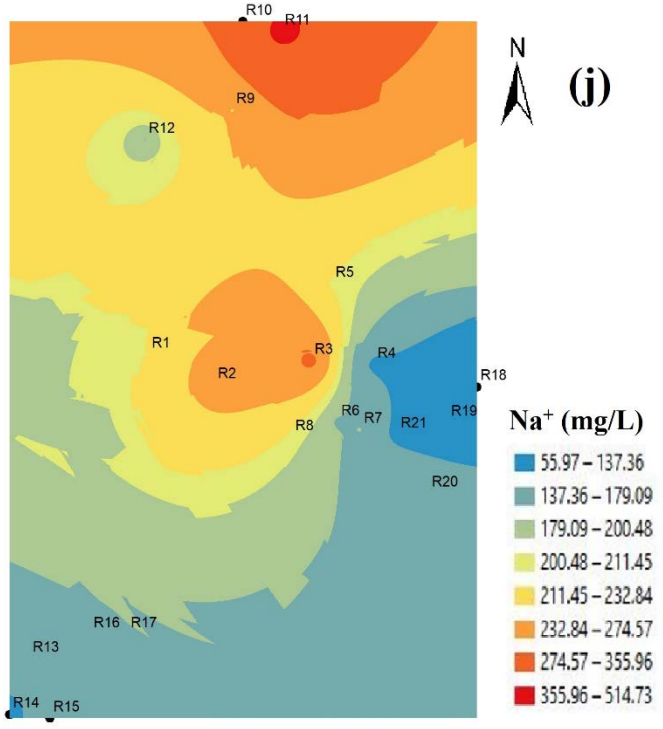
Fig. 2(d–e): (d) Score plot of the water samples (e) Influence plot showing the contribution of the water sample to the principal components

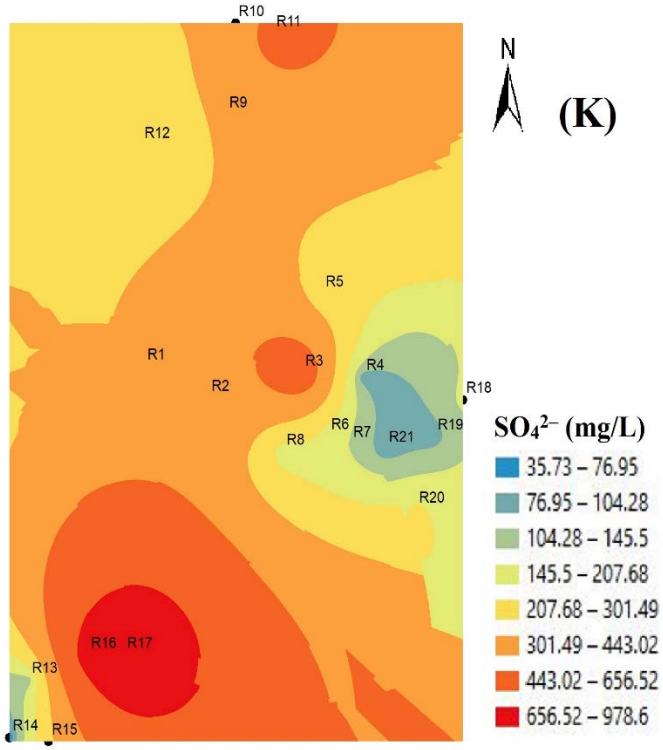

(a)
TDS (mg/L)
498.33 – 594.70
594.70 – 713.07
713.07 – 858.46
858.46 – 1,037.03
1,037.03 – 1,256.35
1,256.35 – 1,525.74
1,525.74 – 1,856.61
1,856.61 – 2,126.00
(b)
pH
5.80 – 6.06
6.06 – 6.28
6.28 – 6.48
6.48 – 6.65
6.65 – 6.80
6.80 – 6.98
6.98 – 7.18
7.18 – 7.40
(c)
EC (µS/cm)
665.33 – 877.70
877.70 – 1,050.61
1,050.61 – 1,262.98
1,262.98 – 1,523.82
1,523.82 – 1,844.20
1,844.20 – 2,237.71
2,237.71 – 2,721.03
2,721.03 – 3,314.67
(d)
K⁺ (mg/L)
2.07 – 3.23
3.23 – 4.31
4.31 – 5.31
5.31 – 6.24
6.24 – 7.09
7.09 – 7.89
7.89 – 8.62
8.62 – 9.3
(e)
Mg²⁺ (mg/L)
21.37 – 25.58
25.58 – 32.44
32.44 – 43.63
43.63 – 61.85
61.85 – 91.54
91.54 – 139.93
139.93 – 218.78
218.78 – 347.27
(f)
F⁻(mg/L)
0.49 – 0.66
0.66 – 0.86
0.86 – 1.11
1.11 – 1.31
1.31 – 1.48
1.48 – 1.62
1.62 – 1.74
1.74 – 1.84
N
R1 R2 R3 R4 R5 R6 R7 R8 R9 R10 R11 R12 R13 R14 R15 R16 R17 R18 R19 R20 R21

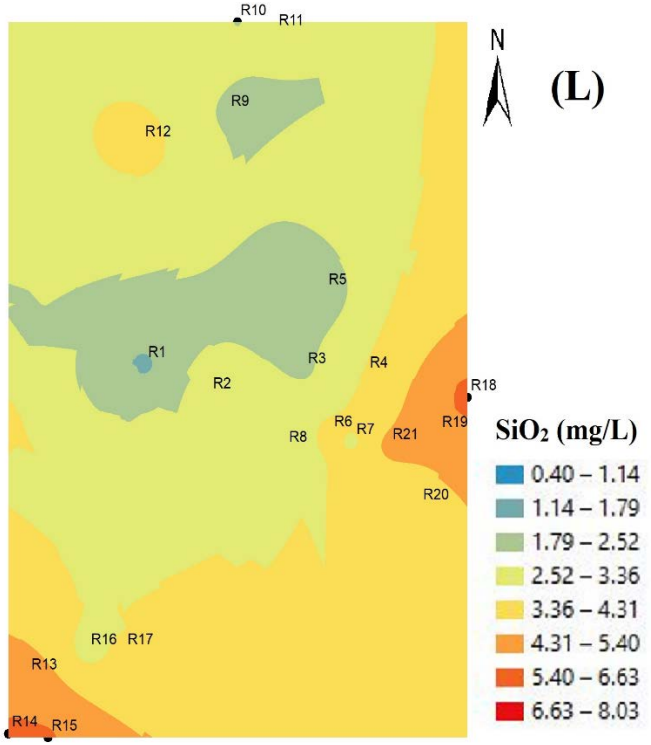

(g)
N
R10 R11
R9
R12
R5
R1
R3
R4
R2
R18
R6 R7 R21 R19
R8
R20
R16 R17
R13
R14 R15
Fe2+ (mg/L)
0.02 – 0.02
0.02 – 0.04
0.04 – 0.07
0.07 – 0.14
0.14 – 0.30
0.30 – 0.64
0.64 – 1.37
1.37 – 2.96


(h)
N
R10 R11
R9
R12
R5
R1
R3
R4
R2
R18
R6 R7 R21 R19
R8
R20
R16 R17
R13
R14 R15
Mn (mg/L)
0.01 – 0.01
0.01 – 0.02
0.02 – 0.02
0.02 – 0.03
0.03 – 0.04
0.04 – 0.06
0.06 – 0.07
0.07 – 0.08


(I)
N
R10 R11
R9
R12
R5
R1
R3
R4
R2
R18
R6 R7 R21 R19
R8
R20
R16 R17
R13
R14 R15
Ca2+ (mg/L)
22.00 – 54.78
54.78 – 64.08
64.08 – 66.72
66.72 – 67.47
67.47 – 70.12
70.12 – 79.42
79.42 – 112.20
112.20 – 227.67


(j)
N
R10 R11
R9
R12
R5
R1
R3
R4
R2
R18
R6 R7 R21 R19
R8
R20
R16 R17
R13
R14 R15
Na+ (mg/L)
55.97 – 137.36
137.36 – 179.09
179.09 – 200.48
200.48 – 211.45
211.45 – 232.84
232.84 – 274.57
274.57 – 355.96
355.96 – 514.73


(K)
N
R10 R11
R9
R12
R5
R1
R3
R4
R2
R18
R6 R7 R21 R19
R8
R20
R16 R17
R13
R14 R15
SO4 2– (mg/L)
35.73 – 76.95
76.95 – 104.28
104.28 – 145.5
145.5 – 207.68
207.68 – 301.49
301.49 – 443.02
443.02 – 656.52
656.52 – 978.6


(L)
N
R10 R11
R9
R12
R5
R1
R3
R4
R2
R18
R6 R7 R21 R19
R8
R20
R16 R17
R13
R14 R15
SiO2 (mg/L)
0.40 – 1.14
1.14 – 1.79
1.79 – 2.52
2.52 – 3.36
3.36 – 4.31
4.31 – 5.40
5.40 – 6.63
6.63 – 8.03

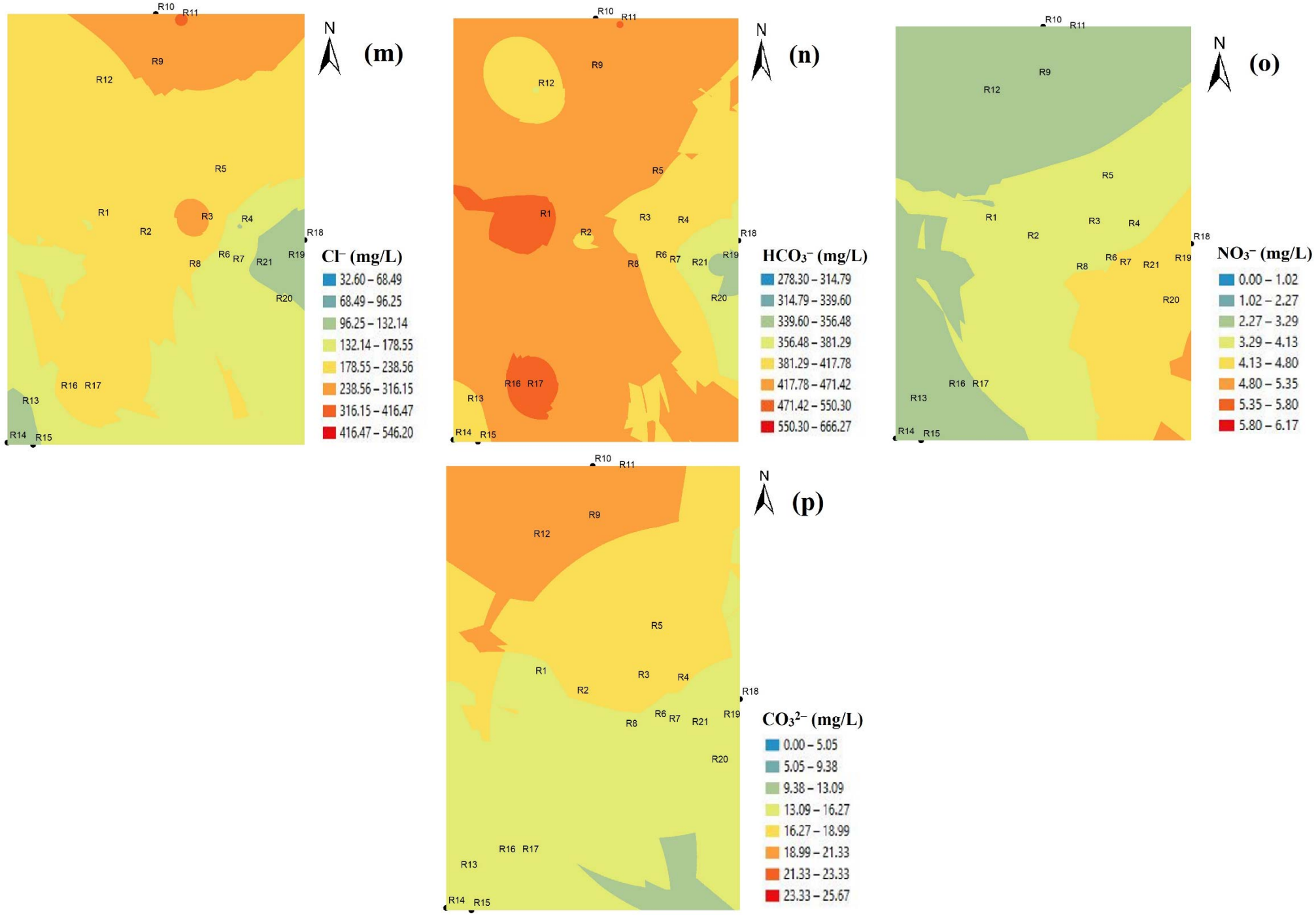
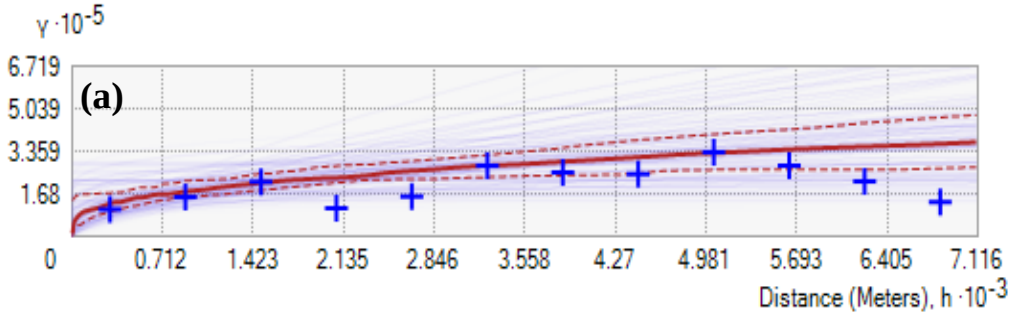
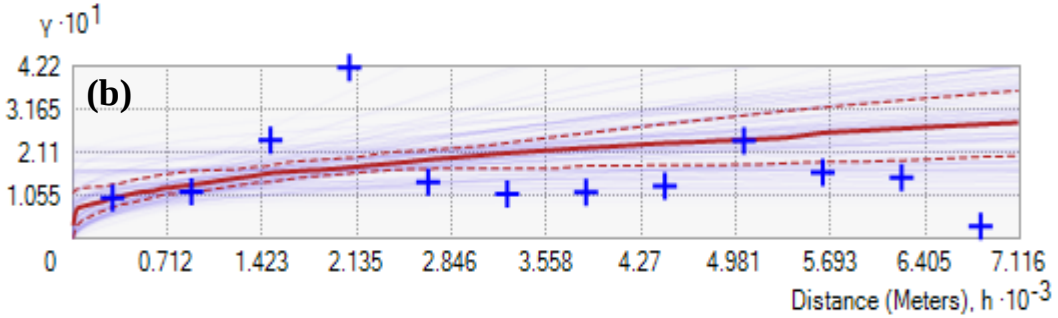
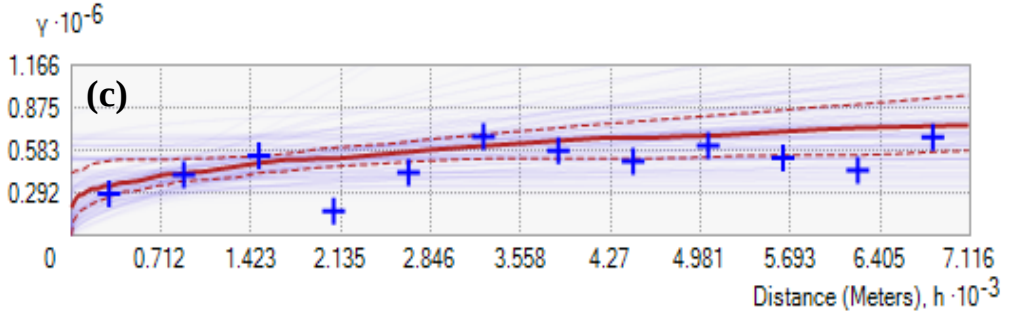
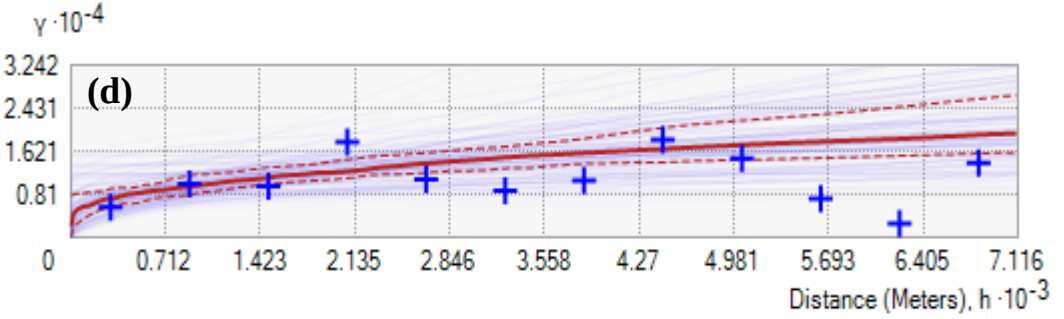


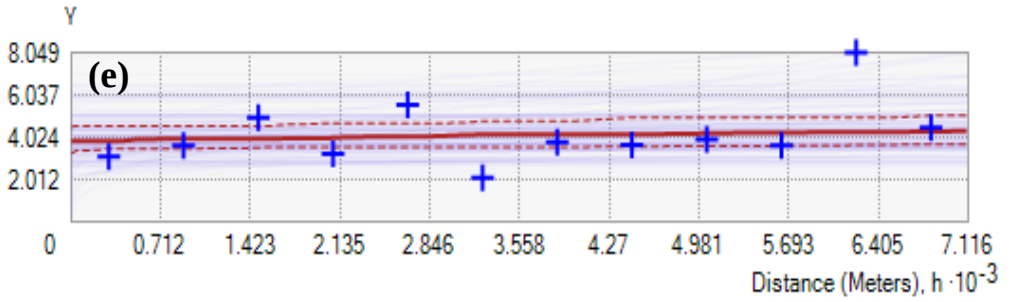
Fig. 3 : Spatial distribution maps of (a) TDS (b) pH (c) EC (d) K^+ (e) Mg^{2+} (f) F^- (g) Fe^{2+} (h) Mn (i) Ca^{2+} (j) Na^+ (k) SO_4^{2-} (l) SiO_2 (m) Cl^- (n) HCO_3^- (o) NO_3^- (p) CO_3^{2-} in the study area.

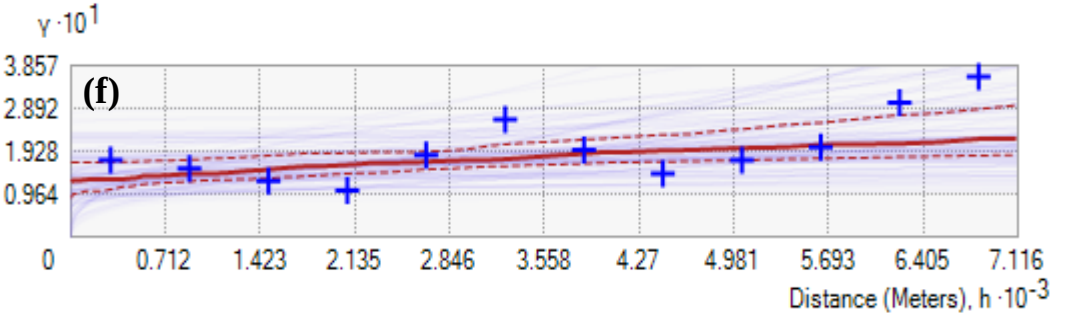

γ ·10-5
6.719
5.039
3.359
1.68
(a)
0
0.712
1.423
2.135
2.846
3.558
4.27
4.981
5.693
6.405
7.116
Distance (Meters), h ·10-3

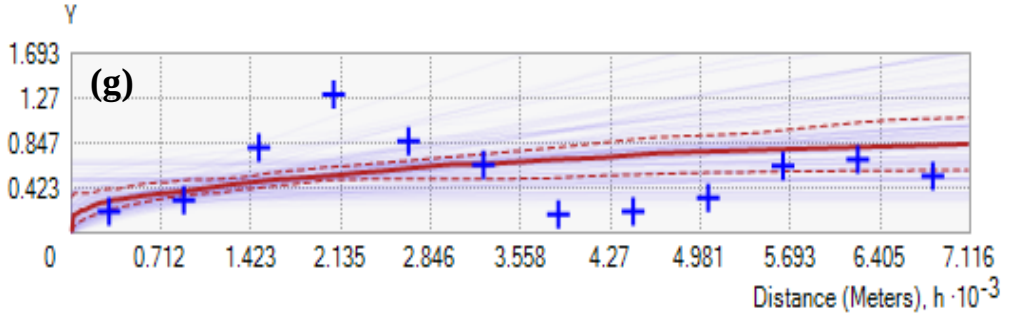

γ ·101
4.22
3.165
2.11
1.055
(b)
0
0.712
1.423
2.135
2.846
3.558
4.27
4.981
5.693
6.405
7.116
Distance (Meters), h ·10-3

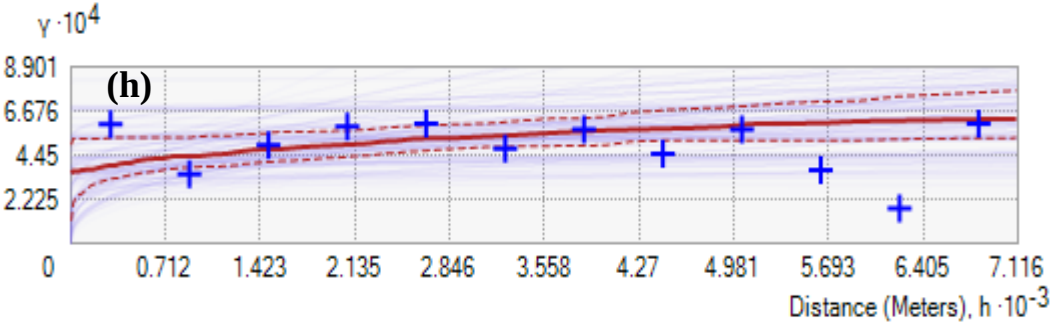

γ ·10-6
1.166
0.875
0.583
0.292
(c)
0
0.712
1.423
2.135
2.846
3.558
4.27
4.981
5.693
6.405
7.116
Distance (Meters), h ·10-3


γ ·10-4
3.242
2.431
1.621
0.81
(d)
0
0.712
1.423
2.135
2.846
3.558
4.27
4.981
5.693
6.405
7.116
Distance (Meters), h ·10-3


γ
8.049
6.037
4.024
2.012
(e)
0
0.712
1.423
2.135
2.846
3.558
4.27
4.981
5.693
6.405
7.116
Distance (Meters), h ·10-3


γ ·101
3.857
2.892
1.928
0.964
(f)
0
0.712
1.423
2.135
2.846
3.558
4.27
4.981
5.693
6.405
7.116
Distance (Meters), h ·10-3


γ
1.693
1.27
0.847
0.423
(g)
0
0.712
1.423
2.135
2.846
3.558
4.27
4.981
5.693
6.405
7.116
Distance (Meters), h ·10-3


γ ·104
8.901
6.676
4.45
2.225
(h)
0
0.712
1.423
2.135
2.846
3.558
4.27
4.981
5.693
6.405
7.116
Distance (Meters), h ·10-3

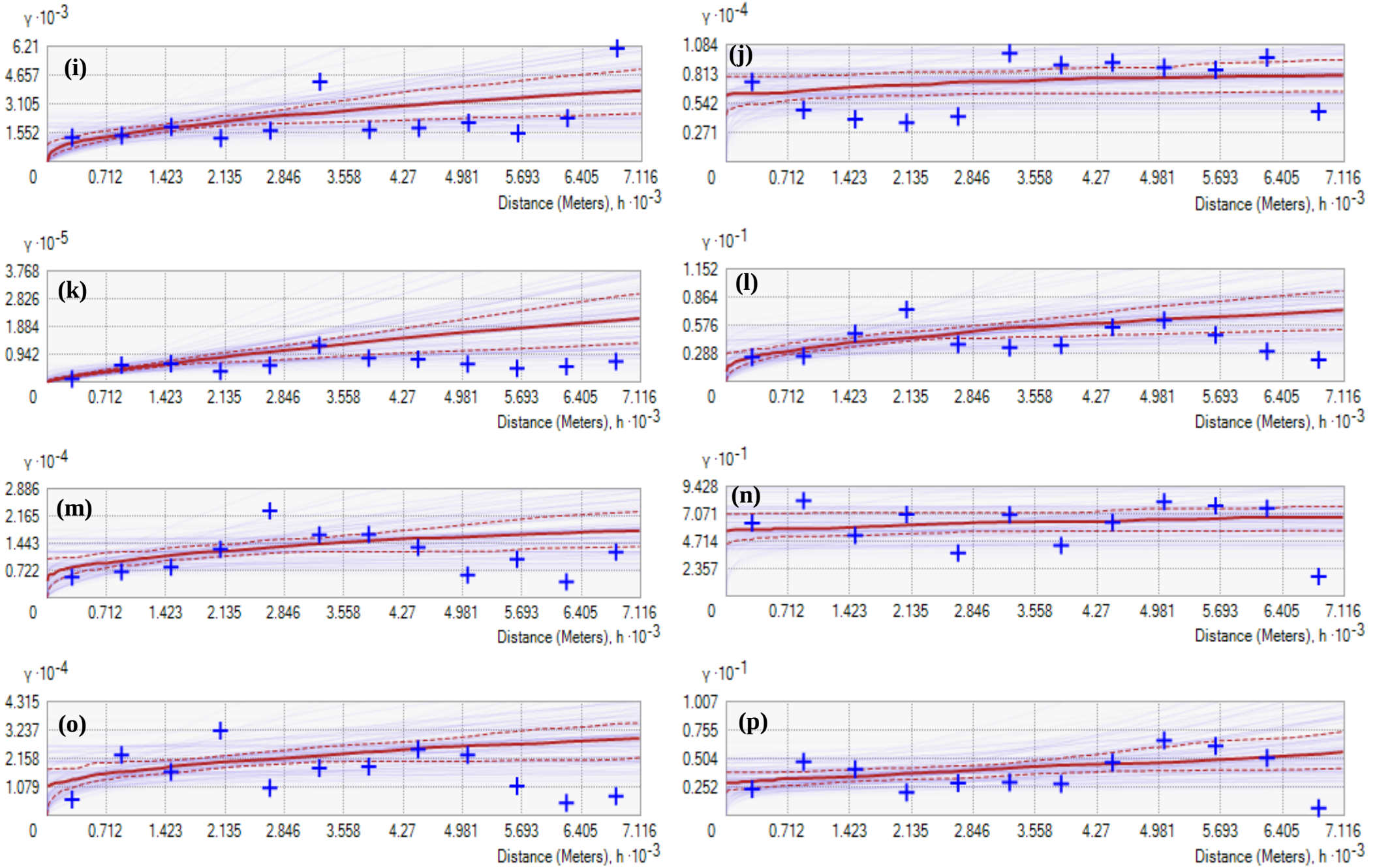


Fig. 4: Semivariograms of the water parameters (a) TDS (b) pH (c) EC (d) K^+ (e) Mg^{2+} (f) F^- (g) Fe^{2+} (h) Mn (i) Ca^{2+} (j) Na^+ (k) SO_4^{2-} (l) SiO_2 (m) Cl^- (n) HCO_3^- (o) NO_3^- (p) CO_3^{2-}

HIGHLIGHTS

- Water quality was assessed and compared with WHO standards
- Water quality parameters is modelled spatially and statistically
- Contaminant sources were inferred using multivariate statistical technique
- Health risk assessment was computed through ingestion and dermal contact route



https://dspace.org/

Fluoride contamination in groundwater sources in Southwestern Nigeria: Assessment using multivariate statistical approach and human health risk

Emenike, Chidozie PraiseGod; Tenebe, Imokhai Theophilus; Jarvis, Peter

2018-04-03
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https://doi.org/10.1016/j.ecoenv.2018.03.022
Emenike CP, Tenebe IT, Jarvis P. (2018) Fluoride contamination in groundwater sources in Southwestern Nigeria: Assessment using multivariate statistical approach and human health risk. Ecotoxicology and Environmental Safety, Volume 156, July 2018, pp. 391-402

Downloaded from a DSpace repository. https://dspace.org/