# Role of agricultural land practices in the behaviour of nitrates in groundwater

Daniel Bonotto<sup>1</sup>, Ayomi Jayarathne<sup>2</sup>, Daniel Bonotto<sup>3</sup>, Buddhi Wijesiri<sup>2</sup>, Ashantha Goonetilleke<sup>2</sup>, and Daniel Bonotto<sup>4</sup>

<sup>1</sup>Departamento de Geologia, IGCE-UNESP <sup>2</sup>Queensland University of Technology (QUT) <sup>3</sup>Instituto de Geociências e Ciências Exatas-IGCE <sup>4</sup>Departamento de Geologia - IGCE/UNESP

November 22, 2022

#### Abstract

Nitrate contamination is a major issue in aquifers that are being exploited for drinking water. Exceeding regulatory levels of nitrates in drinking water can cause acute and chronic health problems. In agricultural areas, aquifers are vulnerable to nitrate contamination due to the excessive use of fertiliser. This research study investigated the potential impacts of anthropogenic nitrates on the giant Guarani Aquifer System (GAS) in Brazil, where nitrogen-based fertiliser use had doubled from 2005–2016. The study results indicated that there exists two different systems for the behavior of nitrates in groundwater, above and below a 150 m depth of the aquifer. For the aquifer depth above 150 m, Cl- (positive influence) and F- (negative influence) were found to significantly influence NO3- contamination (p < 0.05). However, statistically significant relationships between NO3- and other influential factors were not found for the aquifer depth below 150 m. Even though agricultural practices do not pose a direct impact on NO3- contamination of groundwater, it was evident that anthropogenic inputs of NO3- could elevate the concentrations in the aquifer depth reaching 150 m.

#### Hosted file

essoar.10506612.1.docx available at https://authorea.com/users/536342/articles/599040-roleof-agricultural-land-practices-in-the-behaviour-of-nitrates-in-groundwater

#### Hosted file

agusupporting-information\_nitrategroundwater.docx available at https://authorea.com/users/ 536342/articles/599040-role-of-agricultural-land-practices-in-the-behaviour-of-nitratesin-groundwater 1

## Role of agricultural land practices in the behaviour of nitrates in groundwater A. Jayarathne<sup>1</sup>, D. M. Bonotto<sup>2</sup>, B. Wijesiri<sup>1,3</sup>, and A. Goonetilleke<sup>1</sup>

- <sup>4</sup> <sup>1</sup>Science and Engineering Faculty, Queensland University of Technology (QUT), GPO Box
- 5 2434, Brisbane, QLD 4001, Australia.
- <sup>6</sup> <sup>2</sup>Instituto de Geociências e Ciências Exatas-IGCE, Universidade Estadual Paulista-UNESP, Av.
- 7 24-A No. 1515, P.O. Box 178, CEP 13506-900, Rio Claro, São Paulo, Brazil.
- <sup>3</sup>College of Chemistry and Environmental Engineering, Shenzhen University, Shenzhen 518060,
  China.
- 10 Corresponding author: Daniel Bonotto (<u>daniel.bonotto@unesp.br</u>)

## 11 Key Points:

- Quantitative assessment of nitrate contamination in groundwater
- Agricultural practices indirectly influence nitrate contamination in aquifers
- Two major systems for nitrates may exist at aquifer depths above and below 150 m
- 15

#### 16 Abstract

17 Nitrate contamination is a major issue in aquifers that are being exploited for drinking water.

18 Exceeding regulatory levels of nitrates in drinking water can cause acute and chronic health

19 problems. In agricultural areas, aquifers are vulnerable to nitrate contamination due to the

20 excessive use of fertiliser. This research study investigated the potential impacts of

anthropogenic nitrates on the giant Guarani Aquifer System (GAS) in Brazil, where nitrogen-

based fertiliser use had doubled from 2005–2016. The study results indicated that there exists

two different systems for the behavior of nitrates in groundwater, above and below a 150 m depth of the aquifer. For the aquifer depth above 150 m,  $Cl^{-}$  (positive influence) and F<sup>-</sup> (negative

depth of the aquifer. For the aquifer depth above 150 m, Cl<sup>-</sup> (positive influence) and F<sup>-</sup> (negative influence) were found to significantly influence NO<sub>3</sub><sup>-</sup> contamination (p < 0.05). However,

statistically significant relationships between  $NO_3^-$  and other influential factors were not found

for the aquifer depth below 150 m. Even though agricultural practices do not pose a direct impact

on  $NO_3^-$  contamination of groundwater, it was evident that anthropogenic inputs of  $NO_3^-$  could

elevate the concentrations in the aquifer depth reaching 150 m.

#### 30 1 Introduction

31 Groundwater contamination is a major environmental problem worldwide (Mor et al., 2006; Stuart et al., 2014; Wang et al., 2007). In addition to natural sources such as weathering of 32 rocks or soil, human activities are a primary contributor to groundwater contamination. The 33 contaminants percolate through soil layers, degrading groundwater. This can ultimately cause 34 serious health issues when people utilise the contaminated groundwater as a drinking water 35 source (Chen et al., 2016; Majumdar & Gupta 2000). It has been reported that over two billion 36 people worldwide rely on groundwater for their primary water uses such as drinking water, 37 agriculture and food production (Famiglietti, 2014). 38

39 Among different contaminants, nitrate ( $NO_3^{-}$ ) contamination is a critical concern in terms of groundwater pollution and related health impacts due to intensive land use practices and the 40 application of N-based fertiliser in agricultural activities. According to the United Nations Food 41 and Agricultural Organization (FAO), rice, maize and wheat utilise over 85% of N-based 42 fertiliser (FAO, 2006), while global demand for N-based fertiliser has increased from 110 43 million tonnes to 117.1 million tonnes during the period of 2015-2019 (FAO, 2020). These 44 45 statistics indicate that there is a growing trend in the application of N-based fertiliser for agricultural production. Even though nitrogen is an essential element for plant growth, the excess 46 application of N-based fertiliser can negatively influence the denitrification capacity of soil, 47 leading to the leaching of N in the form of nitrates into groundwater during the transformation of 48 N within the soil (Almasri & Kaluarachchi, 2004; Follet & Delgado, 2002; Menció et al., 2016). 49

50 The transport and fate of nitrates in subsurface environments and their influence on groundwater quality have been widely studied (Joshua et al., 2013; Kaçaroğlu & Günay, 1997; 51 Maila et al., 2004). Nitrate is considered to be a highly mobile contaminant in groundwater due 52 53 to its anionic form and solubility characteristics (Canter, 2019). Thus, nitrates can migrate longer distances from the source region when favourable surface water infiltration is present. 54 Additionally, aquifers that have highly permeable subsurface materials and shallow water tables 55 are susceptible to nitrate contamination (Mahvi et al., 2005). Past studies have shown that 56 aquifers found around agricultural lands were contaminated by nitrates, exceeding the drinking 57

water standards of World Health Organization (WHO) (Burkart & Stoner, 2008; Korbel et al.,
2013; McLay et al., 2001).

This study investigated the potential impact of agricultural practices on nitrate contamination in groundwater, while accounting for typical parameters that influence nitrates behaviour in groundwater. The study was based in the giant Guarani Aquifer System (GAS), Brazil, and the research outcomes are expected to contribute to prudent land use management strategies to protect critical groundwater resources in a region.

## 65 2 Materials and Methods

## 66 2.1 Study area

The study was carried out in a catchment located within the Paraná sedimentary basin, 67 South Brazil (Figure 1). The hydro-stratigraphy of the Paraná basin consists of multi-aquifer 68 systems, comprising sandstones and basaltic sediments (Campos, 2000). Groundwater primarily 69 occurs within the interflow zones and along the basalts and diabases joints, where interbedded 70 sediments increase the porosity of rocks. The GAS of Triassic-Jurassic age is the largest aquifer 71 72 system within the Paraná sedimentary basin (extends over 1.2 million km<sup>2</sup> with an average thickness of 300-400 m), which mainly composes of silty and shaly sandstones of fluvial-73 lacustrine origin and variegated quartzitic sandstones (Araújo et al., 1999). A detailed description 74 of the general features of the study area is provided in the Supporting Information. The dominant 75 land use within the study area are agricultural lands, which are widely utilised for sugar cane, 76 soybean and corn, flooded rice farming and pasture, and followed by urban uses. 77

#### 78 2.2 Sample collection and laboratory testing

79 Groundwater samples were collected from 78 free-flowing and pumped tubular wells drilled into the GAS, considering its geological formation within the study area. Geo-coordinates 80 81 of sampling locations and agricultural land use fraction as well the lithologies of sampling wells are provided in Tables S1 and S2 in the Supporting Information. The dominant land use within 82 the study area are agricultural lands, which are widely utilised. The physical parameters of 83 groundwater, including temperature (T), dissolved oxygen (DO), pH, and electrical conductivity 84 (EC) were measured *in situ*. Samples collected were stored in polyethylene bottles (50 L), 85 transported to the laboratory and kept under 4°C until further analyses were carried out. The 86 hydrogeochemical parameters of groundwater, including nitrate ( $NO_3^{-}$ ), sulfate ( $SO_4^{2-}$ ), chloride 87 (Cl<sup>-</sup>), fluoride (F<sup>-</sup>), bicarbonate (HCO<sub>3</sub><sup>-</sup>), dissolved sodium (Na), potassium (K), calcium (Ca) and 88 magnesium (Mg) were analysed. A detailed description of the analytical methods used, quality 89 control and quality assurance procedures adopted, and the data set generated are provided in 90 Tables S3 and S4 in the Supporting Information. 91

## 92 2.3 Data analysis

The factors that potentially influence the NO<sub>3</sub><sup>-</sup> contamination in groundwater were quantitatively assessed using Bayesian Networks (BNs) modelling. BNs develop probabilistic relationships among variables that define a process/system based on the prior knowledge or expert opinion (Bonotto et al., 2018, 2019; Uusitalo, 2007). The relationships between variables are represented by a directed acyclic diagram, and its *Markov Property* indicates that each variable depends only on its immediate parent variables. The model parameters are estimated in terms of conditional probabilities for discrete variables and in terms of conditional regression

coefficients for continuous variables. Further details on BNs modelling can be found elsewhere
(Jayarathne et al., 2019; Scutari, 2010, 2013; Wijesiri et al., 2018).

102



103

Figure 1. Sampling locations in the Guarani Aquifer System (GAS) within the Paraná Paraná
 sedimentary basin, South Brazil.

#### 106 **3 Results and Discussion**

Once released to the surface soils, infiltration plays a key role in transporting nitrates to the aquifer. Figure 2 shows the variations in water table depth and nitrate concentrations at each well investigated in this study. The majority of wells indicate that lower the water table (situated

in the close proximity to ground surface), the higher the nitrate concentration in groundwater.

111 This pattern is more frequent in wells drilled up to a surface depth of 150 m (aquifer depth). This

112 could be due to the larger number of wells above 150 m depth compared to the fewer number of

wells below 150 m of surface depth. On the other hand, there can be two different

114 hydrogeochemical systems of nitrates in groundwater above and below 150 m of surface depth.

115 This was investigated using BNs modelling.



116

Figure 2. Variation in nitrate concentration in groundwater and water table against the depth ofwells.

In the BNs analysis, those wells where the surrounding land use data was not available, 119 were removed. Then, the remaining wells were separated, such that the aquifer depth is above 120 150 m and below 150 m. Further, the two data sets were screened by removing the samples 121 where hydrogeochemical parameters were not available and by taking half the detection limit for 122 the parameters that were measured below the detection limit. The outliers (those outside 1.5 123 times interquartile range) were also removed. The final data sets included 20 wells above 150 m 124 depth and 37 wells below 150 m depth. However, it is important to note that those wells above 125 150 m ranged from 75–150 m, while the deeper wells ranged from 151–4,582 m. As such, the 126 overall data set included 20 wells within the first 150 m of the aquifer, while there were 37 wells 127 within 4.4 km. 128

The two data sets were then fitted with a simple BNs model shown in Figure 3. As such, 129 the model considered direct interdependencies between nitrates concentration in groundwater 130 131 and agricultural land use and 12 hydrogeochemical parameters (T, DO, pH, EC, SO<sub>4</sub><sup>2-</sup>, Cl<sup>-</sup>, F<sup>-</sup>, HCO<sub>3</sub>, Na, K, Ca and Mg) identified based on the research literature (see Figure 3 for relevant 132 references). After performing leave-one-out cross validation (given the limited number of 133 samples), F-test indicated that the overall model performance is satisfactory (p < 0.05) only for 134 the wells reaching 150m depth. Further, model performance for aquifer depths reaching 150 m 135 was also confirmed from the observed/predicted and residuals plots shown in Figure 4. 136



Figure 3. Bayesian Networks (BNs) model of the system of nitrates in groundwater. Note:
interdependencies shown in BNs model were identified based on literature, such that
NO<sub>3</sub><sup>-</sup>/Temperature (Luk & Au-Yeung, 2002; Pfenning & McMahon, 1997); NO<sub>3</sub><sup>-</sup>/pH (Menció et al., 2011; Yesilnacar et al., 2008); NO<sub>3</sub><sup>-</sup>/Dissolved Oxygen (Rivett et al., 2008); NO<sub>3</sub><sup>-</sup>/Electrical

- 152 Conductivity (Menció et al., 2011);  $NO_3^-/HCO_3^-$  (Tang et al., 2012);  $NO_3^-/Cl^-$  and F<sup>-</sup> (Menció et
- al., 2016);  $NO_3^{-}/SO_4^{-2}$  (Menció et al., 2016),  $NO_3^{-}/Na$ , K, Ca and Mg (Menció & Mas-Pla, 2008).

On the other hand, cross validation revealed that the overall model does not accurately replicate the behaviour of NO<sub>3</sub><sup>-</sup> at depths beyond 150m (p – 0.329), implying that there is no statistically significant relationship between NO<sub>3</sub><sup>-</sup> and influential factors. Additionally, to confirm the critical depth of 150 m at which nitrates might exhibit different behaviour, the model was fitted with data of wells above 200 m (26) and wells below 200 m (31). However, the overall model performance was not satisfactory for both cases (above 200 m: p – 0.183 and below 200m: p – 0.826 without statistically significant relationships between influential factors).



#### 162

Figure 4. (a) Observed vs Predicted values plot; (b) Residual plot for aquifer depths above 150
 m.

Table 1 shows the estimated conditional regression coefficients relating to nitrate 165 behaviour within the aquifer depth above 150 m, which indicates statistically significant 166 interdependencies between  $NO_3^-$  and  $Cl^-$  (positive influence) and  $F^-$  (negative influence). As the 167 most influential factors, a 1% increase in Cl<sup>-</sup> concentration in groundwater would likely increase 168 the NO<sub>3</sub><sup>-</sup> concentration by 0.099 mg/L, while a similar increase in F<sup>-</sup> concentration would result 169 in a decrease in the  $NO_3^{-1}$  concentration by 0.17 mg/L. BNs modelling outcomes further indicate a 170 positive influence of  $SO_4^{2-}$  on the prevalence of  $NO_3^{-}$  in groundwater. The prominent sources of 171  $NO_3^-$ , Cl<sup>-</sup> and  $SO_4^{2-}$  ions in groundwater are mostly agricultural fertiliser, animal wastes and 172 industrial and municipal sewage (Ako et al., 2014; Jalali, 2009). As such, the positive 173 relationships between these ions can be attributed to fertiliser application in the study area. 174

Furthermore, the results show that lower  $NO_3^-$  concentrations could be observed in groundwater that is hot and basic (high temperature and pH), while cold and acidic conditions may increase the prevalence of  $NO_3^-$  (see negative relationships for temperature and pH in Table 1). It has been reported that during the nitrification of  $NO_2^-$  and  $NH_4^+$  in the subsurface soil layers drained by groundwater, there is an increase in H<sup>+</sup> concentration in groundwater, indicating that reduced pH can increase  $NO_3^-$  concentration in groundwater (Menció et al., 2011; Stumm & Morgan, 1970).

182 As evident from Table 1, Na and Mg positively influence the prevalence of  $NO_3^-$ , whilst there is a negative influence from K and Ca. This implies that the increase in nitrates in 183 groundwater is accompanied by the increase in Na and Mg in groundwater. Further, Table 1 184 shows that the increase in  $NO_3^-$  would likely decrease  $HCO_3^-$  concentration (negative influence), 185 where there is high dissolution of carbonate rocks in the unsaturated zone (Baalousha, 2008). 186 Such a relationship agrees with the findings by Kim et al. (2005). Based on the reaction 187 188 stoichiometry for nitrification and oxidation of organic matter, Kim et al. (2005) noted that nitrate-generating processes produce an equivalent of nitrate by consuming the same equivalent 189 of alkalinity, justifying the negative correlations between nitrate and alkalinity. 190

191

<sup>a</sup> Conditional density: NO <sub>3</sub> <sup>-</sup>   AGR + T + pH + DO + EC + HCO <sub>3</sub> <sup>-</sup> + Cl <sup>-</sup> + SO <sub>4</sub> <sup>2-</sup> + F <sup>-</sup> + Na + K + Ca + Mg						
Intercept (NO <sub>3</sub> <sup>-</sup> )	AGR	Т	рН	DO	EC	HCO <sub>3</sub> -
0.00547	0.0616	-0.0553	-0.203	0.0820	0.0319	-0.158
Cl	<b>SO</b> <sub>4</sub> <sup>2-</sup>	F-	Na	K	Ca	Mg
0.0989*	0.0507	-0.170*	0.136	-0.0505	-0.122	0.121

#### Leave-one-out cross validation

Significance code: \* 0.05

Overall model p (F-test): 0.0402 (<0.05)

<sup>a</sup> Conditional density refers to the probability density function of a variable given each of its immediate parent variables AGR – agricultural land area

T - water temperature

DO – dissolved oxygen

EC – electrical conductivity

Table 1. Estimated conditional regression coefficients for the proposed Bayesian Networks
 (BNs) model for aquifer depths above 150 m (Gaussian distribution, log-transformed data).

More importantly, the model indicates that a direct relationship between agricultural 194 lands and NO<sub>3</sub><sup>-</sup> is unlikely, meaning that there exist intermediate factors as the sources of NO<sub>3</sub><sup>-</sup> to 195 the aquifer. These could potentially include the mineral weathering through rock/soil-water 196 interaction that generally exerts as an important control on groundwater chemistry, dominating 197 the concentration of the major cations and anions. Silicates and carbonates are widely spread in 198 the study area and their leaching affects the hydrochemical composition of the GAS 199 groundwaters because of their reactivity and abundance in the magmatic and sedimentary rocks 200 of the Paraná basin. However, given that the overall model performs satisfactorily, the estimated 201 conditional regression coefficients indicate that 1% increase in agricultural land uses could 202 elevate  $NO_3^-$  concentration by 0.062 mg/L. The influence of agricultural lands, in terms of 203 fertiliser applications alone on  $NO_3^-$  concentration in groundwater might be low, but it is 204 205 important to note that agricultural practices create a favourable environment for the prevalence of  $NO_3$ , which can be a challenging issue into the future. This is because of the increasing extent 206 of Brazil's agricultural lands (as a fraction of total area: 31% in 1995, 33% in 2005 and 34% in 207 2016) (FAO, 2018), together with increasing urbanisation could change the hydrogeochemical 208 environment in the aquifers. 209

#### 210 4 Conclusions

- Based on the giant Guarani Aquifer System (GAS) in Brazil, this study investigated the impact
- of agricultural land use practices on the nitrate contamination of groundwater, while accounting
- for the aquifer geochemistry. The results derived indicate that there can be two different
- hydrogeochemical systems of nitrates above and below the aquifer depth of 150 m. For the
- aquifer depth above 150 m, significant interdependencies were found between  $NO_3^-$  and  $Cl^-$
- (positive) and  $NO_3^-$  and F- (negative). A 1% increase in Cl<sup>-</sup> concentration in groundwater was
- found to elevate the NO<sub>3</sub><sup>-</sup> concentration by 0.099 mg/L, while a similar increase in  $F^-$
- concentration would decrease  $NO_3^-$  concentration by 0.17 mg/L. Importantly, it was evident that
- the agricultural land use fraction is unlikely to pose a direct impact on  $NO_3$  in groundwater, but
- indicated that there exists intermediate factors such as physicochemical properties of subsurface
- 221 rocks/soils that potentially influence the transport of  $NO_3$  released from agricultural lands.
- However, the study found evidence of the influence of agricultural land use practices on  $NO_3^{-1}$ .

#### 223 Acknowledgments

- The authors thank support given jointly by Fundação de Amparo à Pesquisa do Estado de São
- 225 Paulo (FAPESP), Brazil and Australian Technology Network (ATN) (Sprint Grant ID:
- 226 2016/50327-4), and by Guangdong Basic and Applied Basic Research Foundation, China
- 227 (2019A1515110353). Data sets for this research are included in this paper and its Supporting
- 228 Information file. Queensland University of Technology (QUT), which is the home university of
- three of the co-authors, has a stringent policy on Management of Research Data. It is based on
- the Australian Code for the Responsible Conduct of Research. QUT provides storage for all
- research data. All data of this paper will be registered in QUT's data registry, Research Data
- Finder, and the national data discovery portal, Research Data Australia. Open or mediated access
- to the data itself will be provided, where possible, via an open access repository. Curation of the

- data is guided by the Queensland State Archive University Sector Retention and Disposal
- 235 Schedule for research data, as well as other relevant legislation.

#### 236 **References**

- Ako, A. A., Eyong, G. E. T., Shimada, J., Koike, K., Hosono, T., Ichiyanagi, K., Richard, A.,
- Tandia, B. K., Nkeng, G. E., & Roger, N. N. (2014). Nitrate contamination of groundwater in
- two areas of the Cameroon Volcanic Line (Banana Plain and Mount Cameroon area). *Applied Water Science*, 4(2), 99-113.
- 241 Almasri, M. N., & Kaluarachchi, J. J. (2004). Assessment and management of long-term nitrate
- pollution of ground water in agriculture-dominated watersheds. *Journal of Hydrology*, 295(1-4), 225-245.
- Almeida, F. F. M., & Melo, M. S. (1981). The Paraná basin and Mesozoic volcanism. In IPT
- (Institute of Technological Researches of São Paulo State) (Ed.), *Geological map of São Paulo State* (Vol. 1, pp. 46-81). São Paulo, Brazil: Promocet.
- APHA (American Public Health Association) (1989). Standard Methods for the Examination of
- 248 Water and Waste Water. (17<sup>th</sup> ed.). Washington, DC: APHA.
- 249 Araújo, L. M., França, A. B., & Potter, P. E. (1999). Hydrogeology of the Mercosul aquifer
- system in the Paraná and Chaco-Parana Basins, South America, and comparison with the
  Navajo-Nugget aquifer system, USA. *Hydrogeology Journal*, 7, 317-336.
- Baalousha, H. (2008). Analysis of nitrate occurrence and distribution in groundwater in the Gaza Strip using major ion chemistry. *Global NEST Journal*, 10(3), 337-349.
- Bonotto, D. M. (2006). Hydro(radio)chemical relationships in the giant Guarani aquifer, Brazil.
   *Journal of Hydrology*, 323(1-4), 353-386.
- 256 Bonotto, D. M., Wijesiri, B., & Goonetilleke, A. (2019). Nitrate-dependent Uranium 257 mobilisation in groundwater. *Science of the Total Environment* 693, 133655.
- 258 Bonotto, D. M., Wijesiri, B., Vergotti, M., da Silveira, E. G., & Goonetilleke, A. (2018).
- Assessing mercury pollution in Amazon River tributaries using a Bayesian Network approach.
   *Ecotoxicology and Environmental Safe*, 166, 354-358.
- Burkart, M. R., & Stoner, J. D. (2008). Nitrogen in groundwater associated with agricultural
- systems. In J. L. Hatfield & R. F. Follett (Eds.), *Nitrogen in the Environment* (pp. 177-202).
  Amsterdam, The Netherlands: Elsevier.
- Campos, H. C. N. S. (2000). Hydrogeological map of Guarani aquifer. Acta Geológica *Leopoldensia*, 23(4), 1-50.
- 266 Canter, L. W. (2019). Nitrates in groundwater. London, UK: Routledge.
- 267 Chen J., Wu, H., & Qian, H. (2016). Groundwater nitrate contamination and associated health
- risk for the rural communities in an agricultural area of Ningxia, northwest China. Exposure and
- 269 *Health*, 8(3), 349-359.
- Famiglietti, J. S. (2014). The global groundwater crisis. *Nature Climate Change*, 4(11), 945-948.
- FAO (Food and Agriculture Organization of the United Nations) (2006). *Fertilizer use by crops* -*Fertilizer and plant nutrition bulletin*. Rome: FAO.
- FAO (Food and Agriculture Organization of the United Nations) (2018). *World food and agriculture - Statistical pocketbook 2018*. Rome: FAO.
- FAO (Food and Agriculture Organization of the United Nations) (2020). World fertilizer trends
- and outlook to 2020: Summary Report. Rome: FAO.
- Follett, R., & Delgado, J. (2002). Nitrogen fate and transport in agricultural systems. Journal of
- 278 Soil Water Conservation, 57(6), 402-408.

- 279 Hach (1992). *Water Analysis Handbook*. (2<sup>nd</sup> ed.). Loveland, CO: Hach Co.
- Jalali, M. (2009). Geochemistry characterization of groundwater in an agricultural area of Razan,
- Hamadan, Iran. *Environmental Geology*, 56(7), 1479-1488.
- Jayarathne, A., Wijesiri, B., Egodawatta, P., Ayoko, G. A., & Goonetilleke, A. (2019). Role of
- adsorption behavior on metal build-up in urban road dust. *Journal of Environmental Sciences*,
  83, 85-95.
- Joshua, W., Thushyanthy, M., & Nanthagoban, N. (2013). Seasonal variation of water table and
- groundwater quality of the karst aquifer of the Jaffna Peninsula-Sri Lanka. Journal of the *National Science Foundation of Sri Lanka*, 41(1), 3-12.
- Kaçaroğlu, F., & Günay, G. (1997). Groundwater nitrate pollution in an alluvium aquifer,
  Eskişehir urban area and its vicinity, Turkey. *Environmental Geology*, 31(3-4), 178-184.
- 290 Kim, K., Rajmohan, N., Kim, H. -J., Kim, S. -H., Hwang, G. -S., Yun, S. -T., Gu, B., Cho, M. J.,
- 291 & Lee, S. –H. (2005). Evaluation of geochemical processes affecting groundwater chemistry 292 based on mass balance approach: a case study in Namwon, Korea. *Geochemical Journal*, 39(4),
- 293 357-369.
- Korbel, K., Hancock, P., Serov, P., Lim, R., & Hose, G. (2013). Groundwater ecosystems vary
- with land use across a mixed agricultural landscape. *Journal of Environmental Quality*, 42(2), 380-390.
- Luk, G. K., Au-Yeung, W. C. (2002). Experimental investigation on the chemical reduction of nitrate from groundwater. *Advances in Environmental Research*, 6(4), 441-453.
- Mahvi, A., Nouri, J., Babaei, A., & Nabizadeh, R. (2005). Agricultural activities impact on
  groundwater nitrate pollution. *International Journal of Environmental Science and Technology*,
  2(1), 41-47.
- Maila, Y. A., El-Nahal, I., & Al-Agha, M. (2004). Seasonal variations and mechanisms of groundwater nitrate pollution in the Gaza Strip. *Environmental Geology*, 47(1), 84-90.
- Majumdar, D., & Gupta, N. (2000). Nitrate pollution of groundwater and associated human health disorders. *Indian Journal of Environmental Health*, 42(1), 28-39.
- McLay, C., Dragten, R., Sparling, G., & Selvarajah, N. (2001). Predicting groundwater nitrate concentrations in a region of mixed agricultural land use: a comparison of three approaches. *Environmental Pollution* 115(2), 191-204.
- 309 Menció, A., Boy, M., & Mas-Pla, J. (2011). Analysis of vulnerability factors that control nitrate
- occurrence in natural springs (Osona Region, NE Spain). Science of the Total Environment,
   409(16), 3049-3058.
- Menció A., & Mas-Pla, J. (2008). Assessment by multivariate analysis of groundwater–surface water interactions in urbanized Mediterranean streams. *Journal of Hydrology*, 352(3), 355-366.
- Menció, A., Mas-Pla, J., Otero, N., Regàs, O., Boy-Roura, M., Puig, R., Bach, J., Domènech, C.,
- Zamorano, M., & Brusi, D. (2016). Nitrate pollution of groundwater; all right..., but nothing else? *Science of the Total Environment*, 539, 241-251.
- 317 Mor, S., Ravindra, K., Dahiya, R., & Chandra, A. (2006). Leachate characterization and
- assessment of groundwater pollution near municipal solid waste landfill site. *Environmental Monitoring and Assessment*, 118(1-3), 435-456.
- 320 Pfenning, K., & McMahon, P. (1997). Effect of nitrate, organic carbon, and temperature on
- potential denitrification rates in nitrate-rich riverbed sediments. *Journal of Hydrology*, 187(3-4), 283-295.
- Rebouças, A. C. (1988). Groundwater in Brazil. Episodes, 11, 209-214.

- 324 Rivett, M. O., Buss, S. R., Morgan. P., Smith, J. W., & Bemment, C. D. (2008). Nitrate
- attenuation in groundwater: a review of biogeochemical controlling processes. *Water Research*,
   42(16), 4215-4232.
- Scutari, M. (2010). Learning Bayesian Networks with the bnlearn R Package. *Journal of Statistical Software*, 35(3), 1-22.
- 329 Scutari M (2013) Package 'bnlearn': Bayesian Network Structure Learning, Parameter Learning
- and Inference. Retrieved from http://www2.uaem.mx/r-mirror/web/packages/bnlearn/bnlearn.pdf
- 332 Stuart, M. E., Lapworth, D. J., Thomas, J., & Edwards, L. (2014). Fingerprinting groundwater
- pollution in catchments with contrasting contaminant sources using microorganic compounds.
- *Science of the Total Environment*, 468, 564-577.
- Stumm, W., & Morgan, J. J. (1970). Aquatic chemistry; an introduction emphasizing chemical *equilibria in natural waters*. New York, NY: Wiley.
- Tang, C., Zhang, Z., & Sun, X. (2012). Effect of common ions on nitrate removal by zero-valent iron from alkaline soil. *Journal of Hazardous Materials*, 231-232, 114-119.
- Uusitalo, L. (2007). Advantages and challenges of Bayesian networks in environmental
   modelling. *Ecological Modelling*, 203(3–4), 312-318.
- 341 Wang, Y., Merkel, B. J., Li, Y., Ye, H., Fu, S., & Ihm, D. (2007). Vulnerability of groundwater
- in Quaternary aquifers to organic contaminants: a case study in Wuhan City, China. *Environmental Geology*, 53(3), 479-484.
- 344 Wijesiri, B., Deilami, K., McGree, J., & Goonetilleke, A. (2018). Use of surrogate indicators for
- the evaluation of potential health risks due to poor urban water quality: a Bayesian Network
- approach. Environmental Pollution, 233, 655-661.
- 347 Yesilnacar, M. I., Sahinkaya, E., Naz, M., & Ozkaya, B. (2008). Neural network prediction of
- nitrate in groundwater of Harran Plain, Turkey. *Environmental Geology*, 56(1), 19-25.
- Zobell, C. E. (1946). Studies on redox potential of marine sediments. *Bulletin of the American Association of Petroleum Geologists*, 30(4), 477-513.
- 351
- 352