

COMMENTS

In this manuscript, the authors employed modified modeling technique for pollution potential and vulnerability of the groundwater resources in Senegal Basin in Mali. They used two different but complementary methods: the DRASTIC method (which evaluates the intrinsic vulnerability) and the fuzzy method (which assesses the specific vulnerability taking into account continuity of the parameters) to show the main following findings:

- Fuzzy model is better than classical model when assessing Groundwater vulnerability to pollution.
- Sensitivity analysis is correlated to fuzzy membership.
- Fuzzy membership can be used directly to do sensitivity analysis in place of classical sensitivity analysis.

This paper is mainly a GIS and Statistical based study of hydrogeochemical data to find the main groundwater potential pollution zones in the Senegal River in Mali. Discussions have been made based on classical and fuzzy DRASTIC models and confirmed with fuzzy membership between parameters and nitrate distribution in the study area. It is founded that fuzzy model is better than classical model and sensitivity analysis is correlated to fuzzy membership.

This article very interesting, which is newsworthy, based on an original work. The methodology adapted to the working scale have led with significant results in line with findings from previous studies. The text is understandable and the methods are well presented.

This study fulfills the journal criteria to be published as it is of both local and global interests. Their work is helpful for the water resource management in Mali and the sustainable development. This study can be a guide for decision makers for any socio-economic infrastructure in the study area.

Hereby I accept this paper which would greatly contribute in the vulnerability assessment of ground water especially in the study area which needs more studies to tackle water related issues after slight modifications which are mentioned below. However, I added my comments regarding the text of the manuscript by adding sticky notes in the manuscript pdf file which would led to further improvement of the manuscript, some are mentioned below.

- 1) Introduction lack references, support your literature with exact reference.
- 2) Introduction lack literature about the comparison of Fuzzy and DRASTIC method which is your basic study. Add a paragraph about.

- 3) Information about study area is missing.
- 4) Need to justify your results with discussion and previous studies.
- 5) better to add few recent references too
- 6) Also need to provide study area map with details of nearby water bodies, river or any other hydrogeologic parameters to get acquaintance with the study area and to get idea about recharge and other parameters.
- 7) Few grammatical improvements are pointed through sticky notes in the manuscript.



23 Abstract

24 Vulnerability to groundwater pollution from Senegal basin was studied by two different but
25 complementary methods: the DRASTIC method (which evaluates the intrinsic vulnerability) and
26 the fuzzy method (which assesses the specific vulnerability taking into account continuity of the
27 parameters). The validation of this application has been tested by comparing the membership in
28 groundwater and distribution of different classes of vulnerabilities established as well as the
29 nitrate distribution in the study area. Three vulnerability classes (low, medium and high) have
30 been identified by both the DRASTIC method and by fuzzy method (passing by normalized
31 model). An integrated analysis reveals that high class with 14.64% (for the DRASTIC method),
32 21.68% (for normalized DRASTIC method) and the very high grade 18.92% (for that of fuzzy)
33 are not the most dominant. In addition, a new method for sensitivity analysis was used to identify
34 (and confirm) the main parameters which impact de vulnerability to pollution with fuzzy
35 membership. And the results showed that vadose is the main parameter which impacts
36 groundwater vulnerability to pollution while net recharge has the least contribution to pollution
37 in the study area. It was found also that Fuzzy method better assesses the vulnerability to
38 pollution with a coincidence rate of 81.13% against 77.35% for the DRASTIC method. These
39 results are a guide for policy makers on protection areas sensitive to pollution and identification
40 of the sites before later hosting the socio-economic infrastructures.

41 **Keywords:** DRASTIC MODEL; Fuzzy Concepts; Groundwater Vulnerability; Senegal basin; Mali

42 Introduction

43 A key component to building a territory is the vulnerability map. It's a fundamental water quality
44 assessment document that aids the development of underground water resources. Among the
45 myriad of functions delivered by a Geographic Information Systems are its capability for multi-
46 criteria analysis, a feature that is essential for developing the vulnerability maps for an aquifer
47 system. Water quality information is a basic data requirement for implementing any water
48 management decisions. It provides necessary information for assessing risk of groundwater
49 pollution, and remediation measures needed to control future pollution level. These set of
50 information could be retrieved from the groundwater pollution vulnerability maps. The
51 assessment of the vulnerability of groundwater to pollution, 24 methods exist, which are
52 classified into three groups; • Comparison methods: used mainly for very large study areas and
53 takes into consideration 2-3parameters;

54 • Methods of analog relationship and numerical models: based on simple or complex
55 mathematical laws. Recommended for assessing the vulnerability of radioactive sites;

56 • Method of parametric systems: it is composed of three sub systems:



- 57 o The matrix system: This system, adapted for local use, is based on a limited number of
58 parameters judiciously chosen. The procedure is a combination of classes to define descriptively
59 the vulnerability of aquifers;
- 60 o The class system: for this group, to define a range for each parameter considered necessary for
61 assessing vulnerability, then subdivides each of the intervals selected based on the variability of
62 the parameter. The final score resulting from the summation (or multiplication) of each score for
63 the different parameters should be divided by the number of classes chosen.
- 64 o Weighted class system: this group of methods is based on assigning ratings to the parameters
65 which are retained as necessary for the evaluation of groundwater vulnerability by defining
66 intervals as is the case with other methods cited previously. Subsequently a weight is applied for
67 each parameter according to its importance in the assessment of vulnerability.

68 Water is one of the most important things we need for our daily life. Nowadays water
69 management is coming more and more a big problem because of many reasons as climate,
70 pollution, environmental issues, etc. So, many surface water and groundwater are polluted.
71 Water system is a cycle. So water in air, water on the land and water under the land are all
72 connected Groundwater and surface water are connected through a very complicated
73 hydrogeological system, that can lead to a mutual contamination which means that if
74 groundwater is polluted, it can affect the upper surface water and if surface water is polluted, it
75 can affect the underlying groundwater too.

76 Sustainable management of the Senegal River basin resources is a major issue for the four
77 riparian countries which are Guinea, Mali, Mauritania and Senegal.

78 The multiple uses of water and the multinational nature of the basin led the riparian countries to
79 create the Organization for the Development of the Senegal River (OMVS in french), to sound
80 management of the basin's water resources. For this, each country needs data and information
81 enabling it to monitor and predict the evolution of the resource, also in view of the importance of
82 climate variability in the region marked by the recurrence of drought, the potential impacts of
83 climate change and the increasing impacts of population pressure on water resources. Many other
84 water uses in the basin also require data and information for their activities.

85 The Senegal River Basin in Mali is increasingly dominated by cultures and industries using
86 chemicals. This strong demand for chemicals threatens the quality of groundwater resources.
87 Groundwater reserves are substantial and are being used to cover different needs. They are also
88 used as source of drinking water in the region experiencing rapid population growth with a
89 growth rate of 3% per year (OMVS, 2013). The quality of this groundwater resource is
90 constantly put to the test, because of the growth of both point and diffuse pollution sources. To
91 prevent the risk of pollution of groundwater, an adapted approach is the knowledge vulnerable
92 areas to pollution. Civita(1994) showed that aquifer groundwater's changes(in quality and
93 quantity) in time and space are due to natural process and/or human activities.



94 The work already done in the area (Newton, Joshua T, 2007; UNESCO 2012), mainly concern
95 the quantity, and water resources management. Other studies (Anoh, 2009; Jourda et al., 2007)
96 have focused on the quality of water resources but not in the same exact area or not to found the
97 vulnerability zones.

98 However, none of these studies has been the event of the impact of human and natural activities
99 on groundwater resources in the basin of the river Senegal to Mali. Thus, the present study uses
100 fuzzy and Drastic methods which evaluate the intrinsic and specific vulnerability to pollution to
101 highlight those impacts.

102 The aim of our study is to find useful and relevant information to guide policy choices for
103 prevention and management of risks of pollution of groundwater resources in this area by a
104 sustainable management.

105 **MATERIALS ET METHODS**

106 The working material consists of multiple data sources. This is the piezometric data from
107 piezometric champagne conducted in different years in the region and complemented by those of
108 the database "sigma" of the National Water Directorate (DNH).
109 Drilling data sheets available provided by the various campaigns of supply of drinking water as
110 well as the National Water Laboratory (LNE) allowed to use the drilling depth data, groundwater
111 levels, lithological cuts and pumping test ... These data helped to the achievement of several
112 maps of vulnerabilities.

113 To these data, add map information with the geological map of the region and that of the soil
114 sketch of Mali provided by FAO's work.

115 Thus, the coordinates of Shuttle Radar Topography Mission or SRTM picture
116 (<http://srtm.csi.cgiar.org>) was used for the cover of the study area. His treatment has established
117 a digital elevation model (DEM) resolution of 90 m and highlights the topographic map.

118 The processing of all this data is performed on ArcGIS 10.0 for cartographic processing,
119 processing of satellite images and to generate the slope map and the combination of other
120 thematic maps.

121 For this study we used two different methods: one to assess the intrinsic vulnerability
122 (DRASTIC) and the second to find the specific vulnerability (Fuzzy).

123 The DRASTIC method is a method for mapping the inherent vulnerability of aquifers.

124 This method has already been the subject of several applications through the literature. Mohamed
125 (2001) evaluated aquifer vulnerability to pollution in El Madher (Algeria); Murat et al. (2003)
126 assessed the south-western aquifer pollution in Quebec (Canada); Jourda et al. (2006) and
127 Kouame et al. (2007) also used DRASTIC method to assess respectively Korogho (northern
128 Cote d'Ivoire) and Bonoua (southern Cote d'Ivoire) aquifers vulnerability to pollution. Although
129 if it often changed (Hamza et al., 2007), it remains effective as the vulnerability assessment tool.

130 To test this ability it has been associated to the fuzzy method, which is one of these variants.

131 The joint application of the two methods has the advantage of ensuring complementarity in
132 evaluating the vulnerability of groundwater to pollution. These methods are in the form of
133 numeric rating system, based on the consideration of the various factors influencing the
134 hydrogeological system. In the assessment of the vulnerability process, seven parameters of



135 interest to both the two methods including the depth of the water level, the effective recharge of
136 the aquifer, soil types, topography, impact of vadose zone or the effect of self-purification of the
137 vadose zone, the lithology of the aquifer and the hydraulic conductivity of the aquifer.
138 The drastic method uses formulas that experiment the linear relationship between the parameters,
139 while the fuzzy method uses formulas that take into account the continuity in pollution from one
140 point to another.

141 **Vulnerability assessment by the DRASTIC method**

142 The DRASTIC method is one of weighted classes, which was developed by 'The US
143 Environmental Protection Agency (EPA)' and the 'National Water Well Association (NWWA)'
144 in 1987 to evaluate the groundwater vulnerability to pollution.

145 Although it is not originally designed for Geographic Information Systems, this model is a
146 classic spatial analysis widely used in GIS.

147 The objective of DRASTIC is to give a standard methodology that gives reliable results for
148 efforts to protect groundwater.

149 DRASTIC generates an index or 'score' for the potential pollution of ground water resources.
150 This index covers the entire range from 23 to 226. Note that the vulnerability to pollution is
151 higher for higher notes.

152 The DRASTIC method uses seven hydrological parameters: the depth of the water level of the
153 water table [D], the net recharge [R], the lithology of the aquifer [A], the soil texture [S], the
154 topography slope of the field- [T], the impact of the unsaturated zone [I] and finally the hydraulic
155 conductivity or permeability of the saturated zone [C].

156 In GIS, each parameter is scored on a layer by assigning a weight coefficient corresponding to
157 the parameter, that is to say, its influence on the vulnerability of the aquifer. Then these layers
158 are superimposed on a layer where result will be calculated the index DRASTIC said 'DRASTIC
159 Pollution Index (DPI)'. The layers will obviously have the same cartographic features: a single
160 projection system, identical units of length, identical geographical area and also the same
161 resolution, because this system uses matrix format for all calculations.

162 DPI is dimensionless. The number or the order of magnitude has no meaning in itself. The unity
163 of the DPI occurs when comparing two sites or a site to several other sites. The site with the
164 highest DPI will be considered most susceptible to contamination or pollution.

165 More than 24 vulnerability assessment methods of groundwater to pollution are identified in the
166 international literature. The method most currently used in the world is the DRASTIC method.

167 It is a method that was developed by L. Aller et al in 1987 and is one of the assessment methods
168 (Vulnerability aquifers) Weighted based and assigning a rating to used different parameters
169 (generally between 1 and 10). A Weighting is also allocated according to the relative importance
170 of each of the parameters used. The DRASTIC numerical rating system incorporates seven
171 different physical parameters involved in the transportation process and mitigation of
172 contaminants: water depth, effective recharge, aquifer, soil type. Step 1: A numerical value
173 ranging from 1 to 5 is allocated to each of 7 parameters (parametric Weight D_p , R_p , A_p ...),
174 topography, vadose zone and hydraulic conductivity of aquifer media. Each of these parameters
175 is a weight (predetermined value) of between 1 and 5, which reflects the importance of the
176 parameter in the transport processes and contaminant attenuation. A key parameter is assigned a
177 weight equal to 5 while a setting with less impact on the fate of a contaminant is assigned a
178 weight of 1. 2nd step: At each of the seven parameters is assigned a value ranging from 1 to 10,



179 defined in terms of ranges of values. The smallest value represents lower vulnerability conditions
180 to contamination (Dc, Rc, Ac ...). For each hydrogeological unit, the seven parameters must then
181 be evaluated to give each a rating that can vary from 1 to 10. A rating of 1 corresponds to the
182 least condition of vulnerability while a rating of 10 reflects the most likely to be contaminated
183 conditions. Step 3: DRASTIC is an acronym, where each letter represents one of the seven
184 factors that highlights DPI (Bezalgues et al., 2002): the depth to the water table (D); the effective
185 aquifer recharge (R); the aquifer material (A); the type of soil (S); the slope or topography of the
186 landscape (T); the impact of vadose zone (I) and the permeability or hydraulic conductivity of
187 the aquifer (C).

188 All parameters were reclassified in ArcMap and assigned a score based on rankings ranging from
189 1 to 10 and a weighting to help merge factors together in the DRASTIC equation in GIS. Each of
190 the seven parameters was then assigned a multiplicative factor (w) sets ranging (weight) from a
191 value of 5 for the most significant factors and to 1 for factors that are less so.

192 The DPI was determined according to equation (1) according to Osborn et al. (1998): (Where D,
193 R, A, S, T, I, and C are the seven parameters of the DRASTIC method, "w" is the weight of the
194 parameter and "r" the associated rating). The weights of the parameters of DRASTIC method
195 used (Table 1) are those defined by Go et al. (1987). The reference values of the index
196 DRASTIC used are those provided by Engel et al. (1996) and represent the measurement of the
197 hydrogeological aquifer vulnerability.

198 (1)

$$DPI = D_r D_w + R_r R_w + A_r A_w + S_r S_w + T_r T_w + I_r I_w + C_r C_w$$

199

200 Or (2)

$$DPI = \sum_{k=1}^7 R_k W_k$$

201 Where R is the rating (1 to 10), W is the weight (1 to 5) and k is the parameter (1 to 7)

202 In the final step, the calculation of the DRASTIC index to each hydrogeological unit is obtained
203 by the sum of the products of each side by its weight. DPI represents the level of risk of the
204 aquifer unit to be contaminated. It can take a maximum of 226 (100%) and a minimum value of
205 23 (0%).

206 Polygon maps were initially generated for all the seven DRASTIC maps by geo-referencing,
207 digitizing, and editing.

208 These polygon maps were classified according to their importance on aquifer pollution potential
209 value from 0 to 10 was assigned to each map). So for each parameter we created specific
210 polygon maps by adding these ratings to attribute table in GIS. Specific polygon maps were then



211 converted into raster maps according to their ratings. We assigned weight to these raster maps
212 and combined them then to get the final vulnerability map by using formula (1 or 2).

213 DRASTIC method is frequently used to study groundwater vulnerability. In United States
214 Hearne et al. (1992); Merchant J.W (1994);Atkinson et al. (1994); Kalinski et al., (1994) used
215 this method to assess groundwater vulnerability.

216 The DRASTIC model was already used in many other countries worldwide. It was used for the
217 assessment of groundwater pollution in Anekal Taluk 9n semi-arid area of Bangalore district
218 (Chandrashekhar et al., 1999).

219 Jha et al. (2005) used DRASTIC method to assess Ranchi, Jharkland groundwater vulnerability.

220 To assess DRASTIC parameters we need to identify and study every hydrogeological and
221 meteorological conditions of the study area (Anwar et al., 2003; M. H. Hamza et al. 2006)

222 The following parameters were used for the DRASTIC method:

223 **Depth to water table (D):**

224 It is the distance between ground surface and groundwater table. So it controls the thickness and
225 amount of possible contaminants (Ckakaraboty S et al., 2007). Hence when this distance is high
226 then it is more difficult for surface water to cross (under chemical, biological reactions) all this
227 thickness and to reach groundwater.

228

229 We got depth to water table data from borehole data given by National Directorates in charge of
230 water resources management in Mali.

231 These data show that the depth varies from 1.50m to more than 120m. As said Dhundi et al.
232 (2009), for depth beyond 100 m, we assigned a rating of 0 because it is almost impossible for
233 pollutant to reach groundwater, due to processes like, sorption, filtration, biodegradation,
234 volatilization... Table 1 shows all the values for range and rating for depth to groundwater table,
235 and it map is shown in figure 1.

236 To generate the map we used the inverse distance moving average and a simple inverse power
237 with a limiting search distance of 7 000 m including a high number of input points to get a good
238 accuracy. We assigned sensitivity rating values as did Dhundi et al.(2009): 10 for depth (<1.5
239 m), 9 for depth (1.5–4.6 m), 7 for depth (4.6–9.1 m), 5 for depth (9.1–15.2), 3 for depth (15.2–
240 22.5 m), 2 for depth (22.5–30 m) and 1 for depth (>30 m and the region having no data).

241 **Recharge (R):**

242 The annual average amount of water that infiltrates the vadose zone and reaches the water table
243 (Aller et al. 1987), groundwater recharge or net recharge is the movement of water from ground
244 surface to groundwater. It can easily bring contaminant to groundwater. So, recharge value
245 increases with aquifer vulnerability potential because dispersion, dilution, etc will increase in
246 unsaturated zone also. There are many sources of recharge in the study area including
247 precipitation, irrigation, waste water, return flow, infiltration from surface water (rivers, springs
248 etc.).



249 Net recharge data was taken from hydrogeological synthesis of Mali (Mali Groundwater
250 Resource Investigation, 1990). The different values of net recharge are in table 2. Figure 2
251 represents the recharge map.

252 We used the following formula to calculate net recharge:

253

254 $\text{Net recharge} = (\text{rainfall} - \text{evaporation}) \times \text{recharge rate}$

255

256 **Aquifer media (A):**

257 Aquifer media was defined by many researchers in the world: Aquifer media designates the
258 consolidated and unconsolidated rocks which serve as water storage (Chandrashekhar et al.,
259 1999). According to Heath (1987) an aquifer is a subsurface rock or sediment unit that will yield
260 usable quantities of water to a well or spring. The aquifer is also defined as a rock formation
261 which can yield sufficient quantities of water for use (Anwar et al., 2003). It is very important in
262 attenuating the pollution because it is the media where all reactions take place and grains size
263 and sorting are very important in pollutant attenuation. Also the aquifer media governs flow path
264 and length in an aquifer. Hence Piscopo (2001) indicates that the duration of time available for
265 attenuation is determined by the path length. In this study, we used topographical map and well
266 log data to prepare the aquifer media map. We assigned a high rating values to coarse media and
267 low values to finer media. With the Mali hydrogeological synthesis maps and report on Senegal
268 Basin groundwater simulations, the aquifer media data (table 3) for this research were computed
269 (figure 3) from more than 2300 borehole data.

270 **Soil media (S):**

271 Soil media is the uppermost part of unsaturated zone. The quantity and shrink/swell capacity of
272 clay in soil, soil grain type, sorting and size are both important because they influence
273 groundwater movement, potential dispersion, pollutants migration throughout biological and
274 physico-chemical reactions (sorption, biodegradation, ionic exchange, oxidation, reduction...).
275 The permeability of the soil media was used as basis for assigning ratings on a scale of 1 to 10.
276 The coarsest soils were assigned a rating of 10 and this decreased all the way to the finest media,
277 which were assigned a rating of 1. Details for rating and index are shown on table 4 while soil
278 map is shown on figure 4.

279 **Topography (T):**

280 Topography of an area accounts for the change in slope. It is a determining factor of how rainfall
281 and pollutants will either run-off or infiltrate (Lynch et al., 1994). The longer the water and or
282 pollutant get retained in an area, the greater the chance for infiltration and consequently, the
283 potential for recharge is higher. Gentler slopes (slopes of 0-2 (%)) have higher retaining capacity
284 for water and/or pollutants while steeper slopes (slopes of +18(%)) have lower retention capacity
285 for water and or pollutants. According to Aller et al., 1987, topography has an effect on
286 attenuation since it influences soil development.
287 Slope values extracted from the digital elevation model of the study area were reclassified and
288 ranked on a scale (table 5) of 1 to 10 to build the topography map (figure 5). This served as basis
289 to be included in the multi-criteria analysis, where other DRASTIC factors play a role.

290 **Impact of vadose zone (I):**



291 Unsaturated zone or vadose zone is situated between ground surface and groundwater table. It
292 highly impacts aquifer pollution potential by its permeability, reactions inside, etc. (Corwin, et al.,
293 1997). Because vadose zone is closely related to soil media and groundwater depth, we used the
294 formula developed by Piscopo (2001) to estimate it: (3)

$$I_r = D_r + S_r$$

295 Where: I is the impact of Vadose Zone, D is depth to water table, S is soil media and r is the
296 rating

297 For groundwater depth we chose the following ratings: 5 for depth less than 10 m, 2 for zones
298 with depth between 10 m and 30 m, and 1 for area which groundwater depth is more than 30.
299 Similarly we chose 5, 3 and 1 for respectively high, medium and low permeable soils. And
300 finally we combined the two map layers to get the impact of vadose zone layer (table 6 and
301 figure 6).

302 **Hydraulic conductivity (C):**

303 Hydraulic conductivity expresses the aquifer ability to transport contaminant (Ckarakorty S et
304 al., 2007). It plays an important role in aquifer contamination potential because an aquifer with
305 high hydraulic conductivity is easy to be contaminated and aquifer with low hydraulic
306 conductivity is difficult to be polluted (Fritch et al., 2000).

307 We used transmissivity values instead of hydraulic conductivity to build its map. We adopted the
308 following rating system: for very high values ($>450 \text{ m}^2/\text{day}$) we chose 10; for high values (300--
309 $450 \text{ m}^2/\text{day}$) we chose 8; for moderate values ($100\text{--}300 \text{ m}^2/\text{day}$) we assigned 6; for moderately
310 low values ($30\text{--}100 \text{ m}^2/\text{day}$) we assigned 4; for low values ($20\text{--}30 \text{ m}^2/\text{day}$) we chose 3; for very
311 low values ($10\text{--}20 \text{ m}^2/\text{day}$) we chose 2 and for extremely low values ($<10 \text{ m}^2/\text{day}$) we assigned
312 1 as rating value. The different values and distribution of hydraulic conductivity are shown in
313 Table 7 and figure 7.

314 **Vulnerability assessment by the fuzzy method**

315 DRASTIC method cannot consider the continuity passage from the highest polluted point to
316 lowest one, this property expresses the blurring effect of the aquifer to be potentially polluted. So
317 fuzzy concept can be utilized to evaluate the groundwater pollution potential. For instance, we
318 know that for vulnerability evaluation, when the water table is shallow, recharge rate is high, and
319 if aquifer and soil materials are coarser, groundwater potential to pollution is higher. Also if the
320 hydraulic conductivity, recharge rate and slope are low then groundwater potential to pollution is
321 low. The main concept using fuzzy logic is very simple and it expresses if a statement is true or
322 untrue and also its degree of verity or wrongness for all the inputs (Pathak et al. 2009). A function
323 of membership links all fuzzy sets. We coupled fuzzy optimized model with GIS to evaluate the
324 vulnerability degree by converting the study area into raster map and taking into account
325 membership degrees in continuous passage from highest polluted points to lowest polluted points
326 in hydrogeological settings.

327 **Optimized fuzzy model:**

328 The fuzzy nature of groundwater vulnerability and groundwater vulnerability assessment can be
329 considered as a particular property. For example instead of numerical measurement of factors in
330 DRASTIC method, the fuzzy method describes continuously the links between those factors that
331 affect groundwater.



332 The fuzziness can be expressed continuously by membership degree from 0 to 1. The following
 333 optimized model is used (Pathak et al. 2009):
 334 Given a matrix for factors: (4)

$$X = (x_{ij})_{7 \times n}$$

335 x_{ij} denotes the value of tester j in element i
 336 $i=1, \dots, 7; j=1, \dots, n$ with n the overall number of sampling points.

337 We can classify Drastic factors into two main groups which are:
 338 -group 1 where the increasing of parameter value increases groundwater vulnerability to
 339 pollution.
 340 -group 2 where the increasing of parameter value decreases groundwater vulnerability to
 341 pollution.
 342 This membership degree can be expressed mathematically by:
 343 For the group 1: (5)

$$r_{ij} \begin{cases} 0 & \text{if } x_{ij} \leq x_{minj} \\ \frac{x_{ij} - x_{minj}}{x_{maxj} - x_{minj}} & \text{if } x_{minj} \geq x_{ij} \geq x_{maxj} \\ 1 & \text{if } x_{ij} \geq x_{maxj} \end{cases}$$

344 For the group 2:(6)

$$r_{ij} \begin{cases} 0 & \text{if } x_{ij} \geq x_{maxj} \\ \frac{x_{maxj} - x_{ij}}{x_{maxj} - x_{minj}} & \text{if } x_{minj} \geq x_{ij} \geq x_{maxj} \\ 1 & \text{if } x_{ij} \leq x_{minj} \end{cases}$$

345

346 With r_{ij} the degree of membership for the sample j in factor i
 347 $minj$ is the smallest value of element i (i.e. 1) in Drastic method.
 348 $maxj$ is the maximum value of element i (i.e. 10) in Drastic method.
 349 We can use equations (4), (5) and (6) to get the following connection of factors matrix: (7)
 350

$$R = (r_{ij})_{7 \times n}$$

351 With the following conditions in matrix R :
 352 -if $r_{ij}=1$ then the tester j has the highest potential to groundwater pollution according element i only.
 353 -if $r_{ij}=0$ then the tester j has the lowest potential to groundwater pollution according the element i only.
 354 For example when all element connection degrees to highest potential to groundwater pollution are 1,
 355 then:(6)
 356 $R_{ij}=(1, \dots, 1)$
 357 And when all element connection degrees to lowest potential to groundwater pollution are 0, then: (8)
 358 $R_{ij}=(0, \dots, 0)$
 359 So the membership degree of each or the parameters in sample j is: (9)
 360 $r_j=(r_1, \dots, r_7)T$



361 In Drastic system different parameters have different weights (from 5 to 1) in relation to vulnerability;
 362 these are normalized in evaluation process to sum to one.
 363 Let (10)
 364 $W = (w_1, \dots, w_7)^T$ the weight vector
 365 The distance from one given sample j to the sample with the highest potential to groundwater pollution
 366 can be express as: (11)

$$d_1 = \sqrt[p]{\sum_{i=1}^7 [w_i(r_{ij} - 1)]^p}$$

367 The distance from one given sample j to the sample with the lowest potential to groundwater pollution
 368 can be express as: (12)
 369

$$d_2 = \sqrt[p]{\sum_{i=1}^7 (w_i r_{ij})^p}$$

370 p in (11) and (12) is called distance factor, when $p=1$ the distances are named Hamming distances and
 371 when $p=2$ the distances are called Euclidean distances.
 372 We used Euclidean distances in our study. We can see clearly that if $d_1=0$ then the given sample j has the
 373 highest potential to groundwater pollution and when $d_2=0$ then the given sample j has the lowest potential
 374 to groundwater pollution.
 375 Let the membership degree of the highest potential to groundwater pollution be denoted by u_j for a given
 376 sample j , so the membership degree of the lowest potential to groundwater pollution will be $(1-u_j)$ for the
 377 same given sample.
 378 Membership can be regarded as weight in view of fuzzy concept. So the following equations express
 379 more clearly continuous changes from a given sample j to the highest potential to groundwater pollution
 380 as well as from the same given sample to the lowest potential to groundwater pollution: (13)

$$D_1 = u_j \sqrt[p]{\sum_{i=1}^7 [w_i(r_{ij} - 1)]^p}$$

381 D_1 is the weighted distance to the highest potential to groundwater pollution and: (14)

$$D_2 = (u_j - 1) \sqrt[p]{\sum_{i=1}^7 (w_i r_{ij})^p}$$

382 D_2 is the weighted distance to the lowest potential to groundwater pollution.
 383 To get an optimized solution for u_j the objective function is: (15)

$$\min\{F(u_j) = (D_1^2 + D_2^2)\} = u_j^2 \left\{ \sum_{i=1}^7 [w_i(r_{ij} - 1)]^p \right\}^{2/p} + (1 - u_j)^2 \left\{ \sum_{i=1}^7 [w_i r_{ij}]^p \right\}^{2/p}$$

384 After differentiating (14) and solving it comes: (16)



$$u_j = \left[1 + \left\{ \frac{\sum_{i=1}^7 [w_i(r_{ij}-1)]^p}{\sum_{i=1}^7 (w_i r_{ij})^p} \right\}^{2/p} \right]$$

385 Equation (16) is called fuzzy optimization model and higher the value of u_j , higher the potential of
386 groundwater vulnerability to pollution for a given tester j . This model is joined to GIS and used to
387 evaluate the pollution potential of groundwater. The diagram of procedures used to evaluate this
388 potential maps using DRASTIC and fuzzy methods in GIS is shown in figure 8.

389 Results and Discussions

390 Fuzzy-DRASTIC parameters:

391 Using memberships defined by fuzzy concept depth to ground water table and topography maps
392 were different from those of DRASTIC, but for the other five parameters the fuzzy optimized
393 and DRASTIC maps were identical.
394 The depth to ground water table and topographic map obtained by using fuzziness are shown in
395 figure 9 and figure 10:

396 The aquifer vulnerability maps

397 The final DRASTIC Potential Index (DPI) was obtained by using formula 1 (or 2) in ArcGIS
398 10.0 software on the seven individual map layers to produce the vulnerability map for DRASTIC
399 method. The DPI rating scores were from 72 to 141 and the greater the score, the higher the
400 aquifer vulnerability. We used natural break (jenks) classification to get three main classes
401 namely low vulnerability area ($DPI < 110$), moderate vulnerability area ($110 < DPI < 120$) and high
402 vulnerability area ($120 < DPI < 141$). Table 8 and figure 11 show DPI scores and distribution.
403 These values range from 72 to 141 and are classified into 3 distinct classes.
404 To facilitate and control scientific discussion, we used natural break (jenks) classification to get
405 three vulnerability maps for both methods: DRASTIC method, normalized DRASTIC method
406 and fuzzy DRASTIC method.
407 Under these conditions figure 11(DRASTIC method) shows that high risk area of Senegal basin
408 in Mali are mainly situated in northern and southwestern portion of the basin with 14.64% of
409 total Senegal basin in Mali. The moderate risk areas which cover 6.51% of the total basin are
410 somewhat disseminated and are mostly situated in the central and northern portion of the basin.
411 Certain moderate risk areas are seen in the north eastern and extreme west zone. All the others
412 portions of the Senegal basin in Mali are under low risk (78.85%) which are found in the western
413 and Middle Western parts regions of the basin.
414 For the normalized vulnerability we got: 21.68% for high vulnerability, 15.22% for moderate
415 vulnerability and 63.32% for low vulnerability. The map is shown in figure 12.
416 And for fuzzy DRASTIC method we got: 18.92% for high vulnerability zone, 8.94% for
417 moderate vulnerability zone and 72.11% for low vulnerability zone (figure 13).

418 However, figures 14-16 showed that coincidence ratio with nitrate high concentration for fuzzy
419 DRASTIC method is the highest (81.13%), followed by normalized DRASTIC method (79.54%)
420 and the lowest coincidence ratio for DRASTIC method (77.31%). This confirmed our
421 assertion that fuzzy method better assesses groundwater vulnerability to pollution than simple
422 DRASTIC method.



423 **Sensitivity analysis**

424 Seven hydro-geological parameters influence the transport of the contaminants to aquifers when
425 using the DRASTIC approach. According to Rosen (1994), the great numbers of parameters are
426 intended to decrease indecisions associated with using the individual parameters on the results.
427 But, several researchers (Merchant, 1994; Barber et al. 1994) opine that groundwater risk
428 assessment is possible without using all the seven parameters of the DRASTIC method. Other
429 researchers (Napolitano and Fabbri, 1996) also criticized in what way the weights and the ratings
430 for the seven parameters are assumed for DPI assessment and lead to uncertainties about the
431 precision of the outcomes for pollution risk assessment. Many factors contribute to the output of
432 the DRASTIC model (Rahman A., 2008; Ckkraborty, 2007) including map units in each layer,
433 the weights, the overlay operation type that is performed, the number of data layers, the error or
434 doubt associated to each map unit etc.

435 Sensitivity analysis was adopted to complement trial evidence for DRASTIC method to perfect
436 the uncertainty about model precision.

437 Two (2) sensitivity analyses were then done (Babiker et al. 2005; Lodwick et al. 1990): the map
438 removal sensitivity test and the single parameter sensitivity analysis.

439 The map removal sensitivity test defines the sensitivity of risk map to each parameter by
440 eliminating a single or more layer map and is applied using the following equation: (17)

$$S = \left(\frac{\left| \frac{V - V'}{N - N'} \right|}{V} \right) * 100$$

441
442 With S the sensitivity degree, V and V' are the unperturbed and the perturbed risk indices, N and
443 n define the number of data layers used to calculate V and V'. The unperturbed risk index defines
444 the real index found by using altogether the seven parameters while the perturbed risk index can
445 have a smaller number of parameters for the calculation procedure.

446 To estimate the impact of individual parameter on the risk potential, we used the single
447 parameter sensitivity test. During this test we compared the effective or actual weight of each
448 individual parameter with its theoretical or assigned weight by using the following formula: (18)

$$W = \frac{P_r * P_w}{V} * 100$$

449
450 W = effective weight of the parameter, Pr = Rating, Pw = Weigh, V = Vulnerability index
451

452 The statistical summary of all parameters are shown in table 8 and table 9. We noted that using
453 DRASTIC method and equation 17 the highest vulnerability source is topography which has a
454 mean value of 9.83. The second main parameter affecting the risk is impact of vadose zone with
455 8.14, followed by soil media (5.71). After vadose zone comes depth to groundwater table with
456 5.52 as mean value. The fifth and the sixth positions are occupied respectively by aquifer media
457 (4.27) and hydraulic conductivity (1.93) for their contribution to groundwater pollution potential.
458 Finally net recharge showed the least mean value for contribution to pollution risk in Senegal
459 basin in Mali.

460 The effective weight also called coefficient of variation (equation 18) shows that the main two
461 parameters which impact the most DPI values are the unsaturated zone (or vadose zone) with
462 35.92% and depth to groundwater table with 24.17%. They are followed by aquifer media
463 (11.25%), soil media (10.04%) and topography (8.73%). Hydraulic conductivity and net recharge



464 have relatively low variations with respectively 5.09% and 4.80%. A low percentage means a
465 small influence on variation on DPI across the basin.
466 Table 8 shows statistics and the correlation on the seven parameters used in both Drastic and
467 fuzzy model. The mean values of parameters reveal that vadose zone contributes the most to
468 vulnerability index with a mean value of 35.90% for Drastic and 0.79 for fuzzy membership.
469 Depth to water table (24.17% and 0.5), aquifer media (11.24% and 0.36) and soil media(10.02%
470 and 0.52) have moderate contribution to final vulnerability index. And topography (8.72% and
471 0.02), hydraulic conductivity (5.08% and 0.1), recharge (4.8% and 0.04) have low contribution to
472 final vulnerability index.

473 **Map removal sensitivity analysis**

474 The first step of map removal sensitivity test shows the change in DPI value when we remove
475 only one map layer a time. Table 10 and table 11 give the calculation results. Because the overall
476 mean variation is not more that 1% the test does not describe very clearly DPI variation when
477 removing only one map layer a time, also all mean values are almost the same here. But the
478 maximum value of DPI variation was estimated when we removed unsaturated zone parameter
479 map with a relative mean variation of 3.60%. This can be explained by its relative high
480 theoretical weight in DRASTIC method and the nature of unsaturated zone material in the basin.
481 Moderate variations were seen after removal of depth to groundwater table (1.72%), net recharge
482 (1.58%) and hydraulic conductivity (1.53%). Only minor variations in mean values of DPI were
483 remarked (from 0.67% to 0.92%) after removal of each of the other parameters from
484 computation (table 10).

485 The second step of map removal sensitivity test shows the change in DPI value when we remove
486 one or more map layers (or parameters) a time from calculation. Based on the first step we
487 removed parameters in the second step (Rahman A., 2008; Babiker I.S et al. 2005) by removing
488 preferentially the parameters which produced less variation on the final DPI value and then next
489 smaller etc.

490 The smallest mean effective weight variation was seen after removal of net recharge (4.80%)
491 from de calculation. The more we remove data layers from calculation the more the mean
492 variation value increases because we keep the most effective parameters each time (Babiker I.S
493 et al. 2005)..

494 **Single parameter sensitivity analysis (effective weight)**

495 The importance of each of the seven parameters has been shown in map removal sensitivity
496 analysis. Now we need to understand if the theoretical weight affected to each parameter in
497 DRASTIC model is its actual/real or effective weight after computation.
498 The effective weight is a function of the value of the single parameter with regard to the other six
499 parameters as well as the weight assigned to it by the DRASTIC model (Rahman A., 2008;
500 Babiker 2005). The single parameter sensitivity analysis data are in table 12. The theoretical
501 weights of both impact of vadose zone and depth to groundwater table are 21.73% but their
502 effective weights are respectively 35.92% and 24.17%. Because their effective weight is higher
503 than their theoretical weight we can say that they are the two most effective parameters in this
504 DPI assessment. The soil media parameter (10.04%) and topography parameter (8.73%)



505 similarly indicate great effective weight in comparison to their theoretical weight (8.69% and
506 4.34% respectively). In contrary, the other three parameters presented lesser effective weight.
507 The importance of the four most effective parameters focuses on the need of precise data for
508 building the model. And the low recharge and hydraulic conductivity values in Senegal basin
509 contributes to reduce the significance of these parameters in its groundwater vulnerability
510 assessment.
511 This study has demonstrated the closed and linearly relationship between sensitivity analysis and
512 fuzzy membership (table 9). So instead of sensitivity analysis, we can also use fuzzy membership
513 to find the main parameters which influence the GW potential vulnerability to pollution.

514 **Conclusion**

515 Basically, analyses were done with the purpose of observing the correlation between the intrinsic
516 risk evaluation outcome and groundwater pollution in Senegal basin in Mali. DPI main values
517 were low, moderate and high. In this study, a methodology was adopted to improve DPI
518 calculation to produce pollution potential map. This was achieved by including the homogeneous
519 nature of vulnerability to pollution using DRASTIC factors in a vast area. In addition, field
520 measured nitrate data were used to confirm risk to pollution map of Senegal basin. So we can say
521 that passing from easiest to most difficult groundwater to be polluted can be continuous. This
522 proves in fact the fuzzy nature of risk to groundwater pollution. So, combined GIS built fuzzy
523 design model produces the continuous risk assessment function different stage DRASTIC index
524 more accurate than the simple DRASTIC method. We compared simple DRASTIC, normalized
525 DRASTIC and fuzzy DRASTIC outputs and it appeared that fuzzy index coincides the most with
526 nitrate distribution in the study area. The outputs show that 18.92% of the study area's
527 groundwater aquifer are under high risk to pollution due to fuzzy DRASTIC while 14.64% of the
528 study area's groundwater aquifer are under high risk to pollution from simple DRASTIC method.
529 From this outcome, it can be established that risk assumed by fuzzy method is more consistent
530 than DRASTIC method. For several aspects of the local and regional groundwater resources
531 protection and management, the groundwater risk to pollution maps established in this work are
532 important tools in policy and decision making.

533 **References**

- 534  Aller, Linda, Truman Bennett, Rebecca J. Petty, and Glen Hackett. (1987). "DRASTIC: A
535 STANDARDIZED SYSTEM FOR EVALUATING GROUND WATER POLLUTION
536 POTENTIAL USING HYDROGEOLOGIC SETTINGS". CR-810715. U.S Geological
537 Survey: Environmental Protection Agency.
- 538 Anoh KA. (2009). Évaluation de la vulnérabilité spécifique aux intrants agricoles des eaux
539 souterraines de la région de Bonoua. Mémoire de DEA, Université de Cocody, Cocody, p.68.
540 Anwar, Prem, & Rao, M. Anwar, C.C. Prem and V.B. Rao. (2002). Evaluation of groundwater
541 potential of Musi River catchment using DRASTIC index model. In: B.R. Venkateshwar, M.K.
542 Ram, C.S. Sarala and C. Raju, Editors, *Hydrology and watershed management. Proceedings of
543 the international conference 18–20, 2002*, B. S.Publishers, Hyderabad (2003), pp. 399–409.



- 544 Atkinson, S et al. (1994). An examination of groundwater pollution potential through GIS
545 modeling. ASPRS/ACSM.
- 546 Babiker IS, Mohammed MAA, Hiyama T, & Kato K. (2005). A GIS-based DRATIC model for
547 assessing aquifer vulnerability in Kakamigahara Heights, Gifu Prefecture, central Japan. *Sci*
548 *Total Environ* 2005; 345: 127–140.
- 549 Barber, C., Bates, L.E., Barron, R., Allison, H. (1994). Comparison of standardized and Region-
550 specific methods assessment of the vulnerability of Groundwater to pollution: A case study in an
551 Agricultural catchment. In: Proceedings of 25th IAH Congress water Down under, Melbourne,
552 Australia.
- 553 Bezelgues S, Des Garets E, Mardhel V, Dörfli N. (2002). Cartographie de la vulnérabilité de
554 Grand-Terre et de Marie-Galatie (Guadeloupe). Phase 1 : Méthodologie de détermination de la
555 vulnérabilité.
- 556 Chandrashekhar H, Adiga S, Lakshminarayana V, Jagdeesha CJ, Nataraju C. (1999). A case
557 study using the model 'DRASTIC' for assessment of groundwater pollution potential. In
558 Proceedings of the ISRS national symposium on remote sensing applications for natural
559 resources. June, 1999: 19–21, Bagalore.
- 560 Civita M. (1994). *La Carte Della Vulnerabilità Degli Acquiferi All'inquinamento: Teoria e*
561 *Pratica*. PITAGORA (Editeurs):Bologna.
- 562 Ckakraorty, S. (2007). Assessing aquifer vulnerability to arsenic pollution using DRASTIC and
563 GIS of North Bengal Plain: A case study of English Bazar Block, Malda District, West Bengal,
564 India, Vol.7, No.1, Springer.
- 565 Corwin, D. L., Vaughan, P. J., and Loague, K. (1997). Modeling nonpoint source pollutants in the
566 vadose zone with GIS, *Environmental Science and Technology* 31(8), 2157–2175.
- 567 Dhundi Raj Pathak, Akira Hiratsuka, Isao Awata, Luonan Chen.(2009). Groundwater
568 vulnerability assessment in shallow aquifer of Kathmandu Valley using GIS-based DRASTIC
569 model. *Environmental Geology*, pp 1569-1578.
- 570 Engel BA, Navulur KCS, Cooper BS, Hahn L. (1996). Estimating groundwater vulnerability to
571 non point source pollution from nitrates and pesticides on a regional scale, *Int. Assoc. Hydrol.*
572 *Sci. Publi.*, 235: 521-526.
- 573 Fritch TG, McKnight CL, Yelderman Jr JC, & Arnold JG. (2000). An aquifer vulnerability
574 assessment of the paluxy aquifer, central Texas, USA, using GIS and a modified DRASTIC
575 approach. *Environ Manage* 2000;25:337–345.
- 576 Hamza MH, Added A, Frances A, Rodriguez R. (2007). Validité de l'application des méthodes de
577 vulnérabilité DRASTIC, SINTACS et SI à l'étude de la pollution par les nitrates dans la nappe
578 phréatique de Metline-Ras Jebel-Raf Raf (Nord-Est Tunisien). *Geoscience*, 339: 493-505.
- 579 Hearne, G et al. (1992). Vulnerability of the uppermost groundwater to contamination in the
580 greater Denver Area, Colorado. *USGS water resources investigation report* 92-4143, 241pp.
- 581 Heat, R.C. (1987). Basic Groundwater Hydrology. US Geological Survey Water Supply paper
582 2220, U.S. Department of the Interior, *US. Geological Survey*. Hendrix W.G et al. (1986).
583 Geographic information system technology as a tool for groundwater management., Vol3. *Fall*
584 *church: American Society of Photogrammetric and Remote Sensing*, pp. 230-239
- 585 Hendrix W.G et al. (1986). Geographic information system technology as a tool for groundwater
586 management., Vol3. *Fall church: American Society of Photogrammetric and Remote Sensing*,
587 pp. 230-239.
- 588 Jha Manish Kumar.(2005). Vulnerability Study Of Pollution Upon Shallow Groundwater Using
589 Drastic/GIS.



- 590 Jourda JP, Saley MB, Djagoua EV, Kouame KJ, Biemi J, Razack M. (2006). Utilisation des
591 données ETM+ de Landsat et d'un SIG pour l'évaluation du potentiel en eau souterraine dans le
592 milieu fissure précambrien de la région de Korhogo (nord de la Côte d'Ivoire) : approche par
593 analyse multicritère et test de validation. *Revue de Télédétection*, 5(4): 339-357.
- 594 Jourda JP, Kouame KJ, Adja MG, Deh SK, Anani AT, Effini AT, Biemi J. (2007). Evaluation du
595 degré de protection des eaux souterraines : vulnérabilité à la pollution de la nappe de Bonoua
596 (Sud-Est de la Côte d'Ivoire) par la method DRASTIC. Actes de la Conférence Francophone.
597 SIG 2007/ 10 au 11 Octobre 2007, Versailles-France, 11p.
598 www.esrifrance.fr/SIG2007/Cocody_Jourda.htm.
- 599 Kalinski, R et al. (1994). Correlation between DRASTIC vulnerabilities and incidents of VOC
600 contamination of municipal wells in Nebraska. *Groundwater*, 32(1), 31-34
- 601 Kouame KJ. (2007). Contribution à la Gestion Intégrée des Ressources en Eaux (GIRE) du
602 District d'Abidjan (Sud de la Côte d'Ivoire) : Outils d'aide à la décision pour la prévention et la
603 protection des eaux souterraines contre la pollution. Thèse de Doctorat, Université de Cocody,
604 Cocody, p.250.
- 605 Lodwik WA, Monson W, Svoboda L. (1990). Attribute error and sensitivity analysis of maps
606 operation in geographical information systems—suitability analysis. *Int J Geograph Inf Syst*;
607 4:413–428.
- 608 Lynch SD, Reynders AG, & Schulze RE. (1994). Preparing input data for a national-scale
609 groundwater vulnerability map of southern Africa. Document ESRI 94.
- 610 M.H. Hamza. (2006). A GIS-based DRASTIC vulnerability and net recharge reassessment in an
611 aquifer of a semi-arid region (Metline-Ras Jebel-Raf Raf aquifer, Northern Tunisia).
- 612 Merchant JW. (1994). GIS-based groundwater Pollution hazard assessment a critical review of
613 the DRASTIC model. *Photogramm Eng Rem S*; 60(9):1117–1127.
- 614 Mohamed RM. (2001). Evaluation et cartographie de la vulnérabilité à la pollution de l'aquifère
615 alluvionnaire de la plaine d'El Madher, Nord-Est algérien, selon la méthode DRASTIC. *Sciences*
616 *et changement planétaires. Sécheresse*, 12(2): 95-101.
- 617 Murat V, Paradis D, Savard MM, Nastev M, Bourque E, Hamel A, Lefebvre R, Martel R.
618 (2003). Vulnérabilité à la nappe des aquifères fracturés du sud-ouest du Québec : Evaluation par
619 les methods DRASTIC et GOD. *Ressources Naturelles Canada, Commission Géologique*.
- 620 Napolitano P, Fabbri AG. (1996). Single parameter sensitivity analysis for aquifer vulnerability
621 assessment using DRASTIC and SINTACS. In: *Proceedings of the 2nd HydroGIS conference*.
622 IAHS Publication, Wallingford; 235:559–566.
- 623 Newton, Joshua T. (2007). Case study of Transboundary Dispute Resolution: Organization for the
624 Development of the Senegal River (OMVS). *Transboundary Freshwater Dispute Database*,
625 Oregon State University. [Online] 2007. [Cited: February 26, 2012].
626 <http://www.transboundarywaters.orst.edu/research/case_studies/OMVS_New.htm>
627 O.M.V.S.(2013) le Journal N.8 Octobre.
628 <http://www.portail-omvs.org/actualite/omvs-journal-ndeg-8-octobre>.
- 629 Osborn NI, Eckenstein E, Koon KQ. (1998). Vulnerability assessment of twelve major
630 aquifer in Oklahoma. *Oklahoma Water Resources Boards, Technical Report*.
- 631 Pathak, D.R., Hiratsuka, A., Awata, I., Chen, L. (2009). Groundwater vulnerability assessment in
632 shallow aquifer of Kathmandu Valley using GIS based DRASTIC model. *Environmental*
633 *Geology* 57 (7), 1569e1578. doi:10.1007/s00254-008-1432-8.
- 634 Piscopo, G. (2001). *Groundwater Vulnerability Map Explanatory Notes - Castlereagh*
635 *Catchment. Parramatta NSW: Australia NSW Department of Land and Water Conservation*.



- 636 Rahman, A., 2008. A GIS based DRASTIC model for assessing groundwater
 637 vulnerability in shallow aquifer in Aligarh, India. Applied Geography 28
 638 (1), 32e53.
 639 Rosen, L. (1994). Study of the DRASTIC methodology with the emphasis on Swedish
 640 conditions. 37th conference of the International Association for Great Lakes Research and
 641 Estuaire Research Federation. Buffalo, p. 166
 642 UNESCO. (2012). World Heritage related Category.
 643 <http://whc.unesco.org/en/news/874>.
 644 Vrba J, Zaporozec A. (1994). Guidebook on mapping groundwater vulnerability, international
 645 contributions to hydrology. Heinz Heise, Hannover; 16: 131.
 646 Zhang R, Hamerlinck JD, Gloss SP, Munn L. (1996). Determination of nonpoint source pollution
 647 using GIS and numerical models. J Environ Qual; 25(3):411–418.

648
 649 Table 1: Range and Rating for Depth to Water

Range (m)	Rating	Index
≤ 1.5	10	50
1.6 – 4.6	9	45
4.6 – 9.1	7	35
9.1 – 15.2	5	25
15.2 – 22.5	3	15
22.5 – 30	2	10
≥ 30	1	5

650 Weight: 5

651 Table 2: rang and rating for net recharge

Range(mm/a)	rating	index
20-50	1	3
50-100	3	9
100-300	6	18

652 Weight:3

653 Table 3: Range and Rating for Aquifer Media

Range	Rating	Index
Silty sand	3	9
Fine Sand	4	12
Medium Sand	6	18
Coarse Sand	8	24
Gravel and Sand	9	27
Gravel	10	30

654 Weight: 3

655 Table 4: Range and Rating for soil media

Range	Rating	Index
Gravel	10	20
Sand	9	18
Sandy loam	6	12



656

Loam	5	10
Silty-loam	4	8
Clay-loam	3	6

Weight: 2

657 Table 5: Range and Rating for Topography(slope)

Range (%)	Rating	Index
0-2	10	10
2-4	9	9
10-12	5	5
14-16	3	3

658 Weight: 1 (Source Ckakaraboty S et al. 2007)

659 Table 6: Range and rating for vadose zone

Range	Rating	Index
Clay and Silt	3	15
Sandy/ Clay	4	20
	5	25
Clay Sand	6	30
	7	35
Sand and Gravel	8	40
	9	45
	10	50

660 Weight: 5

661 Table 7: Range and Rating for hydraulic conductivity

Range (transmissivity)	Rating	Index
<10 m ² /d	1	4
10-20 m ² /d	2	8
20-30 m ² /d	3	12
30-100 m ² /d	4	16

662 Weight = 3

663 Table 8: DRASTIC parameters

DRASTIC parameters	Ranges	Rating	Index	Weight
	0-1.5	10	50	
	1.5-4.6	9	45	
	4.6-9.1	7	35	



Depth to gw(m)	9.1-15.2	5	25	5
	15.2-22.5	3	15	
	22.5-30	2	10	
	>30	1	5	
Net recharge(mm/a)	0-50	1		4
	50-100	3		
	100-175	6		
	175-225	8		
	>225	9		
Aquifer media	Silty sand	3	9	3
	Medium sand	6	18	
Soil media	gravel	10	20	2
	Sandy loam	6	12	
	Loam	5	10	
	Clay loam	3	6	
Topography (%)	0-2	10	10	1
	2-4	9	9	
	10-12	5	5	
	14-16	3	3	
Impact of vadose zone (soil+recharge)	15-18	10	50	5
	13-15	9	45	
	10-13	8	40	
	8-10	7	35	
	6-8	5	25	
	4-6	3	15	
	<4	1		
Hydraulic conductivity (transmissivity m ² /d)	<10	1	3	3
	10-20	2	6	
	20-30	3	9	
	30-100	4	12	

664

665 Table 9: Statistical summary of the seven parameters for the two methods

	D		R		A		S		T		I		C	
	d	f	d	f	d	f	d	f	d	f	d	f	d	f
Min	1	0.33	1	0	3	0.22	3	0.22	3	0	3	0.22	1	0
Mean	5.52	0.5	1.36	0.04	4.27	0.36	5.71	0.52	9.83	0.02	8.14	0.79	1.93	0.10
Max	7	1	3	0.22	6	0.55	10	1	10	0.77	10	1	4	0.33
SD	1.41	0.16	0.77	0.08	1.48	0.16	2.20	0.24	0.72	0.08	1.24	0.13	0.87	0.09

666 Noted: Drastic method and f:fuzzy method

667 Table 10: Map removal sensitivity analysis (One parameter is removed at time)

Parameters removed	Variation Index (%)			
	Max	Mean	Min	SD



D	3.69	1.72	0	0.76
R	2.99	1.58	0	0.44
A	3.61	0.67	0	0.42
S	2.99	0.83	0	0.42
T	3.40	0.92	0.06	0.18
I	7.19	3.60	0	0.88
C	4.85	1.53	0.05	0.38

668
 669
 670

Table 11: Map removal sensitivity analysis (One or more parameters are removed at time)

Parameters removed	Variation Index (%)			
	Max	Mean	Min	SD
DASTIC	2.99	1.58	0	0.44
DASTI	5.71	3.73	1.38	0.72
DASI	8.44	6.06	2.92	0.88
DAI	13.18	9.49	4.32	1.54
DI	22.04	15.76	1.94	2.72
I	43.18	21.63	0	5.33

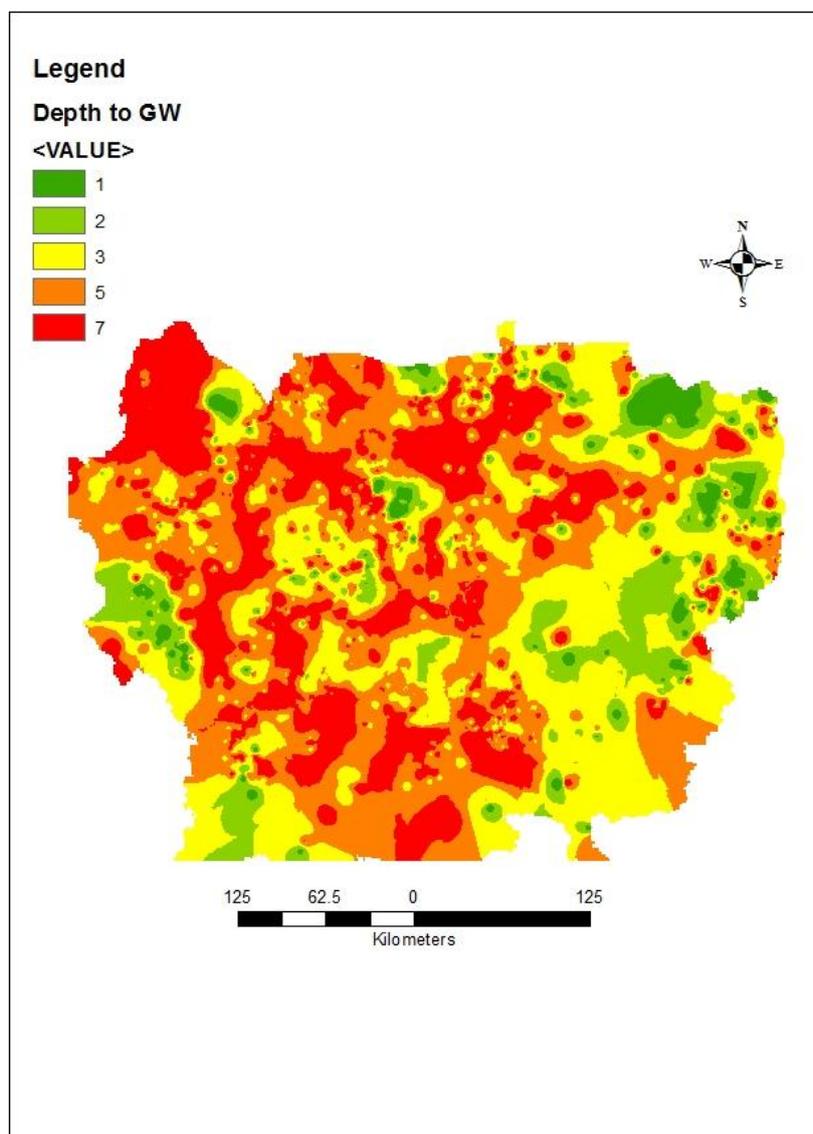
671

672 Table 12: single parameter sensitivity analysis (effective weights)

Parameters	Theoretical weight	Theoretical weight(%)	Effective weight (%)			SD
			Max	Mean	Min	
D	5	21.73(22)	43.20	24.17	4.42	5.59
R	4	17.39(17)	15.58	4.80	2.85	2.65
A	3	13.04(13)	23.37	11.25	6.71	3.65
S	2	8.69(9)	21.97	10.04	4.61	3.70
T	1	4.34(4)	13.88	8.73	2.41	1.09
I	5	21.73(22)	57.47	35.92	14.27	5.37
C	3	13.04(13)	13.95	5.09	2.14	2.27

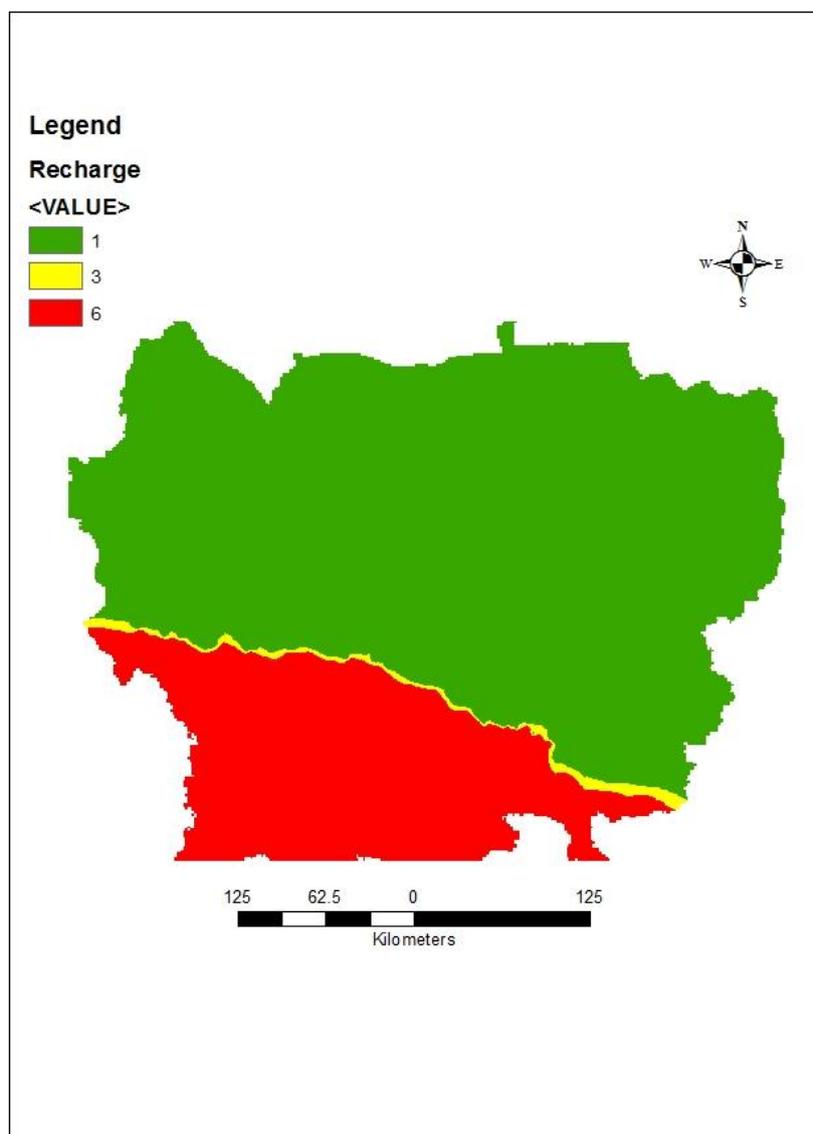
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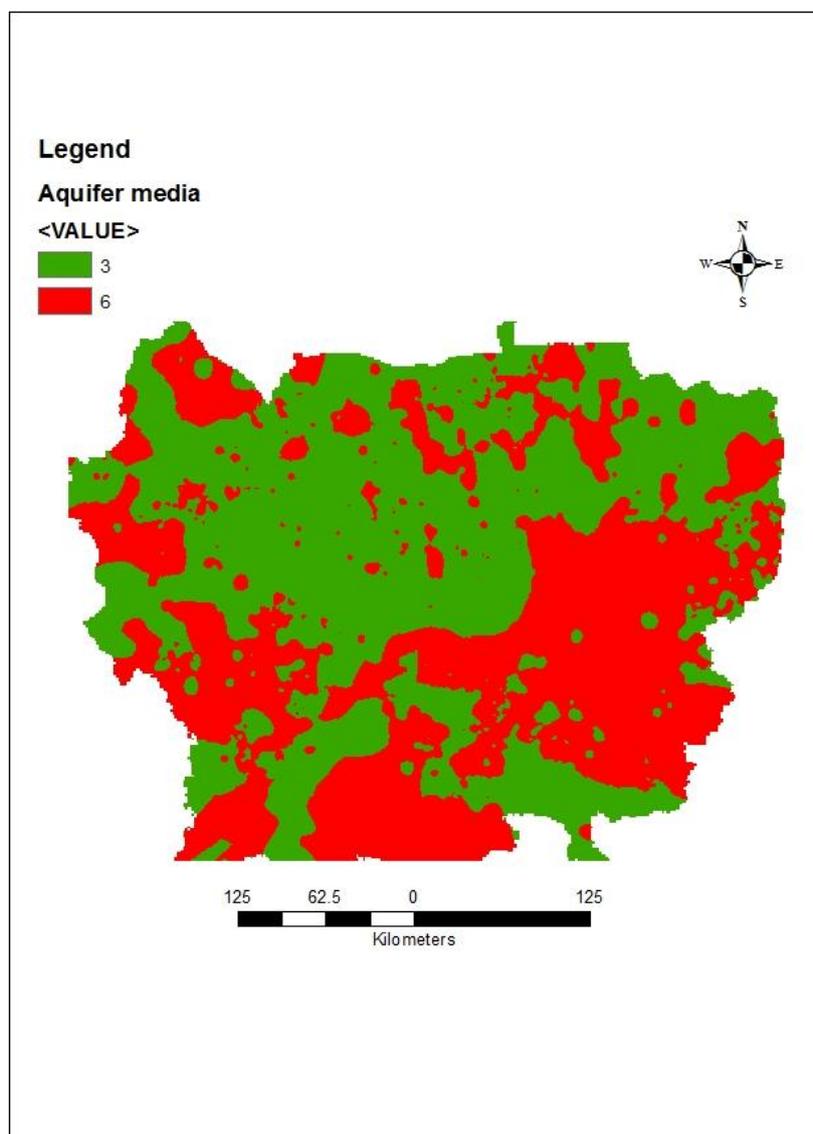
675

676 Figure 1:Groundwater Depth distribution map



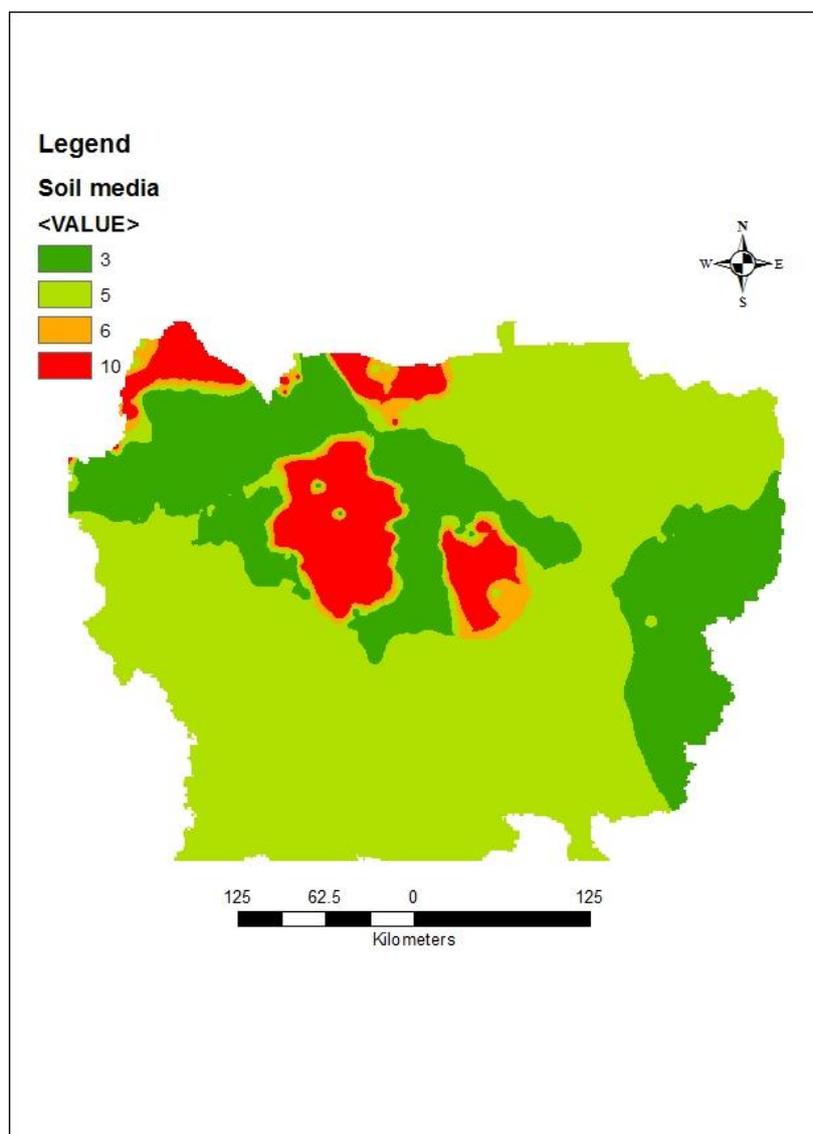
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678 Figure 2:Groundwater Recharge distribution map



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680 Figure 3: Aquifer media distribution map

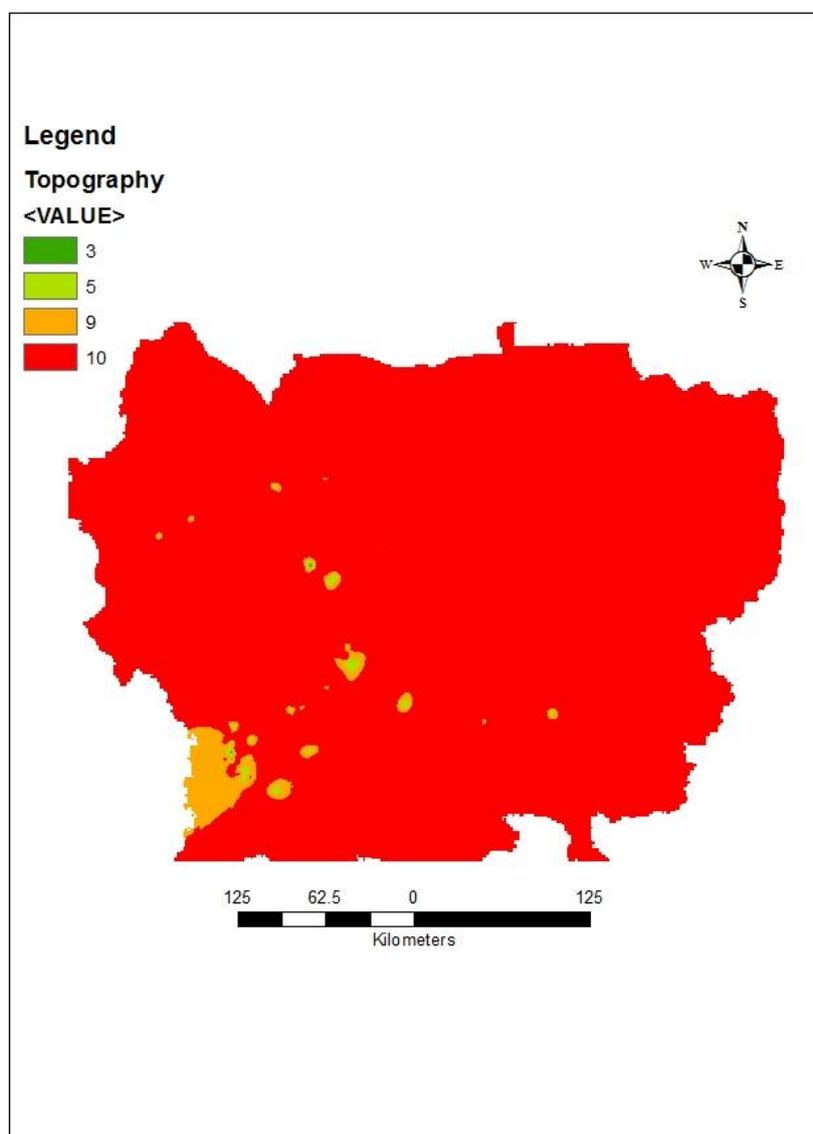


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682 Figure 4: Soil type distribution map

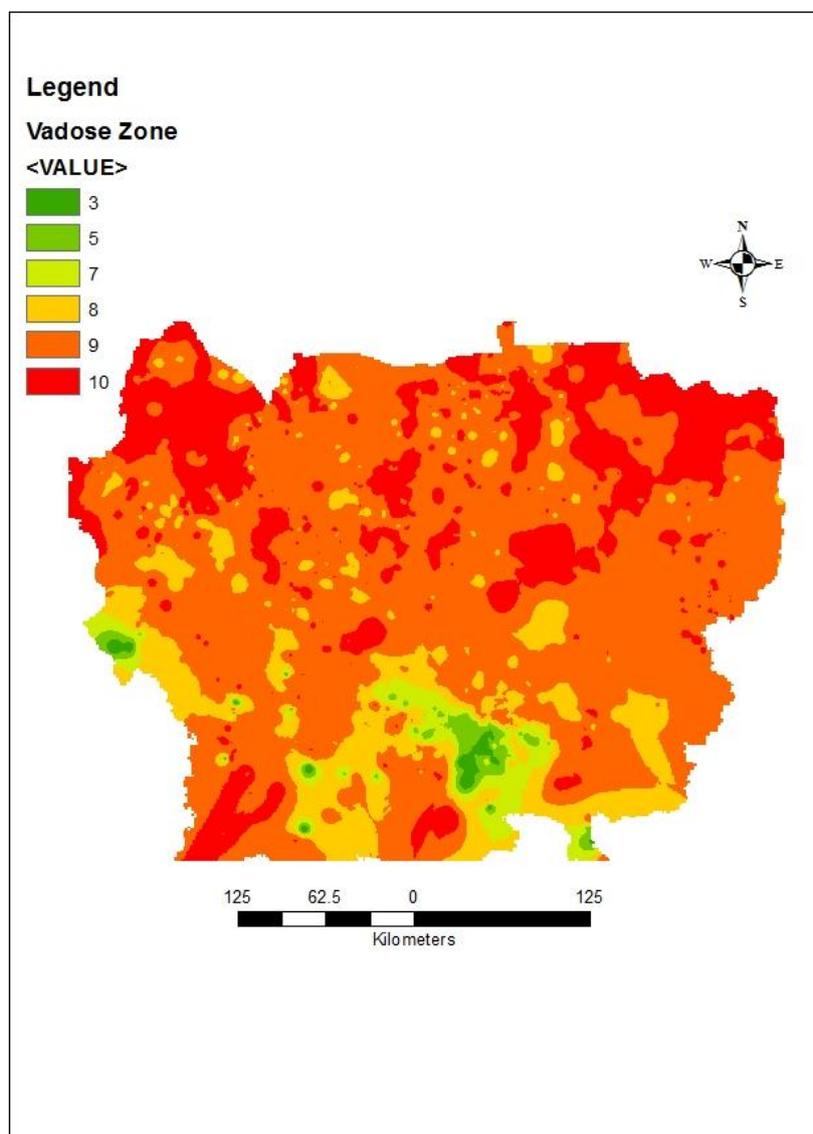


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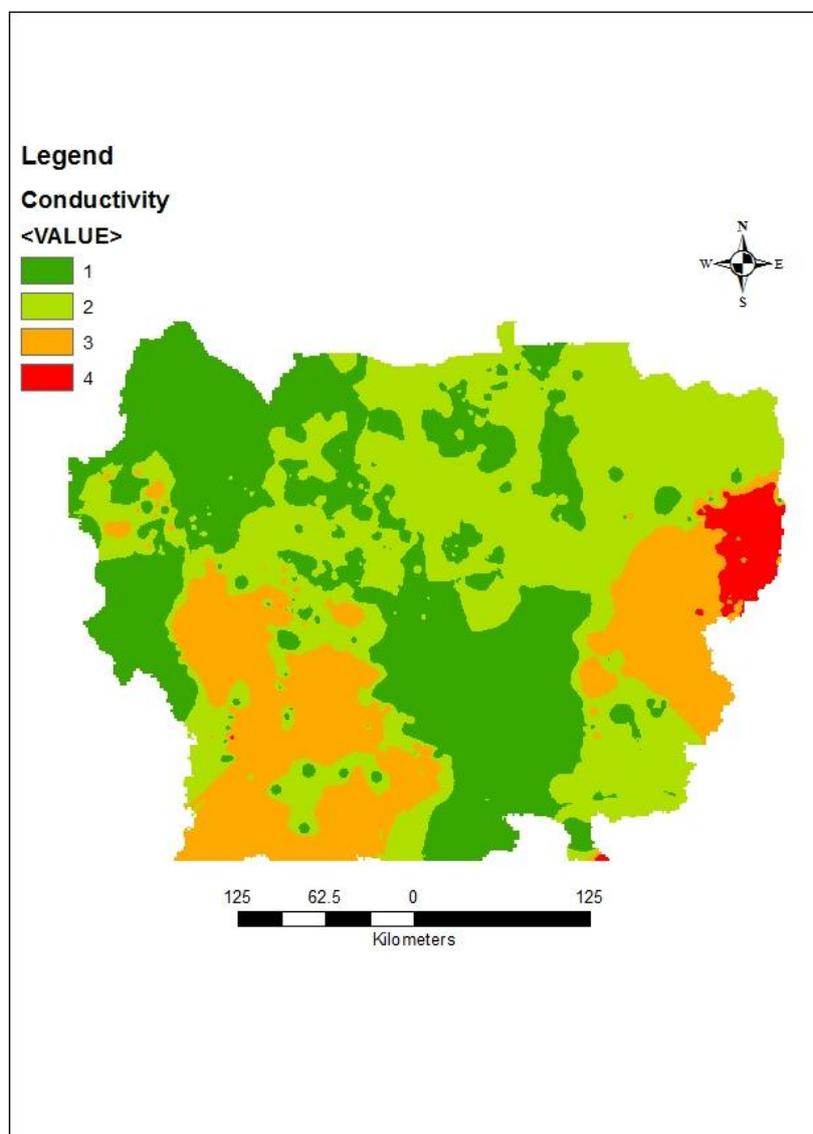


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685 Figure 5:Slope distribution map



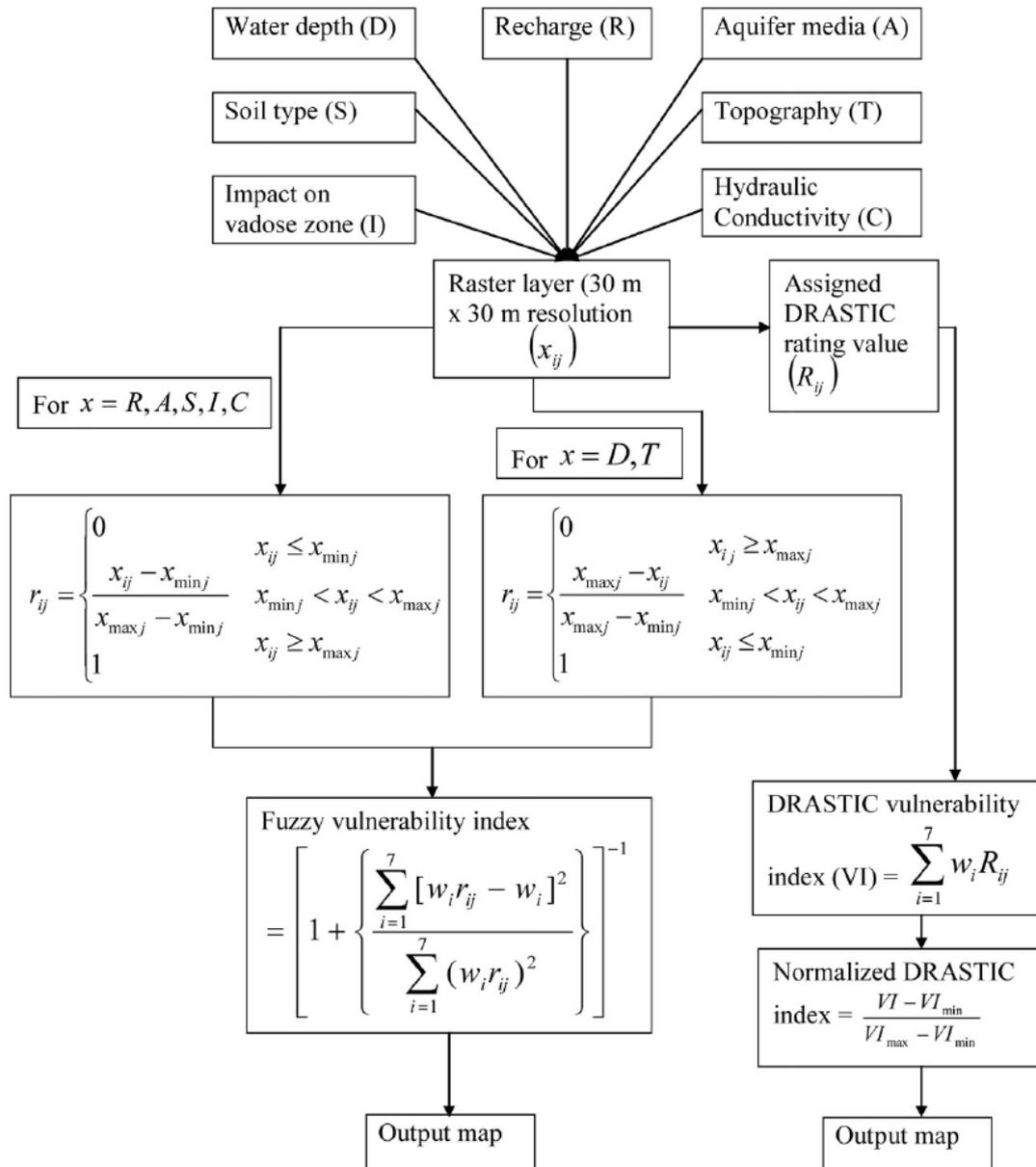
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687 Figure 6: Vadose zone distribution map
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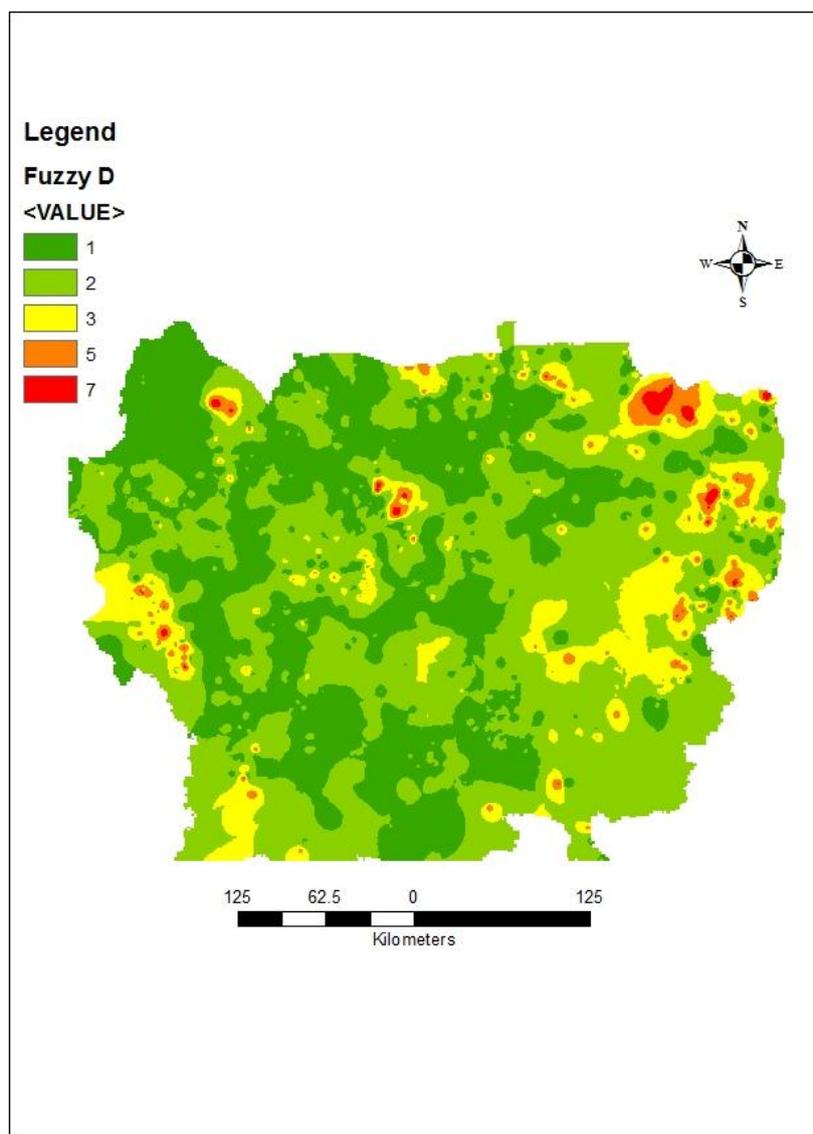
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690 Figure 7: Hydraulic conductivity distribution map
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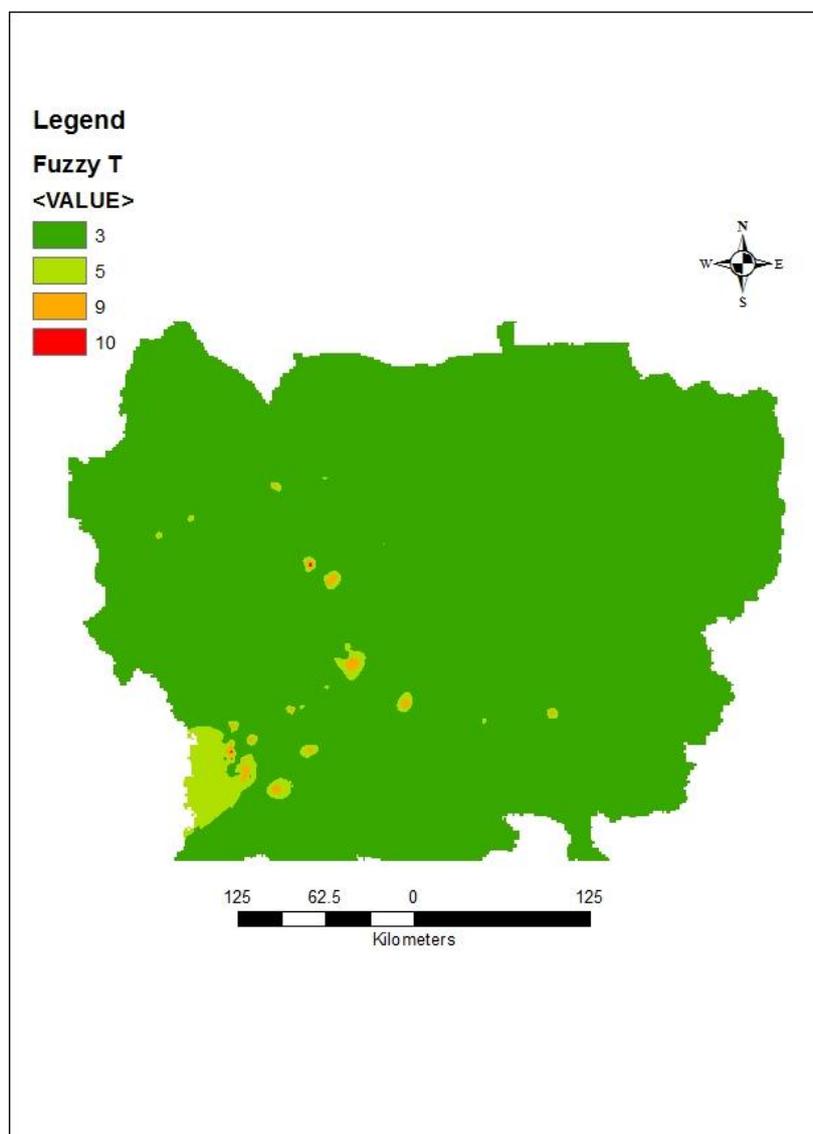
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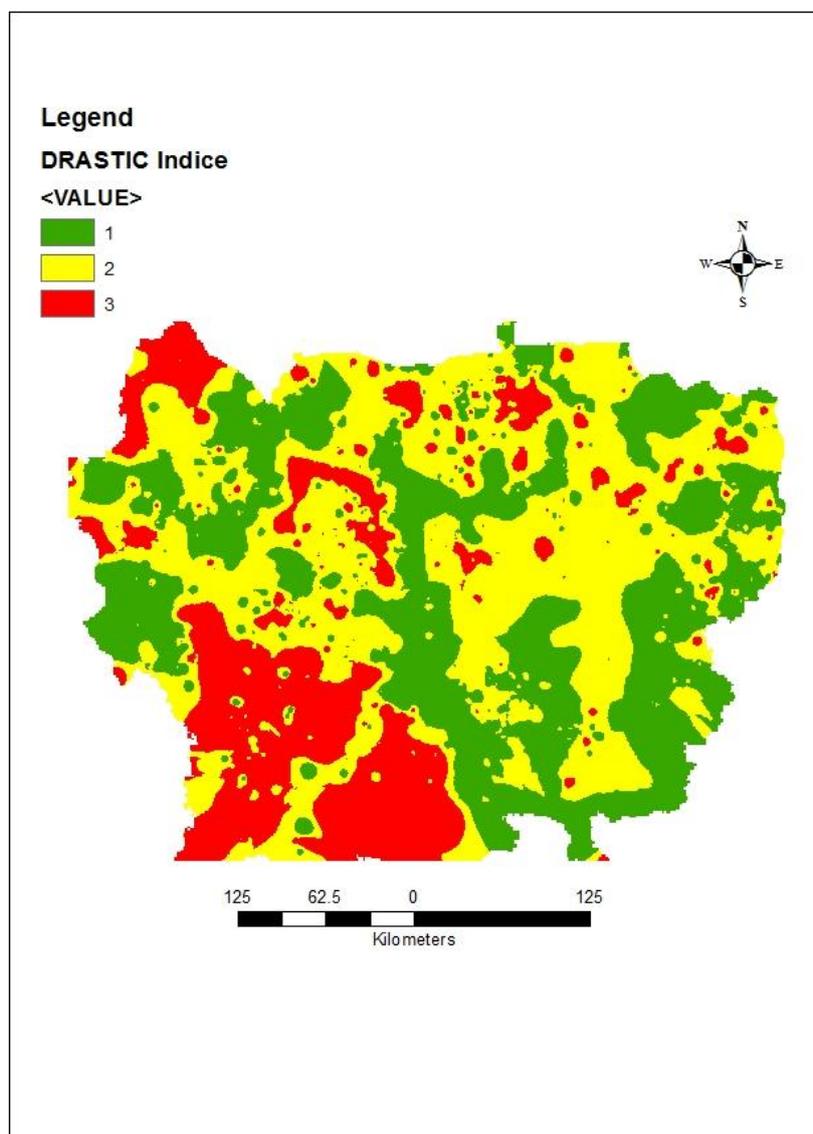
695
 696 Figure8: Flow chart of methodology adopted to develop groundwater contamination potential
 697 map using DRASTIC and fuzzy pattern recognition model in framework of GIS(source Pathak et
 698 al.2009).



699
700 Figure 9:fuzzy concept Ground water depth distribution map

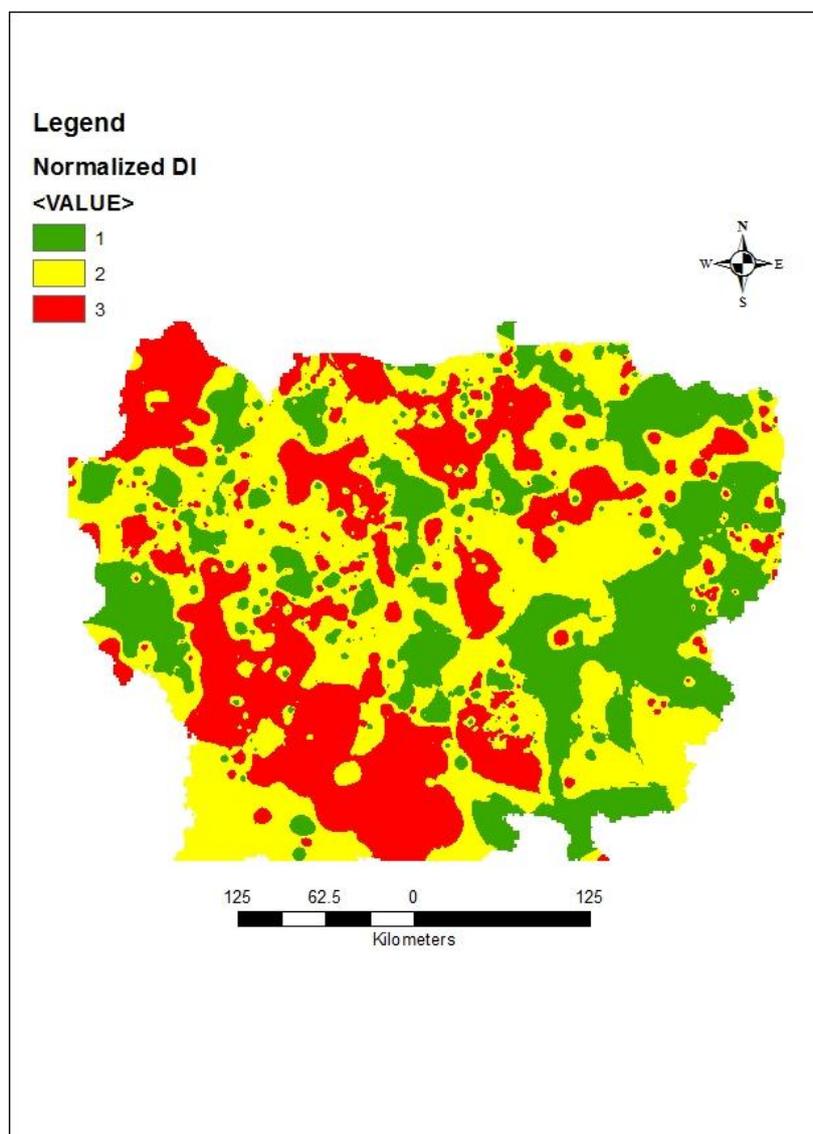


701
702 Figure 10: fuzzy concept topography(or slope) distribution map

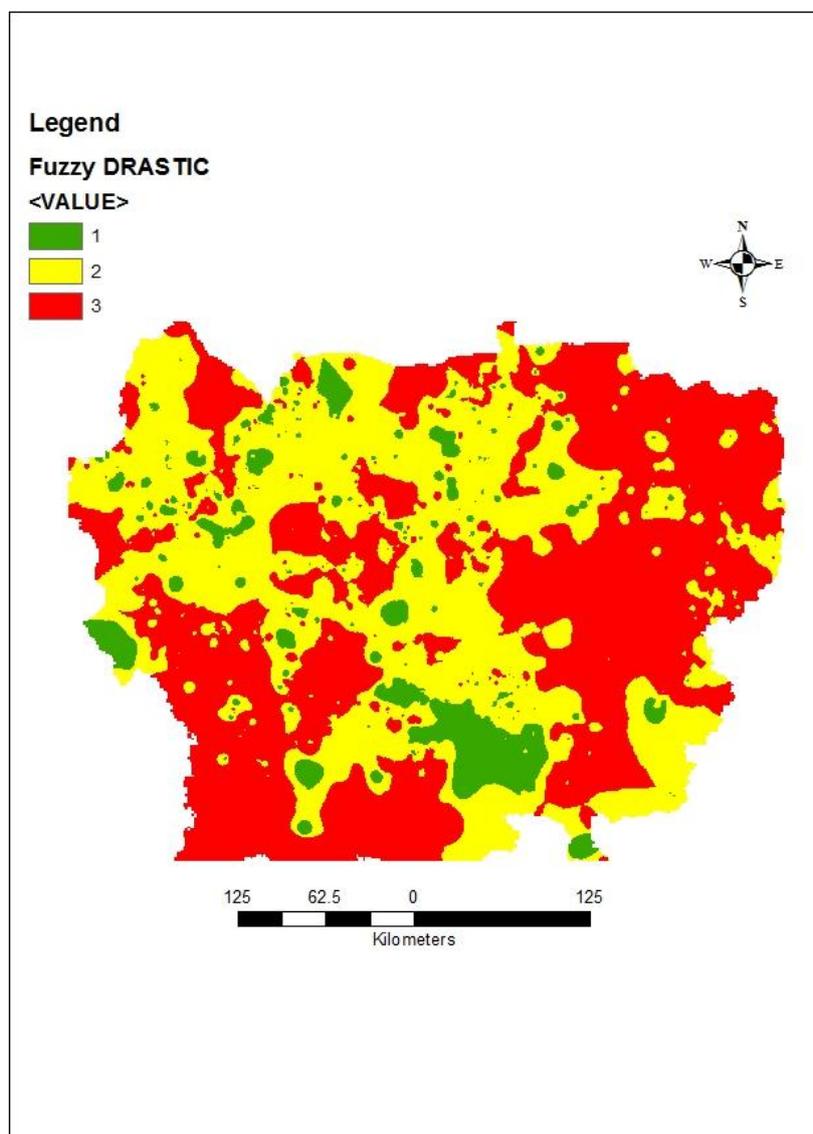


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Figure11:DRASTIC vulnerability map

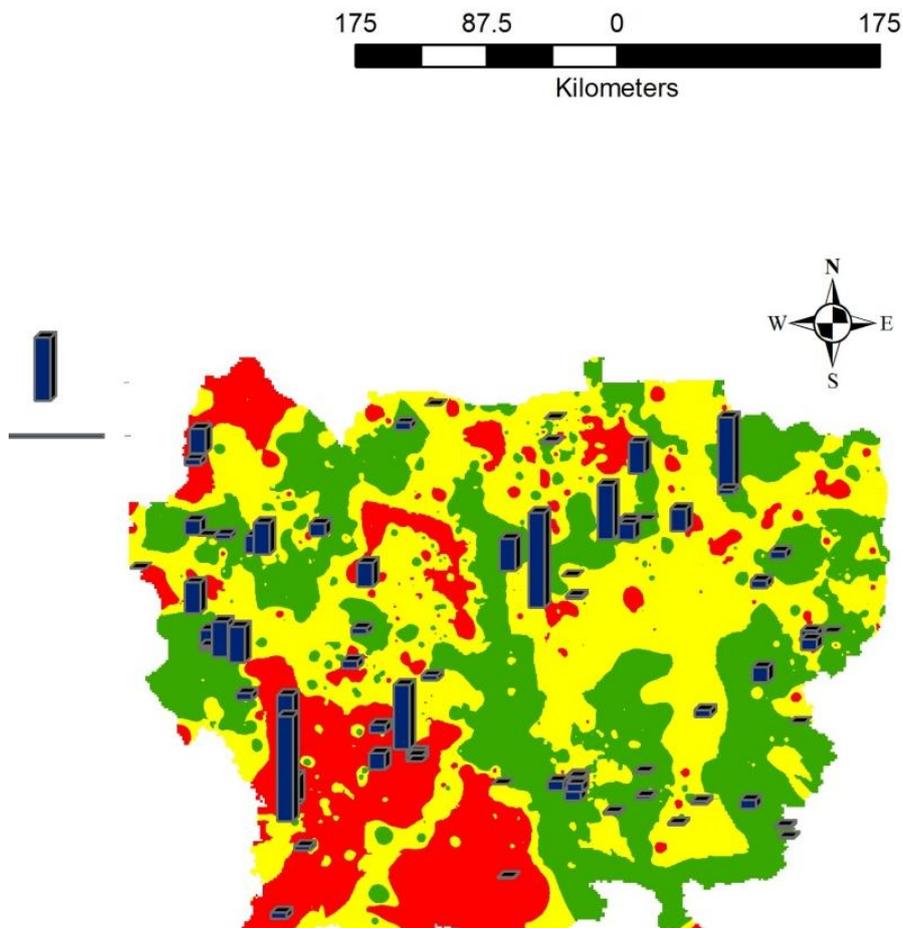


705
706 Figure 12 : Normalized vulnerability map
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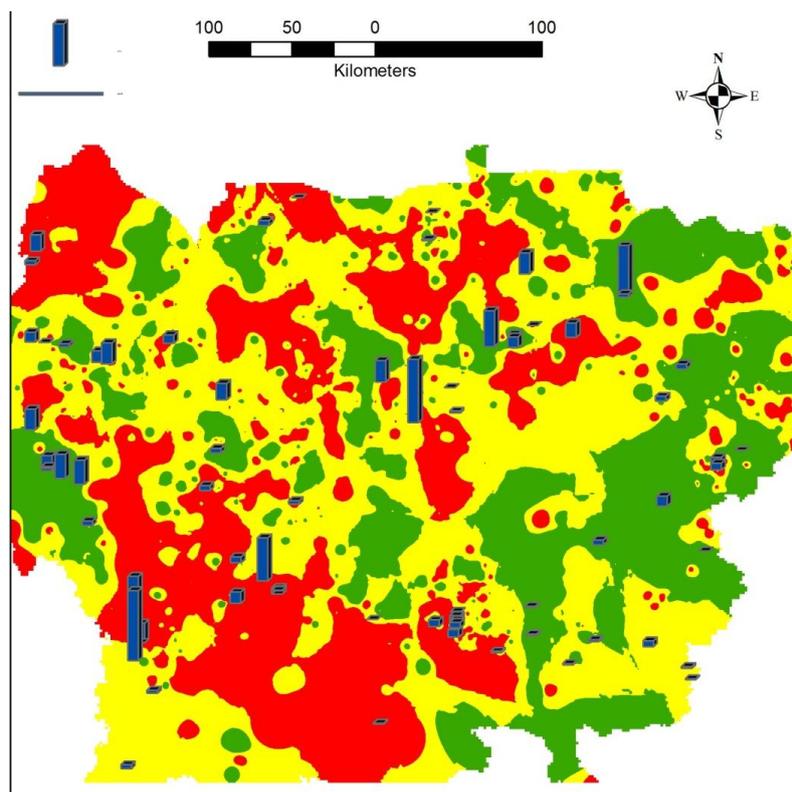
710 Figure 13: fuzzy DRASTIC vulnerability map



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Figure14: Nitrate distribution in DRASTIC model

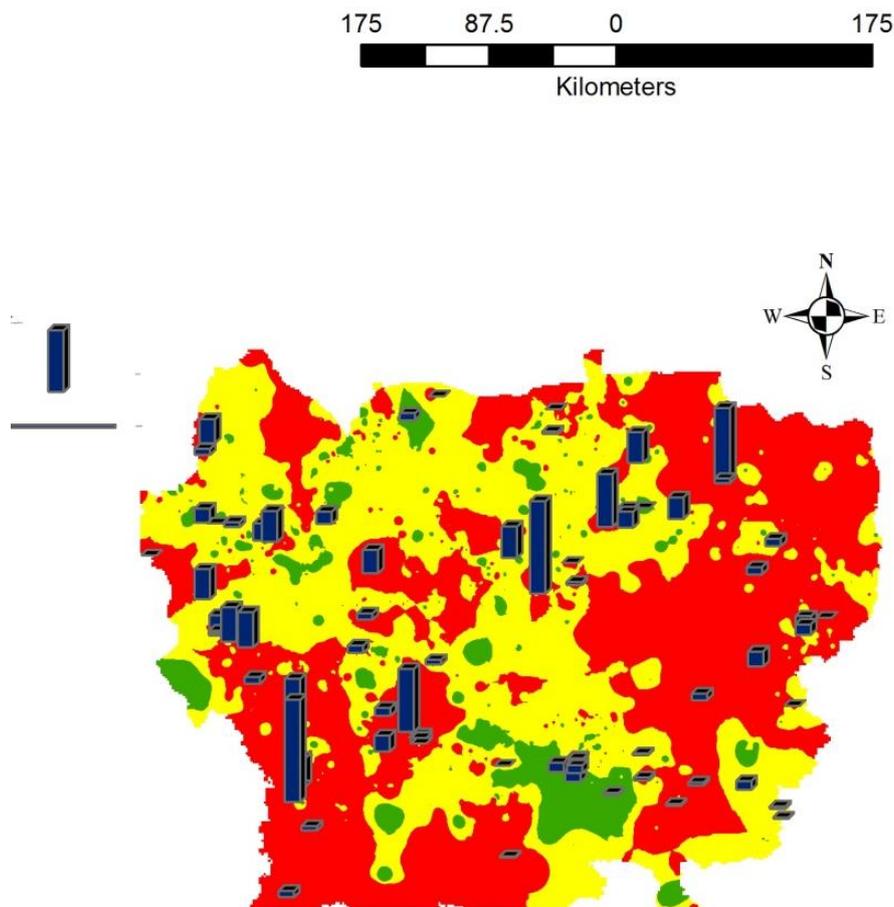


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Figure15:Nitrate distribution in Normalized model



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Figure16: Nitrate distribution in Fuzzy model