

1 **Title:**

2 GIS-based DRASTIC Model for Groundwater Vulnerability and Pollution Risk Assessment
3 in the Peshawar District, Pakistan

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27 **Abstract**

28 Groundwater is the most economic natural source of drinking in urban and rural areas which
29 are degraded due to high population growth and increased industrial development. We applied
30 a GIS-based DRASTIC model in a populated urban area of Pakistan (Peshawar) to assess
31 groundwater vulnerability to pollution. Six input parameters – depth to phreatic/groundwater
32 level, groundwater recharge, aquifer material, soil type, slope and hydraulic conductivity –
33 were used in the model to generate the groundwater vulnerable zones. Each parameter was
34 divided into different ranges or media types, and ratings $R = 1-10$ were assigned to each factor
35 where 1 represented the very low impact on pollution potential and 10 represented very high
36 impact. Weight multipliers $W = 1-5$ were also used to balance and enhance the importance of
37 each factor. The DRASTIC model scores obtained varied from 47 to 147, which were divided
38 into three different zones: low, moderate and high vulnerability to pollution. The final results
39 indicate that about 31.22%, 39.50%, and 29.27% of the total area are under low, moderate, and
40 high vulnerable zones, respectively. Our method presents a very simple and robust way to
41 assess groundwater vulnerability to pollution and helps the decision makers to select
42 appropriate landfill sites for waste disposals, and manage groundwater pollution problems
43 efficiently.

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45 **Key words**

46 Groundwater pollution, Pollution potential; Prioritization of risk areas; GIS for groundwater;
47 Groundwater contamination.

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50 **1. Introduction**

51 Groundwater plays an important role in both human health and many aquatic ecosystems but
52 also facing a high pollution risk from agricultural activities, mining enterprises, industrial
53 development and urbanization (Foster et al. 2002). Different types of activities such as pesticide
54 or fertilizer applications in lawns and agriculture fields, treatment plants, leaking tanks and
55 urban runoff make the groundwater resources more susceptible to pollution. Therefore, it is
56 important to analyze the susceptibility of groundwater resources to pollution for their better
57 planning and management. In addition, such susceptibility analysis could be used to select
58 appropriate locations for different activities such as dumping sites or industrial locations,
59 minimize the adverse consequence on groundwater resources, and achieve the safety and
60 conservation of the aquifer (Gupta and Onta 1997).

61 Groundwater is the water from rainfall or surface water bodies such as lakes and streams that
62 infiltrates into the soil and bedrock and deposits into the subsurface in small pores and spaces
63 between rocks and soil particles. Groundwater vulnerability to pollution is defined as “the
64 direction and probability for pollutants to contact the phreatic level (groundwater table) after
65 infiltration at the ground surface” (Ball et al. 2004). During infiltration, water becomes polluted
66 when it passes through a contaminated soil, carrying the contamination from the surface to the
67 groundwater (Ritter et al. 2002). Similarly, soil or rock particles containing harmful substances
68 also perform as a long-term source of groundwater pollution (Arias-Estévez et al. 2008). The
69 presence of wells accessing the groundwater within an aquifer present higher vulnerability to
70 pollution due to different anthropogenic and natural activities (Rahman 2008). The geological
71 setting of the area also plays an important role in groundwater vulnerability (Baalousha 2011)
72 because it controls the magnitude of time such as the residence time of the groundwater that
73 infiltrates from surface to the subsurface (Prior et al. 2003). The US National Research Council
74 (1993) has defined the groundwater susceptibility to pollution as “the trend or probability for

75 pollutants to contact a specified location in the groundwater system after beginning at a certain
76 place above the highest aquifer”. Many studies are available on the topic of groundwater
77 vulnerability assessment such as Evans and Myers (1990), Piscopo (2001), Al-Adamat et al.
78 (2003), Shirazi et al. (2013), Kumar et al. (2014) and Wu et al. (2014). However, no such
79 attempt has so far been made in the area of Peshawar (Pakistan), and this study case sets
80 precedent for evaluating groundwater vulnerability to pollution using geographic information
81 systems (GIS)-based models in other similar areas as well.

82 According to US National Research Council (1993), groundwater vulnerability assessment
83 approaches include three major categories. (i) Indexed methods, which are based on overlaying
84 and combining maps of different hydro-geologic attributes (for example geology, soil, or depth
85 to water table). These methods include DRASTIC¹ model (Aller et al. 1987), California
86 Hotspot (Cohen et al. 1986), Iowa Ground Water Vulnerability (Hoyer and Hallberg 1991) and
87 Underground Injection Control (UIC) (Pettyjohn et al. 1991). (ii) Process-based simulation
88 methods, which include numerical solution or mathematical models that represent the
89 contaminate transport process, for example, MOUSE (Steenhuis et al. 1987) and RUSTIC
90 (Dean et al. 1989). (iii) Statistical methods that incorporate data from known locations and
91 provide categorization of contamination of the same geographic area. These methods include
92 Discriminant Analysis (Teso et al. 1988) and Regression Analysis (Chen and Druliner 1988).

93 In this study, we focused on DRASTIC model because it is a simple, robust and standardized
94 system developed by Environmental Protection Agency (EPA), USA to assesses contamination
95 potential of large areas using their basic hydrogeologic settings (Aller et al. 1987). Another
96 reason was that a detail physiochemical analysis of water quality parameters would be costly
97 (Baalousha 2010). The model uses a numerical grading system that assigns comparative ratings

¹ Each letter in the name DRASTIC represents a parameter: Depth to groundwater/Phreatic level, net Recharge, Aquifer media, Soil media, Topography, Impact of vadose zone and hydraulic Conductivity (see section 2.2).

98 (*R*) and weights (*W*) to different parameters that are used as input to the model. Thanks to the
99 technological developments of GIS, the DRASTIC model has been used in some GIS-based
100 studies such as the procurement of groundwater vulnerability maps of the Castlereagh
101 catchment in Australia (Piscopo 2001). Similarly, Evans and Myers (1990) used a modified
102 DRASTIC approach and evaluated the pollution potential of groundwater in southern Delaware
103 (US) but they excluded three parameters from the DRASTIC index (recharge, impact of vadose
104 zone and aquifer media) and added new parameters such as septic tank density and land
105 use/land cover (LULC). Wu et al. (2014) also used the DRASTIC model to calculate the
106 intrinsic vulnerability index and they found that intrinsic vulnerability index decreased from
107 east to west with an increase in depth to phreatic level and clay soil thickness. Based on their
108 results, they recommended that the study area should be divided into preferential zones,
109 feasible zones and unfeasible zones for reclaimed water irrigation planning and operation with
110 suggested engineering measures. Shirazi et al. (2013) interpreted groundwater vulnerability of
111 Melaka State (Malaysia) by integrating a land use map with DRASTIC index values to generate
112 a risk map of the study area. They prepared three types of zones such as low, medium and high
113 where the groundwater of important cities was highly vulnerable. Similarly, Kumar et al.
114 (2014) prepared a small-scale map of groundwater vulnerability to pollution using DRASTIC
115 Model in Kancheepuram district, India. They found a strong relationship between the
116 DRASTIC index and nitrate concentration. These studies show that GIS-based DRASTIC
117 method is an important tool for groundwater vulnerability and pollution risk assessment
118 because it uses their basic hydrogeologic parameters which are supposed to have an impact on
119 pollutant transportation from ground surface sources to groundwater (Kalinski et al. 1994) and
120 are easy to obtain (Aller et al. 1987).

121 According to the Pakistan Council of Research in Water Resources (PCRWR), most of the
122 groundwater resources in Pakistan are unsafe for drinking due to biological or chemical

123 contamination (PCRWR 2007), and are continuously degrading due to high population growth
124 and increased industrial development. A huge gap between rural and urban areas, and among
125 provinces, exists with respect to drinking water coverage and accessibility. The Ministry of
126 Planning, Development and Reforms, Government of Pakistan (2013) calculated that 65% of
127 population is considered having access to safe drinking water and 35% are using contaminated
128 water (bacteriological, fluoride, arsenic or nitrate). The sustainability of the existing safe water
129 supply is also a major issue in Pakistan's water sector (Ministry of Environment, Government
130 of Pakistan, 2009). Such inefficient water supply, hygiene and sanitation in Pakistan cause a
131 high rate of diseases as, for example, mortality and morbidity rates and situates a major risk to
132 the existence and growth of children in Pakistan. According to PCRWR (2007) assessment in
133 the Peshawar district, 62% samples were microbiologically contaminated by *Escherichia Coli*,
134 38% samples were contaminated by iron, 23% by calcium and 8% by a high level of total
135 dissolved solids. In this regard, we used a simple approach of GIS-based DRASTIC model to
136 identify those areas where the groundwater resources may be vulnerable to pollution, which
137 would help the decision makers in water-related policies and management.

138 **2. Material and Methods**

139 *2.1. Study area*

140 Peshawar (34° 00' N, 71° 34' E) is the most populated district in the Khyber Pakhtunkhwa
141 province of Pakistan, and extends for an area of 125,700 hectares (**Fig. 1**). The area is divided
142 into four towns named Town 1, Town 2, Town 3, Town 4 and a cantonment area, respectively
143 occupying 3%, 16%, 35% and 45%, and 1% of the total area (Adnan and Iqbal 2014). The
144 mean maximum / minimum temperatures recorded are 30.25 / 1.28 °C in winter and 47.45 /
145 15.60 °C in summer. The average 30 years annual rainfall has been recorded as 400 mm, being
146 higher in winter than summer.

147 ****approximate position of Figure 1*****

148 2.2. *DRASTIC Model*

149 The hydrogeologic settings which established the abbreviation of DRASTIC are;

150 [D] Depth to phreatic level or groundwater table. It is the depth from a ground surface to a
151 phreatic level below ground. The deeper the phreatic level is, the lesser the probabilities of
152 contaminants to affect groundwater.

153 [R] Net Recharge. It is the quantity of water in a unit area that infiltrates the earth surface and
154 arrives the phreatic level. This is the most important cause for contaminants to reach the
155 groundwater table.

156 [A] Aquifer material. These are consolidated or unconsolidated rocks such as gravel, limestone
157 and sand beneath the earth's surface, which retain a quantity of water sufficient for its
158 availability for drinking or irrigation. The smaller and finer the grain size, the higher will be
159 the capacity of the aquifer to reduce contamination because the finer particles restrict the
160 movement of contaminants.

161 [S] Soil material. It is the topmost and weathered part of a ground surface. Soil texture controls
162 the descending movement of pollutants from surface to subsurface.

163 [T] Topography. The gradient of terrain slopes. Flat areas where slope is low are more
164 susceptible to groundwater pollution than steep areas, because they are likely to preserve water
165 for more time, thus allowing high penetration of water recharge and high potential for pollutant
166 migration once contaminated.

167 [I] Impact of vadose zone material. Vadose zone is the portion of soil between the surface and
168 phreatic level, where rock pores and spaces between soil particles are unsaturated. Its influence

169 to the contamination potential is similar to that from soil or aquifer material, since it depends
170 upon its permeability and the granulometric properties of the particles.

171 [C] Hydraulic conductivity. The capacity of aquifer material to transmit water. Thus, an aquifer
172 will be highly vulnerable if the hydraulic conductivity is higher.

173 The DRASTIC model uses the above seven hydrogeologic parameters and evaluates the
174 groundwater vulnerability of the area. **Table 1** shows all the required data layers for modelling
175 which were acquired from various governmental agencies and during the field survey,
176 however, the impact of vadose zone was excluded due to non-availability of data.

177 ****approximate position of Table 1*****

178 Each parameter was divided into different categories, according to ranges or materials of
179 different media types (**Table 2**), and a subjective rating (R_j) from 1 to 10, representing an
180 increasing impact on pollution potential was given to each range or material type (Aller et al.
181 1987; Rahman 2008; Srinivasamoorthy et al. 2011). To balance and enhance the importance of
182 each factor, a weight multiplier (W_j) from 1-5 was used (**Table 2**).

183 ****approximate position of Table 2*****

184 The contaminant potential was measured using a DRASTIC index (D_i), and it is computed as
185 adding the products of weights and rating of each factor, mathematically;

$$186 \quad D_i = \sum_{j=1}^n (W_j * R_j) \quad (1)$$

187 Where, D_i represents DRASTIC index score for a mapping unit i , in the study area. For each
188 parameter j , R_j and W_j represent their rating and weight, respectively. To apply equation (1), a
189 weighted overlay analysis was used in ArcGIS (version 10.2.0.3348, ©1999-2013, ESRI, Inc.,
190 California), which helped to combine these rated parameters and their respective weights.
191 Mapping units having higher DRASTIC index score will have relatively higher pollution

192 potential. The final vulnerability map of the study area was based on experts opinions and
193 literature (Aller et al. 1987; Rahman 2008; Srinivasamoorthy et al. 2011), where the experts
194 opinions were used to select appropriate weights of the parameters and ratings of their ranges
195 or media types according to the hydrogeological setting of the area.

196 *2.2.1. Depth to phreatic level*

197 This parameter is very important in DRASTIC model because the chances of attenuation
198 increase as depth to phreatic level increases since pollutants will require more travel time to
199 reach a deeper phreatic level. To map changes in phreatic level, borehole point data which
200 include depth to phreatic level was collected from Public Health Engineering Department of
201 Peshawar, and their coordinates (latitude and longitude) were collected during the field survey.
202 The data were interpolated using inverse distance weighted (IDW) model, classified into
203 different ranges, and then reclassified into DRASTIC ratings (**Table 2**) (Aller et al. 1987;
204 Rahman 2008; Srinivasamoorthy et al. 2011). This way, a higher rating was given where the
205 phreatic level was less than 5 m because the pollutants will require shortest travel time to reach
206 it, and a lower rating was given where the phreatic level was higher than 25 m as the travel
207 time to reach the phreatic level will be more in this case. Comparing all DRASTIC parameters,
208 the phreatic level was considered by experts as the most influential factor in the vulnerability,
209 and thus it was given the highest weight $W = 5$ (**Table 2**).

210 *2.2.2. Net Recharge*

211 It is the amount of water that infiltrates into the ground surface and reaches the groundwater
212 table level below. Recharge is the main vehicle for transporting and leaching solid or liquid
213 pollutants to the phreatic level as well as responsible for the vertical and horizontal
214 transportation of contaminants. The higher the water recharge is, the higher will be the
215 pollution potential of groundwater. Net recharge was calculated as a product of precipitation

216 and recharge rates while subtracting the amount of water lost during evapotranspiration
217 (Rahman 2008).

$$218 \quad \text{Net Recharge} = \text{Precipitation} - \text{Evapotranspiration} * \text{Recharge rate} \quad (2)$$

219 Since only one rainfall gauge station was available in the study area installed by Pakistan
220 Meteorological Department (PMD) and was not possible to generate a rainfall map. For this
221 reason, Tropical Rainfall Measuring Mission (TRMM) average accumulated rainfall data were
222 downloaded from TRMM Online Visualization and Analysis System (TOVAS). TRMM data
223 was calibrated with PMD data using five years' time series (2004-2008) rainfall data of both
224 PMD station and TRMM points (locations). A linear regression model was developed from
225 both data (**Fig. 2**) and PMD rainfall was predicted at each TRMM point (**Table 3**). The resulting
226 model $PMD = 0.954 \cdot TRMM - 0.5895$ reached a coefficient of determination $R^2 = 0.94$.
227 After predicting the PMD rainfall, surface interpolation for the average annual accumulated
228 precipitation was generated using IDW technique (**Fig. 3**).

229 ***approximate position of Table 3****

230 ***approximate position of Figure 2****

231 ***approximate position of Figure 3****

232 Recharge ratings were calculated from land use/land cover (LULC) map obtained from a
233 Landsat thematic mapper (TM) satellite image of 26 June 2011. The image was classified into
234 five major LULC classes such as wetlands and water bodies, vegetation/agricultural fields,
235 built-up, open lands and mountains, using supervised image classification techniques in Erdas
236 Imagine (Version 13.00.00 Build 281, ©1990-2012 Intergraph Corporation) (**Fig. 1**). The
237 image was reclassified and highest rating value of $R = 5$ was given to highest recharge areas
238 such as wetlands and water bodies due to the lower water table and direct exposure to the

239 pollution; rating $R = 4$ was given to agricultural fields; rating $R = 3$ was given to open lands, and
240 lowest rating $R = 1$ was given to lowest recharge areas, mountains and urban areas, due to
241 either their high slopes or impermeable surfaces.

242 Potential Evapotranspiration (PET) of Willmott and Matsuura (2001) (**Fig. 4**) was used in the
243 study, and final net recharge was calculated using equation 2, which was then reclassified to
244 DRASTIC ratings (**Table 2**). Net recharge was considered as the second most important factor
245 in the analysis and, therefore, a weight $W = 4$ was given to it (**Table 2**).

246 ***approximate position of Figure 4****

247 2.2.3. Aquifer Material

248 The aquifer materials (like gravel, limestone and sand) directly affect the amount of
249 groundwater available in the aquifer and its pollutant attenuation capacity. For example, if the
250 grains sizes of the aquifer materials are larger and having more fractures or openings, then
251 permeability will be high and attenuation capacity will be low.

252 An aquifer map obtained from water and power development authority (WAPDA), Pakistan
253 was digitized and converted to raster using ArcGIS. The raster map was then reclassified into
254 DRASTIC ratings. Aquifer materials were considered as the third most important parameter
255 and was given a weight $W = 3$ (**Table 2**).

256 2.2.4. Soil Media

257 Soil significantly affects the amount of recharge by allowing water to infiltrate and reach
258 phreatic level, thus it affects the pollution movement. The soil permeability is decreased by the
259 occurrence of small textured materials such as silts and clay that restrict the pollutant migration,
260 however, the larger textured materials improve soil permeability as well as pollutant migration.
261 The study area soil map was obtained as a scanned hardcopy from the Institute of Geographical

262 Information System (IGIS) library, National University of Sciences and Technology, Pakistan.
263 The map was digitized, converted to raster, and a lowest DRASTIC rating was given to fine-
264 textured soils such as clay and a higher rating was given to larger textured soil such as gravel
265 and stones (**Table 2**). Soil types and aquifer materials have a similar characteristic, therefore
266 weight $W = 3$ was also given to soil type (**Table 2**).

267 *2.2.5. Topography*

268 Flat areas retain water for longer time, permit high penetration or recharge of water and provide
269 high potential for pollution movement. While areas with steep slopes provide a high amount of
270 runoff, less infiltration and are less vulnerable to groundwater pollution. Slope ranges were
271 calculated from an ASTER digital elevation model (DEM). Lowest rating was given to steep
272 slopes, and highest rating was given to flat areas (**Table 2**) (Aller et al. 1987; Rahman 2008;
273 Srinivasamoorthy et al. 2011). The influence of topography was the lowest than all other
274 parameters and therefore, it was given a lowest $W = 1$ (**Table 2**).

275 *2.2.6. Impact of Vadose Zone Material*

276 The materials in the vadose zone, the area between the surface and the water table, has similar
277 effects on vulnerability to pollution as those in the surface soil or aquifer. Vulnerability to
278 contamination depends on its particle sizes, permeability (Rahman 2008) and thickness (Wu et
279 al. 2014). Due to non-availability of useful data, the impact of vadose zone material was
280 excluded from this study.

281 *2.2.7. Hydraulic Conductivity*

282 Measuring the hydraulic conductivity (HC) provides a quantification of the capacity of aquifer
283 material to transmit water. The amount and interconnection of void spaces in the aquifer are
284 responsible for controlling hydraulic conductivity. Freeze and Cherry (1979) calculated the HC

285 values for different aquifer materials and we obtained those values according to the aquifer
286 materials (gravel-sand with some silt and clay, and sand with minor silt and clay) in our study
287 area. DRASTIC ratings were assigned according to Aller et al. (1987) (**Table 2**) and was given
288 a weight $W = 3$, similar that for the aquifer materials (**Table 2**).

289 **3. Results**

290 *3.1. DRASTIC parameter maps*

291 **Figure 5** shows the maps obtained for each of the parameters considered in the DRASTIC
292 model, wherein **Fig. 5a** illustrates variations in phreatic level which ranged from 1.5 m to 78.0
293 m. The phreatic level was categorized into four ranges such as <5 m, 5-10 m, 10-25 m and >
294 25 m, and were given ratings $R = 10, 9, 7$ and 5 , respectively (**Table 2**). It was found that the
295 phreatic level was much higher (< 5 m) on the northern sides as compared to the southern sides
296 (> 25 m) of the study area.

297 The net recharge map obtained using equation 2 was divided into low, medium and high
298 recharge zones, and were given rating $R = 1, 5$ and 8 , respectively (**Table 2**). The spatial
299 patterns of net recharge shown in **Fig. 5b** are in clear relationship with land use patterns in the
300 district of Peshawar (**Fig. 1**), since medium recharge zones are those mainly covered by
301 agricultural fields, low recharge zones correspond to urban areas in the center and mountains
302 in the south, and the presence of higher recharge values is mainly where wetlands and water
303 bodies are located.

304 The study area is mostly homogenous in terms of aquifer materials and is composed of only
305 two types of aquifer materials. First, gravel along with sand, silt and clay, which occupies most
306 of the area and was given rating $R = 8$. The second type of aquifer material was sand along

307 with some silt and clay, which occupies the northern parts of the study area and was given a
308 rating $R = 5$ (**Fig. 5c**).

309 Different soil types exist in different parts of the study area such as silt loam with some sands
310 in northern parts, silt loam in southern parts, gravel and stones in the eastern parts and silt-clay
311 in the central parts (**Fig. 5d**). The lowest rating $R = 1$ was given to the fine-textured soils such
312 as clay and silt-clay, and the highest rating $R = 10$ was given to the gravels and stones because
313 fine-textured particles help to prevent and the larger particles accelerate the movement of
314 pollutants.

315 The terrain of the area was mostly flat, which was given the highest ratings. However, some
316 steep slopes, greater than 12%, also existed in the southern areas of Peshawar district (**Fig. 5e**).
317 The slope was divided into different ranges such as 0-2, 2-6, 6-12 and > 12%, which were given
318 ratings of $R = 10, 9, 5,$ and $2,$ respectively.

319 Based on the above two types of aquifer material in the study area, the hydraulic conductivity
320 values obtained were 2.3 meters/day for gravel-sand with silt and clay (first type) and 10.4
321 meters/day for sand with minor silt and clay (second type). Therefore, these soil types also
322 depicted in **Fig. 5c** were given ratings $R = 2$ and $5,$ respectively, with regards of their HC (**Fig.**
323 **5f**).

324 ****approximate position of Figure 5*****

325 *3.2. Final Vulnerability Map of the Study Area*

326 The final vulnerability map of DRASTIC model was obtained by combining the parameters
327 ratings and their respective weights as discussed in section 2.2 and 3.1. The DRASTIC scores
328 obtained from the model varied from $D = 47$ to $D = 147$. These values were reclassified into
329 three classes i.e. low, moderate and high vulnerable zones using a quantile classification

330 scheme (**Fig. 6**). This classification resulted in a low vulnerable class corresponding to
331 DRASTIC Index scores $D < 92$, moderate vulnerable being the values from $D = 92$ to $D =$
332 107, and those which were $D > 107$ corresponding to highly vulnerable areas. **Table 4**
333 summarizes the final results obtained after calculating the final areas of vulnerable zones. We
334 found that 31.22% (399.7 km²) of the total area was under low vulnerable zone, where Achni
335 Bala, Pawaka, Sheikhan, Sarband, and Sufaid Dherai are included. Moderate vulnerability
336 corresponded to 39.50% (505.7 km²) of the total area, including Jogani, Mehl Tari, Sheikh
337 Muhammadi, Urmir Miana, Badabher, Hazar Khwani. Finally, 29.27% (374.6 km²) of the total
338 area was classified as being highly vulnerable to pollution, which included Mathra, Panam
339 Dherai, Kaaniza, Haryana, Hasan Garhi, Shaheen Town, Wadpaga, Suraizi.

340 ****approximate position of Table 4****

341 ****approximate position of Figure 6****

342 **4. Discussion**

343 Groundwater is one of the major sources of water which is used for all purposes such as
344 agricultural, commercial, domestic and industrial (Visnuvarthanan et al. 2016), but a high
345 dependency on groundwater resources causes an increasing pressure on its quality and quantity
346 (Piscopo 2001), thus facing a pollution risk from different activities such as agriculture, mining
347 enterprises, industrial development and urbanization, in Pakistan (PCRWR 2007) as well as
348 worldwide (Foster et al. 2002). Therefore, a quantitative and qualitative assessment of
349 groundwater vulnerability to pollution is important for decision-making. Due to high
350 groundwater contamination in the study area, an assessment of groundwater pollution was
351 needed (PCRWR 2007). But a detailed physiochemical analysis of groundwater can be costly
352 and time-consuming (Baalousha 2010). Therefore, we attempted to assess the groundwater
353 vulnerability to pollution using GIS-based DRASTIC model as it has been applied in other

354 various study areas (Aller et al. 1989, Evans and Myers 1990, Shirazi et al. 2013, Wu et al.
355 2014).

356 The phreatic level has a significant impact on groundwater vulnerability because vulnerability
357 to pollution decreases as the phreatic level increases (Wu et al. 2014). In this study case, a huge
358 difference was found between the northern area of the Peshawar district, where the phreatic
359 level was < 5 m, and the south-eastern areas where the phreatic level was > 25 m (**Fig. 5a**).

360 This shows that the chances of vulnerability would much likely to decrease from north to south-
361 east areas because the pollutants will have longer distances to travel before contacting
362 groundwater (Rahman 2008). Whereas, wetland areas and water bodies in the north of the study
363 area (**Fig. 5b**) are highly vulnerable to pollution due to high phreatic level in the north (**Fig.**
364 **5a**). Fertilizers and pesticides used in the nearby agricultural fields can have direct or indirect
365 effects to groundwater as these pollutants infiltrate along with groundwater recharge (Rahman
366 2008). Groundwater recharge is the second most important factor in vulnerability analysis
367 (Rahman 2008; Srinivasamoorthy et al. 2011; Rahmati et al. 2015), and hence a high weight
368 was given in this study (**Table 2**). But a classification of a medium-resolution image (30 m) of
369 Landsat TM satellite and TRMM precipitation data was used for recharge calculations which
370 could have affected the final vulnerability. However, recharge map could be improved further
371 by using a high-resolution image such as Quickbird or SPOT, and field rainfall measurements
372 or direct recharge measurement in the field. In soil type and aquifer material, the particle size
373 greatly affects the vulnerability. Larger sizes of particles leads to higher vulnerability, because
374 the water and pollutants would easily move in larger particle size materials (Rahman 2008;
375 Srinivasamoorthy et al. 2011). Therefore, a higher rating was given to coarse size particles such
376 as gravel and sand, and a lower rating was given to fine size particles such as clay and silt
377 (Jamrah et al. 2008) (**Figs. 5c and 5d**). Based on literature and experts opinions, soil and aquifer
378 material was considered as the third most important factor in vulnerability, and thus it was

379 ranked with a corresponding weight (**Table 2**). The study area is mostly flat (**Fig. 5e**) having
380 higher recharge potential which speeds up vulnerability to pollution, therefore, maximum
381 rating of 10 was given to flat slopes and rating 1 was given to steep slopes (Rahman 2008;
382 Srinivasamoorthy et al. 2011). The lowest weight was given to the effect of topography (**Table**
383 **2**), because its influence is mostly related to rainfall occurrences. For example, if rainfall
384 occurs, water will retain for a longer time in flat areas, as compared to the steep slopes,
385 permitting higher penetration or recharge of water. The hydraulic conductivity which depends
386 on the particle size of the aquifer materials (Odong 2007) was the most challenging and difficult
387 to calculate (Aller et al. 1987). However, since Freeze and Cherry (1979) calculated the HC for
388 different aquifer materials, their values were used to account for the effect of HC according to
389 the aquifer materials in the study area (**Fig. 5f**). Due to the dependency of hydraulic
390 conductivity on aquifer materials, a similar weight was assigned to it (**Table 2**).

391 Spatial variability in groundwater vulnerability to pollution shows that, generally, the southern
392 and western areas are low vulnerable, whereas north-eastern and south-eastern areas are
393 moderately vulnerable, and the higher vulnerable areas are distributed in the northern and
394 eastern zones parts of the of the Peshawar district (**Fig. 6**). This categorization was similar to
395 those carried out by Shirazi et al. (2013) and Wu et al. (2014), in which they divided their study
396 area into low, medium and high vulnerable zones, or, preferential zones, feasible zones and
397 unfeasible zones, respectively. A detailed spatial analysis of groundwater quality in the study
398 area shows that different water quality parameters such pH, electrical conductivity, calcium
399 and magnesium hardness, alkalinity (Adnan and Iqbal 2014), and bacteriological
400 contamination (Khan et al. 2013) are mostly concentrated in the northern and eastern areas. In
401 the present study, DRASTIC model accurately predicted high vulnerability in those areas. This
402 is a good indication that the DRASTIC model can be useful for determining groundwater
403 vulnerability to pollution, as an alternative to the high costs of a more detailed physiochemical

404 analysis (Baalousha 2010). However, the central, western and south-western parts of the study
405 area were predicted as medium to low vulnerable, where Adnan and Iqbal (2014) obtained
406 nitrate concentrations with high to medium values. The final results of our DRASTIC model
407 implementation (minimum and maximum DRASTIC index score) were similar to those
408 obtained by Rahman (2008), Srinivasamoorthy et al. (2011), Shirazi et al. (2013), Wu et al.
409 (2014) or Kumar et al. (2014), which demonstrates consistency and robustness in the method.
410 Thus, we recommend that DRASTIC model may easily be employed at very other different
411 areas. We also observed that our results were very dependent on ratings and weights of the
412 parameters. Therefore, we recommend further research on the sensitivity of the DRASTIC
413 model to different choices for rating and weighting the input parameters, in order to analyze
414 their effects and importance on the final results.

415 **5. Conclusions**

416 The DRASTIC model was useful for determining groundwater vulnerability to pollution, and
417 it can be a good alternative to physiochemical analyses, sparing their high costs. The overall
418 results of our study case were satisfactory, and we are confident that it would help the decision
419 makers in waste disposal sites selection and development of reliable policies for efficient
420 management and exploitation of groundwater resources. Areas found less vulnerable to
421 groundwater pollution thus suitable for industries or waste disposal sites were Achni bala,
422 Sheikhan, Aza khel, Sheikh Muhammadi, Pawaka and Sarband. Highly vulnerable priority
423 areas needing protection and conservation were Panam Dheri, Charghar Matti, Mathra, Kaniza,
424 Dagg, Hasan Garhi, Khazana, Urmar, and Budhni. This study suggests that GIS-based
425 DRASTIC model could be used for groundwater vulnerability assessment and prioritization of
426 vulnerable areas, and it could be easily employed elsewhere.

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541 **Table Titles**

542 **Table 1.** Data types and sources of data used in the DRASTIC Model

Sr. No	Data Type	Source	Output Layer
1	Tube wells locations	Field survey	X,Y coordinates
2	Tube wells Strata charts	Public health engineering department Peshawar	Depth to groundwater
3	Rainfall and satellite image	TRMM rainfall data, PMD rainfall, Landsat thematic mapper (TM) image	Recharge
4	Aquifer map	Water and power development authority (WAPDA), Pakistan	Aquifer material
5	Soils map	University (institute of GIS) library	Soil type
6	DEM	SRTM DEM from internet	Topography
7	Hydraulic conductivity	Literature (DRASTIC manual)	Hydraulic conductivity

543 Note: TRMM, tropical rainfall measuring mission; PMD, Pakistan meteorological department;
 544 SRTM, shuttle radar topography mission; DEM, digital elevation model.

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Table 2. Rating and weights of DRASTIC parameters

1: Depth to groundwater		2: Net recharge		3: Aquifer media		4: Topography (slope)		5: Soil media		6: Hydraulic conductivity	
Ranges	Rating	Category	Rating	Material	Rating	Ranges	Rating	Material	Rating	Value	Rating
(m)										(m/day)	
< 5	10	Urban, built-up and mountain/ low recharge	1	Gravel, sand, clay	8	0—2	10	Clay and silt clay	1	2.3	2
5—15	9	Vegetation/ medium recharge	5	Sand with minor clay	5	2—6	9	Silty clay and clays with sand	3	10.4	5
15—25	7	water bodies and wetlands/ high recharge	8			6—12	5	Silt loam	4		
>25	5					>12	2	Silt loam with sand	5		
								Sandy loam with gravel	6		
								Sand	8		
								Gravel and Stony	10		
Weight = 5		Weight = 4		Weight = 3		Weight = 1		Weight = 2		Weight = 3	

560

561 **Table 3.** TRMM annual accumulated precipitation at various locations and their corresponding
 562 PMD predicted precipitation

S. No	Latitude	Longitude	TRMM Precipitation (mm)	Predicted PMD (mm)
1	33.625	71.375	509.4	485.4
2	33.625	71.625	461.5	439.7
3	33.625	71.875	548.2	522.4
4	33.875	71.375	547.2	521.4
5	33.875	71.625	550.02	524.1
6	33.875	71.875	614.5	585.7
7	34.125	71.375	527.6	502.8
8	34.125	71.625	552.5	526.5
9	34.125	71.875	638.8	608.8
10	34.375	71.375	501	477.3
11	34.375	71.625	559.5	533.2

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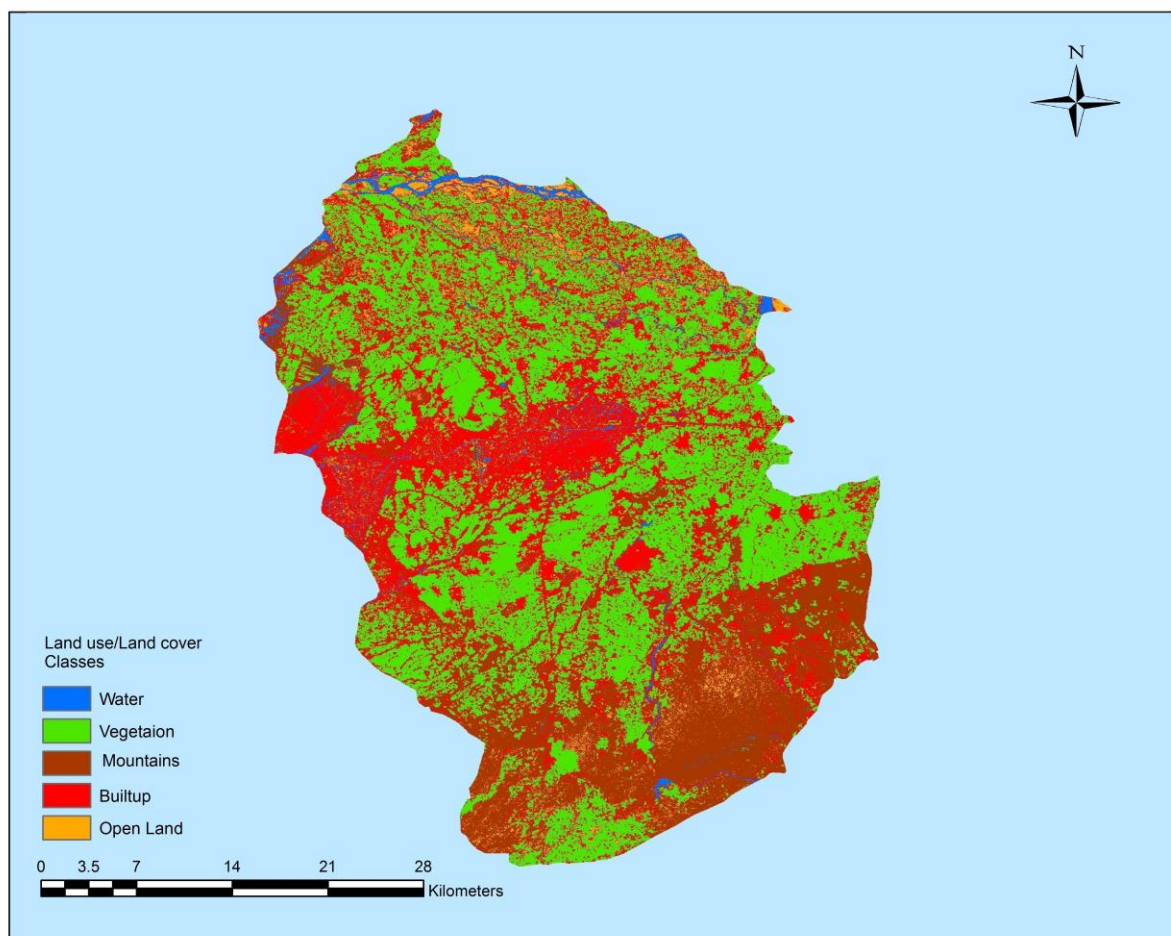
564 **Table 4.** Groundwater vulnerable zones

S. No	Status	DRASTIC Index	Area (km ²)	Union Councils
1	Low Vulnerable	<92	399.7	Achni bala, Sheikhan, Aza khel, Sheikh Muhammadi, Sarband,
2	Moderate Vulnerable	92-107	505.7	Badbher, Hazar Khwani, Shaheen town, Sikandar Town, Faqeer abad, Mehl Tari, Tehkal, Hayatabad
3	High Vulnerable	> 107	374.6	Panam Dheri, Charghar matti, Mathra, Kaniza, Dagg, Hasan Garhi, Khazana, Urmar, Budhni

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567 **Figure Captions**



568

569 **Figure 1.** Map of the study area (Peshawar, Pakistan).

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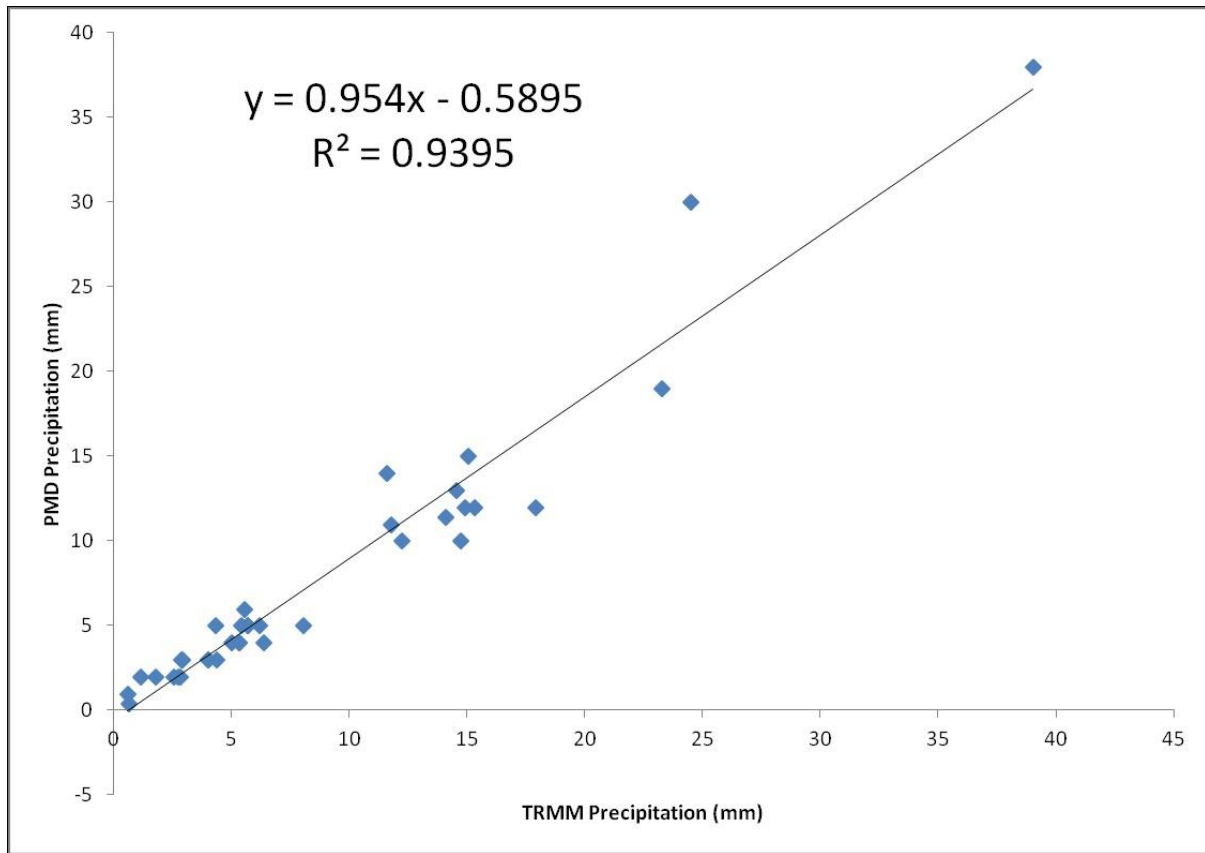
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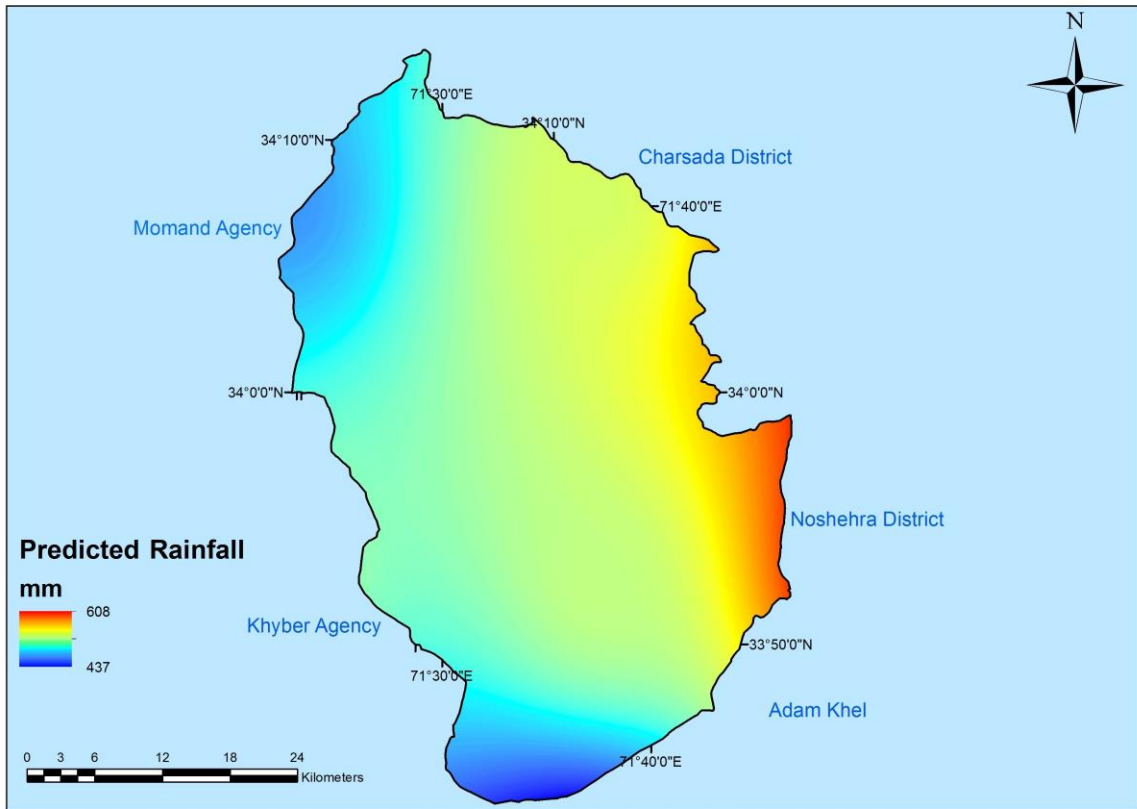
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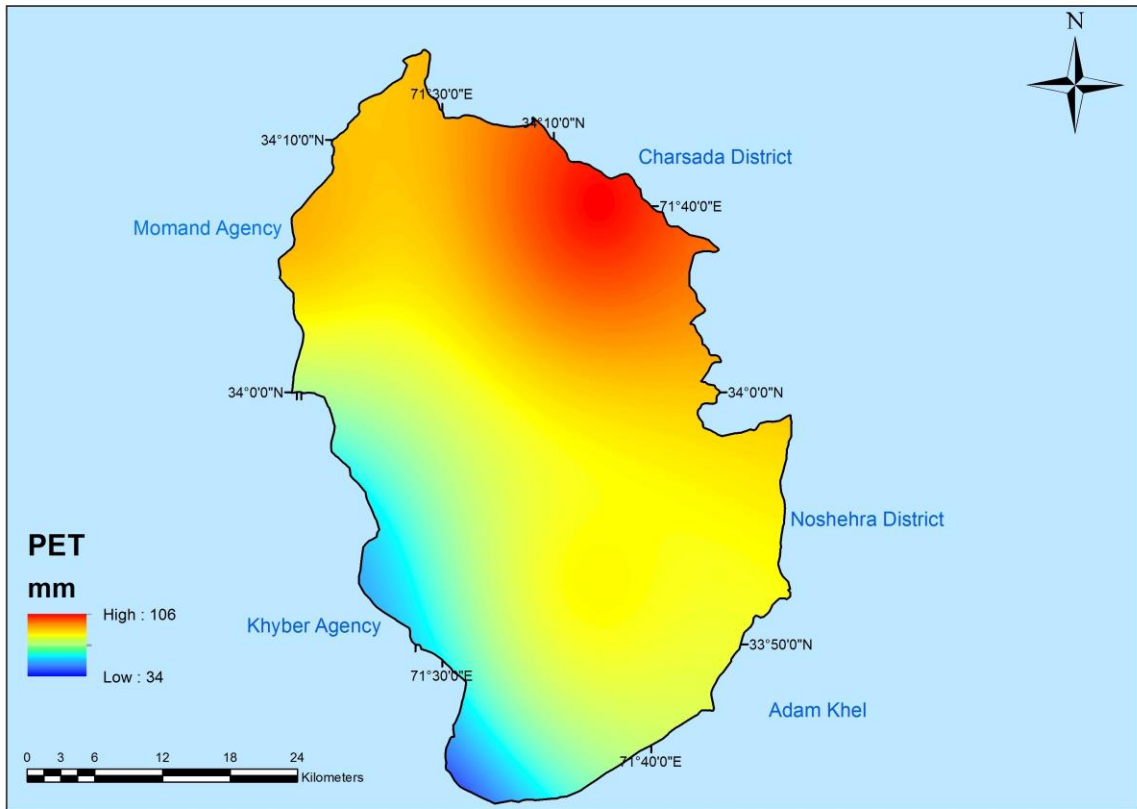
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579 **Figure 2.** Model for predicting the PMD precipitation from TRMM precipitation data.



580

581 **Figure 3.** PMD predicted precipitation map.

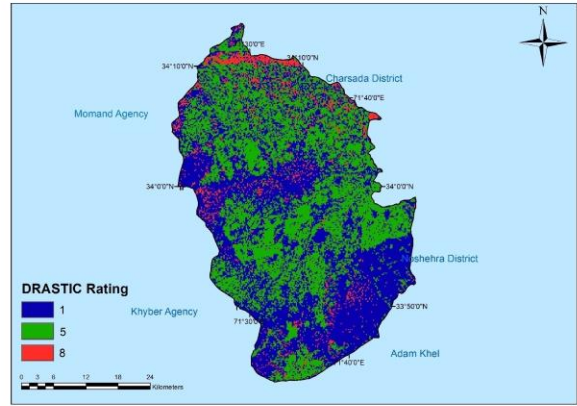


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583 **Figure 4.** Potential evapotranspiration map (Willmott and Matsuura, 2001)



(a)



(b)



(c)



(d)



(e)



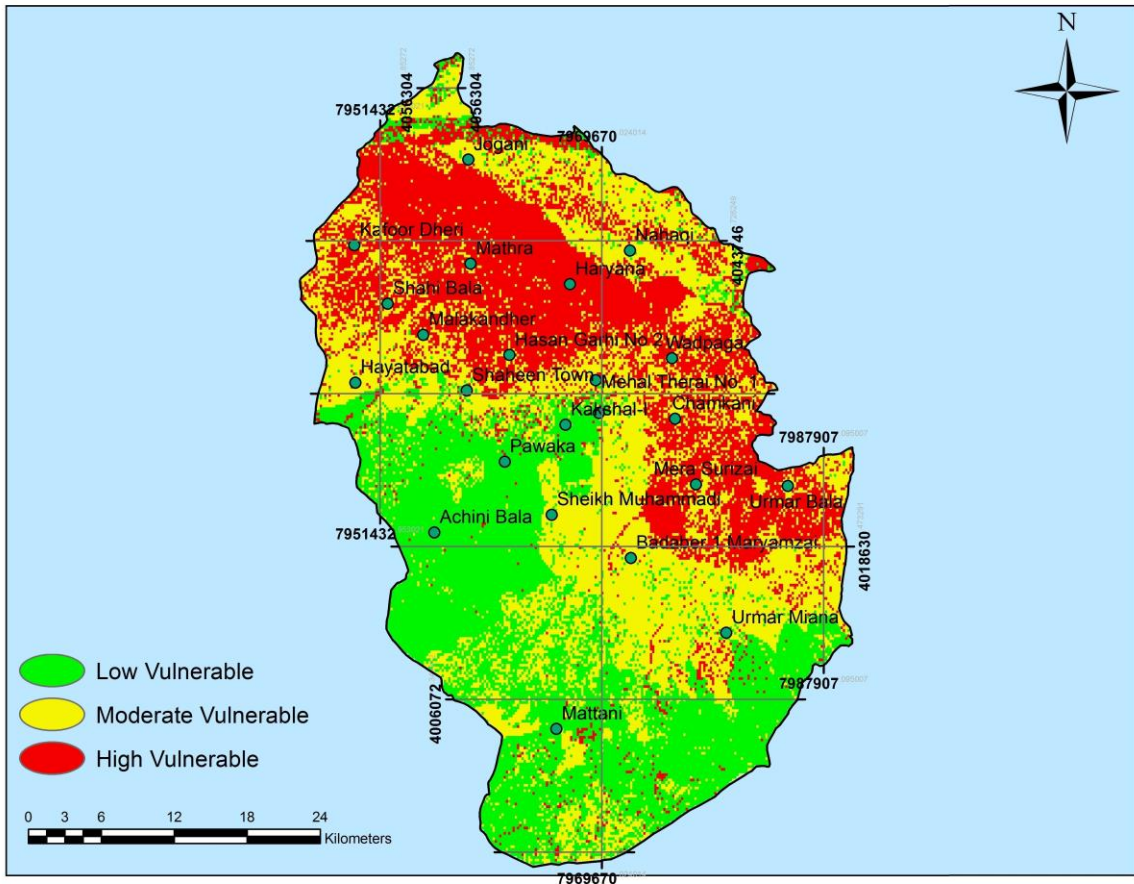
(f)

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585 **Figure 5.** DRASTIC ratings for different ranges or media types of; (a) phreatic levels (b) net

586 recharge (c) aquifer materials (d) soil types (e) slopes, and (f) hydraulic conductivity

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589 **Figure 6.** Final map of groundwater vulnerability in the study area.