

1 **Integrated multi-criteria analysis for groundwater potential mapping in Precambrian**
2 **hard rock terranes (North Gujarat), India**

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14

15 **Abstract**

16 In north Gujarat, local communities depend on Precambrian basement aquifers for their
17 primary source of water supply. Increasing demand for potable water puts stress to explore
18 groundwater from less reliable sources of basement rocks. The objective is to map
19 **groundwater potential zone** (GWPZ) in water-scarce Precambrian terranes using integrated
20 multi-criteria analysis (AHP and Geospatial techniques). Various important thematic has
21 been prepared and the output of the GWPZ map was grouped into six different categories.
22 The results show that very-high and high groundwater potentials extend over an area of
23 4.29% and 12.86% respectively of the total area. The Kappa index method was used for
24 validation and its result showed an effective and reliable result between collected water level
25 data with calculated GWPZ. It shows 83.3% accuracy for very-high, 85.7% accuracy for

26 high, and 90.1% accuracy for moderate GWPZ. The synthesis of the work can be applied
27 anywhere having similar setups for groundwater prospect and management.

28 **Keywords:** Groundwater; Analytical hierarchy process; GIS; Electrical resistivity
29 tomography (ERT); Ambaji basin; Precambrian terrane

30 **1 Introduction**

31 Groundwater is considered being one of the most precious and limited resources stored in the
32 subsurface and its utility to human, aquatic, and terrestrial ecosystems are great extent
33 (Freeze and Cherry 1979, Fitts 2002). The occurrence and distribution of groundwater
34 resources mostly rely on several factors such as lithology, geomorphic features, drainage
35 networks, geologic structures (i.e. lineaments/joints/faults), and slope (Greenbaum 1992,
36 Teeuw 1995, Gyeltshen *et al.* 2019). In the last few decades, water demand has been
37 increased significantly, especially in arid and semi-arid regions of the world. The ever
38 increasing water demand has imposed huge pressure on limited freshwater resources. In
39 developing countries like India, groundwater is considered as a significant source for
40 domestic, agriculture, and industrial uses (Mohanty *et al.* 2008). Further, the groundwater
41 resources are depleting day by day due to an increase in population, industrialization, and
42 overexploitation of aquifer systems which are not restored sufficiently (Rodell *et al.* 2009). In
43 India, about three-fourth of the households do not have a right to access to get potable water
44 for their needs (NITI AAYOG Report 2017-18, Govt. of India). The total annual groundwater
45 recharge has been estimated as 432 Billion Cubic Metre (BCM) and the annual extractable
46 groundwater resource is 393 BCM. Further, the total annual extractable groundwater resource
47 is 249 BCM (as of March 2017) (CGWB 2019). The freshwater situation in India will
48 become challenge and alarming by 2025 if proper measures are not taken (World Bank
49 Report 2006). This poses the need for a better understanding and mapping of the available
50 freshwater resources of the country.

51 In recent years, the northwestern Indian states mainly, Gujarat, Rajasthan, Haryana,
52 and Punjab are heavily stressed with water resources due to less rainfall, high
53 evapotranspiration rates, and overexploitation of the aquifer systems. The overexploitation of
54 aquifers leads to depletion in the groundwater level (Rodell *et al.* 2009), **water exhaustion,**
55 **deterioration of water quality issues, etc.** (World Bank 2011, Souissi *et al.* 2017). Most of the
56 northwestern Indian states heavily depend on groundwater resources than surface water
57 resources for their **domestic and** agricultural needs. Due to the huge groundwater-dependent
58 population, variations in climatic condition and indiscriminate land-use changes with
59 urbanization are among many factors for water scarcity in both qualitatively and
60 quantitatively. **The traditional methods used for identifying groundwater potential zones and**
61 **mapping are, hydrogeological surveys, geophysical surveys, logging, and outcrop mapping,**
62 **etc. which are generally expensive and time-consuming** (Sander *et al.* 1996, Sturchio *et al.*
63 **2004, Israil *et al.* 2006, Jha *et al.* 2010, Singh *et al.* 2013). On the other hand, remote sensing
64 (RS) and geographic information system (GIS) **technology** provide a unique platform **that can**
65 **be used for the evaluation of natural resources due to their ability and efficiency in**
66 **developing spatio-temporal information and complex spatial data analysis of a large area**
67 **within a short period** (Tweed *et al.* 2007, Souissi *et al.* 2018, Wieland and Pittore 2017).**

68 In the recent past, **several researchers have used different bivariate and multivariate**
69 **statistical methods** (Table 1) **with different accuracy to delineate** groundwater potential zones
70 (GWPZ) **and mapping** viz. probabilistic frequency ratio model (Manap *et al.* 2014, Naghibi
71 and Pourghasemi 2015, Moghaddam *et al.* 2015), weight-of-evidence model (Corsini *et al.*
72 2009, Ozdemir 2011a, Tahmassebi *et al.* 2016), fuzzy-gamma (Antonakos *et al.* 2014),
73 multi-criteria decision-making analysis (Kumar *et al.* 2014, Machiwal and Singh 2015,
74 Rahmati *et al.* 2015, Singh *et al.* 2018, Al-Ruzouq *et al.* 2019, Arulbalaji *et al.* 2019,
75 Sandoval and Tiburan 2019), Shannon's entropy (Naghibi *et al.* 2015), artificial neural

76 network studies (Banerjee *et al.* 2009, Lee *et al.* 2012), random forest machine learning
77 techniques (Naghibi and Pourghasemi 2015, Naghibi *et al.* 2016, Rahmati *et al.* 2016), and
78 logistic regression methods (Ozdemir 2011a and 2011b, Park *et al.* 2017, Chen *et al.* 2018).
79 Of the various methods, RS and GIS act as a powerful, cost-effective, and rapid in generating
80 results (Oh *et al.* 2011) compared to the traditional approaches of hydrogeological surveys
81 and has a great significance in groundwater exploration in water-scarce Precambrian hard
82 rock terranes (Jha *et al.* 2010). Several studies have already been done using RS and GIS
83 techniques for delineation and mapping GWPZ in different parts of the world (Saraf and
84 Choudhury 1998, Solomon and Quiel 2006, Madrucci *et al.* 2008, Pradhan 2009, Yeh *et al.*
85 2009, Dar *et al.* 2010, Gupta *et al.* 2010, Mukherjee and Singh 2020, Fildes *et al.* 2020,
86 Lentswe and Molwalefhe 2020).

87 In this study, we integrated the analytical hierarchy process (AHP) and geographic
88 information system (GIS) techniques to properly assess and map groundwater potential
89 zones. In 1980, Thomas Saaty first introduced the AHP method that has a huge advantage in
90 groundwater related studies as it reduces the complex results to a sequence of pair-wise
91 assessments and then generates the results (Saaty 1980). Moreover, the AHP tool is a suitable
92 technique for computing the consistency of the results, thus it reduces the bias in the
93 decision-making process. The multi-criteria analysis (i.e. AHP and GIS) can also apply
94 different environmental management purposes, etc. Further, the integration of geospatial
95 analysis for the preparation of different influencing thematic layers, such as geology,
96 lineament, lineament density, hydrogeomorphology, drainage, drainage density, slope, land
97 use land cover, etc. along with suitable weightage assignment using AHP technique facilitates
98 in demarcating groundwater bearing zones and recharge zones. For example, Mukherjee and
99 Singh (2020) successfully identified GWPZ in drought-prone semi-arid regions of East India
100 using AHP and GIS analysis. Kumar and Krishna (2016) evaluated GWPZ in the coal mining

101 area of Ramgarh and parts of Hazaribagh district, Jharkhand, India. Adiat et al. (2012) used
102 the AHP for groundwater potential mapping in Kedah Peninsula, Malaysia and the result
103 stated that the AHP requires pairwise comparison and expert knowledge to assign weightages
104 to different thematic layers. Also, the AHP technique successfully applied for mapping
105 groundwater recharge zones in different parts of the world (Chenini *et al.* 2010, Chowdhury
106 *et al.* 2010, Rahimi *et al.* 2014, Agarwal and Garg 2016, Lentswe and Molwalefhe 2020).

107 About 90% of the Indian Peninsula is covered with Precambrian hard rocks. In those
108 terranes, delineation of GWPZ and mapping are relatively complicated task due to lack of
109 reliable and existing data sets, and heterogeneity in lithology (Anbazhagan *et al.* 2011). By
110 considering all these, here we studied a basin (i.e. Ambaji Basin) from a semi-arid region of
111 Aravalli (North Gujarat), that lies in the NW part of the Indian Peninsula. The large
112 percentage of area is mostly underlain by granitic rocks. The residents in these regions
113 heavily dependent on groundwater resources and faces huge water scarcity throughout the
114 year as there is no proper understanding of the aquifer systems. Despite being a highly water-
115 scarce and overpopulated area, yet there is no such scientific communication has been made
116 related to groundwater prospect zones, management, and developments in this region. So, the
117 main aim of the study is to identify and demarcate groundwater bearing zones through
118 integrated multi-criteria analysis using AHP and Geospatial techniques in water-scarce
119 Precambrian hard rock terranes of the Ambaji basin (North Gujarat), India as an example for
120 sustainable groundwater development and management.

121 **2 Study area**

122 **2.1 Location**

123 The study area (Ambaji basin) covers a part of Aravalli terrane and is situated in Amirgarh
124 and Danta taluka of Banaskantha District (North Gujarat), India (Figure1). The area extends
125 over an area of 450 km² approximately and lies between latitudes 24° 12'N to 24° 21' N and

126 longitudes 72° 31'E to 72° 48' E in the northwest part of Gujarat. The area covers a total of
127 eighteen (18) major villages and divided into two talukas (Amirgarh taluka and Danta taluka)
128 with a population density of ~42,000 based on 2011 Indian census data. The area falls under
129 the Survey of India (SOI) toposheet no. 45D/11, 45D/12, and 45D/15 (1:50,000 scale). The
130 area has experienced a diverse landscape and is characterized by structural hills, various
131 piedmont zones with alluvium, alluvial plains, and residual hills. The elevation ranges from
132 215 to 809 m above MSL in the west- and east-part, respectively. The area belongs to semi-
133 arid climatic regions where extreme heat in the summer months and dryness in winter, erratic,
134 and scanty rainfalls with high evapotranspiration are the typical characteristic climatic
135 features (Pradhan and Biswal 2018). The average annual rainfall is 771 mm (Figure 2) and
136 mostly receives through the south-west monsoon season. The average temperature ranges
137 from 35-42 °C during the summer and 15-30 °C during the winter season.

138 **2.2 Geological and hydrogeological settings**

139 The Ambaji basin lies in Precambrian terranes of Aravalli range of North Gujarat, India. The
140 area consists of granulites (pelitic, calcareous, and mafic varieties) and granites belonging to
141 Neoproterozoic- Cryogenian age. The rocks are deformed by multiple phases of folding,
142 ductile shearing, and brittle faulting (Singh *et al.* 2010, Tiwari and Biswal 2020). Based on
143 the geographical setting and the south-western monsoon, the study area experiences a heavy
144 water scarcity during summer. In the summer months, the groundwater table lowers down to
145 several meters, which affects directly the livelihood (Machiwal *et al.* 2011). The major aquifer
146 systems are fractured rock granitic aquifers. The main sources of drinking and irrigation use
147 in this region were groundwater resources and less on surface water. In the recent past,
148 surface water (Balaram River) was primarily used for domestic and irrigation use but in
149 recent years the surface water (river) dried up due to less and erratic rainfalls. As the area
150 falls under a semi-arid climatic region, the rivers namely Balaram River and Teliya Nadi

151 flowing in the basin are ephemeral (Pradhan and Biswal 2018). The drainage network is
152 mainly constituted by Banas River and the tributaries therein viz. Balaram River and Teliya
153 Nadi. The aquifer systems are constituted of two hydrogeological units in this area. The first
154 one is shallow unconfined or weathered aquifer units range a few meters. The second
155 hydrogeological unit is the fractured bearing layer ranges from 15 m to 90 m (and above).
156 The groundwater table (below ground level) ranges from 2.5 to 20.65 m (in pre-monsoon)
157 and 0.65 to 17.85 m (in post-monsoon) and the regional flow direction of groundwater is
158 from north-east to south-west (Pradhan *et al.* 2019 accepted).

159 **3 Data and Methodology**

160 ***3.1 Data used and preparation of different thematic layers***

161 For the present work, geospatial methods were used for mapping GWPZ in Precambrian
162 terranes of the Amabji basin (Aravalli terrane) using factor analysis of 10 groundwater
163 influencing layers. To prepare the GWPZ map, different thematic layers viz. geology,
164 lineament density, lineament proximity, LULC, hydrogeomorphology, slope, drainage
165 density, drainage proximity, groundwater level, and groundwater quality (total dissolved
166 solids) were generated from remote sensing image, existing Geological map (Geological
167 Survey of India, 1: 50,000 scale), SOI toposheet, and through field data. For the lithological
168 study, the GSI geological map was used to identify different geological units followed by
169 ground truth verifications. The pre-processing analysis of remotely sensed data of the Ambaji
170 basin was performed using ERDAS Imagine 2011. ArcGIS 10.1 software package was used
171 for preparing thematic layers (Table 2). The hydrogeomorphology and land use land cover
172 features were demarcated from Resource Sat2 LISS IV (acquired from NRSC, ISRO) image
173 through visual interpretations. The Shuttle Radar Topography Mission (SRTM) (30m
174 resolution) data was employed for the demarcation of the basin boundary. The detailed
175 methodology used in this study was presented in a flowchart (Figure 3). The SOI toposheets

176 (45D/11, 45D/12, and 45D/15) were used for the extraction of drainage layers. All the input
177 data sets were georeferenced with a WGS1984 coordinate system. Further, the geospatial
178 results were validated with groundwater level (depth to the water table), well yield data, and
179 subsurface resistivity data. A sum of fifty-eight groundwater level (dug well/bore well) data
180 were collected during 2016-18 through fieldwork and from existing data sets obtained from
181 Central Groundwater Board, Ahmedabad. The data were then used through an Inverse
182 Distance Weighted (IDW) interpolation tool in ArcGIS 10.1 for making the spatial
183 distribution of the groundwater level. A total of 9 well yield data has been collected through
184 fieldwork and from CGWB and GWRDC, Ahmedabad. For subsurface resistivity study, a
185 total of seven (07) electrical resistivity tomography (ERT) traverses were conducted using
186 IRIS SYSCAL Pro 48 multi-electrode channel (France) resistivity system in December 2017
187 and 2018. A combined inversion Wenner-Schlumberger (WS) configuration has been used
188 with an electrode spacing of 5.0 m in this survey. The total spreading length for each profile
189 was 235.0 m. The obtained WS datasets were processed through a RES2DINV inversion
190 software package to determine the distribution of true resistivity from apparent resistivity
191 values (Griffiths and Baeker 1993). Further, to compare with field data, the obtained WS data
192 sets were inverted using multiple iterations to derive a model distribution that produces a
193 section with a root mean square (RMS) less than or equal to 5% (Acworth 2001, Pradhan *et*
194 *al.* 2019 accepted).

195 **3.2 Analytical hierarchy process (AHP) technique**

196 The AHP method is a semi-quantitative and multi-criteria analysis method established by
197 Thomas Saaty in 1980 for analyzing the complicated problems. This enables the user to
198 arrive at a scale of choices drawn from multiple alternatives (Yalcin *et al.* 2011). In recent
199 years, AHP method gained wide applicability in site selection, landslide susceptibility
200 analysis, groundwater potential zone mapping, regional planning, etc. (Ayalew *et al.* 2005,

201 Yalcin *et al.* 2011, Pourghasemi *et al.* 2012, Althuwaynee *et al.* 2014, 2016, Mumtaz *et al.*
202 2019). Based on Saaty's scale of preference, the AHP method was implemented to estimate
203 weight factors and the rankings of the class in every groundwater influencing parameter over
204 the pair-wise comparison matrix (Saaty 1980). The AHP technique helps to integrate all
205 thematic layers and for this study, sums of ten thematic layers were used. The selected
206 thematic layers are expected to be the controlling parameters for groundwater occurrence and
207 movement in this area. The associations of these selected influencing parameters weighted
208 based on their interrelationships and the experts' opinion. An influencing parameter with a
209 high weightage demonstrates high influence whereas a parameter with a low weightage
210 demonstrates low influence on groundwater prospects (Arulbalaji *et al.* 2019). The weightage
211 of each influencing parameter was assigned based on Saaty's scale (Table 3) which varies
212 from 1 (equal importance) to 9 (extremely high importance). In this study, the weightage
213 assignments given based on the existing literature knowledge and field experience. Further,
214 based on the Saaty's classification, different weightages were assigned to different layers
215 based on their significance and water-bearing capacity. Saaty (1980) stated that the scale will
216 be between 1 to 9 in the pair-wise comparison matrix if the influencing factors on the vertical
217 axis have a higher degree of preference with reference to the horizontal axis (Balamurugan *et*
218 *al.* 2017). On the other hand, the scale will be reciprocals (1/2 to 1/9), if the influencing
219 factors in the horizontal axis of the matrix have higher degrees of preference with reference
220 to the vertical axis (Table 4). For this work, the scale preference has been used based on
221 objective and subjective methods. Further, the input values are given based on influencing
222 factors for groundwater potentials as well as subjective knowledge (Balamurugan *et al.*
223 2017). Saaty (1980) formulated the effect of subjectivity through the consistency index (CI)
224 using Eq. (1), and the consistency ratio (CR) using by Eq. (2)

225
$$CI = \frac{(\lambda_{\max} - 1)}{(n - 1)} \text{-----} (1)$$

226
$$CR = \frac{CI}{RI} \text{-----} (2)$$

227 where “ λ_{max} ” denotes the Principal Eigen value and “n” is the number of the causative factors
 228 used in the pairwise comparison matrix.

229 Further, CR can be formulated by taking a ratio of CI/RI, where RI is the random
 230 index based on the number of random samples (Saaty 1980) (Table 4). In the pair-wise
 231 comparison matrix, if $CR > 0.1$, it means there is an existence of inconsistency and requires
 232 further modification in the scale preference (Saaty 1980). To prepare a GWPZ map of the
 233 Ambaji basin (North Gujarat), all ten thematic layers were incorporated with all the scores
 234 acquired by multiplying the weight of factor with a rating of the class using Eq. (3)

235
$$GWPZ = \sum_{j=1}^n (W_j \times w_{ij}) \text{-----} (3)$$

236 where, GWPZ is the groundwater potential zone, “ W_j ” represents the weight value of
 237 causative factor “j”, “ w_{ij} ” denotes the weight value of class “i” in causative factor “j”, and
 238 “n” is the number of causative factors. The final GWPZ map was then classified into six
 239 categories i.e. very-high, high, moderate, moderately-low, low, and very low.

240 ***3.3 Delineation of groundwater potential zone (GWPZ) using AHP technique***

241 For this study, the GWPZ signifies groundwater in quantifiable form, i.e. high groundwater
 242 potential index (GWPI) signifies the higher groundwater potentiality and lower GWPI
 243 signifies the lower groundwater potentiality. The GWPI prepared through the incorporation
 244 of all thematic layers in the raster calculator tool in Arc GIS 10.1 software package by using
 245 Eq. (4).

246
$$GWPI_{AHP} = \sum(GE_{AHP} + HG_{AHP} + LULC_{AHP} + LD_{AHP} + DD_{AHP} + WL_{AHP} + LP_{AHP} + DP_{AHP} +$$

 247 $SL_{AHP} + GWQ_{AHP}) \text{-----} (4)$

248 Where $GWPI_{AHP}$ =Groundwater Potential Index. AHP – Analytical Hierarchy Process, GE_{AHP} -
 249 geology, HG_{AHP} - hydrogeomorphology, $LULC_{AHP}$ - land use and land cover, LD_{AHP} -
 250 lineament density, DD_{AHP} - drainage density, WL_{AHP} - water level, LP_{AHP} - lineament

251 proximity, DP_{AHP} - drainage proximity, SL_{AHP} - slope, GWQ_{AHP} - groundwater quality (total
252 dissolved solids)

253 **4 Results**

254 *4.1 Thematic layers for groundwater potential zone mapping*

255 *4.1.1 Geology*

256 The role of geological features considered as one of the most important factors for a
257 hydrogeological investigation study. The main attention is to be made on lithological units,
258 which control the storage, flow, and chemistry of the groundwater system. The different
259 litho-units demarcated based on the fieldwork, visual interpretations of the satellite image,
260 and existing geological map. Geologically, the study area comes under the Meso-proterozoic
261 age of South Delhi terrane (1100-900 Ma) of Aravalli mobile belt (Singh *et al.* 2010). The
262 rocks found in this terrane were mainly of pelitic-, calcareous- and basic granulites, where
263 three phases of granite intruded namely G1 (gneissic), G2 (medium to coarse-grained) and
264 G3 (fine-grained) (Pradhan and Biswal 2018). Unconsolidated sediments and fractured hard
265 rocks are more suitable for groundwater occurrence and movement than a massive
266 metamorphic or igneous rock. Secondary porosity (faults, fractures, and shear zones) and
267 weathered zones only allow groundwater flow and occurrence in hard rock terrane (Singhal
268 and Gupta 2010). From a geological point of view, the granites (G2, G3, and G1),
269 alluvial/sand, and recent sediments were the major aquifers in this area (Figure 4a). The G1,
270 G2, and G3 granites named after their geologic age. Further, G2 and G3 granites are coarse to
271 fine-grained and highly fractured, whereas G1 granite is gneissic in nature (Pradhan and
272 Biswal 2018). The thickness of weathered zones varies from 5-20 m. The fractures
273 encountered at a depth of 5-150 m in hard rocks. A proper weightage assignment has been
274 given to the hydrogeologically important rock types. Based on the rock types, a high
275 weightage is assigned to G2 granite and alluvial/sand, moderate to moderately-low weightage

276 for G3 granite and recent sediments, and low to very-low weightage assigned for G1 granite,
277 basic granulite, and calc-granulites (Table 6).

278 *4.1.2 Hydrogeomorphology*

279 The geomorphological setting of an area may affect groundwater potentials to a great extent
280 as it reflects various landforms and topography of an area that has a direct impact on the
281 distribution of groundwater. The identification of geomorphic features is essential for tracing
282 potential groundwater bearing zones. The slope, relief, thickness of soils/weathered zones,
283 type of weathering, and different landforms are the result of long-term denudation of fluvial
284 processes. The hydrogeomorphological features of the present study area mapped by using
285 Resource Sat2 LISS IV Satellite image. The different erosional and depositional
286 hydrogeomorphic units viz. structural hills, residual hills, denudational hills, buried
287 pediments, deep buried pediments, shallow pediments, and valley fills are mapped (Figure
288 4b). The highland part mainly consists of hilly terranes with a highly undulating surface,
289 whereas the lowland areas are of less and gently undulating surfaces. The main
290 hydrogeomorphic units in the Ambaji basin are structural hills (44.51%), buried pediments
291 moderate (18.90%), denudational hills (14.42%), deep buried pediments (14.24%), valley
292 fills (4.28%), river (0.74%) and shallow buried pediments (0.52%) (Table 6). Further, the
293 upper catchment of the basin is mainly comprised of structural hills, while the lower
294 catchment is covered with alluvial/sandy zones.

295 *4.1.3 Land-use and land-cover (LULC)*

296 The LULC of a particular area plays an important role in recharge-discharge, infiltration, and
297 runoff, which are mainly controlled by the nature of surface material. The LULC is one of the
298 key parameters in hydrogeological studies because it provides valuable information about the
299 degree of groundwater requirements and utilization (Gupta and Srivastava 2010). Todd and
300 Mays (2005) studied dense vegetation in a land-use map as an excellent indicator of a

301 suitable site for groundwater exploration. The LULC of the study basin was categorized into
302 six different classes viz. agricultural land, fallow land, mixed forest, open forest, river, and
303 scrub forest through visual interpretation of Resource sat2 LISS IV satellite data (Figure 4c).
304 The map was reclassified by assigning different scores for various LULC classes. The scrub
305 forest (64.98%), agricultural land (14.68%), and fallow lands (7.85%) were the major LULC
306 type in the study area. Further, the high score assigned to the river and agricultural lands
307 while for scrub forest and open forest lower score was given due to poor permeability. The
308 various LULC types and their spatial distributions were given in Figure 3c. The different
309 scores assigned to various classes are summarized in Table 6. The hilly terranes and
310 agricultural land are predominantly found throughout the study area. The accuracy was
311 further calculated using the Kappa index method (Balamurugan *et al.* 2017).

312 *4.1.4 Lineament density and lineament proximity*

313 Lineaments are the linear surface features or expression of faults, shear zones, or geological
314 contacts or discontinuities and act as good conduits for the flow of groundwater (Prost 1994,
315 Singhal and Gupta 2010) and have very good water reservoirs. Elewa and Qaddah (2011)
316 mentioned that fractures in a rock enhance the secondary porosity that increases water
317 percolation to the aquifer. Lattman and Parizek(1964) first adopted the linear feature called
318 “lineaments” for groundwater potential in a folded and faulted hard rock terrane. The
319 lineaments generally show the tonal, textural, relief, drainage and vegetation linearity, etc. in
320 a satellite image (Lillesand 1989, Drury 1990, Gupta 1991, Kumar *et al.* 2014). The structural
321 lineaments were mapped with the help of satellite image and field studies for identifying
322 different faults, fractures, and lithological contacts that control the groundwater occurrence.
323 From the satellite data and field mapping, the detected linear features with different sizes,
324 orientations, and distributions were exported and plotted on a map (Elewa and Qaddah 2011).
325 Furthermore, not all linear features in the satellite image are lineaments; hence, there need to

326 be ground truth verification by field visits. The straight course of streams or vegetation
 327 linearity may be due to faulting and fracturing, therefore inferred as the areas of high porosity
 328 and permeability in hard rock terrane (Shekhar and Pandey 2015). The areas around
 329 lineaments and crisscrossed lineaments are considered the most favorable site for
 330 groundwater occurrence due to high infiltration and transmitting capacity. Solomon and Quiel
 331 (2006) studied that high chance for getting groundwater bearing zones are close to the
 332 structural lineaments as the intensity of fractures are decreased with increasing distance from
 333 lineaments. For the present study, lineaments were prepared from Resource Sat2 LISS IV
 334 satellite imagery through visual interpretation based on the existence of linear features and
 335 tonal contrast followed by field verifications. The study area comprises numerous criss-
 336 crossed lineaments in the northern part and north-western part due to transtensional settings
 337 (Pradhan and Biswal 2020, Pradhan *et al.* submitted). Yeh *et al.* (2009) computed lineament
 338 density of an area as follows:

$$Ld = \frac{\sum_{i=1}^{i=n} Li}{A} \text{-----} (5)$$

340 Where Ld is termed as lineament density, $\sum_{i=1}^{i=n} Li$ is refers to the total length of lineaments,
 341 and A is the unit area.

342 Lineament density is one of the important aspects prepared from the lineament map,
 343 which is critically used in groundwater studies especially in hard rock terrane. The
 344 importance of the lineament density study is to calculate the frequency per unit area
 345 (Narendra *et al.* 2013). Based on the lineament density analysis, concentrations of the
 346 lineaments have been prepared for the study area (Figure 4d). The higher lineament density
 347 areas are considered excellent aquifer zones for groundwater resources. Based on the
 348 lineament density of a region higher to lower weightage has been assigned. In the map
 349 (Figure 4d), it has been observed that a high density of lineaments falls in areas of granitic
 350 gneiss, granite, and basic granulites/gabbroic rocks. On the other hand, lower to moderate

351 lineament density is observed in the alluvium and sandy zones in the central and
352 southwestern corners of the study area. Using the Euclidean method, the distance from
353 lineaments was calculated in ArcGIS 10.1. The lineament proximity (Figure 4g) was further
354 categorized into five class viz. 0 – 91.764, 91.764 – 229.411, 229.411 – 416.470, 416.470 –
355 645.882, and 645.882 – 900 m (Table 6) using natural break methods (Jenks 1967).

356 *4.1.5 Drainage density and drainage proximity*

357 Drainage is the pattern formed mainly by streams and rivers in a particular drainage basin.
358 These are mainly controlled by topography dominated by crystalline hard rocks or
359 unconsolidated rocks and the gradient of the land surface. The drainage transmitted water
360 from the hilly regions to the plains (Balamurugan *et al.* 2017). Drainage density has a major
361 role in exploring groundwater resources for a particular region. The drainage systems in hard
362 rock terrane are mainly controlled by different structural entities, i.e. faults and fractures, and
363 hence drainage systems act as a medium for water movement into the subsurface
364 (Balamurugan *et al.* 2017). The drainage of the Ambaji basin was extracted from the SOI
365 toposheet and ASTER DEM and verified with an acquired satellite image.

366 The drainage density is measured as the closeness of channels and higher the drainage
367 density of a basin, lower the infiltration rate. This implies that the drainage density is an
368 inverse function to permeability. Dinesh Kumar *et al.* (2007) mentioned that higher drainage
369 density is not favorable for groundwater prospects. The drainage density and drainage type
370 provide useful information about the runoff, runoff rate, infiltration, and permeability of a
371 particular watershed. The dendritic drainage pattern indicates homogenous rocks, whereas the
372 trellis, rectangular, and parallel drainage patterns indicate structural and lithological controls.
373 The drainage density of the study basin was calculated by the total length of all the rivers and
374 streams in the basin divided by the total area (Figure 4e). The drainage density percentage is
375 computed through the line density tool in ArcGIS 10.1 (Agarwal and Garg 2016). Further, the

376 drainage proximity (Figure 4h) was calculated using the Euclidean method and categorized
377 into five different classes viz. 0 – 88.235, 88.235 – 215.294, 215.294 – 409.411, and 409.411
378 – 900 m (Table 6) based on the natural breaks method (Jenks 1967).

379 *4.1.6 Slope*

380 Slope defines the difference in elevation for a specific area and has a direct influence on the
381 surface and subsurface runoff (Naghibi *et al.* 2016, Balamurugan *et al.* 2017). The slope
382 gives critical information about the regional geologic and geodynamic processes functioning
383 in the subsurface of the earth (Arulbalaji *et al.* 2019). In general, gentler the slope, slower is
384 the movement of water and in turn more percolation into the subsurface (Balamurugan *et al.*
385 2017). On the other hand, a higher or steeper slope promotes an increase in surface runoff and
386 in turn decreases the percolation capability into the subsurface. The slope map was generated
387 from ASTER DEM (30 m resolution) (Figure 4i). The slope degree ranges from 1 to 30 and is
388 classified into seven classes, i.e. <1, 1-3, 3-5, 5-10, 10-15, 15-30, and >30° based on the
389 natural breaks method (Jenks 1967). The high weightage has assigned to a flat slope or gentle
390 slope (<1°) while lower weightage has given to a steep slope (>30°) based on their
391 importance in groundwater recharge and discharge (Table 6).

392 *4.1.7 Depth to the water table*

393 The depth to the water table or groundwater level mainly depends on the hydrogeological
394 conditions, rainfall pattern, and topography of an area (Eleswa and Qaddah 2011).
395 Groundwater level represents the top surface of the saturated zone. Rainfall and heavy
396 pumping play a major role in groundwater fluctuations of water level and variations in
397 groundwater recharge-discharge. The groundwater level data were collected through
398 fieldwork during the pre- and post-monsoon period (2017-18) and from CGWB, Ahmedabad.
399 The groundwater levels were measured from 58 observed wells (Dug wells/ Bore wells). The
400 groundwater level in this region varies from 2.5 to 20.22 m (Figure 4f). These data were used

401 for plotting the groundwater level map and then classified into five classes, i.e. 2.50-7.92,
402 7.92-9.59, 9.59-11.26, 11.26-13.27, and 13.27-20.22 m (Table 6) below the groundwater
403 level. The weightage was assigned based on the groundwater level values. Generally,
404 shallower the groundwater table higher the groundwater potential and deeper the groundwater
405 level indicates less groundwater potential.

406 *4.1.8 Groundwater quality (total dissolved solids)*

407 The groundwater quality of an aquifer system mainly depends on the local geological
408 formation, anthropogenic activities, and drainage networks (Freeze and Cherry 1979). While
409 accessing groundwater resources quantitatively, it is also essential to check the water quality
410 for a healthy society. World Health Organization (WHO) set a different permissible limit for
411 different ion for drinking and domestic uses. The water sample consists of major and minor
412 ions. A total of 42 groundwater samples were collected from different dug well/bore
413 well/hand pumps for water quality study. The total dissolved solids (TDS) of the water
414 samples are calculated by multiplying 0.45 with electrical conductivity (EC). The existence
415 of dissolved solids in groundwater may affect the taste and WHO suggested a scale of limits
416 for domestic use (excellent, <300 ppm, good, 300 - 600 ppm, fair, 600 - 900 ppm, poor, 900 -
417 1200 ppm, and unacceptable, >1200 ppm). The ranks are assigned based on WHO limits of
418 the TDS level to identify and map the quality of the groundwater in the study area. The
419 results show that most of the area falls in good quality of water (301-600 ppm) zone except
420 for some samples with a higher value in the central part of the basin(Figure 4j) (Table 6).

421 **5 Discussions**

422 *5.1 Groundwater potential zone (GWPZ) mapping*

423 Groundwater is considered as a natural resource, but due to anthropogenic activities through
424 aquifer overexploitation for irrigation and domestic use, the life-sustaining resource has been
425 deteriorated considerably day by day. Therefore, a proper understanding of the groundwater

426 resources is required for effective planning and management for sustainable use. The
427 hydrogeological setting of the Ambaji basin shows that the groundwater occurs in top
428 unconfined zones, fractured aquifer zones especially in weathered and fractured granitic
429 rocks, and alluvial/sand, and recent sediments. Also, groundwater occurs in semi unconfined
430 aquifers in the deep-seated fractured rock aquifers (Pradhan and Biswal 2020). The
431 groundwater availability is varied with space and time; hence, a detailed assessment of the
432 groundwater resources is required. For the present study, ten groundwater influencing
433 parameters (geology, hydrogeomorphology, LULC, lineament density, lineament proximity,
434 drainage density, drainage proximity, slope, groundwater quality, and groundwater level) are
435 considered for GWPZ mapping. The weightage overlay method has been employed for the
436 generation of GWPZ in the Ambaji basin (Table 5). These are further classified as very-high,
437 high, moderate, moderately-low, low, and very-low potential zones (Figure 5) and the aerial
438 spread percentage of these categories are 4.29, 12.86, 19.67, 24.98, 25.81, and 12.40
439 respectively (Table 7). In the groundwater potential map, very high and high GWPZ are
440 concentrated in the northern and eastern end of the basin. Very high to high potentials zones
441 are predominantly found in large-scale faults and fractures-prone areas, which helps the
442 infiltration ability to aquifer systems. The moderate potential zones were found mainly in the
443 north-western and south-eastern zones of the study basin. The low to very-low potential
444 zones fell in the central part and the eastern end of the study area. The general flow direction
445 of groundwater in the Amabji basin is from north-east to south-west. All the wells are agreed
446 with the GWPZ categories in the study area.

447 **5.2 Validation of results**

448 *5.2.1 Depth to water table data*

449 Validation of scientific outputs is the most crucial part as far as groundwater prospect is
450 concerned. For validation of the GWPZ map, we have analyzed a total of 58 groundwater

451 level data with GPS locations (Figure 6). The groundwater level ranges from 2.5 m to 20.65
452 m in this region. These are classified into six categories, i.e. 2.5-6.4, 3.5-9.0, 8.2-12, 9.3-13.5,
453 9-18.3, and 11.5-20.6m and were referred as very-high (6 wells), high (14 wells), moderate
454 (11 wells), moderately-low (3 wells), low (12 wells) and very-low (12 wells) water table.
455 These observed data sets were used as a reference point to calculate and compare the
456 accuracy in the classifications. Using the Kappa index method, the water level data overlay
457 on the GWPZ map for comparison of accuracy. The Kappa method suggests a valid and
458 reliable result between collected depths to water table data sets with calculated GWPZ
459 (Balamuruganet *al.* 2017). Based on the Kappa method we have obtained 83.3% accuracy for
460 very-high GWPZ, 85.7% accuracy for high GWPZ, and 90.1% accuracy for moderate
461 GWPZ. Moreover, the field well data also shows the average accuracy of GWPZ of the study
462 area as 86.6%. The overlay analysis shows that water level with shallower to a moderate
463 depth well is within high to moderate GWPZ and water table with higher depths well is
464 within low to very-low GWPZ.

465 *5.2.2 Well yield data and electrical resistivity tomography (ERT) profiles*

466 In addition to the depth to groundwater table data, the delineated GWPZ map has also been
467 validated with two more important field-based data sets i.e. electrical resistivity imaging
468 (ERT) profiles and limited well-yield data (collected from CGWB, GWRDC, and through
469 fieldwork) (Figure 7). The well-yielding capacity of a specific region mainly relies on the
470 permeability and availability of subsurface water in the aquifer system. A total of 9 well yield
471 data (Table 8) were analyzed for this study. The results show that very high to high GWPZ
472 have a yield capacity of 120-200LPM and moderate to moderately low GWPZ have 90-110
473 LPM yield capacity. However, low to very low GWPZ have a yielding capacity of less than
474 90 LPM. Out of 09 observation well, 08 wells has a good correlation with the AHP based
475 GWPZ map and observation wells. A total of 01 well (i.e. PW-7) does not match with the

476 GWPZ map due to one or more other reasons. The application of electrical resistivity
477 tomography (ERT) is considered to be one of the important geophysical surveys for
478 groundwater exploration to demarcate groundwater bearing zones, especially in water-scarce
479 hard rock terranes. To validate with geospatial data, ERT (combined Wenner-Schlumberger
480 configuration) profiles have been used to delineate the weathered and fractured zones in the
481 Ambaji basin. Generally, the subsurface resistivity values depend on the lithological units and
482 fracture systems present in an area (Pradhan *et al.* 2019 accepted). For this study, instead of
483 traditional four-electrode systems (1D) resistivity survey, electrode resistivity tomography
484 (2D and 3D imaging) traverses were conducted as it produces more realistic models to
485 understand a complex hard rock aquifer system like Ambaji basin (Aravalli terrane). A
486 combined inversion Wenner-Schlumberger (WS) mode has been arranged for this study as it
487 provides a very good resolution (signal-to-noise-ratio) of the lateral and vertical resistivity
488 variations of the subsurface (Loke 2004, Chandrasekhar *et al.* 2014, Cardarelli and De Donno
489 2019, Pradhan *et al.* 2019 accepted). The 2D resistivity section (Figure 8) shows that the
490 study area mainly consists of weathered zones, fractured, and fresh basement zones. The
491 weathered zones may be divided into upper and lower units as it shows a variation in
492 resistivity values due to different mineral composition and the degree of weathering of rocks.
493 The results show that all the resistivity models have approximately 1-15m thick layer of
494 weathered zones in the top and followed by fractured zones (5-40 m) and basement rocks
495 (35m onwards). The ERT sections (Figure 8P1-P7) show the electrical resistivity with three
496 different zones (labeled with X: soil and weathered granites, Y: fractured zones, and Z:
497 massive hard rock or basement rocks). The upper zone has low resistivity ranges from 5 to
498 120 Ω .m and a thickness of 20 to 30 m, indicating a water-bearing fractured and weathered
499 zone. The lower zone (labeled as Z) has a higher resistivity, indicating a massive basement
500 and granitic body. Vertical or steeply fractured zones (Figure 8P2-P5) are identified in Surela

501 and Bhedla areas. These zones can be targeted for pinpointing borehole sites. Generally,
502 groundwater storativity in basement rocks increases with an increase in overburden
503 weathered zones and fractured zones. The weathered zones represent basement rocks that are
504 relatively high porosity and permeability and can be used for sitting large-diameter dug wells.
505 A borehole was drilled up to a depth of 65 m encountering soil layer, weathered granite,
506 fractured zone, and fresh granite basement as incurred from the lithology of the well (Figure
507 9) in Surela area (Figure 8P4). This shows a good pact between resistivity values, well yield
508 data, and GWPZ map.

509 Based on the results, it can be inferred that AHP and GIS-based methods are powerful
510 methods for mapping GWPZ qualitatively and quantitatively from a local to regional scales.
511 Further, this method could be applied for better planning and management for long term
512 groundwater sustainability, especially in water-scarce Precambrian hard rock terranes.

513 **6 Conclusions**

514 In this study, groundwater potential zones (GWPZ) have been delineated through integrated
515 multi-criteria approach (AHP and GIS techniques) in water-scarce Precambrian terranes of
516 the Ambaji basin (North Gujarat), India. The results show that combinations of both
517 techniques are effective tools for understanding the groundwater occurrence and behaviors of
518 the study area. A total of ten groundwater influencing parameters (geology,
519 hydrogeomorphology, lineament density, lineament proximity, drainage density, drainage
520 proximity, LULC, slope, groundwater level, and groundwater quality) were taken into
521 considerations to delineate the GWPZ. Besides other influencing parameters, water quality
522 data, which is very crucial for qualitative as well as quantitative studies of groundwater
523 resources, are considered for this study. The GWPZ map was classified into six categories,
524 such as very-high, high, moderate, moderately-low, low, and very low. The very-high zone
525 indicates the most favorable zone, whereas the very-low zone indicates the least favorable

526 zone for groundwater prospects. Very-high and high GWPZ mostly located in the upper
527 catchment of the basin and covers an area of 4.29% and 12.86% respectively of the total area.
528 Moderate and moderately low GWPZ spreads about 44.65% of the total catchment area.
529 Very-low and low GWPZ located in the lower catchment of the basin and covers an area of
530 12.4% and 25.81% of the study area. The validation of the observed depth to water table data
531 sets shows a strong agreement with the GWPZ map. Also, for validation, well yield data and
532 electrical resistivity imaging based on multi-electrode Wenner-Schlumberger (WS)
533 configuration have been used to understand the groundwater system. The basement feature
534 has been delineated through WS configuration that involves saprolites, weathered-, fractured-
535 , steeply contact zones, and fresh basements. These indicate that the WS configuration for
536 resistivity imaging is sensitive to vertical features such as faults and fractured zones and
537 steeply contacts which has a significant role for sitting bore-holes/dug wells for sustainable
538 groundwater use. Further, for a rigorous validation with AHP outputs, a large number of
539 pumping test (well yield) data will be much more needed for the sustainable management of
540 groundwater resources. Further, the proposed methodology is found to be an important and
541 effective method that can be applied to arid- and semi-arid regions, especially in the water-
542 scarce Precambrian hard rock terranes. Finally, the output of the GWPZ map will be useful
543 for local communities as well as decision-makers for planning and management for sitting
544 hand pumps/bore wells/dug wells to meet the groundwater requirements qualitatively and
545 quantitatively.

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552 essential hydrogeological data.

553 **Figure captions:**

554 **Figure 1.** Location map of the study area (Amabji Basin, Aravalli terrane).

555 **Figure 2.** Rainfall distribution in the study area.

556 **Figure 3.** Flowchart of the methodology used for groundwater potential zone (GWPZ) map.

557 **Figure 4.** Thematic layers of the Ambaji basin a. geology, b. hydrogeomorphology, c. land
558 use/land cover, d. lineament density, e. drainage density, f. water level, g. lineament
559 proximity, h. drainage proximity, i. slope, j. Groundwater quality (TDS-total dissolved
560 solids).

561 **Figure 5.** Groundwater potential zone (GWPZ) **map of Ambaji Basin**

562 **Figure 6.** Validation of the GWPZ map.

563 **Figure 7.** Validation of GWPZ map with well yield data and ERT profiles.

564 **Figure 8.** Resistivity profiles in the Ambaji regions (P1; AjapurMota, P2; Surela-1, P3;
565 Surela-2, P4; Surela-3, P5; Bhedla, P6; Kansara, P7; Ghoda), X; weathered and unconfined
566 zones, Y; fractured zones, Z; massive or basement rocks.

567 **Figure 9.** Subsurface lithology profile in the Surela area (Borehole B1; Refer in Figure 8P4).

568

569 **Table captions:**

570 **Table 1.** Recent groundwater studies using remotely sensed data and different statistical
571 methods.

572 **Table 2.** Details of source data.

573 **Table 3.** The scale of preference between two parameters in AHP (after Saaty, 1980).

574 **Table 4.** Random Consistency Index (RI) (after Saaty, 1980).

575 **Table 5.** Weight criteria for different layers.

576 **Table 6.** Weightage of different thematic layers for groundwater potential zone mapping.

577 **Table 7.** Groundwater potential zone derived from AHP.

578 **Table 8.** Observation well locations and the yield data.

579

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