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Assessment of Irrigation Potential in Jewuha Watershed, Middle Awash River Basin, Ethiopia

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1 ABSTRACT

2 Ethiopia is endowed with immense water and land resources that could be tapped and used for irrigation
3 development. However, little has been done so far due mainly to lack of comprehensive knowledge on the potential
4 of these resources. This study was, therefore, taken up to assess the irrigation potential in data scarce Jewuha
5 watershed of the Awash River Basin. To achieve the objective of the study, the suitability of the land for both
6 surface and pressurized irrigations was estimated using a parametric evaluation technique combined with the FAO
7 land evaluation framework in the GIS environment with multi-criteria analysis of irrigation suitability factors. The
8 irrigation water demands of major crops adopted in the area were determined using CROPWAT software. The
9 surface water potential at the sub watershed was estimated using a combination of SWAT model and the spatial
10 proximity regionalization technique. The performance of the SWAT model was checked by statistical parameters.
11 The irrigation suitability analysis reveals that 51372ha of the study area are suitable and 16274ha are unsuitable for
12 surface irrigation system. Furthermore, 52768ha of the area are suitable and 15088ha are unsuitable for a sprinkler
13 irrigation system. For drip irrigation, 52751ha are suitable and 14855ha are unsuitable. The calibration and
14 validation of SWAT model showed that the model has performed well to simulate the hydrology of the watershed
15 with a coefficient of determination (R^2), Nash-Sutcliffe efficiency (NSE) and Percent of bias (PBIAS) of 0.74, 0.73
16 and 0.80 for calibration and 0.71, 0.70 and 7.90 for validation, respectively. From a combined analysis of the
17 available land and water resources, the gross irrigation potential of the area is estimated to be 12,997ha, of which
18 3098ha of land is exclusively suitable for surface irrigation and 9,899ha is suitable for pressurized irrigation
19 systems.

20 **Key words:** Land suitability, SWAT, GIS, CROPWAT, Irrigation potential

21 1 Introduction

22 Over 85 percent of Ethiopia's population lives in rural areas and relies on agriculture for subsistence (Awulachew et
23 al. 2007). The country has huge water resource potential that comprises 12 river basins with an annual runoff
24 volume of 124 billion m^3 of water and over 2.6 billion m^3 of groundwater potential (Melesse et al. 2014; Worku
25 Ayalew 2018). But there is not sufficient water for most farmers to produce more than one crop per year due to a

26 lack of water storage structures and large spatial and temporal variations in rainfall. Therefore, irrigation
27 development and improved agricultural water management practices could provide opportunities to cope with the
28 impact of climatic variability, enhance productivity per unit of land, and increase the annual crop production volume
29 significantly(Awulachew and Merrey 2005).

30 Ethiopia also has a total cultivable land of between 30 to 70 million hectares out of 112 million hectares, but only 15
31 million hectares of land is under cultivation. However, only about 4% of this cultivated area has been irrigated so far
32 (Awlachew et al. 2010), despite the fact that there is no precise figure for the potential and actual irrigated area. This
33 is due to a lack of consistent, reliable inventory, well-studied and documented data. Also, this shows that there is a
34 lack of detailed studies in the area. So, assessment of irrigation potential for irrigation development is important to
35 utilize the land resources efficiently for the sustainable production of crops and to sustain the food security of the
36 rapidly increasing population in the country (Haile 2015). Among the 12 river basins in Ethiopia, the Awash river
37 basin is one of them, covering a total area of 110,000km². The Awash river basin is the most intensively utilized
38 river basin for irrigation development in Ethiopia due to its strategic location, accessibility, available land, and water
39 resources (Awash Basin Authority 2017). However, the most utilized part is the upper part of the basin and relative
40 to the total area, there is very little area to be utilized. Due to this, assessing irrigation potential in the Awash River
41 basin is needed for more effective use of the basin for agricultural development.

42 Irrigation development could improve agricultural productivity and enhance socio-economic development through
43 the growth of production (AU 2020). But most of Ethiopia, particularly in the study area, practiced traditional
44 agricultural activity. This shows that irrigation development in the study area is necessary and to do this, first the
45 potential site for irrigation must be identified. Therefore, the objective of this study is to assess the irrigation
46 potential for different irrigation systems in the Jewuha watershed to easily develop an irrigation project and the
47 surface water resource potential is assessed using the SWAT model (Arnold et al. 2012). Furthermore, the irrigation
48 water demand of the major crops adopted in the study area was estimated using the CROPWAT model (FAO 1992).

49 2 **Materials and Methods**

50 2.1 **Description of the study area**

51 Jewuha watershed is located in the Awash river basin. It is geographically located between 39°44'55" to 40°10'4"E
52 and 10°00'3" to 10°21'10"N. The watershed is 240km away from the capital city Addis Ababa in the northeast
53 direction.

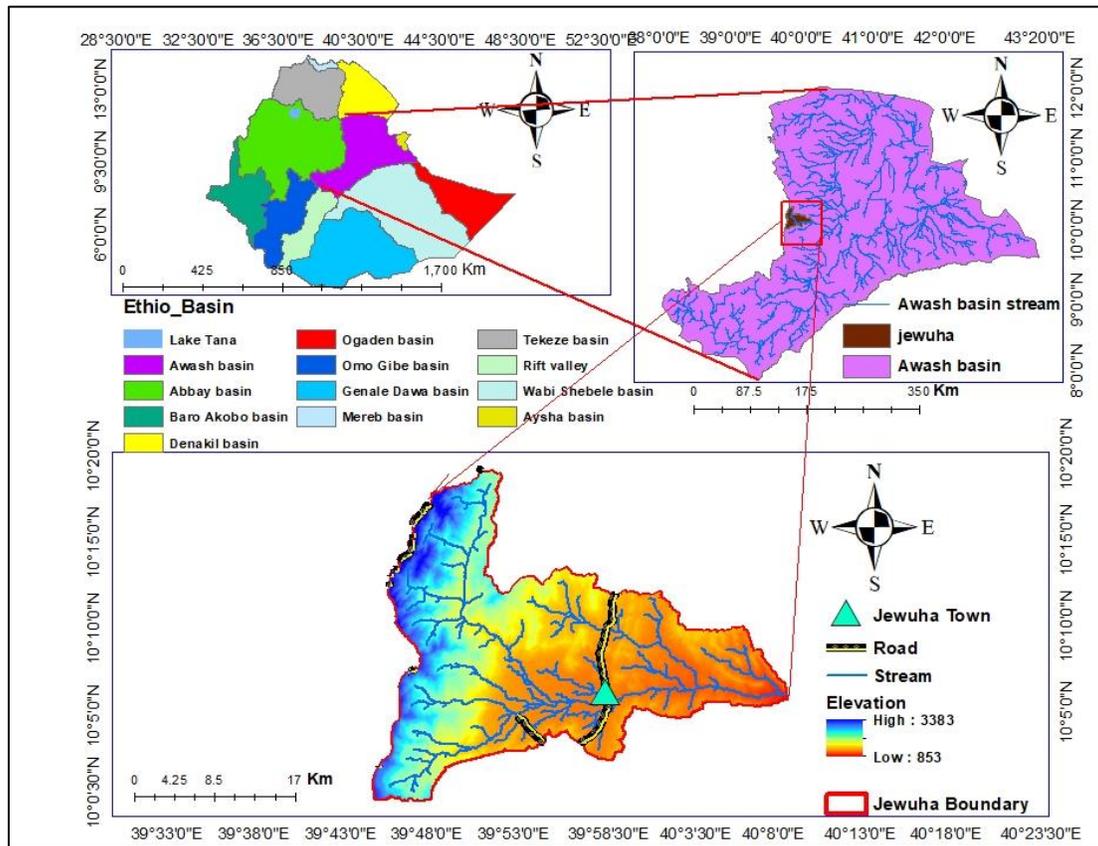


Fig.2.1 Location of the study area

54

55 The topography of the Jewuha watershed is characterized by diverse topographic conditions. The upper part of the
 56 watershed is characterized by mountainous and highly separated terrain with steep slopes and the downstream part is
 57 a gentle slope that is suitable for agricultural activity, with elevation ranges from 3555m in the mountainous area to
 58 1101m in the lowland. The general topography of the catchments is undulating hills and flats.

59 The climate of the Jewuha watershed is characterized by high rainfall with low temperatures in the highland and
 60 low rainfall with high temperatures in the low land area. According to the Ethiopian agro-climatic zone
 61 classification, the climate of the study area ranges from hot temperate (Kola) around the low land of the Jewuha
 62 town area to cool temperature (wurch) in the mountains and escarpment.

63 Rainfall distribution over the area is bimodal, characterized by a short rainy season (Belg) that occurs between
 64 March and April and a long rainy season (kiremt) that occurs between June and August, with a dry season from
 65 December to February. In the dry season, the rainfall ranges between 218 and 259mm, whereas in the wet season it
 66 ranges between 600mm to 1500mm.

67 The temperature in the study area is hot in the low land areas, reaching up to 34.4°C and the lowest temperature is
 68 recorded at Molale station on the highland, reaching up to 7°C and sometimes it is less than this value.

69 The land use land cover map gives the spatial distribution and classification of the various land use land cover
 70 classes. Different land cover types have been found in the study area in terms of areal coverage. The important land
 71 cover units are forest, shrubs, grassland, intensively cultivated, woodland, bare areas, and built-up areas.

72 The types of soil found in the study area are Eutric Fluvisols, Vertic Cambisols, Leptosols, Eutric Cambisols,
73 Chromic Cambisols, Chromic Vertisols, Haplic Xerosols, Eutric Regosols and Cambic Arenosols.

74 **2.2 Data Types and Source**

75 To achieve the objectives of the study, different data inputs were collected from various sources and field
76 observations.

77 **2.2.1 Metrological data**

78 Metrological data were collected from the National Metrological Agency of Ethiopia (NMAE). Shewa Robit and
79 Majeta climate stations were used for WGEN statistics calculation for SWAT using SWAT Weather database. The
80 SWAT Weather Database (Essenfelder 2018) is designed to be a friendly tool to store and process daily weather
81 data to be used with SWAT projects. It is capable of storing relevant daily weather information, easily generating.txt
82 files that are used for input during SWAT project and efficiently calculating the WGEN statistics from one or more
83 gauge stations. The daily weather data such as relative humidity (HMD), precipitation (PCP), solar radiation (SLR),
84 maximum and minimum temperature (TMP), wind speed (WND) and the batch file containing the location of the
85 gauge stations are loaded to SWAT Weather Database to calculate the WGEN statistics. The continuity and
86 consistency of the meteorological data were checked by the normal ratio method and double mass curve,
87 respectively.

88 **2.2.2 Hydrological data**

89 The daily stream flow data of (1985-2003) at Jewuha, 1988-2001 at Ataye, and 1995 – 2011 at Robi gauging station
90 were collected from the Ministry of Water, Irrigation and Energy (MoWIE). Ataye and Robi river were used for
91 validation of the regionalization technique in model parameter transfer for the ungauged sub watershed.

92 **2.2.3 Spatial data**

93 The data which is considered as spatial data includes DEM, land use land cover, soil map, and road map. Digital
94 elevation model (DEM) was obtained from the USGS website. The soil chemical parameters of CaCO₃ and EC
95 covering the study area were obtained from the Harmonized World Soil Database (HWSD) on the FAO website, in
96 Environmental System Research Institute (ESRI) shape file format and Micro Soft Access database (Nachtergaele et
97 al. 2009) and soil physical properties were obtained from the map of the Awash river master plan soil survey study
98 from the GIS and remote sensing department in the Awash basin authority, Adama branch.

99 Land use land covers were obtained from the Water and Land Resource Center (WLRC), Addis Ababa; Ethiopia.
100 The road map is another spatial data set that is used to extract road proximity to assess the irrigation potential of the
101 study area and it is obtained from the DIVA-GIS website.

102 **2.3 Land suitability**

103 Land suitability assessment means evaluating the parcel of land for irrigation or agricultural development (FAO
104 1976). Soil physical and chemical properties and topography, a characteristic of the land includes soil depth, texture,
105 drainage, slope, CaCO₃, EC and slope were taken to assess the suitability of the land in the study area. The digital
106 soil map was prepared and obtained from different organizations, then analyzed in a GIS environment. The soil

107 characteristics that were considered for land suitability assessment were taken from different works of literature
 108 (M.Albaji et al. 2008; Sys et al. 1991; Worqlul et al. 2015; Kassaye et al. 2019; Mandal et al. 2018).

109 Among the different land suitability assessment techniques, this study was interested in parametric evaluation
 110 systems prepared by (Sys et al. 1991) based on the soil characteristics. Parametric procedures usually allocate
 111 numerical ratings to separate land characteristics or land qualities depending on their relevance to the land use
 112 considerations. Then, they are combined into one numerical result using a mathematical equation as shown. These
 113 soil characteristics are rated and used to calculate the capability index (CI). The rating table was obtained from the
 114 tables prepared by (Sys et al. 1991).

$$115 \quad CI = A \times \frac{B}{100} \times \frac{C}{100} \times \frac{D}{100} \times \frac{E}{100} \times \frac{F}{100} \quad (1)$$

116 Where CI is the Capability index for irrigation, A= soil texture rating, B = soil depth rating, C = CaCO₃ rating, D =
 117 EC rating, E = drainage rating, F = slope rating

118 Table 1 Capability indices (CI) class for land suitability

Capability index	Definition	Symbol
>80	Highly suitable	S1
60 - 80	Moderately suitable	S2
45 - 59	Marginally suitable	S3
30 - 44	Currently not suitable	N1
<29	Permanently not suitable	N2

119 2.4 SWAT Model Description

120 The water availability in the study area was assessed using the SWAT hydrological model. SWAT (Arnold et al.
 121 2002) can simulate hydrological cycles, vegetation growth, and nutrient cycling with a daily time step by
 122 disaggregating a river basin into sub-basins and hydrologic response units (HRUs). SWAT uses the following water
 123 balance equation to simulate the hydrologic cycle within a watershed.

$$124 \quad SW_t = SW_o + \sum_{i=1}^n (R_{day} - Q_{surf} - E_a - W_{seep} - Q_{gw}) \quad (2)$$

125 Where; SW_t is the final water content (mm H₂O), SW_o is the initial soil water content on the day i (mm H₂O), t is
 126 time, days, R_{day} is the amount of precipitation on the day i (mm H₂O), Q_{surf} is the amount of surface runoff on the
 127 day i (mm H₂O), E_a is the actual evapotranspiration on the day i (mm H₂O), W_{seep} is the amount of water entering
 128 the vadose (unsaturated) zone from the soil profile on the day i (mm H₂O), Q_{gw} is the amount of return flow on the
 129 day i (mm H₂O).

130 The model reflects the difference in evapotranspiration for various land use and soil type in the subdivision of
 131 watersheds. The runoff was predicted separated from each HRU and routed to obtain the total yield for the
 132 watershed. Hence, increase the accuracy and gives a better physical description of water balance.

133 2.4.1 Sensitivity analysis

134 Sensitivity analysis is the way of determining the rate of change in model results concerning changes in model
135 parameters (Abbaspour 2012). Sensitivity analysis is important to provide information on the most important
136 parameters that affect the process in the study area and helps to decrease the number of parameters in the calibration
137 procedure by eliminating the parameters identified as not sensitive (Abbaspour et al. 2017). The sensitivity of the
138 parameter was selected based on the P-test and t-test values. The value that has a high P and a low t value is the most
139 sensitive parameter. The SUFI-2 (Sequential Uncertainty Fitting 2) program which is linked to SWAT-CUP
140 (Abbaspour 2012) was used for a combined model sensitivity and uncertainty analysis, calibration, and validation
141 procedures. From the two sensitivity analysis techniques in SWAT-CUP; global sensitivity analysis technique was
142 used.

143 2.4.2 Model Calibration and Validation

144 Calibration is an effort to better parameterize a model to a given set of local conditions, thereby reducing the
145 prediction uncertainty (Abbaspour et al. 2017). The prediction of uncertainty of SWAT model calibration and
146 validation results was analyzed by the SWAT calibration uncertainties program known as SWAT-CUP (Abbaspour
147 2012). It is a public domain program that links Sequential Uncertainty Fitting (SUFI-2) (Abbaspour et al. 2007),
148 Particle Swarm Optimization (PSO) and Parameter Solution (ParaSol) (Beven and Binley 1992) and Marko Chain
149 Monte Carlo (MCMC). For this study, SUFI-2 was used for calibration of the model. To show the intimate
150 relationship between the simulation result, expressed as 95PPU, and the observation expressed as a single signal
151 (with some error associated with it), two statistics values are used (Abbaspour et al. 2007). These are the p-factor
152 and r-factor, which give a good measure of the strength of calibration results. The P-factor is the percentage of
153 measured data bracketed by the 95PPU band and the r-factor is a measure of the thickness of the 95PPU. The value
154 of the p-factor and R-factor is between 0 and 1, and 0 to infinity, respectively. A p-factor of 1 and R-factor of 0
155 indicate simulations are exactly corresponding to the observed data (Abbaspour 2012).

$$156 \quad r \text{ factor} = \frac{\frac{1}{n} \sum_{t=1}^n \left(Q_t^{s,97.5\%} - Q_t^{s,2.5\%} \right)}{\sigma_{obs}} \quad (3)$$

157 Where $Q_{ti}^{s,97.5\%}$ and $Q_{ti}^{s,2.5\%}$ are the upper and lower boundary of the 95PPU at time t and simulation i,
158 respectively, n_j is the number of data points and σ_{obsj} is the standard deviation of the j th observed variable

159 Model validation is the process of describing that a given site-specific model is capable of making satisfactory
160 simulations. It is the comparison of model results with an independent data set without further adjustment of the
161 model parameters. Validation embraces running a model using parameters that were estimated during the calibration
162 process and comparing the predictions to observed data not used in the calibration process. The hydrological data of
163 Jewuha River from 1990 – 1997 and 1998 – 2003 was used for the calibration and validation of the SWAT model,
164 respectively.

165 2.4.3 Model performance evaluation

166 Herein, the performance of the model was checked by statistical tests that can be used to judge the SWAT model.
 167 For this study, Nash Sutcliffe efficiency (NSE), percent bias (PBIAS), and coefficient of determination (R^2) were
 168 used as recommended by (Moriassi et al. 2007). The coefficient of determination (R^2) describes the proportion of the
 169 variance in the measured data explained by the model. R^2 ranges from 0 to 1, with higher values indicating less error
 170 variance, and typically, values greater than 0.5 are considered acceptable (Moriassi et al. 2007).

171 Percent bias (PBIAS) measures the average tendency of the simulated data to be higher or smaller than its observed
 172 values. The optimal value of PBIAS is 0.0, with lower magnitude values indicating an exact model simulation. The
 173 negative value of PBIAS indicates the models overestimate the simulated and the positive shows the model
 174 underestimates the simulated flow (Moriassi et al. 2007).

$$175 \quad NSE = 1 - \left[\frac{\sum_{i=1}^N (O_i - P_i)^2}{\sum_{i=1}^N (O_i - \bar{O}_i)^2} \right] \quad (4)$$

$$176 \quad PBIAS = \left[\frac{\sum_{i=1}^N (O_i - P_i) \times 100}{\sum_{i=1}^N O_i} \right] \quad (5)$$

$$177 \quad R^2 = \left[\frac{\sum_{i=1}^N (O_i - \bar{O})(P_i - \bar{P})}{\sqrt{\sum_{i=1}^N (O_i - \bar{O})^2} \sqrt{\sum_{i=1}^N (P_i - \bar{P})^2}} \right]^2 \quad (6)$$

178 Where P_i = simulated flow, O_i = observed flow, \bar{O}_i = the mean of observed data, \bar{P} is predicted flow and the
 179 remaining variable is stated above and N is the total number of observations.

180 The Nash–Sutcliffe efficiency(NSE) (Sutcliffe 2001) indicating how well the model expresses the variance in the
 181 observation. It generally ranges from $-\infty$ to 1 the optimum value is unity and it shows a good explanation of the
 182 observed versus simulated data fits on a one to one line.

183 Table 2 General performance rating; Source:(Moriassi et al. 2007; Arnold et al. 2012)

Performance rating	NS	PBIAS	R^2
Very good	$0.75 < NS < 1$	$PBIAS < \pm 10\%$	$0.75 < R^2 < 1$
Good	$0.65 < NS < 0.75$	$\pm 10\% < PBIAS < \pm 15\%$	$0.65 < R^2 < 0.75$

Satisfactory	$0.5 < NS < 0.65$	$\pm 15\% < PBIAS < \pm 25\%$	$0.5 < R^2 < 0.65$
Unsatisfactory	$NS < 0.5$	$PBIAS > \pm 25\%$	$R^2 < 0.5$

184 2.4.4 Regionalization technique for ungauged sub watershed

185 Regionalization is the process of transferring hydrological information (parameters) of a model from a gauged
186 watershed to an ungauged watershed.

187 Among the different regionalization techniques, spatial proximity and physical similarity methods are widely used
188 (Oudin et al. 2008; Parajka et al. 2005). In this study, spatial proximity with Inverse Distance Weighting (IDW) was
189 used to transfer the calibrated model parameter of the gauged watershed to the ungauged watershed. And IDW was
190 used to estimate the weight of the ungauged watershed. The distance between the two watersheds was determined
191 using GIS. This regionalization techniques were verified using leave-one-out cross validation, in which a single
192 gauged site is considered as ungauged and the transferred parameters to that site are entered into the SWAT-CUP to
193 validate with the observed flow. Observed flow at Jewuha and the two neighbouring watersheds of Shewa Robit and
194 Ataye gauged watersheds was used to estimate the flow in the ungauged sub-watershed of the Jewuha River. The
195 calibrated parameters of the Jewuha and Robi watersheds were transferred to the other ungauged sub watersheds.
196 The parameters transferred by this technique are added into the SWAT model using a manual calibration helper and
197 then the SWAT model is run to obtain the flow for each sub watershed. The general formula for spatial proximity
198 with the IDW method to regionalize the calibrated parameter of the gauged watershed is as follows.

$$199 \quad Z_{ug} = \sum_{i=1}^n w_i z_i \quad (7)$$

200 Where Z_{ug} is the estimated model parameter at the ungauged watershed; n is the total number of observed points
201 (gauges); z_i is the calibrated parameter value at gauged watershed and w_i is the weight contributing to the
202 interpolation

$$203 \quad w_i = \frac{\frac{1}{d_i^2}}{\sum_{i=1}^n \frac{1}{d_i^2}} \quad (8)$$

204 Where d_i is the distance between at the centroids of gauged and ungauged watershed

205 2.5 Estimation of Irrigation Water Demand

206 Irrigation water demand is estimated from the water requirement of the crop. The major crops adopted in the study
207 area include maize, cabbage and onion. Several methods and procedures are available to compute the crop water
208 requirement. The computer program available in FAO Irrigation and Drainage Paper No. 56 “CROPWAT”, has
209 been used for the calculation of crop water requirements.

210
$$ET_c = K_c \times ET_o \quad (9)$$

211 Where K_c is crop coefficients and ET_c is crop evapotranspiration in mm

212 Irrigation water demand is derived from crop evapotranspiration (ET_c) and effective rain fall which is calculated
 213 based on USDA soil conservation service. Then the net irrigation water demand of the crop was calculated as
 214 follows.

215
$$NIWD = ET_c - P_{eff} \quad (10)$$

216 The gross irrigation water demand of the crop has been calculated by considering the loss of water during
 217 application of water to the irrigation field, loss in the canal through seepage and evaporation. Thus, so as to
 218 compensate for this loss, irrigation efficiency was introduced. The irrigation project efficiency is between 0.45 and
 219 0.7 for surface irrigation and 0.7 to 0.9 for pressurized irrigation systems (Luo et al. 2011). Thus, for this study,
 220 average irrigation efficiency was taken as 0.5, 0.75 and 0.85 for surface, sprinkler and
 221 pressurized irrigation systems, respectively.

222
$$GIWD = \frac{NIWD}{\eta} \quad (11)$$

223 **2.6 Irrigation Suitability Area Assessment**

224 The irrigation suitability of the study was assessed by weighing the factors of land suitability, land use land cover,
 225 distance from the source, and distance from the road (Nigussie et al. 2019; Rediet et al. 2020; Kasye Shitu 2020).

226 **2.6.1 Land use land cover suitability assessment**

227 The LULC of the study were reclassified based on the classification system of (FAO, 1976) using the
 228 reclassification tool, which is an attribute generalization technique in ArcGIS, as highly suitable (S1), moderately
 229 suitable (S2), slightly suitable (S3) and not suitable (N).

230 Table 3 Land use land cover suitability rating; Source:(Rediet et al. 2020; Gurara 2020)

LULC type	Definition	LULC rating (r)	Class
Cultivated land	Highly suitable	4	S1
Grassland/bare land	Moderately suitable	3	S2
Shrub/bush/wood land	Slightly suitable	2	S3
Settlement/forest/wetland	Not suitable	1	N

231 **2.6.2 Distance from the water source (river) suitability assessment**

232 The identification of irrigable land that is close to the water supply (rivers) was done by calculating the straight-line
 233 (Euclidean) distance from the streams that is generated from a 20m x 20m cell size DEM in a GIS tool and then
 234 reclassifying. The land which is nearest to the stream was considered the most suitable land for irrigation
 235 development and the land which is far from the stream is slightly suitable (Birhanu et al. 2019; Worqlul et al. 2015).

236 The land from the river was reclassified as highly suitable (S1), moderately suitable (S2), slightly suitable (S3) and
 237 not suitable (N).

238 Table 4 Distance from source suitability rating; Source:(Birhanu et al. 2019; Kassaye et al. 2019)

Class	Definition	Rating (r)	Distance from river (Km)
S1	Highly suitable	4	0 – 2
S2	Moderately suitable	3	2 – 4
S3	Slightly suitable	2	4 – 5
N	Not suitable	1	>5

239 **2.6.3 Distance from road suitability assessment**

240 The road maps obtained from the DIVA-GIS website were reclassified based on the classification system of (FAO
 241 1976) using the reclassification tool in the GIS. It is classified as highly suitable (S1), moderately suitable (S2),
 242 slightly suitable (S3) and not suitable (N) by ratings of 4, 3, 2, and 1, respectively.

243 Table 5 Distance from road suitability rating; Source: (Worqlul et al. 2015)

Class	Definition	Rating (r)	Distance from road (Km)
S1	Highly suitable	4	0 – 3
S2	Moderately suitable	3	3 – 5
S3	Slightly suitable	2	5 – 8
N	Not suitable	1	>8

244 **2.6.4 Weighted overlay of irrigation suitability factors**

245 To find an overall suitable site for irrigation, a suitability model was created using the model builder in the GIS Arc
 246 tools box to overlay the factor to map the suitable land. The weights developed above for each factor were overlaid
 247 in GIS to undertake multi-criteria evaluation (MCE) (Khongnawang and Williams 2015). In a multi-criteria
 248 evaluation, an attempt is made to combine a set of criteria to achieve a single composite basis for a decision
 249 according to a specific objective. The relative importance/weight of criteria and sub-criteria was estimated using
 250 multi-criteria evaluation through AHP, applied by using pairwise comparison of each suitability factor developed by
 251 (Saaty 1977).

252 In pairwise comparison, each factor was matched head-to-head (one to one) with the other and a pairwise or
 253 comparison matrix was prepared to express the relative importance. The diagonal elements of the pairwise
 254 comparison matrix are assigned the value of unity since the diagonal of the matrix value was obtained by the
 255 compared value of itself.

256 To fill the matrix, ratings were given for all factors on a 9 point continuous scale. For example, if one feels that land
 257 suitability is very strongly more important than LULC suitability in determining whether it is suitable for irrigation,
 258 one will enter 7 on this scale. However, if the reverse is true, one will give the value of 1/7. The value is given based
 259 on expert judgment and related literature reviews.

260 Table 6 Pairwise comparison matrix of factors

Factor	Soil	LULC	Distance from river	Distance from road
Soil	1	2	3	4
LULC	1/2	1	2	3
Distance from river	0.333	0.5	1	2
Distance from road	0.25	0.333	0.5	1
Sum	2.08	3.833	6.5	10

261 Calculate the normalized decision matrix of the irrigation suitability factor, which is obtained by all the elements in
 262 each column are divided by the sum of the columns as shown in the Table 9 below. The normalized value is
 263 calculated from the pairwise comparison matrices (A_{ij}).

$$264 \quad N = \frac{\sum j}{c} \quad (12)$$

265 Where N is normalized value, j is the column of the matrix and c is the values of the column of the factors

266 Table 7 Normalized value of factors

Factor	Soil	LULC	Distance from river	Distance from road
Land	0.48	0.52	0.46	0.4
LULC	0.24	0.26	0.31	0.3
Distance from river	0.16	0.13	0.15	0.2
Distance from road	0.12	0.09	0.08	0.1

267 The Eigenvectors and weights of the criteria were calculated from the normalized matrix through summation and
 268 average of the row values of each element in the row, respectively.

269 Table 8 Eigenvector value of criteria

Factor	Eigenvectors
Land	1.86
LULC	1.11
Distance from river	0.64
Distance from road	0.39

270

$$271 \quad W_i = \frac{\sum N}{x} \quad (13)$$

272 Where W_i the weights of the criteria, N is the row values of normalized matrix and x is the number of criteria for
 273 suitability analysis.

274 With a weighted linear combination, factors are combined by applying weight or percent of influence to the
 275 suitability of the irrigation obtained by the pairwise comparison technique. The multiplication was based on the
 276 following equation.

$$277 \quad S_i = \sum_{i=1}^n W_i X_i \quad (14)$$

278 Where S suitability, W_i is the weight of factor, X_i is criteria score of factor available water in the river in volume

279 The map obtained after overlaying was irrigation suitability, which are classified based on their degree of suitability
 280 as highly suitable (S1), moderately suitable (S2), slightly suitable (S3) and not suitable (N) (FAO 1976).

281 Table 9 Irrigation suitability rating

Suitability rating range	Class	Definition
4	S1	Highly suitable
3	S2	Moderately suitable
2	S3	Slightly suitable
1	N	Not suitable

282 2.7 Irrigation Potential Area Assessment

283 Irrigation potential (IP) is the land that is suitable for irrigation and that can be irrigated with the available surface
 284 water at a selected diversion site.

285 Diversion sites are selected by considering different factors. Those includes, diversion site elevation must be larger
 286 than the area being irrigated, it should be easily accessed by road, the site should be not be too far from the
 287 command area of the project (Dai, 2016) and the river cross section should be straight and narrow.

288 After the estimation of suitable land, irrigation water demand and available surface water, the actual irrigation
 289 potential of the study area was estimated. The actual irrigation potential at the selected diversion site was evaluated
 290 using the following formula for each perennial and some intermittent rivers in the watershed.

$$291 \quad IP(\text{ha}) = \frac{AW}{GIWD} \quad (15)$$

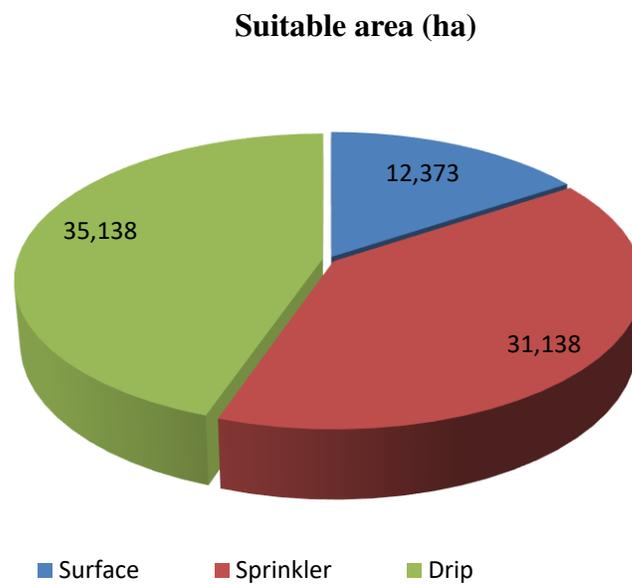
292 Where GIWD (m^3/ha) is gross irrigation water demand and AW (m^3) is available water in the river at selected
 293 diversion site in m^3/ha or mm (1mm = $10\text{m}^3/\text{ha}$)

294 The irrigation potential areas which are estimated for each diversion site might not be found along the river and its
295 delineation is difficult. In such case, manual delineation through trial and error was done by following the contour
296 line generated.

297 3 RESULT AND DISCUSSION

298 3.1 Land Suitability for Irrigation Potential

299 The result of land suitability analysis showed that about 12,373ha of land are ranged in between moderately to
300 marginally suitable for surface irrigation, 31,138ha of land are ranged in between highly to marginally suitable for
301 sprinkler irrigation systems and 35,433ha of land are considered highly to marginally suitable for drip irrigation
302 systems.



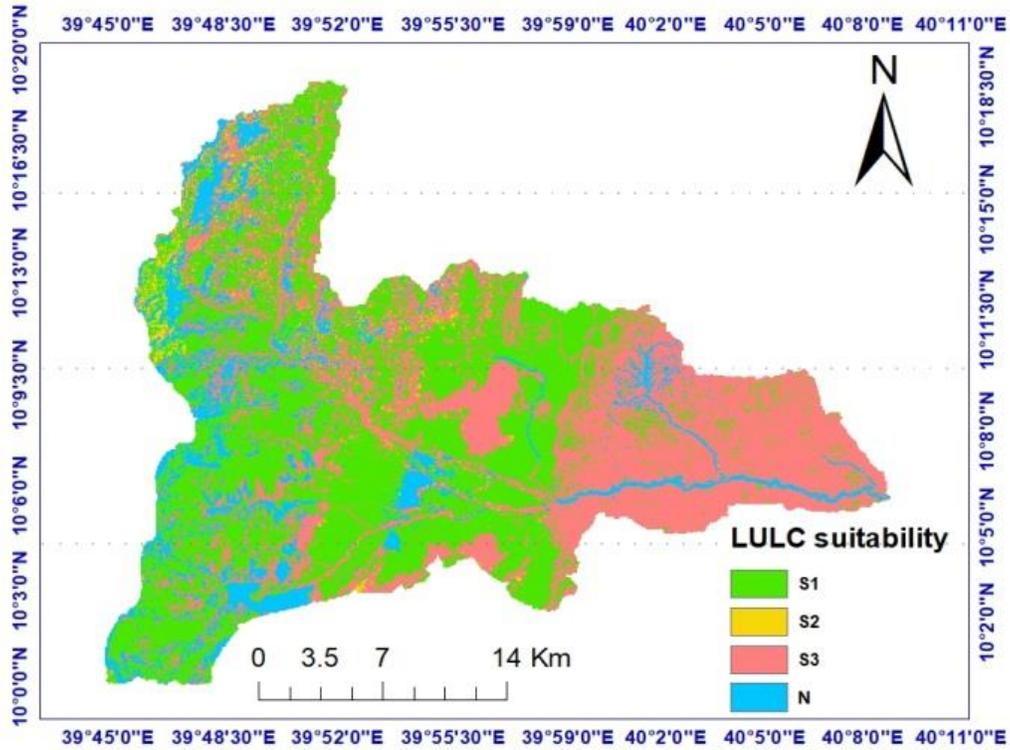
303

304

Fig.2 Land suitability chart

305 3.2 Land use land cover suitability

306 Based on the(FAO 1976) land suitability classification systems, the land use type was reclassified as 31,492ha
307 highly suitable (S1), 5756ha moderately suitable (S2), 30,599ha marginally/slightly suitable (S3), and 105ha not
308 suitable (N).



309

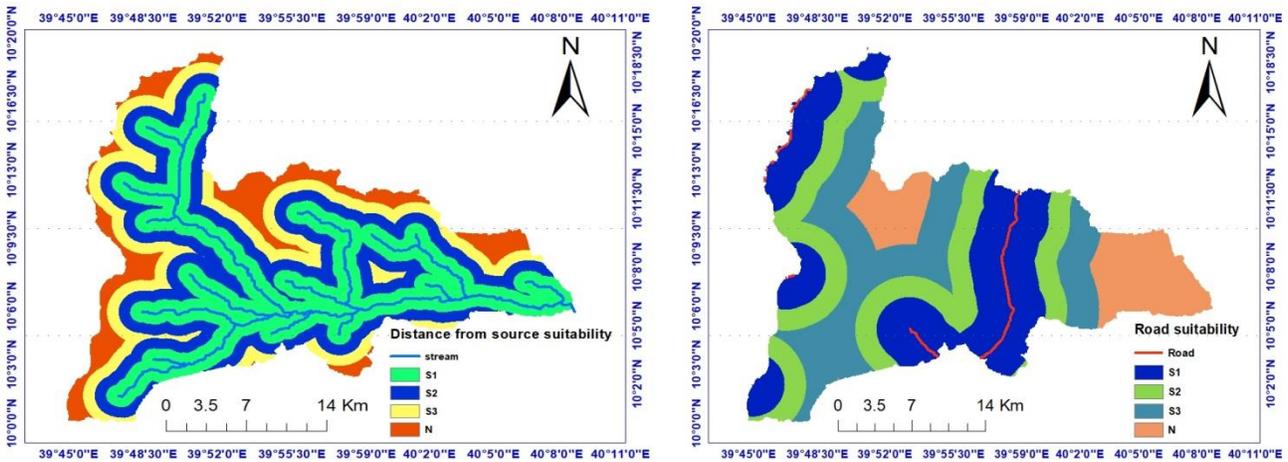
310

Fig.3 Land use land cover suitability

311 **3.3 Distance from River and Road Suitability**

312 The result of the analysis to estimate the suitability of the land from stream revealed that 27,991ha of land are highly
 313 suitable, 18,780ha are moderately suitable 12,412ha of land are slightly suitable and 8779ha are not suitable for
 314 irrigation development.

315 The result of road suitability shows that 21,620ha of land are highly suitable, 17,256ha are moderately suitable,
 316 18,257ha are slightly suitable and 10,830ha are not suitable for irrigation.



317

a. Distance from source (river)

b. Distance from road

318

Fig.4 Distance from source (river) and road suitability

319 **3.4 Surface water potential**

320 Surface water of the study area was assessed using SWAT hydrological model. The model in the study area can be
321 performed well with good performance.

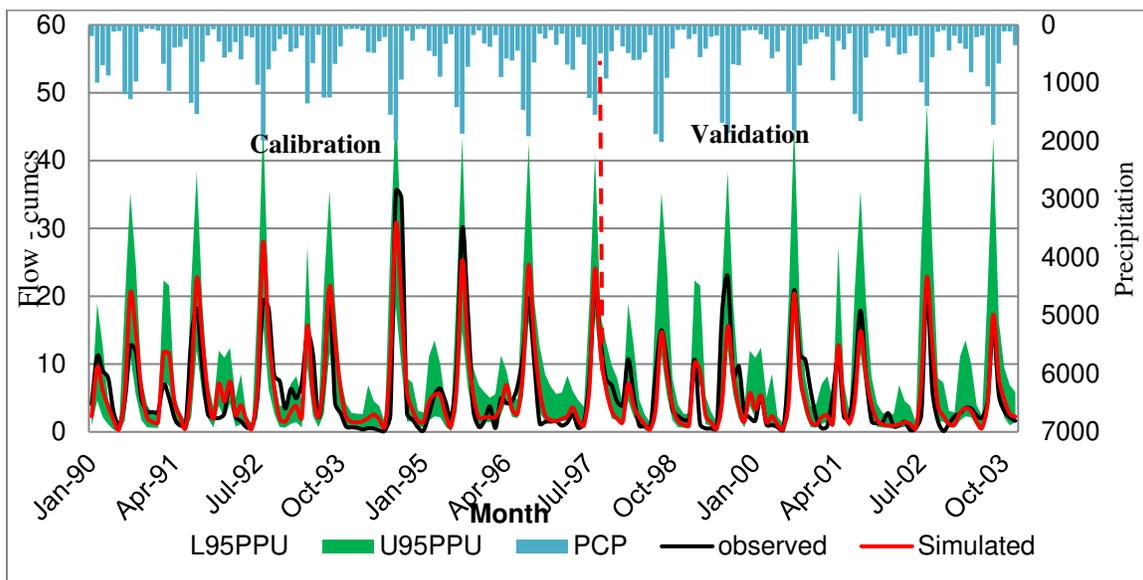
322 **3.4.1 SWAT model sensitivity Analysis**

323 Among twenty one model parameters that were selected for sensitivity analysis, sixteen parameters were found to be
324 sensitive under the category from high to low sensitive.

325 The most sensitive parameters of the study area in the SWAT model are R_CN2, ALPHA_BF, R_ESCO, R_EPCO,
326 V_GW_DELAY, V_ALPHA_BF, V_GWQMN, V_GW_REVAP, V_REVAPMN, V_RCHRG_DP, R_SOL_AWC,
327 R_SOL_K, R_SLSUBBSN, R_SURLAG, R_HRU_SLP, V_SHALLIST.gw and R_OV_N. Among these sensitive
328 model parameters, curve number (R_CN2), saturated hydraulic conductivity (R_SOL_K), Ground water delay
329 (days) (V_GW_DELAY), Manning’s “n” values for overland flow (R_OV_N) and available water capacity of the
330 soil layer (R_SOL_AWC) are the top five sensitive parameters with a p-value less than 0.5.

331 **3.4.2 Model calibration and validation**

332 Using the river discharge data obtained from the Minister of Water, Irrigation and Energy (MoWIE), the SWAT
333 model was calibrated at a monthly time scale.



334

335 Fig.5 Jewuha river calibration and validation

336 **3.4.3 Model performance evaluation**

337 The performance of the model was evaluated using time series plots of observed and simulated value and statistical
338 measures such as coefficient of determination (R^2), Nash-Sutcliffe efficiency (NSE) and percent of bias (PBIAS).
339 The statistical analysis of the watershed showed very good agreement between observed and simulated monthly
340 flow. For the Jewuha watershed, the model overestimates the flow. The model was most affected by small
341 traditional diversion structures in the Jewuha watershed, which do not have sufficient data to enter the value into the

342 SWAT model. This makes the model perform less. The p-factor is a good measure of the strength of calibration
 343 results. The P-factor is the percentage of measured data bracketed by the 95PPU band and its value is range between
 344 0 and 1. When its value ranges between 0.7 and 1 the percentage of uncertainty is very good. As shown in Table 10,
 345 the p-factor was 0.75 for the Jewuha River. This value shows that the 95PPU band is within acceptable ranges in the
 346 watershed.

347 Table 10 Model performance

Objective function	Calibration	Validation
R ² (Coefficient of determination)	0.74	0.71
NSE (Nash Sutcliff Efficiency)	0.73	0.7
PBIAS (Percent of bias)	-0.8	7.9
RSR	0.51	0.54
p-factor	0.75	0.74

348

349 3.4.4 Flow Regionalization

350 The flow from the gauged watershed to the ungauged watershed was estimated through parameter regionalization
 351 using the spatial proximity (SP) technique by Inverse Distance Weighting (IDW). For verification of the
 352 regionalization technique in the watershed, the statistical parameters of the objective function are good and this
 353 shows that applying the spatial proximity technique to transfer the model parameters to ungauged sub watershed to
 354 estimate was acceptable. Thus, the flow in the ungauged sub watershed was estimated using the spatial proximity
 355 (SP) regionalization technique.

356 Table 11 Mean monthly stream flow (m³/s) of ungauged sub watershed using spatial proximity regionalization

Sub watershed	Jan	Feb	Mar	Apr	May	Jun	Jul	Aug	Sep	Oct	Nov	Dec
Gida	0.5	0.97	0.87	0.83	0.8	0.4	1.06	2.62	1.83	1.13	0.8	0.7
Lomi	4.1	4.62	5.75	6.92	4.1	1.8	8.00	21.5	14.54	8.14	5.22	4.3
Gundifit	1.1	2.13	2.52	3.54	2.7	1.6	5.43	8.29	6.24	3.61	2.4	1.6
Ashmaq	0.1	0.42	0.50	0.74	0.5	0.3	1.24	1.77	1.09	0.46	0.3	0.2
Samet	0.4	0.78	0.90	1.30	0.9	0.6	2.35	2.99	1.77	0.90	0.7	0.5

357 3.5 Irrigation Water Demand

358 CROPWAT model results include crop evapotranspiration (ETc) and effective rainfall to estimate the irrigation
 359 water requirement of the crop. As shown in table 12, monthly gross irrigation water requirement of maize, cabbage
 360 and onion was estimated throughout their full growth periods for both surface, sprinkler and drip irrigation systems.

361 Table 12 Gross irrigation water demand (mm/month)

Irrigation Methods	Crop type	Jan	Feb	Mar	Apr	May
Surface irrigation	Onion	135.15	59.3	89.85	37	0
	Maize	69.36	50.66	120.57	144	119.94
	Cabbage	123.74	83.41	144.21	74.6	0
	Average	109.42	64.46	118.21	85.2	39.78
Sprinkler irrigation	Onion	90.1	39.53	59.9	24.7	0
	Maize	46.24	33.77	80.38	96	79.96
	Cabbage	82.5	55.6	96.1	49.7	0
	Average	72.95	42.97	78.79	56.8	26.65
Drip irrigation	Onion	79.5	34.88	52.85	21.8	0
	Maize	40.8	29.8	70.92	84.71	70.55
	Cabbage	72.8	49.1	84.8	43.9	0
	Average	64.37	37.93	69.52	50.14	23.52

362

363 3.6 Irrigation Suitability Potential

364 To find the irrigation suitability potential of the study area; land suitability maps, land use land cover suitability
 365 maps, distance from stream suitability maps and distance from road suitability maps were overlay in a GIS
 366 environment using a weighted overlay analysis tool in the spatial analysis tool. To overlay these maps, the factors
 367 are given a weight.

368 Table 13 Weight developed for factors

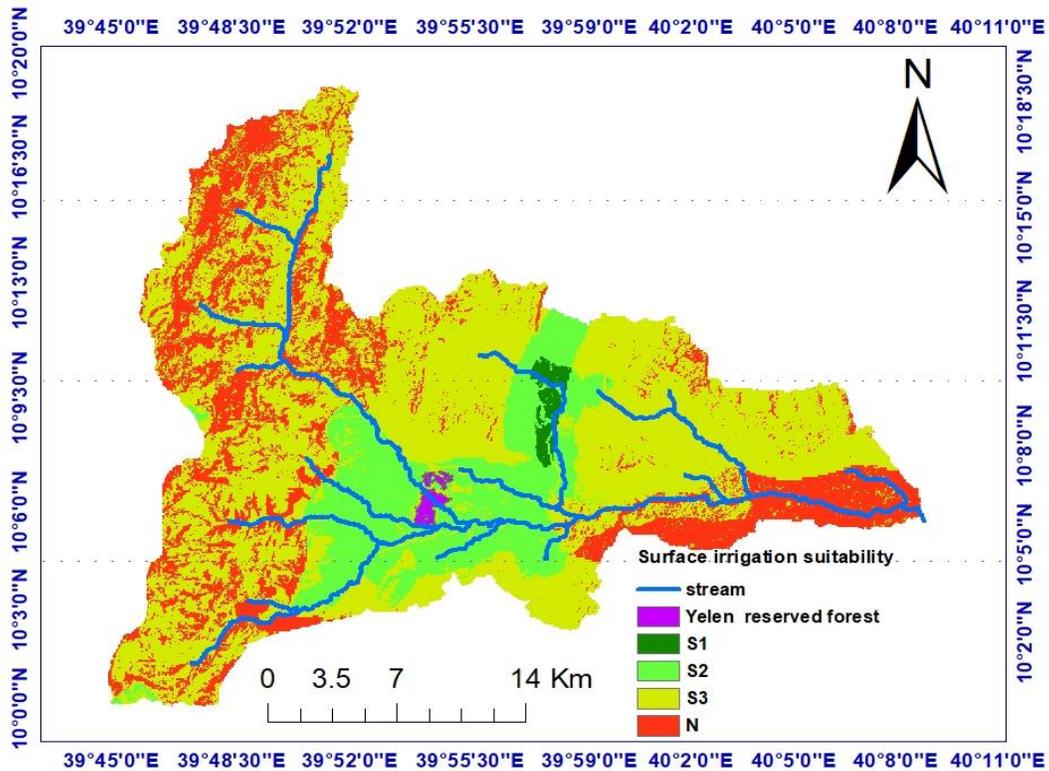
Factors	Wj (Weight (%))
Land suitability	47
Land use land cover suitability	27
Distance from source suitability	16
Distance from road suitability	10
Sum	100

369 The consistency ratio (CR) was 0.02, which is acceptable for weighting the factors to evaluate the irrigation
 370 suitability of the watershed.

371 3.6.1 Surface irrigation suitability

372 Assessment of the study area for surface irrigation suitability revealed that 1% (581ha) are highly suitable, 16%
 373 (10778ha) suitable, 59% (40013ha) moderately suitable and 24% (16274ha) are not suitable to develop a surface

374 irrigation system. Of the total land area of the study, 51372ha of land are suitable for surface irrigation. The analysis
 375 indicates that in the irrigation suitability map, a high portion of the cultivated area is found in the middle low land
 376 area and is deemed suitable for surface irrigation due to its high soil depth and flat slope of less than 8%, which is
 377 suitable for surface irrigation.



378
 379 Fig.6 Surface irrigation suitability

380 **3.6.2 Sprinkler irrigation suitability**

381 Assesment of the study area for sprinkler irrigation suitability revealed that 4.5% (3029ha) are highly suitable, 26.5%
 382 (18042ha) are suitable, 46% (31697ha) are moderately suitable and 23% (15088ha) are not suitable to develop a
 383 sprinkler irrigation system. From the total land area of the study, 52768ha of land can be developed by the
 384 sprinkler irrigation system.

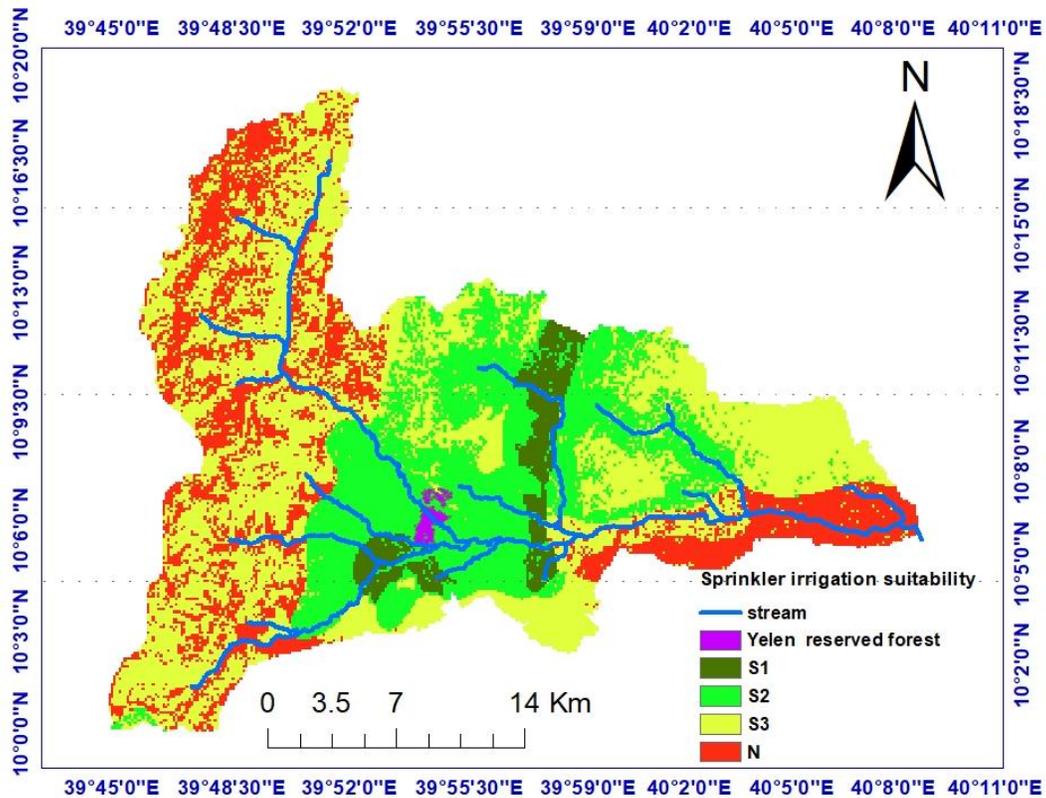


Fig.7 Sprinkler irrigation suitability

385

386

387 **3.6.3 Drip irrigation Suitability**

388 The assessment of drip irrigation suitability in the study area revealed that 13% (8498ha) are highly suitable, 25%
 389 (16971ha) are suitable, 40% (27282ha) are moderately suitable, and 22% (14855ha) are not suitable for drip
 390 irrigation system. From the total land area of the study, 52751ha of land can be developed by the drip irrigation
 391 system.

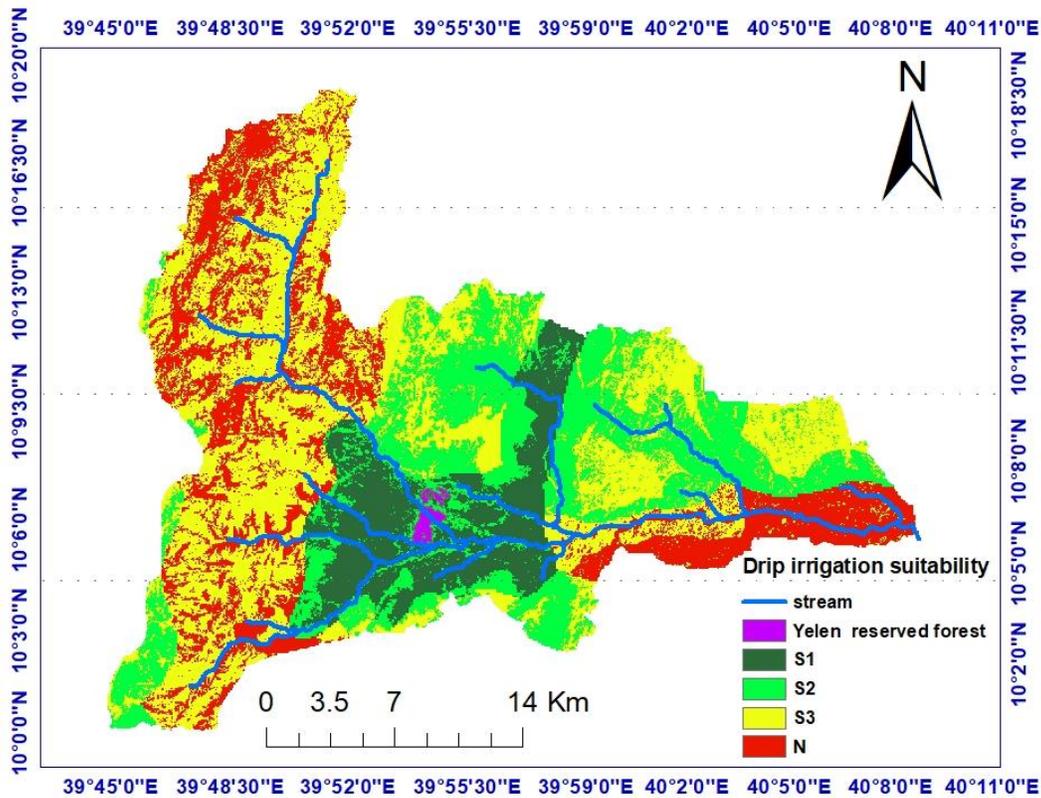


Fig.8 Drip irrigation suitability

3.6.4 Irrigation suitability of river catchments (sub watershed)

The irrigation suitability of river catchments or sub watersheds as shown in table 14 below shows that 315ha of the area in Gida sub watersheds was categorized as highly suitable for surface irrigation, but the remaining sub watersheds are not considered as highly suitable. For sprinkler irrigation suitability system 587ha, 603ha and 2ha of the area in Gida, Gundifit and Ashemaq are categorized under highly suitable for irrigation, respectively.

Table 14 Irrigation suitability of sub watershed

Sub watershed Name	Surface irrigation (ha)				Sprinkler irrigation (ha)				Drip irrigation (ha)			
	S1	S2	S3	N	S1	S2	S3	N	S1	S2	S3	N
Gida	315	678	4409	266	587	3107	1950	25	588	3083	1961	34
Lomi	-	302	10397	7036	-	785	10134	6836	163	729	9928	6909
Gundifit	-	2283	2226	641	603	2213	1721	616	2071	823	1615	641
Ashemaq	-	394	2253	1161	2	467	2203	1138	298	333	2165	1013
Samet	-	163	3769	2225	-	230	3919	2025	32	1255	3018	1848

For drip irrigation suitability systems, all the sub watershed are categorized as highly suitable, from the smallest area, 32ha in the Samet river sub watershed to the largest 2014ha in Lomi river sub watershed. This shows that pressurized irrigation systems are more suitable than surface irrigation systems in the Jewuha watershed. This was

403 due to the fact that, pressurized irrigation method is more suitable for steep slopes than the surface irrigation
 404 systems.

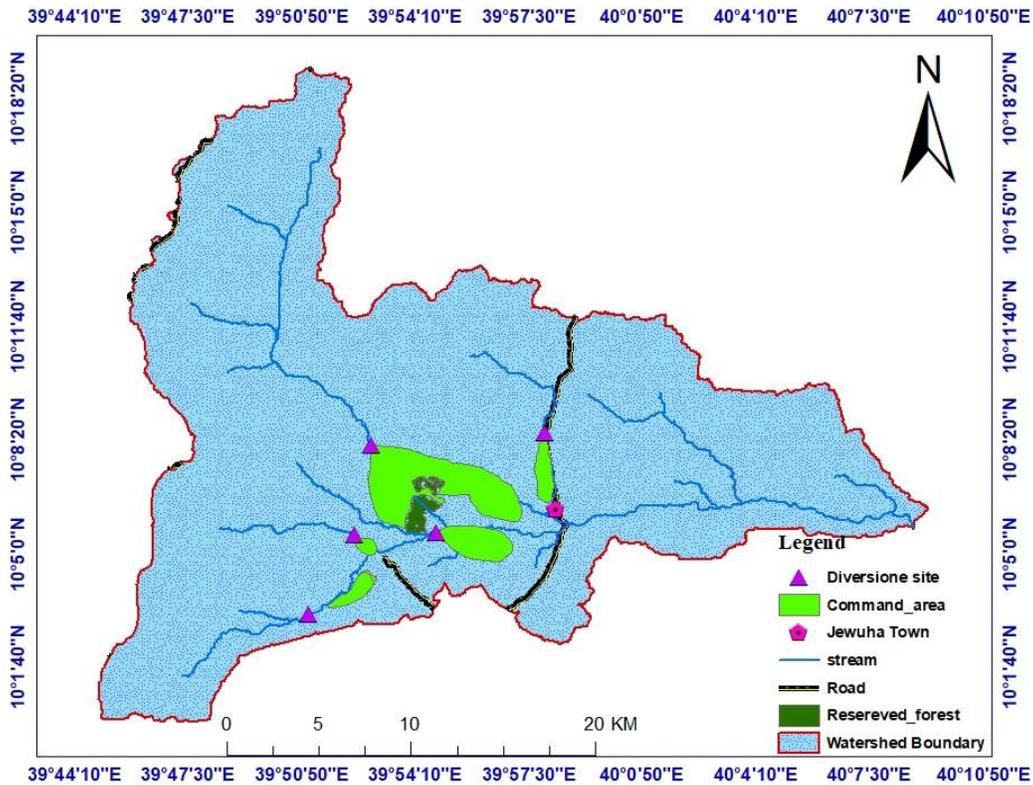
405 **3.7 Irrigation Potential Mapping**

406 From the analysis to estimate the irrigation potential from the available surface water and gross irrigation mater
 407 demand, the irrigation potential of the watershed was estimated as shown in the table 15 and figure 9.

408 Table 15 Irrigation potential and surface water available at selected diversion site and their location

Diversion site	Water potential (Mm ³)	GIWR (m ³ /ha)			Irrigation potential (ha)		
		Surface	Sprinkler	Drip	Surface	Sprinkler	Drip
Gida	1.06	4173	2780	2460	254	381	431
Lomi	8.4	4173	2780	2460	2014	3021	3415
Gundifit	2.34	4173	2780	2460	560	842	951
Ashmaq	0.34	4173	2780	2460	83	122	138
Samet	0.78	4173	2780	2460	187	281	317

409 Therefore, as shown in table 15 the actual irrigation potential of Jewuha watershed was found 3098ha for surface
 410 irrigation, 4647ha for sprinkler irrigation and 5252ha for drip irrigation systems.



411
 412 Fig.9 Irrigation potential maps and diversion site location

413 **Conclusion**

414 In this study, the land suitability was analyzed using the parametric evaluation technique by considering topography
415 and soil physical and chemical characteristics. The result showed that 12,373ha of land are suitable for surface
416 irrigation, 31,138ha of land are suitable for sprinkler irrigation systems and 35,433ha of land are suitable for drip
417 irrigation systems. Suitability assessment of LULC, distance from source and distance from road shows that 67,
418 847ha, 59,183ha and 57,133ha are suitable for irrigation, respectively. From the weighted overlay of suitable land,
419 land use land cover, distance from water resource and distance from road reveals that 1% (581ha) are highly
420 suitable, 16% (10,778ha) are moderately suitable, 59% (40,013ha) are slightly suitable and 24% (16,274ha) are not
421 suitable to surface irrigation system. And 4.5% (3,029ha) are highly suitable, 26.5% (18,042ha) moderately suitable,
422 46% (31,697ha) slightly suitable and 23% (15,088ha) are not suitable for sprinkler irrigation system. For drip
423 irrigation system 13% (84,98ha) are highly suitable, 25% (16,971ha) moderately suitable, 40% (27,282ha) slightly
424 suitable and 22% (14,855ha) are not suitable.

425 To estimate surface water potential, the SWAT model was employed. The model was calibrated and validated by the
426 observed flow. During the calibration and validation, the model performed good to simulate the hydrology of the
427 watershed with a coefficient of determination (R^2), Nash-Sutcliffe efficiency (NSE) and Percent of bias (PBIAS)
428 0.74, 0.73 and 0.8 for calibration and 0.71, 0.7 and 7.9 for validation in Jewuha watershed. Spatial proximity
429 regionalization technique was used to estimate the water potential for each ungauged sub watershed. The surface
430 water potential of Gida, Jewuha, Gundifit, Ashmaq and Samet sub watershed was 1.06Mm³, 8.4Mm³, 2.34Mm³,
431 0.34Mm³ and 0.78Mm³ respectively.

432 The gross irrigation water demand was estimated using three major crops grown in the area, viz.maize, cabbage and
433 onion. The average monthly demands were found to be 417mm, 278mm and 245mm for surface, sprinkler and drip
434 irrigation systems, respectively. Possible diversion site were selected based on the available command area to be
435 irrigated below the proposed diversion site, accessible to road, 80% available flow and narrow river width. As a
436 result, five (5) diversion sites were selected. The irrigation potential of the study area was estimated and mapped
437 based on the available water and gross irrigation water demand to the selected diversion site. This result reveals that
438 3098ha, 4647ha and 5252ha of land are potential for surface, sprinkler and drip irrigation systems, respectively.

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