

RUSLE Model Based Annual Soil Loss Quantification for Soil Erosion Protection in Fincha Catchment, Abay River Basin, Ethiopia.

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Research

Keywords: RUSLE, Quantification, Severity, Significant Factors, Soil Erosion, Soil Loss

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RUSLE model based Annual Soil Loss Quantification for soil erosion protection in Fincha Catchment, Abay River Basin, Ethiopia.

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Abstract

Background: The quantity of soil loss as a result of soil erosion is dramatically increasing in catchment where land resources management is very weak. The annual dramatic increment of the depletion of very important soil nutrients exposes the residents of this catchment to high expenses of money to use artificial fertilizers to increase the yield. This paper was conducted in Fincha Catchment where the soil is highly vulnerable to erosion, however, where such studies are not undertaken. This study uses Fincha catchment in Abay river basin as the study area to quantify the annual soil loss, where such studies are not undertaken, by implementing Revised Universal Soil Loss Equation (RUSLE) model developed in ArcGIS version 10.4.

Results: Digital Elevation Model (12.5 x 12.5), LANDSAT 8 of Operational Land Imager (OLI) and Thermal Infrared Sensor (TIRS), Annual Rainfall of 10 stations (2010-2019) and soil maps of the catchment were used as input parameters to generate the significant factors. Rainfall erosivity factor (R), soil erodibility factor (K), cover and management factor (C), slope length and steepness factor (LS) and support practice factor (P) were used as soil loss quantification significant factors. It was found that the quantified average annual soil loss ranges from 0.0 to 76.5 t ha⁻¹ yr⁻¹ was obtained in the catchment. The area coverage of soil erosion severity with 55%, 35% and 10% as low to moderate, high and very high respectively were identified.

Conclusion: Finally, it was concluded that having information about the spatial variability of soil loss severity map generated in the RUSLE model has a paramount role to alert land resources managers and all stakeholders in controlling the effects via the implementation of both structural and non-structural mitigations. The results of the RUSLE model can also be further considered along with the catchment for practical soil loss quantification that can help for protection practices.

Keywords: RUSLE, Quantification, Severity, Significant Factors, Soil Erosion, Soil Loss

31 **1 INTRODUCTION**

32 Soil erosion is one of the current challenging issues of agriculture causing soil degradation. The
33 severity of soil erosion is very serious in a country where land management is very weak[1]. The
34 decrease in agricultural productivity[2], ecosystem disturbances, and water resources pollutions
35 are some of the major ill impacts of soil erosion that are commonly happening in the world[3]. The
36 topographic conditions, land use land cover, the intensity of the rainfall, and the soil characteristics
37 are major significant factors of soil erosion[4]. The annual dramatic increment of the depletion of
38 very important soil nutrients (fertility) exposes the residents of this catchment to high expenses of
39 money[5] to use artificial fertilizers to increase the yield. The intrusion of runoff from eroded soil
40 into a water source[6] invites harmful pollutants and chemicals to change both the physical and
41 chemical properties of sources of water. Even if soil erosion is a globally happening natural
42 hazard[7] in the world, its effects are very serious in developing countries as a result of the inability
43 of restorative lost soil and valuable nutrients. In Africa, soil erosion is one of the top-ranked
44 problems affecting agricultural sectors[8] and water resources such as lakes and reservoirs and
45 Ethiopia is not exceptional in the problem. Throughout the world, researchers have been using
46 different models for the assessment and investigation of soil loss risk. EUROPEan Soil Erosion
47 Model (EUROPSEM), Limburg Soil Erosion Model (LISEM), Soil and Water Assessment Tool
48 (SWAT), and Water Erosion Prediction Project (WEAP) are commonly used soil loss model[9]s.

49 The application of integrated Geographical Information System (GIS) and remote sensing
50 technologies in areas of the earth's surfaces are getting global attention and are widely
51 used[10],[11]. The simple empirical function called Universal Soil Loss Equation (USLE) is the
52 most commonly used model for loss assessment and later changed into Revised Universal Soil
53 Loss Equation (RUSLE) was adopted to different catchments in Ethiopia[12]. This paper was
54 conducted in Fincha Catchment where the soil is highly vulnerable to erosion, however, where
55 such studies are not undertaken. Knowing the information about the spatial variation of soil erosion
56 severity is a very important tool for implementing a protective measure in land resource
57 management[8]. In the Fincha catchment, two man-made reservoirs and a natural lake reservoir
58 are there, and the majority of water sources in this catchment are serving as water supply sources.

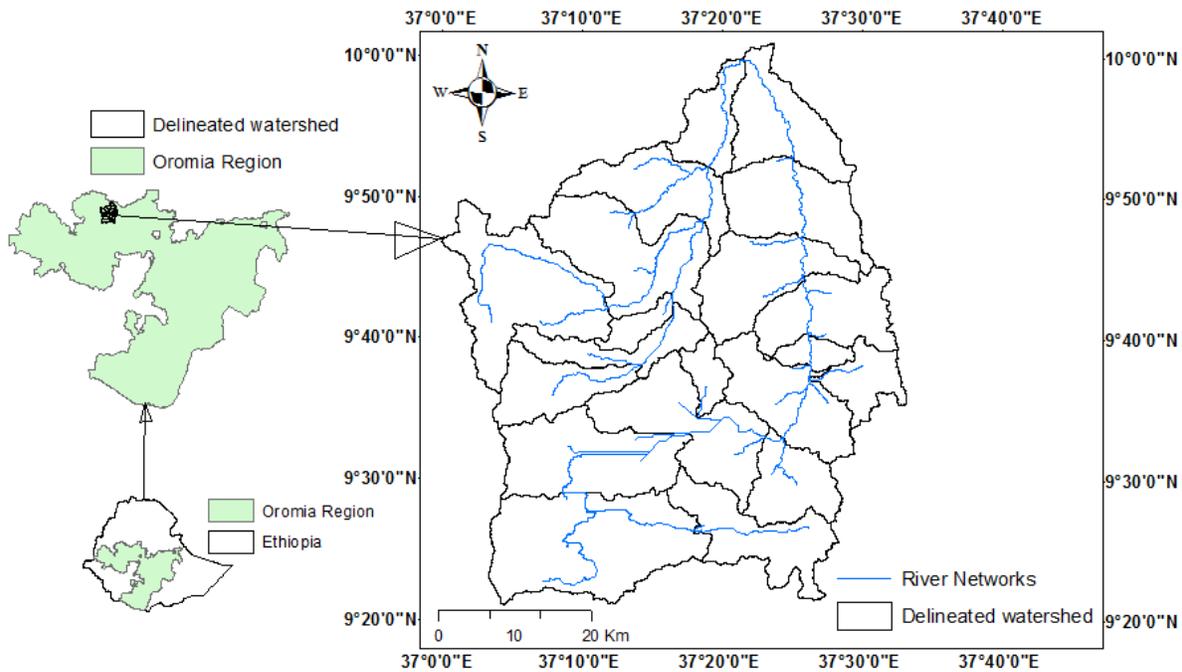
59 As a result of the topographic conditions, the intensity of rainfall, a traditional way of agricultural
60 practices, and the nature of the soil in this catchment; the majority of the agricultural lands are

61 prone to water derived erosion. Therefore, for this particular study area, an integrated GIS and
62 RUSLE model-based soil loss quantification model was used to develop a spatially varied soil
63 erosion severity map which is very important for sustainable land resources management
64 strategies.

65 2 MATERIAL AND METHODS

66 2.1 Study Area

67 This study area is conducted in the Fincha catchment (Fig.1) in Abay River basin, and particularly
68 the study under consideration was delineated fixing the outlet at Fincha River. This delineated
69 watershed is geographically found between 37°0.06'00" E to 37°33'18" E longitude and 09° 21'11"
70 N to 10° 01'00" N latitude. The delineated watershed has a total area of 21.48 km².



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72

Figure 1: Geographical Location of the study area

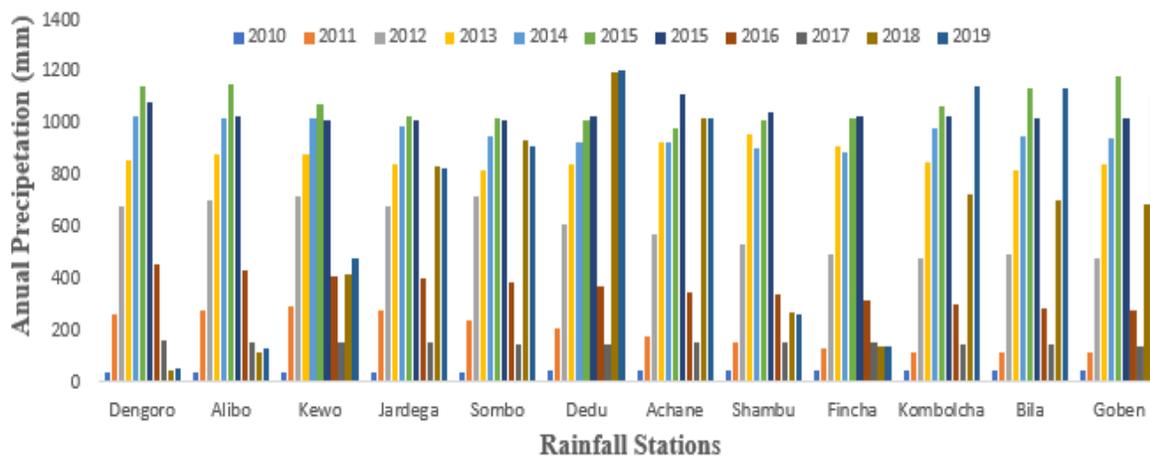
73 2.2 Data Collection

74 To generate soil erosion significant factors namely; R-factor, P-factor, K-factor, LS-factor, and C-
75 factor, annual average precipitation of 11 stations (Table 1 and Fig. 2) in Fincha catchment (*Alibo,*
76 *Kewo, Jardega, Sombo, Dedu, Achane, Shambu, Fincha, Kombolcha, Bila and Goben*), land use
77 land cover (LULC) derived from LANDSAT 8 OLI/TIRS, soil map obtained from ministry of

78 Ethiopian geological Survey, and Digital Elevation Model (DEM) of (12.5 x 12.5) spatial
 79 resolution downloaded from Alaska University online facilities were used as input.

80 Table 1: Rainfall stations in and surrounding Fincha subbasin

No.	Station Name	Long	Lat	Elevation (m)	Annual mean precipitation (mm)
1	Alibo	37.0795	9.8904	2405.53	1754.2
2	Kewo	37.5543	9.8271	2362.15	1353.65
3	Jardega	37.0143	9.8065	2402.98	2030.93
4	Sombo	37.0326	9.7424	2385.78	1988.75
5	Dedu	37.5357	9.6889	2270.19	1491.6
6	Achane	37.3216	9.6547	2398.33	1645.81
7	Shambu	37.0943	9.5707	2556.4	1827.07
8	Fincha-a	37.362	9.5659	2226.33	1528.11
9	Kombolcha	37.4781	9.5097	2390.25	1621.3
10	Bila	37.0279	9.2925	2114.34	1728.16
11	Goben	37.311	9.1741	2610.34	1906.2



81
 82 Figure 2: Mean annual precipitation of stations in and surrounding Fincha catchment

83 **2.3 Soil Loss Models**

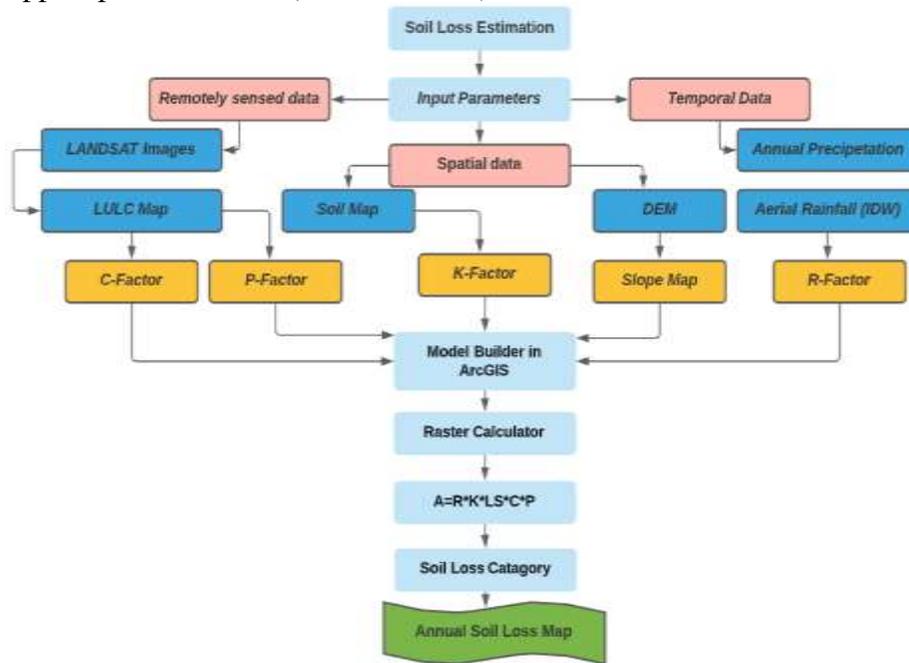
84 Researches have been widely undertaking for developing a reliable soil loss estimation. Different
 85 countries have been attempting to develop empirical equations and other physical-based models
 86 that are applicable at local and global levels[7]. EUROPEan Soil Erosion Model (EUROPSEM),
 87 Limburg Soil Erosion Model (LISEM), Soil and Water Assessment Tool (SWAT), and Water
 88 Erosion Prediction Project (WEAP) are commonly used soil loss models by adaptation
 89 techniques[13][9]. RUSLE model is the most popular soil loss estimation model and can be
 90 developed in ArcGIS. The map algebra in spatial tool analyst of ArcGIS can easily manipulate the
 91 combination of soil loss significant factors by developing a model builder[14].

92 **2.4 RUSLE model**

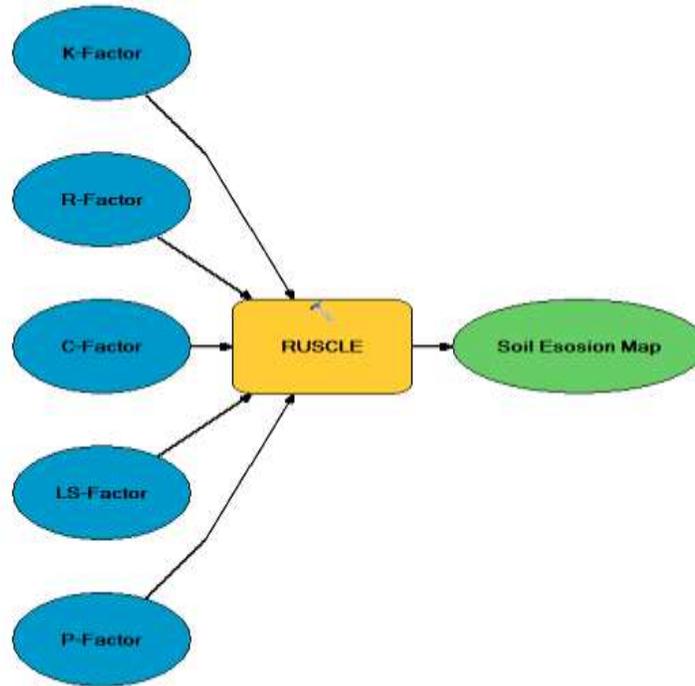
93 Of soil loss models, Revised Universal Soil Equation (RUSLE) is a popular and commonly used
 94 reliable model for annual soil loss quantification[15]. In this study, RUSLE was used to generate
 95 the spatially varied soil erosion severity map by combining five factors. The thematic maps of the
 96 raw input parameters of the significant factors (R-factor, P-factor, K-factor, LS-factor, and C-
 97 factor) were prepared with 12.5 x 12.5 m spatial resolution. A map algebra algorithm in spatial
 98 tool analyst and a model builder (Fig.4) developed in ArcGIS were implemented to combine the
 99 significant factors using the RUSLE equation as shown below. The general flowchart showing the
 100 detailed procedures and data needed is summarized in (Fig. 3).

101
$$A = R * K * LS * C * P \dots \dots \dots (1)$$

- 102 Where A The total annual soil loss (t/ha per year), t is the thickness of lost soil
 103 R Rainfall erosivity factor (MJmm ha⁻¹ h⁻¹ year⁻¹)
 104 K Soil erodibility factor (t haMJ⁻¹ mm⁻¹)
 105 LS Slope length and steepness factor (dimensionless)
 106 C Over and management factor (dimensionless)
 107 P Support practice factor (dimensionless)



108
 109 Figure 3: Detail flowchart of the steps and data needed in the study



110
 111 Figure 4: Model builder developed in ArcGIS for combination of the significant factors

112 **2.5 Rainfall erosivity factor (R)**

113 The rainfall erosivity factor (R) describes the relationship between the rainfall intensity and the
 114 soil responses to it[16][17]. In this catchment, the spatial variability of the intensity of the rainfall
 115 varies from 1353.65 mm to 2030.93mm. There is a positive relationship between the intensity of
 116 the precipitation and the soil characteristics in such a way that if the intensity is very high, there is
 117 the probability of severe soil erosion[8]. Ten years of historical mean annual precipitation of 11
 118 stations contributing to the catchment were considered in this study to generate information about
 119 erosivity in the area. An aerial raster geodatabase of rainfall was generated from the point rainfall
 120 historical data using Inverse Distance Weight (IDW) interpolation technique[18] with 12.5 x 12.5
 121 m spatial resolution using the regression equation developed by (Eq.2)

122
$$R = 1.24 * P^{1.36} \dots \dots \dots (2)$$

123 Where R Rainfall erosivity factor (MJmm ha⁻¹ h⁻¹ year⁻¹)
 124 P Annual mean precipitation (mm)

125 **2.5.1 Soil Erodibility factor (K)**

126 The properties of soil and the degree of erodibility are interconnected parameters. When the drop
 127 of rainfall hits the soil particles, there is a high probability of disintegration in soil particles when
 128 the soil hitting external force exceeds the cohesion forces between soil particles[19]. The ability

129 of soil particles in persisting against rainfall is different in different soil types and this property is
 130 expressed in terms of erodibility factor[20]. In this catchment there are more than 10 soil types are
 131 available and reclassified into 6 dominant types (Fig. 5 and Table 2) and k-values were assigned.

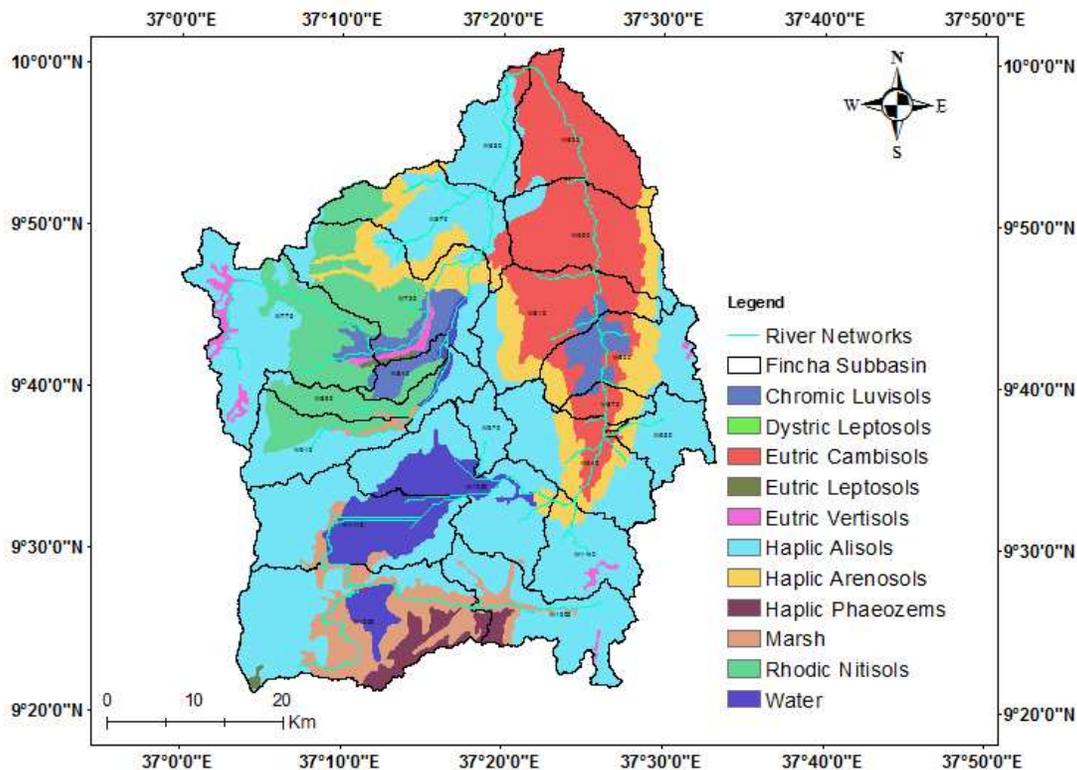


Figure 5: Soil types in the Catchment

Table 2: Soil types and corresponding erodibility factor (k-values) in the catchment

Soil Types	K-factor value	Area (Km ²)	Area (%)
Chromic Luvisols	0.27	3.71	17.27
Dystric Leptosols	0.68	0.74	3.45
Eutric Cambisols	0.35	4.57	21.28
Eutric Leptosols	0.34	0.54	2.51
Eutric Vertisols	0.24	2.42	11.27
Haplic Alisols	0.45	8.58	39.94
Haplic Arenosols	0.67	0.25	1.16
Haplic Phaeozems	0.21	0.14	0.65
Marsh	0.24	0.09	0.42
Rhodic Nitisols	0.34	0.44	2.05

135 2.5.2 Topographic Factor (LS)

136 The severity of the spatial variability of soil erosion highly relies on the topographic conditions of
 137 an area. The steepness or the flatness of an agricultural land governs the degree of the erodibility

138 of soil particles. The speed of the water flowing over soil and the slope of the topography are
 139 dependent parameters[19]. The length of the slope and slope steepness of the area in the study area
 140 was generated using Digital Elevation Model (DEM) of 12.5 x 12.5m spatial resolution and LS-
 141 factor was generated in ArcGIS version 10.4. Flow accumulation and slope (%) are commonly
 142 used input parameters with a fixed cell size with regression equation[17] (Eq.3).

$$143 \quad LS = \left\{ Flow\ accumulation * \left(\frac{Cell\ size}{23.13} \right) \right\}^{0.4} * \left\{ \frac{\sin(slope\ (\%)) * 0.01745}{0.09} \right\}^{1.3} * 1.6 \dots \dots (3)$$

144 Where LS Slope length and steepness factor (dimensionless)

145 2.5.3 Cover and Management factor (C)

146 The types of cover and land use in agricultural land is highly interrelated factors. Raw input DEM
 147 was corrected by applying fill and flow direction in ArcGIS using spatial tool analyst. Slope (%)
 148 and flow accumulation were generated for the study area and reclassified based on the c-values in
 149 spatial tool analyst. The types of land use land cover (LULC) in the study area (Fig.7) and the
 150 corresponding c-values (Table 3) were assigned. In the same fashion, the values for support
 151 practice factor (P) were generated from the land use land cover map.

152 Table 3: Cover and Management (C) and Support practice (P) factors in the study area

Land Use	C-Factor Value	P-factor value
Trees cover areas	0.25	0.27
Shrub cover areas	0.37	0.68
Grass Land	0.45	0.87
Crop Land	0.68	0.48
Swampy areas	0.75	0.91
Breland	0.65	0.57
Built-up areas	0.47	0.67
Open water	0.66	0.69

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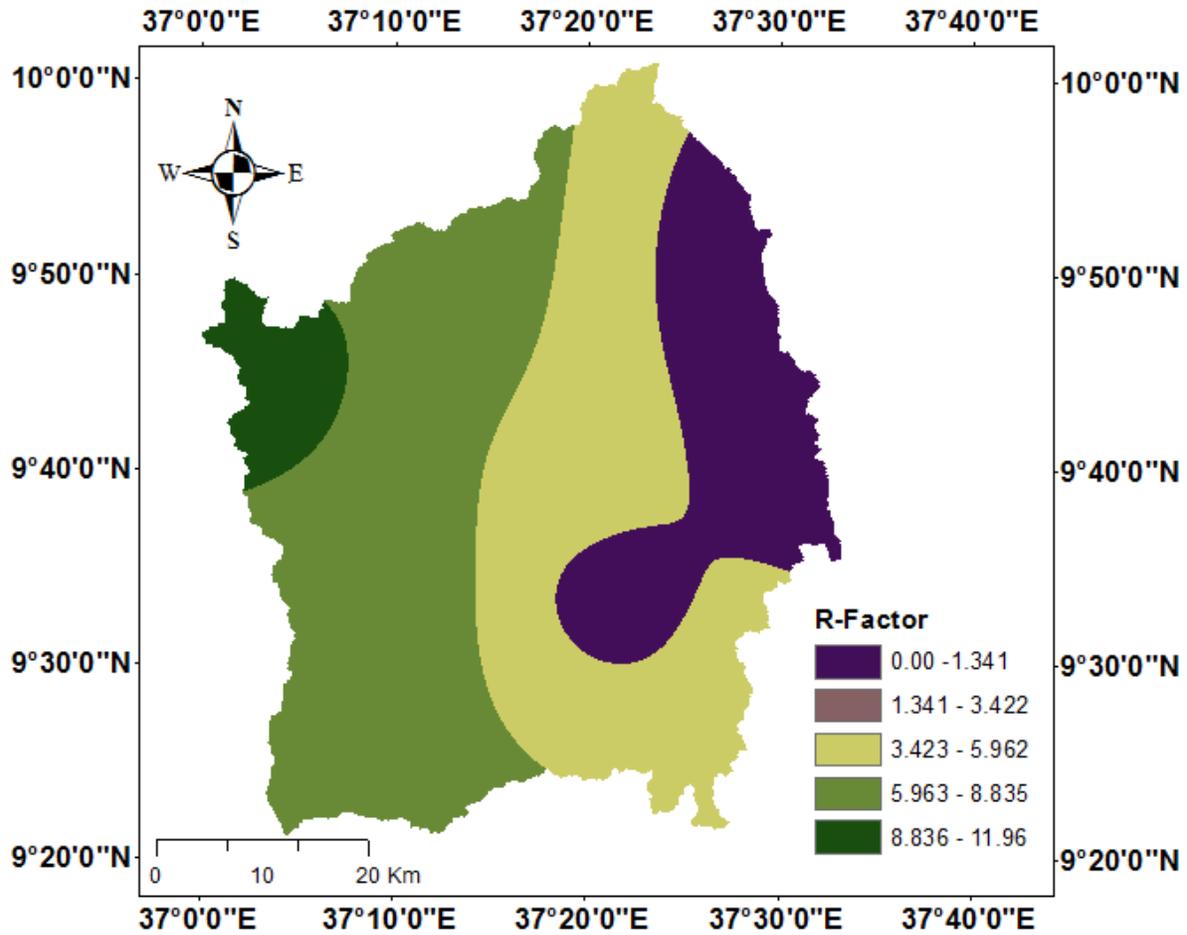
154 3 RESULT AND DISCUSSIONS

155 3.1 Soil Loss Quantification in Fincha catchment

156 The quantified annual soil loss values are generally ranging from 0 to 76.5 in thickness (t) soil per
 157 hectare per year. The total area of this particular catchment is 21.48 km² (2148 ha) and this
 158 indicates that 164, 322 t/ha per year quantity of soil is lost and a similar soil loss estimation was
 159 made in the study conducted by [12][6]. The qualitative based classifications namely; very high,

160 high, moderate, low, and very low for identification spatial variation of soil erosion severity
161 applied in [8][1][9] was also repeated in this study. The area covered in the percentage of soil
162 erosion severity of 10%, 45%, 30%, and 15% as low, moderate, high, and very high respectively
163 were identified (Fig. 5). As revealed in the severity map generated using the RUSLE model (Fig.
164 13), the agricultural lands which cover 65% (13.96 ha) of the total area are highly vulnerable to
165 erosion and the qualitative classification of the area is between high to very high to the soil loss
166 risk. The effect of the erosion is very visible in this catchment when compared to estimated soil
167 loss in other catchments [4]. In terms of the significant factors; rainfall erosivity (R-factor), cover
168 and management(C-factor), and support and conservation practice (P-factor) factors revealed high
169 significance while the other factors are relatively low significant for the initiation of soil erosion,
170 and the values of the corresponding factors were shown in Fig.6-8 respectively. The spatial
171 variability of mean annual rainfall shown in (Fig.6) showed that the majority of croplands and soil
172 types in the lower part of the catchment is very sensitive to soil loss and this fact is observed in the
173 severity map (Fig.13). The rate of soil loss seen in the catchment is higher than the total annual
174 soil formation rate ranges from 2 to 22 t/ha per year for the different land uses units of Ethiopia
175 [8] and special attention should be given to minimize the rate of soil loss in the catchment by
176 implementing soil formation strategies or soil and resources management strategies. Support and
177 conservation practices factor (P-factors) values were assigned based on the soil types in the study
178 area. The soil map was reclassified into six dominant soil types and the corresponding p-factor
179 values were given (Fig. 11).

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181

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Figure 6: Rainfall erosivity factor (R-factor) values in Fincha catchment

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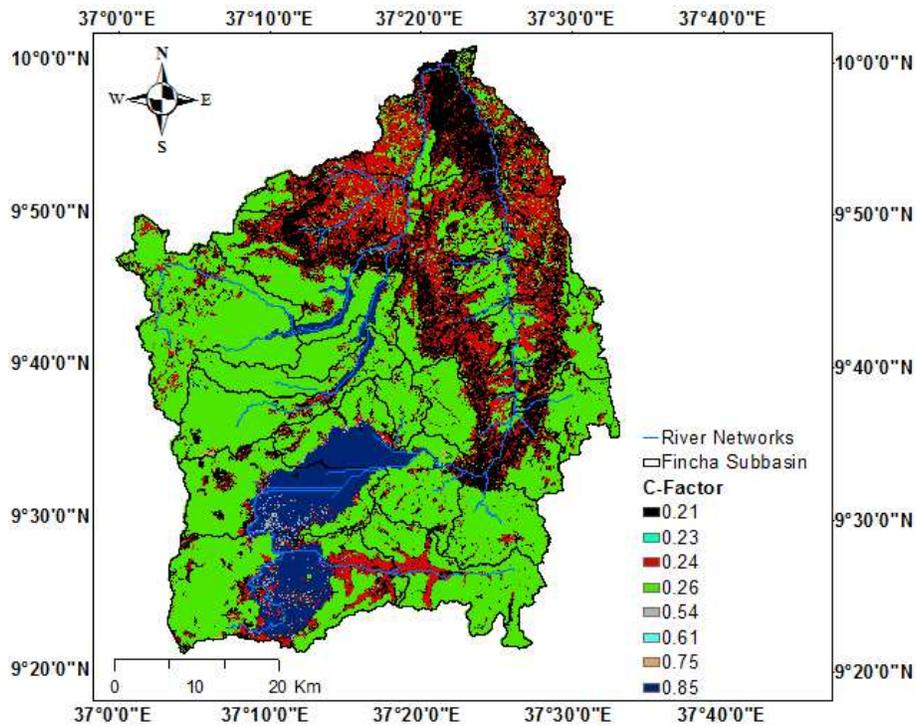
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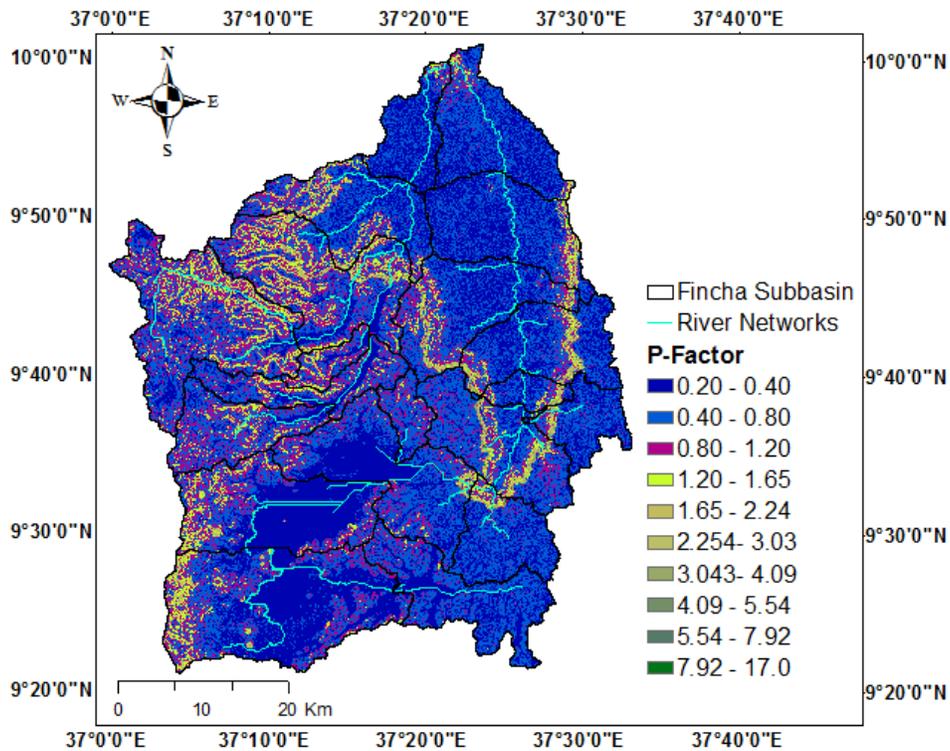
The low-land agricultural areas in this catchment are less vulnerable to soil erosion and relatively less soil loss is visible[16] due to the low velocity of runoff water. The traditional way of agricultural systems and soil conservation practices is very weak in the lower part of the study area, therefore, the support and conservation practices factor (p-factor) reveals that the croplands and bare land are very exposed and sensitive to erosion due to the incoming runoff water from the highland areas[16]; Kayet et al., 2018).



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Figure 7: Major land use land cover (LULC) types and corresponding C-values

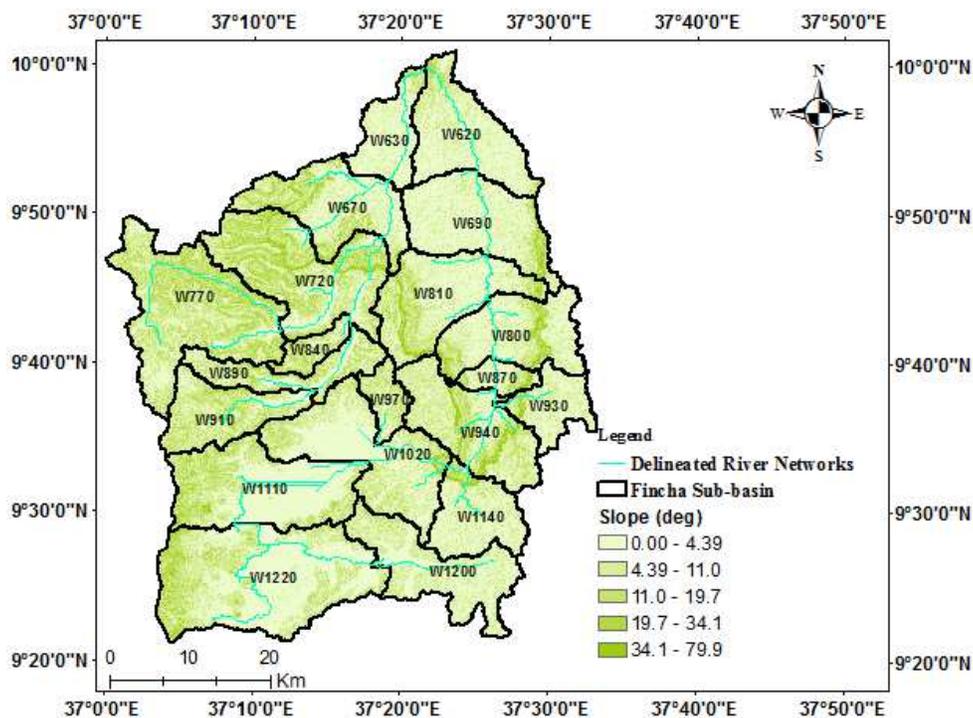


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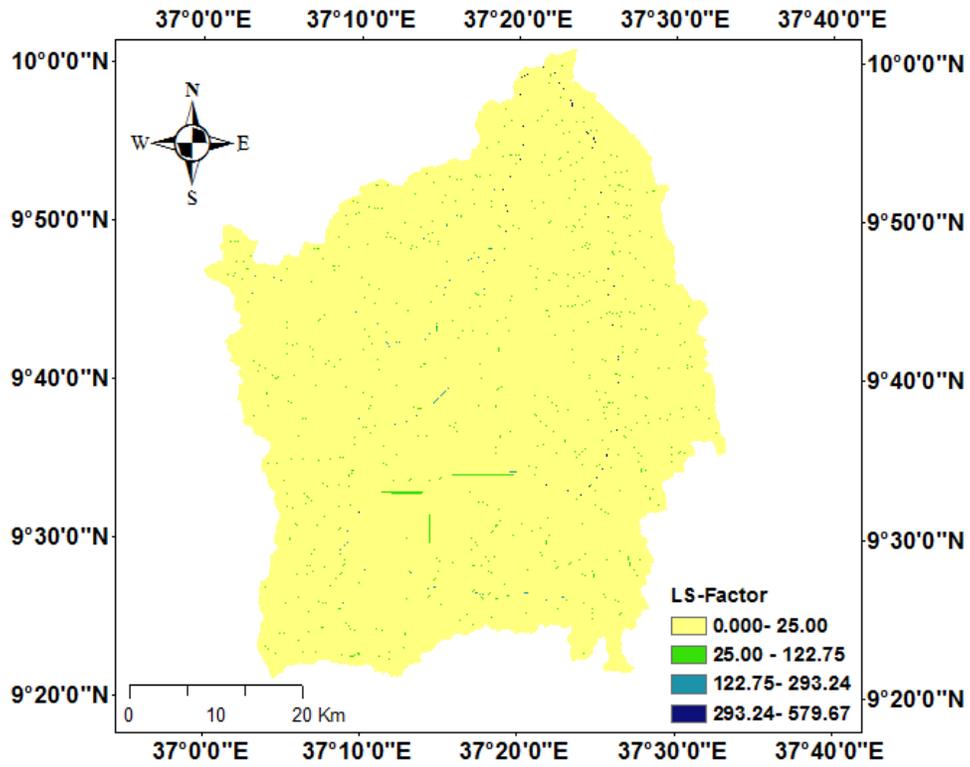
Figure 8: Support and practice factor (P-factor) values in Fincha catchment

193 The slope length and steepness (LS) is also another factor that describes the sensitivity of soil to
 194 erosion that governs the velocity of runoff water exerting high pushing forces on the soil particles
 195 and causing detachment of soil particles, which in turn lead to erosion. In this catchment, the slope
 196 ranges from 0 to 79.9 %, and due to the steepness of the slope, the soil loss is very visible especially
 197 for the slope values of more than 11% [6]. The ranges of slope in degree (Fig.9) and the
 198 corresponding LS-factor values were generated (Fig.10) according to the studies conducted by
 199 (Dinka, 2020;Thapa, 2020). A model builder for the RUSLE model was developed and raster map
 200 algebra was applied in ArcGIS version 10.4 to quantify the total annual soil loss. It was found that
 201 the quantified average annual soil loss (Fig.13) which ranges from 0.0 to 76.5 t ha⁻¹ yr⁻¹ was
 202 obtained in the catchment. In this study, qualitative classification based five erosion severity
 203 classes as very high (> 45 t ha⁻¹ year⁻¹), high (30-45 t ha⁻¹ year⁻¹), moderate (15-30 t ha⁻¹
 204 year⁻¹), low (0-15 t ha⁻¹ year⁻¹) were identified (Fig.12 and Fig. 14). The spatial variation of
 205 the soil loss severity map generated in the RUSLE model has a paramount role to alert land
 206 resources managers and all stakeholders in controlling the effects via the implementation of both
 207 structural and non-structural mitigations. The results of the RUSLE model can also be further
 208 considered along with the catchment for practical soil loss quantification that can help for
 209 protection practices.



211

Figure 9: The range of Slope in Fincha Catchment (derived from DEM (12.5m x 12.5m))

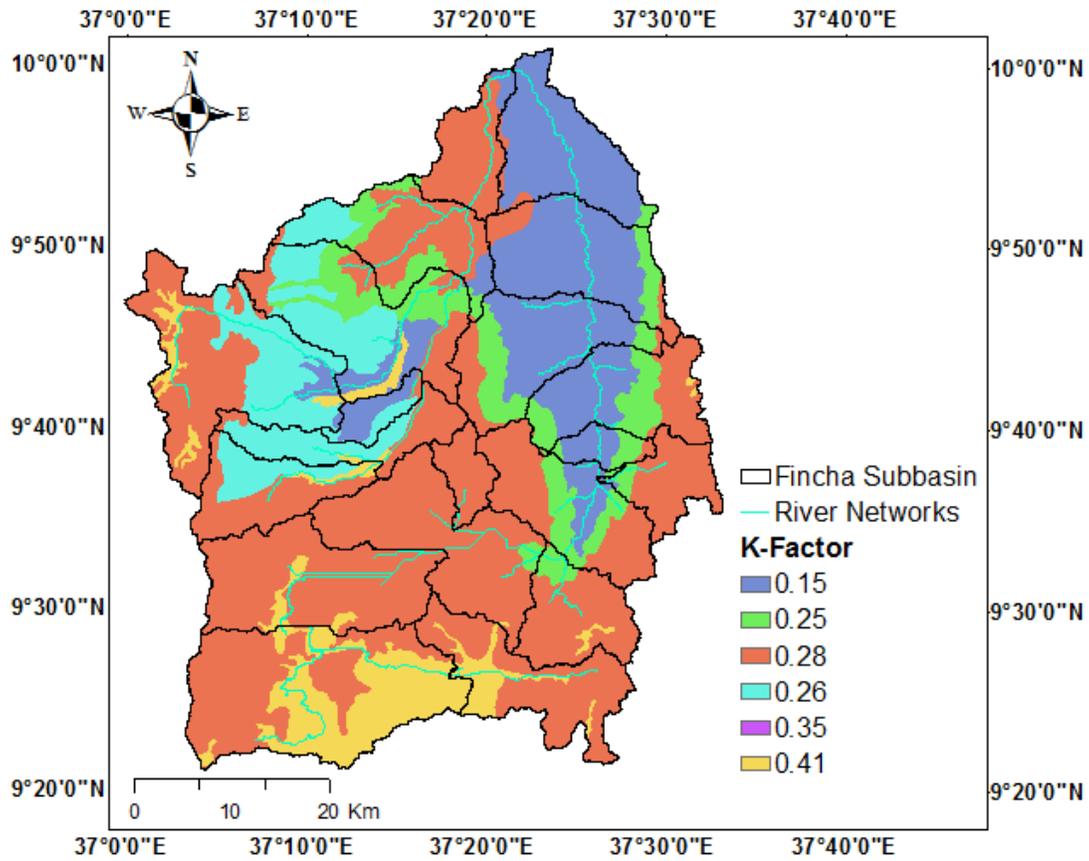


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Figure 10: LS factor generated from slope (%) at 12.5m x 12.5 m spatial resolution

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Figure 11: Soil erodibility factor generated from major soil types in Fincha catchment

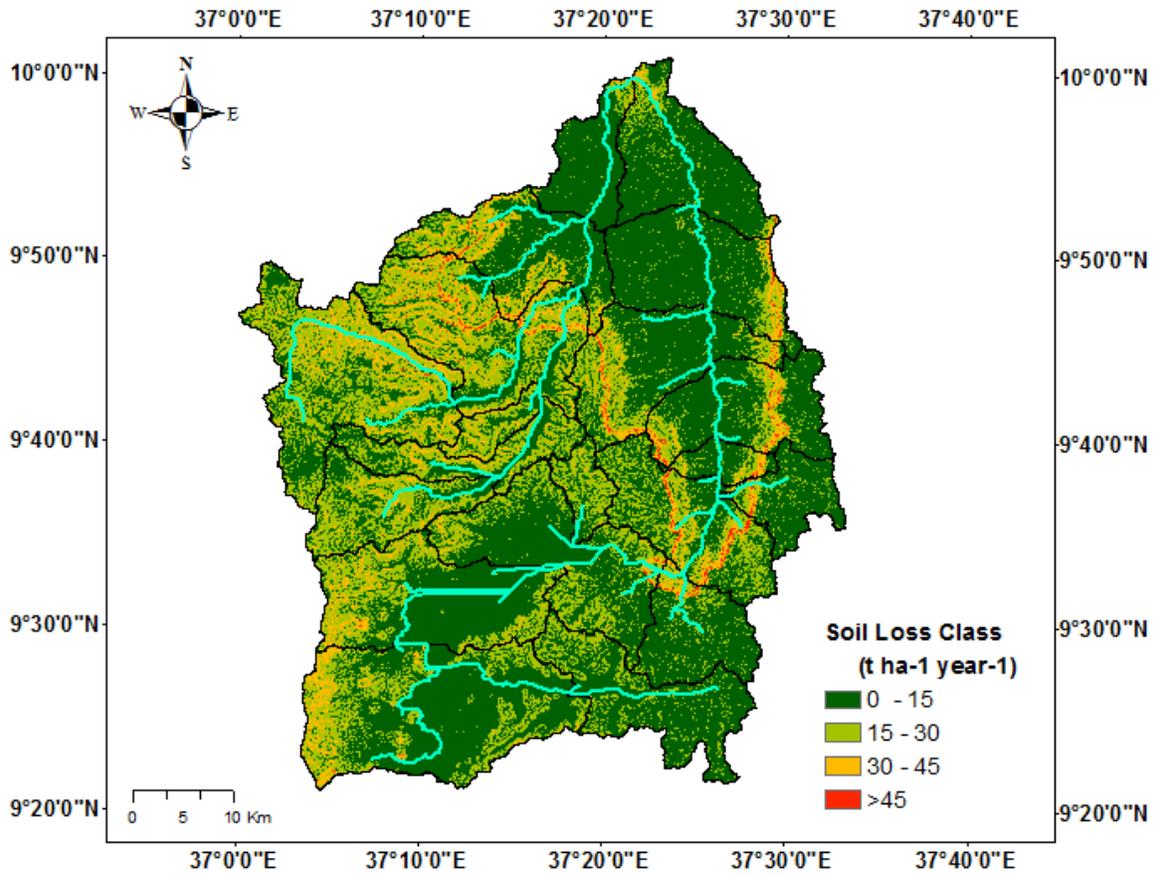


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Figure 12: Soil Erosion class in the catchment

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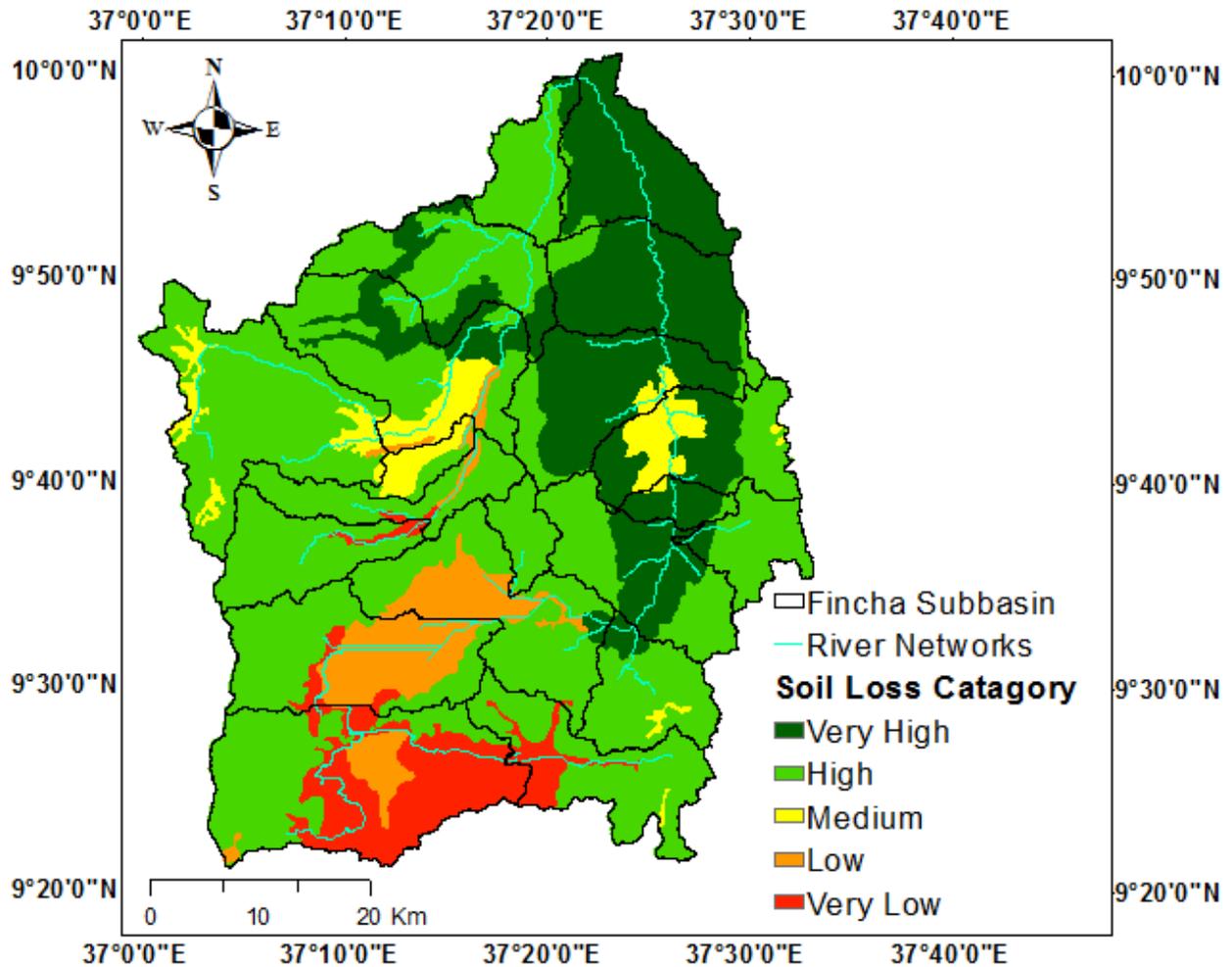


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Figure 13: Soil loss classification map of Fincha catchment

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223

224 Figure 14: Qualitative based spatial variation of soil erosion severity map in Fincha catchment

225 4 Conclusion

226 The quantification of soil loss using an integrated RUSLE model and GIS is successfully provided
 227 a qualitative classification-based identification of soil loss severity understandings in Fincha
 228 catchment of Abay river basin in Ethiopia. In general, five erosion severity classes as very high (>
 229 45 t ha⁻¹ year⁻¹), high (30-45 t ha⁻¹ year⁻¹), moderate (15-30 t ha⁻¹ year⁻¹), low (0-15 t ha⁻¹
 230 year⁻¹). The soil erosion-prone areas map generated in this catchment provides necessary
 231 information for soil and land resources management practices for the implementation of either
 232 structural or non-structural soil conservation measures. From this study, it was found that the upper
 233 and the low-lying areas are highly vulnerable to soil erosion and a soil conservation strategy should
 234 be implemented to control the loss of top fertile soil in the catchment. Additionally, capacity

235 building training should be given for the farmers and soil conservation experts to minimize the
236 man-made soil loss driving factors such as deforestation and traditional way of farming practices.
237 Finally, it was concluded that having information about the spatial variability of soil loss severity
238 map generated in the RUSLE model has a paramount role to alert land resources managers and all
239 stakeholders in controlling the effects via the implementation of both structural and non-structural
240 mitigations. The results of the RUSLE model can also be further considered along with the
241 catchment for practical soil loss quantification that can help for protection practices.

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245 issues in this study.

246 **Authors' contributions**

247 The correspondent author initiates the research idea, reviews relevant literature, design the
248 methods, field data collection, data cleaning, analysis, interpretation, and prepare draft manuscripts
249 for publication. Co-author evaluate the research idea, supervise the whole research activities, and
250 develop the manuscript. All authors read and approved the final manuscript.

251 **Funding**

252 This research fully got fund by Wollega for data collection and processing only.

253 **Availability of data and material**

254 The data generated and processed in this manuscript were included in the manuscript submission.

255 **Ethics approval and consent to participate**

256 This research paper is part of my community-based project entitled 'Natural Resources
257 management strategies '. Therefore, all authors approve to publish the findings, and there is no
258 ethical conflict.

259 **Consent for publication**

260 All authors read the manuscript and agreed to publication.

261

262 **Competing interests**

263 The authors declare that they have no competing interests.

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Figures

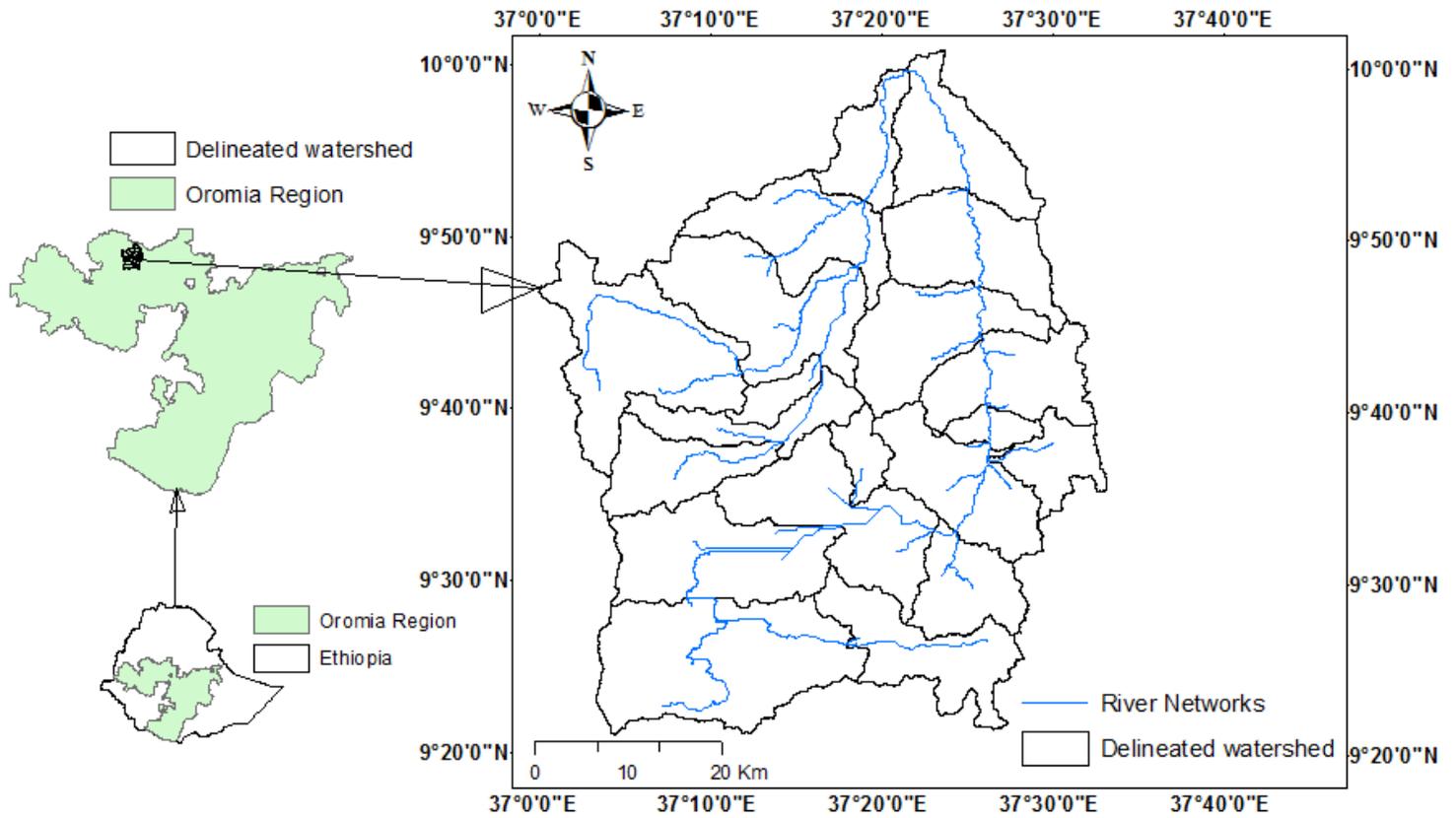


Figure 1

Geographical Location of the study area Note: The designations employed and the presentation of the material on this map do not imply the expression of any opinion whatsoever on the part of Research Square concerning the legal status of any country, territory, city or area or of its authorities, or concerning the delimitation of its frontiers or boundaries. This map has been provided by the authors.

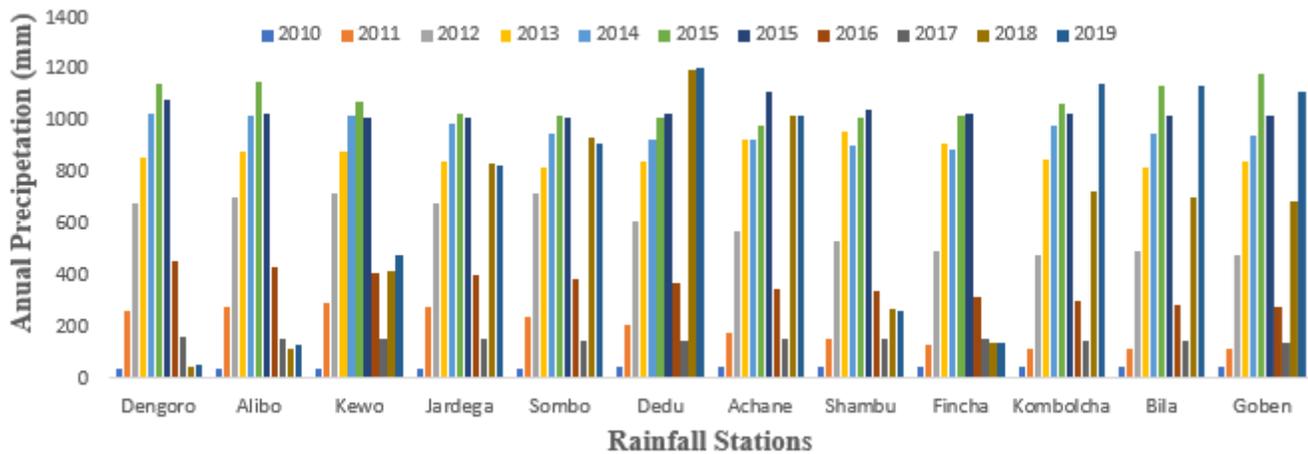


Figure 2

Mean annual precipitation of stations in and surrounding Fincha catchment

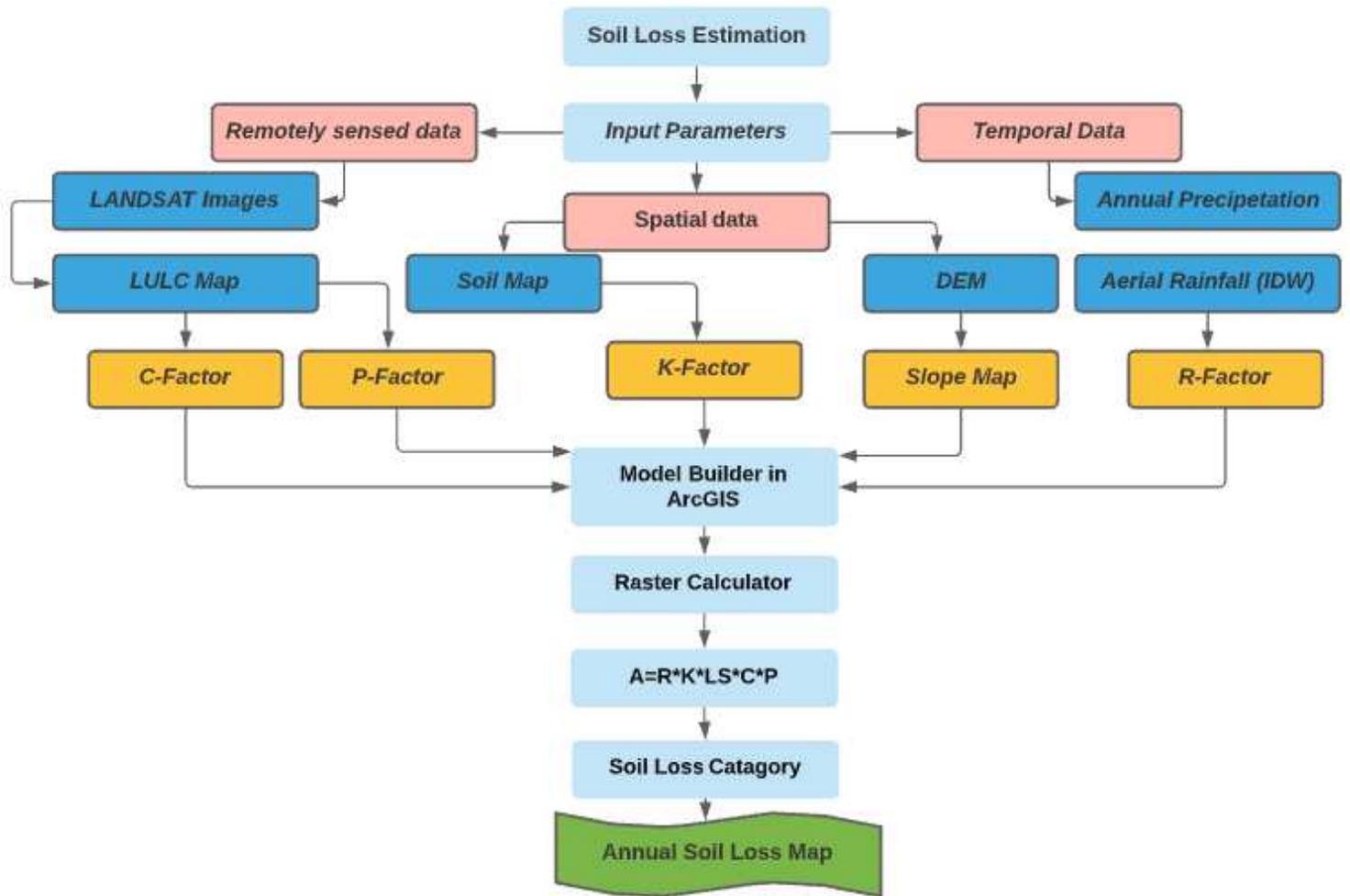


Figure 3

Detail flowchart of the steps and data needed in the study

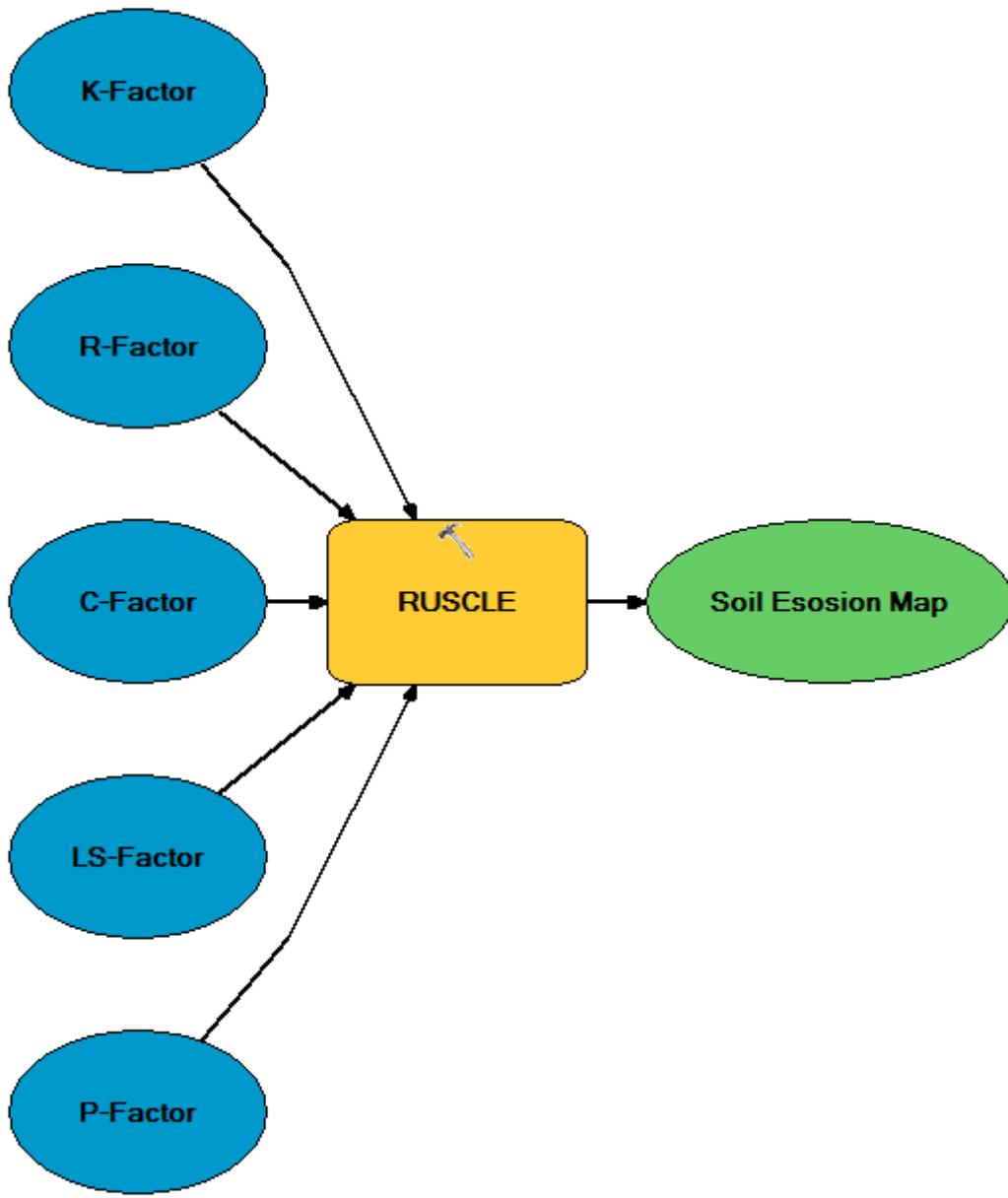


Figure 4

Model builder developed in ArcGIS for combination of the significant factors

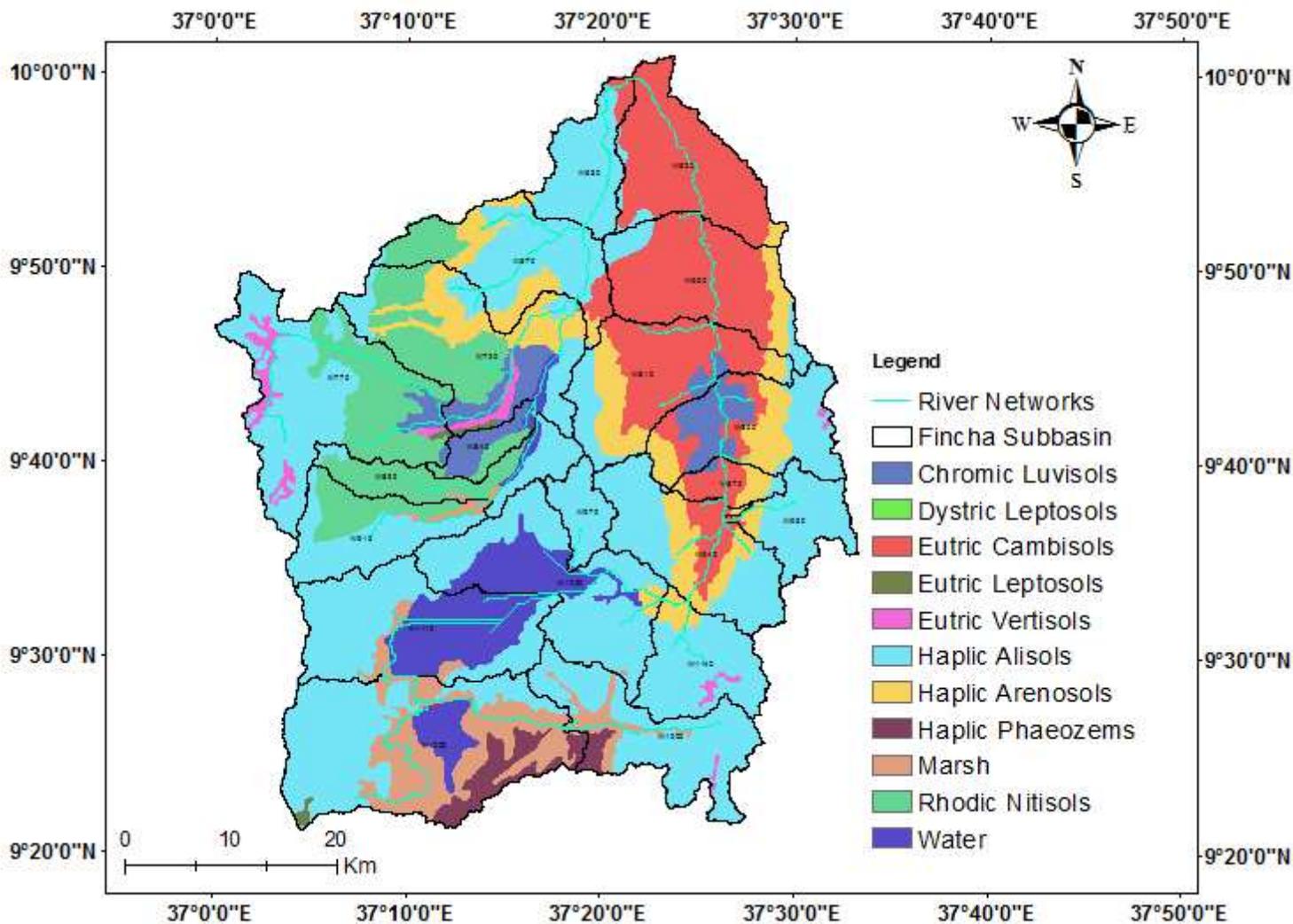


Figure 5

Soil types in the Catchment Note: The designations employed and the presentation of the material on this map do not imply the expression of any opinion whatsoever on the part of Research Square concerning the legal status of any country, territory, city or area or of its authorities, or concerning the delimitation of its frontiers or boundaries. This map has been provided by the authors.

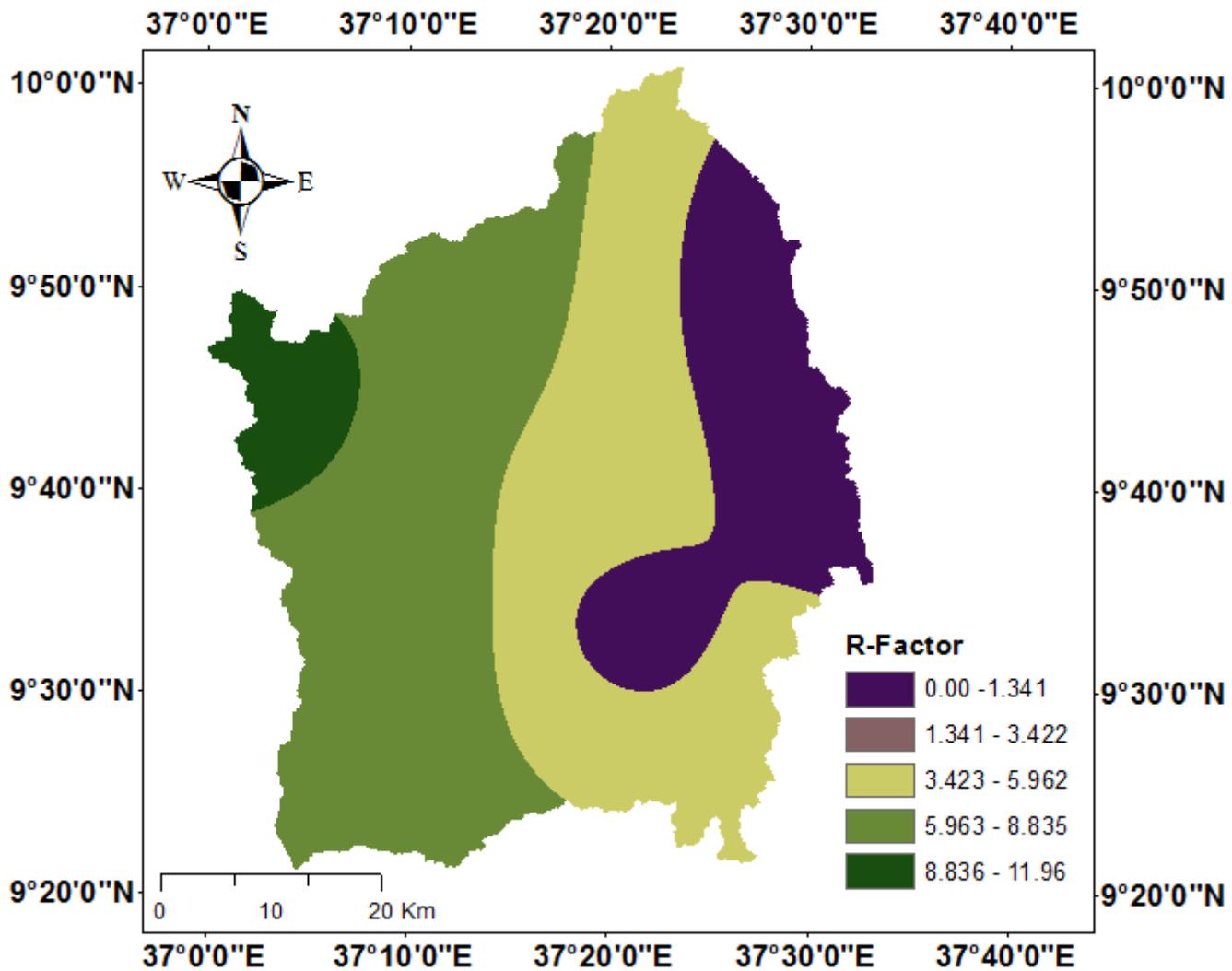


Figure 6

Rainfall erosivity factor (R-factor) values in Fincha catchment Note: The designations employed and the presentation of the material on this map do not imply the expression of any opinion whatsoever on the part of Research Square concerning the legal status of any country, territory, city or area or of its authorities, or concerning the delimitation of its frontiers or boundaries. This map has been provided by the authors.

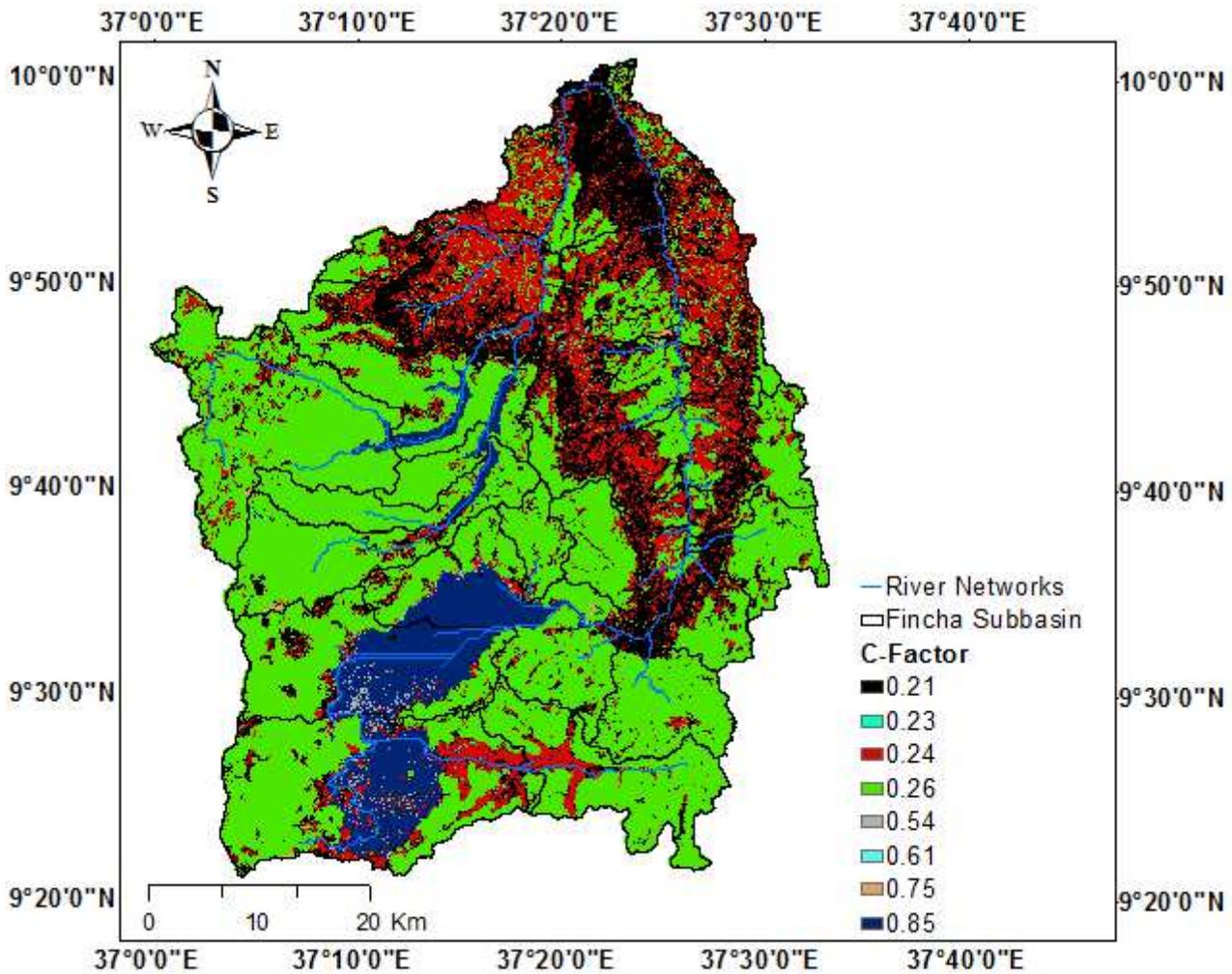


Figure 7

Major land use land cover (LULC) types and corresponding C-values Note: The designations employed and the presentation of the material on this map do not imply the expression of any opinion whatsoever on the part of Research Square concerning the legal status of any country, territory, city or area or of its authorities, or concerning the delimitation of its frontiers or boundaries. This map has been provided by the authors.

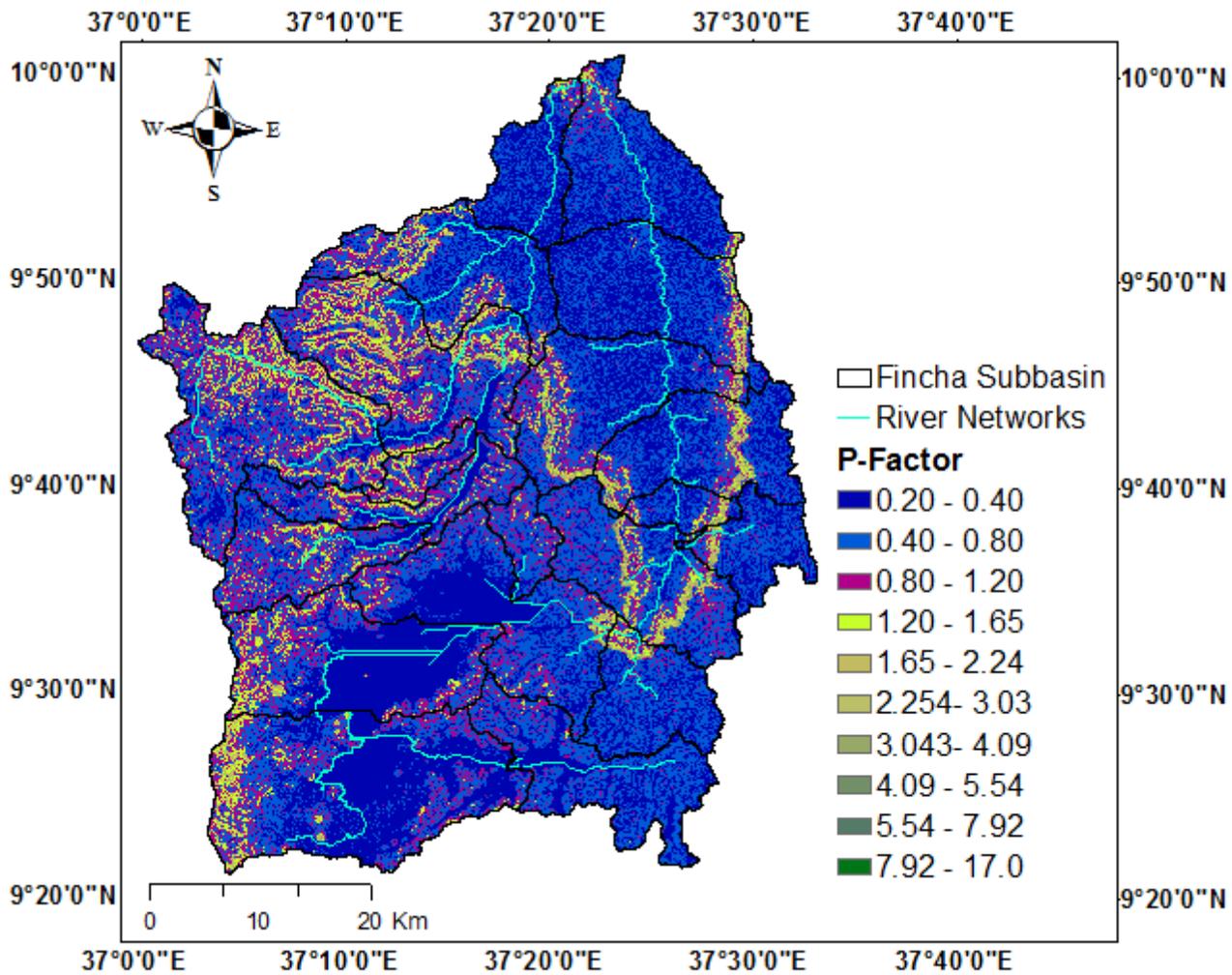


Figure 8

Support and practice factor (P-factor) values in Fincha catchment Note: The designations employed and the presentation of the material on this map do not imply the expression of any opinion whatsoever on the part of Research Square concerning the legal status of any country, territory, city or area or of its authorities, or concerning the delimitation of its frontiers or boundaries. This map has been provided by the authors.

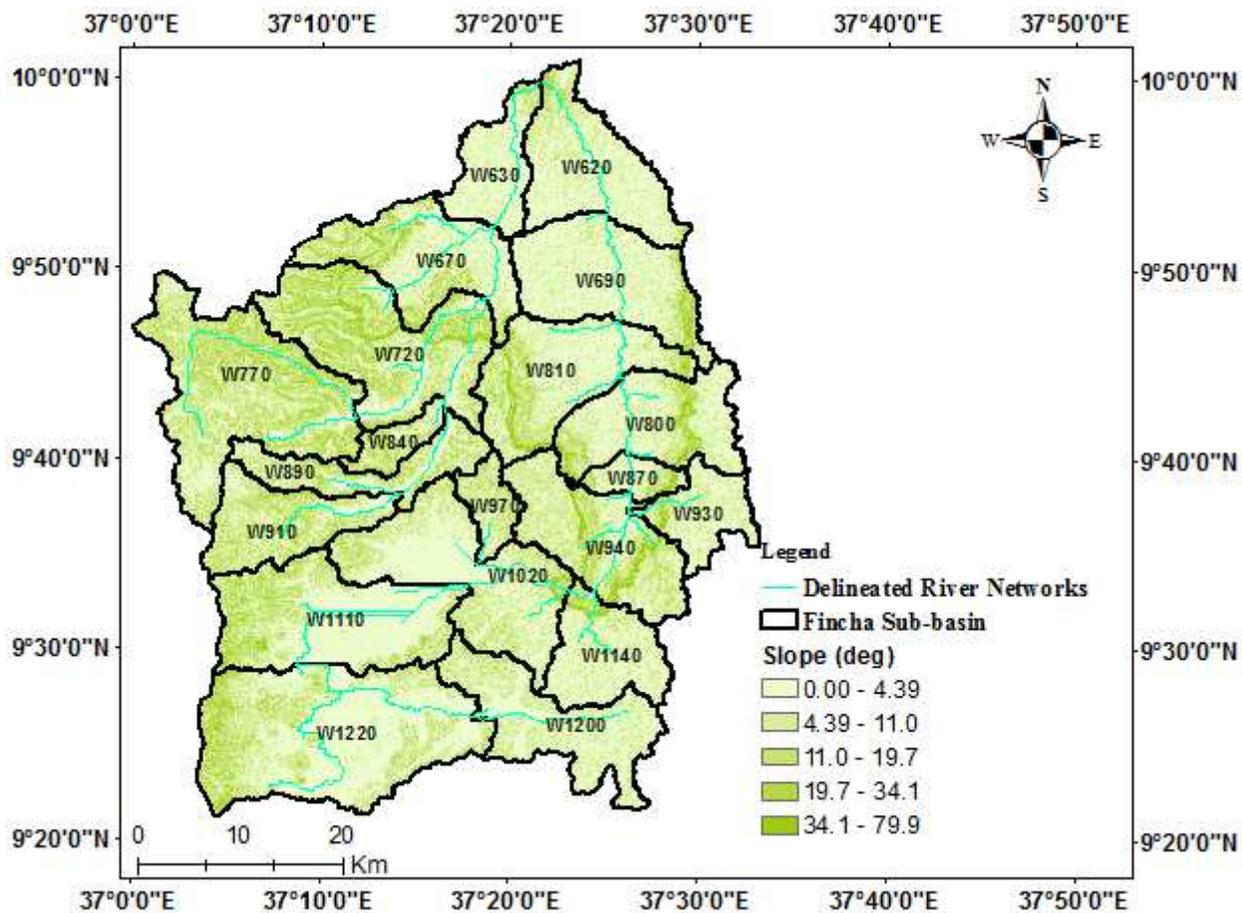


Figure 9

The range of Slope in Fincha Catchment (derived from DEM (12.5m x 12.5m)) Note: The designations employed and the presentation of the material on this map do not imply the expression of any opinion whatsoever on the part of Research Square concerning the legal status of any country, territory, city or area or of its authorities, or concerning the delimitation of its frontiers or boundaries. This map has been provided by the authors.

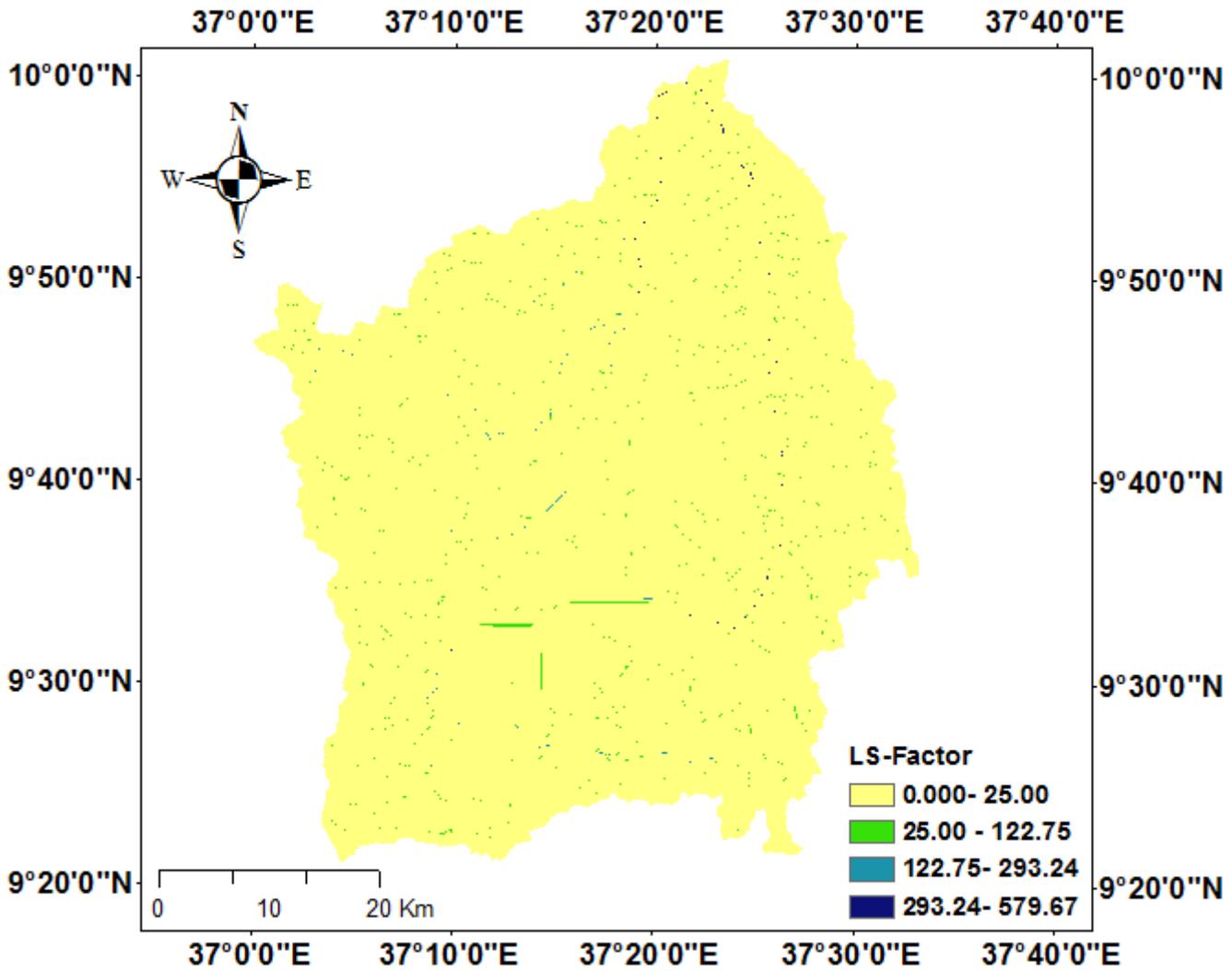


Figure 10

LS factor generated from slope (%) at 12.5m x 12.5 m spatial resolution Note: The designations employed and the presentation of the material on this map do not imply the expression of any opinion whatsoever on the part of Research Square concerning the legal status of any country, territory, city or area or of its authorities, or concerning the delimitation of its frontiers or boundaries. This map has been provided by the authors.

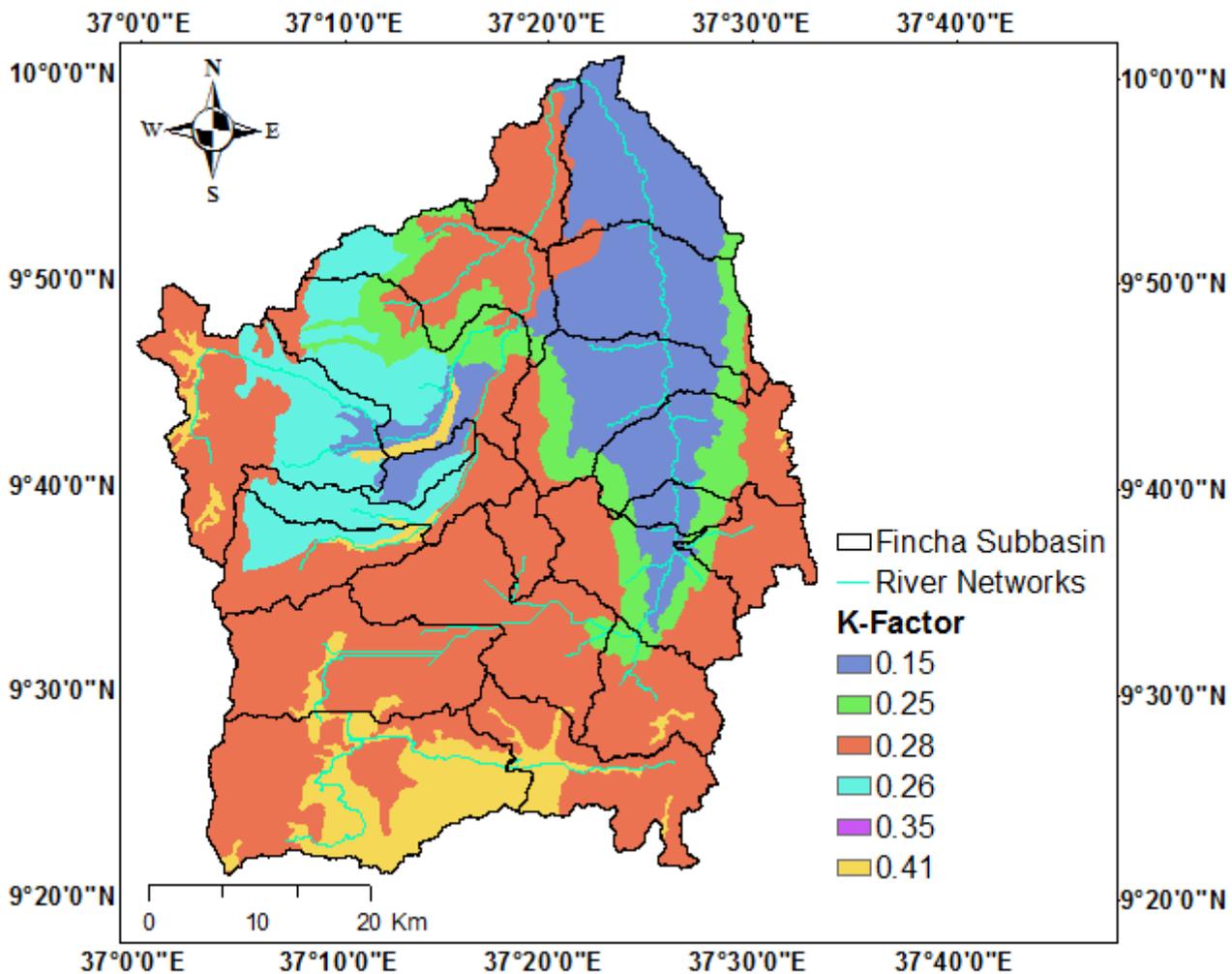


Figure 11

Soil erodibility factor generated from major soil types in Fincha catchment Note: The designations employed and the presentation of the material on this map do not imply the expression of any opinion whatsoever on the part of Research Square concerning the legal status of any country, territory, city or area or of its authorities, or concerning the delimitation of its frontiers or boundaries. This map has been provided by the authors.

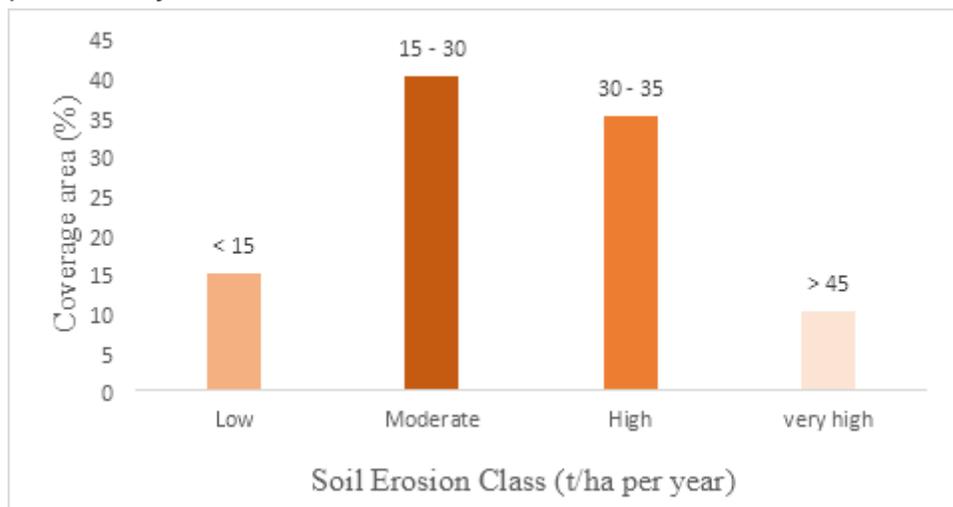


Figure 12

Soil Erosion class in the catchment

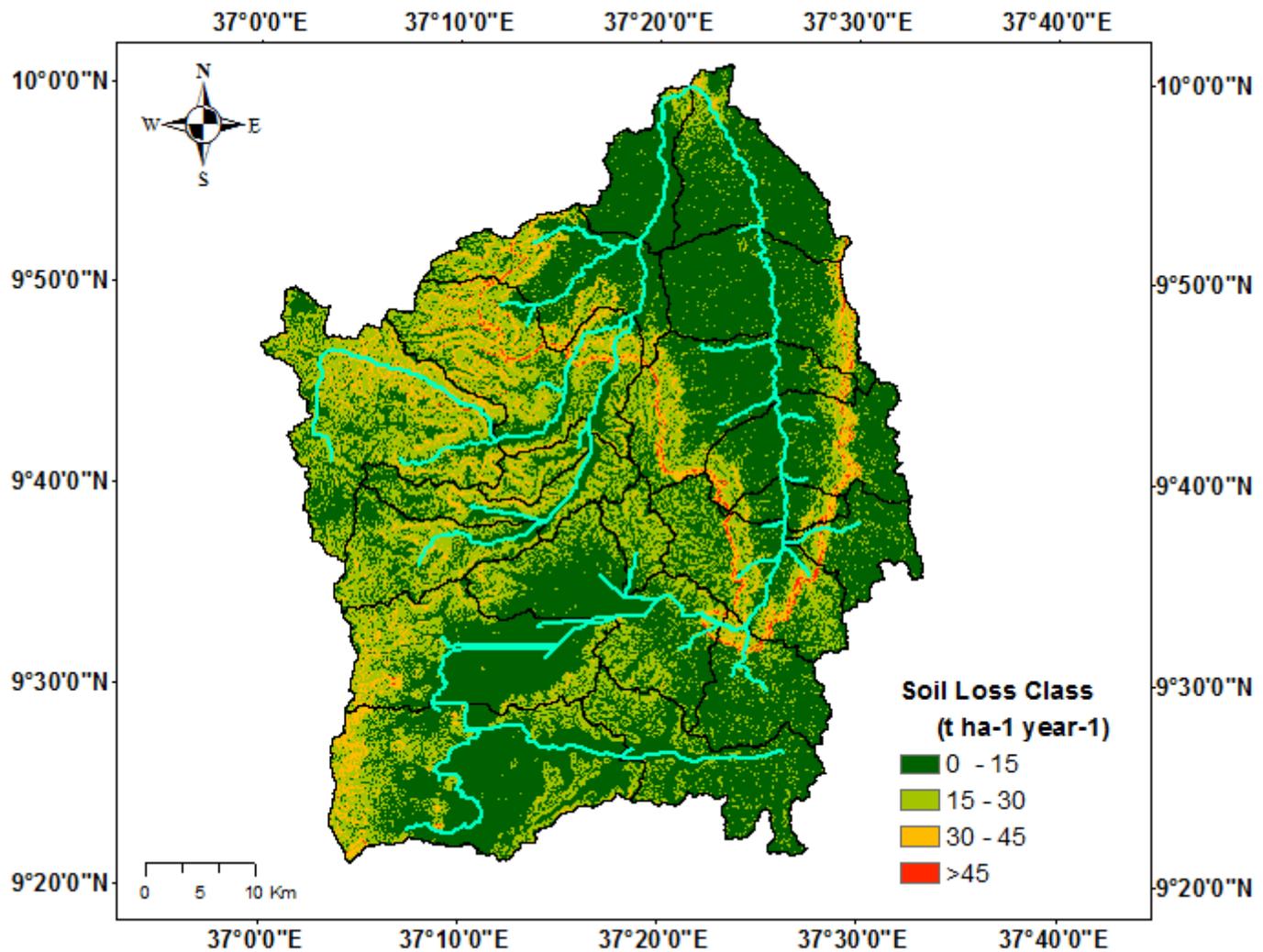


Figure 13

Soil loss classification map of Fincha catchment Note: The designations employed and the presentation of the material on this map do not imply the expression of any opinion whatsoever on the part of Research Square concerning the legal status of any country, territory, city or area or of its authorities, or concerning the delimitation of its frontiers or boundaries. This map has been provided by the authors.

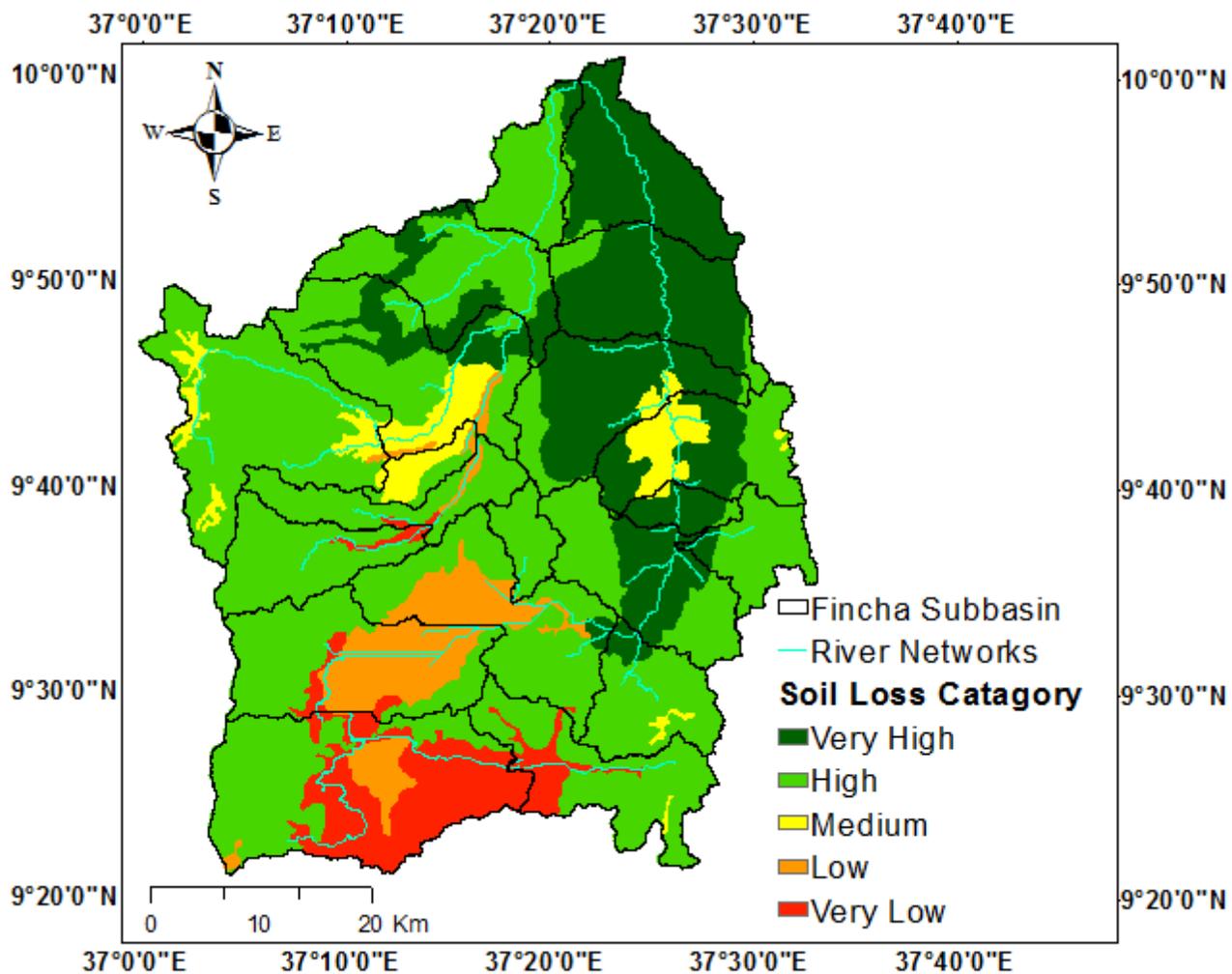


Figure 14

Qualitative based spatial variation of soil erosion severity map in Fincha catchment Note: The designations employed and the presentation of the material on this map do not imply the expression of any opinion whatsoever on the part of Research Square concerning the legal status of any country, territory, city or area or of its authorities, or concerning the delimitation of its frontiers or boundaries. This map has been provided by the authors.