

# PREDICTING SOIL EROSION BY WATER: RUSLE APPLICATION FOR SOIL CONSERVATION PLANNING IN CENTRAL RIFT VALLEY OF ETHIOPIA.

Legass Asmamaw<sup>1\*</sup>, Assen Mohammed<sup>2</sup> and Tesfaye Mathias<sup>3</sup>

<sup>1</sup>Addis Ababa University, Department of Geography and Environmental Studies, Addis Ababa University, P.O.Box 150249, Addis Ababa, Ethiopia

<sup>2</sup>Addis Ababa University, Department of Geography and Environmental Studies, Addis Ababa University, P.O.Box 150116, Addis Ababa, Ethiopia

<sup>3</sup>Addis Ababa University, Department of Geography and Environmental Studies, Addis Ababa University, Addis Ababa, Ethiopia.

## 1 | INTRODUCTION<sup>2,3</sup>

Soil erosion lowers the quality of Ethiopia's agricultural land resources (Haycho et al., 2015). It reduces agricultural and pasture production and exposed people to food insecurity and shortage of livestock feed. Soil erosion by water is a common environmental problem in developing countries including Ethiopia. Water erosion is the leading soil degradation problem in the mountainous regions such as Ethiopia (Hurni, 1988; Mengistu et al., 2015), Vietnam (Pham et al., 2018 citing Trinh, 2015) and Jordan (Farhan et al., 2013). On the top of lowering agricultural production, soil erosion leads to severe siltation of lakes, dams and irrigation canals, and drying both natural and artificially developed water sources (Assen, 2011; Farhan et al., 2013).

Soil erosion is caused and intensified by overexploitation of soil, indiscriminate deforestation, expansion of agricultural land onto ecologically fragile land and poor practice of land management strategies and technologies (Hurni, 1988; Asmamaw and Mohammed, 2019). As a result, soil erosion by water shows a spatiotemporal variability. In Brazil studies confirmed a presence of high (57 t ha<sup>-1</sup> yr<sup>-1</sup>) soil loss by water on cultivated and urban land use/land cover areas (Duque and Melese, 2016). In the highlands of Ethiopia, water erosion ranges from 16 to over 300 t ha<sup>-1</sup> yr<sup>-1</sup> (Hurni, 1988; Mengistu et al., 2015). The local and regional disparity of soil loss rate depends on variations of environmental

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<sup>1</sup> \* Correspondence to Legass Asmamaw, Addis Ababa University, Department of Geography and Environmental Studies, Addis Ababa, Ethiopia.

E-mail: [abahir675@gmail.com](mailto:abahir675@gmail.com); Phone: +251 (0)989133365

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30 and socioeconomic factors (Hurni, 1988). Water induced soil erosion is the result of the combined  
31 interaction of rainfall erosivity, soil erodibility, slope length and steepness, land use/land cover and  
32 land management practices (Renard et al., 1997; Vrieling, 2006). In Ethiopia, soil erosion adversely  
33 affecting agricultural and pastoral and/or agro-pastoral economic activities of high and lowland areas.

34 Several soil erosion models have been developed by scientists (e.g. Nearing et al., 2005) and  
35 applied in different parts of the world. However, the applications of these process-based erosion  
36 models are difficult to employ in developing countries where there is little and poorly measured data,  
37 due to their large and expensive data requirements (Sonneveld et al., 2001). Therefore, it would be  
38 useful to apply models which depend on available and cheap data such as RUSLE (Renard et al.,  
39 1997). Furthermore, knowledge of soil erosion is useful in identifying erosion sensitive hotspot corners  
40 and design appropriate soil and water management plans, strategies and technologies depending on the  
41 degree of the problem and available local resources.

42 In Ethiopia, most of the model and experimental based soil erosion studies have been  
43 concentrated in the highlands of the country (Abate, 2011; Belay & Bewket, 2012; Gebreyesus &  
44 Kirubel, 2009; Hurni, 1983; Mengistu et al., 2015; Moges & Holden 2008). As a consequence, the arid  
45 and semiarid lowlands of the country have been given little attention from the focus of scientific  
46 research (Woldemariam et al., 2018). However, the research carried by Schewel (2019) in Adami Tulu  
47 Jido Kombelcha, East Shewa zone of Oromia region, outside the present study areas in Ethiopia  
48 confirmed the transformation of the pastoral economy to agro-pastoral and finally to permanent  
49 agriculture over time. Some soil erosion assessments made in the lowlands of Ethiopia are too general  
50 and would not be useful for local specific applications (Bhan, 1988). On the other hand, mainly with  
51 the application of available irrigation technologies, present-day agriculture is expanding towards the  
52 arid and semiarid lowlands of Ethiopia (Mekonnen et al., 2019). Therefore, the present research is  
53 profoundly important and unique to fill the existing research gap on soil erosion studies lacking in the  
54 semi-arid and arid lowlands of Ethiopia at large and the Middle Awash Valley of Afar region in  
55 particular.

56 The present study aims to investigate the magnitude of soil loss and main drivers of soil erosion  
57 with the application of RUSLE in the Middle Awash Valley (MAV), Central Rift Valley of Ethiopia.  
58 The results of this research have relevance to (1) understand the degree of soil erosion in the arid and  
59 semi-arid lowlands of Ethiopia, (2) identify the major human and environmental factors accelerating  
60 soil erosion occurrence in semi-arid and arid agro-ecologies, and (3) produce potential soil erosion risk

map and recommend appropriate land management strategies to be undertaken in preventing water-driven soil erosion challenges.

63

## 64 **2 | MATERIALS AND METHODS**

65

### 66 **2.1 | Description of the study area**

67

68 The study is made in the Middle Awash Valley (MAV) of the Awash Fentale and Amibara districts  
69 (locally called *woreda*) of the Afar region, Central Rift Valley of Ethiopia. It lies between 8°30' 12" -  
70 9°50' 03" N latitude and 39°50' 20" - 40°32'0" E longitude (Figure 1). The study area covers  
71 2,148.72km<sup>2</sup> (214, 872ha and elevation ranges from 688m to 1852m asl (meters above mean sea level).  
72 The slope gradient is monotonously flat in large parts of the study area and ranges from almost zero on  
73 flat grounds to 30% to 45% in hilly landforms (FAO, 2006).

#### 74 **Insert FIGURE 1**

75 Considering available meteorological data from Melka Worer station (730masl, 09° 19' 15.5"  
76 N latitude and 40° 11' 56.3''E longitude (located in the centre of the study site), the MAV has a semi-  
77 arid climate with 550.95mm mean annual rainfall and 26.75<sup>0C</sup> mean annual temperature. Temperature  
78 is high throughout the year, which is beyond the optimal requirements of most cultivated plants and  
79 animals. The mean monthly temperature varies from 24°C in December to 32°C in June (Figure 2). The  
80 mean annual rainfall ranges from 238.8mm in 2004 to 818.1mm in 1982, giving a high inter-annual  
81 variability. The low rainfall amount causes shortages of animal feed and water commonly leading to  
82 the toll death of livestock and human food insecurity. The trend of rainfall is irregular and difficult to  
83 predict. As Westphal (1975) discussed, the Middle Awash Valley experiences arid climate, low and  
84 uncertain rainfall and high evaporation rate. The available climatic data for Melka Worer station as  
85 well disclosed the occurrence of rainfall above the average (550mm) within in thirty-five years period  
86 was only in 1982, 1988, 1989, 1996, 2004, 2005 and 2012 (Figure 2).

#### 87 **Insert FIGURES 2a & 2b**

88 The major soils of the study area include Leptosols, Luvisols, Cambisols, Fluvisols and  
89 Andosols (MoA, 2013; FAO, 1984; Figure 3). As the area is poorly vegetated mainly associated with  
90 scanty rainfall, the soils have low organic matter contents. The Leptosols occupy the steeper and higher  
91 grounds of the study area. Fluvisols are commonly found along the Awash River course. Luvisols are  
92 major soils of the flat slopes, whereas Andosols are common where volcanic ashes are locally found.  
93 The foot of local hills and intermediate sloppy lands are occupied by Cambisols (Figure 3).

### Insert FIGURE 3

Analysis of the 2016 satellite imageries of 30 X 30 m cell size resolution reveals that the study area has six land use/land cover (LU/LC) types (Figure 4). The shrub and cultivated lands, respectively, accounting 46.6% and 30.3% of the total area were the predominant LU/LC patterns.. The shrubland includes mixes of natural shrub and invasive *Prosopis juliflora* species. The remaining parts of the study area are covered with grassland (14.4%), urban settlement (4.9%), forestland (3.3%), and water body (0.5%).

### Insert FIGURE 4

Most local communities of the study area are mainly traditional pastoralists. Crop cultivation in the MAV depends on irrigation water of the Awash River and its tributaries. Cotton, sugarcane and sorghum were the main crops. Most tributaries of the Awash River usually dry up in the lowlands as soon as rainfall ceases in their surrounding highlands (Kloos, 1982). Recently, mainly due to the villagization and expansion of irrigation, some pastoralist have started small-scale irrigation agriculture as additional source of livelihoods (Mekonnen et al., 2019). As a result, some pastoralists in fourteen *kebeles* (lower administrative units of Ethiopia) of Amibara and Awash Fentale *woreda* were becoming agro-pastoralists. However, the pastoral communities predominantly depend on livestock production, which includes large herds of camels, cattle, and ruminants such as sheep and goats. The agro-pastoralists practice both livestock rearing and crop production which mainly includes sugarcane, maize, onion, tomato, cabbage, and cotton. Shortages of grazing and irrigation land, and water, lack of access to the market for agricultural products, soil erosion in the form of gullies, flooding and wide invasion of land by *Prosopis juliflora* were challenges experienced by local communities (Mekonnen et al., 2019).

The Afar region with about 1,060,573 people in 1994 was one of the lowest sparsely populated regions of Ethiopia (CSA, 1998). However, the population of the Afar region has increased to 1,390,217 in 2007, and possibly resulting in more demand for basic resources such as forest, water, and land for agriculture and settlement (CSA, 2010).

## 2.2 | Research Methodology<sup>4</sup>

Various spatial datasets were obtained from different organizations and processed using the RUSLE model with the applications of GIS tools and RS techniques. Thus, the RUSLE model is applied to

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<sup>4</sup> Data Availability: the authors are in agreement if the results of this research have been shared for open accessibility

123 estimate soil loss rates, map the six RUSLE factors and spatial variability and mean annual of soil loss  
124 of the study area (sections 2.4.1-2.4.5). Besides, key informant interviews (KII) and focus group  
125 discussions FGD) were conducted with development agents and purposely selected local community  
126 members. The KII and FGD members were recruited based on experience and knowledge merit to  
127 understand the severity and spatial variation of soil erosion rates.

## 128 129 **2.3 | The Revised Universal Soil Loss Equation (RUSLE) Model**

130  
131 The RUSLE is a widely used and validated erosion model in predicting the long term average annual  
132 soil loss rate. Soil loss results from the combined interaction of six RUSLE factors: rainfall erosivity,  
133 soil erodibility, slope length and steepness, land use/land cover and land management practices  
134 (Millward & Mersey, 1999).

135 The modified (RUSLE) model has several merits: 1) it employs cheap and non-complicated data  
136 input generated from easily accessible sources 2) can be easily connected to GIS and RS technologies  
137 which makes the model variables and mean soil loss rate computation to be efficient, manageable and  
138 easy to handle (Pham et al., 2018), and 3) is executed in conjunction with a raster-based GIS to predict  
139 cell by cell potential erosion to identify spatial soil loss variation within the research area (Millward  
140 and Mersey, 1999). The spatial variation of soil loss in the MAV is the result of the spatial  
141 heterogeneity of the RUSLE factors (Farhan et al., 2013). For this research, the MAV has been divided  
142 into a small homogenous unit of 30m by 30m grid cell size before running the computation of the soil  
143 loss (Farhan et al., 2013). This RUSLE model computes the mean annual soil loss rate using equation  
144 1:

$$145 \quad A = R * K * LS * C * P \quad (1)$$

147  
148 Where, A = the average annual soil loss per unit area ( $\text{tons ha}^{-1} \text{ year}^{-1}$ ); R = rainfall erosivity ( $\text{MJ mm}$   
149  $\text{ha}^{-1} \text{ hr}^{-1} \text{ yr}^{-1}$ ); K = soil erodibility ( $\text{t ha hr ha}^{-1} \text{ MJ}^{-1} \text{ mm}^{-1}$ ); LS = slope length-steepness  
150 (dimensionless); C = land use/land cover (dimensionless); and P = conservation/management factors  
151 (dimensionless).

## 152 **2.4 | Determining the six RUSLE Factors**

153 The processes used to generate each of the six RUSLE model parameter values are explained as  
154 hereunder.

### 155 **2.4.1 | Rainfall erosivity (R) factor**

The R factor is the product of the total kinetic energy multiplied by the maximum 30 minutes rainfall intensity (Wischmeier & Smith, 1978). This factor measures the erosivity of average annual rainfall and runoff to cause soil erosion (Farhan et al., 2013). The spatial rainfall distribution of the study area was computed from gridded meteorological data (CHIRPS, <ftp://ftp.chg.ucsb.edu/pub/org/chg/products/CHIRPS-2.0/>). Kriging interpolation method has been used to generate the raster R factor. Kriging is a multistep process including exploratory statistical analysis of the data, variogram modeling, creating the surface, and exploring a variance surface (Burrough & McDonnell, 1998). The annual rainfall erosivity factor can be computed by different methods (Gitas et al., 2009; Parveen & Kumar, 2012). However, the spatial distribution of R factor value (Figure 5) was computed with the ArcGIS raster calculator tool using Hurni (1985) formula expressed in Equation 2:

$$R = -8.12 + 0.562P \text{ (Hurni 1985)} \quad (2)$$

Where R is the calculated rainfall-runoff erosivity factor and P is the mean annual rainfall (mm).

The Hurni's (1985) R factor formula shows the relationship between mean annual total rainfall and R factor. The variation of R factor depicts the difference in the amount and distribution of rainfall across spaces (Farhan et al., 2013).

#### 2.4.2 | Soil erodibility (K) factor

The K factor refers to the susceptibility of the soil to erosion agents. It is mainly associated with some physical soil characteristics (Shabanin et al., 2014) which specifically depends on soil texture, soil structure, soil organic matter content, soil moisture and surface roughness (Lal, 2001; Millward and Mersey, 1999; Renard et al., 1997). The K factor indicates the degree of resistance of soil particles to raindrop detachment and transport capacity of runoff. Soil erodibility (K) value ranges from 0 to 1, where values closer to 0 show the least soil susceptibility to erosion and values closer or equal to 1 are the most erodible soils and highly prone to soil erosion (Farhan et al., 2013; Ganasri and Ramesh, 2016; Mhangara et al., 2012; Zerihun et al., 2018). Soils having better infiltration rates such as sandy textured one becomes less susceptible to water erosion and less surface water accumulation to initiate runoff (Wischmeier & Smith, 1978). As obtaining soil erodibility (K) data is one of the most challenging tasks (Bahrami et al., 2005), the K factor value of the soil types of MAV (Figure 3) estimated by FAO (1984) has been used in this study (Table 1).

**Insert Table 1**

189 In this study, the vector soil data were converted to raster format to produce a continuous spatial  
 190 variability map of soil erodibility (Figure 6). Using the geoprocessing reclassification tool, the soil grid  
 191 data/map had been recategorized based on the K value of each soil type (Table 1; Figure 6).

192

### 193 **2.4.3 | Topographic (LS) factor**

194 The LS factor is comprised of the effects of slope length (L) and slope steepness (S) on soil erosion rate  
 195 (Farhan et al., 2013; Panagos et al., 2015). The LS factor influences the sediment transport capacity of  
 196 the flow (Moore & Wilson, 1992).

197 Slope length is “the distance from the point of origin of overland flow to the point where either  
 198 the slope decreases to the extent that deposition begins or runoff water enters to well-defined channel”  
 199 (Ganasri & Ramesh, 2016). As slope length increases, soil loss per unit area rises as the gradual  
 200 accumulation of runoff down slope increases (Farhan et al., 2013). Slope steepness refers to “the  
 201 gradient of the land immediately surrounding the site” (FAO, 2006). The steeper slope is, the higher  
 202 soil loss would be due to the impact of velocity and erosivity of runoff. However, slope steepness that  
 203 indicates the effects of slope gradient on soil erosion has a greater impact than slope length (Farhan et  
 204 al., 2013; Ganasri and Ramesh, 2016).

205 The topographic (LS) factors do not consider the three-dimensional distribution of the terrain in  
 206 estimating soil loss (Mitasova, Hofierka, Zlocha, & Iverson, 1996). These LS factors assume soil loss  
 207 increases with slope length and/or upslope contributing area (Desmet and Govers, 1996; Moore and  
 208 Burch, 1986). However, slope length and/or upslope contributing factor does not necessarily lead to  
 209 higher soil loss unless the three-dimensional terrain complexity is considered (Mitasova, et al., 1996).  
 210 Therefore, this fact has to be taken as one limitation of the RUSLE model. As suggested by Mitasova  
 211 and Mitas (1996), the LS factor is computed with equation 3 (using ASTER 2016 global raster satellite  
 212 image with 30 x 30m resolution as a basic data source):

213

$$214 \quad LS = \left( \frac{FA \cdot \text{cell size}}{22.13} \right)^m * \left( \frac{\sin(\text{slope angle})/0.01745}{0.09} \right)^n \quad (3)$$

215

216 Where, FA = flow accumulation derived from Digital Elevation Model (DEM) after processing fill and  
 217 flow direction in ArcGIS; cell size is the grid cell size derived from DEM 30 m by 30 m resolution,  
 218 slope angle is in degrees (°), and 0.01745 is the parameter to convert degrees to radians, m and n are  
 219 slope length and slope steepness exponents.

220 The exponent values are ranging from 0.2 to 0.6 for m and from 1 to 1.3 for n (Pham *et al.*,  
221 2018). The lower exponent values are used for prevailing sheet flow and higher values for prevailing  
222 rill flow. The 22.13m (72.6ft.) values and 0.09 radian (5.14°) are the length and slope angle of the  
223 standard USLE plot, respectively (Pham *et al.*, 2018).

224 In the advanced LS equation 3, the slope length was substituted by the upslope contributing  
225 area to consider the impact of flow convergence and diversion on soil erosion in the three-dimensional  
226 complex terrain configurations. Thus, equation 3 has considered the contribution of upstream  
227 contributing area and slope gradient in estimating soil loss by LS factor.

#### 228 229 **2.4.4 | Land use/land cover (C) factor**

230  
231 The C factor reflects the influence of land use/cover types on soil erosion rate (Patil & Sharma, 2014).  
232 The C factor ranges from nearly 0 to 1, where values closer or equal to 1 indicate the absence of land  
233 use/land cover in the area and the surface is considered as barren land. However, the C factor value  
234 closer to zero (0) indicates the existence of a well-protected soil by forest or good plant cover (Ganasri  
235 & Ramesh, 2016). An increase in the C factor, therefore, portrays the higher exposure of soils to  
236 erosion and thus, the rise in potential soil loss (Farhan et al., 2013; Table 2).

#### 237 **Insert Table 2**

238  
239 The Normalized Difference Vegetation Index (NDVI) derived by RS technology (Equation 4) is  
240 the most commonly used indicator of vegetation growth (the C factor value). The 2016 Operational  
241 Landsat TM raster image of 30 x 30 m resolution was used to compute NDVI using equation 4  
242 (Parveenu & Kumar, 2012).

243 NDVI is positively correlated with the amount of green biomass and indicates differences in  
244 green vegetation coverage (Knijff et al., 2000). Thus, NDVI value can be an input to calculate the C  
245 factor. The NDVI value has an inverse relationship with the land use/land cover (C) factor value.  
246 Therefore, the rise in the NDVI value shows the decline of C factor which ultimately indicates the  
247 decrease of soil loss with the improvement in vegetation cover (Farhan et al., 2013). Many researchers  
248 calculated the C-factor with different equations (Durigon, et al., 2014; Knijff et al., 2000). However,  
249 the formula suggested by Durigon et al. (2014) as indicated in equation 5 has been used to compute the  
250 C factor values of the study area. A reconnaissance survey was conducted to validate the computed  
251 LULC (C) factor value with the existing reality on the ground.

252



$$NDVI = \frac{(NIR - RED)}{(RED + NIR)} \quad (4)$$

$$C = \frac{(-NDVI + 1)}{2} \quad (5)$$

Where, C= the land use/land cover (C) factor; NDVI= Normalized Difference Vegetation Index; NIR= the surface spectral reflectance in the near-infrared band; and RED = surface spectral reflectance in the red band was extracted from Landsat images.

#### 2.4.5 | Support practices (P) factor

The P factor is regarded as the impact of farming systems on soil erosion. It measures the effect of conservation practices in influencing the outbreak and prevalence of water-induced soil erosion. P factor adjusts the potential erosion by runoff through the implementation of contouring, strip cropping, and terraced farming (Kuok et al., 2013; Wischmeier & Smith, 1978). Some researchers suggested as the P-value is dependent on the slope inclination (Lufafa et al., 2003; Wenner, 1980; Wischmeier & Smith, 1978), whereas others use farming practices to calculate P value (Stone et al., 2000). If there would not be any erosion control practice, the P-value should be 1 (Table 3). The support practices (P) factor of MAV have been calculated using the combination of the 2016 land use/land cover and slope degrees suggested by Shin (1999) cited in El Jazouli et al., (2019) (Table 3).

#### Insert Table 3

### 3 | RESULTS AND DISCUSSIONS

The RUSLE model has been employed in estimating the magnitude of mean annual soil loss (metric ton ha<sup>-1</sup> year<sup>-1</sup>), map spatial soil loss variation and identify erosion hotspot areas by the combined interplay of the six RUSLE model factors. The impact of each of the main erosion factors on the rate of soil loss has been analyzed hereunder.

#### 3.1 | Contribution of RUSLE factors on the soil loss rate

##### 3.1.1 | Rainfall erosivity (R) factor

The intensity, amount and distribution of rainfall are some of the most important physical factors affecting the rate of soil erosion. As computed using equation 2, the R-factor of the MAV ranges from 471.39 to 817.34mm (Figure 5).

282 **Insert Figure 5**

283         The spatial distribution of R factor value varies across the study area. The northeastern corridor  
284 has experienced the lowest rainfall erosivity whereas the northwestern part has relatively encountered  
285 higher rainfall erosivity. In large parts of the Middle Awash Valley, rainfall erosivity ranges from  
286 540.58 to 609.77 and gradually increases towards the east and west directions (Figure 5). As Batjes  
287 (1996) stated, the rainfall erosivity factor value of 800 or below, as seen in most parts of the present  
288 study area, indicates the occurrence of low rainfall erosivity to erode soil resources. Therefore, the  
289 rainfall erosivity (R) factor is not the main driving agent of soil loss in the Middle Awash Valley of  
290 Ethiopia. (Figure 5).

291 **3.1.2 | Soil erodibility (K) factor**

292         The study area has seven different soil types (Table 1). However, depending on their water erosion  
293 vulnerability, these soil types of the Middle Awash Valley have been reclassified into four soil  
294 erodibility (K) factor classes (Table 1; Figure 6). The relatively low erodible Eutric Cambisols with soil  
295 erodibility factor of 0.15 covers 13.71% of the study area. Eutric Cambisols have a high infiltration rate  
296 because of their relatively high sand contents and low content of clay (Belay, 1998). Eutric Cambisols  
297 have, therefore, better resistant and less susceptibility to the eroding power of rainfall than other soil  
298 types of the study area (Table 1). Chromic Luvisols and Vertic Cambisols have relatively high and  
299 dominant clay content (Muller-Samann & Kotschi, 1994) and would have less infiltration rate, and will  
300 have high K values. The relatively most erodible Chromic Luvisols and Vertic Cambisols ( $K = 0.6$ )  
301 cover 2.36% and 50.55% of the total study area. The other relatively less erodible soils of Chromic  
302 Cambisols and Leptosols ( $K = 0.2$ ) together covered 3.16% of the total study area. Chromic Cambisols  
303 (Asmamaw and Mohammed, 2012; Engdawork, 2002; Mohammed et al., 2005) are clayey with  
304 intermediate infiltration rates. Leptosols have shallow depths which would cause low moisture holding  
305 capacity that will generate more surface runoff (Asmamaw and Mohammed, 2012; Mohammed et al.,  
306 2005). Eutric Fluvisols and Vertic Andosols with soil erodibility factors of 0.3 covered 30.21% of the  
307 total study area (Figure 6). These soil types have an intermediate level of soil erodibility, as they  
308 contain relatively high silt content (Table 1), which is less cohesive and susceptible to detachment than  
309 other soils of the study area.

310 **Insert FIGURE 6**

311

312 **3.1.3 | Topographic (LS) factor**

313 The slope length-steepness (LS) of the study area factor ranges from 0 to 20.71 (Figure 7). Most of the  
314 study area, therefore, has a low topographical factor of soil loss owing to the prevalence of low slope  
315 length-steepness in the largest part of the study area. As a result, the LS factor has a low contribution to  
316 soil erosion occurrence in many parts of the Middle Awash Valley of Afar region, Ethiopia (Fig. 7).

317 **Insert FIGURE 7**

#### 318 **3.1.4 | The land use/land cover (C) factor**

319 The C factor of the study area (Farhan et al., 2013) which has been computed from NDVI of the  
320 Landsat satellite image (Figure 8), ranges from 0.3 in relatively forested areas to 0.6 in low vegetation  
321 cover area. The presence of low land cover which is directly understood from the lowest NDVI value  
322 would negatively affect the occurrence and spread of rainfall-induced soil erosion (Figure 8). As a  
323 result, the trend of soil erosion is increasing with the decline of vegetation cover. The shrub, grassland  
324 and cultivated areas covering 30.3%, 46.6% and 14.4% of the Middle Awash Valley, respectively, are  
325 moderately vulnerable to water erosion. The study carried out by Pamo and Pieper (2000) confirmed  
326 that heavy grazing over the grassland removes the vegetation cover, thereby exposing soil surfaces to  
327 erosion. Similarly, shrublands of semiarid areas have sparse vegetation cover whereas cultivated areas  
328 of annual crops are with no plant cover during their early stage of crop growth and none cropping  
329 periods contribute to the rise of water-induced soil loss. Therefore, it is relevant to create a suitable  
330 balance between resource use and their capacity while implement sustainable use of soil resources  
331 management (Pamo and Pieper, 2000).

332 **Insert FIGURE 8**

333 Low soil erosion is experienced in 3.3 % of the study area covered by forest. Therefore, the  
334 impact of the C factor is moderately significant in triggering water-induced soil loss. The land use/land  
335 cover patterns of the area have to be properly utilized and managed to curb the contribution of the C  
336 factor in halting soil loss and sustainably use the soil resources.

337

#### 338 **3.1.5 | Support practices (P) factor**

339 The P factor values ranges from 0 to 1 (Ganasri & Ramesh, 2016). The P factor value closer to 0  
340 indicates the existence of good conservation practice. However, the P-value of 1 or closer indicates  
341 poor/slight conservation practices (Ganasri & Ramesh, 2016; Hurni, 1988). The support (P) practices  
342 factor of Middle Awash Valley ranged from 0.003 to 1 (Table 2; Figure 9). As portrayed in Figure 9,  
343 areas with the P value of 0.003 have very limited areal coverage with better conservation practices.

344 However, the P-value of 1 portrays the poor/slight land management practice in most of the Middle  
345 Awash Valley areas of Afar region, Ethiopia (Figure 9). As a result, poor land management practices  
346 significantly contribute to the high occurrence of soil erosion by water in the northeastern parts of the  
347 study area.

348 **Insert FIGURE 9**

### 349 350 **3.2 | Magnitude and Spatial Pattern of Soil Loss**

351 The six RUSLE model (RKLSCP) factors are overlaid and multiplied pixel by pixel using the raster  
352 geoprocessing calculator tool in the ArcGIS 10.5 environment to estimate the soil loss rate (Metric tons  
353  $\text{ha}^{-1} \text{ year}^{-1}$ ) and map the spatial soil loss variation of the study area (Figure 10).

354 **Insert FIGURE 10**

355 The mean annual soil loss ranges from close to 0 to over slightly 20 tons  $\text{ha}^{-1} \text{ year}^{-1}$  (Figure  
356 10). Depending on soil loss magnitude, the soil erosion of the study area has been classified into five  
357 soil erosion severity classes (Table 4). The classification of the soil loss risk was carried out to map the  
358 spatial distribution of soil loss and identify soil erosion hotspot areas for land management  
359 prioritization.

360  
361 **Insert Table 4**

362  
363 As confirmed by the RUSLE model result, water-induced soil loss is very low and would not be  
364 considered as the major constraint in about 60% ( $1271.03\text{km}^2$ ) of the study area. In the very low  
365 erosion rate corners, the magnitude of soil erosion accounts for up to  $0.5 \text{ ton soil loss ha}^{-1} \text{ year}^{-1}$  (Table  
366 4). This is mainly attributed to the lower effect of rainfall erosivity and local LS topographical factors.  
367 Such areas have the fifth (V) priority of land management which could be implemented after all the  
368 other soil erosion-prone areas have been conserved.

369 In some of the study areas, sugarcane plantation forms the major cultivated crop. As  
370 sugarcane plantation cover protect soil from raindrop detachment and runoff, the sugarcane cultivated  
371 area has a low rate of soil loss, below  $1 \text{ ton ha}^{-1} \text{ year}^{-1}$  (Figure 10). Similarly, low soil erosion rate has  
372 been experienced in sugarcane farms of semi-arid areas of Morocco (North Africa) (Lahloui, Rhinane,  
373 Hilali, Lahssini, & Khalile, 2015). This contradicts contradictory to results obtained from highland  
374 cereal cultivated areas of Ethiopia as proved by many research findings (Asmamaw & Mohammed,  
375 2019; Bewket & Teferi, 2009; Gelagay & Minale, 2016). Similarly, low soil erosion rate has been

376 experienced in sugarcane farms of semi-arid areas of Morocco (North Africa) (Lahloui, Rhinane,  
377 Hilali, Lahssini, & Khalile, 2015). About 75% of the total area in many semi-arid lowlands of the  
378 world experience slight (0-2) tons of soil loss  $\text{ha}^{-1} \text{ year}^{-1}$  ( Mohammed et al., 2017). On the contrary,  
379 the wetter intensively cultivated and rugged highlands of Ethiopia are highly vulnerable to the risk of  
380 severe and very severe soil erosion than the drier flat lowland areas of the country (Mohammed et al.,  
381 2017; Esa et al., 2018). Thus, over many highland areas, the magnitude of soil loss exceeds both the  
382 tolerable soil loss rate of 18 tons  $\text{ha}^{-1} \text{ year}^{-1}$  and estimated soil formation rate of 2 to 22 tons  $\text{ha}^{-1}$   
383  $\text{year}^{-1}$  (Hurni, 1983).

384           The low to medium soil loss areas with the soil loss rate of 0.5-1 and 1-10 tons  $\text{ha}^{-1} \text{ year}^{-1}$ ,  
385 accounted for only 10.72% (230.24 $\text{km}^2$ ) and 8.07% (173.47 $\text{km}^2$ ) of the Middle Awash of Afar region,  
386 Central Rift Valley of Ethiopia. Generally, a very low to medium rate of soil loss areas covered over  
387 three-quarters (77.94%) of the research area. The high (10-20) and very high (over 20) tons of soil loss  
388  $\text{ha}^{-1} \text{ year}^{-1}$  together covered 22.06% (473.98 $\text{km}^2$ ). The existence of relatively high and very high  
389 erosion rate in some corners of the study area was attributed to the presence of intrinsically less  
390 resistant soils to water erosion, the sparse nature of shrub and vegetation cover, poor support practices  
391 experienced across these parts of the study area (Figures 4, 5, 7 and 8). Therefore, the first and second  
392 priority of soil management has to be given in 4.96% (106.48 $\text{km}^2$ ) and 17.10% (367.50 $\text{km}^2$ ) of the  
393 study area which have very high and high hotspot soil loss rates (Table 4). The rotational use of grazing  
394 lands and enhancement of support and soil water conservation practices would contribute in curbing  
395 the high to very high soil loss rate in 22.06% (473.98 $\text{km}^2$ ) of the study area.

396           The customary dependence only on the present poor land management practices would  
397 drastically lead to the decline in the productivity of grass and cultivated lands. Hence, sustainable  
398 livestock breeding and irrigation agriculture would be challenged in the face of the current adverse  
399 impact of climate change and variability. The productivity of the grasslands can be efficiently enhanced  
400 with effective dryland water conservation strategies, rotation of grazing lands and minimizing the  
401 density of livestock per unit area that focuses on the quality of animal husbandry. Besides, the support  
402 practices and availability of required resources have to be improved along the various land use/land  
403 cover categories to enhance the productivity of cultivated and grasslands. In the Northern highlands of  
404 Ethiopia, the scarcity of loans to farmers by Rural Saving and Credit Cooperative Institutions limits the  
405 access of lighting solar panel and force them to deforest the nearby shrubs (Hishe et al., 2018). Thus,

406 provision of alternative source of light and biomass energy in a long run can minimize the shrub and  
407 forest resources degradation of pastoral, agro-pastoral and farming communities.

#### 408 **4. Conclusions**

409 The arid and semi-arid climate of the present research area commonly has mean annual rainfall below  
410 700mm with dominantly low slope length and steepness. The present study indicates that rainfall  
411 erosivity (R) and slope length and steepness (LS) factors were not the main drivers of water-induced  
412 soil loss. However, the nature of the soils, land use/land covers and lack of required soil management  
413 practices were found to be the main accelerators of moderate to very high soil loss by water. To  
414 minimize water induced soil loss, the modest erodibility of the soils has to be managed through locally  
415 acceptable land management practices. To control the overgrazing of the grasslands, it is important  
416 providing public awareness to the community to transform their economies from owning too many into  
417 few livestock with a focus on quality animal breeding systems. Besides, the rangeland of the semi-arid  
418 areas of the study sites has to be used through rotation. Therefore, applications of locally fitting land  
419 management practices with the consideration of diverse strategies and other measures would minimize  
420 soil loss, enhance land quality, maximize agricultural productivity and promote the livelihood status of  
421 the local community in the study area.

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