



Article

Spatial Estimation of Soil Loss and Planning of Suitable Soil and Water Conservation Interventions for Environmental Sustainability in Northern Karnataka in India Using Geospatial Techniques

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Abstract: The selected study area lies in Karnataka State of Southern India and is frequently subjected to prolonged dry spells, high soil erosion, declining groundwater levels, and reductions in crop yield. In order to make this region sustainable, estimation of soil loss, selection and prioritization of suitable interventions, and its adoption are very important. In this study, spatial soil loss estimation models were developed sub-district-wise using the Revised universal soil loss equation (RUSLE) and GIS for a period of 70 years (1951 to 2020). The observed soil loss data for the period of 2011 to 2015 were used for validation of the model ($R^2 = 0.89$) and were found satisfactory. The average annual rainfall ranged spatially from 420 to 3700 mm, erosivity (R) ranged from 2606 to >15,000 MJ mm ha $^{-1}$ h $^{-1}$ year $^{-1}$, and average annual soil loss varied from <2.0 to >15.0 t ha $^{-1}$ y $^{-1}$ in the northern dry zone of Karnataka. Most of the study area had an average annual rainfall of 550 to 800 mm and the soil loss was <10.0 t ha⁻¹ y⁻¹. A higher erosivity and soil loss occurred in the western part of the selected area where high rainfall is predominant. A considerable variability in rainfall, erosivity, and soil loss was found in high, medium, and low-rainfall regions from 1951 to 2020. The spatial soil loss was estimated catchment-wise and prioritized to determine the vulnerable areas. It was found that 7.69% of the area with soil loss \geq 15.0 t ha⁻¹ y⁻¹ needs top priority for planning interventions (Priority 1) followed by 10.49% of the area with soil loss ranging from 10.0 to 15.0 t ha⁻¹ y⁻¹ (Priority 2) and 42.7% of the area under 5.0 to 10.0 t ha⁻¹ y⁻¹ (Priority 3), and the remaining area has lower priority. In order to make Northern Karnataka more sustainable, suitable site-specific moisture conservation practices and water-harvesting/groundwater recharge structures were planned using geospatial techniques. Among the selected moisture conservation interventions, conservation furrow and contour cultivation are very suitable for all the nine districts followed by compartmental bunding and semi-circular bunds. Out of the total area, conservation furrow was found suitable for 45.3% of the area, contour cultivation for 24.3% of the area, and compartmental bunding and semi-circular bunds for 16.8 and 16.9% of the areas, respectively. The study indicated that a considerable amount of topsoil is lost as erosion and, hence, planning and adoption of suitable in situ soil and water conservation practices and water-harvesting/groundwater recharge structures are the need of the hour for the sustainable management of this region. The identified locations were validated using visual interpretations, ground truth, and recorded data.

Keywords: erosivity; GIS; Northern Karnataka; soil loss; RUSLE; soil and water conservation



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1. Introduction

About 75 billion tons of soil is eroded in a year from the arable regions of the world, and most of the agricultural land is losing soil 13 to 40 times more quickly than the rate of

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regeneration of soil [1,2]. Even though the soil degradation process occurs naturally, human interferences accelerate its extent and its associated impacts on food production [3]. It is also well recognized as a major threat to environmental and soil biodiversity [4,5]. In India, 53.0% of the area is affected from the soil erosion with an average rate of 16.0 t ha $^{-1}$ y $^{-1}$ [6]. Soil erosion is a common type of land degradation across the globe, and for attaining a land-degradation-free world for future needs, execution of immediate action plans such as obtaining precise data of soil loss and planning and adoption of suitable management strategies for erosion control is needed [7–9]. The climate crisis will strongly affect the erosion process. Directly, rainfall erosivity is expected to increase under the future climate projection [10].

The yield of crops mainly depends on two natural resources, namely land and water, and, hence, the conservation and precise utilization of these resources are very important [11–13]. Excessive erosion causes on-site as well as off-site problems [14]. The on-site impact of erosion is the loss of the top fertile soil, which adversely affects the root growth, moisture storage, and crop yields, leads to ecological collapse, and reduces the carrying capacity and storage of streams, reservoirs, tanks, and riverbeds [15–22]. Soil erosion generally depends on different parameters such as rainfall, soil, topography, and land use/land cover [23,24]. Mechanical and vegetative measures are generally used to control erosion.

In India, out of 120.72 M ha of degraded land, 82.57 M ha is deteriorated due to water-induced erosion [25,26]. Among different Indian states, approximately 49.0% of the area of Karnataka State is impacted by erosion at a rate of >10.0 t ha⁻¹ yr⁻¹ and is in the fifth position with respect to soil erosion. The soil loss in different districts of Karnataka ranges from <5.0 to >10.0 t ha⁻¹ yr⁻¹ [25]. This is driven by natural and anthropogenic factors including a high density of population, unemployment rate, agricultural expansion, poor management practices, and climate change impacts [27]. Previous studies have reported that one-third of the districts, namely Koppal, Bagalkote, Bellary, and Dharwad, require soil loss mitigation in a phased manner, whereas two districts, namely Belgavi and Vijayapura, are worst affected and need immediate conservation [25].

The most commonly used methodologies for determination of soil erosion include runoff plots and gauging devices, remote sensing, and modeling [2,28,29]. The models widely used by researchers include the Universal Soil Loss Equation (USLE) [30–32], Revised Universal Soil Loss Equation (RUSLE) [33,34], Water Erosion Prediction Project (WEPP) [35], Kinematic Runoff and Erosion Model (KINEROS) [36], Pan-European Soil Erosion Risk Assessment (PESERA) [37], Coordination of Information on the Environment (CORINE) [38], and Erosion Potential Model (EPM) [39]. These models are applied in different regions across the world. The widely applied models for soil loss estimation include USLE, RUSLE, and the Soil Loss Estimation model of Southern Africa (SLEMSA) [40,41]. Other commonly used models are the Areal Nonpoint Source Watershed Environment Response Simulation (ANSWERS) [42], Agricultural Non-Point Source Pollution (AGNPS) [43], and Chemical, Runoff, and Erosion from Agricultural Management Systems (CREAMS) [44]. ANSWERS and CREAMS consider slope, vegetation, rainfall, soil, and erosion control methods [40].

The selected study area has a limited number of gauging stations and the availability of runoff and soil loss data is very low; hence, estimation of the soil loss is needed [42]. The RUSLE has been commonly applied for soil loss determination at the watershed and regional scale because of its easiness and compatibility with GIS. This study aims to determine the soil loss spatially with RUSLE and prioritize the vulnerable areas for site-specific management. In order to control the soil erosion and to conserve the moisture in the soil, adoption of in situ moisture conservation techniques is essential. Selection of a suitable intervention using surveys for large areas is a time-consuming task and, hence, for planning of site-specific in situ interventions, application of geospatial techniques is well suited. Very limited literature is available in this regard and there is no study carried out in Northern Karnataka. Therefore, a methodology is developed in this study to plan

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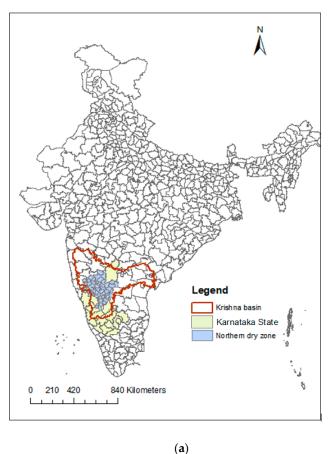
suitable soil and water conservation interventions for a large area in Northern Karnataka using geospatial techniques to minimize the soil erosion.

2. Materials and Methods

2.1. Study Area

The study area is located in Northern Karnataka, a part of the Krishna basin, which lies between $13^{\circ}~47'$ and $17^{\circ}~30'$ N and $74^{\circ}~05'$ and $77^{\circ}~36'$ E with an altitude range from 75 m to 1104 m above MSL (Figure 1). It is a dry zone and a semi-arid region including Vijayapura, Belgavi, Bagalkot, Gadag, Koppal, Bellary, Davengere, Raichur, and Dharwad districts of Karnataka covering an area of $67,884.59~{\rm km}^2$. The mean annual rainfall of these regions ranged from 420 to 3700 mm with the few sub-districts of Belgavi and Dharwad with higher rainfall.

The main soil types in this area are shallow to very shallow black soils, medium deep to deep black soils, and red soils with clayey to loamy in texture (Figure 2a). The major crops grown are green gram, pearl millet, sunflower, pigeonpea, sorghum, chickpea, and *rabi* sorghum [45]. These districts have mainly agricultural land, current fallow, waste land, scrub land, and degraded forest (Figure 2b). Most of the area in the selected districts has a slope less than 5.0% (Figure 2c).



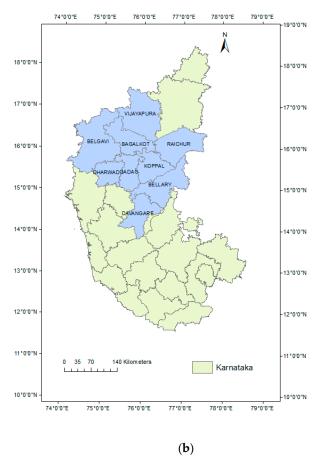


Figure 1. Cont.

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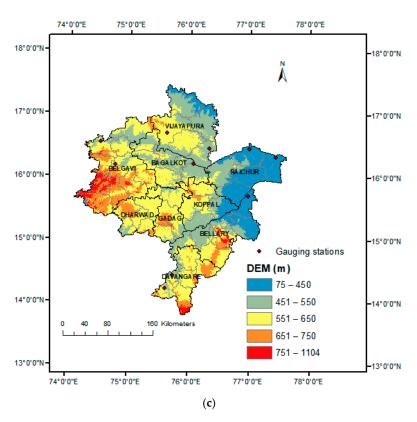


Figure 1. (a–c) Location map and gauging stations of northern dry zone of Karnataka in Krishna basin of India.

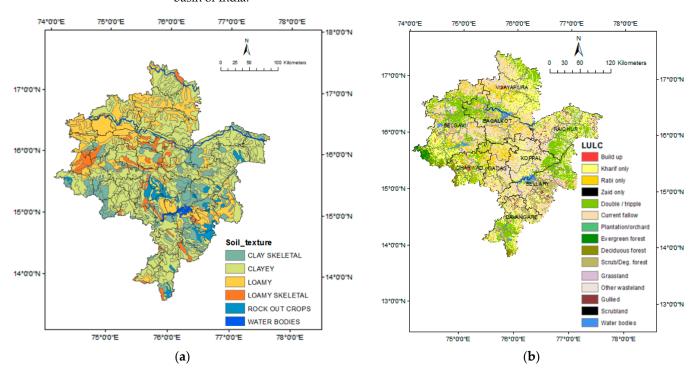


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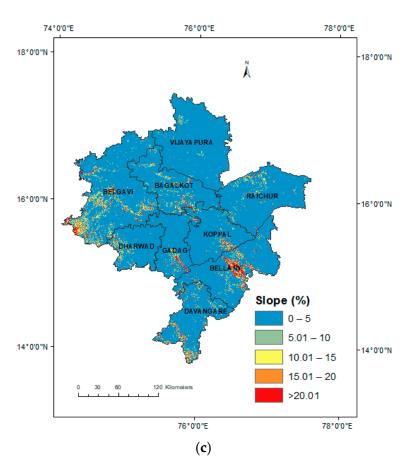


Figure 2. (a) Soil map, (b) land use land cover map, and (c) slope map of northern dry zone of Karnataka.

Different layers were prepared in GIS to estimate the soil loss. ASTER DEM (30 m resolution), the LULC map from the Remote Sensing Centre, rainfall grid data $(0.25^{\circ} \times 0.25^{\circ})$ from the Indian Meteorological Department (IMD) for the period of 1951 to 2020, and MODIS NDVI and the soil map from the National Bureau of Soil Science and Land Use Planning (NBSS and LUP) were used. The flow accumulation map, slope map, drainage lines, and stream orders were generated from DEM.

2.2. Soil Erosion Estimation Using Revised Universal Soil Loss Equation (RUSLE)

The prediction of soil erosion using several models with different techniques is used in different regions across the world. In the present study, RUSLE was used for estimating the erosion [33,34]. The mean annual soil loss in the selected sub-districts was determined spatially and temporally using the GIS and RUSLE equation (Equation (1)), which depends on rainfall, soil, LULC, and topography. The required thematic layers were intersected in GIS and soil loss was estimated.

$$A = R K L S C P \tag{1}$$

where A = mean annual soil loss (t ha⁻¹ y⁻¹); R = erosivity factor (MJ mm ha⁻¹ h⁻¹ y⁻¹) (Equation (2)); K = soil erodibility factor (t ha h ha⁻¹ MJ⁻¹ mm⁻¹); LS = slope length steepness factor (Equation (3)); C = cover management factor; P = conservation practices factor. LS, C, and P are dimensionless factors.

The erosivity factor is one of the key factors in the USLE/RUSLE models for determination of soil loss from catchments [30]. The *R* factor is associated with the kinetic energy of a raindrop's impact and is derived from the individual rainfall events, and it depicts the average annual values over multiple years. The equation for the *R* factor for daily soil loss developed at ICAR-CRIDA, Hyderabad was used [46].

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The *R* factor was estimated for the period of 1951–2020 using the following equations:

$$R = \frac{\sum_{i=1}^{n} \sum_{j=1}^{m} EI_{30}}{n} \times 1000/200.6 \tag{2}$$

where R is the mean annual erosivity (MJ mm ha⁻¹ h⁻¹ y⁻¹); i is the year; j is the days per year i; EI_{30} is the rainfall erosivity at 30 min per day (hundreds t cm ha⁻¹ h⁻¹).

$$EI_{30} = 34.065 EI_{1440} - 0.2695 (3)$$

where EI_{1440} = erosion index per day (hundreds t cm ha⁻¹ h⁻¹).

$$EI_{1440} = 3.856 PI_{1440} - 0.0048 (4)$$

where PI_{1440} = daily precipitation index, cm² h⁻¹.

$$PI_{1440} = (Rainfall)^2 / 24$$
 (5)

where Rainfall is in cm.

The *R* value corresponding to each rainfall event was added to obtain the monthly and annual erosivity for 70 years.

The *K* factor was estimated based on the soil properties such as texture, structural stability, particle size distribution, water transmission characteristics, clay mineralogy, and organic matter content [47]. Generally, the *K* value is lower for sandy and clay soils because clay soils are not easily detachable and sandy soils have less runoff. Silt loamy soils have medium to high *K* values because these soil particles are not resistant to detachment and produce higher runoff. In the present study, the value of the *K* factor selected was 0.015 [48].

The *LS* factor [30] was derived from the flow accumulation and slope [46,49] using Equation (6)

$$LS = \left\lceil \frac{\text{Flow accumulation} * \text{cell value}}{22.1} \right\rceil^m \left(0.065 + 0.045 \text{s} + 0.0065 \text{s}^2 \right)$$
 (6)

where s = slope in degrees and m is a constant depending on slope.

In the case of slopes \leq 3, m is 0.3; for slopes between 3 and 5, m is 0.4; for slopes \geq 5, m is 0.5.

The value of *C* ranges from 0 for a very well-maintained soil to 1.5 for a highly tilled and susceptible soil. In this study, the *C* factor was derived from MODIS *NDVI* using Equation (7) [50].

$$C = \exp\left(-\propto \frac{NDVI}{\beta - NDVI}\right) \tag{7}$$

where α and β are the parameters that determine the *NDVI-C* curve. The monthly images of *NDVI* were used to derive monthly *C* factors for normal and dry years during the selected period. The values of α and β were 2 and 1, respectively. The *P* factor was adopted based on the LULC and the conservation practices. The *P* factor ranged from 0 for good conservation practice to 1 for poor conservation practice.

The different thematic layers generated above were intersected in ARCGIS and soil loss was estimated spatially and temporally using RUSLE for the period of 70 years from 1951 to 2020. The observed soil loss data from the All India Coordinated Research Project for Dryland Agriculture, Vijayapura for the period from 2011 to 2015 were used for validation of the model (Figure 1c). The soil loss from high, medium, and low-rainfall regions during normal, above normal, and drought years were selected and analyzed to find the temporal variability. Among the districts selected, Belgavi was characterized with high, medium, and low-rainfall regions. The annual rainfall of these three rainfall regions was tabulated into different rainfall years. The years with an annual rainfall more than +19% were selected as above normal years, those between -9 and +19% as normal years, and those less than

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> -19% as drought years [51]. Correspondingly, the erosivity and soil loss estimated were also categorized. The statistical significance of the identified trends in rainfall, erosivity, and soil loss was carried out with the Mann-Kendall trend test and Sen's slope [52].

2.3. Prioritization of Vulnerable Areas

The spatial soil loss map and the catchments obtained using GIS were intersected and dissolved to obtain the catchment-wise soil loss. It is to be noted that the catchments with more soil loss needs top priority for taking up erosion control techniques [48]. The catchments affected with a soil loss of >15.0 t $ha^{-1}y^{-1}$ need top priority for implementing erosion control measures followed with areas having a soil loss ranging from 10.0 to $15.0 \text{ t ha}^{-1} \text{ y}^{-1}$ and then from 5 to 10.0 t ha⁻¹ y⁻¹. Regions with a soil loss less than $5.0 \text{ t ha}^{-1} \text{ y}^{-1}$ also need small interventions for controlling erosion.

2.4. Determination of Location-Specific Interventions

To control the soil erosion from different vulnerable districts in the selected area, suitable sites for various soil and water conservation interventions were determined using geospatial techniques [53].

Thematic layers of slope, soil, rainfall, runoff potential, and LULC were intersected using GIS, and the criteria corresponding to selected interventions were used in GIS for finding the potential locations for various in situ interventions such as conservation furrow, adjusted contour/graded bunds, compartmental bunding, contour cultivation, small pits, broad bed furrow (BBF), ridge and furrow, and semi-circular bunds (Figure 3a and Table 1). Similarly, suitable locations for water-harvesting structures were also identified using the criteria provided in Table 2 [54] in GIS. The identified locations were exported to Google Earth and validated by recorded data, visual interpretations, as well as using ground truth.

Table 1. Criteria for planning selected in situ soil and water conservation interventions.

Interventions	Slope (%)	Soil Type	Rainfall (mm)	Soil Depth (cm)
Ridge and furrow	2-15 ^(b)	Loamy/loamy skeletal	>350 (c) and <1000	>50 (field crops)
Semi circular bunds	5–15 ^(b)	exclude sandy soil ^(b)	>200 and <4000 ^(b)	100–150 ^(b) (tree crops)
Small pits	2–10 ^(b)	exclude sandy soil (b)	>350 and <4000	
Broad Bed Furrow (BBF)	≤3 ^(e)	clayey and loamy soil	>750	100–150 (field crops)
Compartmental bunding	≤1 ^(e)	clayey soil ^(e)	>400 and < 750	>50 ^(e) (field crops)
Conservation furrow	≤10 ^(c)	exclude sandy soil ^(b)	≤1500	<100 (field crops)
Contour/graded bunds	1–6 ^(a)	exclude deep clayey soil ^(a)	>200 and <600 ^(c)	<100 (c) (field and tree crops)
Contour cultivation	≤5 ^(d)	exclude sandy soil	>350 and <4000	>100 ^(d) (field and tree crops)

Note(s): Source: Rejani et al. [53]; Rejani et al. [54]; a Shanwad et al. [55]; b Pauw et al. [56]; c Anschütz et al. [57]; ^d Kalgapurkar et al. [58]; ^e TNAU [59]; Rejani et al. [60].

Table 2. Site selection criteria for the planning of water-harvesting/groundwater recharge structures.

Structure	Slope (%)	Stream Order	Catchment Area (ha)	Annual Rainfall (mm)
Farm ponds (lined/unlined)	≤5	1–2 and other potential area	>1-2	>500
Check dams	≤15	3–4	25	>700
Percolation tanks (Light sandy soil)	≤10	1–4	25–40	>700

Note(s): Source: Rejani et al. [53]; Rejani et al. [54]; Rejani et al. [60].

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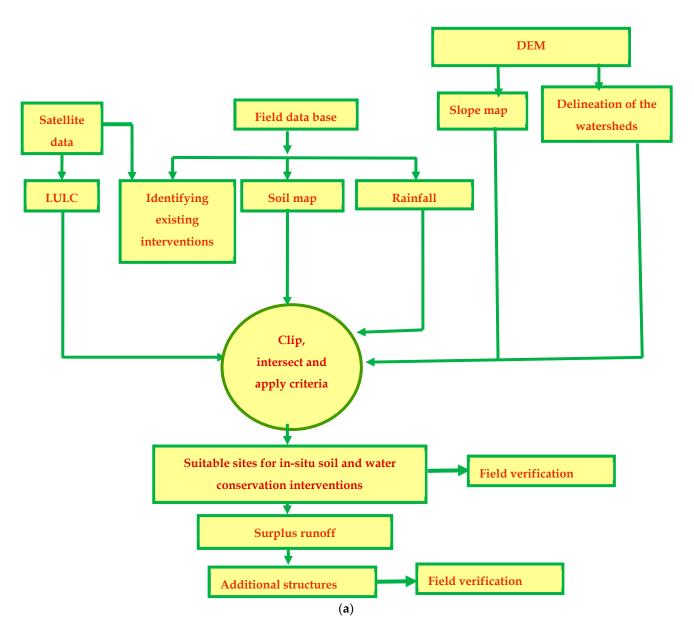


Figure 3. *Cont.*

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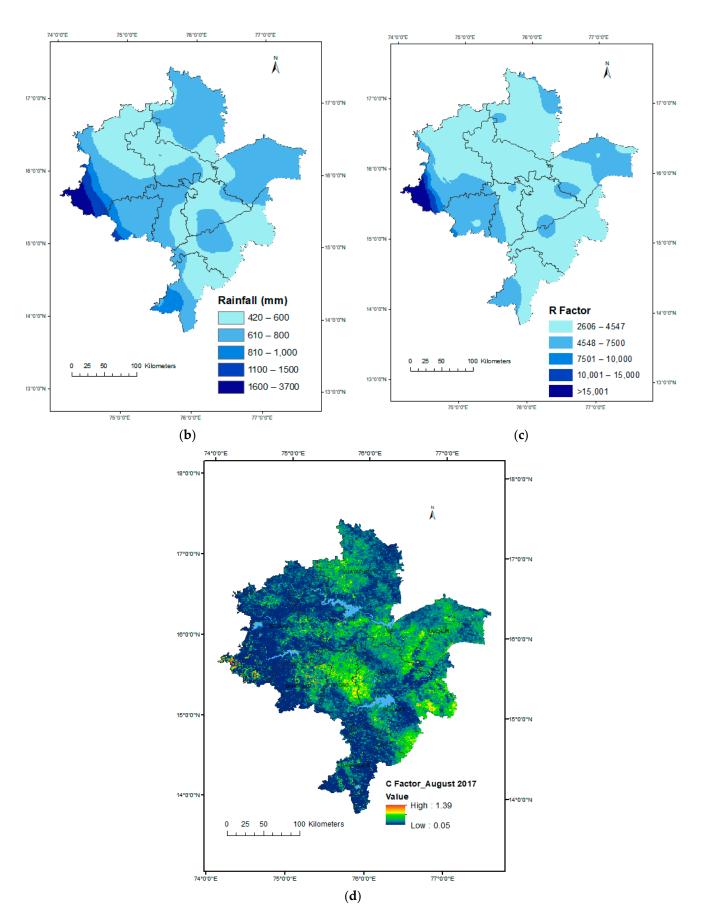


Figure 3. (a). Flow chart depicting the planning of soil and water conservation interventions. (b,c) Rainfall and erosivity (*R* factor) in the study area. (d) *C* factor map for the selected area during August 2017.

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3. Results

3.1. Rainfall, Erosivity, and Soil Loss in Northern Dry Zone of Karnataka

Rainfall and erosivity maps were generated from 148 grids of rainfall data (IMD) for 70 years from 1951 to 2020. The average annual rainfall varied spatially from 420 to 3700 mm and the R factor ranged from 2606 to >15,000 MJ mm ha $^{-1}$ h $^{-1}$ year $^{-1}$ in the northern dry zone of Karnataka (Figure 3b,c). The years were classified into above normal, normal, and dry years based on rainfall and the corresponding C factor was used (Figure 3d). Soil loss was estimated spatially using RUSLE and GIS. The observed soil loss data for the period 2011 to 2015 were used for validation of the model (Figure 1c). A linear regression analysis of the measured and estimated soil loss was carried out and an R² value of 0.89 was obtained, which shows a good match between the observed and simulated values (Figure 4a). Hence, this model could be successfully used for estimating the soil loss. The mean annual soil loss estimated varied from <2.5 t ha $^{-1}$ y $^{-1}$ to >15.0 t ha $^{-1}$ y $^{-1}$ and a major portion of the area (81.8%) had soil loss <10.0 t ha $^{-1}$ y $^{-1}$ (Figure 4b and Table 3).

Erosion Category	Area under Each Category (ha)	% Area under Each Category	Rate of Soil Loss (t $ha^{-1}y^{-1}$)
Low erosion	Low erosion 783,537		≤2.5
Slight erosion	1,851,847	27.47	2.51 to 5.0
Moderate erosion	2,880,111	42.72	5.01 to 10.0
High erosion	707,434	10.49	10.01 to 15.0
Very high erosion	212,607	3.15	15.01 to 20.0
Severe erosion	306,089	4.54	>20.01

Table 3. Quantification of average annual soil loss at northern dry zone of Karnataka (1951 to 2020).

3.2. Variability in Rainfall, Erosivity, and Soil Loss in Different Rainfall Regions

The temporal variability in rainfall, erosivity, and soil loss from high-, medium-, and low-rainfall areas of Belgavi district during above normal, normal, and drought years were assessed separately for the period of 1951 to 2020. In the case of high-rainfall areas, the annual rainfall for the period of 1951 to 2020 varied from 845 mm in 1995 to 3505 mm in 2019 with a mean annual value of 1783 mm. The erosivity varied from 4594 in 1987 to 73,407 in 2019 with a mean of 18,630 MJ mm ha $^{-1}$ h $^{-1}$ year $^{-1}$ (Figure 5a). Even though rainfall was lowest in 1995, erosivity was lower in 1987 due to the low-intensity daily rainfall pattern. The topography of this region is undulating with a slope of more than 7.5% and LS factor of >2. The annual soil loss varied from 4.4 t in 1987 to 30.2 t in 2019 with a mean value of 17.9 t ha $^{-1}$ y $^{-1}$ (Figure 6).

The medium-rainfall area had rainfall ranging from 302 mm in 1985 to 1288 mm in 2019 with a mean annual rainfall of 742 mm. Similarly, the R factor varied from 746 in 1985 to 11,090 in 2019 with a mean of 5084 MJ mm ha $^{-1}$ h $^{-1}$ year $^{-1}$ (Figure 5b). The slope was more than 5% with an LS factor of >0.5. Correspondingly, the soil loss varied from 1.1 t in 1985 to 16.3 t in 2019 with a mean value of 7.3 t ha $^{-1}$ y $^{-1}$ (Figure 6).

In the case low-rainfall regions, rainfall ranged from 260 mm in 1997 to 1084 mm in 1962 with a mean of 521 mm, and the erosivity factor varied from 838 in 2003 to 13,608 in 2009 with a mean R factor of 3737 MJ mm ha $^{-1}$ h $^{-1}$ year $^{-1}$ (Figure 5c). These regions are generally flat with a slope less than 5% and an LS factor of <0.5. The annual soil loss from these low-rainfall regions was much less and varied from 0.1 t in 2003 to 1.4 t in 2009 with a mean soil loss of 0.4 t ha $^{-1}$ y $^{-1}$ (Figure 6). It is visible that considerable variability in rainfall, erosivity, and soil loss exists in high-, medium-, and low-rainfall regions in Belgavi district. Hence, the determination of soil loss spatially for planning and prioritizing interventions is very important for such areas.

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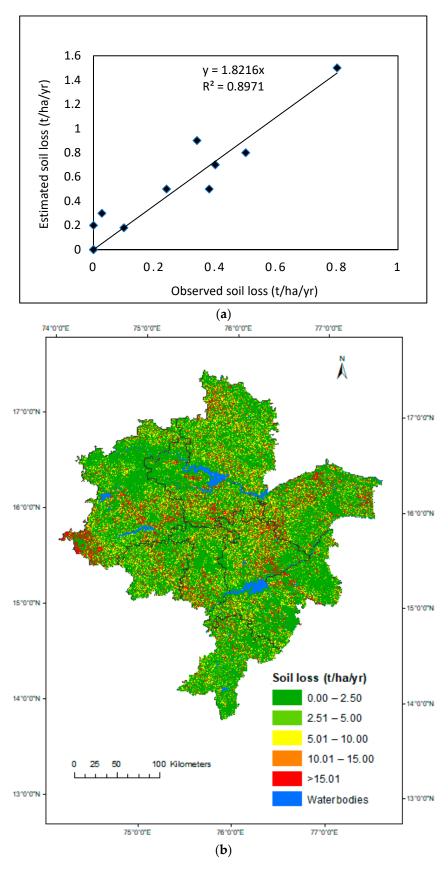
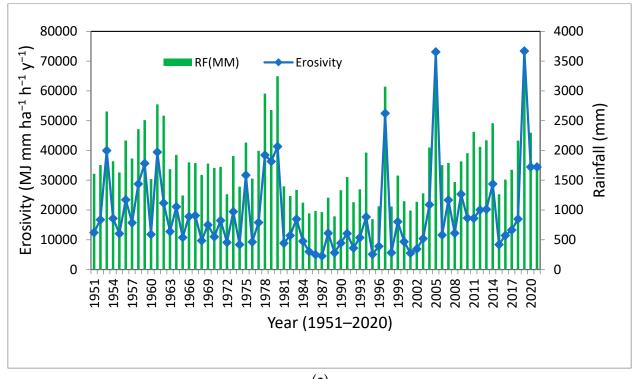


Figure 4. (a). Observed vs. estimated soil loss in northern dry zone of Karnataka. (b). Mean annual soil loss in northern dry zone of Karnataka.

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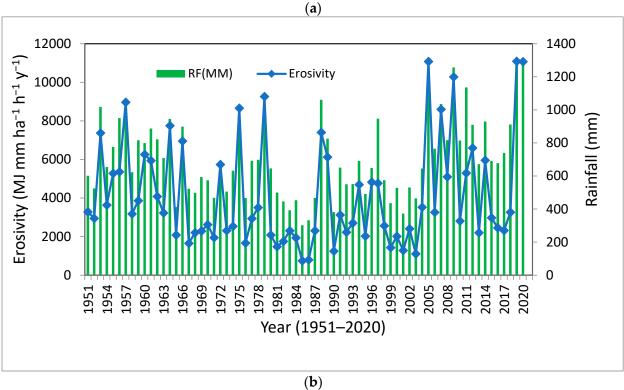


Figure 5. Cont.

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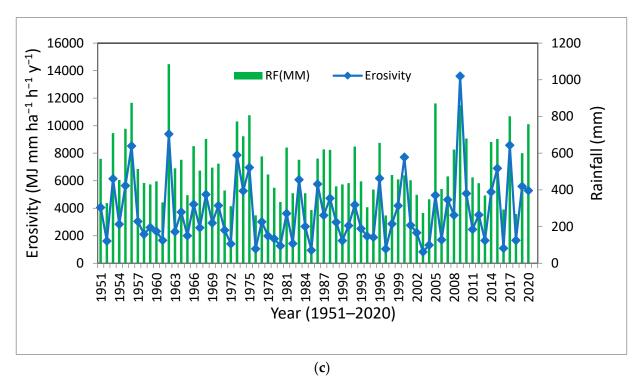


Figure 5. (a). Variation in rainfall and erosivity in high-rainfall area. (b). Variation in rainfall and erosivity in medium-rainfall area. (c). Variation in rainfall and erosivity in low-rainfall area.

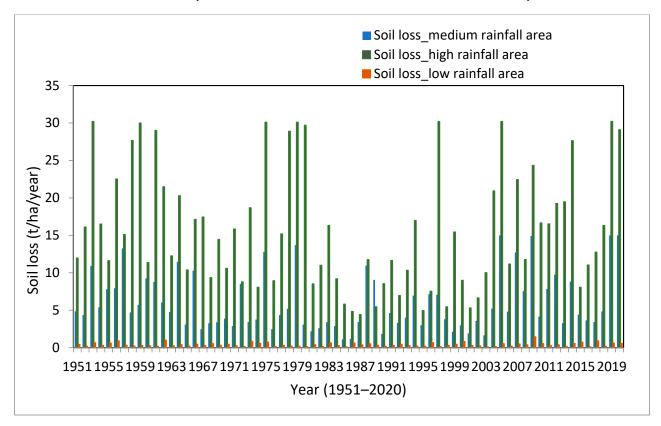


Figure 6. Temporal variation in soil loss at low, medium and high-rainfall areas of Belgavi district.

The rainfall, erosivity, and soil loss were also analyzed separately for high-, medium-, and low-rainfall regions of Belgavi district during different rainfall years. Out of 70 years, low-rainfall regions experienced drought for 31% of the years with a rainfall of 260

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to 419 mm, erosivity of 1047 to 2993 MJ mm ha^{-1} h^{-1} year⁻¹, and soil loss of 0.1 to 0.3 t ha^{-1} y^{-1} , and 39% of the years were normal with a rainfall of 430 to 618 mm, erosivity of 2608 to 4734 MJ mm ha^{-1} h^{-1} year⁻¹, and soil loss of 0.3 to 0.5 t ha^{-1} y^{-1} (Table 4). The remaining 30% of the years were above normal with rainfall varying from 619 to 1085 mm, erosivity varying from 3504 to 9396 MJ mm ha^{-1} h^{-1} year⁻¹, and soil loss varying from 0.37 to 0.99 t ha^{-1} y^{-1} .

Table 4. Temporal variation in rainfall, erosivity, and soil loss for high, medium and low-rainfall
regions of Belgavi district during drought, normal, and above normal years during 1951 to 2020.

Parameters	Low-Rainfall Region			Mediu	Medium-Rainfall Region			High-Rainfall Region		
	Drought Year	Normal Year	Above Normal Year	Drought Year	Normal Year	Above Normal Year	Drought Year	Normal Year	Above Normal Year	
% of Years (1951 to 2020)	31	39	30	36	37	27	33	41	26	
Mean annual rainfall (mm)	260 to 419	430 to 618	619 to 1085	302 to 574	594 to 826	886 to 1288	846 to 1397	1469 to 2060	2131 to 3505	
Mean erosivity (MJ mm $ha^{-1} h^{-1}$ $year^{-1}$)	1047 to 2993	2608 to 4734	3504 to 9396	746 to 2563	2290 to 6122	3267 to 11,090	5147 to 8860	9290 to 25,327	16,698 to 73,407	
Mean annual soil loss (t ha ⁻¹ year ⁻¹)	0.1 to 0.32	0.3 to 0.5	0.37 to 0.99	1.1 to 3.78	3.4 to 6.5	6.0 to 11.2	4.9 to 8.5	8.0 to 15.0	16.3 to 30.2	

In medium-rainfall regions, out of 70 years, 36% of years were drought-predominant with rainfall from 302 to 574 mm, erosivity from 746 to 2563 MJ mm ha $^{-1}$ h $^{-1}$ year $^{-1}$, and soil loss from 1.1 to 3.78 t ha $^{-1}$ y $^{-1}$, and 37% of the years were normal with rainfall from 594 to 826 mm, erosivity from 2290 to 6122 MJ mm ha $^{-1}$ h $^{-1}$ year $^{-1}$, and soil loss from 3.4 to 6.5 t ha $^{-1}$ y $^{-1}$. The remaining 27% of the years were above normal with rainfall varying from 886 to 1288 mm, erosivity from 3267 to 11,090 MJ mm ha $^{-1}$ h $^{-1}$ year $^{-1}$, and soil loss from 6.0 to 11.2 t ha $^{-1}$ y $^{-1}$ (Table 4).

In the case of high-rainfall regions, 33% of the years were drought years with rainfall varying from 846 to 1397 mm, erosivity from 5147 to 8860 MJ mm ha $^{-1}$ h $^{-1}$ year $^{-1}$, and soil loss from 4.9 to 8.5 t ha $^{-1}$ y $^{-1}$, and 41% of the years were normal with rainfall varying from 1469 to 2060 mm, erosivity varying from 9290 to 25,327 MJ mm ha $^{-1}$ h $^{-1}$ year $^{-1}$, and soil loss varying from 8.0 to 15.0 t ha $^{-1}$ y $^{-1}$. The remaining 26% of the years were above normal with rainfall varying from 2131 to 3505 mm, erosivity varying from 16,698 to 73,407 MJ mm ha $^{-1}$ h $^{-1}$ year $^{-1}$, and soil loss varying from 16.3 to 30.2 t ha $^{-1}$ y $^{-1}$ (Table 4). The Mann–Kendall trend test and Sen's slope showed no significant trend for rainfall, erosivity, and soil loss, but a slight increase in rainfall, erosivity, and soil loss was observed in low-, medium-, and high-rainfall regions during the 70-year period.

3.3. Prioritization of Vulnerable Areas

The spatial soil loss map after intersecting with the catchments in GIS was dissolved to obtain the catchment-wise soil loss. The catchments affected with more soil loss of >15.0 t ha⁻¹ y⁻¹ needs special priority for implementation of soil erosion control measures followed with areas with soil loss >10.0 to 15.0 t ha⁻¹ y⁻¹ (Figure 7a). Around 7.69% of the area with soil loss \ge 15.0 t ha⁻¹ y⁻¹ (Priority 1) and 10.49% of the area with soil loss ranging from 10.0 to 15.0 t ha⁻¹ y⁻¹ (Priority 2) need top priority followed by 42.72% of the area under 5.0 to 10.0 t ha⁻¹ y⁻¹ category (Priority 3) (Table 3). It was found that 39.09% of the area with slight erosion with soil loss <5 t ha⁻¹ y⁻¹ (Figure 7b) also needs interventions for controlling erosion (Priority 4).

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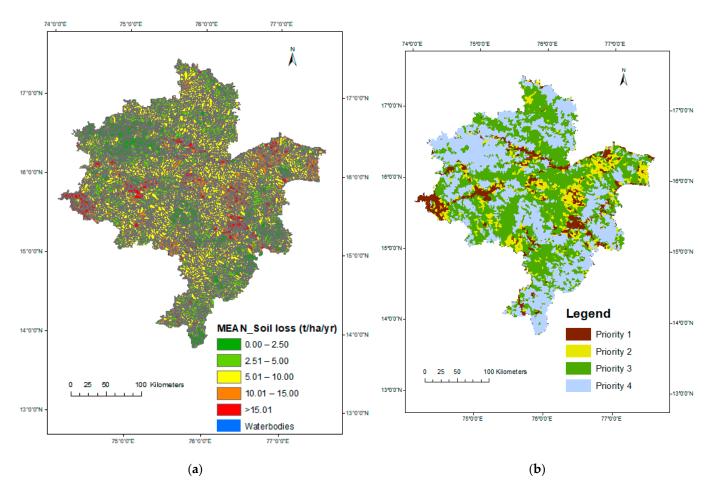


Figure 7. (a). Catchment-wise soil loss in the study area. (b). Prioritization for implementing soil erosion control measures.

3.4. Suitable Locations for In Situ Moisture Conservation Measures and Water-Harvesting Structures

Much research has been performed in the past and has developed several soil and water conservation practices for Northern Karnataka. Out of those practices, the practices that were found better in terms of reducing soil and water loss and producing higher crop yields in multi-location trials were selected and recommended for the farmers to implement. Those practices being followed by farmers in the region were selected as a suitable practice for this specific region. The scope for expanding these practices/interventions for the whole region can be achieved using geospatial techniques.

The major soil and water conservation interventions generally followed in the study area includes broad bed furrow (BBF), compartmental bunding, conservation furrow, ridge and furrow, contour bunds, graded bunds, and semi-circular/crescent bunds (Figure 8a–f. Singh et al. [61] reported the impact of life-saving irrigation from water harvested in farm ponds, check dams, and community ponds for saving the crops from dry spells of more than 20 days in Northern Karnataka. It increased the crop yields by 25 to 54% more as compared to unirrigated fields. Similarly, the rejuvenation of percolation tanks and nala bunds under ICAR-NICRA enhanced the water storage and increased the groundwater recharge, which enhanced the groundwater level in nearby wells used for critical irrigation. In Northern Karnataka, adoption of trench-cum-bunds enhanced the soil moisture and increased the yield of pigeonpea, groundnut, and maize by 29, 25, and 31%, respectively. Similarly, field bunds/contour bunds in Kalburgi district minimized the erosion and increased the yield of pigeonpea by 2.0 q/ha, and compartmental bunding in Gadag, Belgavi, and Kalaburgi districts of Karnataka harvested the rainfall, conserved the moisture, increased the rainwater use efficiency, and increased the sorghum yield by 21 to 36%. Broad bed

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furrow (BBF) for groundnut in Chickaballapur district of Karnataka increased the rainwater use efficiency and economic water use efficiency by 18.7 to 85.5% as compared to flatbed sowing. Ridge and furrow for pigeonpea in Kalburgi of Karnataka saved the crops from 15 days of dry spells and increased pigeonpea yield by 3 q/ha as compared to farmers practices, whereas adoption of conservation furrow for green gram pigeonpea and maize pigeonpea intercropping in Gadag district of Karnataka increased the yields by 25 to 38% more as compared to flatbed sowing or farmers practices [61].



Figure 8. Soil and water conservation interventions existing in Karnataka. (a) Broad bed furrow (BBF), (b) compartmental bunding, (c) conservation furrow, (d) contour bunds, (e) graded bunds, and (f) semi-circular/crescent bunds.

In order to expand the area under these interventions and to make the Northern Karnataka zone more sustainable, suitable in situ moisture conservation measures and water-harvesting structures were planned and prioritized using geospatial techniques. Among the selected in situ soil and water conservation interventions, conservation furrow

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and contour cultivation are very suitable for all nine districts followed by compartmental bunding and semi-circular bunds (Table 5).

Table 5. Prioritization	of different in situ in	terventions for northern of	ry zone of Karnataka
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			Suitable Area (% of District Area) for Different In Situ Interventions								
Districts	District Area	Small Pits	Adjusted Contour/Gra- ded Bunds	Ridge and Furrow	Compart- mental Bunding	Contour Cultiva- tion	Semi- Circular Bunds	Conser- vation Furrow	Broad Bed Furrow (BBF)		
Belgavi	1,340,860.8	6.1	15.5	25.2	10.9	19.1	10.2	35.2	6.1		
Davengere	595,480.6	1.8	2.2	4.2	19.8	35.9	8.9	61.2	5.2		
Bellary	845,086.3	6.8	7.9	10.3	18.7	17.3	14.6	43.8	0.0		
Vijayapura	1,052,237.9	4.6	3.1	8.6	8.6	16.3	31.7	47.4	24.8		
Raichur	845,494.5	14.2	0.6	1.6	26.4	33.6	13.0	61.6	0.0		
Koppal	557,498.7	13.8	3.5	5.6	22.9	24.0	22.0	55.6	0.0		
Dharwad	428,481.3	5.1	0.0	1.4	22.7	40.1	11.0	76.5	8.2		
Gadag	465,183.0	7.7	3.6	10.5	19.2	30.3	25.3	51.1	1.3		
Bagalkot	658,136.8	9.0	18.6	24.8	14.2	19.6	15.8	36.0	0.0		
Total	678,8459.8	7.5	7.1	11.8	16.8	24.3	16.9	49.2	6.1		

Out of the total area, conservation furrow was found suitable for 49.2% of the area, contour cultivation for 24.3% of the area, and compartmental bunding and semi-circular bunds for 16.8 and 16.9% of the areas, respectively. BBF was found to be suitable for 24.8% of Vijayapura district and the ridge and furrow technique for 25.2% of Belgavi and 24.8% of Bagalkot districts. The suitability of conservation furrow varied from 36.0% in Bagalkot to 61.2% in Davengere, whereas contour cultivation varied from 16.3% in Vijayapura to 40.1% in Dharwad. Semi-circular bunds were found as preferable options for Vijayapura and Gadag districts, whereas compartmental bunding was preferable for Raichur, Koppal, and Dharwad districts. It was found that the ridge and furrow technique was suitable for 11.8% of the total area. Adjusted contour/graded bunds with terraces were suitable in 18.6% of Bagalkot, 15.5% of Belgavi, and 7.9% of Bellary area (Figure 9a) (Table 5). Further, suitable locations for water-harvesting/groundwater recharge structures such as farm ponds, percolation tanks, and check dams were determined using the stream order generated using GIS and the criteria available in Table 2 (Figure 9b,c) [54]. The identified locations were converted to *kml file and validated in Google Earth by visual interpretations, ground truth, and recorded data (Figure 9d and Table 6).

Table 6. Validation of different in situ interventions in different villages of northern dry zone of Karnataka.

SI.No.	Name of the Village	Latitude	Longitude	Soil and Water Conservation Interventions Implemented
1.	Vijayapura	16° 49′	75° 43′	In situ moisture conservation measures (compartment bunds, BBF, ridges and furrows, contour and graded bunds), farm ponds and percolation tanks
2.	Kavalagi, Vijayapura	16° 48′	75° 45′	In situ moisture conservation measures (compartment bunds, BBF, ridges and furrows), farm ponds
3.	Honnutagi, Vijayapura	16° 45′	75° 50′	In situ conservation measures (compartment bunds, BBF, furrows and ridges, graded bunds), farm ponds
4.	Hegdyal, Vijayapura	16° 44′	75° 49′	In situ moisture conservation measures (compartment bunds, BBF, ridges and furrows), farm ponds

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Table 6. Cont.

SI.No.	Name of the Village	Latitude	Longitude	Soil and Water Conservation Interventions Implemented
5.	Honwad, Vijayapura	16° 48′	75° 25′	In situ moisture conservation measures (compartment bunds, BBF, ridges and furrows, graded bunds), farm ponds
6.	Hullur, Gadag	15° 44′	75° 40′	In situ moisture conservation measures (compartment bunds, BBF, ridges and furrows, graded bunds), farm ponds

Note(s): Source, AICRPDA Centre, Vijayapura.

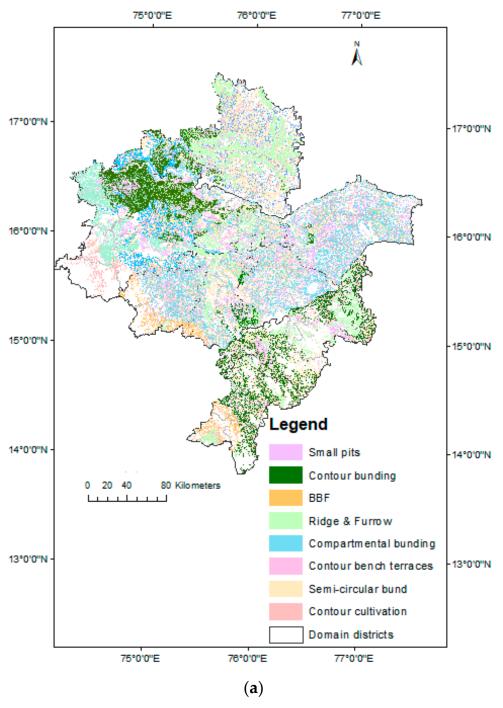


Figure 9. Cont.

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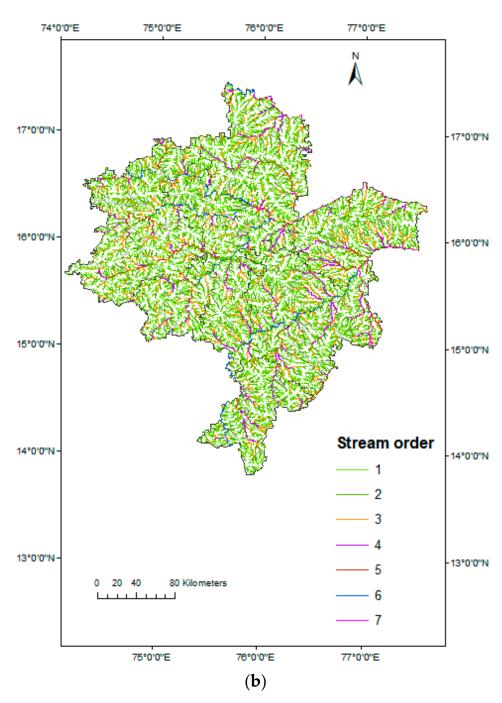


Figure 9. Cont.

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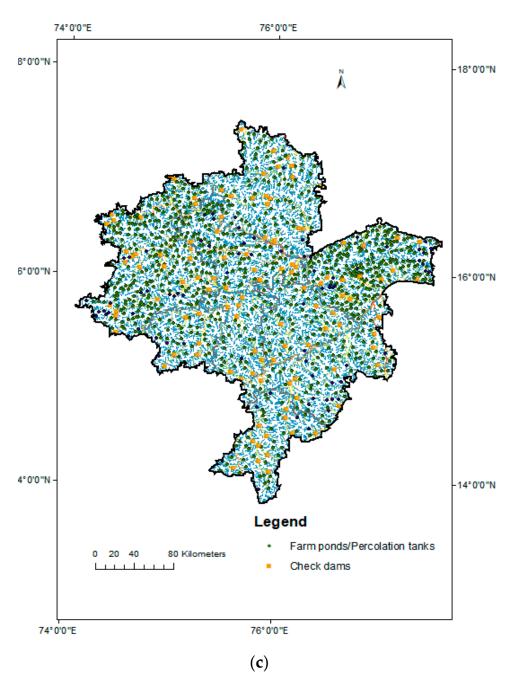


Figure 9. Cont.

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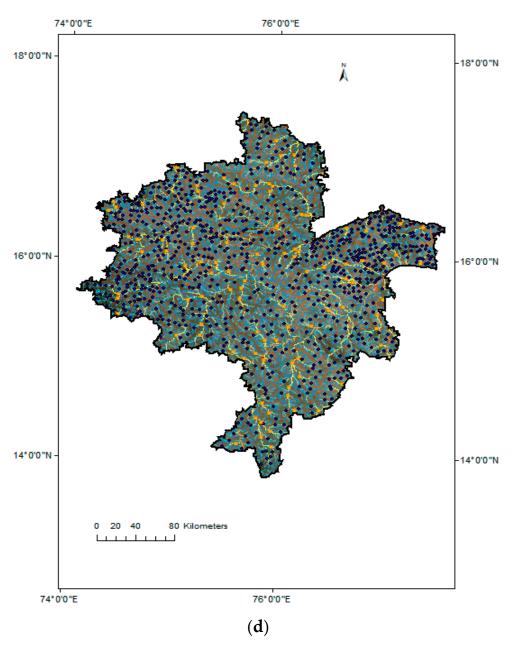


Figure 9. (a) Suitable in situ interventions identified for northern dry zone of Karnataka. (b) Stream order map of the study area. (c) Potential locations for water-harvesting/recharge structures identified for northern dry zone of Karnataka. (d) Validation of potential locations for water-harvesting/recharge structures.

4. Discussion

4.1. Soil Loss Estimation Using RUSLE

In the present study, the RUSLE model and GIS were applied to determine the soil loss in the northern dry zone of Karnataka. RUSLE helps to predict the soil erosion from ungauged watersheds at a reasonable cost and better accuracy by considering the climatic conditions and heterogeneity of the soil. Though USLE/RUSLE is an empirical model, it is the model used worldwide in soil loss estimations because of its compatibility with GIS. The integrated application of remote sensing, GIS, and RUSLE makes soil loss estimation and its spatial distribution easier with better accuracy even in large catchments [6,42,62].

Generally, researchers use empirical equations with monthly or annual rainfall for estimating the erosivity factor [63–65]. Reddy et al. [48] used empirical equations with annual rainfall for estimating the R factor for Andhra Pradesh and Telangana in India. Demirci

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and Karaburun [63] used an empirical formula with monthly and annual average rainfall to estimate the R factor for Beiluo River Basin and the Buyukcekmece Lake watershed of northwest Turkey, respectively. Mohammed et al. [64] also used a similar empirical formula with monthly and annual average rainfall (216 to 541 mm) and determined the erosivity for Southern Syria as 374.1 ha $^{-1}$ h $^{-1}$ year $^{-1}$, and soil loss was less than 5.0 t ha $^{-1}$ y $^{-1}$ for 95% of the area. Singh et al. [12] determined the annual erosivity for Maharashtra State of India from empirical equations using annual rainfall and found that the erosivity ranged from 301.5 to 1509.6 MJ mm ha $^{-1}$ h $^{-1}$ year $^{-1}$ in 2011, and 80% of the area had a soil loss less than 10.0 t ha $^{-1}$ y $^{-1}$. In the present study, the R factor was estimated using empirical equations with daily rainfall developed at ICAR-CRIDA [46], the mean annual erosivity for 70 years varied spatially from 2606 to >15,000 MJ mm ha $^{-1}$ h $^{-1}$ year $^{-1}$, and the average annual rainfall varied spatially from 420 to 3700 mm.

Many researchers used the C factor derived from NDVI for calculating the soil loss similar to the present study [66–69]. Toubal et al. [70] used RUSLE and found that the average annual soil loss in the Wadi Sahouat basin of North-West Algeria ranged from 0 to 255 t ha⁻¹ y ⁻¹. The higher values were found in steep slopes greater than 25% and elevations from 600 and 1000 m. Erosion becomes more intense corresponding to the rainfall intensity and slope characteristics, which complies with the present study. Kumar et al. [71] used remote sensing and geographic information system based RUSLE modelling for estimation of soil loss in western Himalayas. Researchers proposed several conservation measures such as terraces, gabion [72], conservation tillage, contour ploughing, cover crops, and afforestation for enhancing productivity [73]. These soil conservation measures aim to reduce the runoff, trap topsoil, promote the formation of natural terraces, and finally reduce the sedimentation in drains and waterbodies [63,73]. Moisa et al. [74] estimated the annual soil loss using RUSLE for a Sub-basin in Western Ethiopia, and it ranged from 0 and 932.6 t ha $^{-1}$ y $^{-1}$ with a mean value of 83.7 t ha $^{-1}$ y $^{-1}$ and categorized 43.6% and 8.4% of the area as very severe and severe erosion type. Wang et al. [75] applied RUSLE to assess the erosion risk in mountain areas and its response to climate change in a region located in the southern side of Tibetan Plateau, determined the average soil loss as 29.1 t ha⁻¹ y $^{-1}$, and found a higher erosion in wet and cold years. In the present study, it was found that 39.09% of the area had a slight erosion of <5 t ha⁻¹ y⁻¹ followed by 42.72% of the area under 5.0 to 10.0 t ha⁻¹ y⁻¹, 10.49% of the area with 10.0 to 15.0 t ha⁻¹ y⁻¹, and 7.69% of the area with soil loss \geq 15.0 t ha⁻¹ y⁻¹, and the study area covers low-, medium-, and high-rainfall regions with flat, medium, and high slopes. Based on the variation in soil loss, the vulnerable areas were prioritized for implementing the soil and water conservation interventions.

4.2. Planning Site-Specific Soil and Water Conservation Interventions

The planning and adoption of site-specific soil and water conservation practices in different sub-districts will help to control the runoff, soil loss, and nutrient loss from agricultural land, therefore minimizing the land degradation. This will lead to the conservation of moisture in the field and finally enhance the crop productivity. Previous research studies across the world proved that adoption of in situ interventions such as ridge and furrow, BBF, contour cultivation, compartmental bunding, and conservation furrow decreases the runoff velocity, enhances soil moisture, and recharges the groundwater [12,64,75]. Milkias et al. [76] reported that ridge and furrow, contour ridge, and tied ridge increased the soil moisture by 134.5, 128.5, and 121.8% and the grain yield of maize by 143.1, 131.4, and 121.1%, respectively, over the control treatment at Ethiopia. The studies by Qin et al. [77] showed that the integrated use of ridge and furrow planting with manure in northwest China increased the maize yield by 18% compared to flat planting. The prioritization of catchments based on soil loss will help the planners and policymakers to identify the erosion-prone areas for implementing the interventions on a priority basis [46,48], and the spatial maps of site-specific interventions generated in this study would enable the stakeholders to choose the best interventions for adoption. Many researchers used geospatial

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techniques for identifying suitable locations for water-harvesting and recharge structures with the aid of geospatial techniques [54,78,79], but limited studies are available for planning in situ moisture conservation interventions using these geospatial techniques [56]. Pauw et al. [56] integrated Expert Knowledge with GIS to find the potential for water harvesting. Raghavan et al. [80] estimated the runoff potential in Northern Karnataka using SCS-CN and GIS and classified the area based on runoff under the current scenario and climate change scenarios. Rejani et al. [53] developed a methodology for planning different in situ interventions suitable for Adilabad District of Telangana State using geospatial techniques. As the interventions are site-specific, a similar approach with different criteria was used in the present study to derive the maps of in situ soil and water interventions suitable for the sustainability of the northern dry zone of Karnataka.

5. Conclusions

Estimation of soil loss and planning of suitable interventions for in situ moisture conservation are of prime importance for the resilience of Northern Karnataka. Due to the insufficient data availability, soil loss was estimated sub-district-wise with the aid of RUSLE and GIS. For most of the region, soil loss was <10.0 t ha $^{-1}$ y $^{-1}$ and was estimated catchment-wise to prioritize the vulnerable areas. The results showed that 7.69% of the area with soil loss ≥ 15.0 t ha $^{-1}$ y $^{-1}$ needs top priority for planning interventions followed by 10.49% of the area with soil loss ranging from 10.0 to 15.0 t ha $^{-1}$ y $^{-1}$ and 42.72% of the area with 5.0 to10.0 t ha $^{-1}$ y $^{-1}$. In order to control erosion and to conserve the topsoil, suitable in situ moisture conservation measures and rainwater-harvesting/groundwater recharge structures were planned using geospatial techniques. Interventions such as conservation furrow and contour cultivation are very suitable for all the selected districts followed by compartmental bunding and semi-circular bunds. Similarly, adoption of suitable structures such as percolation tanks, farm ponds, and check dams is also essential for the sustainable water management of Northern Karnataka.

Author Contributions: R.R.: methodology and writing the manuscript draft; K.V.R.: interpreting the results; M.S.S., G.R.C., and K.A.G.: data collection and survey; K.S.R., M.O., M.P., and V.K.S.: supervised the study, and reviewed and edited the manuscript. All authors have read and agreed to the published version of the manuscript.

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Data Availability Statement: The datasets used and analyzed during the present study are available from the first author on reasonable request.

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Conflicts of Interest: The authors declare no conflict of interest.

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