

Research Article

Predicting & Analysing Discharge and Sediment of Sore Catchment, Baro-Akobo Sub Basin Using Saturation Excess Model, Western Ethiopia

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Abstract

Soil erosion, land degradation, and soil loss are major problems in Ethiopia's Baro-Akobo Basin. In the Ethiopian highlands, many hydrological models have been employed to forecast sediment and discharge in watersheds of varying sizes. PED mode was used for this study because it requires fewer data and is well recognized by academics in Ethiopia's Upper Blue Nile. The parameter efficient semi-distributed watershed model (PED) was used to study the prediction of discharge and sediment in the Sore watershed. The objective of this study was to analyze and forecast the sediment and discharge from the Sore catchment in the western Ethiopian region of the Baro-Akobo Sub basin. The selected watershed was Sore (1665.5 Km²) in the Baro-Akobo River Basin. The stream flow and sediment data for the Sore watershed from 2005 to 2013 and 2014 to 2018 were used for the model's calibration and validation by the Ministry of Water and Electricity. For the Sore watershed, the daily time step scale model efficiency of PED-W was found to be NSE= 0.40 and NSE= 0.44 for the calibration of sediment and discharge, respectively. Similarly, NSE values for the Sore watershed's NSE= -0.82 and PED-W's discharge and sediment during the validation period were found. As a result, there was a range of uneven to acceptable agreement between the simulated and observed discharge and sediment at daily time steps. In comparison to the daily time step scale, the models' performance varied and performed better on the monthly time step scale. According to the overall model performance, the PED-W model was shown to be less accurate than the SWAT model for estimating stream flow and sediment output. The PED model, on the other hand, was exactly the same as the previously employed SWAT model and was used for calibration and validation at monthly time steps.

Keywords

Hydrological Model, PED-WM, Sore, Baro Akobo Basin

1. Introduction

The Nile Basin is one of the largest and most complex river basins in the world because of its vastness and diverse climatic, hydrological, topographical, and demographic condi-

tions. Given the rapid population growth, urbanization, industrialization, both economic and food crises, climate change, and its possible impact on water resources in the Nile Basin, a

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great number of water resource development projects are under planning and implementation on both national and trans boundary levels within the Nile Basin in order to respond to these challenges and achieve sustainable socioeconomic development on both national and regional levels [1].

Since 1986, the Baro river basin in southwest Ethiopia has undergone ongoing LULC change as a result of displacement, an increase in population, resource overuse, and the growth of commercial farming [2]. Unusual floods that exceed riverbanks outside of the typical flood zones have a significant impact on the human settlements and agricultural regions downstream. Floods, soil erosion and other hydrological hazards are significantly influenced by surface runoff in their onset and progression [3, 4].

According to Steenhuis et al. [5] shows that the number of hydrological models simulating the discharge from watersheds in Blue Nile and other river basin in Ethiopia has increased exponentially in recent years. But, most of these models were originally developed for application in temperate regions. Beginning from rational method to more complex models, such as SWAT [6, 7] and water balance approaches [5, 8, 9].

The fact that permanent gullies (of medium to large size), mass movements (earth slide, earthquake, etc.), and river bank erosion are frequently overlooked is a significant constraint of measurement and prediction of catchment sediment supply. On the RUSLE, which was created for the prediction of mean annual soil loss due to sheet and rill erosion on hill slopes, empirical models are frequently built. Such models cannot be utilized to forecast catchment sediment yield since sediment deposition is not taken into account. While some models clearly depict sediment deposition and transport. The majority of these models are assessed using information from tiny catchments or erosion plots, without taking sediment deposition into account [10]. It would still be challenging to apply sediment transport and deposition elements at the watershed scale due to the complexity and interactions of the processes involved, as well as a lack of adequate input data. Additionally, the majority of models for soil erosion rate and sediment yield do not replicate the interaction between erosion, deposition, and topography [11].

With the exception of WEPP, erosion has been predicted by Universal Soil Loss Equation in the majority of the previous models used in the Ethiopian Highlands, including AnnAG-NPS and SWAT [6, 12] SHETRAN model [13].

This is because the underlying runoff mechanism in these models is based on infiltration excess. The Parameter Efficient Semi-Distributed (PED) model was created just recently. It uses saturation excess principles as its runoff mechanism. Three minor upland watersheds in the Ethiopian highlands, including the Blue Nile basin near the border with Sudan, had good performance from this PED model in predicting discharge and sediment concentrations [14]. Because it accurately represented the main hydrological processes in the basin, the PED-W model was chosen. In basins with mon-

soonal climates, this model is acknowledged. However, contemporary advancements and the use of the idea of saturated surplus runoff have backed the PED model in numerous investigations [2].

Using saturation excess, or the PED-W model, this study attempts to forecast and examine the regional discharge and sediment of the Sore watershed. Additionally, it makes an effort to address estimates of discharge and sediment yield (SYLD), modify the proper catchment-wide soil erosion control strategies, and demonstrate the model's applicability using statistical model performance criteria. Finally, this paper provides quite useful information about discharge and sediment transport as a reference for researchers and policy-makers to pick appropriate natural resource conservation strategies, implement water resource management policies for watershed managers, and select suitable discharge and sediment prediction models for engineers. This research article was preprinted for publication [14].

2. Materials & Methodology

2.1. Description of the Study Area

Baro-Akobo Sub Basin lines in the south west of Ethiopia between latitudes of $5^{\circ}31'$ and $10^{\circ}54'$ N, and longitudes of $33^{\circ}0'$ and $36^{\circ}17'$ E. The basin area is about 76,000 km² and is bordered by the Sudan in the West, northwest and southwest, Abbay and Omo-Gibe Basins in the east. The major rivers within the Baro-Akobo basin are Baro and its tributaries Alwero, Gilo and the Akobo. These rivers, which arise in the eastern part of the highlands, flow westward to join the White Nile in Sudan. The mean annual runoff of the basin is estimated to be about 23 km³ as gauged at Gambella station. Elevation of the study area varies between 440 and 3000 m a.m.s.l. The higher elevation ranges are located in the North East and Eastern part of the basin while the remaining part of the basin is found in lower elevation. In the study area, there is high variability in temperature with large differences between the daily maximum and minimum temperatures [15].

One of the tributaries of the Baro River is a river in southwestern Ethiopia, which defines part of Ethiopia's border with South Sudan. The Baro River is created by the confluence of the Birbir and Gebba rivers, east of Metu in the Illubabor Zone of the Oromia Region. From its source in the Ethiopian Highlands it flows west for 306 kilometers (190 mi) to join the Pibor River. The Baro-Pibor confluence marks the beginning of the Sobat River, a tributary of the White Nile. The Baro and its tributaries drain a watershed 41,400 km² (16,000 sq. mi) in size. The river's mean annual discharge at its mouth is 241 m³s (8,510 ft³s).

The Sore watershed (approximately 1665.5 km²), a sub-watershed of the Baro river basin, was chosen for this particular study (Figure 1). The Sore watershed is between 937 and 3001 metres above sea level in elevation. Sore's climate is divided into warm to chilly, semi humid areas

rainfall distribution zones based on altitude and local temperature. The Sore watershed experiences annual rainfall ranging from 1533 mm to 2046 mm on average. The studied

area's rainfall distribution pattern has a unimodal trend with constant highs from March to October

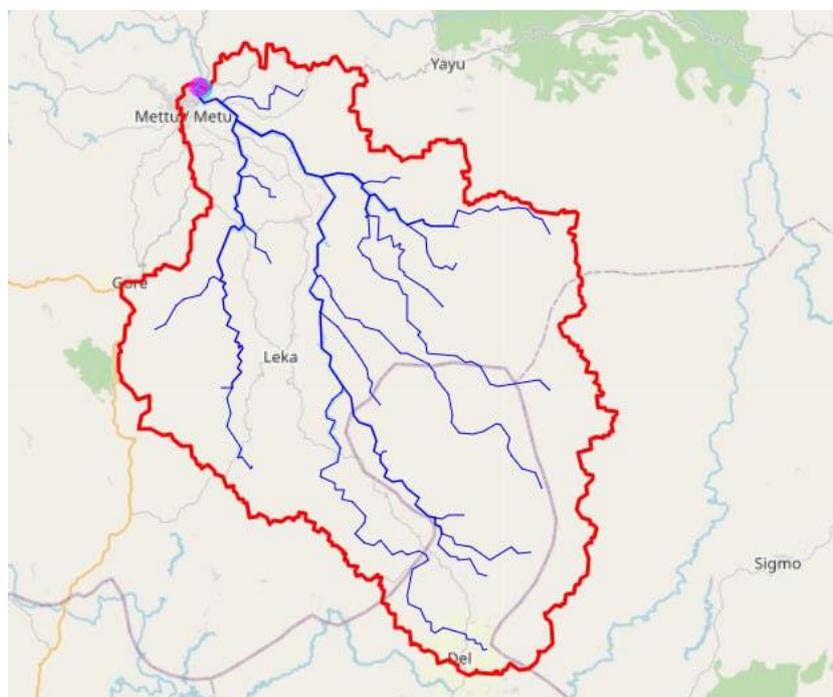


Figure 1. Location of Sore Watershed.

2.2. Data Sources and Availability

For this study, it was necessary to collect meteorological (Rainfall, Maximum & Minimum temperature, Evapotranspiration), hydrological (River discharge & sediment) and GIS (Land use land cover, Soil) data. These data were collected from Meteorological Agency of Ethiopia and Ministry of Water & Electricity. To calibrate and validate the model we used 2005-2018 years of recorded period.

Meteorological data

The required meteorological data, including daily precipitation, daily maximum, and daily minimum air temperatures, were collected from the Gambella branch of the Ethiopian National Meteorological Agency, which provided the inputs for PED models. It is the major source of meteorological data. These data were used for hydrological model development. The reason for selecting the representative meteorological stations that represent the catchments depends on the availability of climatic variables, the length of the record period, and others. The 13 years of meteorological data from 2005–2018 for the Sore watershed were used for calibrating and validating the selected hydrological models.

Hydrological data

The Sore catchment's daily stream flow and sediment (produced by the sediment rating curve) were used to calibrate and validate the model (PED-WM) using data gathered from

the Ministry of Water and Electricity (MoWE) for the Sore watershed from 2005 to 2018 at Sore near the Metu measuring station.

Digital Elevation Model (DEM)

The digital elevation model (DEM) is the digital representation of a topographic surface and it is specifically made available in the form of raster or regular grid of spot heights. The DEM data was used for watershed delineation and river network generation for the selected catchment. The Digital Elevation Model was obtained from Ministry of Water and Electricity and it has a horizontal resolution of 30m x30m.

Soil map/Soil shape file

The soil map (shape file) data was used to show the distribution of the soil type in the catchments and the various parameters describing the soil's hydrological and textural properties. The shapefile, which describes the distribution of soil in the study area, was obtained from the base line maps available at the Ministry of Water and Electricity at a scale of 1:250,000.

2.3. Data Quality Analysis

Initial data screening has been conducted by visually observing the data for outliers and missing values. Missed values and outstanding outliers at initial screening was systematically ignored and refilled by simple normal ratio (rainfall), arithmetic mean and XLSTAT excel model (maximum &

PET, potential Evapo-transpiration (mm/ time step), and Δt is the time step.

$$S_{t=} = S_{t-\Delta t} \left[\exp\left(\frac{P-PET}{S_{max}}\right) \right] \text{ when } P < PET \quad (5)$$

Base flow QB is calculated as a first order reservoir and interflow QI as a zero order reservoir [2, 5, 17]. The groundwater storage and the recharge to the interflow compartment, *Perc*, calculations depend on whether the groundwater storage has reached its maximum value of BS_{max} . Recharge to the Interflow only occurs when the base flow reservoir is full.

PED-W Sediment model

In the sediment model, the two runoff source areas are considered the main sources of sediment. The seepage from the subsurface flow to the stream channels unlike earlier application of the model. Sediment yield is depending on the suspended sediment concentration and discharge of the rivers [14]. Erosion originates from the run-off producing zones. Erosion is negligible from the non-degraded hill slopes because almost all water infiltrates before it reaches the stream. The maximum sediment concentration that a stream can carry sediment derived from the stream power function.

$$C_s = a_t qr^n \quad (6)$$

Where, transport limiting capacity (C_t , g L⁻¹) qr is the runoff rate per unit area from each runoff producing area (mm d⁻¹), and a_t is a variable derived from the stream power [(g L¹) (mm d⁻¹)- n]. Variable a_t is a function of the slope, Manning's roughness coefficient, slope length, and effective deposit ability (Guzman et al., 2013).

When cohesive soils eroded from the watershed sediment concentration will not always reach the transport limit. The sediment concentration is at the transport limit only when rills are actively formed in newly plowed soils (Tebebu et al., 2010). Once the rill networks has been formed no further erosion takes places, the sediment source becomes limited and the sediment concentration falls below the transport limit. In other cases, when the sediment concentration below transport limit, it does not down from sediment sources.

$$C_s = a_s qr^n \quad (7)$$

Where, a_s is the source limit and is assumed to be independent of the flow rate for a particular watershed (as compared to plots). Introducing a new variable, H , defined as the fraction of the runoff-producing area with active rill formation, the sediment concentration from the runoff-producing area (C_r , g L⁻¹) can then be written as:

$$C_r = C_s + H(C_t - C_s) \quad (8)$$

$$C_r = [a_s + H(a_t - a_s)] qr^n \quad (9)$$

Base flow and interflow play an important role in the calculation of the daily sediment concentration and sediment yield. In monsoon climate, base flow at the end of the rainy season can be a significant portion of the total flow.

Sediment load per unit watershed area (Y , g m⁻² d⁻¹) can be obtained by multiplying sediment concentration from the runoff-producing area (C_r , g L⁻¹) by the relative area and the flux per unit area (Zimale et al., 2016).

$$Y = A_1 qr_1 [a_{s1} + H(a_{t1} - a_{s1})] qr_1^n + A_2 qr_2 [a_{s2} + H(a_{t2} - a_{s2})] qr_2^n \quad (10)$$

Where, qr_1 and qr_2 are the runoff rates expressed in depth units for contributing area A_1 (fractional saturated area) and A_2 (fractional degraded area), respectively. Theoretically, for both turbulent flow and a wide field, n is equal 0.4 (Tilahun et al., 2015).

Potential Evapo-transpiration (PET)

For this specific study temperature based (Enku & Melesse, 2014) was used to calculate the daily potential evaporation during PED –W model calibration and validation.

$$ET_o = \frac{[T_{max}]^n}{k} \quad (11)$$

Where, ET_o is the reference evapo-transpiration (mm/day), n is 2.5 which can be calibrated for local conditions, k is Coefficient which can be calibrated for local conditions. The coefficient, k could be approximated as $k = 48 * T_{min} - 330$ where T_{min} is the mean annual maximum temperature.

2.4.2. PED-W Model Sensitivity Analysis

Sensitivity analysis is defined as the technique of identifying the responsiveness of different parameters involving in the simulation of a hydrological model. Sensitivity analysis performed to limit the number of optimized parameters to obtain a good fit between the simulated and measured data. It helps to determine the relative ranking of which parameters most affect the output variance due to input variability [18], which reduces uncertainty of the out puts of the model and provides parameter estimation guidance for the calibration step of the model. In Parameter Efficient semi-Distributed watershed (PED-W) model the sensitivity was done manually by changing one parameter and all other parameters are kept constant.

2.4.3. PED-W Model Calibration and Validation

In order to know the model performance the model must be proven for its reliability, accuracy and predictive capabilities by using model calibration and validation. Calibration is tuning of model parameters based on checking results against observations to ensure the same response over time. Initially the model would not often give satisfactory result. Due to these reason hydrological models requires adjustment of the values of model parameters to match model output with

measured data for selected period. After calibrating, the model should be validated against independent data without making farther adjustments to test model simulation capabilities. So, for this study model was calibrated and validated using 9 & 4 years recording discharge and sediment data of a river.

2.5. Model Evaluation and Selection Criteria

The performance of PED-WM was evaluated by using statistical measures to determine the quality and reliability of prediction when compared to observed values. In this study, the goodness of fit was quantified by four model evaluation statistics including: Nash-Sutcliffe efficiency (NSE), the ratio of root mean square error to the standard deviation of measured data (RSR) and coefficient of determination (R^2).

The Nash-Sutcliffe simulation efficiency (NSE) indicates how well the plot of observed versus simulated value fits the 1:1 line. If the measured value is the same as all predictions, ENS is 1. If the ENS is between 0 and 1, it indicates deviations between measured and predicted values. If ENS is negative, predictions are very poor, and the average value of output is a better estimate than the model prediction (Nash and Sutcliffe, 1970). The R^2 value is an indicator of strength of relationship between the observed and simulated values. NSE is calculated using Equation (12).

$$NSE = \frac{(\sum_{i=1}^n O_i - P_i)^2}{(\sum_{i=1}^n O_i - \bar{O})^2} \quad (12)$$

Where, NSE is the prediction efficiency, O_i is the observed condition at time i , \bar{O}_i is the mean of the observed values over all times, P_i is the predicted value at time I . The index i refer to storm number for calculating the prediction efficiencies for sediment yield, and refer to time during the storm for calculating the efficiency of a hydrograph for a particular storm.

Root mean square error (RMSE) observations standard deviation ratio (RSR) is calculated as the ratio of the RMSE and standard deviation of measured data, as shown in Equation (13).

$$RSR = \frac{\sqrt{\sum_{i=1}^n (O_i^{obs} - O_i^{sim})^2}}{\sqrt{\sum_{i=1}^n (O_i^{obs} - O_i^{mean})^2}} \quad (13)$$

Many researchers use different models selection criteria to predict discharge and sediment yield in watershed [2, 11, 14, 19]. These criteria are including: Prediction accuracy as indicated by the level of agreement between measured and predicted discharge & soil erosion or sediment yield data, Potential to provide information about dominant soil erosion and sediment transport processes within catchments, Detail and amount of required input data and model calibration requirements and models potential for use in scenario studies of changing climate, land use, and land management. Appropri-

ate models are need for the better assessment of the hydrology and soil erosion processes as well as decision support system for planning and implementations of appropriate measures [6].

For this study PED-W model was used to predict the discharge and sediment yield in Sore watershed. The selection is based on: i) the model simulated the major hydrological process in the catchments. ii) PED-W model is suitable for monsoonal climate watersheds. iii) Because of its semi-distributed nature and its structure is more physically-based than the structure of lumped model, and it is less demanding on input data than fully distributed model. On the other hand, PED model have been selected by many researchers due to recent innovation and account saturation excess runoff principles than others (infiltration process principles) [2, 5, 20]. PED model performed well in predicting discharge and sediment concentrations for three small upland watersheds in the Ethiopian highlands as well as for the entire Blue Nile Basin at the Sudan Ethiopian border [2].

3. Result and Discussion

The findings and analyses of this study are reported in the following three main sub-sections: The results of this study were given for the PED-W hydrological model and included model calibration, validation, and efficiency evaluation. PED-W models were used to present the predicted stream flow and sediment for the Sore watershed, respectively. By comparing anticipated and actual discharge and sediment flow at the outlet point, the model's appropriateness for the Sore watershed in the Baro Akobo basin was determined.

3.1. PED-W Model Results of Sore Watershed

Both the observed and the simulated daily discharge and sediment for the Sore watershed show satisfactory agreement for the calibration period 2005 to 2013 (NSE = 0.40, 0.44), using a runoff contributing area of 0.088% (0.024% saturated and 0.064% degraded) and an aquifer half-life of 70 days and 64 days, respectively. Once more, throughout the validation period of 2014 to 2018, the anticipated and actual daily discharge and sediment yielded inadequate results (NSE = 0.22, -0.82). In the Sore watershed, monthly calibration and validation findings for discharge indicate acceptable agreement between observed and predicted values (NSE = 0.52, 0.47), whereas the results for sediment calibration and validation show reasonable agreement (NSE = 0.30, 0.19). The calibration of discharge and sediment in the Sore watershed revealed good and satisfactory agreement between the observed and simulated values according to the results of the coefficient of determination R^2 and root mean square error (RMSE).

PED-W Calibrated Parameters:

The fractional areas for Sore watershed add up to only 0.69 or 69% (Table 1). An area proportion of one means the cal-

culated interflow, base flow and storm flow are equal to the long term discharge measured at the outlet. In other words since the long term average of the discharge in the PED-W model equals the average of the net precipitation (i.e., rainfall minus evaporation), all precipitation reaches the outlet eventually. However, the total contributing fractional areas of Sore watershed 0.69 or 69% means that the net input precipitation is much more than the discharge at the outlet (Sore near mettu gaging station). Because, the unaccounted net precipitation either flows subsurface under the gauge to the Sore near mettu

or the discharge was not measured correctly. According to Moges et al., [17] the under prediction of high flows for Ribb river is the consequence of the increased bed levels, can observed clearly by [21]. The result shows that measured flow is much less than the predicted. The final reason for contributing areas not adding up to one is the missing rainfall, this is facilitated subsurface flows and reduces the flow at the gauging stations, and also for this study the same phenomenon was happened.

Table 1. PED parameters value of discharge and sediment load for Sore River.

| Hydrology | Parameters | Unit | Calibration | | |
|------------|------------------------------|-------------------|------------------------------|------------------------------|-----|
| Discharge | Saturated area | Area A1 | % | 0.024 | |
| | | S_{max} , A1 | mm | 300 | |
| | Degraded area | Area A2 | % | 0.064 | |
| | | S_{max} , A2 | mm | 50 | |
| | Infiltration area | Area A3 | % | 0.6 | |
| | | S_{max} , A3 | mm | 5 | |
| | | BS_{max} , | mm | 20 | |
| | | Subsurface | $t^{1/2}$ | days | 70 |
| | | | τ^* | days | 64 |
| | | Sub- surface flow | α_B | $(gL^{-1})(mmd^{-1})^{-0.4}$ | 0.1 |
| α_I | $(gL^{-1})(mmd^{-1})^{-0.4}$ | | 0.0 | | |
| α_t | $(gL^{-1})(mmd^{-1})^{-0.4}$ | | 1.0 | | |
| α_s | $(gL^{-1})(mmd^{-1})^{-0.4}$ | | 1.5 | | |
| Sediment | Saturated area | α_t | $(gL^{-1})(mmd^{-1})^{-0.4}$ | 4.0 | |
| | | α_s | $(gL^{-1})(mmd^{-1})^{-0.4}$ | 1.5 | |
| | Degraded area | α_t | $(gL^{-1})(mmd^{-1})^{-0.4}$ | 4.0 | |
| | | α_s | $(gL^{-1})(mmd^{-1})^{-0.4}$ | 1.5 | |

A1, A2 and A3 are area fractions of the saturated, degraded and recharge hillside areas respectively. S_{max} is the maximum water storage capacity; BS_{max} is the maximum base flow storage of linear reservoir; $t^{1/2}$ is time taken in days to reduce volume of the base flow reservoir by half under no recharge conditions or base flow half-life time in days; τ^* is the duration of the period after a single rainstorm (until interflow ceases) or inter flow in days; α_s and α_t are coefficient of sediment source and transport model for components of base-flow (B), interflow (I), saturated area (1) and degraded area (2).

Both sediment parameter coefficients, transport limiting and source limiting, were found to demonstrate difference between the two basins (Upper Blue Nile and Baro-Akobo Basin). The transport and source limiting capacity from saturated area and degraded areas at the outlet point of Sore

catchment were found to be $\alpha_t = 1$ & $\alpha_s = 1.5$ and $\alpha_t = 4$ & $\alpha_s = 1.5$ respectively. The result is found to be less than Ribb ($\alpha_t = 1$) and Kessie ($\alpha_t = 5$) watersheds in Upper Blue Nile Basin shown by [17]. There is no clear reason why this should be the case, but it may be the active gully formation and it is the reason why in Sore watershed (smallest area) time of concentration is much more than Ribb and Kessie (largest area).

3.1.1. PED-W Sensitivity Analysis

The results of sensitivity analysis in stream flow gave the degree of sensitivity of nine parameters which was important for the manual calibration methods. Among these parameters the saturated area (A1), degraded area (A2), hill- side area (A3), maximum soil storage for base flow (BS_{max}) and base flow half-life ($t^{1/2}$) are sensitive parameters. From nine inputs

PED- W model parameter five parameters were found the most sensitive, including: saturated area (A1), degraded area (A2), hillside (A3), maximum soil storage in saturated area (S_{max}) and recession coefficient (K). Portion of infiltration area, A3 was the most sensitivity parameters for both watersheds, meaning maximum runoff were generated from the hillside area. Similarly, in sediment load analysis sensitivity

was done by manual calibration methods by using two parameters: including sediment transport and sediment source from A1 (saturated area) and A2 (degraded area) (at, as). From these parameters sediment transport (transport limit) from degraded area is the most sensitivity for the watershed. The degree of sensitivity of the parameters is different in different watershed (Table 2).

Table 2. Calibrated parameters for PED-W model in to sensitivity watersheds and their sensitivity rank (discharge and sediment).

| | Parameters | Sensitivity rank |
|---------------------|----------------|------------------|
| Discharge component | Area, A1 | 2 |
| | S_{max} , A1 | 9 |
| | Area, A2 | 3 |
| | S_{max} , A2 | 8 |
| | Area, A3 | 1 |
| | S_{max} , A3 | 5 |
| | BS_{max} , | 4 |
| | $t^{1/2}$ | 7 |
| | τ^* | 2 |
| Sediment component | α_t | 1 |
| | α_s | 2 |

3.1.2. PED-W Model Calibration and Validation

For PED-W model, all nine input parameters for discharge and six parameters for sediment were calibrated. The starting values for calibrating parameter were based on [5, 15] for discharge and [14] for sediment. These initial values were changed manually through randomly varying input parameters in order that the best “closeness “and “goodness of-fit” was achieved between predicted and observed suspended sediment flow in the watersheds. The hydrograph shape for the Sore watershed from 2005 to 2013 was calibrated in the PED-W model after the sediment was calibrated. From 2014 to 2018, the calibrated model underwent

validation. With an NSE of 0.44 for the calibration period and -0.82 for the validation period, the sediment was either predicted or simulated at a daily time step. The model was calibrated with an NSE of 0.47 and 0.19 for the Sore watershed during the calibration period, showing a reasonable and usable difference between the observed and simulated sediment load throughout the validation period. The observed and simulated sediment result of the Sore watershed was lower than [17] at the Ribb watershed, Upper Blue Nile Basin, throughout the calibration and validation period of the PED-W model on a daily time step NSE value of (0.44, -0.82).

Table 3. Calibration & validation performance of discharge (PED-W model).

| description | Daily | | | Monthly | | | |
|-------------|-------------|------|------|---------|------|------|-------|
| | R^2 | NSE | RMSE | R^2 | NSE | RMSE | |
| Sore | Calibration | 0.44 | 0.40 | 2.16 | 0.59 | 0.52 | 10.04 |
| | Validation | 0.50 | 0.22 | 2.53 | 0.63 | 0.30 | 12.58 |

Table 4. Calibration and validation performance of sediment yield by PED-W model.

| description | Daily | | | Monthly | | |
|------------------|----------------|-------|------|----------------|------|------|
| | R ² | NSE | RMSE | R ² | NSE | RMSE |
| Sore Calibration | 0.52 | 0.44 | 0.04 | 0.66 | 0.47 | 0.18 |
| Sore Validation | 0.49 | -0.82 | 0.00 | 0.66 | 0.19 | 0.22 |

Based on the Nash and Sutcliffe efficiency, the coefficient of determination and R2 results from the calibration period of 2002 to 2011 (0.62 and 0.55), and the hydrograph pattern for the Sore watershed, it can be seen that the relationship between the daily observed and simulated sediment load was

satisfactory (Figure 3). For the Sore watershed over the calibration and validation periods, the hydrograph displays the observed and simulated discharge and sediment linked peak to peak, rising to rising, and recession to recession.

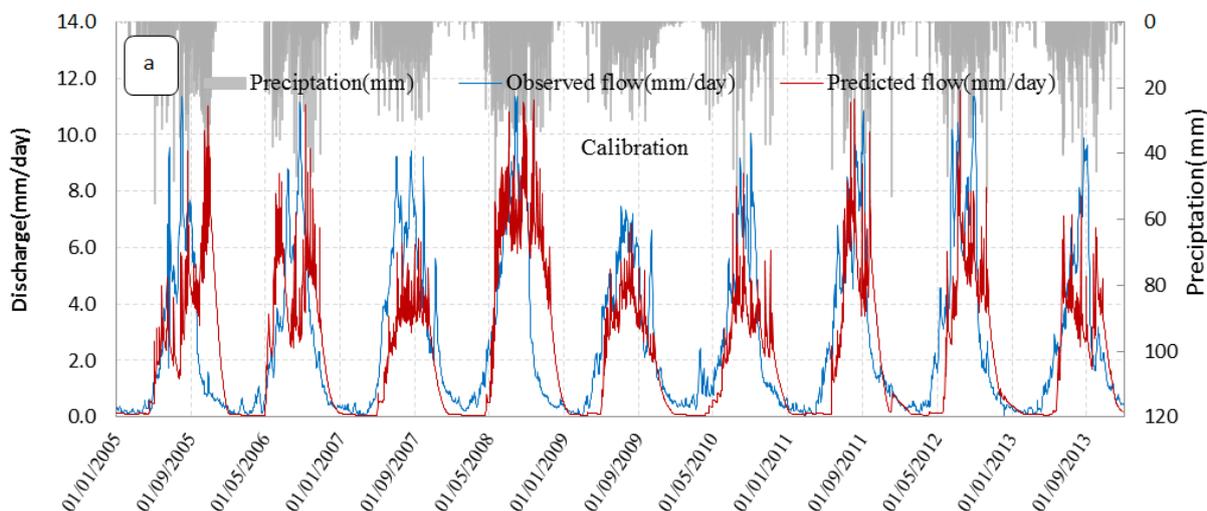
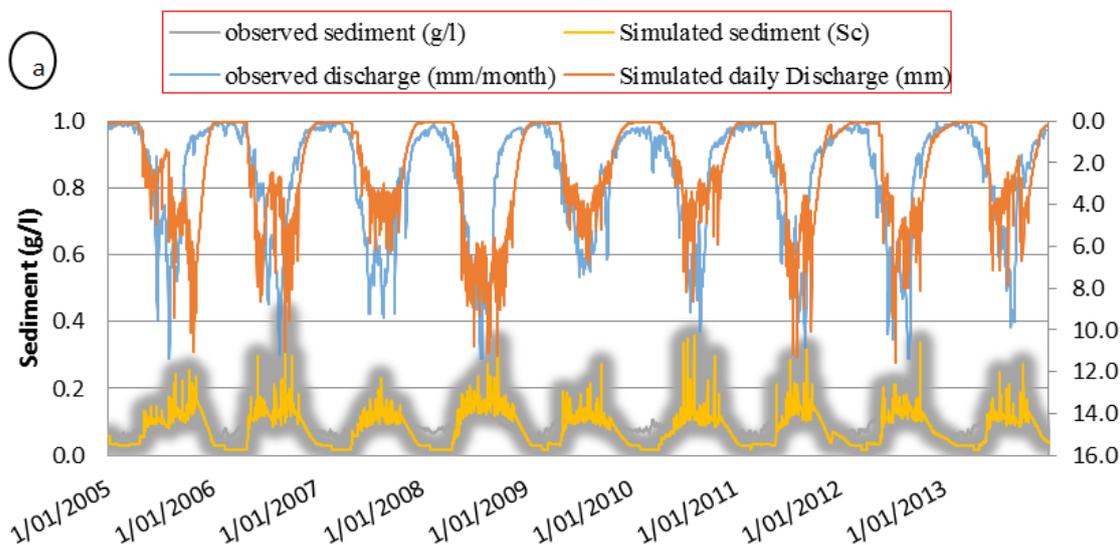


Figure 3. Daily predicted and observed stream flow for Sore watershed during (a) calibration and (b) validation period (2005-2013) and (2014-2018).



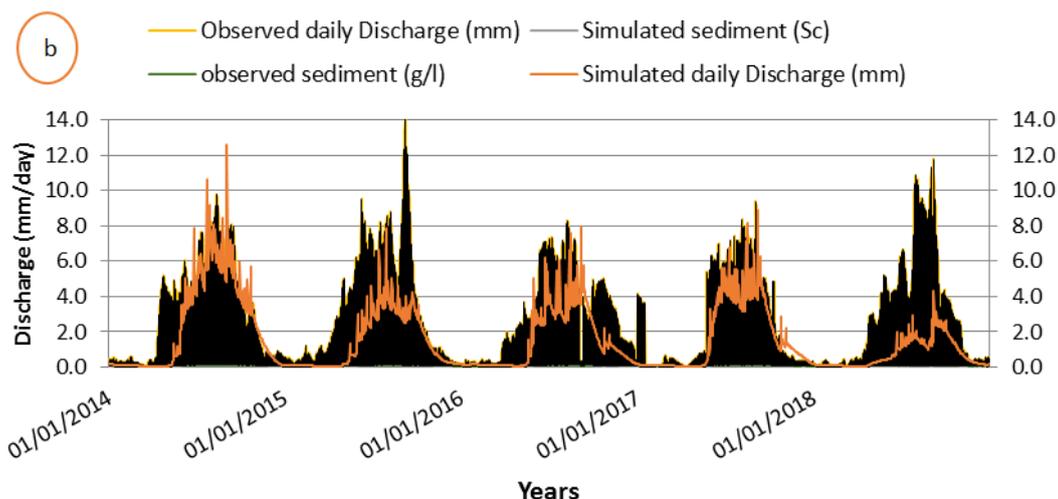


Figure 4. Daily predicted and observed sediment for Sore watershed during (a) calibration and (b) validation period (2005-2013) and (2014-2018).

The Sore watershed's monthly calibration reveals that the model was able to simulate the peak flows and sediment loads in a moderately accurate manner (Good for stream flow and sediment load), as well as the characteristics of dry periods. For flow and sediment load, respectively, the Nash-Sutcliffe Efficiency (NSE) values were 0.52 and 0.47. This demon-

strates a good match between the simulated and measured flow and sediment. Results that were predicted outperformed those that were simulated. Problems with discharge and sediment measurement as well as shortcomings in the chosen hydrological models were to blame for this.

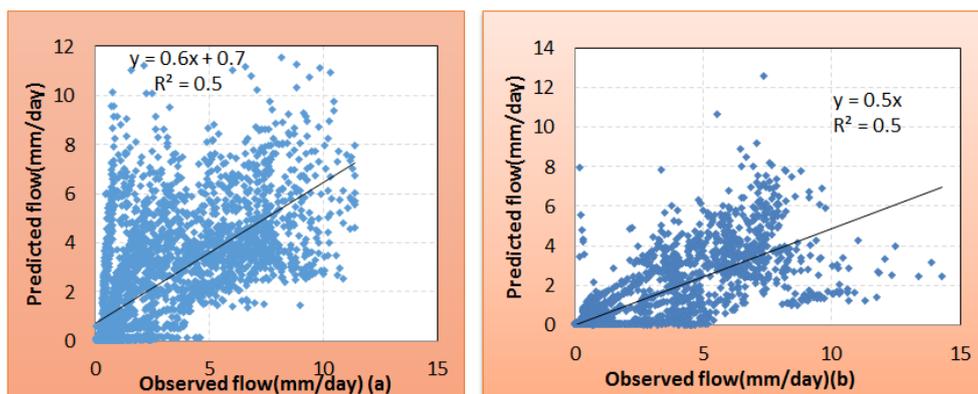
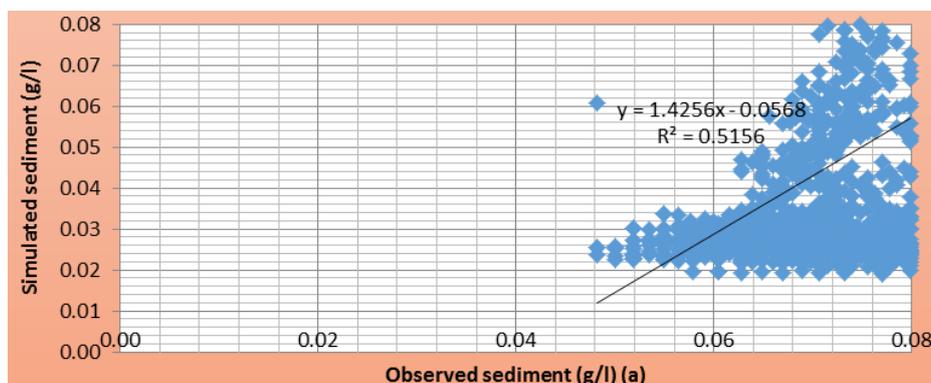


Figure 5. Daily observed and simulated discharge for Sore watershed a) Calibration and b) Validation period (2005-2013) and (2014-2018).



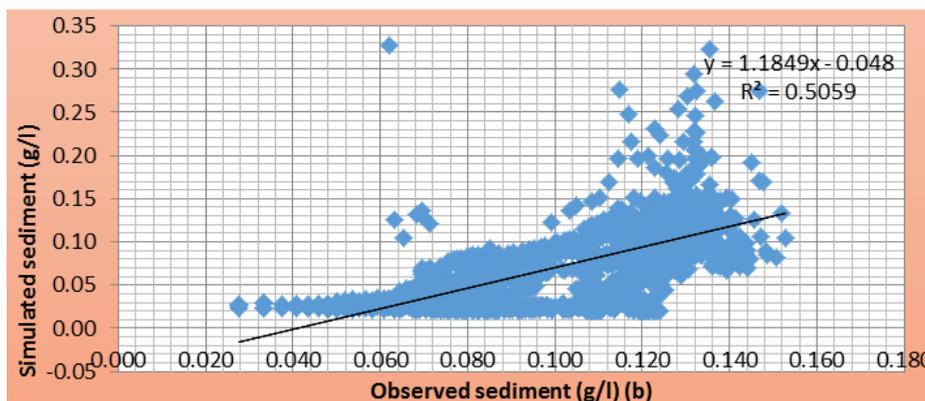


Figure 6. Daily observed and simulated sediment for Sore watershed a) Calibration and b) Validation period (2005-2013) and (2014-2018).

3.1.3. Discharge & Sediment Yield

In this study, the PED models were used to forecast the sediment yield at Sore near Mettu as well as their relative distribution among the sub basins. These models were calibrated and validated at the Sore watershed gauging station. Based on this simulation, the daily and monthly average yearly sediment yields for the Sore watershed at the Sore

near Mettu outlet are 0.08 g/l and 2.94 g/l on average from 2005 to 2013, respectively. Similarly, simulated discharge levels at the same outlet point are higher than those that were observed during both the calibration and validation periods. The lack of prior research on the watershed by PED made it challenging to compare and build the daily and monthly simulated values of the PED models.

Table 5. Average values of observed and simulated discharge and sediment yield by PED-W model.

| Parametres | Observed values | | | | Simulated values | | | |
|----------------|-----------------|---------|------------|---------|------------------|---------|------------|---------|
| | Calibration | | Validation | | Calibration | | Validation | |
| | Daily | monthly | Daily | monthly | Daily | monthly | Daily | monthly |
| Discharge (mm) | 2.68 | 81.55 | 3.06 | 93.09 | 2.26 | 68.74 | 1.56 | 47.4 |
| Sediment (g/l) | 0.10 | 2.49 | 0.1 | 2.14 | 0.08 | 2.96 | 0.07 | 3.04 |

3.2. Model Results Comparison

During the Sore watershed's calibration and validation periods, a daily time step of the observed sediment was reproduced using the PED model with NSE of 0.44 and -0.82, R2 of 0.52 and 0.49, and RMSE of 0.04 and 0.01. Similar to that, the catchment was calibrated and verified at monthly time steps, with RMSE of 0.18 and 0.22 at the outflow point, NSE of (0.47 and 0.19), R2 of 0.66 and 0.66, and NSE of (0.47 and 0.19). The result of PED-W model on discharge & sediment prediction was less than the similar studies in Baro Akobo Basin, Baro catchment [20, 21]. They applied SWAT model to predict discharge of the catchment, very good agreement between observed and simulated values.

Although, PED-W model data input requirement is less relative to SWAT model, the overall model performance

indicated that PED-W model at daily time step was less accurate and unsuitable to simulate flow and sediment in sore watershed at daily time steps. Because, PED-W model runoff simulations mechanisms were based on saturation excess and the sediment model scaling up plot erosion estimates. For monthly time steps of discharge and sediment calibration and validation of PED model result was proportionally the same as SWAT model applied previously.

4. Conclusion and Recommendation

4.1. Conclusion

PED-W watershed hydrology model was evaluated predicting stream flow and sediment yield in Sore watershed in the Baro- Akobo Basin; where direct runoff were generated

from saturation (PED-W) excess principles. At a daily time step the sediment yield was simulated using PED- W with NSE of 0.44 and -0.82 during calibration and validation periods, respectively. The predicted discharge and sediment for Sore watershed for each outlet runoff amount and sediment yield were compared with observed data.

By dividing the record time of discharge and sediment into calibration and validation periods (64.3% for calibration and 35.7% for validation), the watershed properties of the models were evaluated. After model calibration, optimized parameters for watershed models were verified, and the outcome showed a satisfactory correlation between observed and simulated hydrological phenomena (stream flow and sediment). The generated hydrograph result and the observed hydrographs may not match up after model calibration. This could be as a result of the input data not accurately reflecting a unique circumstance or the model structure not perfectly capturing the characteristics of the catchments.

To evaluate the effects of land use change, management practices (land and soil conservation), and climate change on flow and sediment dynamics in the watershed, utilize the calibrated and validated PED-W model for the Sore watershed.

Researchers generally agree that the PED model at the daily time step is the best for predicting discharge and sediment in the Upper Blue Nile basin, but not in the Baro Akobo basin (like the Sore watershed). Discharge and sediment calibration and validation, however, were proportionate to the SWAT model that had previously been utilized at the monthly time step. This resulted from PED-W being oversaturated and scaling up plots, as is the case in Ethiopia's highlands.

4.2. Recommendation

In future research to forecast or evaluate surface water resources and sediment production, the effectiveness of the model in the Sore catchment should be assessed at different upstream and downstream measurement stations. The calibrated models can be utilized for further investigation of stream flow and soil erosion consequences of climate and land use change, as well as other various management scenarios. The operation of the gauging stations (both hydrological and meteorological) requires additional maintenance and qualified personnel. Consideration of PED model parameters has to be updated. The study's findings may aid various researchers, stakeholders, policymakers, and managers in developing and putting into practice appropriate soil and water conservation strategies. They may also aid the designer of hydraulic structures in choosing suitable and effective hydrological models to simulate the amount of sediment that will accumulate in reservoirs and the yield of sediment from the basin.

Abbreviations

| | |
|-----|---------------------------------|
| GIS | Geographical Information System |
| NSE | Nash and Sutcliffe efficiency |

| | |
|-------|---------------------------------------|
| PED | Parameters Efficient semi-Distributed |
| RUSLE | Revised Universal Soil Loss Equation |
| SWAT | Soli Water Assessment Tool |

Conflicts of Interest

The authors declare no conflict of interest.

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