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Evaluation of the Impacts of Land Use Land Cover Change on Hydrology—A Case Study of the Nashe Watershed ⁺

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Abstract: Changes in land use and land cover (LULC) have a significant impact on a watershed's 12 hydrological processes. Investigating the impact of land use land cover change on hydrological com-13 ponents and the responses of watersheds to environmental changes is crucial for water resource 14 planning, land resource utilization, and maintaining hydrological balances. The impacts of land use 15 land cover on hydrological parameters in the Nashe watershed, Blue Nile River Basin, are explored 16 in this study. Historical and future land use land cover change scenarios that represent baseline, 17 current and future periods have been implemented into a calibrated Soil and Water Assessment 18 Tool (SWAT) model using Digital Elevation Model (DEM), land use land cover maps, soil data, 19 weather data, and hydrological data. The result showed the land use land cover changes analyzed 20 for the time period of 2019 to 2035 reveal a decline in ground water flow, lateral flow, evapotranspi-21 ration, and increment of surface runoff, and water yield. This depicts that the land use land cover 22 change will occur in the future by decreasing forest land and increasing agricultural land and urban 23 area that will increase the vulnerability of the watershed. 24

Keywords: LULC change; surface runoff; SWAT model; water balance

1. Introduction

Land use land cover change effect on hydrology is revealed at different scales and 28 has become a worldwide concern because of its varied environmental effects. Many hy-29 drological processes such as rainfall, evapotranspiration, and runoff are significantly af-30 fected by LULC change. The most frequent analysis in hydrology is runoff estimation in 31 a watershed depending on rainfall distribution [1]. The assessment of LULC change and 32 the drivers that have direct consequences on the natural environment and human socie-33 ties are the focus of the current scientific examination of scientists [2]. Therefore, it is im-34 portant to investigate the impacts of LULC change on the hydrology of the catchment to 35 address water resource operation and management issues. 36

Assessment of the historical, current, and potential future LULC change dynamics is 37 essential to manage LULC and water resources efficiently in a watershed [3]. Analyzing 38 and predicting the future watershed hydrology through advanced tools over a long pe-39 riod is significant to attain sustainable water resources at the catchment scale [4]. Soil and 40 Water Assessment Tool (SWAT) is used to investigate the effects of LULC change on hy-41 drological processes in small and large watersheds [5]. The model is also mostly used for 42 modeling and analyzing hydrological processes in the context of changing LULC and land 43 management with high efficiency. The findings of different studies that compared the 44

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Copyright: © 2021 by the authors. Submitted for possible open access publication under the terms and conditions of the Creative Commons Attribution (CC BY) license (https://creativecommons.org/licenses/by/4.0/). SWAT performance to other hydrological models conclude that the SWAT model simulates stream flow better than other hydrological models [2,6]. 46

Therefore, based on the criteria specified for the Nashe watershed, the SWAT model 47 has been adopted. Ethiopia experienced serious environmental problems including soil 48 erosion, land degradation, loss of soil fertility, and deforestation due to LULC changes [7]. 49 The investigations directed on the hydrological processes of watersheds dependent on 50 LULC change show an increment of flow in the wet season and surface runoff potential 51 that relates to the agricultural and urban area expansion at the expense of forest lands [8]. 52

The Blue Nile River Basin is the most substantial and diverse water resource for the region and continent also serving as the largest catchment to the Nile River Basin. Humaninduced land degradation has occurred in the Nashe watershed that subsidizes a large amount of water to the Blue Nile River Basin [7]. Furthermore, in the Nashe catchment, the agricultural land and urban expansion at the expense of range land, forest land, and grass land is the common problem and this will also be predictable to continue in the future [3].

This paper aimed to assess different features of LULC change impacts on hydrological parameters at various Spatio-temporal scales and to develop LULC scenarios to explore the change of LULC effect on hydrological parameters of the watershed. Therefore, analyzing the impacts of LULC on the hydrological processes at different periods and prioritization of the sub-basins will contribute to identifying strategies of hydrological responses of the watershed. 65

2. Materials and Methods

2.1. Study Area

The study was conducted in the upper Blue Nile River Basin, Nashe catchment in Oromia Regional State, Ethiopia. Nashe catchment lies in 9°35′ to 9°52′ North latitude and 37°00′ and 37°20′ East longitude covering 94578 Ha areas (Figure 1). The Nashe watershed is the major tributary of Blue Nile River Basin of Ethiopia which is situated about 300 km from Addis Ababa. The watershed area varies in elevation from 1600 m in the lower plateau under the escarpment to the hills and ridges of the highland climbing to over 2500 m. 73



Figure 1. Location map of the study area.

2.2. Data

The main data used for the study to characterize the basin was weather data, Digital 77 Elevation Model, land use land cover maps, soil data, and hydrological data. The historical LULC images were obtained from Landsat images and classified using supervised 79 classification in Earth Resource Data Analysis System (ERDAS) imagine model [3]. The 80 future LULC was predicted based on the classified historical satellite images using Land 81 Change Modeler (LCM) integrated TerrSet model (Figure 2 b). The historical (1990, 2005, 82 and 2019) and future (2035 and 2050) LULC maps developed by Leta et al. [3] were used. 83

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Figure 2. Nashe watershed a) DEM b) LULC c) Soil.

2.3. Modeling

SWAT is physically based, a spatially semi-distributed, daily time step hydrological 87 model designed to simulate a range of various parameters such as lateral flow, ground 88 water, surface runoff, and soil water. Similarly, the SWAT model was developed to pre-89 dict the impact of land use and management on water, sediment, and agricultural chemi-90 cal yields at catchment scale at daily, monthly, and annual time increments. 91

The major model components include DEM, weather, hydrology, soil properties, and 92 land management. Depending on the extent of the watershed and detail of available geo-93 graphical input data, the SWAT model splits a watershed into several sub-basins, which 94 are further divided into smaller areas denoted as hydrologic response units (HRUs) [9,10]. 95 As a result, depending on the topographical information data, the SWAT divides the 96 Nashe catchment into 23 sub-watersheds that are then divided into a total of 321 HRUs 97 based on their soil type, land use land cover, and slope. 98

$$SWt = SWo + \sum_{i=1}^{n} (Rday - QSurf - Ea - Wseep - Qgw)$$
(1)

where: *SWt* is the final soil water content(mm), *SWo* is the initial water content (mm), *t* is 99 the time (days), *Rday* is the amount of precipitation on day *i* (mm), *Qsurf* is the amount of 100 surface runoff on day i (mm), Ea is the amount of evapotranspiration on day i (mm), Wseep 101 is the amount of water entering the vadose zone from the soil profile on day i (mm) and 102 *Qgw* is the amount of return flow on day *i* (mm). 103

2.4. Sensitivity Analysis, Calibration and Validation

Due to a large number of flow parameters in SWAT, ascertaining the most sensitive 105 parameters is crucial to improve the calibration of the hydrological model. The Sequential 106 Uncertainty Fitting (SUFI-2) integrated in the SWAT-CUP (Calibration and Uncertainty 107 Program) has been used to achieve sensitivity analysis, calibration and validation [11]. 108 The process of estimating hydrological model parameters by comparing the model pre-109 diction with the observed data is known as Calibration. Whereas, testing the calibrated 110 model without further parameter adjustments with an independent dataset is known as 111 validation. The observed stream flow of 1985-2008 was divided into a warm-up (1985-112 1986), calibration period (1987–1999), and validation period (2000–2008). 113

The simulation of the model fitness with the observed stream flow was expressed by 114 statistics like coefficients of determination (R²), Nash-Sutcliffe efficiency (NSE), and per-115 cent bias (PBIAS). The performances of the model ratings were: R^2 varies between 0 and 116 1, where higher value shows less error. NSE ranges from negative infinity to 1, where 1 117 indicates the best. PBIAS varies from negative infinity to positive infinity. Where the value 118close to 0 shows the best simulation, a negative and positive value indicates overestima-119 tion, and underestimation respectively. The two measurements used to assess the quality 120 of uncertainty analysis are the p-factor and the r-factor. The p-factor is a proportion of 121

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measured data bracketed by the 95PPU that varies from 0 to 1, with 1 being the optimal result. The r-factor ranges between 0 to infinity, and it is the average thickness of the 95PPU band to the standard deviation of the corresponding measured data.

The calibrated and validated model with the historical (1990, 2005, 2019) and future 125 (2035, 2050) LULC maps were used to reveal the effects of LULC variations on watershed 126 hydrology. In this study five LULC scenarios were developed, representing baseline, cur-127 rent, and future LULC conditions. The first two scenarios (1990 and 2005) are considered 128 as a baseline. The third scenario corresponds to 2019 represents the current LULC. The 129 2035 and 2050 projected under Business as Usual Scenario represents the fourth and fifth 130 scenarios representing the future LULC change. Correspondingly, to develop linear cor-131 relations between dependent variables and independent variables the pair-wise Pearson 132 correlation matrix was implemented. 133

3. Results and Discussions

3.1. Sensitivity Analysis, Calibration and Validation

Sensitivity analysis (r_CN2.mgt, v_GW_DELAY.gw, r_SOL_K (...).Sol, v_AL-136 PHA_BF.gw, v_CH_N2.rte, v_GWQMN.gw, r_SOL_AWC (...).Sol, r_SLSUBBSN.hru, 137 r_RCHRG_DP.gw) was conducted and the top three most sensitive parameters are 138 CN.mgt, GW_DELAY.gw, and SOL_K(..).Sol. The parameters include those governing 139 sub-surface and surface hydrological processes and stream routing. The simulated and 140 observed graphical and statistical comparison shows a good agreement both in calibration 141 and validation periods (Figure 3). The evaluation of simulated and observed stream flow 142 computed through the statistical values of objective functions are in the recommended 143 range based on the performance assessment criteria. Therefore, the performance indices 144 obtained indicates a good performance rate of the model in simulating the impacts of 145 LULC changes [7]. 146



Figure 3. Calibration and validation of average monthly stream flow.

3.2. Land Use Land Cover Change Effects on Hydrological Responses

The effect of LULC change on hydrological parameters of the Nashe watershed was 150 assessed based on the LULC classes of different periods. The seasons in Ethiopia are cat-151 egorized into three in the year based on the rainfall magnitudes. The wet season is from 152 June to September, the season of short rain (February to May), and the dry season from 153 October to January. The hydrological parameter variability due to LULC change was as-154 sessed based on these three rainfall seasons. The findings revealed that above 80% and 155 40% of the surface runoff and ground water happens throughout the wet season, whereas 156 less than 10% of the surface runoff happens in the dry and short rainy season. 157

The surface runoff in the wet season was increased by 2.15% from 2019 to 2035 LULC 158 change (Figure 4). The extraction of forest land, range land, grass land, and agricultural 159 coverage and urban area expansion highly influence surface runoff, peak flow, and base 160 flow following rainfall events [9]. The reduction in forest land decreases infiltration and 161 evapotranspiration rates, resulting in a decrease of base flow and an increase in impervi-162 ous surface covers. Evapotranspiration has reached a minimum from October to January 163 and maximum in the period of February to May. The urban area expansion increases high 164 stream flow and decreases low stream flow. 165



Figure 4. Average annual hydrological components change (%) under land use land cover scenarios.

The average surface runoff of the catchment was increased by 7.94%, 10.45%, and 168 9.17% in 2019, 2035, and 2050 respectively, compared to the baseline scenario (1990). In 169 contrast, the average annual lateral flow of the watershed declined by 2.27%, 5.55%, and 170 18.24% in 2019, 2035, and 2050 with the baseline scenario (1990). The surface runoff will 171 decrease by 1.41 % from 2035 to 2050 due to the gradual increase of grass land and range 172 land starting from the year 2035. Gyamfi et al. [12] and Leta et al. [9]indicated that the 173 surface runoff, groundwater, and base flow parameters were affected by LULC changes. 174 It was also observed from the result a strong positive Pearson correlation factor was found 175 between agricultural land and surface runoff. Similarly, a strong negative correlation has 176 also happened between forest land and surface runoff. 177

In hydrological components, changes in rainfall are the dominant factor that induces 178 changes in water balance components. Evapotranspiration is the foremost water availa-179 bility determinant in the watershed since it negatively influences surface runoff. The de-180 crease in forest land, grass land, and range land combined with an increase in slope length 181 and steep slopes cause surface runoff increment [9]. The monthly peak flows happened in 182 July and August and the maximum monthly discharges occurred in 2050, while the min-183 imum flow occurred in 1990. Generally, the increase of surface runoff in wet seasons may 184 result in flooding and a decline in the dry season may affect water scheme practices. 185

4. Conclusions

The SWAT hydrological model was used to simulate historical and future continuous 187 fluctuations in stream flow through time. The relation of LULC categories and hydrolog-188 ical components revealed that the surface runoff was highly attributed to change in the 189 agricultural land with a higher correlation coefficient. Similarly, it was observed that the 190 increment of surface runoff and decline of ground water observed during the rainy season 191 in the Nashe watershed of the Blue Nile River Basin may lead to increasing extreme 192 weather events, sedimentation, runoff, siltation, and water shortages may occur during 193 the dry season and obstruct socio-economic development in Ethiopia. The suitable man-194 agement policy should be prepared depending on the usually LULC change of the water-195 shed. Additionally, appropriate conservation measures of water and soil are extremely 196 essential and should be flexible and adaptable to changing insights on the impacts. 197

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