



Article

# Impact of Land Use Change on Flow and Sediment Yields in the Khokana Outlet of the Bagmati River, Kathmandu, Nepal

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**Abstract:** Land use changes are a key factor for altering hydrological response, and understanding its impacts can help to develop a sustainable and pragmatic strategy in order to preserve a watershed. The objective of this research is to estimate the impact of land use changes on Bagmati river discharge and sediment yield at the Khokana gauging station of the Kathmandu valley outlet. This study analyzes the impact of land use changes from the year 2000 to 2010 using a semi-distributed hydrological, Soil Water Assessment Tool (SWAT) model. The Load Estimator (LOADEST) simulates sediment loads on limited available sediment data. Sensitivity analysis is performed using the ParaSole (Parameter Solution) method within SWAT Calibration and Uncertainty Procedure (SWAT-CUP), which shows that Linear parameters for calculating the maximum amount of sediment that can be re-entrained during channel sediment routing is a most sensitive parameter that affect the hydrological response of the watershed. Monthly discharge and sediment data from 1995 to 2002 are used for calibration and remaining monthly discharge and sediment data from 2003 to 2010 are used for validation. Four statistical parameters including the Coefficient of Determination (R<sup>2</sup>), Nash-Sutcliffe Efficiency (NSE), RMSE-observations' standard deviation ratio (RSR), and Percentage Bias (PBIAS) are estimated in order to evaluate the model performance. The results show a very good agreement between monthly measured and simulated discharge data as indicated by  $R^2 = 0.88$ , NSE = 0.90, RSR = 0.34, and PBIAS = 0.03. The model shows nearly the same performance also with sediment data. The land use change data shows about a 6% increase in built-up areas from the years 2000 to 2010, whereas the remaining areas such as Forest, Shrub, Grass, Agriculture, Open Field, and Rivers/Lakes are decreased. The surface runoff contribution to stream flow and sediment yields are increased by 27% and 5% respectively. In the contrary, lateral flow contribution to stream flow and groundwater contribution to stream flow are decreased by 25% and 21%, respectively.

Keywords: GIS; land use change; SWAT; Bagmati watershed; Khokana; calibration; validation

#### 1. Introduction

Overexploiting natural resources for livelihood and economic activities is a common problem in developing countries Water resources are facing serious stress because of the rapid urbanization and soaring population growth rate, which has altered the hydrological response. Various hydrological indicators such as altering discharge, sediment, and nutrient yields can be used to investigate the impact of land use change. These indicators can be obtained either through on-site monitoring or using simulation models. On-site monitoring is generally unfeasible because it is labor intensive, time consuming, and expensive; therefore, the hydrological simulation model is gaining popularity to solve these problems. Selecting a proper model in the specific watershed has always been challenging, while field verification of the models' parameters is useful to researchers in selecting the model for their purposes.

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Land use change impacts various areas including hydrology, socio-economics, ecological, and environmental [1]. This is a triggering factor that leads to a change in the flow pattern due to the temporal variation of the discharge distribution [2]. Land use changes are a common problem in developing countries whose economies are generally based on agriculture [3]. Land use changes from agriculture to settlement alter the hydrological regime by increasing peak flow and decrease infiltration [4]. It accelerates soil erosion, which leads to various negative consequences such as increasing sedimentation, turbidity, and nutrients. Land use changes can lead to undesirable effects in the ecosystem, which cause water, soil, and air pollution [5].

The analysis of the land use change, which has an impact on the discharge, sediment, and nutrient yields, is enhanced by implementing modeling tools that incorporate both spatial and temporal dynamics. It is widely accepted that land use changes in the basin result in flooding, which is a major carrier of sediment [6]. The anthropogenic activities are mainly responsible for land use change in the river basin, and have been widely studied [7]. The hydrological model has become a useful tool in understanding the natural process in the river basin and its response of anthropogenic activities [8]. Geographic Information System (GIS)-based hydrological models are gaining popularity because of their capacity to easily incorporate spatial variability and estimate the land use change impact on the discharge. The GIS-based SWAT model is being used extensively in such conditions. SWAT has been widely used for simulating discharge, nutrient, and sediment from the basin [9,10]. This model has also shown to be an effective tool for assessing land use changes, water resources, and non-point pollution problems [11].

This paper is trying to evaluate the impact of land use change on the discharge, and sediment yields from the highly dense Kathmandu mountainous valley where approximately three million residents rely on the Bagmati river for their livelihoods and socioeconomic activities.

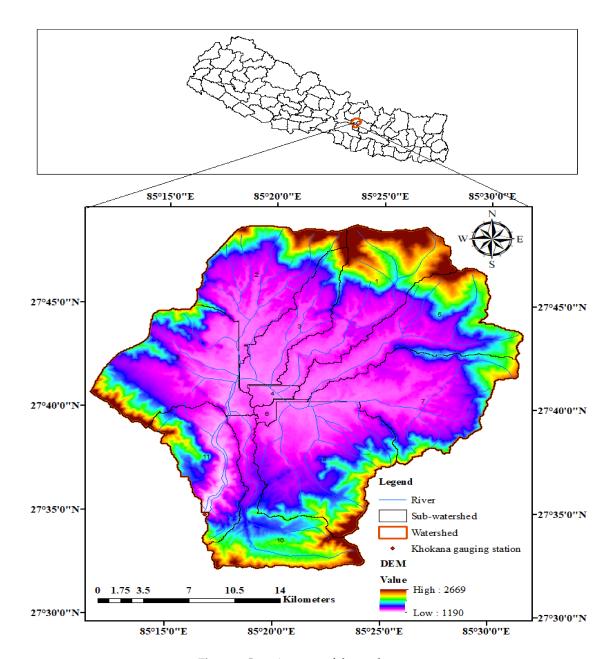
The influential factors of land use change are an increasing population, the expansion of urban areas, and climatic change, which is happening at alarming rates in Kathmandu. The expansion of built-up areas and overexploitation of natural resources to meet the demands of a soaring population will likely accelerate over the coming years. Therefore, scientific research is essential to estimate the impact of land use change on discharge. The finding of this research will disseminate knowledge to the policy makers to undertake effective measures for minimizing the undesirable impact of future land use change.

#### 2. Methodology

#### 2.1. Study Area

Bagmati is a medium-sized spring and monsoon rain-fed river flowing through the heart of the capital city, Kathmandu; it consists of many tributaries such as Bishnumati, Manohara, and Tukuchha (Figure 1).

The Bagmati River originates at the Mahabharat Range of Shivapuri at an altitude 2669 m a.s.l. and has a catchment area at the Khokana gauging station that is 629 km², which is situated at Latitude  $27^{\circ}37'44''$ , Longitude  $85^{\circ}17'41''$ , and 1190 m a.s.l. elevation. The mean monthly maximum air temperature reaches up to 29.3 °C and mean monthly minimum temperature reaches up to 0.9 °C. Likewise, the average annual rainfall is around 1300 mm [12]. In this research, the 2000 and 2010 years' land use data are used in order to investigate its impact on the discharge and sediment yield. Comparing land use data of years 2000 and 2010, it shows that all classes of land use areas are decreased except the built-up areas, which have significantly increased from 131.26 km² (20.87%) to 171.88 km² (27.32%) (Table 1 and Figure 2). The study area is divided into 11 sub-watersheds, which vary from 3.12 km² to  $95.22 \text{ km}^2$ , followed by  $49.49 \text{ km}^2$  sub-watershed number 11, which is situated at the outlet and 118 Hydrologic Response Unit (HRU) based on soil and land use data.



**Figure 1.** Location map of the study area.

**Table 1.** Land use classes in the Bagmati watershed at Khokana gauging station.

	Land Use					
	2000		2010		Difference	
	Area (km²)	%	Area (km²)	%	km <sup>2</sup>	%
Forest	229.23	36.44	223.00	35.45	-6.23	-0.99
Shrubland	0.53	0.08	0.005	0.001	-0.53	-0.08
Grassland	11.08	1.76	7.02	1.12	-4.06	-0.65
Agriculture	254.85	40.51	225.83	35.90	-29.03	-4.61
Open Field	1.57	0.25	1.21	0.19	-0.35	-0.06
Rivers/Lakes	0.53	0.08	0.12	0.02	-0.41	-0.07
Built-Up Areas	131.26	20.87	171.88	27.32	40.62	6.46
Total	629.05		629.05			

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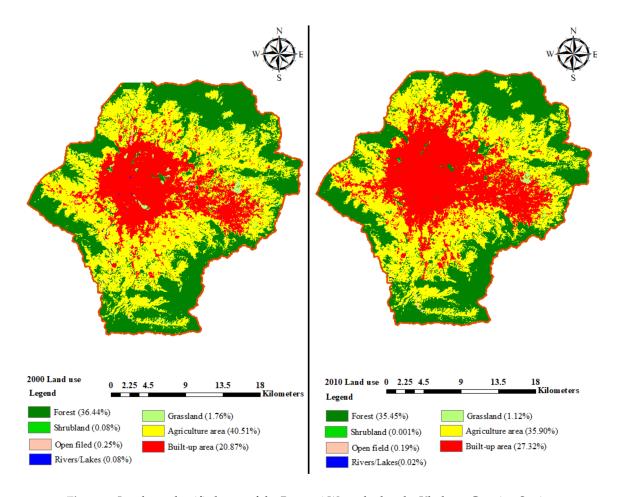


Figure 2. Land use classified map of the Bagmati Watershed at the Khokana Gauging Station.

## 2.2. SWAT Model

The SWAT model is used in this research in order to estimate the impact of land use change on discharge and sediment yield in the Bagmati river at the Khokana Gauging Station. SWAT is a conceptual, physically-based, semi-distributed hydrological model. This was developed in order to estimate the impact of land management practices on hydrology and water quality yields in large and complex watersheds having heterogeneous soil and land use conditions [13].

Runoff generation is an important part of the model that uses the Natural Resources Conservation Service (NRCS), and the formal Soil Conservation Service curve number (SCS-CN) method to calculate surface runoff and infiltration [14]. The CN depends on soil type, land use, and management practice, which distinguishes the amount of infiltration and surface runoff in a particular rain event. Higher CN value infers higher runoff and less infiltration [15]. The total discharge at the outlet is the discharge accumulated from each HRU, sub-surface flow, lateral flow, and return flow. The flow in SWAT is either routed through Muskingum or a variable storage coefficient method. Similarly, sediment yield from each HRU and erosion is estimated from the Modified Universal Soil Loss Equation (MUSLE), which uses a modification of Bagnold's sediment transport equation for routing through the channel.

The GIS-based SWAT model has given researchers freedom to incorporate spatial variation and predict the hydrological process such as surface discharge, percolation, deep aquifer flow, evapotranspiration (ET), and channel routing [16]. The model strength for studying the long-term impact of land use change on discharge has given preference to select the SWAT model in this research. This is also a robust model for simulating sediment load and has been widely implemented in different parts of the world. The model performed well in monthly sediment loads rather than daily in the North Bosque River Watershed in North Texas [17]. The model divides the watershed into a number of

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sub-watersheds, which are connected by surface flow [18]. These sub-watersheds are further divided into number of hydrological response units (HRU) based on topography, land use, and soil input data. The model lumps the parameters into HRUs. Traditionally, HRUs are defined by the coincidence of soil type [19] and land use.

### 2.3. Model Setup

#### Data

The Digital Elevation Model (DEM), soils, land use, and weather are essential inputs data for the SWAT model. A Digital Elevation Map (DEM) of 30 m × 30 m resolution of the watershed was downloaded from NASA's web site (https://reverb.echo.nasa.gov/reverb/). This is one of the critical input data required by the SWAT model to delineate the watershed and sub-watershed, and analyze the drainage pattern, slope, stream length, and width of the stream [20]. Land use maps of the years 2000 and 2010 are used to estimate the effect of land use change during these periods and were obtained from the International Centre for Integrated Mountain Development (ICIMOD, http://rds.icimod.org/). The soil data that defines the physical properties of the soil layers are obtained from the United State Department of Agriculture (https://www.nrcs.usda.gov/wps/portal/nrcs/detail/soils/survey/?cid=nrcs142p2\_053627).

Discharge data of the Khokana Gauging Station and meteorological data such as daily rainfall, minimum and maximum temperature, solar radiation, wind speed, and relative humidity of the Kathmandu Airport and Khumaltar stations were obtained from the Department of Hydrology and Meteorology (DHM). Only 2003 to 2006 sediment concentration data are obtained from DHM, which are simulated to the calibration and validation period using LOADEST. The source and description of the data used in the SWAT model are highlighted in the Table 2.

Data	Duration	Resolution	Source
Digital Elevation Model (DEM)		30 m × 30 m	NASA
Land use	2000 and 2010		ICIMOD
Soil	2000		NASA
Meteorological data of Kathmandu airport and Khumaltar	1992-2010		DHM
Discharge data of Khokana gauging station	1995-2010		DHM
Sediment data of Khokana gauging station	2003-2006		DHM

Table 2. Input data for the SWAT model at Khokana gauging station.

#### 2.4. Sensitivity Analysis

ParaSole (Parameter Solution) optimization method within the SWAT Calibration and Uncertainty Procedure (SWAT-CUP) is used for sensitivity analysis with 1000 trials. Sensitivity analysis is an important part of the model that reduces the number of parameters for calibration. This approach determines the most significant and sensitive parameters altering the water quantity and quality yields. The calibrated parameter range is used as a default parameter values in the Parasole. All parameters are selected for the sensitivity analysis. The results show that the SPCON is a most sensitive parameter (Rank 1) followed by EPCO, ESCO, GW\_REVAP, SPEXP, REVAPMN, SOL\_AWC, SURLAG, ALPHA\_BF, RCHRG\_DP, GW\_DELAY, and CANMX. The remaining parameters have no significant effect on the stream flow and sediment simulation (Table 3).

**Table 3.** List of calibrated parameters and their values.

Calibration					
Parameters	Definition	Unit	Range of Values	Calibrated	Rank
SPCON(.bsn)	Linear parameter for calculating the maximum amount of sediment that can be reentrained during channel sediment rounting		0.0001 to 0.01	0.01	1
EPCO(.hru)	Plant uptake compensation factor		0.01 to 1	0.1	2
ESCO(.hru)	Soil evaporation compensation factor		0.01 to 1	0.1	3
GW_REVAP(.gw)	Groundwater revap coefficient		0.02 to 0.2	0.02	4
SPEXP(.bsn)	Exponent parameter for calculating sediment re-entrained in channel sediment routing		1 to 2	0.72	5
REVAPMN(.gw)	Threshold depth of water in the shallow aquifer for "revap" or percolation to the deep aquifer		0 to 500	100	6
SOIL_AWC(.sol)	Available soil water capacity		0-0.12	0.02	7
SURLAG(.bsn)	Surface runoff lag coefficient	day	1 to 12	5	8
APHA_BF(.gw)	Baseflow alpha factor for recession constant	ř	0 to 1	0.048	9
RECHRG-DP(.gw)	Deep aquifer percolation fraction		0 to 1	0.05	10
GW_DELAY(.gw)	Groundwater delay	day	0 to 500	31	11
CANMX(.hru)	Maximum canopy storage	-	0 to 100	1	12
CH-N2(.mgt)	Manning's roughness coefficient for the main channel		0 to 0.1	0.1	13
GWQMN(.gw)	Threshold depth of water in the shallow aquifer for return flow		0 to 5000	1000	14
CN2(.mgt)	SCS runoff curve number		39 to 98	$\pm 10$	15
CH-K2(.rte)	Effective hydraulic conductivity in the main channel		0 to 150	2	16
CH-EROD(.rte)	Channel erodibility factor		0 to 1	0.18	17
CH-COV(.rte)	Channel cover factor		0 to 1	0.18	18

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SPCON is a linear parameter that estimates the maximum amount of sediment transports from the reach segment during the channel sediment routing, and an exponential parameter (SPEXP) is for calculating sediment re-entrained. These parameters affect the movement and separation of the sediment fraction in the reach and are adjusted within their acceptable limit to match the measured sediment yield. Baseflow recession coefficient (ALPHA\_BF) is a direct index of the groundwater flow response to change in recharge [21]. This factor determines the ratio corresponding to the base and surface flow contribution to the total flow. GW\_DELAY is a lag time between water exiting from the soil profile and entering the shallow aquifer. Deep aquifer percolation fraction (RCHRG\_DP) is a fraction from the root zone that recharges the deep aquifer. Plant uptake compensation factor (EPCO) shows the relation between water uptake by the plant to meet its transpiration demand and the amount of water available in the soil.

ESCO adjusts the evaporation depth distribution of the soil in order to show the effect of capillary action, crusting, and cracking [18]. GW\_REVAP explains that the fraction of water that can be moved from a shallow aquifer into the overlaying unsaturated soil layer is a function of evapotranspiration. REVAPMN is a threshold water depth in the aquifer for "revap" or percolation to a deep aquifer. SOL\_AWC is a key soil parameter associated with generating runoff and evaporation [22]. A higher value of it infers a higher water holding capacity of soil, which means less water available for the surface runoff and percolation. SURLAG is a sub-watershed level parameter and controls the fraction of the total water available that is allowed to enter the stream in a specific time period. CANMX is the maximum canopy storage that significantly affects infiltration, runoff, and evapotranspiration because of reduction in the erosive energy of droplets and traps a portion of the rainfall within the canopy.

#### 2.5. Model Calibration and Validation

The model performance is generally evaluated with the data collected at the outlet of the watershed. Bagmati River in Khokana is only an outlet of Kathmandu Valley and is a sampling site. Daily discharge and few years sediment concentration data are obtained from DHM. Sediment data are simulated to the entire calibration and validation period using the LOADEST regression model [23]. This regression model simulates nutrients and sediment load in the discharge using time series discharge and constituent concentration data as inputs. This model is widely accepted for simulating load from a limited amount of nutrient and sediment data. Monthly time step discharge and sediment data are used for calibration and validation.

In order to understand the uncertainty characters of the basin, three years (1992 to 1994) of data are used for the warm up period. Monthly data from 1995 to 2002 are used for calibration and remaining data of from 2003 to 2010 are used for validation. Calibrated parameters of the model are shown in Table 3. Manual calibration and validation techniques are adopted by varying parameters within their acceptable ranges, assuming that the optimal set of parameters exists for the model for describing hydrology of the study area that is shown by the good agreement between measured and simulated data both in calibration and validation period with discharge and sediment data. This is accomplished by comparing the measured data with the SWAT output of the same duration [24]. Model robustness and performance are evaluated both in the calibration and validation graphically and quantified using statistical parameters [25]. These statistical parameters are: Coefficient of determination (R<sup>2</sup>), Nash–Sutcliffe simulation efficiency (NSE), RMSE-observations standard deviation ratio (RSR), and Percentage of Bias (PBIAS) [26]. The coefficient of determination (R<sup>2</sup>) explains the degree of collinearity between the measured and simulated data [27]. Typically,  $R^2$  greater than 0.5 is considered an acceptable limit for model performance [28,29]. The model performance is classified as very good, good, satisfactory, and unsatisfactory if the NSE value is larger than 0.75, between 0.65 and 0.75, between 0.5 and 0.65, and less than 0.5, respectively [27]. The zero RMSE value indicates perfect agreement between measured and simulated data. RMSE less than 0.5 of standard deviation of the measured data indicates good model performance [30,31]. PBIAS indicates the average tendency of the simulated data to be larger or smaller than the observed data. Its zero value indicates perfect

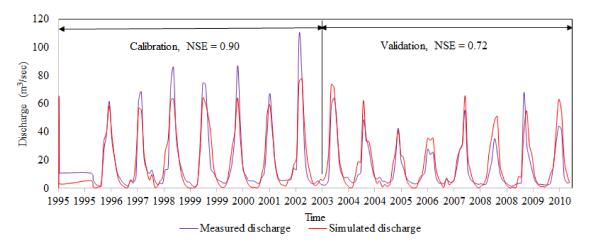
matching of measured and simulated data. Positive values indicate underestimation bias and negative value indicates overestimation bias [32].

#### 3. Results and Discussion

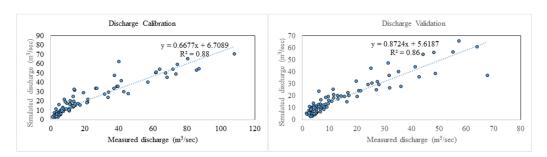
In the calibration period, the model was unable to capture the peak of measured discharge. In contrast, the model over-forecasts the peak discharge in the validation period. This problem may be increasing in built up area creates impervious layers which reduces infiltration and percolation that cause increases in surface runoff. This is probably a shortcoming of the SWAT model for forecasting the hydrograph of the present watershed. However, results indicate the good agreement between the measured and simulated discharge for both calibration and validation periods. The statistical parameters  $R^2$ , NSE, RSR, and PBIAS are estimated as 0.88, 0.90, 0.34, and 0.03 in calibration, and 0.86, 0.72, 0.53, and -0.25 in the validation period, respectively (Table 4). These values indicate a very good level of model performance with the discharge data. The graphical comparison of measured and simulated discharge data during calibration and validation period, regression analysis for finding  $R^2$ , and comparison of the yearly discharge data are shown in Figures 3–5, respectively.

Table 4. Calibration and Validation statistics of the SWAT model using monthly discharge data.

		Evaluated	l Statistics	
Stage of Model	R <sup>2</sup>	NSE	RSR	PBIAS
Calibration (1995–2002)	0.88	0.90	0.34	0.03
Validation (2003–2010)	0.86	0.72	0.53	-0.25



**Figure 3.** Observed and simulated monthly discharge data of Bagmati Watershed at the Khokana Gauging Station.



**Figure 4.** Scatter plots of observed and simulated monthly discharge data of the calibration and validation period.

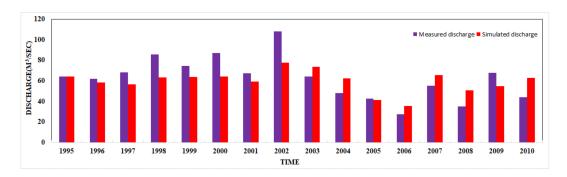
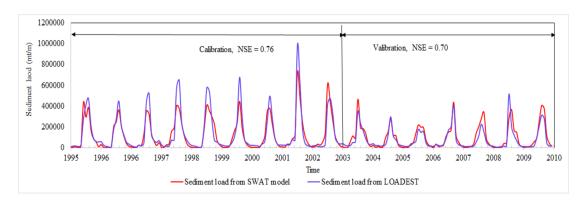
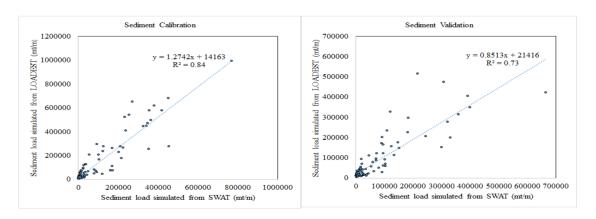


Figure 5. The relationship between the calibrated and simulated yearly discharge data.

Similarly to the discharge, the performance of SWAT model is also evaluated graphically and statistically with the monthly sediment data. The graphical comparison of monthly sediment data during calibration and validation period, regression analysis for finding  $R^2$ , and comparison of the yearly sediment data are shown in Figures 6–8, respectively. Table 5 shows statistical parameters estimated in order to evaluate the SWAT performance with the sediment data, which shows 0.84, 0.76, 0.49, and 0.30 in calibration and 0.73, 0.70, 0.55, and -0.21 in validation, respectively. The statistical parameters indicate that the SWAT model shows a very good level of performance also with sediment data. Four parameters such as CH\_EROD, SPCON, SPEXP, and CH-COV are mainly adjusted to their default limit in order to match the LOADEST simulated sediment load with the SWAT simulated (Table 3).



**Figure 6.** Observed and simulated monthly sediment load of Bagmati Watershed at the Khokana Gauging Station.



**Figure 7.** Scatter plots of observed and simulated monthly sediment data in the calibration and validation period.

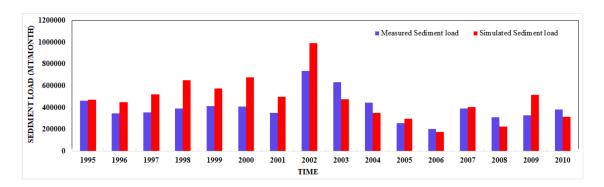


Figure 8. The relationship between the calibrated and simulated yearly sediment data.

Table 5. Calibration and validation statistics of the SWAT model using monthly sediment data.

0. (1.1.1.1	Evaluated Parameters				
Stage of Model	$R^2$	NSE	RSR	PBIAS	
Calibration (1995–2002)	0.85	0.76	0.49	0.30	
Validation (2003–2010)	0.73	0.70	0.55	-0.21	

Table 3 indicates the list of parameters adjusted during calibration with discharge and sediment data. Land use data of 2000 and 2010 show some noteworthy changes in the land use classes. The results show that, during these periods, all land use classes (Forest, Shrub, Grass, Agriculture, Open Field, and Rivers/Lakes) are reduced by  $40.62~\rm km^2$  (6.46%), which are converted to Built-Up Areas. The impact of land use change is estimated by running the model using 2000 land use along with climate data from 1992 to 2002 and using 2010 land use along with climate data from 2000 to 2010 keeping the parameters, DEM, and Soil data constant.

The research has found the greatest impact on the surface runoff contribution to stream flow, which is increased from 172 mm/year to 219 mm/year, representing 27% (Table 6). The built-up areas increase by  $40.62 \text{ km}^2$  (6.46%) is attributed to the increasing surface runoff contribution to stream flow. The increase in built-up areas creates impervious layers, which reduces infiltration and percolation that cause increases in surface runoff. This finding supports the statement that an increase in built-up areas increases impervious surfaces, which accelerates surface runoff [33,34].

**Table 6.** Estimated discharge, sediment yield at different land uses.

Commont	Land	9/ Chara	
Component -	2000	2010	% Change
Surface runoff contribution to stream flow SURQ (mm/year)	171.99	219.17	+27.43
Lateral flow contribution to stream flow LATQ (mm/year)	455.51	340.14	-25.33
Ground water contribution to stream flow (mm/year)	390.66	309.09	-20.88
Sediment yield (mt/ha)	2.99	3.15	+5.41

Lateral flow contribution to stream flow is decreased from 455 mm/year to 340 mm/year, representing 25%. Sediment yield is increased from 2.99 mt/ha to 3.15 mt/ha, representing 5%, which is supported by the Wolman statement as the urban landscape is built out, the sediment yield increase [35]. In the contrast, groundwater contribution to stream flow is decreased from 391 mm/year to 309 mm/year (21%). This is attributed to less soil infiltration and more surface runoff, and an increasing rate of groundwater abstraction from the shallow and deep aquifer for water supply and irrigation. These scenarios are happening in the research area and are expected to accelerate in the future.

#### 4. Conclusions

In this research, a GIS-based SWAT model is used to estimate the impact of land use changes on the discharge and sediment yield of the Bagmati river at the Khokana Gauging Station in Nepal. Online data sources are used to download DEM, Soil, and Land use data, whereas discharge and meteorological data are obtained from the Department of Hydrology and Meteorology (DHM). Data from 1992 to 1994 are used for the warm up period, whereas 1995 to 2002 data are used for calibration and the remaining 2003 to 2010 data are used for validation. Four statistical parameters (R<sup>2</sup>, NSE, RSR, and PBIAS) are used to evaluate the level of agreement between measured and simulated discharge and sediment data, which are estimated as 0.88, 0.90, 0.34, and 0.03 in the calibration period and 0.86, 0.72, 0.53, and -0.25 in the validation period, respectively, with the discharge date. Nearly similar model performances are also achieved with the sediment data. These values indicate that the model performs very well both with the discharge and sediment data. Comparison of land use data between 2000 and 2010 shows the noteworthy land use changes where all the land use classes are decreased except built-up areas, which are increased by 40.62 km<sup>2</sup> (6.46%). Therefore, surface water contribution to stream flow and sediment yields is increased by 27% and 5%, respectively. It is believed that expansion of the urban attributes increases surface water contribution to stream flow, which carries more sediment. In the contrast, groundwater contribution to stream flow is decreased by 25% because of decreasing infiltration due to the expansion of urban areas.

The continuation of land use change is a serious threat in the Bagmati Watershed. Due to continuous sediment accumulation, the river bed has been risen. Therefore, small flood is more likely to topple the river and are become a too serious threats to the residents particularly in the low land, and impact on their economic activities. Contamination of the river due to the direct disposal of effluent and sewerage results in extinction of aquatic life and leads to a serious threat for those downstream who are dependent on the river for their livelihood. Therefore, a sustainable and pragmatic strategy should be developed in order to preserve the watershed. A well-calibrated SWAT model can simulate discharge and sediment related to the impact of land use change, which is expected to be a valuable asset and information to the policy makers.

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