Evaluating Hydrological Model Performances on Stream Flow Simulation with Agricultural Management

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Abstract
Two comprehensive agricultural watershed-scale hydrological models: Soil and Water Assessment Tool (SWAT), and Annualized Agricultural Non-Point Source Pollution Model (AnnAGNPS), have been widely used for evaluating agricultural Best Management Practice (BMP). In this study, models were developed for the Big Sunflower River Watershed (BSRW) in Mississippi, USA. Both hydrological models were successfully calibrated and validated for predicting monthly stream flow discharges from three USGS gages (Merigold, Sunflower, and Leland) within the BSRW. The model performances were reasonable with R² and NSE values range from 0.45 to 0.84 for SWAT, and from 0.34 to 0.66 for AnnAGNPS. Both models predicted lower stream flow discharge with conservation tillage practice as compare to conventional tillage practice BMP evaluated in this study. Overall, results showed that both SWAT and AnnAGNPS models successfully simulated hydrological conditions of the watershed with agricultural BMP, and these models can be useful for making hydrological decisions related to agricultural watershed managements.

Keywords: SWAT, AnnAGNPS, tillage, streamflow, hydrology, BMP

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1. Introduction
With greater understanding of watershed hydrology, various modeling tools were developed to simulate watershed hydrological processes. Watershed models have their emphasis on different aspects of watershed representation such as TR-55 (NRCS, 1986) and TR-20 developed by Natural Resources Conservation Service (NRCS) are common watershed runoff models for single event focusing on small urban or urbanizing watershed. Hydrologic Modeling System (HEC-HMS - Feldman, 2000) developed by the United States Army Corps of Engineers Hydrologic Engineering Center is a runoff simulating tool with several options including continuous simulation. Agricultural activities such as intensive fertilization and tillage operation could have great influences on surface water quality as well as water quantity (Ayers & Westcot, 1985; Shipitalo & Edwards, 1998; Vaché et al., 2002). Agricultural Best Management Practices (BMPs), as solutions helping reduce the negative impacts of crop production, have been widely implemented in United States. In order to evaluate BMPs impacts on watershed processes, watershed simulation with BMP scenarios have considered as an effective approach (Arnold et al., 1993; Arabi et al., 2008). The selected simulating tool need to account for both hydrologic and agricultural activities during simulation.

Within various modeling tools, there are two modeling tools fulfilling the requirements have the most applications in U.S. The first one, Soil and Water Assessment Tool (SWAT), a process-based watershed-modeling tool, allows simulating agricultural activities such as agricultural tillage operation and crop rotation (Ullrich & Volk, 2009; Kirsch et al., 2002). Previous studies have demonstrated the abilities of SWAT model to assess impacts of agricultural activities on surface water quality and quantity (Kirsch et al., 2002; Arabi et al., 2008; Neitsch et al. 2011). Another one is Annualized Agricultural Non-Point Source Pollution Model (AnnAGNPS). It is a user-friendly watershed-modeling tool emphasizing on agricultural area, which also contains modules simulating BMPs such as conservation tillage and crop rotation scenarios (Yuan et al., 2001; Srivastava et al., 2002; Bingner et al., 2015).

Performances of different watershed models on hydrology simulations have been compared by some previous studies. Van Liew et al. (2003) compared the performance of HSPF and SWAT on runoff simulation in an agricultural watershed. They compared HSPF and SWAT model performances predicting various time-scale stream flows using standard statistical parameters such as deviation and coefficient of efficiency. They reported that the SWAT model was a more robust tool than HSPF in their study. Nasr et al. (2007) compared SWAT, HSPF, and Systeme Hydrologique European TRANsport (SHETRAN) model efficiency in predicting phosphorus yield by comparing model results with observed data. They recommended the SWAT model as a good tool to estimate phosphorus yield. Parajuli et al. (2009) compared the performances of SWAT and AnnAGNPS using various statistical parameters including coefficient of determination (R²), and Nash–Sutcliffe model efficiency coefficient (NSE). They also summarized the differences in methods used to simulate same physical process in two models. Both models had fair to good performances simulating stream flow and sediment yield. All of these previous
studies were mainly focused on comparing the performance among models. Performances of different models indicate the ability to simulate watershed condition, which is important for choosing appropriate models for a specific watershed application. After selecting an appropriate model for a specific application, making decisions based on the scenario simulations is the next step. Thus, the evaluations of the results of concerned scenarios from different models are necessary. Thus, the specific objectives of this study included: (i) evaluating models performance for predicting stream flow discharges, and (ii) comparing models for simulating stream flow discharge with tillage practices.

2. Methodology

2.1 Model parameters

The Big Sunflower River Watershed (BSRW, Fig. 1) was the target watershed in this study located in Mississippi with crop production as mainstay of its economy (Parajuli & Jayakody, 2012). The watershed contains approximately 80% croplands, 15% of forest, and other land use areas (USDA/NASS, 2016). With the agricultural water use as the main concern in simulation, to ensure that two hydrological models (SWAT and AnnAGNPS) were comparable, the same input data including both for hydrology simulation and agricultural management practices were applied in both models. The watershed simulation data include Digital Elevation Model (DEM) (USGS, 1999), SSURGO soil type (USDA, 2005), Crop Data Layer (CDL - USDA/NASS, 2016) for land use and cover, and climate information from Climate Forecast System Reanalysis (CFSR) database (NCDC, 2016) and Global Historical Climatology Network (GHCN) – daily database (NOAA, 2016). The general crop management practices in the watershed were generated based on previous publication (Parajuli et al., 2013); and on-site annual crop trial reports (MAFES, 2000-2014).

There are similar methodologies used in both models to describe physical modeling procedures. Both models used original or revised version of the Soil Conservation Service (SCS) curve number (SCS, 1985) based method to simulate surface runoff (King et al., 1999; Parajuli et al., 2009); and Universal Soil Loss Equation (USLE - Wischmeier & Smith, 1978) based method to simulate sediment yield (Borah & Bera, 2003; Parajuli et al., 2009). Thus, the curve number and USLE method related factors were kept consistent in both models.

2.2 SWAT model

Soil and Water Assessment Tool (SWAT) is a physical-based hydrological and water quality model (Arnold et al., 1993). It allows predicting impacts of agricultural management practices (e.g. short-term, long-term) on stream water quantity and quality in large and complex watershed. Moreover, it considers spatial parameters, such as soil type and land use and cover, as semi-distributed, thus, provides high computational efficiency using various mathematical equations to describe each hydrological process in hydrological response unit (HRU). HRU as the unit of simulation is defined by variation of surface elevation, land use and cover, and soil type (Neitsch et al., 2011). In addition, the model is designed to be user-friendly with readily available input data (e.g. surface elevation, land use and cover, soil type, weather). After initial introduction during 1980’s, it has been continuously developed, which helped to improve the estimation under various locations (e.g. freezing region, high elevation region), specific agricultural management practices (e.g. tile drainage), and with the change of climate and land use.

In this study, the BSRW was divided into 1,799 HRUs with 22 sub-basins based on spatial variation of soil type, land use and cover, and surface elevation (Fig. 1). The threshold area values of 5% for soil type, 1% for land use, and 5% for slope length of each sub-basin area were used to define HRU. Calibration process included both auto and manual calibration. SWAT-Cup SUFI2 algorithm was applied to obtain simulation with highest R2 and NSE values by comparing the simulated model results and observations at the USGS gaging stations. This study utilized monthly stream flow data from 2008 to 2015. Afterward, SCS curve number was manually calibrated.

2.3 AnnAGNPS model

Annualized Agricultural Non-point Source model (AnnAGNPS) is a process-based, watershed scale, continuous time pollutant-loading model that focus on agricultural watershed (Binger et al., 2018). The AnnAGNPS model utilizes surface elevation, land use, soil type, weather information, and pollution input type data for each spatial modeling unit of the watershed to quantify fate and transport process of non-point source loads (e.g. pesticide, sediment, and nutrient). The AnnAGNPS model provide emphasis on combining various simulating modules including upstream agricultural management practices (e.g. irrigation, tile drainage, tillage); and hydrological processes (e.g. surface runoff, subsurface flow) to demonstrate the impact of pollutant sources on stream water quality. The AnnAGNPS model has been available or developed during 1980’s. Recent updates of the AnnAGNPS model are focused on additional features including gully erosion simulation to improve overall model performances.

In this study, the BSRW was divided into 193 cells as the modeling units in the subsequent simulation in AnnAGNPS model, based on surface elevation by TopAGNPS module (Fig. 1). The spatial data including soil type, weather stations and land use and cover were summarized and utilized similar to the SWAT model to the
AnnAGNPS cells. The manual calibration was performed for AnnAGNPS model.

2.4 Tillage management scenarios
In this study, two tillage management scenarios namely conventional and conservation tillage management practices were simulated using SWAT and AnnAGNPS models. Tillage operation affects surface water runoff (Shipitalo et al., 1998) and water quality including nitrogen, phosphorus, and sediment (Tan et al., 2002; Vaché et al., 2002). Conventional tillage practice was applied in about 70% of the area of Mississippi Delta region (Snipes et al., 2005), while conservation tillage is a commonly implemented BMP in U.S to prevent soil erosion. Thus, the baseline scenario was with conventional tillage, while conservation tillage management was considered as a BMP scenario. In both model, SCS curve number was considered as the parameter distinguishing conventional and conservation tillage management scenarios (Kirsch et al., 2002). In addition, its value was based on soil type and land use combination, which was set based on previous study by Ni and Parajuli (2018).

2.5 Model calibration and validation
Both SWAT and AnnAGNPS models were calibrated and validated by monthly stream flow discharge from USGS gages. Models were calibrated using monthly stream flow discharge from Merigold station, while validated at Sunflower and Leland gages. The $R^2$ and NSE values were used to evaluate each model performance (Parajuli, 2012; Kim and Parajuli, 2014).

3. Results and Discussion
3.1 Calibration and validation
The SWAT and AnnAGNPS models calibration results for stream flow showed reasonable performance based on $R^2$ and NSE values (Moriasi et al., 2015; Table 1). The SWAT model showed higher $R^2$ and similar NSE value when compared to AnnAGNPS. The similar results were expected since both models used SCS curve number (SCS, 1985) based methods to estimate surface runoff (Bingner et al., 2015; Neitsch et al., 2011).

The SWAT and AnnAGNPS models simulated monthly stream flows were compared with the observed monthly stream flow at Sunflower gage station (Fig. 2). The model simulated average monthly stream flows were 38.9 cms by SWAT, 29.1 cms by AnnAGNPS, and 28.5 cms by observation. The $R^2$ values in SWAT simulation were higher than AnnAGNPS, while NSE values were similar. With using same curve number, homogeneous modeling units, same land use, and soil type in both models, the differences might be resulted from the different modifications of SCS curve number methods used to simulate surface runoff in the SWAT and AnnAGNPS models. The AnnAGNPS model considered the retention parameter that is an important parameter in SCS curve number method varies with soil moisture content, while SWAT model considered the retention parameter varies with plant potential evapotranspiration in addition to soil moisture content (Bingner et al., 2015; Neitsch et al., 2011).

3.2 BMP Scenarios comparison
In order to obtain comparable results for the BMP scenarios simulated using SWAT and AnnAGNPS models, the comparison were conducted at three USGS gaging stations, where the baseline scenario was compared with observed stream flow data during calibration and validation processes. Fig. 3 shows the scenario comparisons between conventional tillage and conservational tillage using SWAT and AnnAGNPS models at three USGS gaging stations. The change variation among three stations were small with average reduction of monthly stream flow of 43%, 42% and 32% for Merigold, Sunflower and Leland respectively by using the AnnAGNPS model, and 12%, 11% and 13% by using the SWAT model. Both models showed the reduction of stream flow in conservational tillage as compare to conventional tillage. The stream flow was more sensitive to curve number in the AnnAGNPS model than in the SWAT model. This might be caused from the difference between modeling units in two models. The modeling units in the AnnAGNPS model were dependent on the surface elevation, while the modeling units in the SWAT model were dependent on the combination of land use, soil type and slope length. The SCS curve number method calculates surface runoff by considering the different infiltration conditions of land cover, soil group and land treatment (SCS, 1985). Thus, the modeling units in the SWAT model were more compatible with the SCS curve number method. The AnnAGNPS model considers majority of land use and soil type as its modeling units, which may have caused that changing of curve numbers might be not be compatible with the tillage management occurring on the cropland.

4. Conclusion
Both the SWAT and the AnnAGNPS models determined acceptable stream flow simulation results during model calibration and validation processes with $R^2$ and NSE values range from 0.45 to 0.84 for the SWAT, and from 0.34 to 0.66 for the AnnAGNPS. Despite similar inputs in the two hydrological models, simulated stream flow outputs were slightly varied with BMP scenarios due to different modifications of the SCS and USLE methods applied in the SWAT and the AnnAGNPS models. For tillage scenarios, both model showed the reduction of
stream flow with conservational tillage scenario as compared to conventional tillage.

This research study resulted that both hydrological models have ability to simulate BMP such as tillage management practices at the watershed scales, but the modeler need to pay attention to the different response from different models. The SWAT and AnnAGNPS models have different sensitivities to the parameters and inputs, which modelers may need to understand while applying model for the hydrological investigations.

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Table 1. Stream flow calibration and validation performance for SWAT and AnnAGNPS models

|                      | SWAT            | AnnAGNPS       |
|----------------------|-----------------|----------------|
|                      | $R^2$           | NSE            | $R^2$           | NSE            |
| Calibration--Merigold| 0.69            | 0.64           | 0.53            | 0.51           |
| Calibration--Sunflower| 0.65            | 0.45           | 0.57            | 0.51           |
| Calibration--Leland  | 0.73            | 0.61           | 0.59            | 0.54           |
| Validation--Merigold| 0.68            | 0.50           | 0.57            | 0.55           |
| Validation--Sunflower| 0.84            | 0.70           | 0.46            | 0.66           |
| Validation--Leland   | 0.70            | 0.46           | 0.48            | 0.34           |
Figure 1. Location of watershed delineated by (a) SWAT, and (b) AnnAGNPS models with USGS gages and weather stations

Figure 2. Stream flow predicted by SWAT and AnnAGNPS models as compared to observed data
Figure 3. Conventional and conservational tillage stream flow simulated by SWAT and AnnAGNPS for: (a) Merigold (b) Sunflower, and (c) Leland gage stations.