

Runoff and sediment yield modeling in a medium-size mediterranean watershed

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Abstract

The AnnAGNPS model was used to estimate runoff, peak discharge and sediment yield at the event scale in the Carapelle watershed, a Mediterranean medium-size watershed (506 km²) located in Apulia, Southern Italy. The model was calibrated and validated using five years of runoff and sediment yield data measured at a monitoring station located at Ordona - Ponte dei Sauri Bridge. A total of 36 events was used to estimate the output of the model during the period 2007-2011, in comparison to the corresponding observations at the watershed outlet. The model performed well in predicting runoff, as was testified by the high values of the coefficients of efficiency and determination during the validation process. The peak flows predictions were satisfactory especially for the high flow events; the prediction capability of sediment yield was good, even if a slight over-estimation was observed. Finally, the model was used to evaluate the effectiveness of different Management practices (MPs) on the watershed (converting wheat to forest, using vegetated streams, crop rotation corn/sovbean, no tillage). While the maximum reduction in sediment yield was achieved converting wheat to forest, the best compromises between soil conservation and agriculture resulted to be crop rotations.

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Introduction

Soil erosion can lead to reduction of soil fertility, loss of nutrients, and declines of crop yields in farmlands. In a review of mechanized agricultural systems in which wheat, corn, soybean and barley were planted, Bakker et al. (2004, 2005) found that on average, soil erosion reduced crop productivity by about 4% for each 10 cm of soil lost. In recent years, it is widely recognized that more site-specific approaches are needed to assess variations in erosion susceptibility in order to select the most suitable land management method (Pandey et al., 2008). Structural and non structural measures to control negative impacts of runoff and erosion processes can be properly addressed through reliable prediction models. Although there has been considerable effort, additional work is needed to assess and improve the reliability of available prediction models in different environmental contexts. Reliable prediction models can help to select the most practical and effective tools in reducing erosion problems and developing appropriate land use planning (Licciardello et al., 2007).

Continuous distributed simulation models (as AnnAGNPS, WEPP, SWAT) provide great advantages as they allow watersheds response to be studied over a longer time period and can help to select the most practical and effective tools in reducing erosion problems (Zema *et al.*, 2010).

AnnAGNPS has been implemented to assess runoff and water quality as well as sediment yield in small to large watersheds under different environmental conditions. Assessments of model performance, frequently coupled with calibration/validation trials in monitored watersheds ranging from 32 ha to 2500 km², have recently been published (Licciardello *et al.*, 2007; Yuan *et al.*, 2008; Parajuli *et al.*, 2009; Zema *et al.*, 2010)

Few studies have been conducted using AnnAGNPS at the watershed scale in semi-arid environments. Here the quantification of sediment transport is quite difficult due to the high variability of rainfalls and water flows. The semi-arid areas receive low annual rainfall, have nutrient-poor soils, short-grass or shrublands vegetation and the actual evapotranspiration is recognised as the main hydrologic loss (50-60 % of the mean annual rainfall). In particular, after long-lasting periods with no rains and high temperatures, the soil can be very dry at the time of the storm. On the contrary in humid and hyper-humid climates, because of the high number of rainy days, the rainfall responsible for the flood peak is likely to occur when the infiltration rate into the soil is close to the gravitational capacity (Fiorentino and Iacobellis, 2001).

With particular reference to the Mediterranean environment, tests of the single-event model AGNPS were carried out in Italy, where hydrological effects of different land uses in an alpine environment (Cazorzi and Dalla Fontana, 1996; Cazorzi, 1996; Lenzi and Di Luzio, 1997) as well as soil erosion in southern small watersheds characterized by ephemeral streams (Morgagni *et al.*, 1993; Licciardello and Zimbone, 2002) were successfully predicted.

In order to estimate erosion and sediment transport processes in semi-arid environments the AnnAGNPS - Annualized Agricultural Non-



point Source model was selected and applied in the Carapelle watershed (Southern Italy). The model structure is suitable as it contains both empirical and quasi-physically based algorithms, it is fully distributed with land surface runoff and sediment processes modelled for the individual grid cell, and the output is routed to the catchment outlet. The data requirements and computational complexity of the AnnAGNPS allow the model to be used as a tool for watershed management planning. (Bisantino *et al*, 2013).

The AnnAGNPS model was developed to analyze and provide estimates of runoff with primary emphasis upon sediment and nutrients transport from agricultural watersheds, and to compare the effects of various conservation alternatives. Simulations under various combinations of different scenarios of land and water management can provide comparative analysis of different options and prove to be very useful as a guide to what Best Management Practices (BMPs) can be adopted to minimize pollution from point and nonpoint sources (Shrestha et al., 2006).Best Management Practices (BMPs) are structural and nonstructural approaches used to reduce pollutant loads in watersheds draining both urban and rural areas. The Soil and Water Conservation Society (SWCS) defines a BMP as "a practice or combination of practices that are determined by a state or designated area-wide planning agency to be the most effective and practicable (including technological, economic, and institutional considerations) means of controlling point and nonpoint source pollutants at levels compatible with environmental quality goals" (Barry et al., 2001).

Vegetation plays a critical role in flow resistance, particularly in relation to relaxation times and thresholds for flood erosion effects. Because of the resistance of vegetation-mantled banks and flood plains, there may be progressive, long-term encroachment of vegetation into the channel, especially on small streams. This vegetation increases roughness so reduces flow conveyance (Wolman and Gerson, 1978). The vegetation could retain from 30 to 70 percent of the deposited sediments. The ability of vegetation to entrap and retain sediment is related to the length and cross-sectional area of the vegetation (Thornton et al., 1997). Early research on the hydrologic impacts of vegetation management practices began in the 1910s, was expanded into the 1930s and 1940s, and continuing in the 1980s to further evaluate the effects of vegetation manipulations on the basin's water resources and other multiple uses (Zou et al., 2010). The impact of vegetation on the system is overwhelming. Vegetation produces an erosion-resistant peat layer, stabilizes channel banks and slows down the water flow. Vegetation also stimulates aggradation of bed load material on the channel bottom, and contributes to avulsion by blocking the channels. The channel network owes its origin to repeated though infrequent avulsion (Gradzinski et al., 2003).

Another important management practice is the no-till farming. Notill farming, due to an associated increase in surface residue and reduction in surface runoff, has been recommended as a best management practice to reduce soil erosion. Surface residues affect erosion by decreasing the soil surface area susceptible to raindrop impact, reducing the velocity of runoff and hence its transport capacity, and by creating mini-ponds that result in deposition behind clumps of residue (Fu *et al.*, 2006).

Another conservation practice is the crop rotation (often called conservation crop rotation), that is defined as the use of different crops in a specified sequence on the same farm field. There are several reasons for using crop rotations; although the primary one is to reduce soil erosion, thereby reducing the quantities of sediment and sediment-bound pollutants such as nitrogen, phosphorus and pesticides. (Barry *et al.*, 2001).

The objective of this paper is to evaluate the AnnAGNPS model prediction capability for runoff and sediment yield using a five years data base in a middle sized Mediterranean watershed located in Southern Italy. A continuous simulation process of runoff and sediment yield has been carried out comparing the simulation outputs to the corresponding observed data measured during the period 2007-2011. A calibration process for the model parameters that have a large impact on the prediction capacity of the model has been performed at the event scale utilizing the events recorded in the period 2007-2008, then a validation process to evaluate the model performance using the events recorded in the period 2009-2011. Finally, after validation, the model has been used to evaluate the effectiveness of alternative MPs on sediment yield at the watershed scale.

Materials and methods

AnnAGNPS Model Description

The Annualized Agricultural Nonpoint Source Pollution model (Theurer and Cronshey, 1998; Bingner and Theurer, 2005; USDA-ARS, 2006) was developed by the USDA Agricultural Research Service (ARS) and Natural Resources Conservation Service (NRCS) to predict sediment and chemical delivery from un-gauged agricultural watersheds up to 300,000 ha (Bosch et al., 2001). AnnAGNPS is a continuous simulation, grid-based, batch-process computer program where runoff, sediment, nutrients and pesticides are routed from their origins in upland grid cells through a channel network to the outlet of the watershed (Binger and Theurer, 2005). The climatic data requirements for simulations include daily maximum and minimum temperature, precipitation, average daily dew point temperature and wind speed, and sky cover (Bingner and Theurer, 2005). The ArcView interface for AnnAGNPS incorporates the Generation of weather Elements for Multiple applications (GEM) climate generation model (USDA-ARS, 2005) which generates daily precipitation, maximum and minimum temperature, and solar radiation. AnnAGNPS users also have the option to input measured climate data by uploading the data into the input editor.

AnnAGNPS hydrology is based on a simple bookkeeping of inputs and outputs of water during the daily time steps (Bingner and Theurer, 2005). The hydrologic processes simulated in the model include interception, evaporation, surface runoff, and evapotranspiration, subsurface lateral flow and subsurface drainage (Yuan et al., 2006). In AnnAGNPS, runoff is predicted using the SCS curve number technique (USDA-SCS, 1986), and sheet and rill erosion are predicted with the Revised Universal Soil Loss Equation (RUSLE, Renard et al., 1997). Soil moisture balance is calculated on a sub-daily time step using a simple constant-time step procedure for both the tillage and below tillage composite soil layers (Bingner and Theurer, 2005). Sediment transport in channels is computed using a modified Einstein equation, and the Bagnold (1966) equation is used to estimate sediment transport capacity of the flow (Bingner and Theurer, 2005). AnnAGNPS utilizes the HUSLE (Hydro-geomorphic Universal Soil Loss Equation) model (Theurer and Clarke, 1991) to determine sediment delivery ratios of total sediment to the stream network.

Study area

The Carapelle watershed is located in Puglia region (Southern Italy). It originates from the flyschoid formations of the Daunia Mountains and crosses the Tavoliere floodplain before flowing into the Adriatic sea (Figure 1; Table 1). Soils predominantly belong to the class of Entisols and have a fine clayey-loamy texture, are low in organic matter content, natural fertility and water-holding capacity. The plain and the low hilly areas (80% of the watershed surface) are mainly used for cereal cultivation and olive orchards, whereas in the higher slopes deciduous oaks



and hardwoods (*Quercus pubescens* and *Quercus cerris*) and pasture conditions are present. The climate is typically Mediterranean, with rainfalls ranging from 450 to 800 mm/year and average temperatures from 10 to 16° C. The flow regime is torrential and flood events are mainly associated with intensive, short-term rainfalls. Suspended sediment, even during floods, is mainly characterized by fine particles.

Monitoring stream flow and suspended sediment

Continuous stream flow and sediment load data derived from the monitoring station located in the Carapelle torrent at Ordona-Castelluccio dei Sauri Bridge (Figure 2). The station is equipped with an ultrasound water stage meter and a water stage recorder, with remote data transmission. Runoff is determined by converting the record of water levels into a record of flows using the experimental water discharge rating curve given by the National Hydrografic Service, that is responsible for the runoff measurements. An infrared optic probe (Hach-Lange SOLITAX Hs-line) is used to monitor suspended sediments. The probe measures suspended sediments by coupling backscattering and nephelometric photodetectors (Gentile et al., 2010). The instrument is contained in a shelter tube through a pulley, a float and a counterweight group that is anchored to a bridge pier to protect the instrument from the impact of coarse material in the flow and to prevent any potential measuring errors caused by incident radiant energy straying into the infrared field. The instrument is controlled through a data acquisition system that is powered by solar panels and is able to measure high solid concentrations with the capability to reduce watery medium and light interferences.

Gentile *et al.* (2010) tested the probe in the laboratory in order to evaluate the functional capacity of the instrument and to assess the effects of the different grain size and solid fractions on measurements. The instrument was field calibrated during the flood events of 2007-2009 to evaluate the efficacy of the housing system, to identify a calibration curve of the instrument for the specific torrent and to assess the type of the relationship between the SSC measured by the instrument and the gravimetric SSC. The SSC of all samples was measured using the gravimetric method and compared with the data observed by the optical sensor.

Thirty-six events observed during 2007-2011 were used for the application of the AnnAGNPS model, this does not include all runoff and sediment yields that occurred in the watershed, some storm events were not sampled due to equipment malfunctions or temporary lack of power to the sensor, caused by the solar panels. In general, the small number of flood events during the rainy season is another characteristic of Mediterranean watersheds.

Considering that AnnAGNPS do not consider base flow, the surface runoff separation from baseflow was performed (Figure 3) using the filtering algorithm developed by Eckhardt (2005). Baseflow separation

Table 1 Main characteristics of the Carapelle watershed, mouth at Ordona bridge.

Watershed area	km2	506.2
Maximum altitude	m a.s.l	1075.0
Average altitude	m a.s.l.	466.0
Minimum altitude	m a.s.l	120.0
Main channel length	km	52.2
Main channel slope	%	1.8
Mean watershed slope	%	8.2
Time of Concentration	hour	10

is required in numerous widely used hydrological and erosive models and must be considered in monthly models (Mouelhi *et al.*, 2006). The filtering algorithm has the following equation:

$$b_{k} = \frac{(1 - BFI_{\max})ab_{k-1} + (1 - a)BFI_{\max}Q_{t}}{1 - aBFI_{\max}}$$
(1)

where b_k is the base flow at time stamp k; b_{k-1} is the base flow at the previous time step; Q_t is the measured total flow; BFI_{max} is a constant that can be interpreted as the maximum value of long term ratio of base flow to total stream flow; a is the recession constant. The filter parameter "a" and BFI_{max} were calculated using the hydrograph recession curve analysis and the optimization module developed by Kyoung *et al.* (2010).



Figure 1. Carapelle watershed in Southern Italy.



Figure 2. Continuous stream flow and sediment load data monitoring station.



Input Data Preparation

Topography

The topographic features were defined using the Digital elevation map (90m) of the Carapelle watershed provided by the SRTM (Shuttle Radar Topographic Mission) project carried out by the NASA (National Aeronautics and Space Administration) and the NGA (National Geospatial-Intelligence Agency).

The watershed discretization into homogeneous drainage areas ("cells") and the hydrographic network segmentation into channels ("reaches") was performed using the GIS interface incorporated into AnnAGNPS. The geometry and the density of the drainage network were modeled by setting the critical source area to 50 ha and the minimum source channel length to 250 m, resulting in 1006 cells and 416 reaches, which allowed a suitable representation of the watershed.

Land Use and Field Management

Land use data are based on the 1:100 000 CORINE Land Cover data set (CLC2000). The accuracy of the data set has been validated in other studies by comparing images with ground based photography and field surveys (EEA, 2006). Based on the CLC2000 dataset, land uses were grouped in six main classes: cropland (winter wheat and olive-groves), rangeland, forest, urban, fallow and pasture. Figure 4 reports the surface area covered by each land use. The CORINE data set only distinguishes between arable land and agricultural or non-agricultural land use types, therefore information on crop growth and cropping methods were needed. In particular the crop data and management information required by the model include the units harvested, surface and subsurface decomposition, crop residue, root mass, canopy cover, management scheduling and agricultural operations.

The winter wheat crop parameters were based on RUSLE guidelines and internal databases (Renard *et al.*, 1997) while four management practices were assigned to represent the local conditions of the watershed (Table 2).



Figure 3. Measured stream flow, suspended sediments and calculated baseflow for some flow events registered during the period 2007-2011

Planting operations occurred in September and harvesting operations occurred in June. After harvest, the land is prepared with other management practices (tillage, semi-deep drill). The tillage effects are linked to the management of crop residues, control of competing vegetation, incorporation of amendments, preparation of the seedbed and, in semi-arid zones, moisture conservation.

A new database was created for the olive grove crops describing: the root density, the estimated aerial coverage of the crop canopy and the average rainfall drop height, which were assumed to remain constant respectively at 30000 kg ha-1, 50 % and 1 m (Galvagni *et al.*, 2006). Tillage operations and organic fertilizers applications were scheduled (Table 2) as they are required to aerate soil, improve water storage,



Figure 4. Layout of the discretized watershed with the surface area covered by each land use and soil texture class.

Table 2. Management scheduling for cropland.

Winter	wheat	Olive grove			
Event date	Management scheduling	Event date	Management scheduling		
06/01	Harvest grain	01/01	Organic fertilizer application		
09/01	Tillage	04/01	Tillage operation		
09/20	Begin crop growth	06/01	Shallow tillage operation		
12/15	Semi-deep drill	08/01	Shallow tillage operation		
11/01	Harvest	12/01	Organic fertilizer application		

Table 3. Soil properties for each textural class.

Soil structure	K factor (t h MJ ⁻¹ mm ⁻¹)	Wp (%)	Fc (%)	ks (mm h ⁻¹)
Clay	0.035	0.3	0.42	4.5
Sandy clay	0.034	0.26	0.37	14.89
Loam	0.043	0.11	0.24	12.7
Clay Loam	0.03	0.2	0.34	2.02
Silty clay	0.035	0.27	0.42	1.49
Silty-loam	0.044	0.1	0.27	9.88
Silty-clay-loam	0.043	0.18	0.37	4.59
Sandy-clay-loam	0.033	0.17	0.26	4.83
Sandy-loam	0.006	0.2	0.1	32.92



remove weeds and avoid soil compaction.

Single non-cropland databases were assigned to rangeland, forest, urban, fallow and pasture field types. The root system parameters of the forest field type were derived from literature (Galvagni *et al.*, 2006).

The crop management factor C for each period was calculated by the model based on land use, canopy cover, surface cover, surface roughness and soil moisture conditions. The P factor used was set to 1 since no significant management operations were implemented to reduce soil erosion.

Soil properties

Soil parameters such as the textural classes, saturated hydraulic conductivity and soil depths were extracted from the project ACLA2 (scale 1:100 000), a research program funded by the Puglia Region aimed at agro-ecological characterization of the region on the basis of laboratory tests, field observations and photo interpretation of aerial photographs and satellite images (Caliandro *et al.*, 2005). The soil depth is considered as the portion of soil that allows the development of functional and organic roots, where the term "functional" refers to soil moisture dynamics while "organic" considers the interactions involved in the organic matter production (Caliandro *et al.*, 2005). On the basis of the USDA triangle the mean percentages by weight of sand, clay and silt, were assigned to each textural class. The percentage of organic matter was derived from the project Octop of the European Soil Data Centre (ESDAC, 2003).

In the Carapelle watershed soils have weak or no diagnostic subsurface layers and are generally well drained. Hydraulic conductivity decreases as soil pore diameters decreases: in the root zone (h<110 cm) saturated hydraulic conductivity can be assumed to have high values (k_{sat}>3.6 cm/h), but the drainage can be restricted below the root zone (k_{sat}<0.036 cm/h).). For this reason the permeability of the soil layer used to simulate subsurface flows and to account for the groundwater processes was referred to the less permeable layer at 150 cm. The average soil hydraulic properties, water content at wilting point Wp, field capacity Fc and saturated hydraulic conductivity ks, were calculated for each soil type (Table 3) using the Saxton and Rawls (2006) pedotransfer functions.

The RUSLE soil erodibility factor "K", was estimated using the Lal and Elliot (1994) equation. Nine types of soils were identified and the average erodibility factors were calculated (Table 3). Based on each land use (cropland, fallow, rangeland, forest, pasture, urban) and Hydrologic Soil Group (A, B, C, D) the initial Curve Numbers were defined (Table 4).

Climate

Meteorological data such as daily maximum temperature, daily minimum temperature, daily precipitation, daily sky cover, daily wind speed and daily dew point temperature were used as the input climate data

Table 4. Initial curve number values.

Cover type	Initial curve numbers for hydrologic soil groups				
				D	
Cropland	72	81	88	91	
Fallow	76	85	90	93	
Rangeland	35	56	70	77	
Forest	43	65	76	82	
Pasture	49	69	79	84	
Urban	89	92	94	95	

for the simulations. The spatial distribution of climatic data were assessed using the Thiessen weighting procedure related to the eight rain gauges located in the watershed or in the surrounding area.

The rainfall erosivity factor R was estimated by (Bisantino *et al.*, 2013) considering the mean monthly precipitation of the period 1979-1999, according to Ferro *et al.* (1999), resulting in 960.70 MJ mm ha⁻¹year⁻¹.

Model calibration and validation

Both the hydrological and erosion components of AnnAGNPS were calibrated/validated in a logical order taking into account a previous sensitivity analysis performed by Bisantino *et al* (2013) for the most meaningful parameters of the model (R, K, C and P factors of USLE equation, CN curve number and MN Manning's roughness coefficient). The parameter calibration order used was first surface runoff, then peak flow and finally sediment load. Input parameters affecting surface runoff and peak flow were first calibrated because of their influence on the other output.

The calibration/validation process of runoff was carried out by modifying the initial values of CN, which represent a key factor in obtaining accurate prediction of runoff and sediment yield (Yuan *et al.*, 2001; Shrestha *et al.*, 2006) and the most important input parameter to which the runoff is sensitive (Yuan *et al.*, 2001; Baginska *et al.*, 2003).

For the calibration of peak flows and sediment yields, both 24 h rainfall distributions (types I and Ia) typical of a Pacific maritime climate with wet winters and dry summers outlined by the Natural Resource Conservation Service (NRCS) and described by SCS (1972) were considered. Based on the analysis of observed rainfall events at different rain gauges performed by Bisantino *et al* (2013) it was found that both storm types well represent the meteorological conditions of Mediterranean zones. The storm type "I" was found to give better predictions of peak discharge so it was set during simulations.

The sediment yields were evaluated at the event scale by adjusting the Manning's roughness coefficient (whichaffects the RUSLE C factor).

The observed flood events have runoff volumes ranging from 0.2 to 8.6 mm (94593 to 4336938 m³), peak discharges between 1.6 and 73.6

Table 5	. Coeffici	ents and	l difference	measures	for	model	evaluation	and
their ra	nge of var	iability.						

Coefficient	Equation	Range of variability
Coefficient of efficiency (Nash and Sutcliffe, 1970)		-∞ to 1
Willmott index 1982		0 to 1
Coefficient of determination R ²	(0 to 1
Coefficient of residual mass (Loague and Green, 1991)		-∞ to ∞
Root mean square error		0 to ∞



m³/s and sediment loads between 202 to 103216 t (0.4 to 204 g m²). The years considered (2007-2011) had precipitation rates ranging from 544.0 mm in 2007 to 873 mm in 2010 .

Model Performance Assessment

The model performance was evaluated at the event scale by qualitative and quantitative approaches. The qualitative approach consisted of visually comparing observed and simulated values. For a quantitative evaluation, a range of both summary and difference measures were used. The summary measures utilized were the mean and standard deviation of both observed and simulated values. For difference measures five evaluation criteria were used to evaluate the model performance: the coefficient of determination (\mathbb{R}^2) , the Nash-Sutcliffe coefficient of efficiency (NSE), the Willmott index of agreement (W), the coefficient of residual mass (CRM), and the root mean square error (RMSE). The coefficient of determination, R^2 , describes the proportion of the total variance in the observed data that can be explained by the model, R² is an insufficient and often misleading evaluation criterion as large values of R² can be obtained even when the model-simulated values differ considerably in magnitude; the Nash and Sutcliffe (1970) coefficient of efficiency (NSE) was also used to assess the model efficiency (Table 5). In particular, some authors discussed that NSE is more sensitive to extreme values (Legates and McCabe, 1999; Krause et al., 2005). Willmott (1982) sought to overcome the insensitivity of correlation-based measures to differences in the observed and modelsimulated means and variances by developing the index of agreement. The Coefficient of Residual Mass (CRM) was used to indicate a prevalent model over- or underestimation of the observed values (Loague and Green, 1991). The values considered optimal for these criteria were one for "R²", "NSE" and "W" and zero for "CRM". According to common practice, simulation results are considered good for values of NSE greater than or equal to 0.75, satisfactory for values of NSE between 0.75 and 0.36, and unsatisfactory for values below 0.36 (Van Liew and Garbrecht, 2003). Finally, the RMSE describes the difference between the observed and simulated values in the unit of the variable, and it ranges from 0 to ¥, where zero indicates that there is no difference between model simulations and field observations. To quantify the model accuracy in simulating runoff, peak discharge and sediment load, AnnAGNPS was applied to simulate the entire period 2007-2009. The output efficiency was evaluated as relative error (RE). The RE is the ratio of the total difference between simulated and observed values versus the total observed value. It ranges from minus one to ∞ while zero indicates that there is no difference between model simulation and field observation. The smaller the absolute value of a RE, the better performance of the model is:

$$RE = \frac{|P-O|}{O} \tag{2}$$

where P is the predicted value and O is the observed value. The relative error was used to solve the problems of significance and units, as it is the ratio between the absolute error and the absolute value of the correct value.

Management Practices

The management practices have an important role when applied as a plan of soil and water conservation. The aim of the simulation was to understand the entity of the sediment yield reduction at a watershed scale in a Mediterranean environment when applying agricultural or environmental measures of soil erosion control; for this reason these measures were applied at a very large scale in the watershed. The different scenarios that have been considered are representative of situations that it is difficult to find in real cases; nevertheless, they could be



Figure 5. Comparison of 36 observed and simulated events: (a) peak flow, (b) runoff and (c) sediment yield in the Carapelle watershed



a good starting point to design a combination of agricultural and environmental measures that can have a good impact on the reduction of sediment yield in the watershed. Such a process, that should be carefully carried out to take into account the real conditions of the watershed in terms of physical, environmental, agricultural and socio-economic features, could drive to the definition of the so called Best management practices.

To evaluate the effectiveness of Management Practices (MPs) on sediment yield, the following alternative scenarios were simulated using AnnAGNPS at the monitoring site, and the sediment yield at the watershed outlet was compared with the results obtained from the validation simulation file:

1st scenario (S1) the wheat crop was assumed to be converted into forest. During this simulation, the input file was modified to replace the field land use type "Winter wheat" into the "Forest" land use type; CN and MN were set to forest. This scenario can be considered representative of the situation before the establishment of agriculture in the area.

2nd scenario (S2) using the Vegetated Streams. In this simulation, the reach vegetation code was changed to assign a code for a vegetated stream instead of no-vegetated streams.

 3^{rd} scenario (S3) A crop rotation of Corn & soybean was assumed to be planted instead of the winter wheat crop. The winter wheat cover crop was not planted and the field was fallow, with residue from soybeans and cotton left on the surface. The data base of corn and soybean were based on the internal data base of the model: the corn was planted in 15th of April of the 1st year of simulation and harvested in the 25th of October of the same year, while the Soybean was assumed to be planted in the 1st of April of the following year and harvested in the 25th of October of the same year. For both crops, no tillage operations were supposed, and the plant residues were left in place.

4th scenario (S4) no tillage in winter wheat. In this simulation, the

input file was modified to exclude any crop tillage operation in the Management operation data.

Results and Discussion

Calibration and validation

The model performance was calibrated at the event scale utilizing 11 observed erosive events registered during the period 2007 -2008. In uncalibrated mode, the model tends to over-predict the runoff volumes, so the initial CNs were properly decreased to get runoff results closer to the observed ones. The same results were found in semi-arid conditions by Licciardello *et al.* (2007).

At the end of calibration, runoff depths were in general slightly over predicted (see the negative value of the CRM coefficient in Table 6). The mean value and the standard deviation of the simulated runoff depths were close to the observed corresponding values, with a difference equal approximately to 10 and 2 % respectively. The coefficient of determination and the Nash-Sutcliff efficiency factor achieved after the runoff calibration were good (Table 6). Similar results were found at the event scale by other authors like Yuan *et al.* (2001), Shrestha *et al.* (2006) and Shamshad *et al.* (2008).

After calibration, the mean predicted value of peak flow was 14 % different from the mean observed value, however, the difference between predicted and simulated values raises for the standard deviation to be more than 40 % (Table 6). High efficiency is shown by the coefficient of determination R^2 value (R^2 =0.81), while other statistical indexes (NSE and RMSE) show a satisfactory prediction (Table 6). Other authors (Zema *et al.*, 2010; Shrestha *et al.*, 2006; Licciardello *et al.*, 2007) found that the model unsatisfactory predicted peak flows. To calibrate the sediment yield, the MN's roughness coefficient of each cell was modified starting from the initial values taken from TR55

Table 6. Statistics concerning the AnnAGNPS calibration and validation in the Carapelle watershed.

Calibration			Runoff (mm)					
Observed	1.5	1.9	-	-	-	-	-	
Default simulation	2.1	2.4	0.38	0.65	-0.42	1.4	0.88	
Calibrated model	1.7	1.9	0.76	0.78	-0.1	0.91	0.93	
			Peak flow (m ³ /s)					
Observed	16	14	-	-	-	-	-	
Default simulation	27	31	-1.4	0.86	-0.75	21	0.78	
Calibrated model	18	20	0.54	0.81	-0.14	9.2	0.92	
	Sediment yield (kg/m2)							
Observed	0.018	0.025	-	-	-	-	-	
Default simulation	0.037	0.044	-0.81	0.62	-1.04	0.33	0.77	
Calibrated model	0.024	0.028	0.67	0.74	-0.35	0.01	0.92	
Validation			Runoff (mm)					
Observed	2.56	2.22	-	-	-	-	-	
Simulated	2.28	2.18	0.81	0.82	0.11	0.99	0.95	
	Peak flow (m ³ /s)							
Observed	29	18	-	-	-	-	-	
Simulated	26	22	0.69	0.82	0.11	10.2	0.93	
Sediment yield (kg/m ²)								
Observed	0.036	0.05	-	-	-	-	-	
Simulated	0.038	0.04	0.86	0.86	-0.06	0.02	0.96	



(SCS, 1986). For forest and rangeland, MN values respectively equal to 0.8 and 0.13 were set , while for urban areas, cropland and pasture the initial value of 0.15 was considered. Increasing the value of MN for the different land uses and especially for the cropland (wheat) and urban areas, the tendency of the model to overestimate the suspended sediment yield that is clear in the un-calibrated mode was reduced (see the CRM values before calibration in default simulation and after calibration in Table 6). Generally, a good correlation between observed and simulated data was obtained, as reported by the Nash Sutcliffe efficiency index and the coefficient of determination R²; the value of the root mean square error was close to zero, showing a good model efficiency in predicting sediment yield.

The results obtained in the calibration phase were used for model validation, that was carried out using 25 events recorded during the period 2009-2011. As confirmed by the statistical indexes, an excellent performance of the model was observed in simulating runoff volumes (Table 6). The model performance in predicting peak discharges is satisfactory (see RMSE and NSE values) although according to the R² values a good correlation between observed and simulated data exist. Sediment yield values were perfectly predicted and an acceptable overestimation is observed.

Figure 5 reports the visual comparison between simulated and observed data for the entire period (thirty-six events). As expected using the calibrated SCS-CNs, the model prediction is good for runoff and sediment yield and satisfactory for peak discharge.

After the validation process, it was observed, comparing the statistical indexes, that the model prediction was better in the validation than in the calibration process. To better understand this finding, the relative error (RE) was calculated for two groups of events and represented on a box plot (Figure 6). The first group (26 events) has peak discharges $Q_p \leq 30 \text{ m}^3 \text{ s}^{-1}$ and is representative of low flow events, as previously stated by Gentile *et al.* (2010) in the same watershed; the second group (10 events) has peak discharges $30 < Q_p \leq 73.6 \text{ m}^3 \text{ s}^{-1}$ and is representative of high flow events. A difference in the RE values between the two groups was observed, showing that the model better predicts sediment yield for high intensive events and that a larger scatter between simulated and observed sediment yield values exists for low flow events. The same behaviour was observed for peak flows.

As the number of high flow events included in the validation process (25 events, 3 years) is greater than those included in the calibration (11 events, 2 years), this could be the reason why the model did not perform as well in calibration than in validation.

Sediment yield response to alternative Management Practices

Figure 7 shows the average annual sediment yield (t/ha) at a watershed scale obtained applying four alternative scenarios of Management practices that could be used to reduce soil erosion and sediment loads in the watershed.

In the first scenario (S1), when all the cropland wheat is substituted with forest, the sediment yield is reduced in the watershed, with an average value of 60 % during the whole period of simulation; this land use change would save a total amount of 9.3 t/ha of sediments.

In the second scenario (S2) the use of vegetation along the whole stream network (vegetated streams) reduces the average annual sediment yield by 18.5% (a total amount of 3 *t*/ha), due to the effect of vegetation in retaining sediments from being eroded and conveyed to the watershed outlet.

In the third Scenario (S3) a crop rotation of corn-soybean is hypothetically applied instead of the overall cropland wheat. The average annual sediment yield is reduced by 34 %, due to the effect of the continuous cover offered by the crop residue that helped in protecting soil from being eroded. As a quantity, a total amount of 5.3 t/ha of sediments could be saved.

The last scenario (S4) is corresponding to the application of no tillage practices in cropland. Herethe sediment yield could be reduced by up to 18.5%, preventing approximately 3 t/ha of sediments from being lost.

The S1 scenario, that means converting all the cropland into forest and gives the better results in terms of soil erosion control, could be considered as representative of the ancient conditions in the Puglia region (eighteen century) when forest was the prevailing land use and agriculture was not yet extensively estabilished.

The S3 and s4 scenarios can be considered as agriculture-targeted, both give good results in terms of sediment yield reduction, even if the efficiency of crop rotations is approximately two times that of the conservation tillage (no-tillage) practices. Finally, the S2 scenario, that could be considered as environmentally-targeted, has the same efficiency of the agricultural S4 scenario.



Figure 6. Average relative error (RE) for low flow events (a) and high flow events (b).



Figure 6. Average relative error (RE) for low flow events (a) and high flow events (b).





Conclusions

The AnnAGNPS Model was implemented in the Carapelle, a medium size Mediterranean watershed in southern Italy, using 36 erosive events recorded during the period 2007-2011. The objective of this work was to evaluate the model capacity of predicting sediment yield, runoff, and peak flow in the watershed through a continuous monitoring of runoff and sediment loads. The evaluation process included 2 phases, the first one was the model calibration and the second one was the model validation. The model calibration was performed following an order in which runoff was calibrated first by adjusting the initial curve number values modifying the retention S factor. The impact of the SCS types Ia and I synthetic rainfall distribution was investigated; the SCS storm type I gave the highest efficiency in predicting peak flows and was used in the simulation. Finally, the sediment yield was calibrated by adjusting the Manning roughness coefficient. Comparing predicted to observed data visually and using different statistical indexes, the results showed a good model performance in predicting runoff and sediment yield on the event basis in calibration and validation, while its performance was satisfactory for peak discharge. Generally, the model showed a tendency to better predict high flow events while a great scatter was found between observed and predicted values in case of low flow events. This result is important in semi-arid conditions where the annual sediment transport rate is mostly concentrated in a small number of high erosive events.

Successively the effectiveness of applying some Management practices on the sediment yield reduction at a watershed scale was determined. Based on the simulation results, it is concluded that substituting cropland wheat with forest achieved the highest reduction in sediment yield. A good compromise between agriculture and soil conservation was represented by the crop rotation of corn and soybean, while the scenarios corresponding to the revegetation of the stream banks ("environmentally targeted") and the conservation tillage (no-tillage) in cropland achieved almost similar results. Further analyses need to be performed to evaluate the management practices that are more suitable with the real agriculture and land conditions in the watershed and consequently to assess possible Best management practices (BMPs) in terms of soil erosion control and sediment yield reduction.

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