

Reply to comments of Anonymous Referee #3

We thank anonymous Referee #3 for reviewing our manuscript. The authors are grateful for the insightful comments which provide great suggestions to improve the manuscript.

General Comments

The authors present a modeling study of an urban catchment with the SWAT. The study aims to compare the performance of daily simulations (using the curve number method) and hourly simulations (using the GAML method). The paper is within the scope of the journal, address relevant scientific questions and the presentation is of quality.

The Methods and assumptions are well described and understandable. However, Results and Discussion could be developed: at the moment they read a little like a list of comments (especially section 3.3) without enough physical discussion. The authors find that different modeling time steps and different runoff generation methods impact model parameters and performance, which is somehow expected and established. The novelty of the contribution therefore needs to be detailed and explained.

The reasons why the sub daily model did not perform as well as the daily model deserve more physical discussion about hydrological processes and implications. In addition, is it really fair to compare hourly and daily performance metrics directly? Shouldn't the results of the GAML method be aggregated to the daily timestep to really be compared with the CN method?

Reply: We thank the referee for the constructive feedback and comments. We agree that the Results and Discussion should be improved. We will re-organize Section 3.2 and Section 3.3 in a more clear way and we will discuss the physical meaning of the results in the revised manuscript. We will explain the objectives and the contribution of this study in Introduction, Results and Discussion section. In addition, we will explain the peak flow underestimation and we will discuss the disadvantages of the GAML method. We will also aggregate the hourly results to a daily time step in order to compare the performances in the revised manuscript. We will make the suggested changes in the following comments.

Specific comments

Abstract

Comment: The abstract does a good job at summarizing the work in concise and clear language. The novelty of study could be explained in the Abstract and numerical values of Results could also be given.

Reply: We thank the referee for the constructive feedback. We agree with the referee that the novelty of the study should be more emphasized. We will make the following changes in the revised manuscript:

Lines 11-22: “SWAT (Soil and Water Assessment Tool) is a continuous time, semi-distributed river basin model that has been widely used to evaluate the effects of alternative management decisions on water resources. This study examines the application of SWAT model for streamflow simulation in an experimental basin with mixed-land-use characteristics (i.e., urban/peri-urban) using daily and hourly rainfall observations. The main objective of the present study was to investigate the influence of rainfall resolution on model performance in order to analyze the mechanisms governing surface runoff at the catchment scale. The model was calibrated for 2018 and validated for 2019 using the SUFI-2 algorithm in the SWAT-CUP program. Daily surface runoff was estimated using the Curve Number method and hourly surface runoff was estimated using the Green and Ampt Mein Larson method. A sensitivity analysis conducted in this study showed that the parameters related to groundwater flow were more sensitive for daily time intervals and channel routing parameters were more influential for hourly time intervals. Model performance statistics and graphical techniques indicated that the daily model performed better than the sub-daily model (daily model: $NSE = 0.86$, $R^2 = 0.87$, $PBIAS = 4.2\%$; sub-daily model: $NSE = 0.6$, $R^2 = 0.63$, $PBIAS = 11.7\%$). The Curve Number method produced higher discharge peaks than the Green and Ampt Mein Larson method and estimated better the observed values. Overall, the general agreement between observations and simulations in both models suggests that the SWAT model appears to be a reliable tool to predict discharge in a mixed-land-use basin with high complexity and spatial distribution of input data”.

Comment: L12: Is 'demonstrate' the right word? The paper rather "examines" the influence of precipitation time-step on performance metrics.

Reply: We agree with the referee. The verb “demonstrate” is not the right word. This study rather investigates or examines the influence of precipitation time step on model performance. We will make the following changes:

Lines 12-14: “This study examines the application of SWAT model for streamflow simulation in an experimental basin with mixed-land-use characteristics (i.e., urban/peri-urban) using daily and hourly rainfall observations”.

Comment: L21: "long periods of time": in the paper the modeling period is 3 years which is not very long.

Reply: We agree with the referee that three years is not a long period of time. We believe that we should focus in the land use characteristics of the study area and in the ability of SWAT to predict discharge in an urban/peri-urban environment. We will make the following changes in the manuscript.

Lines 21-22: “Overall, the general agreement between observations and simulations in both models suggests that the SWAT model appears to be a reliable tool to predict discharge in a mixed-land-use basin with high complexity and spatial distribution of input data”.

1. Introduction:

Comment: L74: water level data of water flow data?

Reply: We thank the referee for this comment. The river gauge at the outlet of the basin measures water level data at a 15 min time step. Then, using the Manning's equation we calculated the flow velocity and the flow rate.

Manning's Equation:

$$Q = V * A = \left(\frac{1}{n}\right) * A * R^{\frac{2}{3}} * S^{\frac{1}{2}}$$

Where:

Q = Flow Rate, (m³/s)

V = Velocity, (m/s)

A = Flow Area, (m²)

n = Manning's Roughness Coefficient

R = Hydraulic Radius, (m)

S = Channel Slope, (m/m)

2. Materials and Methods:

Comment: L95-96: "in different times and under different weather conditions": is the monitoring continuous?

Reply: The monitoring stations measure water level at a 15 minute time step and the data are provided freely from Open Hydrosystem Information Network (OpenHi.net).

Comment: L101-106: The authors could consider adding a sentence about velocity measurement? Can the real life accuracy of the probes be somehow estimated (especially since observational errors are mentioned later in the discussion)?

Reply: We thank the referee for this comment. The station at the outlet of the basin measures water level data. Then, using the Manning's equation we calculated the flow velocity and the flow rate. We will also exclude the reference to the water velocity in the revised manuscript since we didn't use water velocity in the calibration process. We will shorten the instruments description. We will clarify this statement in the manuscript in the following lines:

Lines 95-106: “The study area includes four water level monitoring stations that provide continuous recordings of the river stage at pre-selected time-intervals (15mins time-step) (Fig. 1). The stations were installed at the end of 2017 under the supervision of the School of Mining of National Technical University of Athens (NTUA). The network was developed under the EU H2020 RIA Program SCENT (Smart Toolbox for Engaging Citizens in a People-Centric Observation Web). The station which is located at the outlet of the study area was selected as the most suitable for further analysis in this study, because the three upstream stations experienced some mechanical problems that affected the calibration and validation process. The monitoring stations are part Open Hydrosystem Information Network (OpenHi.net) which is a national integrated information infrastructure for the collection, management and free dissemination of hydrological data (OpenHi.net) in Greece.”

Comment: L133: and sub-daily?

Reply: We thank the referee for this comment. The SWAT model operates on a daily and sub-daily time step. We will make the following changes in the revised manuscript:

Lines 133-134: “The model operates on a daily time step, and it has been recently updated to sub-daily time step computations (Jeong et al., 2010). SWAT has been developed to evaluate the impact of management practices on water, sediment, and agricultural chemical yields in large river basins over long time periods.”

Comment: L198: mean and standard deviation of daily discharge? In m^3/s ?

Reply: We thank the referee for this comment. The numbers refer to discharge. We changed line 198 in the following way:

Line 198: “Mean and standard deviation of discharge for 2018 were 1.25 and 0.46 and for 2019 were 1.42 and 0.74 respectively”.

3. Results and Discussion

Section 3.1

This section describes the results of the sensitivity analysis. It shows that the daily simulation seems to be more sensitive to runoff generation parameters whereas the sub-daily simulation is more sensitive to channel routing parameters. The section could be better structured: for example it starts with a discussion on CH_N2, then discuss GWQMN and GW_REVAP, then again CH_N2 and in the same paragraph discuss CN2, which makes it hard for the reader to follow the reasoning.

Reply: We thank the referee for the suggestions and we agree that this section should be better structured. We joined the first two paragraphs, we deleted the repetition in lines 241-242 and we explained the difference in the calibrated values of the two models. We will make the following changes in the revised manuscript in Section 3.1:

Lines 231-257: “The most sensitive parameters obtained in daily and hourly simulation are presented in Table 4. Sensitive parameters are characterized by large t-Test and small p-Value. The parameters were characterized as significantly sensitive when the p-value was less than 0.03. In the daily model, the most sensitive parameters were deep aquifer percolation fraction (RCHRG_DP), groundwater delay time (GW_DELAY), lateral flow travel time (LAT_TTIME), average slope steepness (HRU_SLP) and moist bulk density (SOL_BD). These parameters were connected to groundwater flow, runoff generation and channel routing. In the sub-daily model, the significantly sensitive parameters were average slope steepness (HRU_SLP), Manning’s “n” value for the main channel (CH_N2), effective hydraulic conductivity in main channel alluvium (CH_K2) and lateral flow travel time (LAT_TTIME). These were all related to channel routing.

The differences in the sensitivity of the calibrated parameters of the two models reflect the impact of the operational time step on model performance (Boithias et al., 2017; Jeong et al., 2010). In particular, the hourly model is characterized by larger GWQMN and GW_REVAP values than the daily model. GWQMN is the threshold depth of water in the shallow aquifer required for return flow to occur and GW_REVAP controls the water movement from the shallow aquifer into the overlying unsaturated soil layers. As these parameters increase, the rate of evaporation increases up to the rate of potential evapotranspiration, resulting in a corresponding decrease of the baseflow. Furthermore, the fitted value of CH_N2 in the hourly simulation was $0.11(m^{-1/3}s)$ and was larger than $0.08(m^{-1/3}s)$ in the daily simulation. The CH_N2 parameter affects the rate and the velocity of flow (Boithias et al., 2017). Therefore, the larger CH_N2 value was connected to smaller flow velocity. According to Boithias et al. (2017), the CH_N2 parameter is more sensitive at the hourly time step rather than the daily time step, because at the daily time step the flow peak is influenced by other processes decreasing the sensitivity of the CH_N2. In addition, the value range for CN2 was smaller for the sub-daily model, leading thereby to lower peak flows. Other differences were average slope steepness

(HRU_SLP), average slope length (SLSUBBSN), groundwater delay time (GW_DELAY) and Manning's "n" value for overland flow (OV_N). Their values were all smaller in sub-daily simulation. Overall, the differences between the two models lay mostly in the different runoff estimation methods used by the two models.”

Comment: L241-242: repetition from lines 234-239.

Reply: We agree with the referee. We deleted the repetition from lines 234-239 as mentioned above.

Comment: L252: is this difference physically meaningful?

Reply: Channel and hillslope velocities define the time of the peak and the shape of the hydrograph. In the SWAT model, the CH_N2 parameter (Manning's “n” value for the main channel) affects the rate and the velocity of the flow (Boithias et al., 2017). In this study, the CH_N2 parameter showed the highest sensitivity in the hourly model. This outcome is similar to those of previous studies (Boithias et al., 2017; Jeong et al., 2010). In particular, according to Boithias et al. (2017) the CH_N2 parameter is more sensitive at the hourly time step rather than the daily time step, because at the daily time step the flow peak is influenced by other processes decreasing the sensitivity of the CH_N2. Therefore, the lower peak flows of the hourly model comparing to the daily model could be attributed to the different calibrated value of the CH_N2 parameter.

Comment: L253: CH_N2 or CN2?

Reply: It is CN2 (Curve Number) parameter. This parameter explains the reason why the sub-daily model has lower peak flows than the daily model.

Section 3.2

This section presents the performance metrics of the daily and subdaily simulations. The authors conclude that the CN method is better than the GAML method. But, as stated above, is it fair to compare daily simulations with hourly simulations? Shouldn't the hourly simulation be aggregated to a daily timestep to have a 'fair' comparison?

Reply: We thank the referee for the suggestion. We will aggregate the hourly results to a daily time step in order to compare the performances in the revised manuscript. We present below in Figure 3 the observed versus simulated daily discharge aggregated from hourly outputs during the calibration and validation processes. We also show in Figure 4 the flow duration curve for the daily discharge aggregated from hourly outputs. We present the statistics for the daily aggregated discharge in Table 5. We will include these changes in the revised manuscript.

Comment: L272: an explanation for the underestimation?

Reply: The GAML method requires detailed soil data and high resolution precipitation data which can be difficult to obtain and may affect the accuracy of the model's results. We will explain the peak flow underestimation and we will discuss the disadvantages of the GAML method in lines 295-299 in the following way in the revised manuscript:

Lines 295-299: “In this study, the daily model produced higher discharge peaks than the hourly model and generally estimated better the observed values. These results could be due to drawbacks of the GAML method, such as the requirement for detailed soil information and high resolution rainfall data in a sub-daily time step (King et al., 1999). The GAML method assumes that the soil profile is characterized by homogeneity and that the previous soil moisture is distributed uniformly in the soil profile (Jeong et al., 2010). Therefore, the uncertainty in the resolution of the rainfall data, the heterogeneity of the soil formations and the upcoming difficulty in determining the parameters' values for parameterization could affect the method's efficiency. The selection of sub-daily precipitation input time step as well as the resolution of the precipitation data have a great impact on model results when using the GAML method and it should be based on the scale and characteristics of the watershed (Bauwe et al., 2016; Jeong et al., 2010; Kannan et al., 2007). Furthermore, observational errors in the model input data (i.e., weather, soil and land use data) include inaccuracies in the estimation of channel and hillslope velocities and channel geometry, in the nature of the sensor, environmental conditions and data collection (Guzman et al., 2015). These errors can generate variability, lead to undesired trends, and influence the model calibration and validation results (Kamali et al., 2017). In addition, the complex land use characteristics and processes of an urban/peri-urban environment and assumptions made during the model structure/parameterization process (e.g., selection of parameters for calibration, objective function, and conceptual simplifications) increase the uncertainty of the results.”

Comment: L274: it is expected that a daily timestep performs better than a subdaily timestep. It could be interesting to compare both methods at the same timestep, as stated above.

Reply: We agree with the referee and we will aggregate the hourly results to a daily time step in order to compare the performances in the revised manuscript. We will make the following changes in the revised manuscript:

Lines 266-277: “Quantitative statistics and criteria recommended by Moriasi et al. (2007, 2015) were used to evaluate the model performance. In order to investigate the influence of rainfall on model performance and compare daily outputs to hourly outputs, the hourly outputs were aggregated to daily averages. Figure 2 shows the temporal dynamics of the hydrographs reproduced by both infiltration methods. The high flow season is observed during winter and spring. The low flow season is observed in summer and early fall due to high evapotranspiration. Figure 3 shows the observed versus the simulated daily discharge aggregated from hourly outputs during the calibration and validation processes. Figure 4 presents the flow duration curves of the models, indicating good agreement between observed and simulated values. Generally, in the sub-daily model, the simulated discharge peaks did not always match the observed values and were sometimes considerably lower.

The performance statistics are illustrated in Table 5 and indicate reasonable calibrated models for both infiltration approaches. Model performance using the CN method showed better results than the GAML method. In particular, the NSE and R^2 indices for the daily model were 0.84 and 0.79 for the calibration period and 0.87 and 0.86 for the validation period. For the sub-daily model the NSE and R^2 indices were 0.53 and 0.49 for the calibration period and 0.63 and 0.6 for the validation period respectively. In addition, when the hourly outputs were aggregated to daily averages the NSE was improved comparing the NSE of the sub-daily model (e.g., sub-daily model: $NSE_{\text{calibration}} = 0.49$ and $NSE_{\text{validation}} = 0.6$, daily averages: $NSE_{\text{calibration}} = 0.66$ and $NSE_{\text{validation}} = 0.78$). However, the daily model outperformed the daily aggregated discharge during both calibration and validation periods. Furthermore, the daily model showed smaller modeling uncertainties with P-factor 0.79 and R-factor 1.58 (compared to 0.83 and 1.71 respectively for the sub-daily model).”

Table 5. Model evaluation statistics of the daily, sub-daily and daily aggregated from hourly outputs SWAT models for the calibration and validation periods.

Time-step	Period	p-Factor	r-Factor	R^2	NSE	PBIAS(%)
Daily	Calibration	0.74	1.41	0.84	0.79	6.4
	Validation	0.79	1.58	0.87	0.86	4.2
Sub-Daily	Calibration	0.72	1.33	0.53	0.49	16.9
	Validation	0.83	1.71	0.63	0.6	11.7
Daily averages	Calibration	-	-	0.76	0.66	16.8
	Validation	-	-	0.82	0.78	11.6

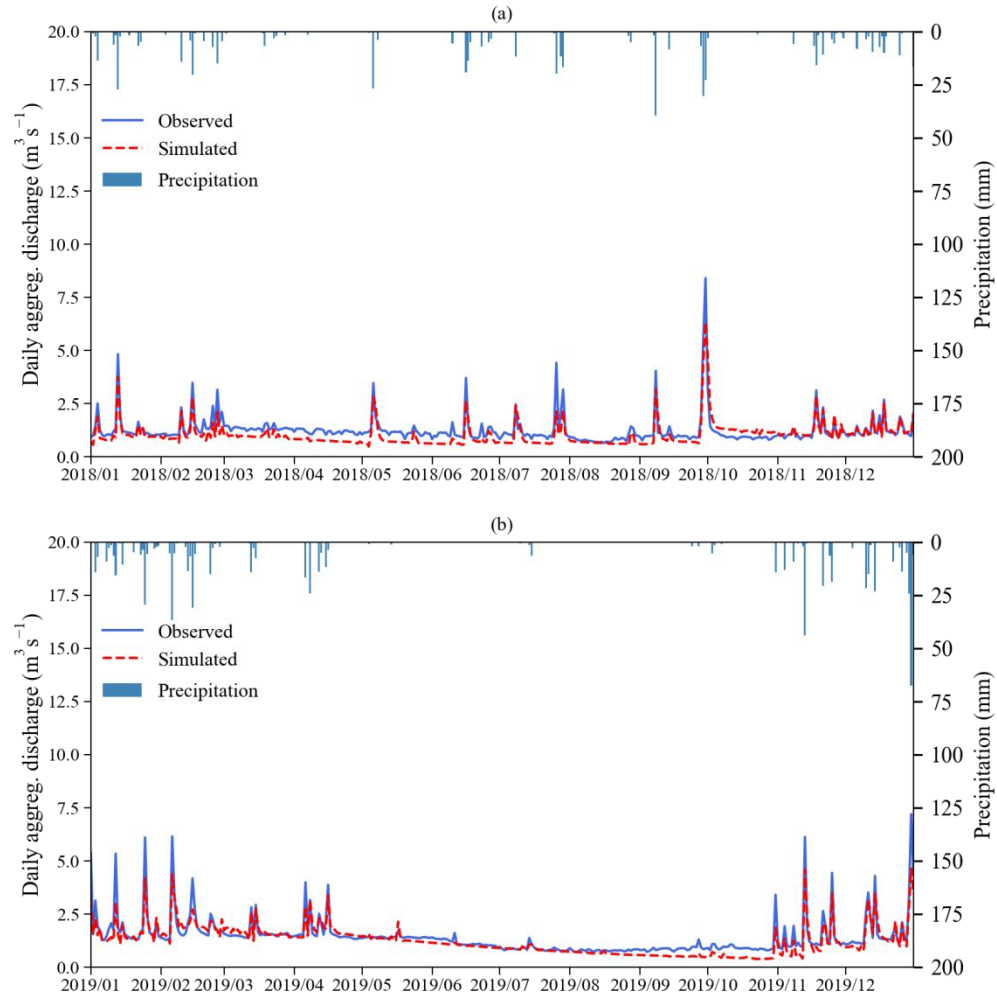


Figure 3. Observed and simulated daily discharge ($\text{m}^3 \text{s}^{-1}$) aggregated from hourly outputs: calibration period (a) and validation period (b).

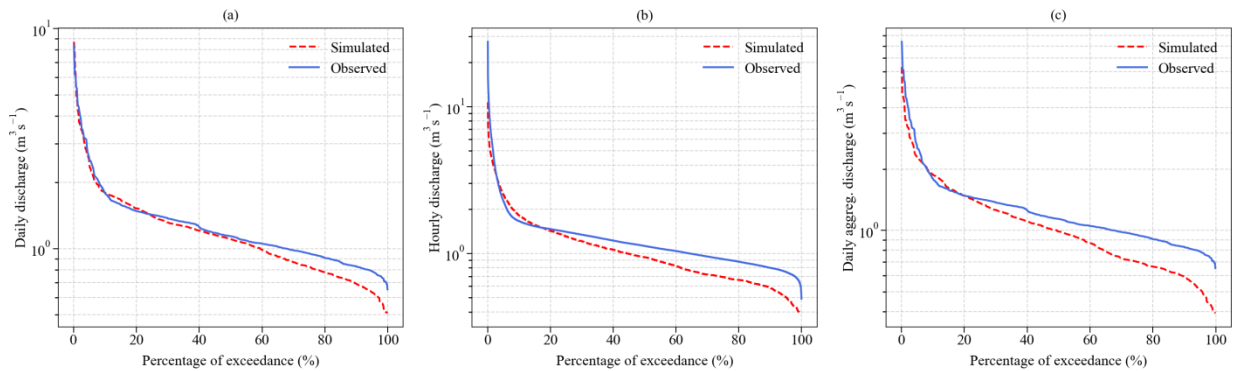


Figure 4. Observed and simulated flow duration curves ($\text{m}^3 \text{s}^{-1}$) at the daily time step (a), at the hourly time step (b), and at the daily discharge aggregated from hourly outputs time step (c).

Comment: L284: performance metrics are satisfactory, but performance metrics also depend on what we want to use the model for: for example, though the model here replicates the timeseries quite well, it could not be trusted for flood analysis (poor performance on hourly peaks).

Reply: We thank the referee for this comment. This study aims to investigate the complex hydrological processes that take place in a mixed-land-use basin in order to understand the mechanisms that create surface runoff. We agree with the referee that the aims and novelty of this study should be more emphasized. We will make the following changes in the Introduction section in the revised manuscript:

Lines 72-73: “In this study, the SWAT 2012 model (rev 681) in the QSWAT interface was used to simulate streamflow in an experimental basin using daily and sub-daily (hourly) rainfall observations. The main objectives were to (i) calibrate and validate the SWAT model using streamflow data, (ii) examine which parameters are more sensitive in different time steps, (iii) estimate the influence of rainfall resolution on model performance, (iv) compare the Curve Number method and Green and Ampt Mein Larson method for runoff simulation, (v) examine the accuracy of the sub-daily model and compare the peak discharges and time of peak of the two models in selected rainfall events, and (vi) investigate the suitability of the SWAT model for hourly simulation in a mixed-land-use basin (i.e., blended combinations of land use). Hence, this study will provide essential hydrological knowledge and contribute to the understanding of the earth surface processes of an urban/peri-urban hydrological system with complex land use in order to analyze the mechanisms governing surface runoff at the catchment scale”.

Comment: L289: what is 'ET runoff generation'?

Reply: We thank the reviewer for this comment. Kannan et al. (2007) identified a suitable combination of evapotranspiration (i.e., Penman-Monteith, Hargreaves) and infiltration (i.e., Curve Number, GAML) methods for runoff generation. We believe that the line 289 is not necessary. We will make the following changes in lines 289-293:

Lines 289-293: “Kannan et al. (2007) conducted a sensitivity analysis to identify the best combination of evapotranspiration and infiltration method for runoff generation and concluded that the CN method performed better than the GAML method for streamflow because the GAML method tends to hold more water in the soil profile and predict a lower peak runoff rate”.

Comment: L296: This is interesting but a little unclear: what is meant by 'too large'? Would the results be better with, say, 10 min rainfall? Why?

Reply: We thank the referee for highlighting this point. As we mentioned above the GAML method comparing to the CN method requires detailed soil data and precipitation data of high resolution. We wanted to emphasize the impact of rainfall resolution and accuracy on model performance. The selection of sub-daily precipitation time step is very crucial when using the

GAML method and it should be based according to the characteristics of each basin. We will rewrite the manuscript to be as follows:

Lines 295-299: “In this study, the daily model produced higher discharge peaks than the hourly model and generally estimated better the observed values. These results could be due to drawbacks of the GAML method, such as the requirement for detailed soil information and high resolution rainfall data in a sub-daily time step (King et al., 1999). The GAML method assumes that the soil profile is characterized by homogeneity and that the previous soil moisture is distributed uniformly in the soil profile (Jeong et al., 2010). Therefore, the uncertainty in the resolution of the rainfall data, the heterogeneity of the soil formations and the upcoming difficulty in determining the parameters' values for parameterization could affect the method's efficiency. The selection of sub-daily precipitation input time step as well as the resolution of the precipitation data have a great impact on model results when using the GAML method and it should be based on the scale and characteristics of the watershed (Bauwe et al., 2016; Jeong et al., 2010; Kannan et al., 2007). Furthermore, observational errors in the model input data (i.e., weather, soil and land use data) include inaccuracies in the estimation of channel and hillslope velocities and channel geometry, in the nature of the sensor, environmental conditions and data collection (Guzman et al., 2015). These errors can generate variability, lead to undesired trends, and influence the model calibration and validation results (Kamali et al., 2017). In addition, the complex land use characteristics and processes of an urban/peri-urban environment and assumptions made during the model structure/parameterization process (e.g., selection of parameters for calibration, objective function, and conceptual simplifications) increase the uncertainty of the results.”

Section 3.3

This section focuses on six "heavy" rainfall events in which the authors describe, in text, peak flow values and average flow values during the events. The hourly model underestimates peak flows. This section comes as a surprise for the reader as it is not mentioned in the Methods. Moreover, it is unclear how the events were selected: are they the 6 larger events of the timeries? It would be interesting to have an estimation of their return period to define "heavy"? From line 312 to line 335, the text simply describes hydrographs, without comments or analysis. Maybe the authors could consider a Table instead with rainfall characteristics (totals, duration, return period, etc.) and standard descriptors of hydrographs (peaks flows, difference between peaks, etc.)?

Reply: We thank the referee for the suggestions. We agree that Section 3.3 is not mentioned in the Methods. Therefore, we will mention this section in the Introduction in lines 72-73. We will discuss the model performance in Section 3.2 and we will discuss the selected rainfall events in Section 3.3. We also agree that the term "heavy" should be more explained. The rainfall events that were selected were the events with the highest rainfall intensity, discharge and return period. According to intensity-duration-frequency (IDF) curves of the study area the approximate return period of the selected episodes was ten years ($T=10$ years). We will also include a table with the rainfall characteristics of each event (i.e., peak discharge, time of peak and average discharge).

We will make the following changes in the revised manuscript in Lines 265-299:

Lines 265-299:

"Section 3.2 Daily and sub-daily model performances

Quantitative statistics and criteria recommended by Moriasi et al. (2007, 2015) were used to evaluate the model performance. In order to investigate the influence of rainfall on model performance and compare daily outputs to hourly outputs, the hourly outputs were aggregated to daily averages. Figure 3 shows the temporal dynamics of the hydrographs reproduced by both infiltration methods. The high flow season is observed during winter and spring. The low flow season is observed in summer and early fall due to high evapotranspiration. Figure 4 shows the observed versus the simulated daily discharge aggregated from hourly outputs during the calibration and validation processes. Figure 5 presents the flow duration curves of the models, indicating good agreement between observed and simulated values. Generally, in the sub-daily model, the simulated discharge peaks did not always match the observed values and were sometimes considerably lower.

The performance statistics are illustrated in Table 5 and indicate reasonable calibrated models for both infiltration approaches. Model performance using the CN method showed better results than the GAML method. In particular, the NSE and R^2 indices for the daily model were 0.84 and 0.79 for the calibration period and 0.87 and 0.86 for the validation period. For the sub-daily model the NSE and R^2 indices were 0.53 and 0.49 for the calibration period and 0.63 and 0.6 for the

validation period respectively. In addition, when the hourly outputs were aggregated to daily averages the NSE was improved comparing the NSE of the sub-daily model (e.g., sub-daily model: $NSE_{\text{calibration}} = 0.49$ and $NSE_{\text{validation}} = 0.6$, daily averages: $NSE_{\text{calibration}} = 0.66$ and $NSE_{\text{validation}} = 0.78$). However, the daily model outperformed the daily aggregated discharge during both calibration and validation periods. Furthermore, the daily model showed smaller modeling uncertainties with P-factor 0.79 and R-factor 1.58 (compared to 0.83 and 1.71 respectively for the sub-daily model).

Overall, the general agreement between the observed and the simulated values during the calibration and the validation period indicate that the choice of the calibration and validation periods was relevant. According to Moriasi et al. (2015) model performance can be evaluated as “satisfactory” for flow simulations if daily, monthly, or annual $R^2 > 0.60$, $NSE > 0.50$, and $PBIAS \leq \pm 15\%$ for watershed-scale models. These ratings should be modified to be more or less strict based on evaluation time step. Typically, model simulations are poorer for shorter time steps than for longer time steps (e.g., daily versus monthly or yearly) (Engel et al., 2007). Considering these guidelines, the daily and sub-daily models showed satisfactory performance for both calibration and validation periods.”

We will make the following changes in the revised manuscript in Lines 308-340:

Lines 308-340:

“3.3 Comparison of selected rainfall events

Figure 5 shows the hydrographs of selected high rainfall events that occurred in the years 2018 and 2019 (Tatoi station, Lagouvardos et al., 2017). According to intensity-duration-frequency (IDF) curves of the study area the approximate return period of the selected episodes was ten years ($T=10$ years). These events were investigated in order to examine the accuracy of the sub-daily model and to compare the peak discharges and time of peak of the two models. Table 6 displays the rainfall characteristics of each event (i.e., peak discharge, time of peak and average discharge).

Generally, the hourly model underestimated the peak flows with values much lower than the observations for the majority of the events. These results confirm that the CN method estimated better the observed values than the GAML method and was able to estimate with greater accuracy the peak discharge in most of the events. The better performance of the CN method in comparison to the GAML method in this study is consistent with the results of other studies (Bauwe et al., 2016; Ficklin and Zhang, 2013; Kannan et al., 2007; King et al., 1999). Bauwe et al. (2016) evaluated both CN and GAML methods and highlighted that the CN method performed slightly better than the GAML method. Ficklin and Zhang (2013) generally suggested that for daily simulations the CN method predicted more accurately streamflow as compared to the GAML model. Kannan et al. (2007) identified a suitable combination of ET runoff generation methods and reported that the CN method performed better than the GAML method.

Kannan et al. (2007) conducted a sensitivity analysis to identify the best combination of evapotranspiration and runoff method for hydrological modeling and concluded that the CN method performed better than the GAML method for streamflow because the GAML method tends to hold more water in the soil profile and predict a lower peak runoff rate. King et al. (1999) concluded that the GAML method appeared to have more limitations in accounting for seasonal variability than the CN method.

In this study, the daily model produced higher discharge peaks than the hourly model and generally estimated better the observed values. These results could be due to drawbacks of the GAML method, such as the requirement for detailed soil information and high resolution rainfall data in a sub-daily time step (King et al., 1999). The GAML method assumes that the soil profile is characterized by homogeneity and that the previous soil moisture is distributed uniformly in the soil profile (Jeong et al., 2010). Therefore, the uncertainty in the resolution of the rainfall data, the heterogeneity of the soil formations and the upcoming difficulty in determining the parameters' values for parameterization could affect the method's efficiency. The selection of sub-daily precipitation input time step as well as the resolution of the precipitation data have a great impact on model results when using the GAML method and it should be based on the scale and characteristics of the watershed (Bauwe et al., 2016; Jeong et al., 2010; Kannan et al., 2007). Furthermore, observational errors in the model input data (i.e., weather, soil and land use data) include inaccuracies in the estimation of channel and hillslope velocities and channel geometry, in the nature of the sensor, environmental conditions and data collection (Guzman et al., 2015). These errors can generate variability, lead to undesired trends, and influence the model calibration and validation results (Kamali et al., 2017). In addition, the complex land use characteristics and processes of an urban/peri-urban environment and assumptions made during the model structure/parameterization process (e.g., selection of parameters for calibration, objective function, and conceptual simplifications) increase the uncertainty of the results.”

Table 6: Rainfall characteristics of selected events for the years 2018 and 2019.

Events	Observed			Simulated		
	Average discharge (m ³ /s)	Peak discharge (m ³ /s)	Time of peak (UTC)	Average discharge (m ³ /s)	Peak discharge (m ³ /s)	Time of peak (UTC)
12/01-14/01/2018	2.6	10.7	6:00	2.2	7.1	5:00
05/05-07/05/2018	2.2	11.1	20:00	2.1	6.6	21:00
29/09-01/10/2018	5.7	17.2	18:00	5.2	8.9	18:00
05/02-07/02/2019	3.6	16.2	1:00	2.9	6.2	00:00
12/11-14/11/2019	2.9	12.3	3:00	2.4	3.5	3:00
29/12-31/12/2019	4.9	14.8	21:00	3.6	6.9	21:00

It is concluded that the underestimation of peak flows is due to uncertainty in observed data or input data (rainfall), but without many arguments or any estimation of these uncertainties. How can one be sure that the errors are due to the data and not to the model? There probably exists a parametrization which can replicate high flows much better, with poorer performance on low flows?

Reply: We thank the referee for highlighting this point. The issue of uncertainty is a major topic in hydrological modeling and has been discussed in many publications (Harmel et al., 2006, 2014; Kamali et al., 2017; Kouchi et al., 2017; Tan et al., 2020). In general, model errors are due to inaccuracies (i) in the quality of input data (i.e., weather, soil and land use data), (ii) in the conceptual model, (iii) in the choice of objective function, (iv) the observed data and (v) in the parameterization. The correct selection and combination of the values of the parameters that influence surface runoff, groundwater, channel routing and evapotranspiration is a critical point in model calibration (Polanco et al., 2017). There are many possible combinations of parameters that can replicate high flows much better but in every case the ranges of the calibrating parameters should be kept in reasonable limits using quantitative statistics and graphical comparisons in order to ensure that hydrological processes represent the characteristic of the study area (Daggupati et al., 2015). In this study, the initial ranges of the calibrating parameters were set according to literature and sensitivity analysis. Then, based on the performance of the default model, specific parameters were parameterized using calibration protocols (Abbaspour et al., 2015). We will emphasize in the revised manuscript the many sources of uncertainty that exist in this study and we will discuss them in the following comment.

Comment: L336-340: It is correct that errors in a model can be explained by 1/ uncertainty in the observed data 2/ uncertainty in input data or 3/ the model structure/parameterization. But what about this particular study? This could be further discussed.

Reply: We thank the referee for this comment. The values of the calibrated parameters and their sensitivities are influenced by the type and quality of input data, the conceptual model, the choice of the objective function and inaccuracies in measured input data used for calibration and validation (Abbaspour et al., 2015; Arnold et al., 2012; Polanco et al., 2017). In this study the errors can be explained by the uncertainty of the input data (e.g., quality of the precipitation data, land use data, observed discharge data, Manning's equation for flow estimation), the complex land use characteristics of an urban/peri-urban environment which are difficult to simulate in SWAT, and the differences behind the mechanisms of the CN method and the GAML method for streamflow estimation. In addition, assumptions made during the model structure/parameterization process (e.g., selection of parameters for calibration, objective function, and conceptual simplifications) increase the uncertainty of the results.

We will make the following changes in the revised manuscript in lines 336-340:

Lines 336-340: “In this study, the daily model produced higher discharge peaks than the hourly model and generally estimated better the observed values. These results could be due to drawbacks of the GAML method, such as the requirement for detailed soil information and high resolution rainfall data in a sub-daily time step (King et al., 1999). The GAML method assumes that the soil profile is characterized by homogeneity and that the previous soil moisture is distributed uniformly in the soil profile (Jeong et al., 2010). Therefore, the uncertainty in the resolution of the rainfall data, the heterogeneity of the soil formations and the upcoming difficulty in determining the parameters' values for parameterization could affect the method's efficiency. The selection of sub-daily precipitation input time step as well as the resolution of the precipitation data have a great impact on model results when using the GAML method and it should be based on the scale and characteristics of the watershed (Bauwe et al., 2016; Jeong et al., 2010; Kannan et al., 2007). Furthermore, observational errors in the model input data (i.e., weather, soil and land use data) include inaccuracies in the estimation of channel and hillslope velocities and channel geometry, in the nature of the sensor, environmental conditions and data collection (Guzman et al., 2015). These errors can generate variability, lead to undesired trends, and influence the model calibration and validation results (Kamali et al., 2017). In addition, the complex land use characteristics and processes of an urban/peri-urban environment and assumptions made during the model structure/parameterization process (e.g., selection of parameters for calibration, objective function, and conceptual simplifications) increase the uncertainty of the results.”

We will also change the Conclusion section to be as follows:

Lines 347-369: “Experimental catchments provide long term time series of hydrological data which are essential for improved application of best management practices and the development and validation of watershed models. In this study, discharge was monitored for three years (2017-2019) in an experimental basin with mixed-land-use characteristics (i.e., urban/peri-urban), located in Athens, Greece. Discharge simulation, calibration and validation were achieved with the application of SWAT model, which has been increasingly used to support decisions on various environmental issues and policy directions. Daily and hourly rainfall observations were used as inputs to investigate the influence of rainfall resolution on model performance in order to analyze the mechanisms governing surface runoff at the catchment scale. Surface runoff was estimated using the CN method for the daily model and the GAML method for the hourly model.

A sensitivity analysis conducted in this study showed that the parameters related to groundwater flow were more sensitive for daily time intervals and channel routing parameters were more influential for hourly time intervals. These findings indicate that the model operational time step affect parameters' sensitivity to the model output, thus demonstrating the need for different model strategy for the simulation of sub-daily hydrological processes.

Quantitative statistics of the observed and the simulated records indicate that the calibration and validation processes produced acceptable results for both infiltration methods. Additionally, graphical techniques at the outlet station show that both models succeed in capturing majority of seasonality and peak discharge. Generally, the daily model performed better than the sub-daily model in simulating runoff. The CN method produced higher discharge peaks than the GAML method and estimated better the observed values. The differences in the calibrated values of the two models lay mostly in the different runoff estimation methods used by the two models. In addition, errors in the quality of input data, the complex land use characteristics of an urban/peri-urban environment and assumptions made during the model structure/calibration process may increase the uncertainty of the outputs.

Overall, the general agreement between observations and simulations in both models suggests that the SWAT model appears to be a reliable tool to predict discharge in a mixed-land-use basin with high complexity and spatial distribution of input data. Furthermore, this study contributed to the understanding of the mechanisms controlling surface runoff and the parameters that influence the hydrological processes that take place in an urban/peri-urban hydrological environment. It should be noted that several factors such as data limitation, observational errors in input data, complexities of spatial and temporal scales, and inaccuracies in model structure may lead to uncertainty in model outputs. In the future, emphasis will be placed in the quantification of the parameter uncertainty by including more observed variables in the calibration process such as evapotranspiration and soil moisture satellite data.”

4. Conclusions:

Comment: L366: 3 years is not really "long time".

Reply: We agree with the referee that three years is not a long period of time. We believe that we should focus in the land use characteristics of the study area and in the ability of SWAT to predict discharge in an urban/peri-urban environment. We will make the following changes in the manuscript in line 366.

Line 366: “Overall, the general agreement between observations and simulations in both models suggests that the SWAT model appears to be a reliable tool to predict discharge in a mixed-land-use basin with high complexity and spatial distribution of input data.”

Comments on Figures:

Comment: Figure 1 and Figure 2 could be merged

Reply: We agree with the referee. We merged Figure 1 and Figure 2. We will add the following figure in the revised manuscript:

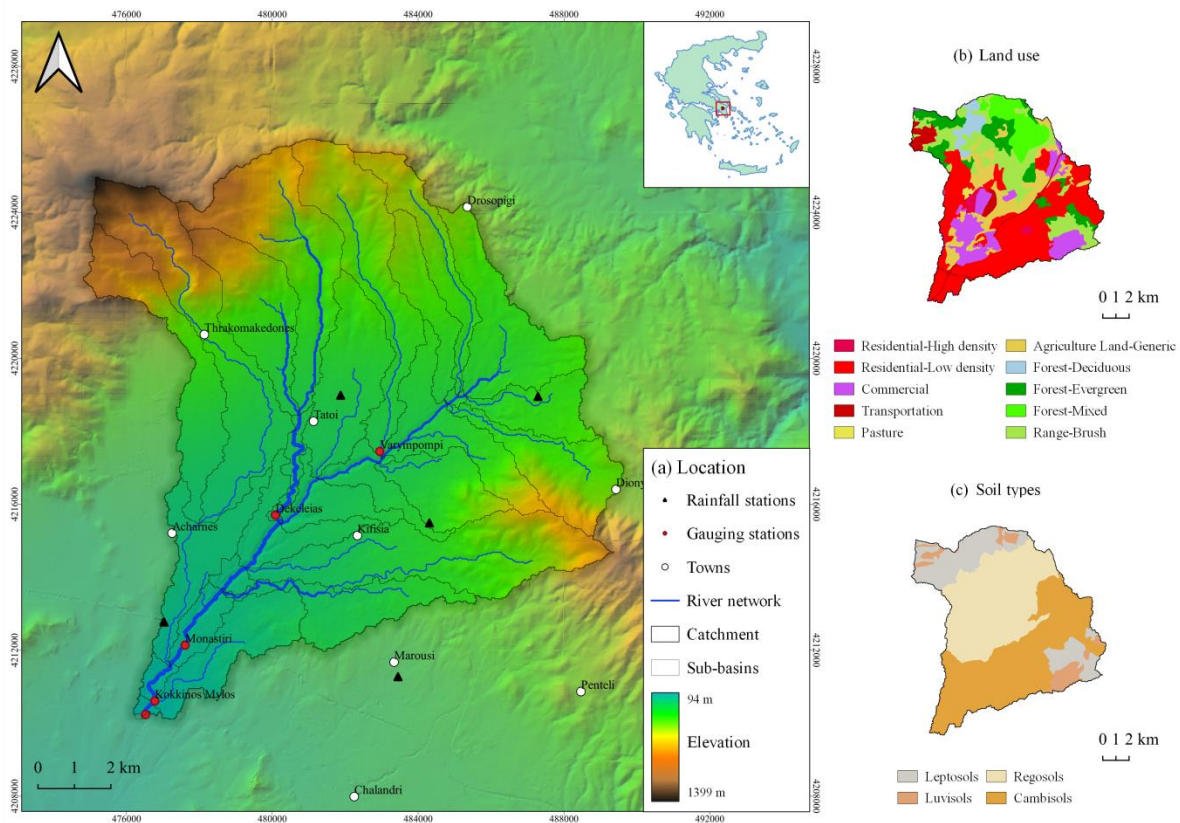


Figure 1. Geographical location of the study area (a) and spatial distribution of land use (b) and soil (c).

We will also make the following changes in the manuscript in lines 80-87:

Lines 80-87: “The study area includes the upper part (NW sub-basin) of the Kifissos River basin, located in Athens Greece (Fig. 1a). The Kifissos River basin occupies an area of 380 km² and its route is approximately 22 km, of which at least 14 km are within an urban area. The elevation ranges from 94 m to 1399 m with plains in the south and hills in the north part of the basin. The mean annual temperature is 16.4 °C and the mean annual rainfall across the basin is 577.2 mm.

The study area is characterized as an urban/sub-urban area, with residential areas, shrubland and agriculture accounting for 34.1, 15.9 and 12.4 % of its land use coverage, respectively (Fig. 1b). It includes mainly four soil types, Cambisols, Regosols, Leptosols and Luvisols (Fig. 1c). The dominant soil formations are characterized by good soil permeability and high contents of clay and sand.”

Comment: Figure 3: It could be worth to add rainfall?

Reply: We agree with the referee. We added rainfall in Figure 3. We will add the following figure in the revised manuscript:

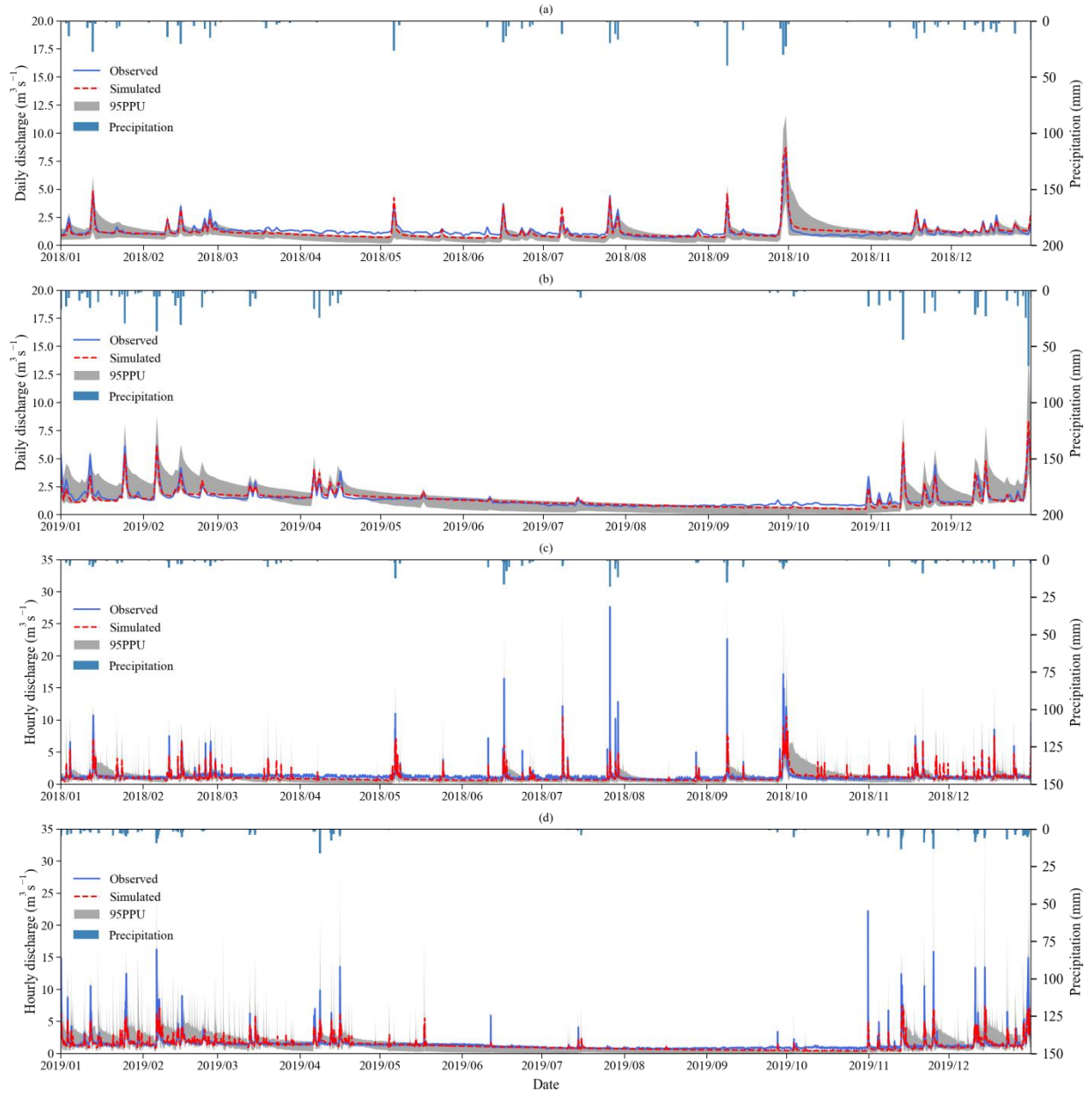


Figure 2. Observed and simulated discharge ($\text{m}^3 \text{s}^{-1}$) at the daily time step (a, b) and at the hourly time step (c, d).

Minor comments and typos

Comment: L12: this study demonstrates (remove the comma)

Reply: We agree with the referee. We will make the following changes in lines 12-14:

Lines 12-14: “This study examines the application of SWAT model for streamflow simulation in an experimental basin with mixed-land-use characteristics (i.e., urban/peri-urban) using daily and hourly rainfall observations.”

Comment: L29: they are 'used to monitor', not 'able'

Reply: We agree with the referee. We will make the following changes in lines 29-30:

Lines 29-30: “They are also used to monitor the major components of the surface hydrological cycle by using remote sensing and geophysical measurements (Tauro et al., 2018).”

Comment: L30: they monitor groundwater

Reply: We agree with the referee. We will make the following changes in lines 30-31:

Lines 30-31: “Furthermore, they monitor groundwater and river water quality with the use of tracer experiments which can estimate the residence and travel times of water in different components of the hydrological cycle (Hrachowitz et al., 2016; Stockinger et al., 2016).”

Comment: L81: "route" is unclear

Reply: We thank the referee for this comment. We will rephrase line 81 in the following way:

Line 81: “The Kifissos River basin occupies an area of 380 km². Kifissos River route is approximately 22 km, of which at least 14 km are within an urban area.”

References

- Abbaspour, K. C., Rouholahnejad, E., Vaghefi, S., Srinivasan, R., Yang, H. and Kløve, B.: A continental-scale hydrology and water quality model for Europe: Calibration and uncertainty of a high-resolution large-scale SWAT model, *Journal of Hydrology*, 524(May), 733–752, doi:10.1016/j.jhydrol.2015.03.027, 2015.
- Arnold, J. G., Moriasi, D. N., Gassman, P. W., Abbaspour, K. C., White, M. J., Srinivasan, R., Santhi, C., Harmel, R. D., Van Griensven, A., Van Liew, M. W., Kannan, N. and Jha, M. K.: SWAT: Model Use, Calibration, and Validation, *Transactions of the ASABE*, 55(4coper), 1491–1508, doi:10.13031/2013.42256, 2012.
- Bauwe, A., Kahle, P. and Lennartz, B.: Hydrologic evaluation of the curve number and Green and Ampt infiltration methods by applying Hooghoudt and Kirkham tile drain equations using SWAT, *Journal of Hydrology*, 537, 311–321, doi:10.1016/j.jhydrol.2016.03.054, 2016.
- Boithias, L., Sauvage, S., Lenica, A., Roux, H., Abbaspour, K., Larnier, K., Dartus, D. and Sánchez-Pérez, J.: Simulating Flash Floods at Hourly Time-Step Using the SWAT Model, *Water*, 9(12), 929, doi:10.3390/w9120929, 2017.
- Daggupati, P., Yen, H., White, M. J., Srinivasan, R., Arnold, J. G., Keitzer, C. S. and Sowa, S. P.: Impact of model development, calibration and validation decisions on hydrological simulations in West Lake Erie Basin, *Hydrological Processes*, 29(26), 5307–5320, doi:10.1002/hyp.10536, 2015.
- Engel, B., Storm, D., White, M., Arnold, J. and Arabi, M.: A Hydrologic/Water Quality Model Application, *Journal of the American Water Resources Association*, 43(5), 1223–1236, doi:10.1111/j.1752-1688.2007.00105.x, 2007.
- Ficklin, D. L. and Zhang, M.: A Comparison of the Curve Number and Green-Ampt Models in an Agricultural Watershed, *Transactions of the ASABE*, 56(1), 61–69, doi:10.13031/2013.42590, 2013.
- Guzman, J. A., Shirmohammadi, A., Sadeghi, A. M., Wang, X., Chu, M. L., Jha, M. K., Parajuli, P. B., Harmel, R. D., Khare, Y. P. and Hernandez, J. E.: Uncertainty considerations in calibration and validation of hydrologic and water quality models, *Transactions of the ASABE*, 58(6), 1745–1762, doi:10.13031/trans.58.10710, 2015.
- Harmel, R. D., Cooper, R. J., Slade, R. M., Haney, R. L. and Arnold, J. G.: Cumulative uncertainty in measured streamflow and water quality data for small watersheds, *Transactions of the ASABE*, 49(3), 689–701, 2006.
- Harmel, R. D., Smith, P. K., Migliaccio, K. W., Chaubey, I., Douglas-Mankin, K. R., Benham, B., Shukla, S., Muñoz-Carpena, R. and Robson, B. J.: Evaluating, interpreting, and communicating performance of hydrologic/water quality models considering intended use: A review and recommendations, *Environmental Modelling & Software*, 21, 40–51, 2014.
- Hrachowitz, M., Benettin, P., van Breukelen, B. M., Fovet, O., Howden, N. J. K., Ruiz, L., van der Velde, Y. and Wade, A. J.: Transit times-the link between hydrology and water quality at the

catchment scale, *Wiley Interdisciplinary Reviews: Water*, 3(5), 629–657, doi:10.1002/wat2.1155, 2016.

Jeong, J., Kannan, N., Arnold, J., Glick, R., Gosselink, L. and Srinivasan, R.: Development and Integration of Sub-hourly Rainfall-Runoff Modeling Capability Within a Watershed Model, *Water Resources Management*, 24(15), 4505–4527, doi:10.1007/s11269-010-9670-4, 2010.

Kamali, B., Abbaspour, K. C. and Yang, H.: Assessing the uncertainty of multiple input datasets in the prediction of water resource components, *Water (Switzerland)*, 9(9), doi:10.3390/w9090709, 2017.

Kannan, N., White, S. M., Worrall, F. and Whelan, M. J.: Sensitivity analysis and identification of the best evapotranspiration and runoff options for hydrological modelling in SWAT-2000, *Journal of Hydrology*, 332(3–4), 456–466, doi:10.1016/j.jhydrol.2006.08.001, 2007.

King, K. W., Arnold, J. G. and Bingner, R. L.: Comparison of Green-Ampt and curve number methods on Goodwin Creek Watershed using SWAT, *Transactions of the American Society of Agricultural Engineers*, 42(4), 919–925, doi:10.13031/2013.13272, 1999.

Kouchi, D. H., Esmaili, K., Faridhosseini, A., Sanaeinejad, S. H., Khalili, D. and Abbaspour, K. C.: Sensitivity of calibrated parameters and water resource estimates on different objective functions and optimization algorithms, *Water (Switzerland)*, 9(6), 1–16, doi:10.3390/w9060384, 2017.

Moriasi, D. N., Arnold, J. G., Van Liew, M. W., Bingner, R. L., Harmel, R. D. and Veith, T. L.: Model Evaluation Guidelines for Systematic Quantification of Accuracy in Watershed Simulations, *Transactions of the ASABE*, 50(3), 885–900, doi:10.13031/2013.23153, 2007.

Moriasi, D. N., Gitau, M. W., Pai, N. and Daggupati, P.: Hydrologic and water quality models: Performance measures and evaluation criteria, *Transactions of the ASABE*, 58(6), 1763–1785, doi:10.13031/trans.58.10715, 2015.

Open Hydrosystem Information Network, (OpenHi.net): Observed streamflow data, [online] Available from: <https://openhi.net/> (Accessed 20 December 2020), n.d.

Polanco, E. I., Fleifle, A., Ludwig, R. and Disse, M.: Improving SWAT model performance in the upper Blue Nile Basin using meteorological data integration and subcatchment discretization, *Hydrology and Earth System Sciences*, 21(9), 4907–4926, doi:10.5194/hess-21-4907-2017, 2017.

Smart Toolbox for Engaging Citizens in a People-Centric Observation Web, (SCENT): Discharge stations, [online] Available from: <https://scent-project.eu/> (Accessed 20 December 2020), n.d.

Stockinger, M. P., Bogen, H. R., Lücke, A., Diekkrüger, B., Cornelissen, T. and Vereecken, H.: Tracer sampling frequency influences estimates of young water fraction and streamwater transit time distribution, *Journal of Hydrology*, 541, 952–964, doi:10.1016/j.jhydrol.2016.08.007, 2016.

Tan, M. L., Gassman, P. W., Yang, X. and Haywood, J.: A review of SWAT applications, performance and future needs for simulation of hydro-climatic extremes, *Advances in Water*

Resources, 143(June), 103662, doi:10.1016/j.advwatres.2020.103662, 2020.

Tauro, F., Selker, J., van de Giesen, N., Abrate, T., Uijlenhoet, R., Porfiri, M., Manfreda, S., Caylor, K., Moramarco, T., Benveniste, J., Ciruolo, G., Estes, L., Domeneghetti, A., Perks, M. T., Corbari, C., Rabiei, E., Ravazzani, G., Bogena, H., Harfouche, A., Brocca, L., Maltese, A., Wickert, A., Tarpanelli, A., Good, S., Lopez Alcala, J. M., Petroselli, A., Cudennec, C., Blume, T., Hut, R. and Grimaldi, S.: Measurements and Observations in the XXI century (MOXXI): innovation and multi-disciplinarity to sense the hydrological cycle, *Hydrological Sciences Journal*, 63(2), 169–196, doi:10.1080/02626667.2017.1420191, 2018.