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*Supplement of*

## **Impact of modellers' decisions on hydrological a priori predictions**

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## 1 **Supplementary material (A)**

2 Here we describe the features of the ten models and document the underlying assumptions  
3 and parameterisation used for the three consecutive predictions. The change of prediction  
4 quality induced by modifying some assumptions and parameters is rated with the Root Mean  
5 Square Error (RMSE) and the Nash-Sutcliffe Efficiency Index (Eq. 1 and 2). Note that we  
6 present two of each error estimates for the first year, one based on a reduced record excluding  
7 the extreme snow melt event of January 27, 2006 (abbreviated with w/o SME) and one using  
8 the entire record. For the two following years the entire record was used and hence only one  
9 error estimate is presented,

### 10 **A-1 Catflow**

- 11 • *Model user*

12 T. Blume (GFZ German Research Centre for Geosciences, Potsdam)

- 13 • *Basic model features*

14 Physically-based model with detailed process representation, e.g. soil-water dynamic based  
15 on Richards equation, evapotranspiration on Penman-Monteith equation and surface runoff  
16 on convection-diffusion equation. The model allows simulation of infiltration excess  
17 runoff, surface saturation, lateral subsurface flow, reinfiltration of surface runoff, and  
18 return flow (Maurer, 1997).

- 19 • *Assumptions and parameters (1<sup>st</sup> prediction)*

20 Basic assumption: Best representation of catchment properties yields best prediction.

21 Main attempt was to minimize the influence of the modeller's choices. No decision on  
22 dominant process(es) using the model primarily as a platform for hypothesis testing. Soil  
23 properties were derived from Carsel and Parrish (1988) ( $K_{\text{sat}} = 146$  mm/h for sand and  
24  $K_{\text{sat}} = 13$  mm/h for sandy clay loam). Soil crusts were considered to be dominant but not  
25 included into the model assuming that only the provided data set should be used.

- 26 • *Assumptions and parameters (2<sup>nd</sup> prediction)*

27 Introduction of the crust layer ( $K_{\text{sat}} = 0.6$  mm/h according to Simunek et al. (1998)) to  
28 allow infiltration excess.  $K_{\text{sat}}$  was reduced to limit subsurface flow. Modifying the lower  
29 boundary condition to better represent the influence of the clay dam.

- 30 • *Assumptions and parameters (3<sup>rd</sup> prediction)*

31 The spatial resolution of the model grid (upper slope 10 m, middle slope 5 m and lower  
32 slope 1 m resolution) was changed to 5 m in order to reduce the numerical problems

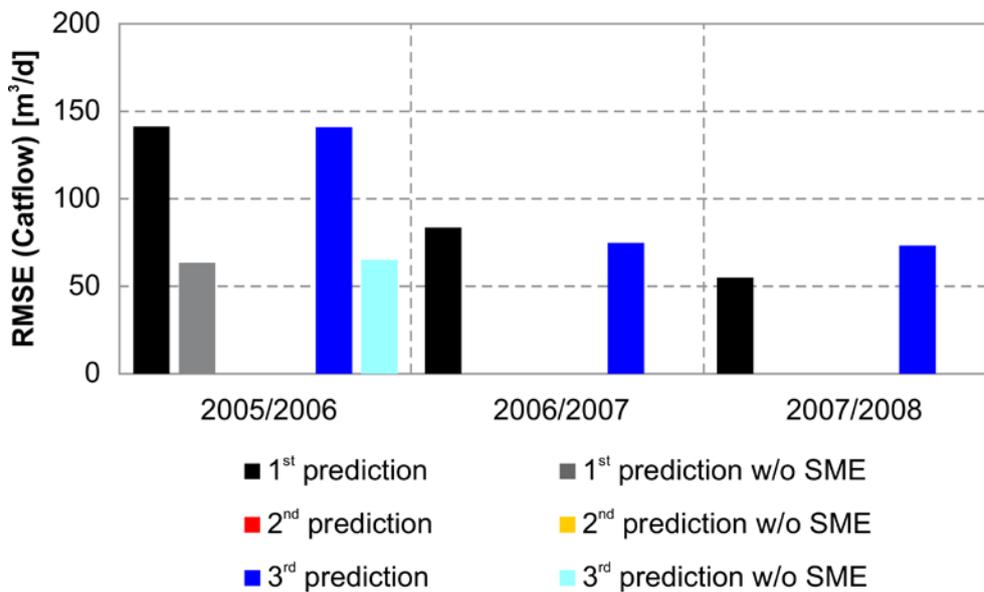
1 caused by the implementation of the surface crust. Soil crust (5 cm) was resolved with  
2 vertical increments of 1 cm (before: top 20 cm with 4 cm resolution). This resulted in an  
3 18% reduction of the total number of nodes.

4 Only the soil hydraulic data and the infiltration rates were selected. The virtual costs were  
5 the second lowest of all modellers. These data were used to parameterize the soil and the  
6 soil crust.

7 • *Results*

8 In the 1<sup>st</sup> and 3<sup>rd</sup> prediction, Catflow discharge was among the largest of all models (e.g.,  
9 262 mm/y and 255 mm/y in 2006/2007) mainly due to the large base flow of 20 to 25 m<sup>3</sup>/d.  
10 Due to numerical problems Catflow's 2<sup>nd</sup> prediction is missing.

11 For the first two years the model performance improved from the 1<sup>st</sup> to the 3<sup>rd</sup> prediction  
12 (Fig A-1a and A-1b). Neglecting the snowmelt event (SME) discharge was simulated for  
13 nearly all years with a similar prediction uncertainty. Q<sub>90</sub> (77 m<sup>3</sup>/d, 3<sup>rd</sup> prediction stage)  
14 was the second largest of all events and Q<sub>5</sub> was >20 m<sup>3</sup>/d. Therefore, the base flow was  
15 very high compared to surface runoff, which was the lowest relative runoff of all models  
16 (<1% in the 1<sup>st</sup> year to 25% in the 3<sup>rd</sup> year). The initial conditions used a higher  
17 groundwater table than the observed one and reaches complete water saturation of the  
18 whole aquifer early in 2008 (Supplementary Material (E)).



19  
20 Fig. A-1a: RMSE for all predictions of Catflow

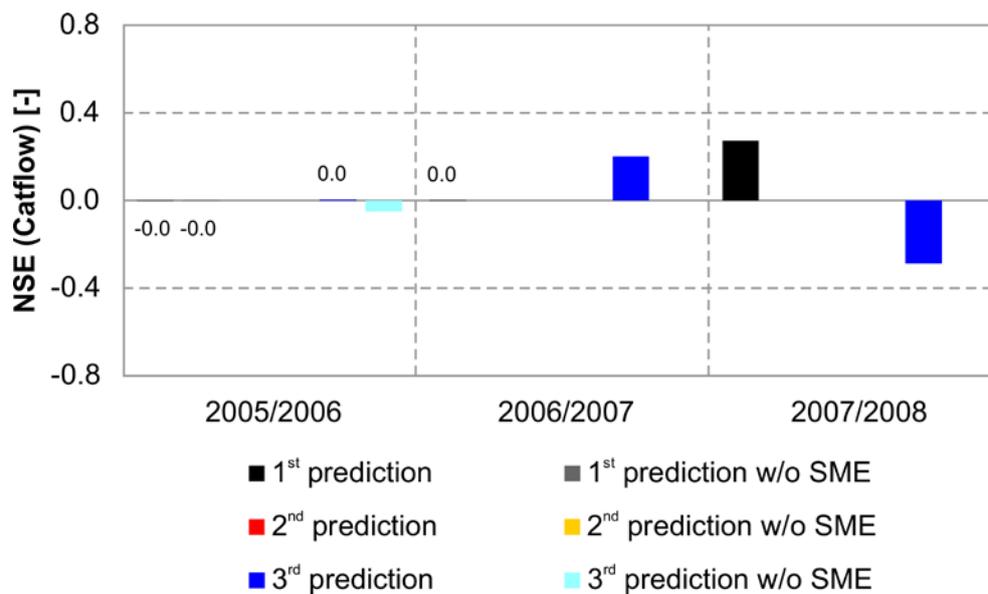


Fig. A-1b: NSE for all predictions of Catflow

## A-2 CMF

- *Model user*

P. Kraft (University of Giessen)

- *Basic model features*

Distributed hydrological model that calculates infiltration and unsaturated percolation using Richards equation and the lateral saturated flow using Darcy's law. Infiltration excess is routed to the stream network (Kraft et al., 2008).

- *Assumptions and parameters (1<sup>st</sup> prediction)*

Opposite to the other models CMF was developed by the user himself using an irregular grid. The modeller intended to evaluate how CMF performs in a region where the model was not used before. His initial parameterization followed a traditional procedure, e.g. using average soil properties derived from texture data ( $K_{sat} = 416.7$  mm/h, AG Boden (1994)) and not considering soil freezing on  $K_{sat}$ .

- *Assumptions and parameters (2<sup>nd</sup> prediction)*

Although the field visit showed the importance of surface runoff due to the deep gullies, the soil crust (to reduce infiltration) could not be implemented due to the model structure. However, the modeller expected that water exfiltrates through preferential flow pathways into the gullies even when the topsoil was still unsaturated. This substantially increased surface runoff through the gullies. In order to more realistically represent the newly constructed catchment, the modeller changed the initial state to quasi-dry conditions (pF

1 2.5).  $K_{sat}$  was reduced to  $60 \pm 30$  mm/h to diminish the large discharge. Finally, the leaf  
 2 area index (LAI) was set to 1.

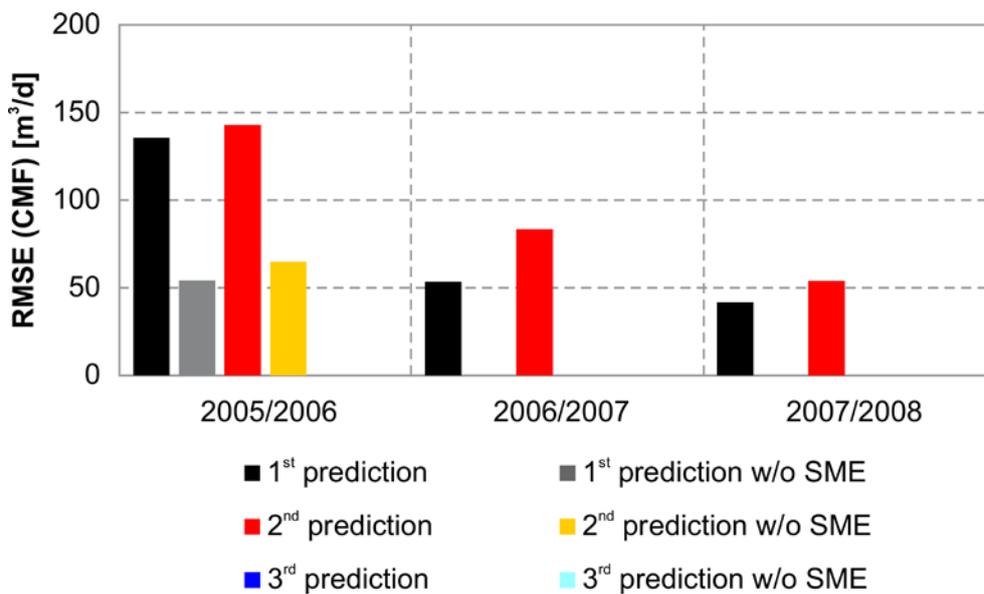
3 • *Assumptions and parameters (3<sup>rd</sup> prediction)*

4 After 2<sup>nd</sup> stage prediction the CMF modeller completed his PhD-program and left the  
 5 modelling group.

6 • *Results*

7 In the 1<sup>st</sup> prediction stage, the modeller reported one of the largest annual discharge (~220  
 8 mm/y), a peak discharge of 461 m<sup>3</sup>/d, and a Q<sub>5</sub> of 20 m<sup>3</sup>/d. In 2<sup>nd</sup> prediction peak discharge  
 9 was only 43 m<sup>3</sup>/d, one of the lowest of all models (e.g. 109 mm/y in the 2006/2007). The  
 10 Q<sub>5</sub> of about 1 m<sup>3</sup>/d shows that the predicted slow flow components were low compared to  
 11 the other models. PET prediction was also lowest (2<sup>nd</sup> prediction, 443 mm/y in 2005/2006)  
 12 while most of the other models predicted 600 to 800 mm/y. Since CMF predicted the  
 13 largest AET/PET ratio of all models (60 to 80%), the AET was average. However, both  
 14 statistical measures show a decrease in prediction quality from the 1<sup>st</sup> to the 2<sup>nd</sup> prediction.

15 The groundwater module of CMF was not responding to drying periods so that the  
 16 simulated groundwater table was since spring 2008 in a quasi-steady state at a depth of  
 17 ~30 cm.



18

19 Fig. A-2a: RMSE for all prediction of CMF

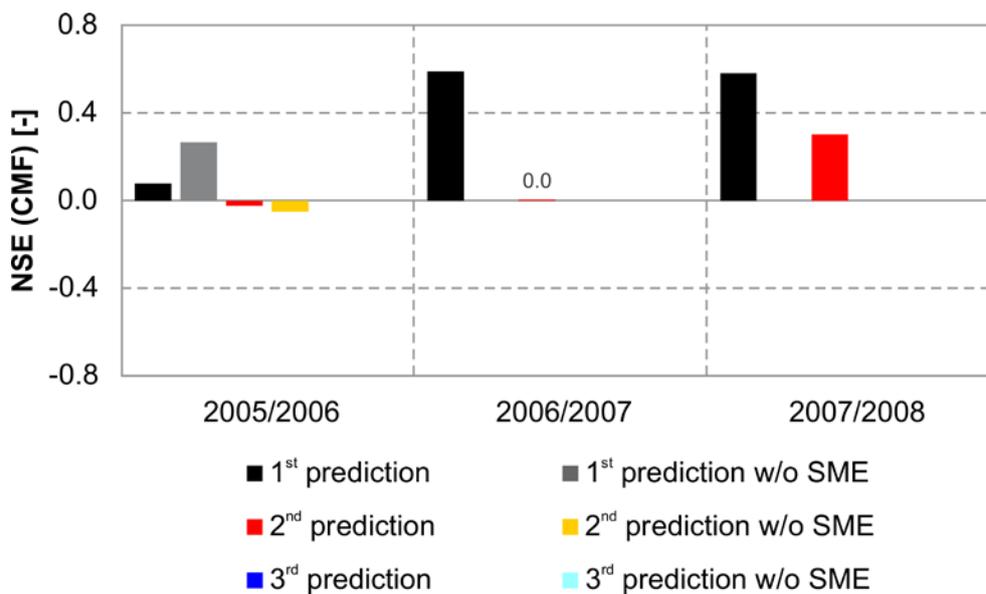


Fig. A-2b: NSE for all prediction of CMF

### A-3 CoupModel

- *Model user*

D. Gustafsson (Royal Institute of Technology KTH, Stockholm)

- *Basic model features*

Physically based model which represents the soil-water dynamic based on Richards equation. Lateral fluxes are considered as drainage to the downstream column. Model accounts for soil freezing, including effects on the thermal and hydraulic conductivity (Jansson and Moon, 2001).

- *Assumptions and parameters (1<sup>st</sup> prediction)*

Average soil properties were derived from texture data ( $K_{sat} = 84$  mm/h, Lundmark and Jansson (2009)) since the sandy soil was assumed to be homogeneous. Soil and snow evaporation are considered to be important.

- *Assumptions and parameters (2<sup>nd</sup> prediction)*

The different development of the vegetation in the western and eastern half of the catchment resulted in a different parameterizations. The modeller assumed dry initial condition and reduced  $K_{sat}$  to slow downstream subsurface flow.

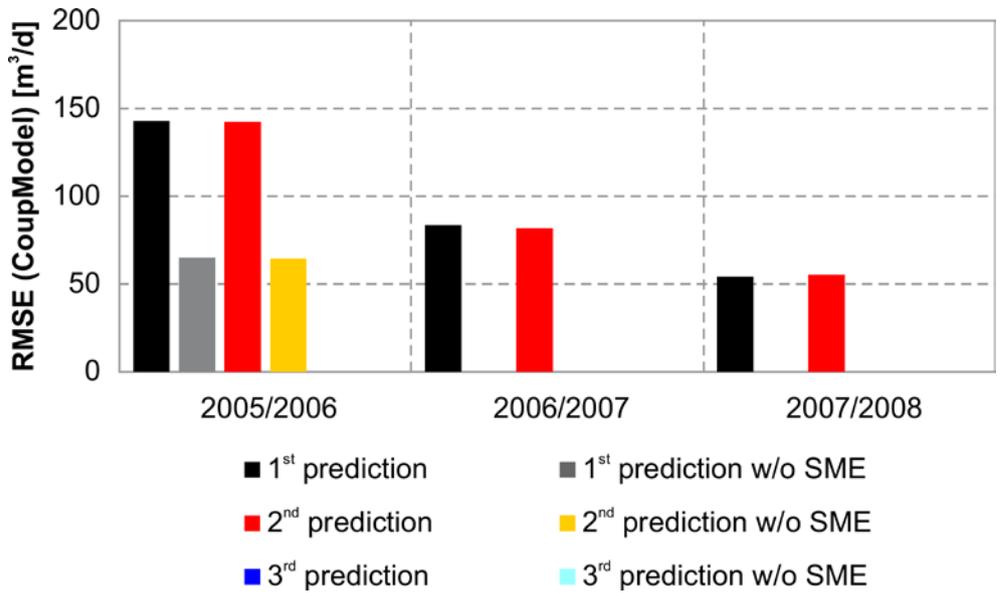
- *Assumptions and parameters (3<sup>rd</sup> prediction)*

After 2<sup>nd</sup> stage prediction the CoupModel modeller left the modelling group due to time constraints.

1 • *Results*

2 The model predicted the lowest annual discharges in the 1<sup>st</sup> prediction stage with less than  
3 15 mm/y base flow and 60 mm/y of surface run-off. CoupModel was the only model,  
4 which included snow melt. However, the discharge during the SME was strongly under-  
5 predicted. Neglecting the SME resulted therefore in a lower RMSE (Fig. A-3a).

6 Although the discharge was increased to an average of 145 mm/y the RMSE and the NSE  
7 do not show a significant change (Fig. A-3a and A-3b). While the peak discharge  
8 decreased slightly from 77 m<sup>3</sup>/d to 65 m<sup>3</sup>/d from the 1<sup>st</sup> to 2<sup>nd</sup> prediction, the slower flow  
9 events increased (e.g.  $Q_{25} = 1 \text{ m}^3/\text{d}$  in the 1<sup>st</sup> prediction and  $Q_{25} = 11 \text{ m}^3/\text{d}$  in the 2<sup>nd</sup>  
10 prediction). Still, the model predicted also 2% of all events with zero discharge. The  
11 CoupModel used a too small storage coefficient and predicted too small groundwater table  
12 fluctuations (Supplementary Material (D)).



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Fig. A-3a: RMSE for all prediction of CoupModel

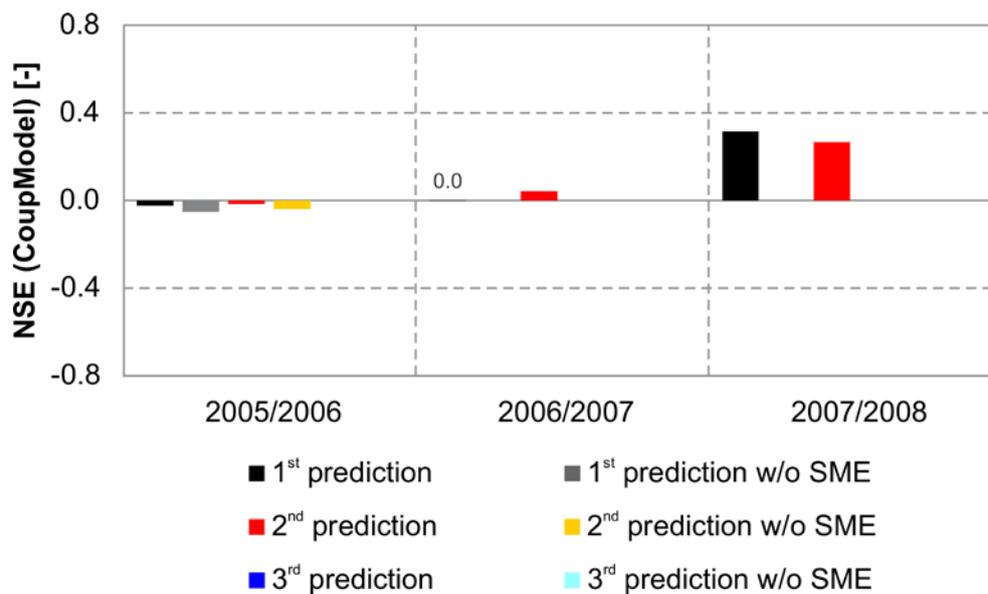


Fig. A-3b: NSE for all prediction of CoupModel

#### A-4 MIKE SHE / Hill-Vi

- *Model user*

S. Stoll (University of Freiburg / ETH Zurich)

Using Hill-Vi for the 1<sup>st</sup> prediction and MIKE SHE for the 2<sup>nd</sup> and 3<sup>rd</sup> prediction

- *Basic model features*

Hill-Vi: explicit grid cell approach, parameterised based on van Genuchten (1980) for unsaturated and on Dupuit-Forchheimer for saturated storage compartments (Weiler and McDonnell, 2004).

MIKE SHE: fully coupled, physically-based groundwater – surface water model

- *Assumptions and parameters (1<sup>st</sup> prediction)*

Initial model focused on the question how the model performs in a region where it has not been used before.

Employing pedotransfer functions (PTF) for assessing soil hydraulic properties (mean  $K_{sat} = 210.6$  mm/h, Schaap et al. (2001)). Snow and frost not considered. Initial conditions generated by warm-up runs.

- *Assumptions and parameters (2<sup>nd</sup> prediction)*

The field visit made the user aware of deep gullies and different vegetation cover in the western and eastern half of the catchment.

The original model could neither adequately describe the actual evapotranspiration (AET), soil crusting, nor the clay dam effects on the groundwater dynamics. Therefore, the

1 modeller switched to using MIKE SHE, redefined the initial conditions, and made small  
 2 changes to reduce AET and PET (potential evapotranspiration). The soil hydraulic  
 3 properties stayed constant.

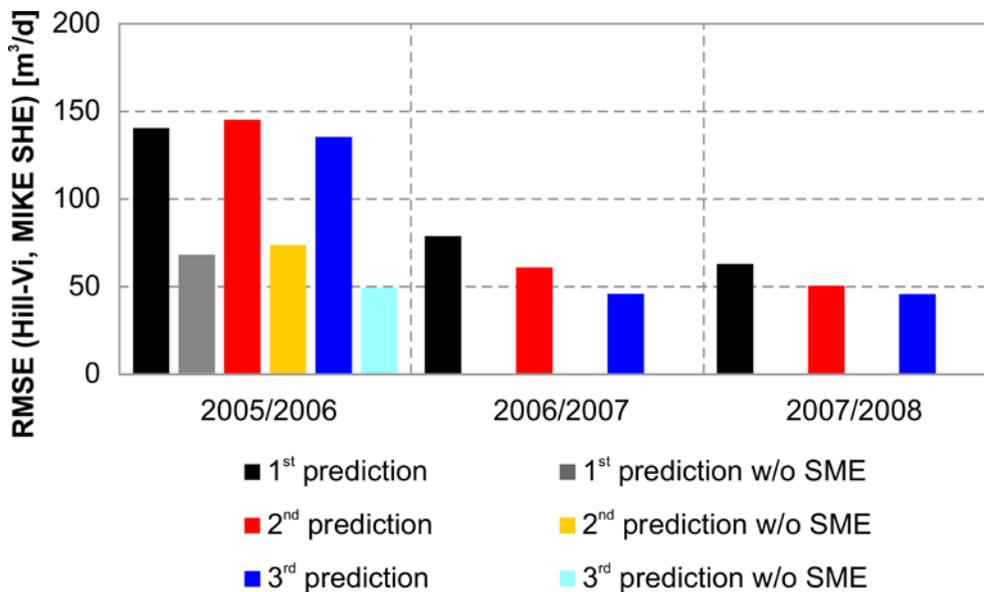
4 • *Assumptions and parameters (3<sup>rd</sup> prediction)*

5 Soil crust newly parameterized using information which was later published by Gerwin et  
 6 al. (2011) and Mazur et al. (2011) and PET adjusted by increasing the vegetation cover.

7 All soil hydraulic data were selected, but none of the others. This resulted in the lowest  
 8 virtual costs of all modellers.

9 • *Results*

10 The results calculated by Hill-Vi (1<sup>st</sup> prediction) suffered from overestimating interflow  
 11 and base flow, which were reduced from about 360 mm/y (1<sup>st</sup> prediction) to 40 mm/y (2<sup>nd</sup>  
 12 prediction) by modifying PET and adding a soil crust. This improved the predictions  
 13 significantly decreasing RMSE (Fig. A-4a) and increasing NSE accordingly (Fig. A-4b).  
 14 Excluding the snow melt event (SME) early in 2006 reduced the errors massively because  
 15 neither Hill-Vi nor MIKE SHE could handle snow melt. This resulted in nearly similar  
 16 RMSE values compared to the other years. The additional data provided for the 3<sup>rd</sup>  
 17 prediction primarily reduced the slow flow components to about 15 mm/y and therefore  
 18 improved the prediction quality. Slow flow dominated the flow regime in this catchment.



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20 Fig. A-4a: RMSE for all prediction of Hill-Vi/MIKE SHE

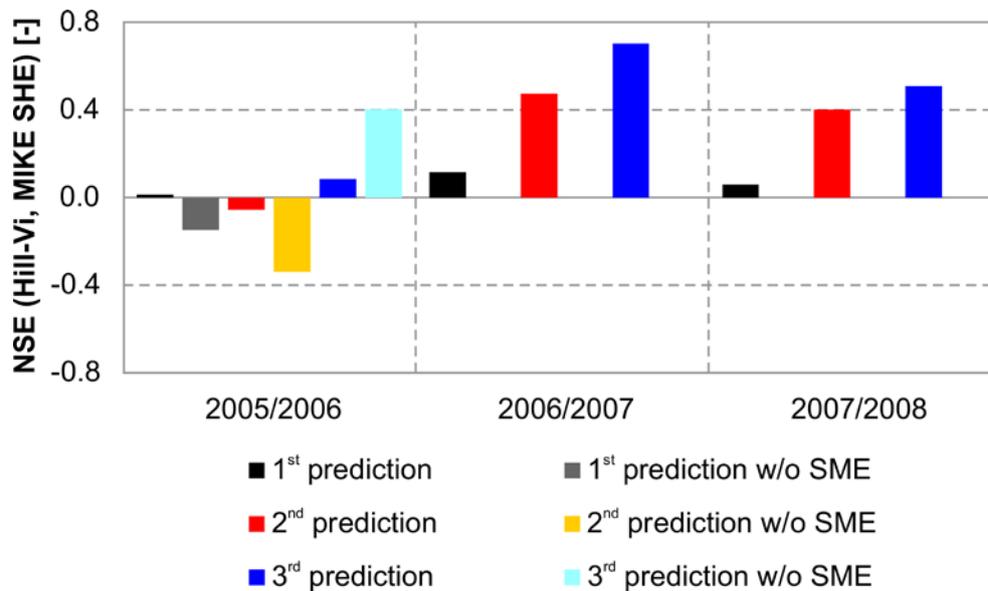


Fig. A-4b: NSE for all prediction of Hill-Vi/MIKE SHE

#### A-5 NetThales

- *Model user*

G.B. Chirico (University of Naples)

- *Basic model features*

Rainfall infiltrates completely into the soil except when the soil column is entirely saturated. Vertical distribution of soil water within the soil column is not simulated. Lateral and subsurface flows are both calculated one-dimensionally (Chirico et al., 2003).

- *Assumptions and parameters (1<sup>st</sup> prediction)*

Modeller tried to match the annual water balance with that of other catchments having a similar rainfall regime.

Unsaturated zone was not simulated (model limitations). The soil properties were assumed uniform along the of a soil column. Hydraulic parameters were derived from texture data using PTF ( $K_{sat} = 50$  mm/h, Rawls and Brakensiek (1985)). Initial soil moisture and groundwater levels were assumed to be uniformly distributed throughout the catchment.

- *Assumptions and parameters (2<sup>nd</sup> prediction)*

The initial state was changed from pre-wetted to quasi-dry.  $K_{sat}$  was reduced to slow down discharge and the influence of the clay dam was increased.

1 • *Assumptions and parameters (3<sup>rd</sup> prediction)*

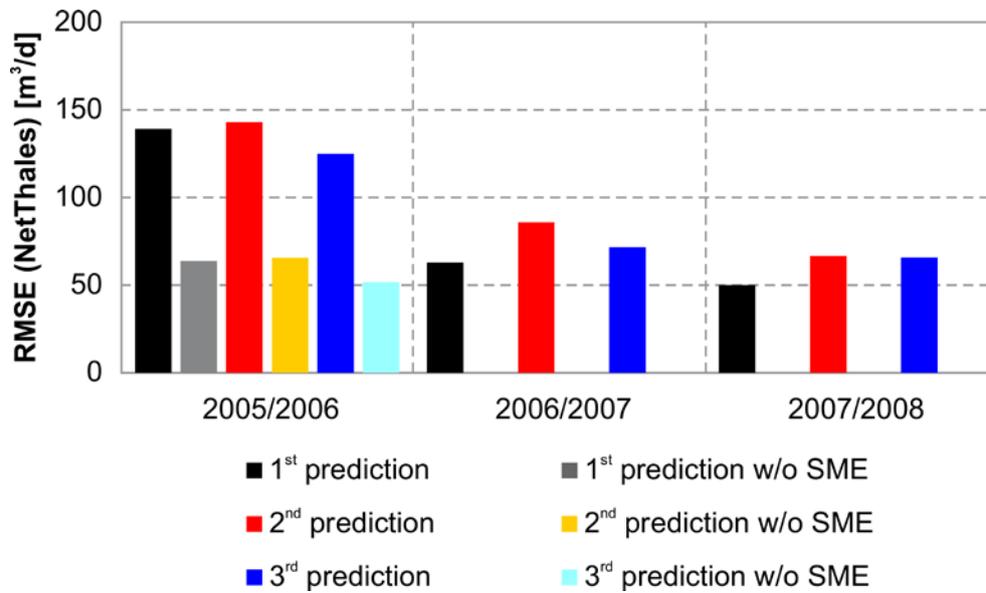
2 Modeller realised that the model is not suitably structured for an efficient description of the  
3 groundwater dynamics, particularly in the initial stage of the emerging groundwater table  
4 and the clay dam required a spatial structure of the subsurface compartment. Changing the  
5 model structure (introducing soil freezing, implementing a soil crust having a hydraulic  
6 conductivity of 3 mm/h, and changing  $K_{\text{sat}}$  to 100 mm/h) caused substantial costs (time)  
7 without a significant reduction of the prediction uncertainty due to the uncertainty in the  
8 model parameterisation based on the apparently still insufficient data set.

9 As a result, the modeller tried to obtain the best fit with the best set of the available data  
10 thereby using all additional soil data including the actual measurements were selected  
11 (average cost compared to other modellers' choices).

12 • *Results*

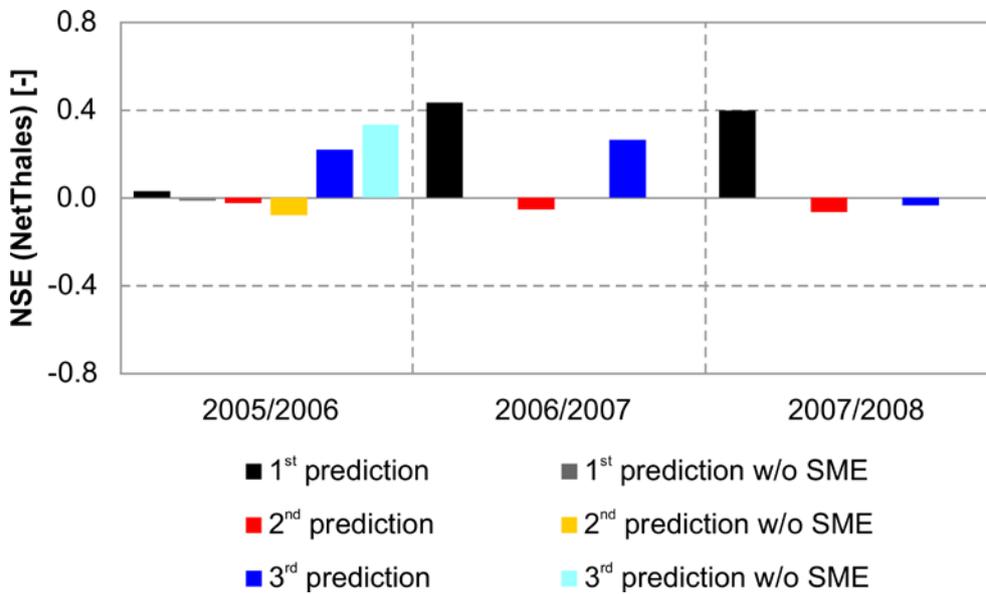
13 During the 1<sup>st</sup> prediction stage, the slow flow components were too prominent because the  
14 model structure allows infiltration until the soil is completely saturated. The high hydraulic  
15 conductivity allowed the groundwater to drain rapidly. The changes in parameterization  
16 resulted in higher peak flow (287 m<sup>3</sup>/d) and little slow flow (e.g. 8% of all days show zero  
17 discharge). However, this did not reduce the RMSE (Fig. A-5a) compared to the 1<sup>st</sup>  
18 prediction state. Similarly, the NSE was lowered significantly so that the NSE for all years  
19 is even below zero for the 2<sup>nd</sup> prediction (Fig. A-5b).

20 Making use of the additional soil data increased discharge more than in all other models  
21 (+88 mm/y) and also produced the largest peak discharge (1481 m<sup>3</sup>/d) exceeding by far the  
22 largest observed value (27<sup>th</sup> May 2007: 897 m<sup>3</sup>/d). Introducing a soil crust reduced the  
23 infiltration. This resulted in slower flow events (<75% of all events had a discharge of less  
24 than 1 m<sup>3</sup>/d). This was the main cause why the RMSE slightly decreased and the NSE  
25 increased between the 2<sup>nd</sup> and 3<sup>rd</sup> prediction stage (Fig. A-5a). Only the NetThales  
26 modeller addressed soil freezing in the 3<sup>rd</sup> prediction, which improved the predictions for  
27 the first hydrological. However, excluding the SME period still results in a better RMSE  
28 (Fig. A-5a).



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Fig. A-5a: RMSE for all prediction of NetThales



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Fig. A-5b: NSE for all prediction of NetThales

## 6 A-6 SIMULAT

### 7 • Model user

8 H. Bormann (University of Oldenburg, now: University of Siegen)

### 9 • Basic model features

10 Physically-based SVAT model (Soil Vegetation Atmosphere Transfer) which solves the  
 11 Richards equation to estimate infiltration and soil-water fluxes, lateral groundwater flow is  
 12 addressed by concentration time, and surface runoff is calculated by a semi-analytical

1 solution of the Richards equation (Diekkrüger and Arning, 1995; Bormann, 2008).  
2 Evapotranspiration is calculated according to the Penman-Monteith approach.

3 • *Assumptions and parameters (1<sup>st</sup> prediction)*

4 The modeller wanted to minimize the influence of the modeller's choice and did not decide  
5 a priori on the dominant process(es) based on his experience from previous studies where  
6 the model produced reliable results without prior calibration (Bormann et al., 1999).

7 The modeller used the pedotransfer function according to Rawls and Brakensiek (1985) to  
8 derive soil hydraulic parameters. A national soil data-base (Adhoc AG Boden, 2005) was  
9 used to estimate bulk density. These assumptions resulted in a mean  $K_{\text{sat}}$  of 60.8 mm/h  
10 while spatially distributed soil parameters were used for the modelling study (25 m grid).  
11 The model started with unsaturated conditions but allowed for partial saturation in case of  
12 storage based lower boundary condition. Additionally, the modeller implemented the  
13 dynamics of the lake at the catchment outlet.

14 • *Assumptions and parameters (2<sup>nd</sup> prediction)*

15 The modeller assumed a shallow soil layer above the dam. After field visit he considered  
16 soil crusts as the major cause for the observed soil erosion. Since the modeller visited the  
17 catchment in spring, he noticed the fast development of the vegetation.

18 He used  $K_{\text{sat}} = 2.1$  mm/h taken from Hölzel et al. (2011). To account for catchment  
19 heterogeneity he used information on spatial variability of  $K_{\text{sat}}$  with a standard deviation of  
20  $\sigma_{K_{\text{sat}}} = 62.5$  mm/h as given by Cosby et al. (1984). The user increased the LAI to  $> 1$  for  
21 the year 2008 at selected grid points.

22 He changed the lower boundary condition of the individual soil columns to a linear storage  
23 based boundary condition a solution possible due to the 1-dimensional nature of this  
24 model. This accounts for the steadily rising groundwater level and it enhanced the  
25 damming effect of the subsurface V-shaped clay dam. Finally, the modeller changed the  
26 volume of the already implemented lake to match the volume at the spillway. The detailed  
27 description of parameterisation of the 2<sup>nd</sup> stage prediction of SIMULAT can be found in  
28 Bormann (2011).

29 • *Assumptions and parameters (3<sup>rd</sup> prediction)*

30 The modeller noticed that the modelled results still varied significantly among the various  
31 models. The simulated water balance was consistently wrong, changes in  
32 evapotranspiration parameterisation did not seem to be appropriate, and the subsurface  
33 storage needed to be better adapted. As a result, the modeller updated the variability in  
34 surface  $K_{\text{sat}}$  and the lower boundary condition of the soil columns once more in order to  
35 better describe the infiltration as well as subsurface storage (Bormann, 2011). Additionally,  
36 the initial condition was re-defined as dry soil.

1 The modeller chose the data based on its usefulness for complementing the model set-up.  
2 This resulted in the 2<sup>nd</sup> highest virtual costs. The additional soil physical data were used to  
3 confirm the magnitude of the soil hydraulic parameters in the preceding simulations. The  
4 data from the infiltration experiments were used to improve the description of the  
5 hydraulic properties of the surface layer. The field data were complemented by literature  
6 values to parameterise the spatial variability of  $K_{sat}$  (Cosby et al., 1984). Finally, the  
7 modeller used soil moisture in two steps. First, he evaluated his model with these data and,  
8 in a 2<sup>nd</sup> step, he used them to calibrate the model by adjusting the lower boundary  
9 conditions of the soil columns (Bormann, 2011). This resulted in a hydraulic conductivity  
10 of the soil crust of 11.6 mm/h. Although the modeller opted for using the vegetation data,  
11 he did not use them after reviewing these data.

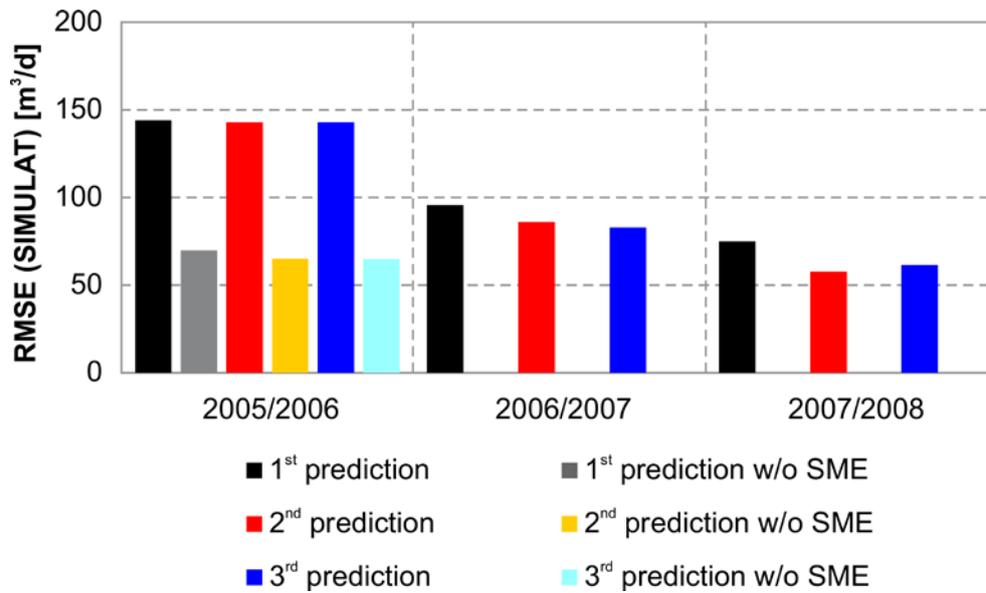
12 • *Results*

13 The 1<sup>st</sup> predictions suffered by too much discharge, twice the observed value. More or less  
14 all discharge was considered to be base flow.

15 The base flow was then effectively reduced to about 20% of the total discharge, which was  
16 the lowest of all models (157 mm/y) in the 2<sup>nd</sup> prediction. Since interflow was negligible,  
17 about 80% was primarily surface runoff. Although SIMULAT predicted zero discharge for  
18 many days, it rose rapidly and produced the largest peak discharge of all models (1433  
19 m<sup>3</sup>/d in the 2<sup>nd</sup> year). SIMULAT predicted primarily surface runoff (~ 80%) and negligible  
20 interflow. However, like all other predictions with peak discharge of >400 m<sup>3</sup>/d, the  
21 discharge between  $Q_{100}$  and  $Q_{95}$  went down by nearly two orders of magnitude. Due to  
22 reduction of the vegetation cover AET was lowered considerably and was the lowest of all  
23 models (157, 266 and 260 mm/y for the 1<sup>st</sup>, 2<sup>nd</sup>, and 3<sup>rd</sup> year, respectively).

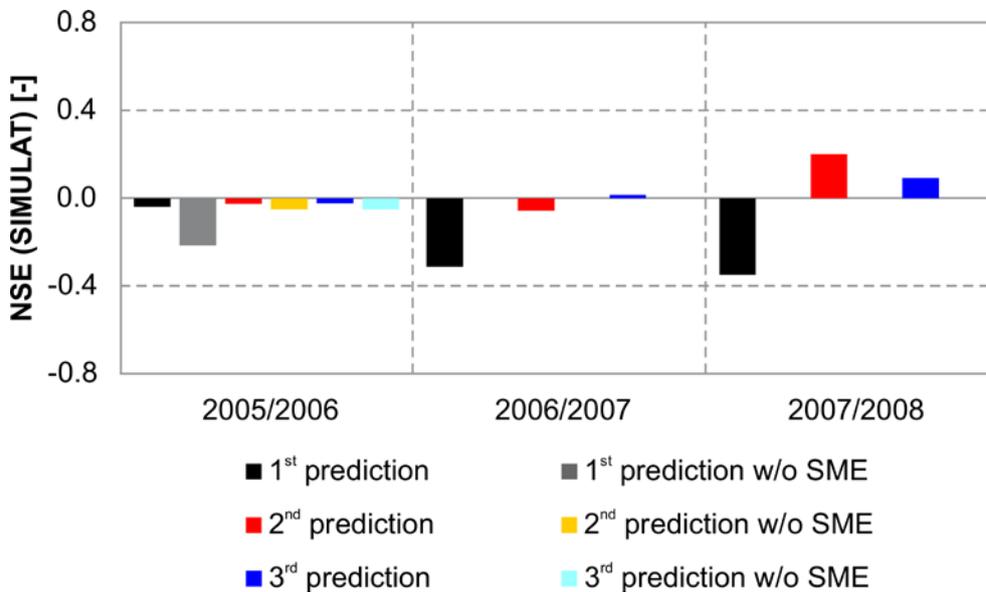
24 PET simulated for 2006/2007 did not change from the 2<sup>nd</sup> to the 3<sup>rd</sup> prediction.  
25 SIMULAT predicted in two of the three years the largest discharge (e.g. 291 mm/y in  
26 2006/2007) and an average increase of discharge by 21 mm/y. This was a consequence of  
27 using the soil data. SIMULAT generated little surface runoff (22%) and mainly subsurface  
28 runoff and therefore the opposite of what it simulated in the 2<sup>nd</sup> prediction. This resulted in  
29 the lowest peak discharge (106 m<sup>3</sup>/d) of all models (Supplementary Material (C)). This is a  
30 reduction by >90% compared to the 2<sup>nd</sup> prediction

31 The prediction quality for the 1<sup>st</sup> year did not improve but for the two following years it did  
32 (Fig. A-6a). The prediction quality from the 2<sup>nd</sup> to 3<sup>rd</sup> prediction shows no significant  
33 change. Similar results are also shown by NSE (Fig. A-6b).



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Fig. A-6a: RMSE for all prediction of SIMULAT



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Fig. A-6b: NSE for all prediction of SIMULAT

## 6 A-7 SWAT

### 7 • Model user

8 J.-F. Exbrayat (University of Giessen, now: University of Edinburgh)

### 9 • Basic model features

10 Physically-based semi-distributed model which divides each sub-catchment into  
 11 Hydrological Response Units. Lateral flow is calculated by the drainage equation,  
 12 infiltration d by SCS (Soil Conservation Service) curve number method, and the soil-water  
 13 fluxes are integrated as a bucket model depending on the soil-water content and other soil

1 properties (Arnold et al., 1998). SWAT simulates plant growth and its effects on the water  
2 balance based on the EPIC model.

3 • *Assumptions and parameters (1<sup>st</sup> prediction)*

4 Assumptions made in SWAT make it more adapted to simulate mesoscale catchments.  
5 Although it cannot be considered a perfect pick for small catchments, applications of  
6 SWAT range from hill slope case studies to large basins such as the Mississippi River. The  
7 modeller therefore intended to further test its performance for small catchments. He  
8 minimized the influence of the modeller's decision and used default values available from  
9 the SWAT user manual. The modeller employed PTF (mean  $K_{\text{sat}} = 74.5$  mm/h, Rawls and  
10 Brakensiek (1985)) for assessing soil hydraulic properties with a cluster analyses to check  
11 the soil homogeneity assumption. The dynamics of the lake at the catchment outlet was  
12 implemented with a total volume estimated from the DEM.

13 • *Assumptions and parameters (2<sup>nd</sup> prediction)*

14 The user became aware of the deep gullies during the on-site inspection of the catchment.  
15 Soil crusting is not included into the model.  $K_{\text{sat}}$  stayed constant but he tried to account for  
16 a larger surface runoff by allowing infiltration from gullies. The modeller removed the  
17 warm-up period completely so that the initial state was changed from pre-wetted to quasi-  
18 dry conditions. Finally, the modeller changed the volume of the already implemented lake  
19 to the actual value provided during the first workshop.

20 • *Assumptions and parameters (3<sup>rd</sup> prediction)*

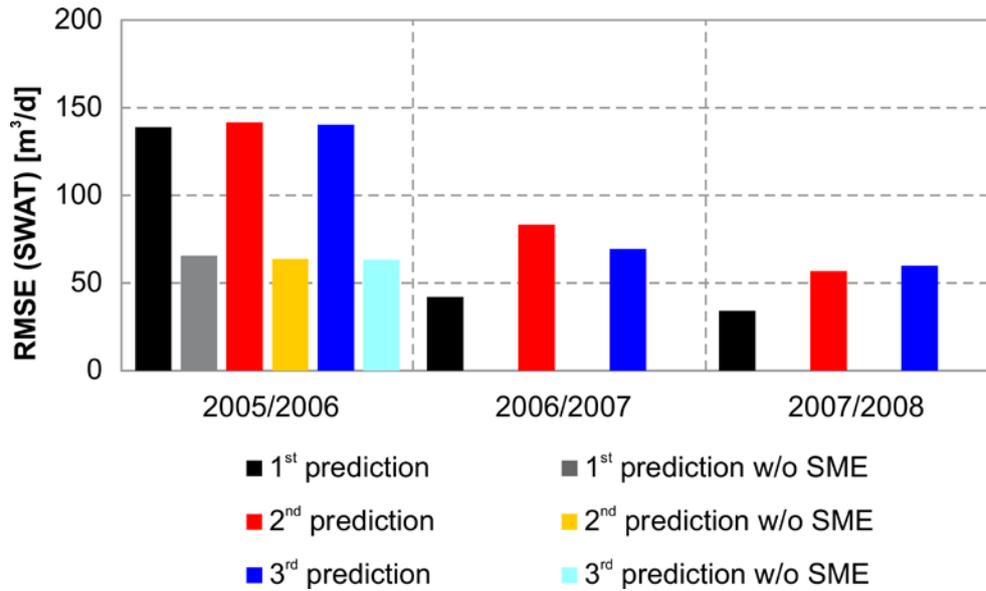
21 The modeller kept the model setup as used for the 2<sup>nd</sup> prediction, but corrected some  
22 parameter values based on the provided dataset that may affect both physical and plant  
23 processes (e.g. the organic carbon content). The modeller selected all soil hydraulic  
24 parameters except the water retention curves. By choosing the weather and the vegetation  
25 date, the modeller came up with the highest virtual costs (32,540 Euro). The vegetation  
26 data were then used to parameterize the prevalent plant species (*Trifolium arvense*) in  
27 terms of max. LAI, rooting depth, and stomatal conductance based on literature values  
28 from the PlaPaDa database (Breuer et al., 2003).

29 • *Results*

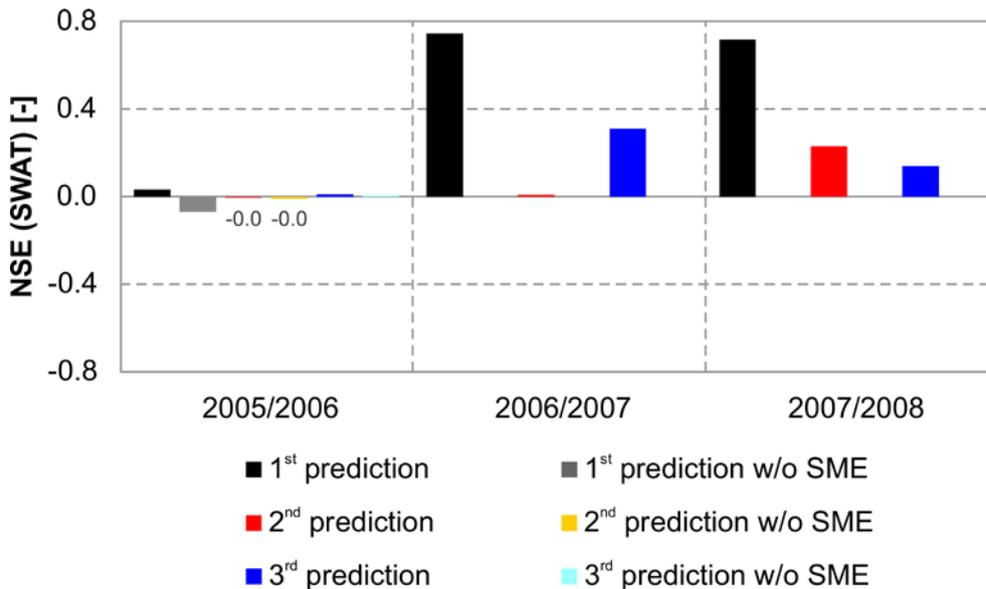
30 The predictions by SWAT showed too large peak discharge and, as in case of the other  
31 models, the snow melt problem. However, especially in the 2<sup>nd</sup> and 3<sup>rd</sup> year the results  
32 were the best of all models as shown by RMSE and NSE (Fig. A-7a and A-7b) despite the  
33 fact that assumptions made in the structure of SWAT can be judged inadequate to represent  
34 small catchments.

35 Although SWAT changed the initial conditions to quasi-dry conditions in the 2<sup>nd</sup> prediction  
36 stage, the storage changes were negative in the 2005/2006 (-24 mm/y). Later they became

1 positive. The peak discharge was strongly reduced (from  $\sim 900 \text{ m}^3/\text{d}$  in the 1<sup>st</sup> prediction to  
 2  $< 100 \text{ m}^3/\text{d}$  in the 2<sup>nd</sup>) but the slow flow components strongly increased, e.g.  $Q_{25} = 9 \text{ m}^3/\text{d}$ .  
 3 This resulted in a strong decrease of RMSE and increase of NSE (Fig. A-7a and A-7b).  
 4 SWAT had similar discharge components comparing the 3<sup>rd</sup> and 2<sup>nd</sup> prediction, producing  
 5 about 50% of surface runoff and base flow. The additional weather data only slightly  
 6 affected PET (+30 mm/y to 847 mm/y) but increased by +33 mm/y in 2005/2005 and +118  
 7 mm/y in 2006/2007. This decreased discharge in the average by -101 mm/y, the second  
 8 lowest peak discharge, with  $Q_{25} < 1 \text{ m}^3/\text{d}$ ). This reduced the prediction uncertainties for the  
 9 first two years slightly but increased them for 2007/2008 (Fig. A-7a and A-7b).



10  
 11 • Fig. A-7a: RMSE for all prediction of SWAT



12  
 13 • Fig. A-7b: NSE for all prediction of SWAT

## 1 **A-8 Topmodel**

### 2 • *Model user*

3 Wouter Buytaert (University of Bristol, now: Imperial College London)

### 4 • *Basic model features*

5 A semi-distributed hydrological model that assigns a combination of storage compartments  
6 such as the root zone, unsaturated and saturated zone. Infiltration is determined by the  
7 Green-Ampt equation and a time delay function controls flow within a vertical soil  
8 column. The lateral subsurface flow is estimated by an exponential transmissivity function.  
9 The model does not consider the influence of soil freezing on  $K_{\text{sat}}$  nor snow (Beven et al.,  
10 1995).

### 11 • *Assumptions and parameters (1<sup>st</sup> prediction)*

12 The modeller wanted to minimize the influence of the modeller's choice and used  
13 published PTFs for parameterization ( $K_{\text{sat}} = 58 \text{ mm/h}$ , Saxton et al. (1986)).

### 14 • *Assumptions and parameters (2<sup>nd</sup> prediction)*

15 The modeller visited the catchment only in a spring period. The rapid appearance of  
16 vegetation would be simulated, but the soil crust could not be introduced into the model.  
17 This modeller did not include the clay dam for the 1<sup>st</sup> prediction. He added it in the 2<sup>nd</sup>  
18 stage. The subsurface response was slightly increased by increasing the areal average of  
19 the local transmissivities ( $\text{m}^2/\text{h}$ ) at saturation from  $\ln Te = -2.5$  to  $-2$ . This corresponds with  
20  $Te = 0.082$  to  $0.135 \text{ m}^2/\text{h}$ . The clay dam delays the subsurface response and the modeller  
21 tried to mimic this by changing  $\ln Te$ . This is a merely intuitive approach since there was  
22 no guidance on how to adapt the parameter. Additionally, a surface runoff of 10% of the  
23 precipitation were generated by changing  $K_{\text{sat}}$  at the surface ( $5 \text{ mm/h}$ ) and the capillary  
24 drive  $CD$  ( $1 \text{ mm}$ ) (Morel-Seytoux and Khanji, 1974).

### 25 • *Assumptions and parameters (3<sup>rd</sup> prediction)*

26 The modeller selected only the information that appeared to be most important since the  
27 model is limited by its conceptual nature. This made it difficult to integrate additional data.  
28 Therefore, the modeller chose the soil hydraulic parameter without the water retention  
29 curves and the soil moisture data set.

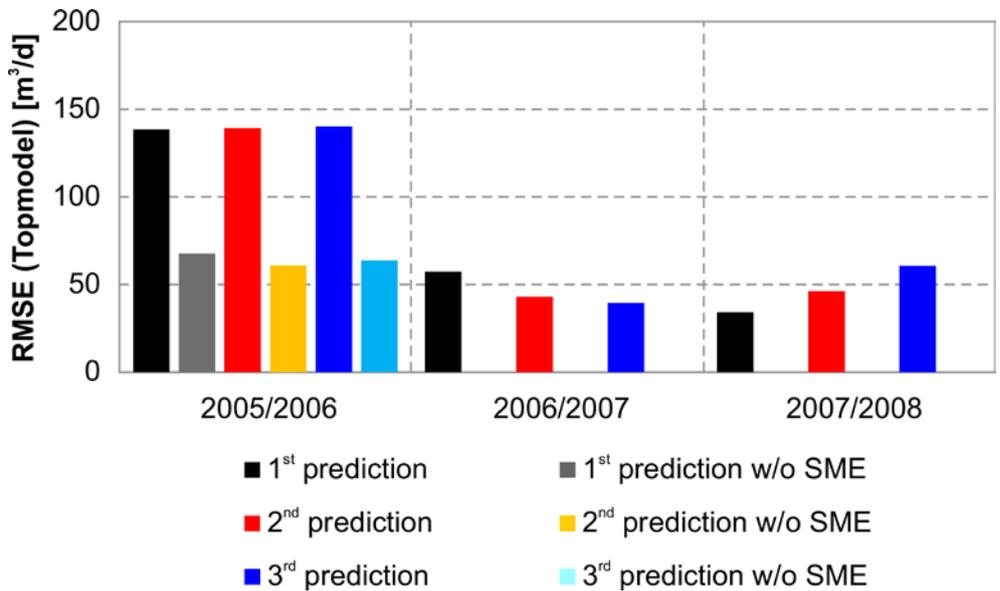
30 The modeller used these data for calibration: all TDR measured at 10 cm depth and the  
31 groundwater levels of observation well L4 were averaged and used as proxies for the  
32 storage deficit. He expected that these two observations are inversely related. He then  
33 performed a Monte Carlo sensitivity test based on the correlation coefficient as  
34 performance measure. The modeller deduced from this that only the amount of water  
35 (expressed as a depth), which the soil can hold within the root zone ( $S_{r_{\text{max}}}$ ) is a sensitive

1 parameter. Finally,  $Sr_{max}$  was chosen to 0.02 m. Subsequently, the initial root zone storage  
 2 deficit ( $Sr_0$ ) and the initial subsurface flow per unit area ( $q_{s0}$ ) were updated to be  
 3 compatible with  $Sr_{max}$ .

4 • *Results*

5 Topmodel showed one of the best predictions during the 1<sup>st</sup> prediction stage with a RMSE  
 6 about 50 m<sup>3</sup>/d (Fig. A-8a). The main problem were the large discharges, e.g. the 2<sup>nd</sup> largest  
 7 peak discharge (777 m<sup>3</sup>/d). One reason was the rather low PET of ~570 mm/y.

8 The changes between the 1<sup>st</sup> and 2<sup>nd</sup> prediction stage had nearly no impact in 2005/2006, a  
 9 positive one in 2006/2007 but a negative one in 2007/2008 expressed as RMSR and NSE  
 10 (Fig. A-8a and A-8b). Modifying the vegetation values strongly changed PET (e.g. 1014  
 11 mm/y in 2005/2006). It became the largest in the 2<sup>nd</sup> prediction stage. This increased also  
 12 AET (e.g. 465 mm/y in 2006/2007). Topmodel predicted one of the largest peak discharges  
 13 (958 m<sup>3</sup>/d) with a strong decrease in discharge between  $Q_{100}$  and  $Q_{95}$  of nearly two orders  
 14 of magnitude. About an equal amount of discharge was due to surface runoff and due to  
 15 cumulated interflow and base flow. These relations did not change in the 3<sup>rd</sup> prediction  
 16 stage. Also the prediction quality did not change significantly. The quality measures show  
 17 no changes for the 1<sup>st</sup> year, slight improvement in the 2<sup>nd</sup>, and a slight quality decrease in  
 18 the 3<sup>rd</sup> year (Fig. A-8a and A-8b). These changes in the prediction quality resulted from an  
 19 average increase in discharge of 48 mm/y. The major problem was still the too “fast“ slow  
 20 flow components so that  $Q_{25}$  was 8 m<sup>3</sup>/d while less than 1 m<sup>3</sup>/d was observed.



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Fig. A-8a: RMSE for all prediction of Topmodel

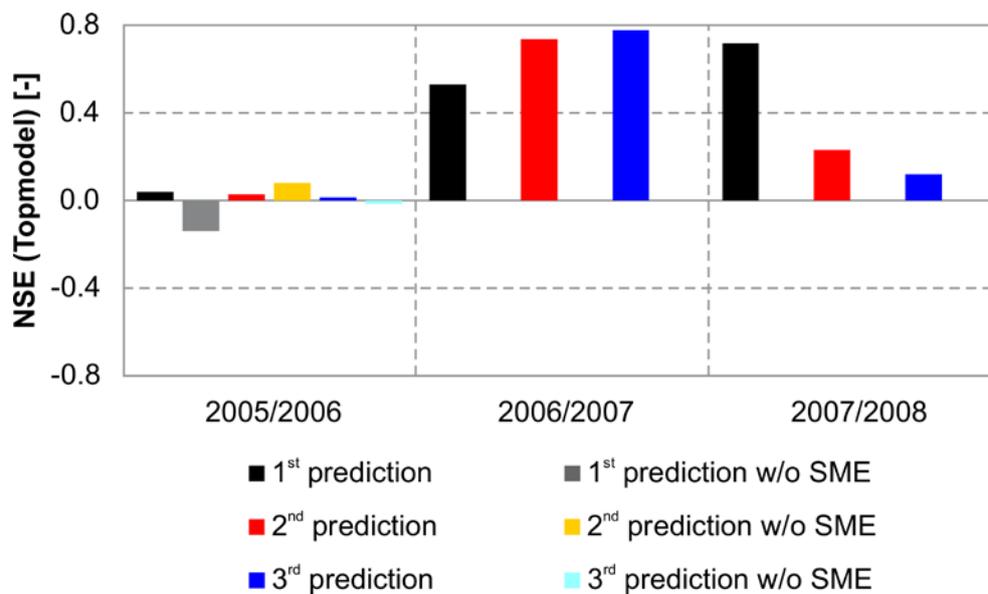


Fig. A-8b: NSE for all prediction of Topmodel

#### A-9 WaSiM-ETH (Richards)

- *Model user*

H. Hölzel (University of Bonn, now: Vattenfall Europe Mining AG)

- *Basic model features*

Spatially distributed hydrological model. All algorithms except the saturated soil zone routine are physically-based. Infiltration is calculated with the equations by Green-Ampt and the unsaturated zone by that of Richards. The lateral flow is determined by a linear storage approach (Schulla and Jasper, 2007).

- *Assumptions and parameters (1<sup>st</sup> prediction)*

The modeller relied on the description of the physical processes in his model and thereby minimized the influence of his own decisions. He derived soil hydraulic parameters from the texture data ( $K_{\text{sat}} = 118 \text{ mm/h}$ , Adhoc AG Boden (2005)). The model was initialized by several warm-up runs to achieve steady state conditions. The clay dam was not considered.

- *Assumptions and parameters (2<sup>nd</sup> prediction)*

The on-site inspection of the catchment showed the extent of erosion due to surface runoff and the modeller noticed the different vegetation development in the western and eastern half of the catchment. Therefore, the model was split into a vegetation-free (before 2008) and a vegetation period with a low density grass canopy. To account for the surface runoff,  $K_{\text{sat}}$  was reduced and a soil crust having with a hydraulic conductivity of 20 mm/h based on NAW (2008) was introduced.

1 The initial state was changed from pre-wetted to quasi-dry and the lake was included. The  
2 formerly constant layer thickness was defined according to the Digital Elevation Model  
3 (DEM) in order to account for the clay dam. The detailed description of parameterisation  
4 of the 2<sup>nd</sup> stage prediction of WaSiM-ETH (Richards) can be found in Hölzel et al. (2011).

5 • *Assumptions and parameters (3<sup>rd</sup> prediction)*

6 The modeller emphasized the relevance of the groundwater dynamic by a more realistic  
7 representation of the clay dam (Holländer et al., 2009). He replaced the conceptual 1-D by  
8 a process-based 2-D groundwater approach (Hölzel et al., 2013).

9 The modeller still relied on the description of the physical processes and therefore, he  
10 requested data which allowed improving the model such as  $K_{\text{sat}}$  derived from slug tests in  
11 the field and  $K_{\text{sat}}$  determined in the laboratory, water retention, and soil moisture data. He  
12 used these data to have control on the soil moisture dynamic. Additionally, the data of  
13 weather station II were selected.

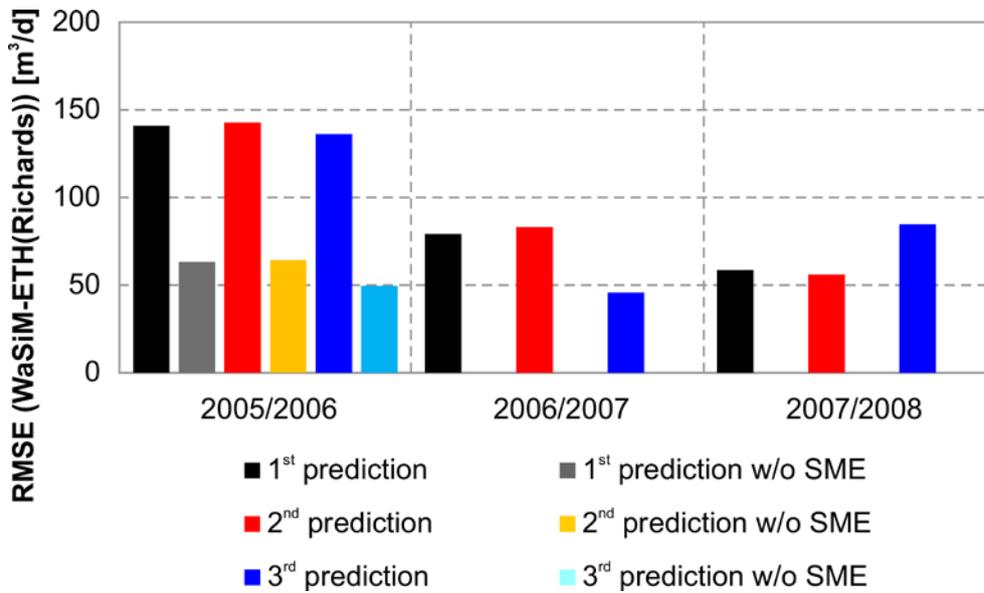
14 • *Results*

15 In the first prediction WaSiM-ETH (Richards) overestimated discharge discharge mostly  
16 predicting  $>10 \text{ m}^3/\text{d}$  but peak discharge was rather low ( $140 \text{ m}^3/\text{d}$ ).

17 The soil crust with a  $K_{\text{sat}}$  of 200 mm/h did not effectively reduce the infiltration into the  
18 soil. This lowered AET significantly (234, 273, and 285 mm/y) and discharge increased  
19 (153, 309, and 306 mm/y), the largest of all models. Therefore, peak discharge increased to  
20  $1028 \text{ m}^3/\text{d}$ , the second largest of all models. Also the base flow was larger ( $15 \text{ m}^3/\text{d}$ ) than  
21 in the 1<sup>st</sup> prediction. Surface runoff and interflow were obtained each at 20% and 60% base  
22 flow. The modeller identified the 1-D groundwater approach used in the 2<sup>nd</sup> prediction to  
23 be a major problem because it prevented the implementation of the clay dam. Including it  
24 would have reduced the base flow. The prediction quality did not significantly change from  
25 the 1<sup>st</sup> to 2<sup>nd</sup> prediction (Fig. A-9a and A-9b).

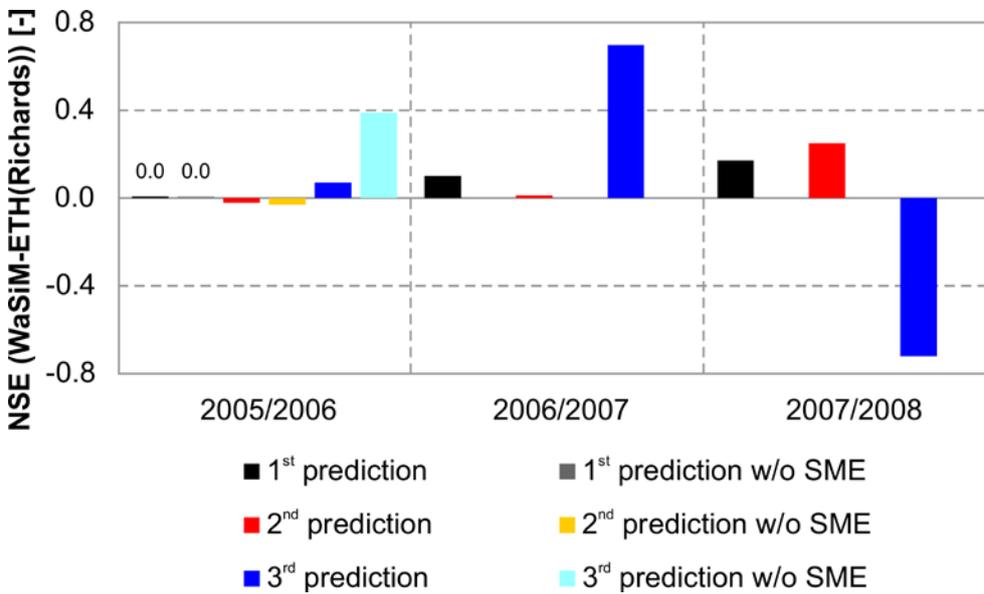
26 Using the weather data from the new station for the 3<sup>rd</sup> prediction reduced PET by about 30  
27 mm/y to 682 mm/y and discharge by -65 mm/y. This affected both, the peak flow ( $1210$   
28  $\text{m}^3/\text{d}$ ) and the base flow ( $10 \text{ m}^3/\text{d}$ ). However, the predicted discharge was still exceeded the  
29 observed values. The reduction in discharge improved the prediction quality for the first  
30 two years but got worse in the third year (Fig. A-9a and A-9b) despite the apparent benefit  
31 of additional and more specific data.

32 The contribution of the discharge components to total discharge changed significantly over  
33 the period of all three years: in the 1<sup>st</sup> year surface runoff, interflow, and base flow were  
34 similar. Surface runoff dominated in the 2<sup>nd</sup> year ( $\sim 45\%$ ) with base flow the lowest ( $\sim$   
35  $20\%$ ) and interflow dominated in the 3<sup>rd</sup> year ( $\sim 45\%$ ). In the 3<sup>rd</sup> year the surface runoff  
36 component was the lowest ( $\sim 25\%$ ).



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Fig. A-9a: RMSE for all prediction of WaSiM-ETH (Richards)



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Fig. A-9b: NSE for all prediction of WaSiM-ETH (Richards)

## 6 A-10 WaSiM-ETH (Topmodel)

- 7 • *Model user*

8 Thomas Krauß (University of Technology Dresden)

- 9 • *Basic model features*

10 A spatially distributed hydrological model which uses the Topmodel attempt (refer to A-8)  
11 (Schulla and Jasper, 2007).

- 12 • *Assumptions and parameters (1<sup>st</sup> prediction)*

1 The WaSiM-ETH (Topmodel) modeller joined the modelling group only in the 2<sup>nd</sup>  
2 prediction stage.

3 • *Assumptions and parameters (2<sup>nd</sup> prediction)*

4 The modeller attended the 1<sup>st</sup> workshop as an observer and subsequently joined the project.  
5 Due to the large variation in the reported water budget and runoff, the modeller concluded  
6 from the discussions that a completely physically based model for the unsaturated zone is  
7 not needed in this case. During the on-site inspection of the catchment the modeller  
8 became aware of the deep gullies, noticed the different vegetation development in the  
9 western and eastern half of the catchment, and the rapid changes of the soil surface  
10 apparently depending on soil texture.

11 The modeller used the initial conditions generated by a one-year pre-run using the 1<sup>st</sup> year  
12 data despite the fact that the catchment was initially dry. Soil hydraulic parameters were  
13 estimated by PTFs according to Wösten and Nemes (2004) and Saxton et al. (1986) and  
14 used a set up based on the DEM data.

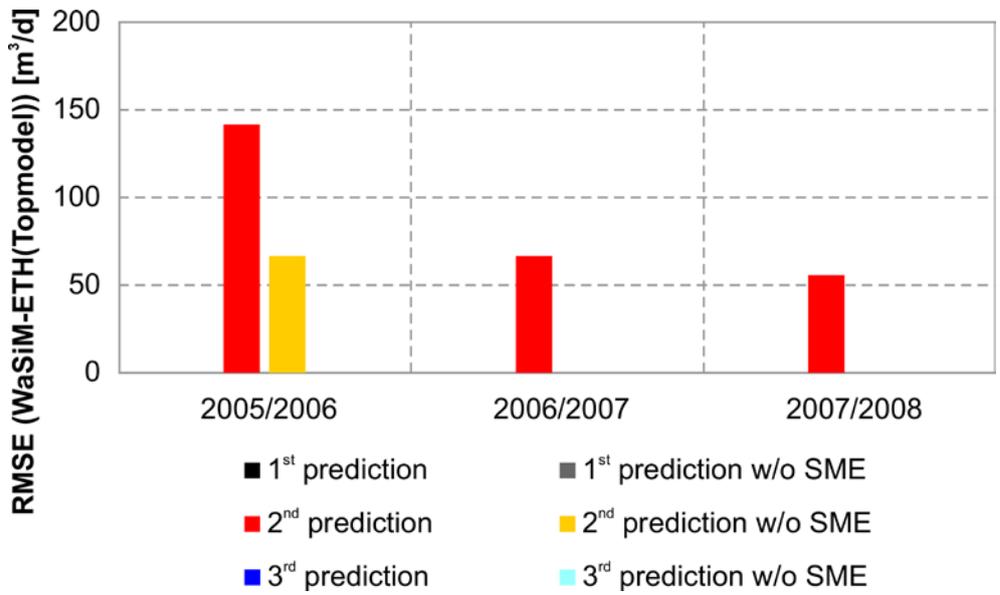
15 • *Assumptions and parameters (3<sup>rd</sup> prediction)*

16 After 2<sup>nd</sup> stage prediction the WaSiM-ETH (Topmodel) modeller left the modelling group  
17 due to time constraints.

18 • *Results*

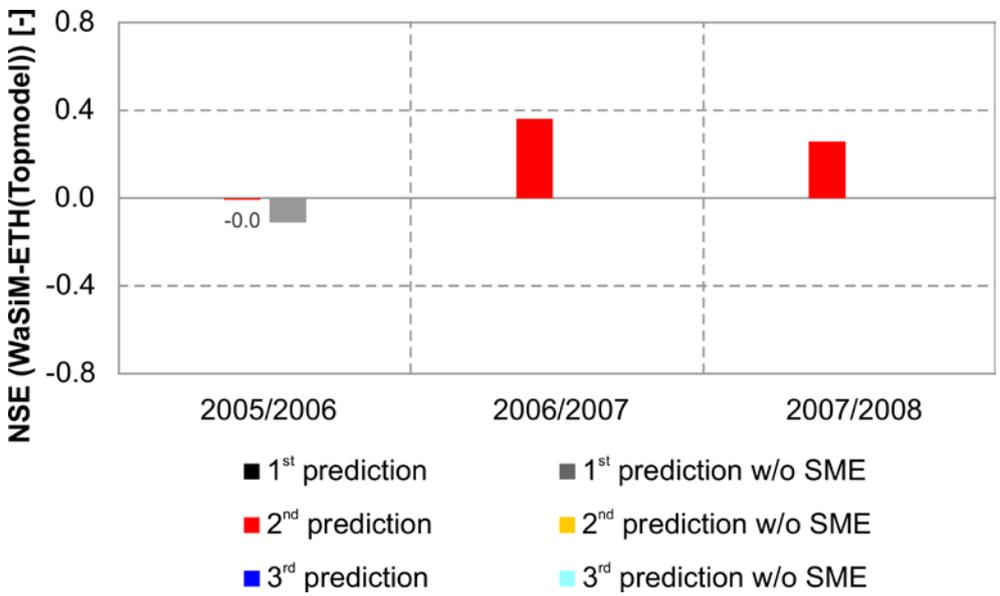
19 The non-dry initial conditions resulted in a higher AET in the 1<sup>st</sup> year (364 mm/y)  
20 compared to any other model during the 2<sup>nd</sup> prediction stage and increased to 459 mm/y in  
21 the 2<sup>nd</sup> year which was nearly 60% of PET. Discharge was relatively constant and varied  
22 between 170 and 200 mm/y during the three years. The error in the mass balance was 17%  
23 in 2005/2006 and about 5% in the other years.

24 The largest base flow was predicted by WaSiM-ETH (Richards) at least 15 m<sup>3</sup>/d. The peak  
25 discharge was 701 m<sup>3</sup>/d. Therefore, the modeller obtained at 60% base flow, 21% surface  
26 runoff, and 19% interflow.



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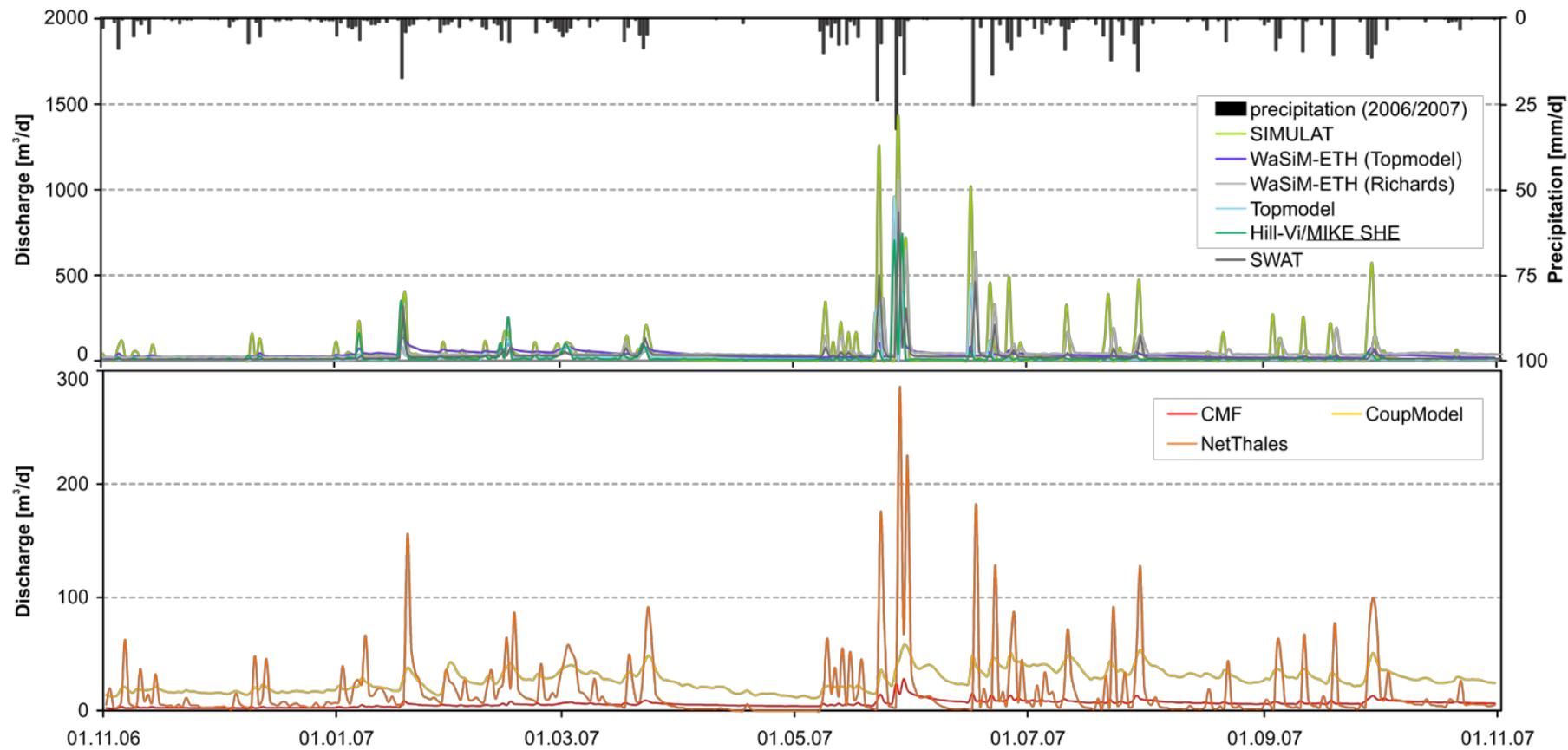
Fig. A-10a: RMSE for all prediction of WaSiM-ETH (Topmodel)



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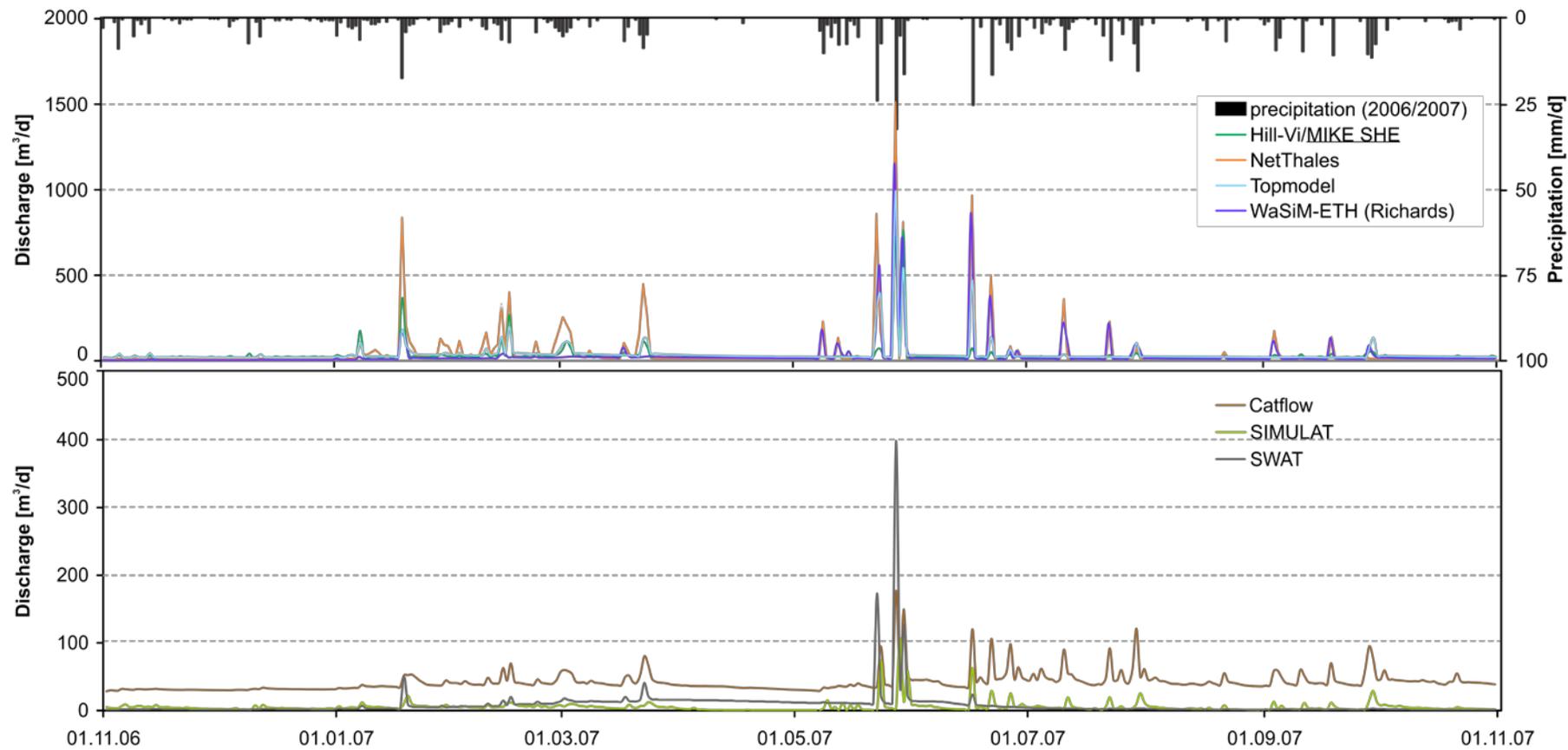
Fig. A-10b: NSE for all prediction of WaSiM-ETH (Topmodel)

1 Supplementary material (B). Discharge predicted for the hydrological year 2006/2007 (2<sup>nd</sup> prediction).



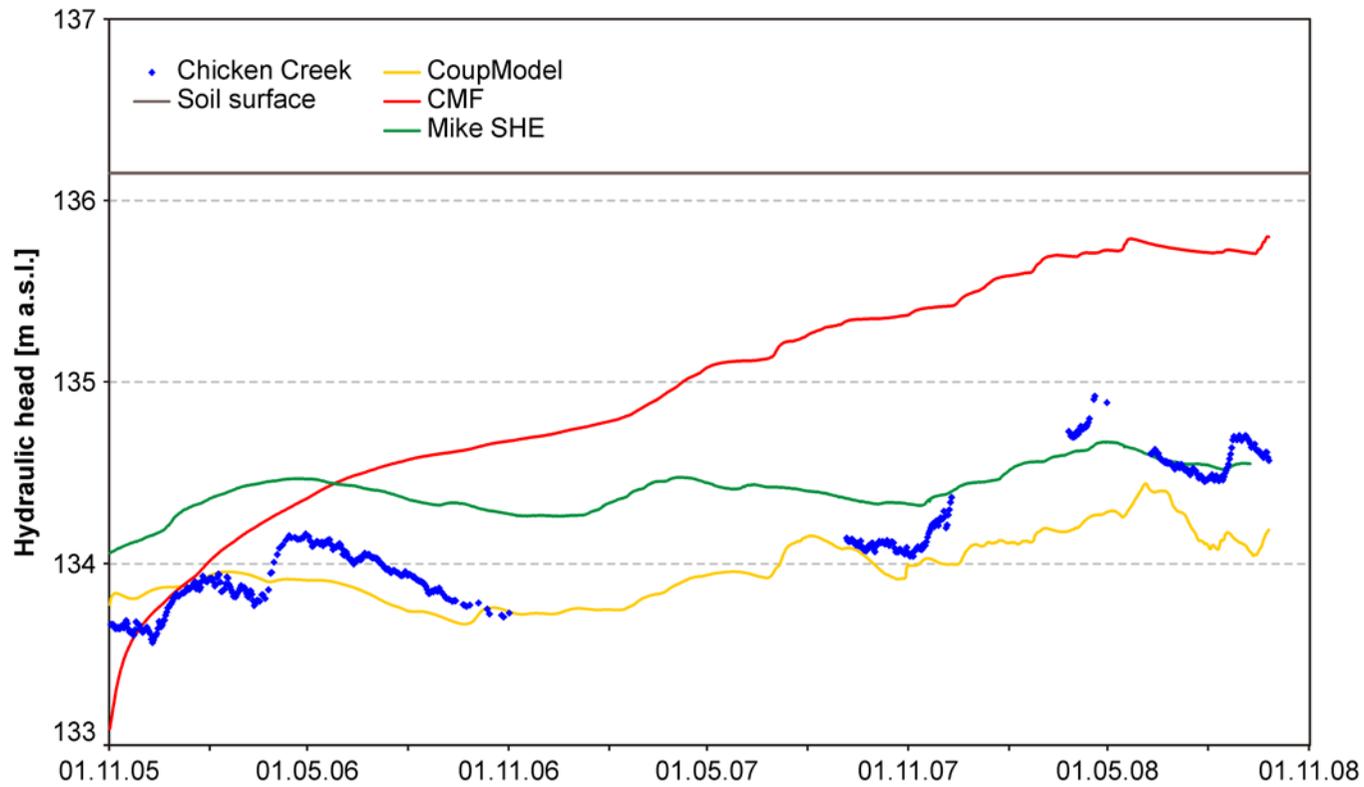
2  
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1 Supplementary material (C). Discharge predicted for the hydrological year 2006/2007 (3<sup>rd</sup> prediction).



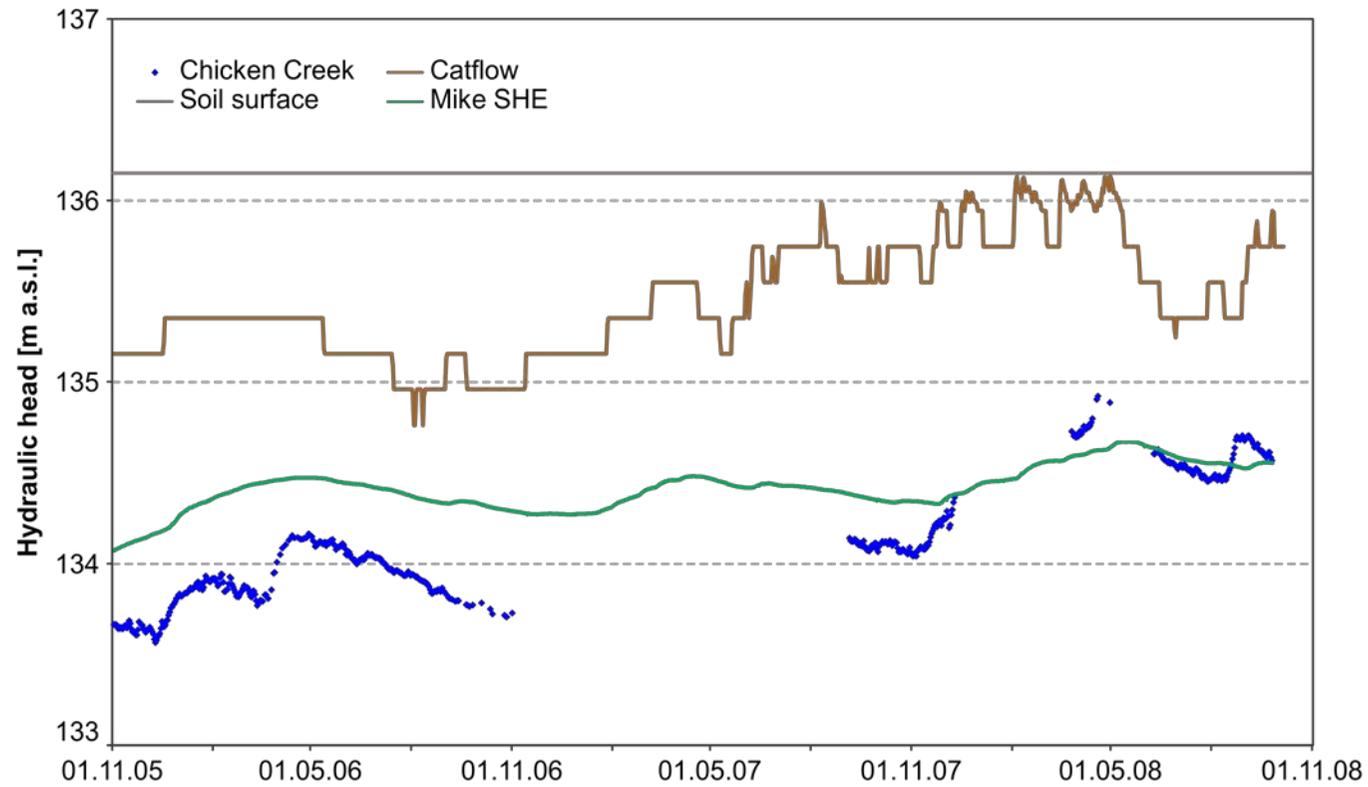
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1 Supplementary material (D). Hydraulic heads (2<sup>nd</sup> prediction) measured and predicted at the observation wells F4 (Soil surface elevation given  
2 as reference).



3  
4

1 Supplementary material (E). Hydraulic heads (3<sup>rd</sup> prediction) measured and predicted at the observation wells F4 (Soil surface elevation given  
2 as reference).



3

4

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