

# 1 **Assessment of SWAT spatial and temporal transferability for** 2 **high altitude glacierised catchments**

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9 **Abstract.** In this study, we investigated the application and the transferability of the Soil Water and Assessment  
10 Tool (SWAT) in a partly glacierised alpine catchment, characterised by extreme climatic conditions and steep  
11 terrain. The model was initially calibrated for the 10 km<sup>2</sup> watershed of the Damma glacier Critical Zone  
12 Observatory (CZO) in central Switzerland using monitoring data for the period of 2009–2011 and then was  
13 validated with the measurements collected during 2012–2013 in the same area. Model performance was found to  
14 be satisfactory against both the Nash Sutcliffe criterion (NS) and a benchmark efficiency (BE). The transferability  
15 of the model was assessed by using the parameters calibrated on the small watershed and applying the model to  
16 the approximately 100 km<sup>2</sup> catchment that drains into the hydropower reservoir of the Göschenalpsee and  
17 includes the Damma glacier CZO. Model results were compared to the reservoir inflow data from 1997 to 2010  
18 and it was found that the model predicted successfully snowmelt timing and autumn recession but could not  
19 accurately capture the peak flow for certain years. Runoff was slightly overestimated from late May to June, when  
20 it is dominated by snowmelt, due to the fact that only one melt factor for both snowmelt and glacier melt was used.  
21 Finally, we investigated the response of the greater catchment to climate change using three different climate  
22 change scenarios and the results were compared to those of a previous study, where two different hydrological  
23 models, PREVAH and ALPINE 3D, were used. Predicted changes in future runoff and peak flow as well as  
24 seasonal dynamics are similar between the two studies. It is concluded that the methodology presented here, where  
25 SWAT is calibrated for a small watershed and then applied for a bigger area with similar climatic conditions and  
26 geographical characteristics, could work even under extreme conditions like ours. However, a greater attention  
27 should be given to the differences between glacial melt and snowmelt dynamics, since our findings indicate that  
28 the performance of the model as well as its transferability could be improved if different parameters for snowmelt  
29 and glacial melt were applied.

## 30 **1 Introduction**

31 The use of calibrated watershed models enables researchers and stakeholders to assess the impact of natural and  
32 management induced environmental changes and, as many studies have pointed out, is of high importance in water  
33 management (i.e. Arnold et al., 1998; Abbaspour et al., 2007). Climate change simulations provide crucial  
34 information for the assessment of its impact on water resources, water quality, and aquatic ecosystems (Farinotti  
35 et al., 2012; Aili et al., 2019). However, watershed modelling in high altitude alpine areas is rather challenging  
36 due to the rough terrain, heterogeneous land cover, extreme climatic conditions and glacier dynamics (Viviroli and  
37 Weingartner, 2004; Farinotti et al., 2012; Rahman et al., 2013), with the main challenge to be the lack of observed

38 and sufficient quality data in ungauged watersheds (Sivapalan et al., 2003; Viviroli et al., 2009b; Bocchiola et al.,  
39 2011).

40  
41 Modelling and predicting the runoff of ungauged watersheds is one of the big challenges that hydrologists face  
42 today (Sivapalan et al., 2003; Hrachowitz et al., 2013). A common approach to address this problem is to calibrate  
43 a hydrological model for a gauged watershed using observed data and then transfer the model to the ungauged  
44 watershed by transferring the model parameters (Merz and Blöschl, 2003; Sivapalan et al., 2003). A great number  
45 of methods have been suggested for transferring model parameters, which include regression techniques between  
46 the model parameters and catchment attributes (e.g. Parajka et al., 2005; Deckers et al., 2010; Zhang et al., 2018)  
47 and similarity approaches such as spatial proximity and physical similarity (e.g. Bárdossy, 2007; Wagener et al.,  
48 2007; Patil and Stieglitz, 2014). However, as Thirel et al. (2015) point out, it is essential to assess and evaluate the  
49 ability of the hydrological models to perform efficiently under conditions different from those in which they were  
50 developed or calibrated.

51  
52 The Soil and Water Assessment Tool (SWAT) developed by the USDA Agricultural Research Service (ARS) is a  
53 public domain and open source integrated model and has been used worldwide for various applications. As a semi-  
54 distributed model, it allows the spatial variation of the parameters by dividing the basin into a number of sub-  
55 basins (Arnold et al., 1998; Srinivasan et al., 1998). It is equipped with a snowmelt algorithm based on a simple  
56 temperature-index approach, which, although simple, is proved to be very effective in numerous studies (Hock,  
57 2003) especially when net solar radiation is the dominant driving energy for snowmelt (Debele et al., 2010).

58  
59 SWAT has been widely used in many studies for the simulation of runoff and nutrient cycling in agricultural and  
60 forested sites. Although there is an increasing interest in applying SWAT on snow-dominated (Grusson et al.,  
61 2015) and glacierised watersheds (Rahman et al., 2013; Garee et al., 2017; Omani et al., 2017), its transferability  
62 under the extreme conditions of these high altitude environments has not been tested yet. In this study, we have a  
63 quite unique situation of a small well gauged watershed, the Damma glacier watershed, which is part of the larger  
64 catchment feeding the Göschenalpsee reservoir, for which we have hydrological data thanks to its use by the  
65 hydroelectric power plant. This way we were able to assess the transferability and upscaling of SWAT, by  
66 calibrating the model for the Damma glacier watershed and then transferring it to the greater area feeding the  
67 Göschenalpsee reservoir. Subsequently, climate change simulations were conducted in order to assess the  
68 transferability of the model on a temporal scale. The assessment was conducted by comparing our findings with  
69 those of a previous study for the same area, which used two other hydrological models with different  
70 characteristics, PREVAH and ALPINE3D (Kobierska et al., 2013).

71

## 72 **2 Study Site**

73 The Damma glacier watershed (Fig. 1a) is situated in the central Swiss Alps in Switzerland and was one of the  
74 Critical Zone Observatories established within the European project SoilTrEC (Banwart et al., 2011). It is located  
75 at an altitude between 1790 m and 3200 m above sea level, has a total area of 10 km<sup>2</sup> and a typical alpine climate  
76 with an average yearly temperature of 1 °C and yearly precipitation of 2400 mm (Kobierska et al., 2013). Damma

77 glacier covers 50 % of the watershed and due to climate change has retreated at an average rate of 10 m per year  
78 in the last 90 years. However, during 1920–1928 and 1970–1992 the recession was interrupted and the glacier  
79 grew, resulting in two moraines (Kobierska et al., 2011). After the retreat of the glacier a soil chronosequence is  
80 developed, which has a total length of 1 km (Bernasconi et al., 2008; Bernasconi et al., 2011; Kobierska et al.,  
81 2013). The bedrock is coarse-grained granite of the Aare massif and is composed of quartz, plagioclase, potassium  
82 feldspar, biotite and muscovite (Schaltegger, 1990). Our study site was extensively described in Bernasconi et al.  
83 (2011).

84

85 The Göschenalpsee (Fig. 1b) is a hydropower reservoir of a volume of 75 million m<sup>3</sup>. A 100 km<sup>2</sup> and 20 %  
86 glacier covered catchment drains into the reservoir. It includes the watersheds of the Damma, Chelen and Tiefen  
87 glaciers and the Voralptal watershed. The Tiefen glacier and Voralptal watersheds do not drain directly into the  
88 reservoir but their runoff is redirected through two tunnels. The site is described extensively in Kobierska et al.  
89 (2013).

## 90 **3 Model and Data**

### 91 **3.1 SWAT model**

92 In this study, we used SWAT 2012 coupled with the ArcView SWAT interface, a GIS-based graphical user  
93 interface (Di Luzio et al., 2002) that enables the delineation of the watershed, definition of subbasins, and initial  
94 parameterisation. It is a semi distributed, time continuous watershed simulator operating on a daily time step.

95

96 Each watershed is divided into subbasins, for which slope, river features, and weather data are considered.  
97 Furthermore, the watershed is divided into hydrologic response units (HRUs), which are small surface units with  
98 distinctive soil-land use combinations and necessary to capture spatially explicit processes. Each process is  
99 simulated for each HRU and then summed up for the subbasin by a weighted average. Subsequently the amount  
100 of water, sediment and nutrients that come out from each subbasin enter the respective river.

101

102 A modified SCS curve number method is used to calculate the surface runoff for each HRU, based on land use,  
103 soil parameters, and weather conditions. The water is stored in four storage volumes: snow, soil moisture, shallow  
104 aquifer and deep aquifer. The processes considered within the soil profile are infiltration, evaporation, plant uptake,  
105 lateral flow, and percolation. What is important in our study is that melted snow is handled by the model the same  
106 way as the water that comes from precipitation regarding the calculation of runoff and percolation. The factors  
107 controlling snow melt are the air and snowpack temperature, the melting rate and the area covered by snow. The  
108 updated snow cover model takes into account shading, drifting, topography and landcover to create a nonuniform  
109 snow cover (Neitsch et al., 2011). Furthermore, runoff from frozen soil can also be calculated by defining if the  
110 temperature in the first soil layer is less than 0°C. Even though the model still allows significant infiltration when  
111 the frozen soils are dry, the runoff of frozen soils is larger than that of other soils. A detailed description of the  
112 theory behind the model is found in detail in Arnold et al. (1998) and Srinivasan et al. (1998).

113

114 Snow processes in high alpine areas are strongly influenced by the terrain features (Ahl et al., 2008; Zhang et al.,  
115 2008). Fontaine et al. (2002) revealed the importance of improving SWAT algorithms to include in the model the

116 influence of elevation and season on the dynamics of the snowpack.. They found that the definition of elevation  
117 bands within the model subbasins can significantly improve the performance of the model in watersheds at high  
118 altitudes and with large elevation gradients. With the improved snow melting algorithm (Fontaine et al., 2002),  
119 streamflow in alpine regions can be successfully simulated by SWAT (Rahman et al., 2013; Grusson et al., 2015;  
120 Omani et al., 2017).

## 121 **3.2 Input data**

122 The input data required by SWAT are: topography, soil, land use and meteorological data.

### 123 **3.2.1 Topography**

124 The topography of both study areas was defined using a high precision Digital elevation model (DEM) with 2 m  
125 grid cells (swissALTI3D), produced by the Swiss Federal office for Topography  
126 (<http://www.swisstopo.admin.ch/internet/swisstopo/en/home/products/height/swissALTI3D.html>).

### 127 **3.2.2 Soil and land use map**

128 In order to better describe the glacier forefield and to reduce the uncertainty of the calibration for the Damma  
129 glacier watershed, detailed soil and land use maps were created based on the observations, field and experimental  
130 data from the Biglink and SoilTrEC projects (Bernasconi et al., 2011; Dumig et al., 2011; Andrianaki et al., 2017).  
131 The soil map was created by adding new soil types to the SWAT database while the land use classes were based  
132 on existing types in the database. For the greater area feeding the Göschenalpsee, the soil map used, was  
133 produced and provided by the Swiss Federal Statistical Office at a scale of 1:200,000  
134 (<http://www.bfs.admin.ch/bfs/portal/en/index.html>). For land use, we used the Corine land cover dataset 2006  
135 (version 16, 100m resolution) produced by the European Environmental Agency (<http://www.eea.europa.eu/data-and-maps/data/corine-land-cover-2006-raster-2>).

### 137 **3.2.3 Climate data**

138 Meteorological data from one local weather station and one station of the SwissMetNet network were used. The  
139 weather stations are located at the Damma glacier watershed (2025 m a.s.l.) and at Gütsch (2283 m a.s.l.). The  
140 meteorological data of the weather Gütsch were provided by MeteoSwiss. The selection of the weather station  
141 Gütsch was based on the results of previous research that showed that it has the best correlation in comparison to  
142 other weather stations located in the area (Magnusson et al., 2011) with a long enough record for this study. The  
143 data from both stations consist of sub-hourly records of air temperature, precipitation, wind speed, relative  
144 humidity, incoming short-wave radiation and incoming long-wave radiation from 2007–2013 for Damma weather  
145 station and 1981–2010 for Gütsch. The lapse rates for temperature and precipitation, which are very important  
146 parameters in SWAT model since they affect snow and glacier melt, and the interpolation methods were based on  
147 the findings of Magnusson et al. (2011) who carried out non prognostic hydrological simulations for the Damma  
148 glacier watershed. The precipitation and temperature lapse rate parameters of the model are PLAPS and TLAPS  
149 and were set to 5 mm km<sup>-1</sup> and -5.84 °C km<sup>-1</sup> respectively.

150 **Climate change scenarios:** The climate change predictions were provided by the EU regional climate modelling  
151 initiative ENSEMBLES (van der Linden and Mitchell, 2009) and were based on the emission scenario A1B. The  
152 model chains produced by the ENSEMBLES project are a combination of a general circulation model (GCM) with

153 a regional climate model (RCM). In Switzerland, model chain data were interpolated to the locations of the  
154 MeteoSwiss stations and the Swiss Climate Change Scenarios CH2011 were created (CH2011, 2011). The delta-  
155 change method was used for the creation of the datasets (Bosshard et al., 2011). Temperature and precipitation  
156 predictions are calculated using daily temperature changes  $\Delta T$ , and precipitation scaling factors  $\Delta P$ . Incoming  
157 short-wave irradiation, wind speed and relative humidity were left unchanged. In Switzerland it is predicted that  
158 the mean temperature will increase 2.7–4.1°C and the precipitation during the summer months will decrease 18%–  
159 24% by the end of the century, in the case when no actions for the mitigation of climate change are taken (CH2011,  
160 2011).

161  
162 In this study, three climate scenarios with interpolated data for Gütsch weather station are used. These scenarios  
163 are: the CNRM ARPEGE ALADIN scenario, the ETHZ HadCM3Q0 CLM scenario, which predicts the highest  
164  $\Delta T$  and  $\Delta P$  in comparison to the other two, and the SHMI BCM RCA scenario, which predicts the lowest  $\Delta T$  and  
165  $\Delta P$ , referred to as CNRM, ETHZ and SHMI scenarios respectively. The CNRM, ETHZ and SHMI scenarios were  
166 chosen to be in agreement with the previous study of Kobierska et al. (2013), to be able to carry out a direct  
167 comparison of the three models. The following periods were selected:

168 Reference period (T0): 1981–2010

169 Near future period (T1): 2021–2050

170 Far future period (T2): 2070–2099

171  
172 Similarly to the predictions for Switzerland, the scenarios for Gütsch weather station predict warmer and dryer  
173 summers and slightly increased precipitation in autumn. The highest  $\Delta T$  for the near future period is 1.5°C in the  
174 mid-summer, 2.5°C in late spring, and below 1.0°C in early summer for the CNRM, ETHZ and SHMI respectively  
175 and for the far future period is approximately 5°C in the mid-summer, 4°C along the whole summer and 3°C in  
176 early summer respectively. The biggest temperature increase is predicted at the end of the century when the  
177 strongest agreement between the different model chains is observed. Precipitation changes for the near future  
178 period are within the natural variability apart from a clear trend in dryer summers. The trend of dryer summers is  
179 most prominent for the far future period. Furthermore, most model chains predict slightly higher precipitation in  
180 autumn. The average  $\Delta P$  value for the near future period is 1.0 and for the far future period is 0.99. The climate  
181 change data were also used for different sites in the Alps (Bavay et al., 2013; Farinotti et al., 2012).

### 182 **3.2.4 Runoff data**

183 Runoff of the Dammareuss stream that drains the Damma glacier watershed was measured every half an hour at a  
184 gauging station at the outlet of the watershed (Magnusson et al., 2011). The runoff of the total area that feeds the  
185 Göschenalpsee is the inflow of the reservoir and the data from 1997–2010 were provided by the energy company  
186 responsible for the management of the reservoir.

### 187 **3.2.5 Glacier extent**

188 Data on the glacier extent for the present period but also for the two periods of the climate change scenarios were  
189 provided by Paul et al (2007). They estimated the evolution of the Swiss glaciers by using hypsographic modelling,  
190 based on the shift of the equilibrium line altitude. However, SWAT is not a model that considers glacier flow  
191 dynamics and therefore, in this study, the glaciers were incorporated in SWAT as the initial snow content in each

192 subbasin and for each elevation band. The initial snow is given as the snow water equivalent in mm instead of  
193 snow as the density of snow can be variable. For this reason, the calculation of the snow water equivalent was  
194 conducted by considering an average density of ice.

#### 195 **4 Methodology**

196 The purpose of this study is to assess the transferability of SWAT in temporal and spatial scales at a high altitude  
197 alpine and glacierised site. This way it is tested whether the model can be transferred and is capable for the  
198 simulation of runoff but also for further climate change studies on an ungauged glacierised watershed.  
199 Furthermore, this methodology tests its robustness under these extreme climatic and geographical conditions. For  
200 this reason, SWAT was initially calibrated for the small Damma watershed, which is well monitored through the  
201 CZO projects, and then it was upscaled and applied for the greater area feeding the Göschenalpsee reservoir and  
202 includes the Damma glacier watershed. The upscaling of the model was verified by comparing model results with  
203 the reservoir data provided by the managing company.

204  
205 Since the Damma glacier watershed is part of the greater Göschenalpsee feeding catchment, the parameters of  
206 the model were transferred using the spatial proximity approach, with no further regionalisation procedure. In this  
207 case, the initial setup of SWAT for the greater catchment was conducted using the input data presented in section  
208 3.2 and only the parameters presented in Table 1 were changed to the calibrated values derived from the calibration  
209 of the Damma glacier watershed. The initial parameterisation of the model during the setup and the watershed  
210 delineation assisted in the transferability of the model since a number of parameters is already defined based on  
211 the topography, land use and soil data.

212  
213 Subsequently, in order to assess its transferability on a temporal scale, climate change simulations were conducted  
214 and results were compared with those of a previous study for the same area, which used two other hydrological  
215 models with different characteristics, PREVAH and Alpine 3D (Kobierska et al., 2013).

216  
217 This methodology is a modified version of the proxy-basin test introduced by Klemeš (1986), which is one of the  
218 proposed testing schemes for the enhancement of the calibration and validation procedure in hydrological  
219 modelling. According to Klemeš (1986) the proxy basin test can be used to test the geographical transposability  
220 of the model between two regions, for subsequent simulation of the streamflow in ungauged watersheds with  
221 similar characteristics. The model is calibrated and validated for two different but similar watersheds and if the  
222 results are acceptable it is then considered safe to be transferred and used at a third watershed with similar  
223 characteristics.

#### 224 **5 Model setup, calibration and validation**

225 SWAT was initially setup for the Damma glacier CZO and the greater area feeding the Göschenalpsee using the  
226 topography, soil and land use data presented in section 3.2. Following the delineation procedure, the Damma  
227 watershed and the greater area were divided into 5 and 25 subbasins respectively. By setting the lowest possible  
228 thresholds for land use, slope and soil, 48 HRUs were created for Damma watershed and 285 HRUs for the greater  
229 area. Finally, six elevation bands were defined for each subbasin of both study sites. The setup was complete with

230 the addition of the meteorological input and the definition of the initial snow for each elevation band of each  
231 subbasin. For the climate change simulations, the meteorological input consists of the climate change scenarios  
232 described in section 3.2.3 and the initial snow that corresponds to the first year of each future period, as calculated  
233 by the glacier extent data described in section 3.2.5.

## 234 **5.1 Model calibration**

235 SWAT was calibrated for the Damma watershed only, using the meteorological data from 2009 to 2011 and  
236 validated with the data from 2012 to 2013. Data for the years 2007 and 2008 were used for the warm-up and the  
237 stability of the model. For the better identification of the parameters that influence the hydrology of the site the  
238 calibration was first conducted manually. The most sensitive parameters during this step were related to snow melt  
239 such as: i) TIMP, the snow pack temperature lag factor, ii) SMFMX, the snow melt factor on the 21st of June  
240 ( $\text{mmH}_2\text{O} / ^\circ\text{C day}^{-1}$ ), iii) SMFMN, the snow melt factor on the 21st of December ( $\text{mmH}_2\text{O} / ^\circ\text{C day}^{-1}$ ), CN\_FROZ,  
241 which was set to active in order and finally the snow fall and snow melt temperatures SFTMP and SMTMP  
242 respectively. Because most of the subbasins of the Damma glacier watershed, delineated during the initial setup  
243 of the model, were partially glacier covered, it was decided to follow a simple approach and apply the same snow  
244 parameters for all the subbasins. This means that the same parameters were applied for both glacier and snow  
245 dynamics.

246  
247 Groundwater flow parameters such as the GW\_DELAY, the groundwater delay time, ALPHA\_BF, the base flow  
248 alpha factor and the SURLAG, the surface runoff lag coefficient, were also found to play an important role on the  
249 performance of the model. Evapotranspiration (ET) related parameters were not significant since our study site is  
250 above the tree line and ET is relatively minor.

251  
252 The manual calibration was followed by an automatic calibration and uncertainty analysis using the SWAT-CUP  
253 software with the Sequential Uncertainty Fitting ver. 2 (SUFI-2) algorithm for inverse modelling (Abbaspour et  
254 al., 2007). Starting with some initial parameter values, SUFI-2 is iterated until (i) the 95% prediction uncertainty  
255 (95PPU) between the 2.5th and 97.5th percentiles include more than 90% of the measured data and (ii) the average  
256 distance between the 2.5th and 97.5th percentiles is smaller than the standard deviation of the measured data. A  
257 model is considered calibrated when the chosen criterion between the best simulation and calibration data reaches  
258 the best value (Abbaspour et al., 2007). The parameters introduced in SWAT-CUP as well as their range are the  
259 ones that were identified during the manual calibration as the most important.

260  
261 The criterion used for the calibration with SWAT-CUP is the Nash-Sutcliffe (Nash and Sutcliffe, 1970) model  
262 efficiency (NS), since it was the criterion available in SUFI-2 that is commonly used in hydrological studies. The  
263 NS shows the relationship between the measured and the simulated runoff (Eq. 1). The performance of the  
264 calibrated model was further evaluated by the square of Pearson's product moment correlation  $R^2$ , which represents  
265 the proportion of total variance of measured data that can be explained by simulated data. Better model  
266 performance is considered when both criteria are close to 1. NS coefficients greater than 0.75 are considered  
267 "good," whereas values between 0.75 and 0.36 as "satisfactory" (Wang and Melesse, 2006).

$$268 \quad NS = 1 - \frac{\sum(y-\hat{y})^2}{\sum(y-\bar{y})^2}, \quad (1)$$

269 where  $y$  is the individual observed value,  $\hat{y}$  for the individual simulated value and  $\bar{y}$  the mean observed value.  
 270 However, as Schaepli and Gupta (2007) pointed out, the NS criterion is not enough to judge the efficiency of the  
 271 model when simulating runoff with high seasonality like the one in high altitude watersheds. Therefore, as an  
 272 additional criterion for the performance of the model, a benchmark efficiency indicator was calculated, according  
 273 to Eq. 2:

$$274 \quad BE = 1 - \frac{\sum_{t=1}^n [q_{obs}(t) - q_{sim}(t)]^2}{\sum_{t=1}^n [q_{obs}(t) - q_b(t)]^2}, \quad (2)$$

275 where  $q_{obs}$  is the observed runoff;  $q_{sim}$  is the simulated runoff by SWAT; and  $q_b$  is runoff given by the benchmark  
 276 model. The calendar day model was chosen as benchmark (Schaepli and Gupta, 2007), which is the observed  
 277 interannual mean runoff for every calendar day.

278  
 279 Table 1 shows the default and the after calibration values of the SWAT parameters that were changed during  
 280 calibration. TIMP was set to a very low value indicating that the glacier is not affected by the temperature of the  
 281 previous day as much as the snowpack would be. Snow and glacier melt in Damma watershed occurs from April  
 282 to September, a fact that explains the low value of the SMFMN parameter (0.1 mmH<sub>2</sub>O / °C-day), the minimum  
 283 melt factor, while the SMFMX is set to the value of 4.7 mmH<sub>2</sub>O / °C-day. SURLAG and GW\_DELAY play an  
 284 important role in the model performance as they control the melted snow routing process and the hydrologic  
 285 response of the watershed. Damma glacier watershed has a fast response and therefore GW\_DELAY was set to  
 286 0.5 days. SMTMP is also sensitive since it is the controlling factor for the initialisation of the snow melt,  
 287 considering the availability of snow for melting on a specific day. As a result, model-generated peak runoff is  
 288 significantly influenced by the variation in SMTMP. Finally, ALPHA\_BF was set to value 0.95, which is a typical  
 289 value for a fast response watershed.

290  
 291 The results of the calibrated model for the daily runoff and the observed data are presented in Fig. 2(a), while  
 292 cumulative runoff is presented in Fig. 2(c). The fit of the model to the observed data is satisfactory and the results  
 293 of the calibrated model matched the observed data throughout most of the year. The graph of the cumulative runoff  
 294 (Fig. 2c) shows that runoff is slightly overestimated in July and August, when it is dominated by glacier melt. Best  
 295 results occur for the years 2009 and 2010. 2011 is characterised by unusually warm and dry months of September,  
 296 October and November which resulted in a slight underestimation of the runoff. Overall SWAT performance for  
 297 the calibrated period is considered very satisfactory since the NS efficiency is 0.84 and R<sup>2</sup> is 0.85. BE for this  
 298 period is 0.22, a value that we consider to be satisfactory and is comparable to that of the previous model, calibrated  
 299 for the greater area of Göschenalpsee.

## 300 **5.2 Sensitivity analysis**

301 The automatic global sensitivity analysis was conducted with SWAT-CUP software and 17 input parameters were  
 302 analysed. It revealed that the most sensitive parameters are the same as the ones observed during manual  
 303 calibration. More specific the most sensitive ones in descending order are TIMP, GW\_DELAY, SMTMP,  
 304 SMFMX, ALPHA\_BF and SURLAG with p values 0 for TIMP and very close to 0 for the remaining parameters.  
 305 The least sensitive parameters were left to their default value.



### 306 **5.3 Model validation**

307 SWAT was validated using the meteorological data for 2012 and 2013 and the results of the model as well as the  
308 measured runoff are presented in Fig. 2(b). Figure 2(d) presents the cumulative graphs. The SWAT model for this  
309 period performed efficiently, similarly to the calibration period, with a Nash-Sutcliffe efficiency of 0.85,  $R^2$   
310 0.86 and the BE 0.25. A small inconsistency is observed in the late spring of 2012, when estimated runoff is  
311 underestimated, probably due to the extremely wet May in that year that cannot be efficiently simulated. Although,  
312 due to the lack of longer monitoring data, the total calibration-validation period 2009-2013 is short, it still includes  
313 a relatively large variability in the weather conditions and precipitation amounts and despite this variability the  
314 overall model performance is very satisfactory. The small seasonal differences in model performance are due to  
315 the evolution of runoff generation throughout the season: runoff in spring and early summer (May, June) comes  
316 mainly from snowmelt while in July and August it stems mainly from glacier melt. Although there are two different  
317 water sources during the two different periods, we can only assign one set of parameters. We can nevertheless  
318 conclude that SWAT can be successfully applied for a partly glacierised watershed.

## 319 **6 Results and Discussion**

### 320 **6.1 Upscaling SWAT to the greater catchment feeding the Göschenalpsee reservoir**

321 The results of the model for the greater area that feeds the Göschenalpsee, are presented in Fig. 3(a) together  
322 with the measured inflow in the reservoir. The observed and predictive cumulative flow is presented in Fig. 3(b).  
323 Model performance criteria were lower than for the calibration period as NS dropped to 0.49 and the  $R^2$  to 0.72.  
324 The cumulative graph shows that there is an overall good agreement between model results and the measured  
325 reservoir inflow. Both Figures 3(a) and 3(b) show that there is an overestimation of total runoff for the period  
326 1999-2002, which appears to be linked to the higher precipitation amounts during this period. Measured  
327 precipitation measured at Gütsch weather station for this period is up to 46 % higher than the average precipitation  
328 of 1981-2010.

329  
330 The predictability of the model was further tested by analysing key parameters related to median runoff such as  
331 spring snowmelt timing, timing of peak flow, autumn recession period and the centre of mass (COM), which can  
332 indicate temporal shifts in the hydrological regime. Table 2 shows the difference in days between the observed  
333 and simulated values of the above parameters for each year of the period 1997-2010. A 15day moving average  
334 window was applied to daily runoff. Snowmelt timing and autumn recession are predicted successfully since the  
335 differences for most years are zero or close to zero, except for 2000 and 2002 for autumn recession. Peak flow  
336 timing shows some inconsistencies between observed and simulated data for certain years, which are mainly  
337 related to the fact that for these years and during the snowmelt period, SWAT produces results with higher peaks.  
338 Finally, the COM of the simulated data is in good agreement with that of the observed data, with an average  
339 difference of 4 days.

340  
341 On the whole, SWAT performance is considered to be satisfactory and it was successfully transferred to the greater  
342 Göschenalpsee feeding catchment. One of the main reasons for the deterioration of the model performance  
343 during the years with higher precipitation, 1999-2002, is that SWAT doesn't differentiate between snow and  
344 glacier dynamics and only one parameter for both snowmelt and glacial melt rate was applied. This becomes more

345 important in our study, since there is a difference between the percentage of glacial coverage of the two catchments,  
346 with the Damma glacier watershed being 50% covered while the greater catchment 20%. In Omani et al. (2017)  
347 this issue was partly addressed by applying different snow parameters to the glacier covered subbasins. However,  
348 the subbasins in our calibration watershed, the Damma glacier watershed, were partly glacierised and for this  
349 reason it was decided to apply only one set of snow parameters for the whole watershed.

350

351 Furthermore, some inconsistency is caused by the fact that for the two out of the four of the watersheds of the  
352 greater area feeding the Göschenalpsee, runoff is drained through tunnels into the reservoir. In addition, there is  
353 a difference in the hydrological response between the Damma glacier watershed in comparison to the greater area.  
354 Damma is characterised by very steep slopes (even up to nearly 80 degrees) and runoff originates mainly from  
355 snowmelt, glacial melt and rainfall (Magnuson et al., 2012). The ALPHA\_BF parameter of SWAT was set to a  
356 high value and the GW\_DELAY to low, parameter trends that characterise a watershed with a high response. On  
357 the other hand, the Göschenalpsee feeding area is less steep on average. The combination of these two factors  
358 might be the reason, why some of the simulated peaks are higher but also narrower compared to the observed  
359 inflows into the reservoir.

360

361 Finally, SWAT results were compared to results from PREVAH and ALPINE3D models, already published in  
362 Magnusson et al. (2011) and Kobierska et al. (2013) (Fig. 4). PREVAH is a semi-distributed conceptual  
363 hydrological model suited for applications in mountainous regions (Viviroli et al., 2009a; Viviroli et al., 2009b)  
364 while ALPINE3D is a fully-distributed energy-balance model (Lehning et al., 2006).

365

366 Figure 4 shows the interannual average of the period 1997-2010 daily runoff for each model. SWAT overestimated  
367 the runoff of the snowmelt period, from May to the beginning of July, while from mid July to late September its  
368 results are close to the observed values and in agreement with the other two models. Finally, in October runoff is  
369 slightly underestimated. The seasonality in variation between model results and observed values is linked to the  
370 application of only one melt rate for both snowmelt and glacial melt periods. The best fit of the model is observed  
371 when glacial melt is the major contributor to runoff, while it is overestimated during the snowmelt period, which  
372 is the reason of the excessive simulated runoff during the 1999-2002 period of high precipitation (Fig.3), as  
373 discussed above. Seasonal variability in model performance is observed not only for SWAT but also for  
374 ALPINE3D and PREVAH, as ALPINE3D underestimated runoff during the snowmelt period, from May to June,  
375 while on the other hand runoff was slightly overestimated by PREVAH in October and November (Kobierska et  
376 al., 2013).

377

378 Furthermore, a combination of the factors discussed above about the applied snowmelt parameters and the  
379 deviation in hydrological response between the two areas because of human intervention and  
380 topographical/geographical features is the reason why SWAT doesn't simulate efficiently the winter low flows.

## 381 **6.2 SWAT transferability on a temporal scale**

382 As a next step, we assessed whether SWAT can be transferred at a temporal scale, by running climate change  
383 scenarios for the greater area that feeds the Göschenalpsee. In order to verify the model transferability, results  
384 were compared with the climate change study in Kobierska et al. (2013) using the same time periods as follows:

385 Reference period (T0): 1981–2010  
386 Near future period (T1): 2021–2050  
387 Far future period (T2): 2070–2099

388

389 The results of SWAT model are presented as the interannual average runoff for each different scenario along the  
390 whole period in Fig. 5(a) for the near and in Fig. 5(b) for the far future periods.

391

392 During the reference period runoff peaks in early July when snowmelt is combined with glacier melt. For the near  
393 future period T1, the main difference happens from July to September when runoff is dominated by glacier melt.  
394 During this period, predicted runoff for all scenarios, and in particular for the warmer ETHZ scenario, is lower  
395 than the reference period, indicating that the glacier melt cannot compensate the predicted decrease in precipitation.  
396 From September until the end of the season, simulated stream flow of all scenarios is higher than the reference  
397 period, which is explained by the higher predicted precipitation during autumn. The annual peak remains in early  
398 July, since the glacier has not melted away yet, providing glacier melt.

399

400 For the far future period T2, runoff from spring to June is predicted to increase significantly for all three scenarios  
401 due to more intense snowmelt. In addition, higher precipitation is predicted by the climatic data for this period.  
402 Based on the available glacier extent data described in section 3.2.5, we estimated that in 2070, the total glacier  
403 volume will be reduced to almost half, resulting in less glacial melt between July and late August. For this reason,  
404 and in combination with the significant decrease in precipitation, predicted by all scenarios for this period, the  
405 simulated runoff is lower than that of the reference. Finally, the snow free period will extend until December  
406 instead of September.

407

408 At the end of the T2 period, the average temperature increase in our site will be 3.35°C and only a small part of  
409 the glacier will remain in high elevation. The date of peak flow will shift to be in the beginning of June. The main  
410 runoff volume is expected to be observed in spring and early summer while during the glacier melt period,  
411 streamflow is significantly lower than that of the reference period. Overall the total water yield for the scenarios  
412 in T2 period is predicted to decrease.

413

414 To better observe the seasonal changes of estimated runoff, Fig. 6 shows the interannual average runoff for a)  
415 May-June, b) July-August and c) September-October for the T1 and T2 future periods divided by the average of  
416 the reference period of the same months for all the three scenarios. In May and June, as mentioned above, runoff  
417 is mainly dominated by snowmelt. The three climate change scenarios predict increased temperatures and higher  
418 precipitation during May and June which result in faster snowmelt and therefore in the increased predicted runoff,  
419 as observed in Fig. 6(a). The increase is higher in the far future due to the higher temperatures. The only exemption  
420 to that is the SHMI scenario for the near future period, since it is the colder scenario that predicts the lowest  
421 temperature and precipitation changes. In July and August, climate change scenarios predict a significant decrease  
422 in precipitation, which is also depicted in the predicted runoff. The scenario that has the most drastic effect is the  
423 ETHZ because it is the scenario that predicts the highest increase in the temperature and decrease in the  
424 precipitation. For September and October, results do not show a clear trend for the warmer ETHZ scenario,  
425 however for the CNRM and SHMI scenarios, predicted runoff is lower than the reference. Finally the predicted

426 runoff of the far future period T2 shows higher fluctuations from year to year than that of the near future period  
427 especially from September to October.

428

429 The climate change predictions of SWAT and the subsequent conclusions show many similarities in the seasonal  
430 variations with that of ALPINE3D and PREVAH. This observation is very promising since it demonstrates that  
431 SWAT could be applied to climate change studies in ungauged high altitude watersheds. There are however  
432 uncertainties and differences between the models. ALPINE3D and PREVAH models predict the spring peak flow  
433 to shift approximately by 3 and 6 weeks for the near and far future periods respectively. On the other hand, the  
434 shift in peak flow with SWAT is smaller and especially for the near future period a 10 day shift is predicted only  
435 with the warmer ETHZ scenario (Fig. 5).

## 436 **7 Conclusions**

437 This study is an assessment of the transferability or upscaling of SWAT on a spatial and temporal scale for a partly  
438 glacierised catchment at a high altitude. For this reason, we followed an approach similar to the proxy-basin test  
439 introduced by Klemeš (1986).

440

441 Firstly, SWAT was calibrated and validated for the Damma glacier watershed and it was demonstrated that despite  
442 the extreme conditions of this high alpine watershed, SWAT performed successfully, with satisfactory NS and BE  
443 efficiencies. Subsequently, we assessed the transferability of the model by upscaling and applying SWAT for the  
444 greater area that drains into the Göschenalpsee reservoir and includes the Damma glacier watershed. By  
445 comparing model results with existing inflow data, we showed that the model was able to predict key parameters  
446 such as the snowmelt timing, autumn recession period and the peak flow timing. However, overestimation of  
447 runoff during the snowmelt period, especially in wet years, highlights the importance of taking into account the  
448 difference in snow and glacier dynamics. It showed that better performance could have been achieved if different  
449 parameters for snow and glacial melt had been applied. This observation is quite important for study sites where  
450 streamflow is greatly dependent on both snow- and glacier melt. Model performance was potentially affected in  
451 the greater catchment due to hydropower infrastructure such as tunnels.

452

453 The temporal transferability of SWAT was analysed by assessing the impact of climate change on the hydrology  
454 of the greater catchment and comparing these results with a previous climate change study conducted for the same  
455 area. Climate change predictions showed that the hydrological regime will change significantly in the future  
456 especially towards the end of the century. Daily runoff during May and June is predicted to increase because more  
457 intense snowmelt and the predicted wetter springs. Projected runoff from July to October, mainly for the far future  
458 period but also for the near future, is significantly decreased. These results show many similarities with those  
459 previously published.

460

461 In conclusion, our findings indicate that SWAT is a model that can be successfully transferred to simulate  
462 streamflow and climate change impact for high altitude glacierised ungauged watersheds. The upscaling  
463 methodology used here, where SWAT is calibrated for a small watershed and then applied for a greater area that  
464 includes the calibration watershed, is a simple but still effective approach. It can be valuable in predicting

465 streamflow of ungauged watersheds, in large scale hydrological simulations and for policy makers working in  
466 water management.

#### 467 **Author Contributions**

468 Maria Andrianaki applied SWAT model, analysed data and prepared the manuscript with contributions from all  
469 co-authors. Juna Shrestha reviewed the manuscript and assisted in the modelling procedure. Florian Kobierska  
470 provided meteorological and runoff data. Nikolaos P. Nikolaidis provided guidance for the research goals. Stefano  
471 M. Bernasconi was the supervisor of the research project and provided the funding that lead to this publication.

#### 472 **Competing interests**

473 The authors declare that they have no conflict of interest.

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**Table 1 The default and calibrated values of the most sensitive SWAT parameters**

<b>Parameter</b>	<b>Unit</b>	<b>Cal. Value</b>	<b>Default</b>
SFTMP	°C	-0.5	1
SMTMP	°C	2.5	0.5
SMFMX	mm H <sub>2</sub> O / °C day <sup>-1</sup>	4.7	4.5
SMFMN	mm H <sub>2</sub> O / °C day <sup>-1</sup>	0.1	4.5
TIMP		0.011	1
SURLAG		0.001	4
CNCOEF		0.5	1
SNOCOVMX	mm H <sub>2</sub> O	500	1
SNO50COV	%	0.3	0.5
ALPHA_BF	days	0.95	0.048
GW_DELAY		0.5	31
GW_REVAP		0.02	0.02
LAT_TTIME		0.0001	0
CN2		35	Na
SLSOIL	m	5	Na
ESCO		1	0.95
SOL_AWC	mm H <sub>2</sub> O/mm soil	0.05	Na

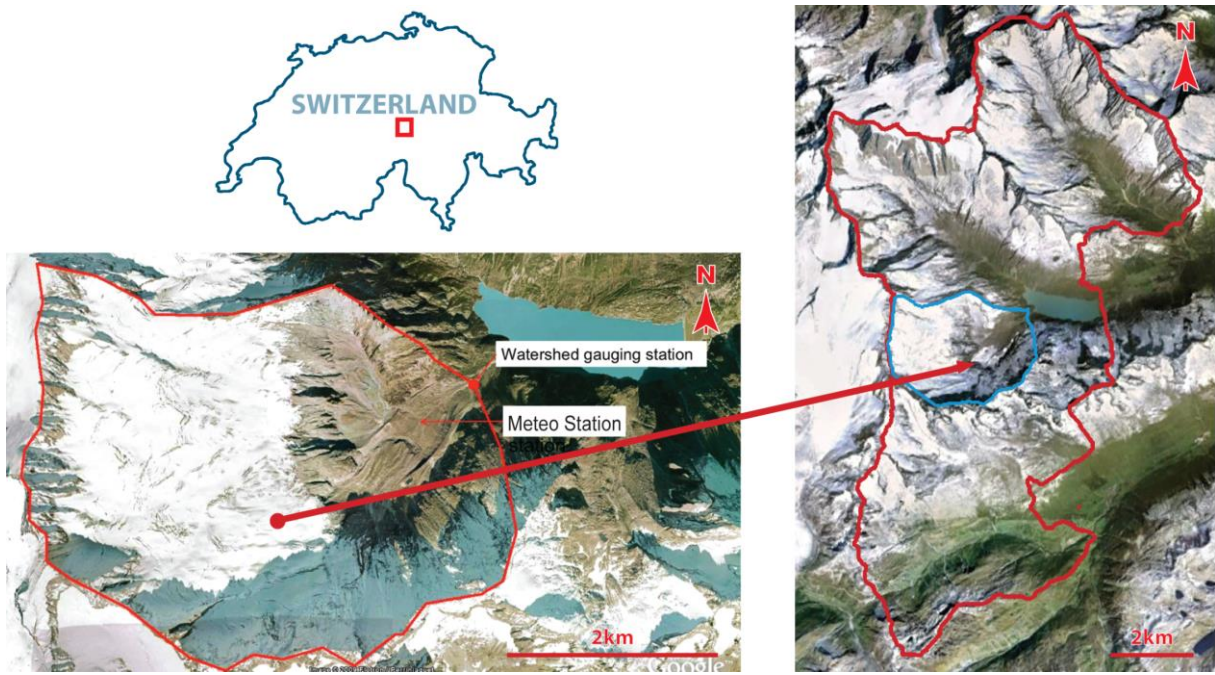
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**Table 2 Absolute difference in days between simulated and observed values of the snowmelt timing, autumn recession period, peak flow timing and the centre of mass (COM), for the greater catchment feeding the Göschenalpsee.**

<b>Year</b>	<b>Snowmelt timing</b>	<b>Autumn recession period</b>	<b>Peak flow timing</b>	<b>COM</b>
1997	0	1	48	7
1998	2	1	2	4
1999	4	0	27	1
2000	0	16	19	3
2001	0	1	1	1
2002	0	19	0	8
2003	2	5	2	1
2004	1	4	21	2
2005	1	0	1	4
2006	3	1	3	4
2007	3	1	7	8
2008	2	0	2	3
2009	1	0	13	5

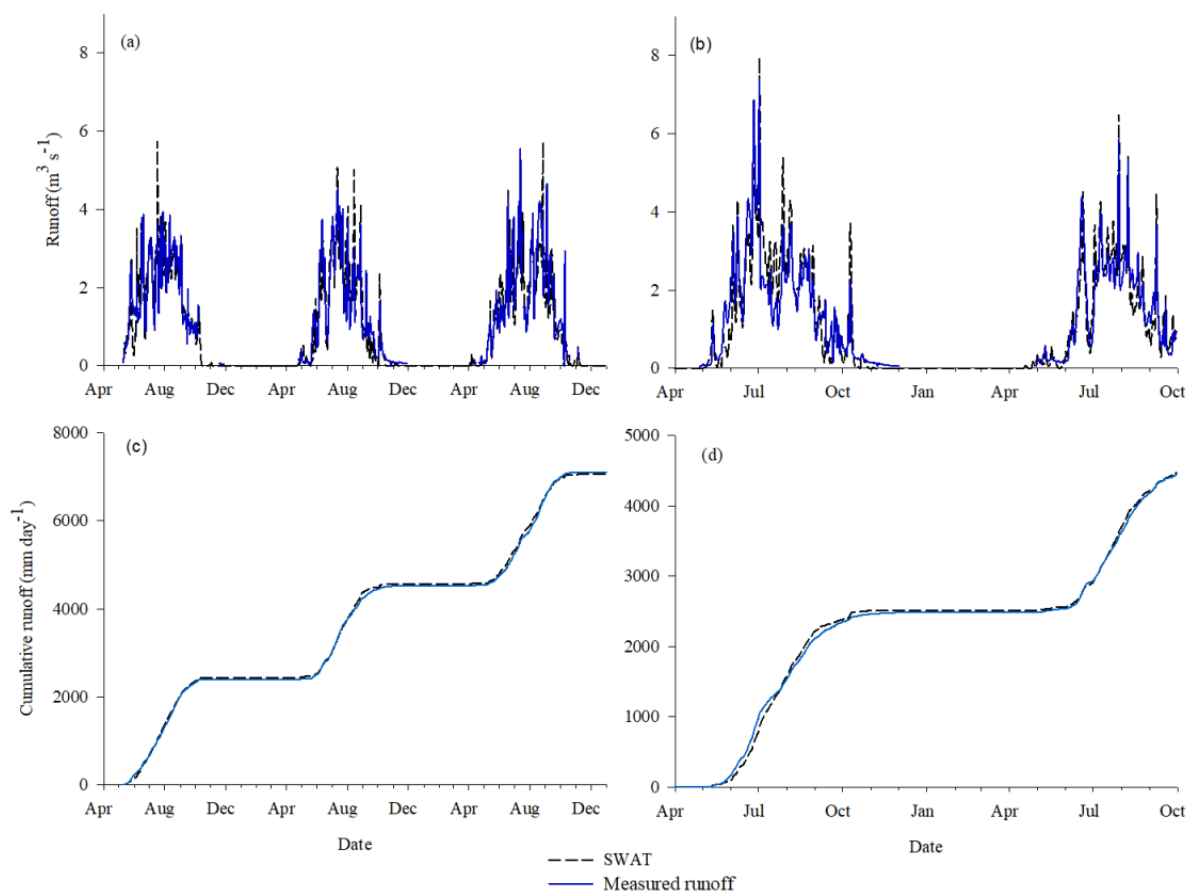
2010	2	0	1	6 630
Average	1.4	3.5	11.0	4.0

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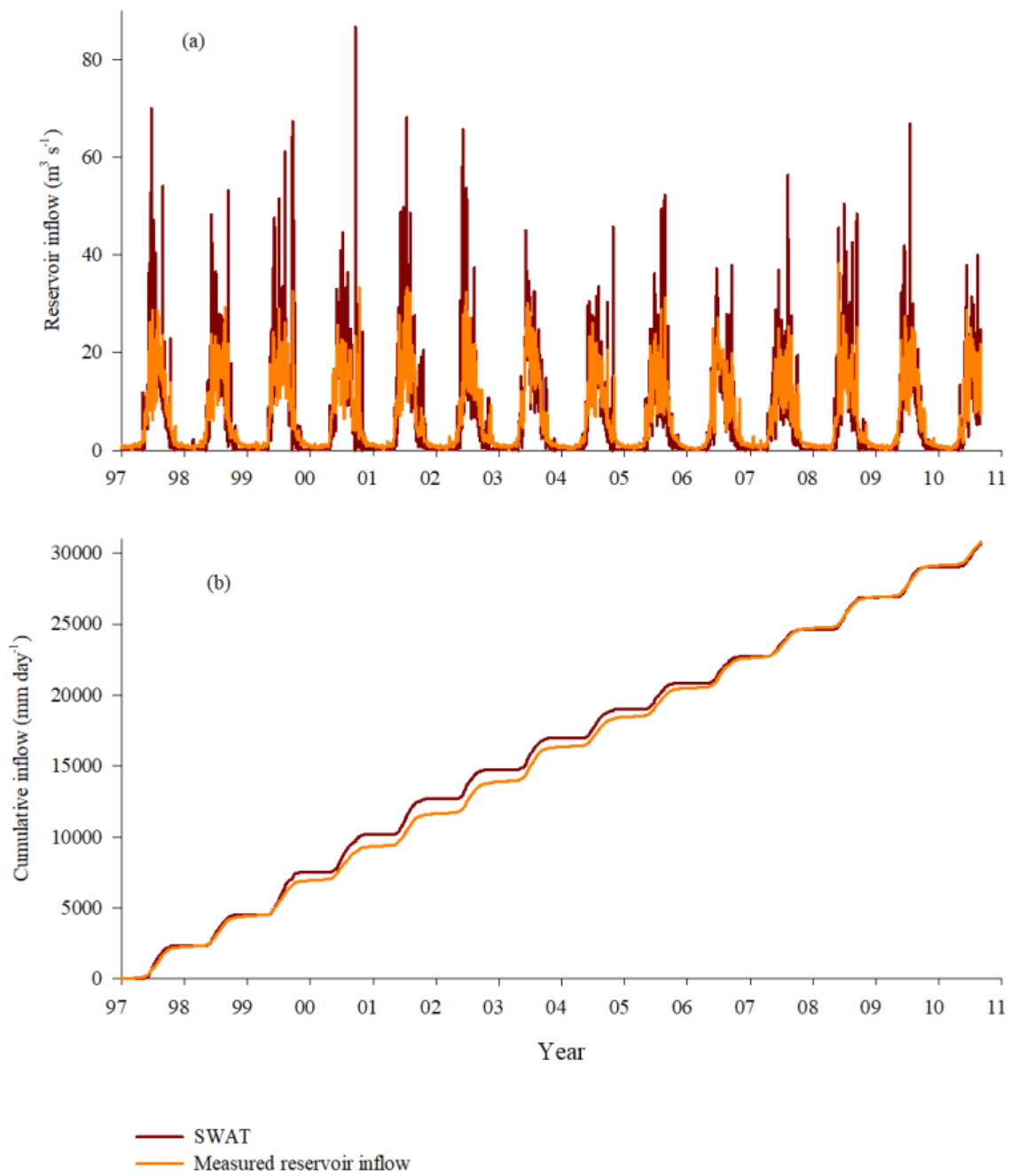
**Figure 1** Map showing the Damma glacier watershed on the left and the greater area that feeds the Göscheneralpsee on the right.



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651 **Figure 2** Observed and cumulative runoff for a) and c) the calibration period 2009-2011 and for b) and d) the validation  
 652 period 2012-2013

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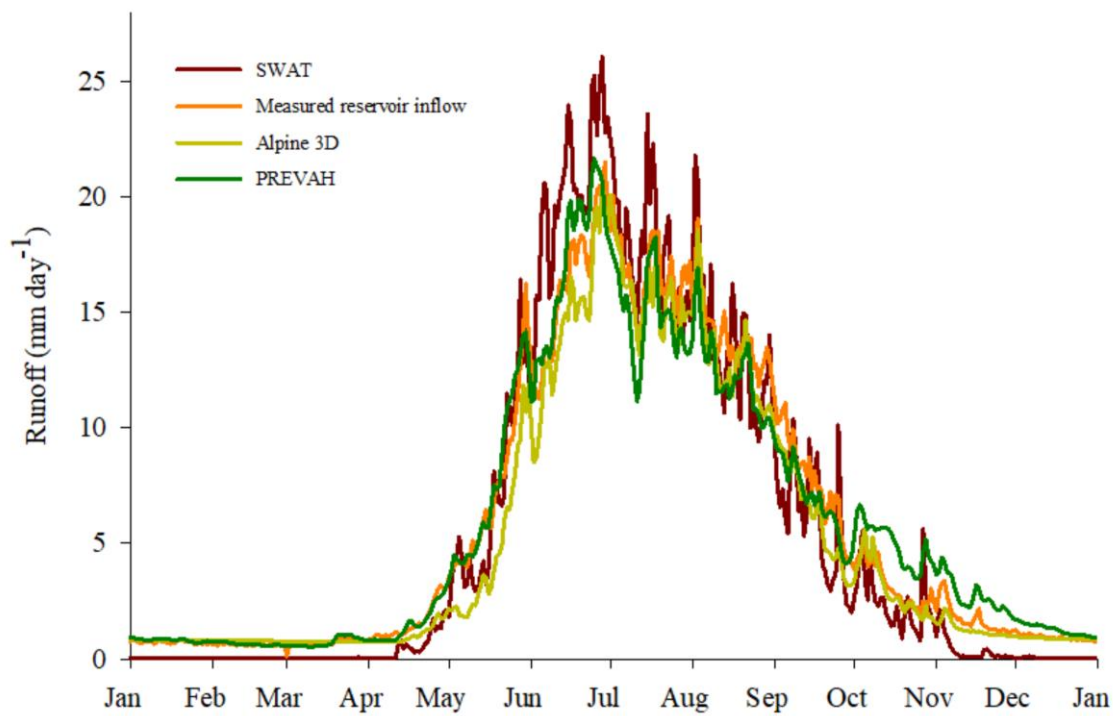


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656 **Figure 3** SWAT results and measured inflow of the feeding catchment of the Göschenalpsee reservoir for the period  
 657 **1997-2010. Graphs in (b) show the observed and simulated cumulative runoff over this period.**

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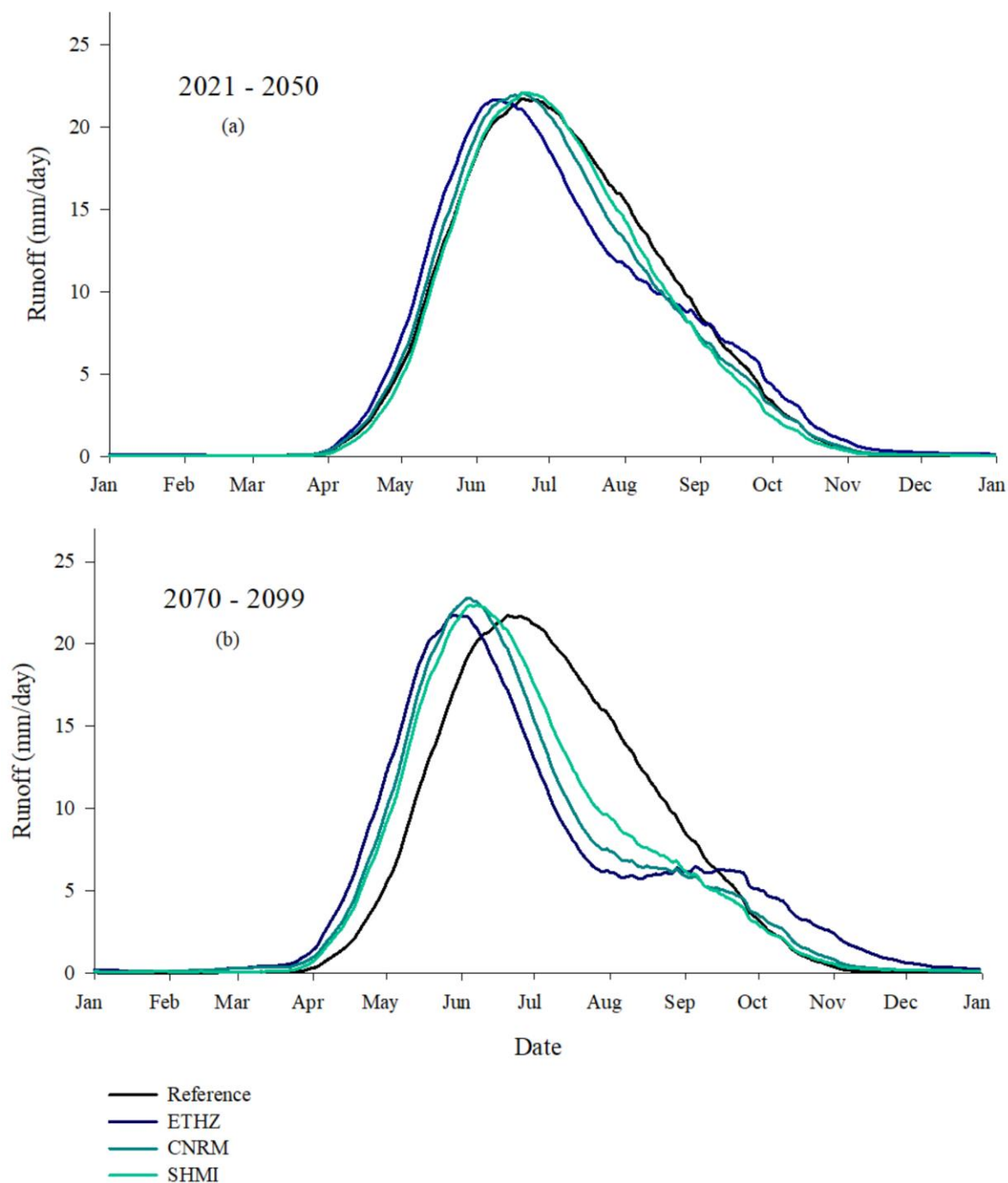
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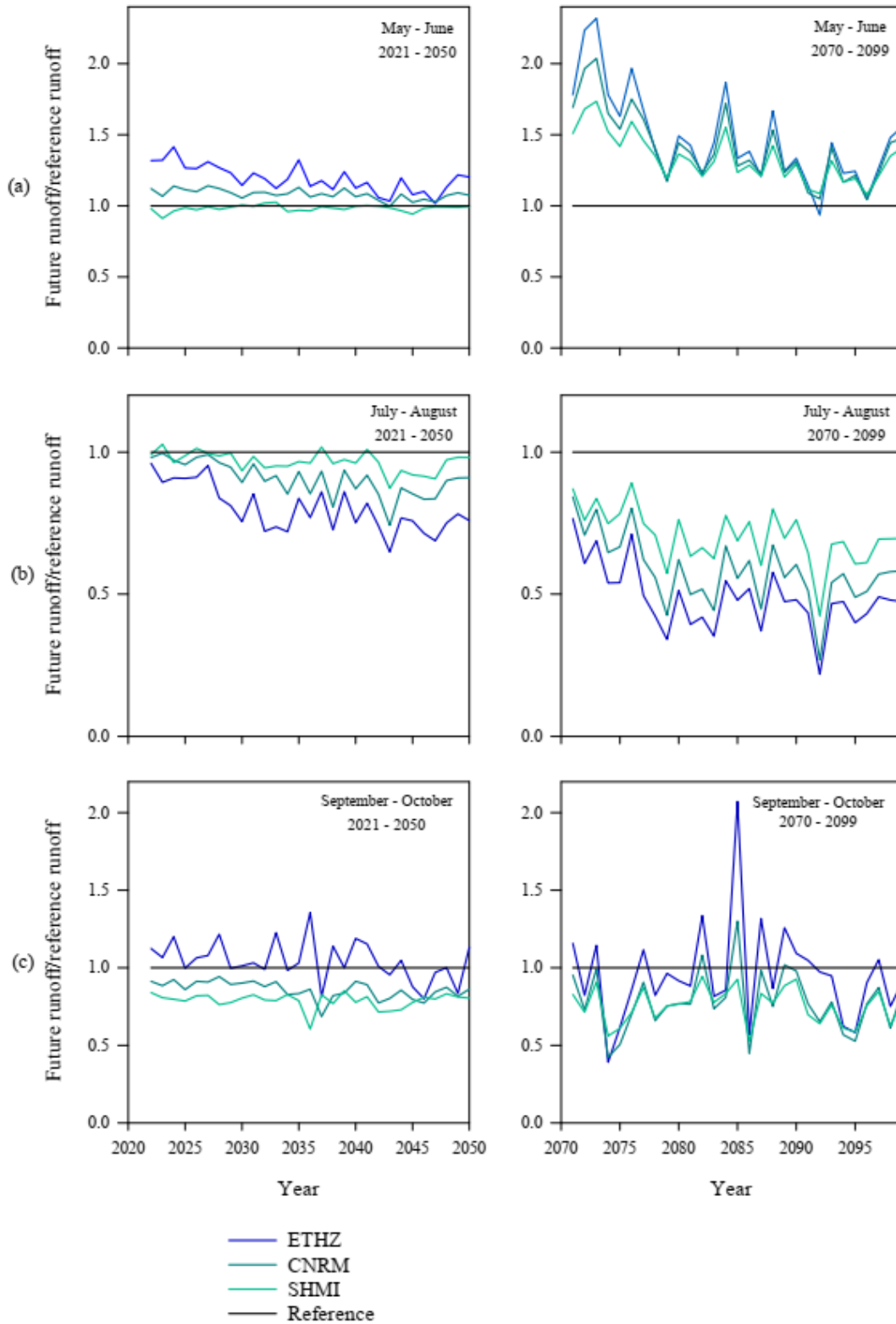
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662 **Figure 4 Interannual average of the results of SWAT, ALPINE3D and PREVAH models and the measured runoff of**  
663 **the Göschenalpsee feeding catchment for the 1997-2010 period.**

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 667 **Figure 5** Interannual average of SWAT results of the three climate change scenarios and the reference period T0 for  
 668 the Göschenalpsee feeding catchment a) for the T1 period 2021-20150 and b) for the T2 period 2070-2099. A 30 day  
 669 average window is applied.



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671 **Figure 6** Seasonal changes of the simulated with SWAT runoff of the Göscheneralpsee feeding catchment for the  
 672 **reference T0 and future periods T1 and T2 for all three climate change scenarios. The interannual mean of the months**  
 673 **a) May and June, b) July and August and c) September and October is taken.**

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