Glacier and runoff changes in the Rukhk catchment, upper Amu-Darya basin until 2050

Wilfried Hagg a,⁎, Martin Hoelzle b, Stephan Wagner c, Elisabeth Mayr a, Zbynek Klose d

a Department of Geography, Ludwig-Maximilians-University, Luisenstr. 37, 80333 Munich, Germany
b Department of Geosciences, University of Fribourg, Chemin du Musée 4, 1700 Fribourg, Switzerland
c Itos-GmbH, Bitzi 2555, 9642 Ebnat-Kappel, Switzerland
d Department of Water Resources and Environmental Modeling, Czech University of Life Science, Kamycka 1176, 16521 Prague, Czech Republic

A R T I C L E   I N F O

Article history:
Received 15 February 2012
Received in revised 22 February 2013
Accepted 13 May 2013
Available online 17 May 2013

Keywords:
glacier parameterization
hydrological model
Pamirs
Aral-Sea
runoff scenarios

A B S T R A C T

A conceptual hydrological model was set up in the upper Panj catchment, the main tributary of Amu-Darya river. After a manual calibration procedure involving model runs with different restrictions, the model reproduced both daily hydrographs of Tanimas river at the Rukhk gauging station (NSE = 0.86) and the snow water equivalent of the Irkt station (R² = 0.85) in a very satisfactory way. Based on two glacier inventories from the mid-20th century (WGI, World Glacier Inventory) and from 2003 (GLIMS, Global Land Ice Measurements from Space), a simple parameterization scheme based on steady state conditions was applied to infer the ice volumes and glacier areas for these different time periods. Assuming temperature rises of 2.2 °C and 3.1 °C, which mark the extreme values of regional climate scenarios, the same method was used to extrapolate glacierization to the year 2050. The results show that these temperature rises will reduce the current glacier extent of 431 km² by 36% and 45%, respectively.

To assess future changes in water availability, the hydrological model input was modified according to the regional climate scenarios and the resulting glacier changes. The use of an elevation distributed deglaciation pattern is a clear improvement over methods used previously, where the impact on runoff was tested by excluding either the lower half or the total glacier area. The runoff scenarios reveal only a slight reduction in annual runoff, because the glacier area decrease is almost balanced out by enhanced melt rates. However, there is an important seasonal shift of water resources from summer to spring, unfavorably affecting agriculture and irrigation in the lowlands.

© 2013 Elsevier B.V. All rights reserved.

1. Introduction

In many mountain ranges at mid-latitudes, glaciers represent important water reservoirs on several time scales. They are nourished by solid precipitation and release the water again through melting. In climate zones where accumulation and ablation occur at different times of the year, this leads to a seasonal delay of runoff (Jansson et al., 2003). When glacier mass balance is disturbed in one direction over a longer period, the glacier volume changes. In the case of a prolonged phase of mass loss as currently observed on most mountain glaciers, additional water is released and contributes to streamflow. Lambrecht and Mayer (2009) determined the portion of this excess melt in Austrian catchments. From 1969 to 1998, they found annual values of 1.5–9% of total discharge, depending on the relative glacier cover. In individual months, this fraction reached 20% in moderately glaciated basins. A prolonged mass loss gradually lowers the amount of excess water until the glaciers disappear completely and runoff is controlled by precipitation and snow-melt alone.

In climates with relatively high precipitation amounts, the contribution of glacier melt to total runoff is high only in heavily glacierized sub-basins. Weber et al. (2010) modeled the present-day fraction of ice melt in the upper Danube catchment and determined a value of 35% for a high Alpine head watershed (Vent, glacierization: 35%). As the relative glacier cover decreases downstream, the portion of ice melt also drops sharply and reaches 2% after the confluence of the Inn river with the Danube at Passau (glacierization: 0.5%). In summer months the respective values for the two gauges are 50% and 6–7% (Weber et al., 2009). The ice melt signal is quickly superimposed by high precipitation sums at the northern margin of the Alps that turn the glacial runoff regime into a nivo-pluvial one.

In reverse, this means that the relative importance of glacier melt is highest in regions where the ablation season is dry and/or where the mountains are surrounded by arid lowlands. A showcase of such conditions can be found in the basin of the Amu-Darya river, the main tributary to the Aral Sea. In a worldwide comparison of the hydrological significance of mountains, Viviroli and Weingartner (2004) classified the Amu-Darya as one of the rivers with the highest share of...
mountain discharge (>90%), which is defined as the portion of total runoff which is formed in the mountainous part of a catchment. According to Froebrich and Kayumov (2004), about 80% of the annual runoff of Amu-Darya river is formed by snow- and glacier melt in the Pamir mountains. Kaser et al. (2010) estimated the contribution potential of seasonally delayed glacier meltwater to total runoff in 18 large river systems on all continents and introduced a population impact index to quantify the potential human dependence on glacier melt. They found that in the Aral Sea basin, this value is by far the highest of all catchments investigated. A population of more than 10 million would suffer from water shortages if the glaciers disappeared. Careless consumption and misuse of water resources make the Aral Sea basin a well-known area of substantial water problems (UNESCO, 2009). The growing demand for water and water use conflicts along the transboundary river increase the need for interdisciplinary approaches to sustainable land and water use in the future (e.g. Martius et al., 2009). This extreme climatological and societal situation makes the region especially interesting for water balance studies. Whereas many authors in the past worked in lower parts of the Amu-Darya basin where the water is mainly used (e.g., Conrad et al., 2007; Bekchanov et al., 2010), there is a lack of investigations in the headwaters. This might be due to logistic obstacles to accessing the glacier region of the Pamir, to political instability and to the sparse data availability, which restricts scientists to very simple methodology. In remote high-mountain environments, conceptual hydrological models are widely used due to their limited data requirements. They simulate the hydrological cycle by use of calibrated and lumped models are widely used due to their limited data requirements. They simulate the hydrological cycle by use of calibrated and lumped models (Le Meur et al., 2004) and the input requirement for fully coupled models increases strongly and processing time is in order of magnitudes higher than for the parameterizations. We use a parameterization scheme based on Haeberli and Hoelzle (1995) to predict elevation-distributed glacierization for the mid-21st century. This glacier model was set up with the help of past and recent glacier inventories (WGMS, 1989; Raup et al., 2007) and digital elevation models (ASTER-GDEM).

The overall aim of this contribution is to estimate future hydrological conditions for the middle of this century, a time horizon important for water resource planning. This requires a classical scenario-led (“top–down”) impact assessment, where regional climate scenarios determine the input of hydrological models which subsequently quantify future consequences. An alternative approach would have been “bottom–up”, where allowed consequences are defined first and in a second step, the likelihood of future conditions to be within the defined margin is determined with an ensemble of regional climate scenarios (e.g. Prudhomme et al., 2010). Such studies are used to assist vulnerability management and to create adaption strategies (Carter et al., 2007) and are often preferred by policy and decision makers, but they are not suitable to estimate the range of possible reactions or the most probable hydrological responses.

2. Regional setting

Amu-Darya is one of the two inflows into the Aral Sea; 68% of total runoff in the Aral Sea basin is formed in its catchment (UNEP, 2006). While almost 90% of the total annual discharge of 78 km³ originates from the mountain areas of Tajikistan and Afghanistan (UNEP, 2006), most of the water consumption takes part in the highly arid zones of the downstream riparians (Turkmenistan, Uzbekistan). From Kerki to Nukus, Amu-Darya loses almost all of its water, mainly by artificial use. According to the Interstate Commission for Water Coordination in Central Asia, up to 4 mill. km² of land in the basin are irrigated (ICWC, 2010), for the most part still using the outdated and inefficient infrastructure from the Soviet period. Another major source of water loss is the Karakum canal which diverts 8–12 km² per year from the Aral Sea Basin to Turkmenistan (Glantz, 2005). As a result, Aral Sea has lost 90% of its volume from 1960 to 2006 (Micklin, 2007).

Amu-Darya is formed at the confluence of the rivers Vakhsh and Panj, the latter contributing 986 m³/s or 60% to total annual runoff. According to the World Glacier Inventory (WGMS, 1989) prepared in the mid-20th century, the Panj catchment has a glacier coverage of 3913 km², corresponding to a glacierization of 3.4%. A comparison of data from the GLIMS project (Zeltbyhina, 2005) with the World Glacier Inventory (WGI) revealed that from the mid 20st century until 2003, glacier area in the whole Panj basin decreased by 8.2% only (Hoelzle et al., 2010), equaling a mean rate of less than 2% per decade.

Rukkh is a village and hydrological post at the Tanimas river, which becomes the river Bartang at the confluence with the Murghab river (Fig. 1). The Rukkh catchment, which is investigated in this study is 4306 km² in size, it stretches from 2650 to 6940 m asl and has a glacier
cover of 431 km² (10%). In contrast to the adjacent Fedchenko glacier, which is completely debris-mantled on its lower parts, most glaciers in the Rukhk catchment are debris-free or have only marginal debris covers. Some glaciers on the orographic right side of Tanimas river have more heavily debris covered tongues, but they are small in number and size. Five years of daily discharge measurements (1985/86–1989/90, Tajik Met Service) are available at the Rukhk gauge. The hydrographs show a distinct glacial regime with maximum values in July and August. Meteorological data were taken from the Irkht station (3290 m asl) close to Lake Sarez (Fig. 1). This station is located outside the Rukhk catchment, but the close proximity (approx. 23 km from Rukhk) suggests that the data are still representative for the meteorological conditions inside the test basin. Irkht offers a valuable long-term series (since 1939) from this strongly continental mountain region. Mean monthly temperatures (1939–1994) range from −13.6 °C in January to 14.4 °C in July, the resulting inter-annual amplitude is 28 K. Precipitation is very low due to the leeward location with respect to the Central Pamir (Schiemann et al., 2008). Mean annual sums in the five years of discharge observations are only 139 mm, 72% of which occur from December to May, mostly in solid form. Only 12 mm or 8.6% fall in July to August, when 46% of annual runoff is formed and when the demand for irrigation water is highest. These numbers impressively demonstrate the importance of ice melt for runoff formation and seasonal redistribution.

The dominance of melt processes is a benefit for hydrological modeling, because they can be simulated satisfactorily even in data sparse regions. The relatively low horizontal variability and the well-known vertical lapse rates of air temperature make it feasible to extrapolate this parameter, also from a single meteorological station. Air temperature is a good index for melt, because it is correlated with all terms of the energy balance except for latent fluxes, which can disturb this relation in both directions (Ohmura, 2001). For this reason, melt rates in continental climates, where evaporation dominates throughout the year, are even stronger correlated to air temperature than in maritime settings, where both condensation and evaporation occur. This justifies the application of simple degree day models in highly continental regions such as the Amu-Darya basin. The dry climate on the lee side of the highest ranges further limits spatial differences in precipitation. The moisture content of the air is very low and even big relative differences in precipitation sums mean only small absolute differences compared to a more humid climate. This gives further confidence in simulating a relatively large catchment such as the Rukhk catchment with data from a single meteorological station.

Aglatseva et al. (2005) conducted regional climate modeling to assess future climate change in the mountainous part of the upper Amu-Darya basin. In order to partly moderate uncertainties of climate models, the averaged results of six General Circulation Models (HadCM3,ECHAM4,CSIRO-TR,GFDL-TR,GCM1-TR,CCSR-NIES) were used (source: WCRP CMIP3 multi-model dataset, Meehl et al., 2007b). Five emission scenarios (A1B,A1FI,B1,A2,B2, see Nakicenovic et al., 2000) formed the basis for characterizing climate changes by three time intervals in the future (2030,2050 and 2080) with respect to the baseline scenario 1961–1990. Statistical downscaling was performed by multivariate linear regression and is based on observational data from 22 stations in the case of air temperature, and 21 stations in the case of precipitation (Agaltseva et al., 2005). For mid-century, the mean warming in the mountain region ranges from 2.2 °C ("moderate scenario") to 3.1 °C ("hot scenario") and is evenly distributed over the year. To cover the full range of possible reactions, we selected these two extreme scenarios to estimate changes in glacierization and runoff. The corresponding precipitation changes for Fedchenko station (4156 m asl) are given in Table 1. Expected annual changes are low increases, ranging between 3 and 4%. Seasonal values reveal slight increases in all cases except for spring in the hot scenario (−3.7%), winter snow accumulation will be augmented by 3.0% (moderate scenario) and 8.5% (hot scenario).

![Fig. 1. Sketch map of the Rukhk catchment.](image-url)
3. Materials and methods

3.1. Processing of input data

The glacier parameterization uses World Glacier Inventory data of the investigation area from two time frames. The older data (1940s to 1960s) were derived mainly from aerial photographs, while the outlines (polygons) from the younger period (2000–2003) were updated from the GLIMS database (Zhelyzhina, 2005). Nearly all existing glacier polygons had to be corrected for their position and updated using self-digitized polygons, which were manually derived from LANDSAT satellite images from 2003, representing the same basis as in the GLIMS-database. The parameterization scheme was applied to some representative glaciers in both inventories and the results were extrapolated to the whole investigation area to provide an estimation of the total ice volume and the ice volume change for the period 2003 to 2050.

Topographic input for the hydrological model consists of a distribution of area by altitude and orientation classes, separately for glaciated and non-glaciated areas. For the Rukhkh catchment, this data were calculated from digital elevation models (ASTER-GDEM version 1) an elevation model with 30 m ground resolution and a product of METI and NASA (http://www.gdem.aster.erdac.or.jp/). The catchment was divided into 300 m elevation bands and into three orientation classes (South, North and East–West horizontal) (Fig. 2). Terrain with slopes smaller than 5° was classified as horizontal. The ASTER images used to generate the elevation models were taken in the 2000–2006 period, some 10–20 years after the available hydrometeorological data. Since no maps or elevation models from the 1980s are available and due to the lack of better options, this inaccuracy was accepted.

3.2. Glacier parameterization

The parameterization scheme developed by Haeberli and Hoelzle (1995) provides the possibility to analyze glacier inventory data stored in the databases of the World Glacier Monitoring service (WGMS) and the National Snow and Ice Data Center (NSIDC). These past inventories are mainly based on aerial photography, in contrast to the newer inventories, which rely on satellite images Paul (2010). Detailed information about the complete parameterization scheme can be found in Haeberli and Hoelzle (1995), Hoelzle and Haeberli (1995), Hoelzle et al. (2003, 2007) and Baumann and Winkler (2010). Here, a brief summary of the most important steps is provided. Only five measured input variables from the inventories were used, namely maximum glacier altitude (Hmax), mean altitude (Hmean), minimum altitude (Hmin), length (L0) and total surface area (F). All other variables are calculated or taken from measurements. Where specific equations are not given, the corresponding references are cited.

This approach considers step changes after full dynamic response and new equilibrium of the glacier have been achieved, when mass balance disturbance Δb leads to a corresponding glacier length change ΔL that depends on the original length L0 and the average annual mass balance (ablation) at the glacier terminus b. Therefore, this approach does not consider any transient changes. The term b is calculated as b = db/dh (Hmean − Hmin), where db/dh is the mass balance gradient:

\[ Δb = b \cdot ΔL/L₀. \]  (1)

Glacier thickness (h) is determined according to Eq. (2) (Paterson, 1994) where α is the slope, τ the basal shear stress, ρ the density of ice and g the acceleration due to gravity.

\[ h = \frac{τ}{ρ \cdot g \cdot \sin α}. \]  (2)

The dynamic response time tresp is calculated after Jóhannesson et al. (1989), where hmax is a characteristic ice thickness, usually taken at the equilibrium line where ice depths are near maximum, hmax is calculated as 2.5 h, as estimated from known thickness measurements on various Alpine glaciers worldwide (March, 2000; Bauder et al., 2003; Farinotti et al., 2009).

\[ t_{\text{resp}} = \frac{h_{\text{max}}}{b}. \]  (3)

Assuming a linear change of the mass balance from b to zero during the dynamic response, the average mass balance -b can be calculated according to Eq. (4). -b values are annual ice thickness change (meters of water equivalent per year) averaged over the entire glacier surface, which can be directly compared with values measured in the field. Although the method is quite simple, the results compare very well with long-term observations (Hoelzle et al., 2003). The factor nresp denotes the count of possible response times for each glacier within the considered time period.

\[ b = \frac{Δb}{2 \cdot n_{\text{resp}}}. \]  (4)

Due to his representative location and typical topographic features, Abramov glacier was considered a reference glacier for the Pamir-Alay during the Soviet era and also serves as such in this study. Although it lies 115 km outside the basin under investigation, it is the nearest glacier for which the required data are available. A mass balance gradient of 9.0 mm m⁻¹, derived from measured data at Abramov (Pertziger, 1996), was chosen for the entire sample. The value is in good agreement with the mass balance gradient of 8.8 mm derived by the HBV-ETH model. However, we decided to use the measured value as input for the glacier parameterization in order to keep the methods independent from each other. The parameterization was applied to 82 selected glaciers, where glacier information from both inventories could be compiled. The resulting differences of the modeled volumes were then extrapolated to an overall volume change for the whole test basin.

Mean specific mass balances were estimated using a relation found by Kuhn (1990), which states that a temperature change of +1 °C increases the equilibrium line altitude (ELA) by 170 m with an accuracy of ±50 m per degree Celsius. The corresponding change in mass (Δb) was calculated using the mass balance gradient (db/dh) after Eq. (5).

\[ δb = α \cdot \frac{dELA}{dT_{\text{air}}} \cdot ΔT_{\text{air}}. \]  (5)

where dELA/dTair describes the vertical shift of ELA per degree Celsius and integrates the change in all climate parameters, i.e. radiation, humidity, accumulation and air temperature, as well as feedback effects for any temperate glaciers (Kuhn, 1993).

The calculation of the -b value was done by taking into account each individual response time and multiples thereof (Eq. (4)). According to Eq. (5) an increase in temperature of 2.2 and 3.1 °C with a corresponding ELA shift of 374 m and 527 m would lead to a Δb of 3.37 m and 4.74 m, respectively.

**Table 1**

<table>
<thead>
<tr>
<th>Scenario</th>
<th>Winter</th>
<th>Spring</th>
<th>Summer</th>
<th>Autumn</th>
<th>Year</th>
</tr>
</thead>
<tbody>
<tr>
<td>Moderate scenario (+2.2 °C)</td>
<td>3.0</td>
<td>0.2</td>
<td>4.7</td>
<td>7.7</td>
<td>3.1</td>
</tr>
<tr>
<td>Hot scenario (+3.1 °C)</td>
<td>8.5</td>
<td>−3.7</td>
<td>6.6</td>
<td>9.2</td>
<td>3.9</td>
</tr>
</tbody>
</table>
The output of the used parameterization, mainly the calculated glacier area in 2050 after full adjustment to the new climate setting as presented above, serves as input for the HBV-ETH model.

3.3. Hydrological model

The HBV-ETH model (Braun and Renner, 1992) is a further development of the worldwide used HBV model (Bergström, 1976). Required input consists of topographical data (distribution of area and orientation classes at altitude, separately for glaciated and total area) and hydrometeorological data (daily values of air temperature and precipitation to drive the model, and runoff for calibration).

The first of three model components is the snow and glacier routine. Here, a threshold air temperature (T0) distinguishes solid precipitation from rainfall and both aggregational states of precipitation are corrected by factors (SCF, RCF) to account for gauge undercatch errors and for the representativeness of the weather station for the catchment. Precipitation is assumed to increase with elevation by a linear lapse rate (PGRAD). Melting of snow and ice is calculated with a temperature index approach using a seasonally variable degree day factor, forming a sinus curve with a maximum (CMAX) and a minimum (CMIN) value at the summer and winter solstices, respectively. The degree day factor (DDF) for each day is calculated by Eq. (6)

\[
DDF(i) = \frac{(CMA<sub>x</sub> - CMIN)}{2} \cos(2\pi i / \text{maxDay}) + CMIN + \frac{(CMA<sub>x</sub> - CMIN)}{2}
\]

where i is the day, numbered from 0 (on 21st June) until maxDay and maxDay is the number of days in a year minus 1.

Snow-melt (M) on plain terrain is estimated by the degree day approach shown by Eq. (7)

\[
M = DDF(i) \cdot (TM - T0)
\]

and occurs only if daily mean temperature (TM) is above the threshold temperature (T0).

Differences in incoming solar radiation due to topography are considered by the factor REXP (> 1). On south-facing slopes, melt is multiplied by REXP, while on north-facing slopes it is divided by the same value. To account for the lower albedo of ice compared to snow, melt on glaciated areas is multiplied by the factor RMULT = 1.2. While precipitation and melt in the snow and glacier routine are treated semi-distributed for different elevation bands and orientation classes, the following steps of the model are performed on a lumped basis for the whole catchment area. In our relatively large study area, this seems to be justified by the uniform landscape pattern due to the absence of vegetation.

The second model component is the soil routine. It receives the sum of melt-water and precipitation (RS) from the snow and glacier routine and computes actual evapotranspiration (EA) as a function of potential evaporation (EP) and the fill level of the soil moisture reservoir (SSM) after Eq. (10).

\[
EA = EP \cdot SS/M/LS
\]

LP is the particular fill level of the reservoir, below which EA is assumed to equal EP.

Potential evapotranspiration (EP) varies sinusoidally throughout the year with a minimum of zero on 30th January and a maximum (ETMAX) on 1st August according to Eq. (11)

\[
EP = 0.5 \cdot ETMAX \cdot \left[1 + \sin(j \cdot 2\pi / \text{maxDay})\right]
\]

where j is the day, numbered from 0 (on 2 May) until maxDay.

The third model component is the response function, where the remaining water is transformed into a field capacity (FC) and a coefficient (BETA) according to Eq. (12).

\[
DSUZ = RS \cdot SSM / FC \cdot BETA
\]

The response function has an upper and a lower storage and three outflows with different response times controlled by constants (k0, k1, k2). Quick runoff (Q0) from the upper storage is determined by Eq. (13).

\[
Q_0 = k_0 \cdot (SUZ - LUZ)
\]

and only appears if the fill level of the upper storage (SUZ) exceeds a certain threshold (LUZ). Intermediate outflow (Q1) is also released from the upper storage and is calculated accordingly (k1 SUZ). Percolation from the upper to the lower (ground water) storage is controlled by the
constant CPERC and from the lower storage, slow runoff (Q2) is formed as the product of k2 and the fill level of the lower storage (SLZ).

The three outflows are summed up and form total runoff.

3.4. Calibration of the hydrological model

The HBV-ETH model was calibrated manually using the model efficiency (NSE) after Nash and Sutcliffe (1970) and the percent bias (PBIAS) in runoff volume (Gupta et al., 1999) as objective functions. The ratio of the root mean square error to the standard deviation of measured data (RSR) was further used to assess model performance as proposed by Moriasi et al. (2007). The lower the root mean square error and RSR, the better the model performance. A perfect fit between modeled and observed hydrographs would yield a NSE of 1.0 and PBIAS and RSR of 0% and 0.0, respectively. The equations used were:

\[
\text{NSE} = 1 - \frac{\sum_{i=1}^{n} (Q_{m} - Q_{s})^2}{\sum_{i=1}^{n} (Q_{m} - \bar{Q})^2}
\]

\[
\text{PBIAS} = \frac{\sum_{i=1}^{n} (Q_{m} - Q_{s})}{\sum_{i=1}^{n} Q_{m}} \times 100
\]

\[
\text{RSR} = \sqrt{\frac{\sum_{i=1}^{n} (Q_{m} - Q_{s})^2}{\sum_{i=1}^{n} Q_{m}^2}}
\]

where \(Q_{m}\) is the observed discharge, \(Q_{s}\) the simulated discharge, \(n\) the total number of calculated time steps and \(\bar{Q}\) the total number of years.

Since both basin precipitation and glacier melt are difficult to simulate correctly, verification with measured mass balance data helps to avoid error compensation (Hagg et al., 2004; Schaefl and Huss, 2010; Mayr et al., 2013).

To check the plausibility of the calculated basin precipitation, we compared it with assimilated and gridded precipitation data sets. Reanalysis data are an important source of precipitation estimates in areas with few observations and have also been used in Central Asia before (e.g. Schär et al., 2004; Schemmann et al., 2008). A gridded (0.25°) precipitation dataset for Asia was constructed from interpolated observations within the APHRODITE project (Asian Precipitation – Highly-Resolved Observational Data Integration Towards Evaluation of the Water Resources). We used outputs from the ERA40 reanalysis (Uppala et al., 2005), the REMO regional climate model (Mannig et al., 2013–this issue) and APHRODITE precipitation products (Takashima et al., 2009) to check the basin precipitation resulting from our parameter settings (see calibration mode 2 below).

Direct mass balance observations are not available in the investigation area, but mass losses comparable to those in more maritime mountains were confirmed by repeat photographs in the Tanimas valley (Braun and Hagg, 2009) and by a GPS survey on Fedchenko glacier in 2009 (Lambrecht et al., 2010). As a consequence, the only frame we had to constrain parameters in terms of ice melt was to avoid strongly negative mass changes, balanced conditions, or a gain in glacier mass. To check if the model is able to reasonably describe accumulation and depletion of snow, we used daily snow height observations from the meteorological station Irkht. The snow heights were transformed into snow water equivalents using snow densities computed as a function of mean snow age, following the formulation of Douville et al. (1995). The numerical criterion for the simulation of the snow cover is the squared correlation coefficient (R²) between modeled snow water equivalents and snow water equivalents derived from observations.

In the HBV-ETH model, calibration of the free parameters is conducted in optimization runs, where two parameters are tested against each other simultaneously. For each parameter, a start value and an increment must be defined. Ten values of each parameter are tested in every optimization run. The results are two matrixes for NSE and PBIAS, each showing 100 values corresponding to each parameter value combination. A color code for the goodness of fit quickly visualizes if the optimum is within the parameter ranges or not. Subsequently, the modeler can narrow the value ranges by choosing smaller increments and a finer resolution of the values, maybe resulting in higher NSE or lower PBIAS values. This step can be repeated until the best value combination is found. Since highest NSE values are typically not reached with the same parameters as the lowest volume error, the modeler has to find a compromise here. This introduces a certain subjectivity, but also lets the modeler decide if his priority is more the shape of the hydrograph or the runoff volume, a question which can depend on the goal of the study. It is important to note that after each optimization, hydrographs are also interpreted visually. This enhances the understanding of the parameter values and their changes. The storage parameters (k0, k1, k2), for example, can be tuned very effectively by an optical assessment of the declining branches of the hydrograph.

In a next step, the optimization is performed with the next pair of parameters. The most sensitive parameters (RCF, SCF, TO, PGRA, TGRA, CMX) are calibrated first. In this context, it is useful to choose two parameters which are related to different processes (e.g. one controlling precipitation, the other controlling melt) to see if there is more than one optimum, meaning that the model is equivariant. In this case, parameter ranges need to be constrained by the help of other criteria (e.g. glacier mass balance, snow observations, assimilated precipitation data). After all free parameters are calibrated, the whole procedure is repeated with other combinations of parameter pairs. If the calibration is robust, the parameter values and also the numerical criteria (NSE and PBIAS) change only very slightly during this second round.

To ensure independent calibration and validation, the measured time-series is usually divided into two sets (Klemes, 1986). We divided the period of runoff observations into the two cooler (1988/89 and 1989/90) and the three warmer (1985/86–1987/88) years and used them in turn for calibration and validation, only judging the ability of the model to reproduce runoff. The results of this split-sample test are shown in Table 2. When compared to calibrating over the full series, splitting the 5 years leads to a slightly better performance in the calibration period and a slightly lower performance in the validation period. To avoid biases by the specific meteorological conditions in short periods,

<table>
<thead>
<tr>
<th>Criterion</th>
<th>Split sample 1</th>
<th>Split sample 2</th>
<th>Full series</th>
</tr>
</thead>
<tbody>
<tr>
<td>NSE</td>
<td>0.94</td>
<td>0.87</td>
<td>0.92</td>
</tr>
<tr>
<td>PBIAS (%)</td>
<td>3.1</td>
<td>7.0</td>
<td>-4.6</td>
</tr>
<tr>
<td>RSR</td>
<td>0.26</td>
<td>0.36</td>
<td>0.29</td>
</tr>
</tbody>
</table>
we decided to use the complete time-series for further calibration purposes.

To spin up the model, we looped it through the hydrological year 1987/88 for 10 consecutive iterations before we started the model runs that were analyzed for calibration. The model soon reached a dynamic equilibrium, after two iterations the changes were already marginal.

3.5. Simulation of future runoff

Runoff scenarios were generated by modifying the meteorological model input by a constant temperature shift throughout the year, according to the evenly distributed warming rates found by Agaltseva et al. (2005). Precipitation was adjusted seasonally according to Table 1.

Glacier changes (Fig. 3) were considered in the area–altitude distribution and are based on the glacier parameterization scheme presented in Section 3.2.

4. Results

4.1. Deglacierization

According to the glacier parameterization, temperature rises of 2.2 °C and 3.1 °C until 2050 reduce the current glacier extent by 36% and 45%, respectively. The deglacierization in the elevation classes of the HBV-ETH model is depicted in Fig. 3.

In the two lowest intervals below 4400 m asl, glaciers practically disappear. In the following three intervals up to 5000 m asl, glacier area is reduced by more than 50%, from 5000 to 5300 m asl by 28–36% and above 5300 m asl there are only very minor changes. Regarding the current distribution, the largest absolute losses (56–74 km²) will occur between 4700 and 5000 m asl.

The availability of areal changes by altitude is a real improvement over former studies, where deglacierization was simulated by “cutting” the lowest glacier parts. In order to determine the differences between these two approaches, additional model runs have been performed with the old deglacierization procedure (“cutting lowest”), where glacier area remains unchanged above 5000 m asl and is completely deleted below 4700 m asl. In the elevation band 4700–5000 m asl, glacier cover is reduced to 39 km² (+2.2 °C) and 4 km² (+3.1 °C).

The parameterization scheme used to derive future glacierization does not allow us to differentiate between the orientation classes used in the hydrological model. It was assumed that relative deglacierization is similar for all orientations.

4.2. Runoff simulations in the period with observations

We conducted four calibration modes (1–4), the model performance and the values of the water balance terms are compiled in Table 3. When the model is tuned by runoff only (1), the highest NSE that could be reached was 0.91. In this calibration mode, any parameter value within very broad ranges was accepted. For example, the best rain correction factor (RCF) found was 0.22, meaning that if precipitation is assumed liquid, only 22% of the measured value was taken into account. Solid precipitation was multiplied by 1.2 (SCF). Such parameter values cannot be explained by physical processes, but they can result if the calibration is conducted freely from any restrictions. Using this parameter set, glacier mass balance is unrealistically high. If the basin value of −192 mm is related to the glacier area, mean net balances of almost −2 m/a result. Furthermore, the snow cover at Irkht cannot be reproduced, indicating that meteorological variables are not extrapolated correctly.

In a second calibration mode (2), parameter values were restricted to narrower ranges which are more realistic according to our modeling experience in Central Asia. In the case of precipitation correction factors, for example, this means that values should be above 1 unless it seems plausible that the meteorological station receives more precipitation than the basin average due to an especially high and/or wind exposed location. Furthermore, the basin precipitation should be closer to the values derived from APHRODITE (290 mm), ERA40 (360 mm) and REMO (439 mm). These values vary considerably, but indicate that a reasonable basin value should at least be in the order of 300 mm. Other demands for calibration mode (2) were less negative glacier balances and a better simulation of the recorded snow cover at Irkht. By reducing ice and snow melt, the glacier mass balance could be reduced to −105 mm and the snow properties can be simulated well. This is illustrated by the correlation between monthly snow water equivalents (Fig. 4b) and by the proper reproduction of the daily course (Fig. 4a). In all years, the timing of snowfalls, the onset of melt and the duration of the snow cover is reproduced very well. HBV-ETH tends to slightly overestimate snow water equivalents, because the model can neither account for snow deflation by wind nor for sublimation losses. Therefore, the snow water equivalent cannot be reduced during periods with air temperatures below zero. As expected, these additional calibration criteria reduce the ability of the model to mimic runoff: the best compromise found has a mean Nash–Sutcliffe value of 0.86. The reduced glacier melt is mainly balanced by increased precipitation.

In order to test the ability of the model system to compensate for errors in basin precipitation, an additional calibration run (3) was
performed. The target was to double basin precipitation, which was realized by increasing the values of RCF, SCF and PGRAD (to 1.5, 1.5 and 15.5%, respectively). The other parameters were tuned again and the results proved that still relatively high model efficiencies of 0.80 can be achieved. However, glacier mass balance becomes positive, which is highly unrealistic. Additionally, evapotranspiration is increased by almost 90% compared to the previous mode.

In a last calibration run (4), the glacier mass balance is forced to reach realistic values similar to those in (2), while the doubled basin precipitation is kept constant. This resulted in a clear divergence of the hydrographs, the goodness of fit dropped to the mean NSE of 0.65 and PBIAS was highest with 33%. Even more water than in mode (3) needs to be evaporated, yielding mean basin values that exceed runoff by a factor of more than two.

All further model applications were conducted with the parameter values of calibration mode (2), they are listed in Table 4 and the hydrographs are displayed in Fig. 5.

Since daily variations in runoff (which are assessed by the Nash–Sutcliffe criterion) are not essential for the evaluation of future water availability, the ability of the model to reproduce monthly runoff was tested (Fig. 6). The comparison of modeled and measured monthly runoff shows a high coefficient of determination ($R^2 = 0.96$) and proves that the model performs well on a monthly time step.

While the use of daily data enables a differentiated evaluation of the model performance, the hydrological regime and the temporal distribution of water resources become more clearly visible if monthly means are calculated (Fig. 7). The only month with a minor deviation among the mean values is August, where the model underestimates runoff by 9% compared to the observational data. The mean (maximum) annual share of ice melt in total runoff is 32% (43%), not considering if this water flows through the ground water body on its way to the basin outlet or not. The highest contributions of ice melt in the Tanias river are reached in August, with mean (maximum) shares of 68% (88%).

### 4.3. Runoff scenarios for 2050

The monthly hydrographs of the scenarios are displayed in Fig. 8. Numerical changes are listed in Table 5. For the sake of a better comparability, absolute runoff values have been transformed into relative changes.

It is important to note that annual runoff remains stable, while the seasonal distribution shifts towards a water surplus in May–June and a shortage in July–September. In August and September, both scenarios

![Fig. 4. Snow water equivalents modeled by HBV-ETH and derived from snow height observations at Irkht (3290 m a.s.l.). a: Daily course of snow water equivalents in an example year (1987/88), b: Monthly snow water equivalents during the calibration period (1985/86–1989/90).](image-url)
show an identical decrease in monthly streamflow compared to the current situation.

In the “cutting lowest” mode, total summer (Jun–Sep) runoff is underestimated by 8.1% in the moderate scenario and by 5.4% in the hot scenario (Table 6).

The HBV-ETH model simulates not only runoff, but also other terms of the water balance. Their average values are listed in Table 7. According to the presented combination of climate modeling, glacier area parameterization and runoff modeling, annual runoff will decrease by 5% until 2050. All values in Table 7 relate to the entire catchment. Therefore, the glacier storage changes have to be regarded as basin values. Considering the respective glacierization (calibration period: 10%, moderate scenario: 6.4%, hot scenario: 5.5%), the specific glacier net balances are −1100 mm/a for the calibration period, −1484 mm/a for the moderate scenario and −1782 mm/a for the hot scenario.

The mean shares of glacial runoff at the basin outlet, calculated as ice melt minus evapotranspiration of melted ice are 21% (+2.2 °C) and 24% (+3.1 °C).

5. Discussion and conclusion

The calibration exercise with four different modes has shown that the terms of the water balance can become entirely wrong if the model is tuned by runoff only. Including glacier mass balance and snow measurements as additional criteria significantly reduces parameter uncertainty, especially if parameter values are kept within a realistic array.

The water balance in the calibration period (Table 7) is rather special for a high alpine catchment, which can be attributed to the unique combination of vast glacierization and strong aridity. Our modeled annual values for precipitation (295 mm/a) and runoff (215 mm/a) are very low, but this scale is confirmed by the latest hydrometeorological observations in Tajikistan from 1990–2005 (Finaev, 2006). Even for Central Asia, these values are exceptional. Outside the Pamir, such arid conditions can only be found in the most central parts of the Tian Shan, where mean precipitation sums of 296 mm/a were recorded at an elevation of 3614 m asl (Fujita et al., 2011). Earlier applications of the HBV-ETH model (Hagg et al., 2006) have yielded basin values for precipitation of more than 500 mm in the Eastern Tian Shan (Glacier No. 1) and of around 1000 mm in the Western (Oigaing) and Northern (Tuyuksu) Tian Shan. These values match well with long-term observations (Kotlyakov, 1997; Glazyrin, 2006; Severskiy, 2006) and follow the general gradient from dry interior mountain parts to a wetter climate at the peripheral ranges (Aizen et al., 1997; Sorg et al., 2012). Modeled evapotranspiration in the Rukhk catchment (187 mm) is within the relative broad range of values reported from the more maritime European Alps, where estimates range from 120 mm/a (Braun et al., 2007) to 259 mm/a (Kuhn, 2000). Other studies from Central Asia (e.g. Kuzmichenok, 2006) confirmed evapotranspiration sums similar to those in more humid environments. This can be explained by the low water availability. While potential evaporation is surely much higher than in a moist climate, actual evaporation is strongly reduced by the low availability during the dry summers. Glacier mass balance (−105 mm/a) is in the same order of magnitude as in the Alps, which was a requirement for model calibration. Since precipitation is strongly reduced in the Pamir, this term has a greater relative importance: in our study, the glacier storage changes are 1/3 of basin precipitation or 1/4 of total water input to the system (sum of precipitation and glacier wastage). In the Alps, typical corresponding values would be 1/5 and 1/6 (Braun et al., 2007). This comparison once more shows how much this region depends on melt water and how much glacier wastages currently contribute to water supply.

Based on climate and glacier projections for 2050, the hydrological model generates annual runoff values which are only a little below the ones in the 1980s. The slightly increased precipitation is
overcompensated by a moderate increase in evapotranspiration and a small decrease in glacier runoff. The areal glacier reduction is almost compensated by enhanced melt rates in the warmer atmosphere. In other words, glaciers will melt more intensely but over a smaller area.

The runoff increase in spring and early summer is stronger in the hot scenario, it can be attributed to an earlier and intensified snow-melt. In July, the glacier area decrease in the two scenarios overcompensates enhanced melt rates due to warming, resulting in lower runoff values compared to the calibration period. The stronger decline in streamflow occurs in the hot scenario, where peak runoff is shifted from July to June, turning the glacier runoff regime into a nivo-glacial.

Konovalov (1985) has determined the fraction of glacial runoff, defined as the share of firm- and ice-melt in total runoff, by regional melt modeling for numerous catchments in Central Asia. Kudara is a hydrological post at Tanimas river some 15 km above Rukkh, with a glacialization of 13.8% (compared to 10% at Rukkh). Here, Konovalov (1985) determined a mean annual glacial contribution of 37% which is very well to our result of 32%. These fractions are very high compared to more humid climates. In the European Alps, similar contributions of annual glacier melt to total runoff were very high compared to more humid climates. In the European Alps, similar contributions of annual glacier melt to total runoff were reported for a glacialization of 35% (Weber et al., 2010).

Reducing glacier cover by simply cutting off the lowest areas leads to an exaggerated loss of the most melt-relevant glacier parts. An elevation-distributed deglaciation also affects higher areas and as is also the case in nature. The error between the two approaches is not large and it diminishes with deglaciation, because the absolute differences between the approaches also decrease with total glacier area. However, the elevation distributed deglaciation approach is a step towards more realistic runoff scenarios.

Uncertainties arise at all stages of the modeling process, beginning with the climate models (Meehl et al., 2007a). Especially in mountain regions, the confidence in the projected precipitation change remains weak. The GCMs fail in part to model precipitation accurately across varying elevations (Agaltseva et al., 2005). Differing precipitation sums and intra-annual changes would influence future runoff in two main ways: solid precipitation contributes to accumulation in winter and reduces ablation in summer, both affecting the glacier mass balance and consequently the future glacier extent. Liquid precipitation that falls on ice and rock surfaces runs off immediately and in this arid region even small absolute changes would have a large impact on runoff formation. In this context, it is important to state that the presented results are no predictions, but have to be regarded as estimates of possible changes. More uncertainty is caused by the use of only one meteorological station and by the assumption that parameter values found in the calibration period are still valid in the future. Possible changes in the relation between air temperature and melt rates could yield substantial errors in the estimates of future glaciarization and streamflow response. Not considering the effect of supraglacial debris cover overestimates both the area retreat and future melt rates. On the catchment scale, these errors partly compensate for each other and are generally small due to the very limited extent of supraglacial debris.

The mean runoff reduction in July and August of approximately 25% would raise major problems for irrigation and drinking water supply. Irrigation capacities are already limited by summer runoff and a further decrease will intensify the situation in the lowlands. Furthermore, evapotranspiration losses in the downstream regions will be significantly higher than in the headwaters treated in this study, which will further limit water availability in summer. The transnational character of the Amu-Darya requires multilateral planning of adaption strategies and impact mitigation to secure the livelihood of the population.

We trust that the applied methodology is a practical but also reliable way to estimate future water availability in regions with sparse data. The incorporation of real estimates of future glacier areas and the consideration of elevation distributed area changes in particular improved the validity of the results and the power of the approach to provide scenario estimates.

Acknowledgments

The authors thank Susan Braun-Clarke for editing the English in a very professional way. Astrid Lambrecht provided glacier shapes which are displayed in Fig. 1. Birgit Mannig, Mark Reyes and Reinhard Schiemann gave valuable advice and assistance for modeled and gridded climate data. The work was supported by the “Research Network Water in Central Asia (CAWa)”, funded by the German Federal Foreign Office. Elisabeth Mayer was supported by the Deutsche Forschungsgemeinschaft (DFG, German Research Foundation) in the frame of the AKSU-TARIM project bundle (project HA 5061/2-1). ECMWF ERA-40 data used in this study have been provided by ECMWF. The valuable comments of Markus Konz, Nina Köplin and an anonymous reviewer are greatly acknowledged.