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Citation: Ragetti, Silvan, Immerzeel, Walter and Pellicciotti, Francesca (2016) Contrasting climate change impact on river flows from high altitude catchments in the Himalayan and Andes Mountains. Proceedings of the National Academy of Sciences of the United States of America, 113 (33). pp. 9222-9227. ISSN 1091-6490

Published by: National Academy of Sciences

URL: <http://dx.doi.org/10.1073/pnas.1606526113>
<<http://dx.doi.org/10.1073/pnas.1606526113>>

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1 Classification: PHYSICAL SCIENCES, Environmental Sciences

2 **Contrasting climate change impact on river flows from high**
3 **altitude catchments in the Himalayan and Andes Mountains**

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1 **Abstract**

2 Mountain ranges are world's natural water towers and provide water resources for millions of people.
3 Yet, their hydrological balance and possible future changes in river flow remain poorly understood
4 because of high meteorological variability, physical inaccessibility and the complex interplay between
5 climate, cryosphere and hydrological processes. Here we use a state-of-the art glacio-hydrological
6 model informed by data from high altitude observations and CMIP5 climate change scenarios to
7 quantify the climate change impact on water resources of two contrasting catchments vulnerable to
8 changes in the cryosphere. The two study catchments are located in the Central Andes of Chile and in
9 the Nepalese Himalaya in close vicinity of densely populated areas. Although both sites reveal a
10 strong decrease in glacier area, they show a remarkably different hydrological response to projected
11 climate change. In the Juncal catchment in Chile runoff is likely to sharply decrease in the future and
12 the runoff seasonality is sensitive to projected climatic changes. In the Langtang catchment in Nepal,
13 future water availability is on the rise for decades to come with limited shifts between seasons. Owing
14 to the high spatio-temporal resolution of the simulations and process complexity included in the
15 modelling the response times and the mechanisms underlying the variations in glacier area and river
16 flow can be well constrained. The projections indicate that climate change adaptation in Central Chile
17 should focus on dealing with a reduction in water availability, whereas in Nepal preparedness for flood
18 extremes should be the policy priority.

19 **Significance Statement**

20 Changes in the hydrology of high altitude catchments may have major consequences for downstream
21 water supply. Based on model projections with a higher spatio-temporal resolution and degree of
22 process complexity than any previous intercontinental comparative study we show that the impacts of
23 climate change cannot be generalized. These impacts range from a high climatic sensitivity,
24 decreasing runoff and significant seasonal changes in the Central Andes of Chile to increasing future
25 runoff, limited seasonal shifts but increases in peak flows in the Nepalese Himalaya. This study
26 constrains uncertainty about response times and mechanisms controlling glacier and runoff response to
27 climate and sets a benchmark for process-based modeling of the climate change impact on the
28 hydrology of high altitude catchments.

1 \body

2 **Introduction**

3 Glaciers and seasonal snow cover change their water storage capacity under a warming climate. When
4 glacier mass balances are negative, glaciers contribute additional water to rivers. However, negative
5 mass balances lead to a reduction in glacier volume and area, which eventually reduces the total
6 meltwater from glaciers. A warming climate may therefore lead to either rising or decreasing river
7 flows, depending on the state of glacier retreat (1). Due to variability in glacier characteristics and
8 differences in climate, high altitude regions respond differently to climatic changes across the world
9 (2). Factors which potentially retard glacier response to global warming are the presence of thick,
10 insulating layers of supraglacial-debris (3), topographic shading due to extreme topography (4) or
11 steep and large headwalls, which cause the glacier's response to be dictated by local avalanche
12 processes (5). Factors which increase melt rates are decreases in albedo due to mineral dust and black
13 carbon depositions (6) and prolonged melting seasons (7). Furthermore, glacier sensitivity to global
14 warming depends strongly on precipitation seasonality (7). Temperature increase leads to a change in
15 precipitation phase over summer accumulation-type glaciers (such as in the Central and Eastern
16 Himalaya), whereas this is not necessarily the case with winter precipitation. The rate of change of net
17 glacier mass loss depends also on initial glacier hypsometry (8): glacier mass balances become very
18 negative if the equilibrium line altitude (ELA) rises above the elevations where most initial glacier
19 area is located. This seems the case currently over most of Central Europe and at Low Latitudes,
20 whereas in Alaska and High Mountain Asia a large portion of glacier area is above the current-day
21 ELA (8). Given the multitude of control factors, often acting at relatively small scales, current
22 projections of global glacier change models do not constrain the mechanisms underlying the variations
23 in river flow from high altitude catchments well (2, 8, 9). It is therefore uncertain if such models can
24 correctly capture the response times and the magnitudes of future changes. Modeling studies at a high
25 spatio-temporal resolution and degree of process complexity are crucial to assess the impact of climate
26 change on the hydrological balance of high altitude catchments and to refine the projections of the
27 coarser scale models (10, 11).

1 Here we compare the impact of climate change on the magnitude and timing of catchment runoff for
2 the two climatically contrasting high altitude areas in the Himalayan and Andean Mountains with a
3 model configuration that substantially exceeds the spatio-temporal resolutions and degree of
4 complexity of previous intercontinental comparative studies (2, 8, 9). Whereas global scale modeling
5 studies work with routines and parameters that are assumed to be generally valid, the simulations
6 presented here are based on model setups which have been thoroughly evaluated in three companion
7 papers (12–14) to correctly reproduce present observed catchment runoff, meteorological processes,
8 snow cover variability and glacier mass balances. Model construction and calibration benefited from
9 extensive short-term field campaigns, which allowed overcoming the problem of data scarcity typical
10 of poorly accessible high altitude regions.

11 We make projections of twenty-first century runoff and glacier changes using the state-of-the art
12 glacio-hydrological model TOPKAPI-ETH (12–15), the newest climate change scenarios (Table S2)
13 and performing simulations at a spatial resolution of 100 m and hourly time steps. The high spatial
14 resolution ensures that processes acting at relatively small scales such as gravitational snow
15 redistribution or topographic shading are taken into account. The high temporal resolution allows to
16 reproduce the strong sub-daily variability of melt rates and to accurately determine the altitude of the
17 freezing level and its diurnal cycles. Short-term variability of temperature has a large effect on the
18 duration of melting episodes (16). Simulations provided here allow for an examination of future runoff
19 extremes, which have been little investigated to date in the Himalayan (17) and Andean Mountains.

20 Uncertainty in model projections is quantified by taking into account a large subset of climate models,
21 representative of the uncertainty range of the original 5th Climate Model Intercomparison Project
22 (CMIP5) ensemble. We analyze two main scenarios (representative concentration pathways, RCPs)
23 that may lead to a radiative forcing of 4.5 W/m^2 (RCP45) and 8.5 W/m^2 (RCP85) in 2100. The two
24 RCPs chosen are the climate scenarios for which the largest number of climate models are available.
25 For each RCP and each watershed twelve global climate models (GCMs) are considered and
26 stochastically downscaled to the location of the study catchments. The stochastic approach allows
27 accounting for the uncertainty due to the natural variability of the climate system (15, 18) and is

1 particularly useful for an analysis of extremes since the multiple model runs for each GCM-RCP
2 combination allow a statistical assessment of peak events which occur with a low frequency.

3 Two study catchments (the Juncal catchment in the Central Andes of Chile and the Upper Langtang
4 catchment in the Nepalese Himalaya) are selected for their proximity to densely populated areas
5 (Figure 1) and their representativeness in terms of meteorology, hydrology and glacier processes for
6 the same elevations in the two large climatic regions (Figure S2). The Central Andes of Chile are
7 amongst the most vulnerable regions to changes in the cryosphere, since meltwater from glaciers and
8 snow is critical to maintain river flows during the dry summers (10, 19) and because of increasing
9 water demand in the downstream regions (20). Large glaciers exist due to high accumulation area
10 ratios in El Niño years and local effects like topographic shading and accumulation through
11 avalanches and wind (19, 21, 22). In the Nepalese Himalaya, glaciers influence the hydrology of the
12 Upper Ganges Basin (17), which is marked by intense competition on water resources because of high
13 population density and extensive irrigation needs (23). However, more than 70% of the annual
14 precipitation occurs during the warmest period of the year, which dampens the relative contribution of
15 snow- and icemelt on the annual water yield (12). The large glaciers with low reaching tongues in this
16 climate are characterized by heavy debris cover that protects them from melting (12, 24).

17 **Results**

18 The temperature projections reveal similar warming trends for the Langtang and Juncal region (Figure
19 1a and b). Between 2010 and 2100 mean temperatures are projected to increase between 1.1°C and
20 3.4°C for RCP45 and between 3.4°C and 7.2°C for RCP85 for the two regions, depending on the
21 GCM. Precipitation projections show a negative trend for the Central Andes of Chile and a contrasting
22 positive trend for the Nepalese Himalaya, and in both cases those trends are pronounced for the
23 RCP85 scenario. However, the uncertainty in precipitation projections is generally large, in particular
24 for Central Chile, where projected end-of-century precipitation changes vary between -50% and +15%
25 in comparison to the beginning of the century (Figure 1a).

1 **Fate of Glaciers.** Both catchments respond with a clear decreasing trend in glacier area to the changes
2 in climate. Multi model median results indicate 53% (RCP45) or 70% (RCP85) glacier area loss in the
3 Juncal region between 2001-2010 and 2091-2100 (Figure 3). For the same period the simulations
4 indicate a decrease in glacierized area by 35% (RCP45) or 55% (RCP85) for the Langtang region
5 (Figure 3). Debris-covered glacier area (representing 27% of the total glacier area in Langtang) is less
6 sensitive to the changes in climate and decreases only by 25% (RCP45) or 33% (RCP85) until the end
7 of the century. It is typical of heavily debris-covered glaciers with stagnant low-gradient termini that
8 fronts are more stable (25). High air temperatures prevailing on the low reaching tongues and
9 enhanced melting on exposed ice cliffs and beneath supra-glacial lakes can substantially mitigate the
10 shielding effect of supraglacial debris (26–29). However, in the Langtang region melt rates of debris-
11 covered ice are much lower than of non-debris-covered ice (12). In the long run this leads to glacier
12 tongues that are disconnected from the accumulation areas (Figure 1iii and iv, Figure S12).

13 **River Flow Projections.** The Upper Langtang catchment exhibits increasing runoff in the first half of
14 the 21st century for all climate scenarios (between +15% and +70% in comparison to 2001-2010,
15 Figure 2). Under RCP45 conditions the simulations indicate a possible runoff decrease after the
16 peaking in 2051-2060. Under RCP85 conditions Langtang River runoff remains relatively stable
17 during the second half of the 21st century, but with an increasing GCM uncertainty. In contrast to the
18 projections for the Nepalese Himalaya, decreasing water availability is a plausible scenario for the
19 Central Andes of Chile. The multi-model median projections for the Juncal catchment indicate stable
20 runoff volumes until 2021-2030 and then a steadily decreasing trend. Multi-model median runoff
21 under RCP45 conditions declines by 40% between 2001-2010 and 2091-2100, and by 62% under
22 RCP85 conditions. While the described trends are consistent across GCMs for the Upper Langtang
23 catchment, the runoff trends projected for the Juncal catchment differ between climate models. Here,
24 the projections by five out of twelve climate models lead to a stable runoff response until the end of
25 the century rather than a steady decline.

26 **Glacier Contributions to River Flow.** Icemelt from glaciers represents roughly one third of total
27 simulated streamflow during the reference period (2001-2010) in Langtang, and one fifth in Juncal

1 (Figure 2). We show that total icemelt is on a rising limb in Langtang at least until 2041-2050 and
2 starts to decrease again after 2051-2060 (Figure 3). These results confirm the findings by a previous
3 modelling study (30). In Juncal, however, total icemelt was already beyond its tipping point at the
4 beginning of the 21st century according to our simulations. This contrasting response to climate
5 warming can be explained by differences in the elevation distribution of the glaciers in the two
6 regions. In Juncal, many glaciers are melting up to the highest elevations already during the reference
7 period. Increasing melt rates due to higher air temperatures cannot compensate the continuous loss of
8 glacier area. In the Langtang catchment, large sections of the glaciers at high elevations are currently
9 not exposed to melt, but will be in the future, thus compensate for the loss of glacier area at lower
10 elevations (Figure S11). Total glacier area contributing to melt in Langtang peaks in 2051-2060
11 (Figure 3b and c). In Juncal, glacier area contributing to melt decreases steadily (Figure 3e and f).

12 The decline in total icemelt in Langtang after 2051-2060 is compensated by a very pronounced
13 increase in total rainfall (Figure S10c) which is due to the projected increase in precipitation by most
14 climate models. During the second half of the century the contribution of icemelt to total water inputs
15 decreases by 10% while the contribution of rainfall increases by 10% according to median projections
16 (Figure 2). Other studies have argued that in a monsoon dominated climate, the glacier contribution to
17 water availability is minor (2, 31). Our results confirm that the effect of runoff decrease due to glacier
18 decline is dampened by high relative (increasing) contributions of rain. However, if the contributions
19 of glaciers in the Langtang catchment remained constant and were not decreasing after mid-century
20 (Figure 3), runoff would further increase rather than decrease or remain constant (Figure 2 a and b). In
21 the Juncal catchment the decline in total icemelt until the end of the century (Figure 3d) explains
22 30-40% of annual runoff declines (Figure 2c and d).

23 **Future runoff seasonality.** The changes in the hydrology of the Juncal catchment are such that peak
24 runoff during the austral summer disappears gradually (Figure 4). By the end of the century the
25 differences between winter, spring and summer runoff become very small (under RCP85 conditions).
26 The climate projections by five GCMs (RCP85) lead to an earlier peak runoff shifted by two to three
27 months, while the simulations associated to a majority of climate models (both RCP45 and RCP85)

1 indicate an anticipation of the seasonal runoff peak by one month. The shifts in the seasonality can be
2 explained by earlier snowmelt onset and a change of phase from liquid to solid precipitation. Total
3 rainfall amounts will likely increase (Figure S10g) - in spite of mostly decreasing precipitation. In the
4 Upper Langtang catchment, on the other hand, the changes in climate lead to almost no changes in
5 runoff seasonality (Figure 4), since the timing of the monsoon period and of the main melting season
6 essentially remains unaffected by climate change. However, the hydrological regime changes to a
7 more rainfall dominated regime, which results in a faster transition of precipitation to runoff and a
8 higher susceptibility to peak flows.

9 **Future runoff extremes.** Our simulations indicate a substantial increase in the magnitude of peak
10 runoff events for the Upper Langtang catchment, particularly under RCP85 conditions (Figure 5a,
11 Figures S8-S9). The model runs indicate increasing annual maximum daily flows with increasing peak
12 flows especially during the post-monsoon season. Annual maximum daily flows with recurrence
13 intervals of 10 years or peak October flows are projected to increase by 100% until the end of the
14 century (Figure 5a, Figure S9). The higher peak flows can be related to precipitation state changes
15 from solid to liquid during extreme precipitation events related to cyclonic disturbances (32, 33). This
16 result is therefore not affected by the common limitations of climate models and bias correction
17 techniques in predicting new precipitation extremes outside empirical ranges (34), or by the large
18 uncertainties in precipitation projections. In Juncal, simulated future annual peak flows remain
19 approximately constant in magnitude (Figure 5b), due to shifts in seasonal flows (Figure 4 and Figure
20 S9) in spite of total annual runoff decreases (Figure 2). For future dry years (cumulative probability of
21 annual runoff less than 50%) the model projects no reduction in water availability in comparison to the
22 present in Langtang, but 10% - 30% stronger decreases than for average years in Juncal (Figure S8).

23 **Discussion**

24 Our findings point to the necessity of identifying coping strategies to a reduction in water availability
25 in the Central Andes of Chile. All simulations for the Juncal region indicate a significant decrease in
26 summer runoff until the end of the century, and more than 90% of the runs a decrease in total annual

1 runoff. In contrast to the Central Andes of Chile, there are no signs of decreasing water availability in
2 the Nepalese Himalaya. Here, future research will have to focus on changes in the return periods of
3 water related natural hazards and assess the downstream impact of strong increases in post-monsoon
4 peak flows from high elevation catchments as projected by our study.

5 Global glacier change models (2, 8, 9) project strong glacier area and volume decreases until the end
6 of the century for the Southern Himalaya (e.g. ref. 8, RCP45, -70% glacier area by the year 2100),
7 which also leads to projections of continuously decreasing glacier runoff (2). However, such
8 projections are representative of very large regions and therefore the comparability with the outputs of
9 the present study is limited. Yet, both Juncal and Langtang are located near urban centers and
10 therefore have a high interest for stakeholders of water resources. Here, large scale studies are not
11 suitable to inform policy making regarding climate change adaptation, given also the large differences
12 between our projections and the regionally averaged outputs of global models. Apart from different
13 glacier ensembles considered, the much stronger glacier retreat projections by global models for the
14 Southern Himalaya in comparison to our study could be explained by the non-consideration of the
15 insulating effect of supraglacial debris and/or the application of simple temperature index models
16 which tend to be oversensitive to temperature fluctuations (35).

17 There are also important differences in the outputs of previous detailed climate change impact
18 modeling studies focusing on the Upper Langtang catchment (30, 36). Water released by glaciers plays
19 a more important role for future runoff from the Upper Langtang catchment than suggested by two
20 previous studies (30, 36) and as a consequence future runoff is not entirely governed by the future
21 precipitation trend. Our study projects stagnating (RCP85) or slightly decreasing (RCP45) runoff after
22 mid-century, while refs. 30 and 36 indicate a consistent increase throughout the 21st century (both
23 RCPs). Here, differences in model structure and between calibration strategies explain the differences
24 in model projections. The consideration of a separate term for shortwave radiation in the melt
25 algorithms used by this study assures that the relationships between air temperature and melt are
26 robust in time and thus suitable for long-term modelling (35). The higher spatio-temporal resolution
27 and degree of process complexity allows representing the heterogeneous snow- and ice melt patterns

1 controlled by solar radiation or supraglacial debris. Glacio-hydrological models in high altitude
2 regions are also particularly sensitive to assumptions about temperature gradients (Figure S5, refs. 12–
3 15). While in this study temperature gradients are based on measured detailed information about air
4 temperature distribution, previous catchment scale models in the region use substantially different
5 approaches to determine temperature lapse rates. Parameterizations were obtained through model
6 calibration (30, 36) which may lead to equifinality and error compensation (37). Other studies use
7 reanalysis products (17, 38) or remotely sensed data (39), which are however more uncertain data
8 sources and available only at relatively coarse resolutions.

9 A distributed model characterized by a high degree of complexity such as TOPKAPI-ETH requires a
10 good knowledge of internal states to correctly represent the basin-internal dynamics (40). We therefore
11 highlight the utility of in-situ data to inform glacio-hydrological models for climate change impact
12 assessments. We argue that more such data collection efforts are required for climate change impact
13 assessment across climates and regions. A smart integration of field based studies with the catchment
14 scale at key basins in the world has great potential and can reveal the full magnitude of the impacts of
15 climate change on mountain water resources.

16 **Materials and Methods**

17 **Main model:** We use a fully distributed, high-resolution (100 m, hourly time step) glacio-hydrological
18 model, TOPKAPI-ETH (12–15), in two glacierized catchments of the Himalaya and the Andes. The
19 model simulates all major hydrological and glaciological processes at the watershed scale. As input
20 variables the model requires distributed fields of air temperature, precipitation and cloud transmittance
21 factors.

22 **Snow- and icemelt:** Snow- and ice ablation is modelled with an enhanced temperature-index (ETI)
23 approach where melt in each grid cell is the sum of a temperature-dependent term and a shortwave
24 radiation dependent term (41). The approach considers therefore the fully distributed shortwave
25 radiation balance which is calculated from the position of the sun relative to the considered grid cell,
26 topographic shading, cloudiness and surface albedo. Snow albedo decreases over time if air

1 temperatures are above the melting point (42) and constant ice albedo is used by the ETI to calculate
2 icemelt once snow is depleted.

3 Supraglacial debris: Sub-debris ice melt is calculated using a debris-ETI approach where ablation rates
4 decrease in function of debris thickness. Debris thickness is reconstructed by an inverse Ostrem
5 approach (12), relating observed surface elevation change rates to debris thicknesses according to an
6 Ostrem curve established using a debris-energy balance model (43). Stagnant glacier area is chosen for
7 this purpose to limit the error due to ice thickening in response to compressive flow regimes. Surface
8 elevation changes are obtained from differencing two high resolution digital elevation models from
9 UAV surveys (44). Debris thicknesses of areas without information about surface elevation changes
10 are parameterized based on the position relative to the snout and the presence of supra-glacial lakes
11 identified from Landsat ETM+ multispectral data (12). The spatial density of supraglacial lakes is used
12 as a proxy for spatial variations in debris thickness because these features (in combination with
13 supraglacial cliffs whose presence correlates with the presence of lakes) have been shown to greatly
14 influence downwasting rates of debris-covered glacier ice (27, 28, 45).

15 Avalanching: Gravitational snow redistribution is taken into account with a mass conservation
16 algorithm based on slope dependent maximum snow holding depths (46). If the threshold depth is
17 exceeded, snow is moved to the next model grid cell downwards.

18 Glacier dynamics: We assume a linear increase of glacier thinning rates below a given threshold
19 elevation and a constant ice thickness above, which guarantees that ice accumulated above the
20 equilibrium line altitude is redistributed to lower elevations, and that declines in glacier area are
21 delayed by flow dynamics (47). The approach is mass-conserving which implies that thinning rates
22 depend on annually accumulated ice volumes. Snow to ice conversion takes place where snow remains
23 on a glacier for longer than a year. Ice depths remain constant if the ice mass balance of a given
24 glacier for a given year is zero. Initial ice thicknesses are estimated using the model GlabTop2 (48),
25 which is essentially a slope-dependent ice thickness estimation approach. The glacier dynamics
26 algorithm is executed once a year at the end of each hydrological year.

1 Routing, groundwater and evapotranspiration: Reference evapotranspiration (RET) depends on
2 incoming shortwave radiation, albedo and air temperature. Vegetation coefficients determine the ratio
3 between potential evapotranspiration and RET. Actual evapotranspiration depends on available soil
4 moisture content within a superficial soil layer calculated internally by TOPKAPI-ETH. A second soil
5 layer accounts for runoff originating from percolation to deeper soil and into fractured bedrock. Water
6 routing in is based on the kinematic wave concept, whereby subsurface flow, overland flow due to
7 saturation excess and channel flow is represented by non-linear reservoir differential equations (49).

8 GCM data: A subset of twelve GCMs (Table S2) is selected for the RCP45 and RCP85 scenarios from
9 the latest climate model ensemble generated for the Intergovernmental Panel on Climate Change
10 (IPCC) fifth assessment report provided through phase five of the Climate Model Intercomparison
11 Project (CMIP5). In order for the subset to be representative of the uncertainty range of the original
12 CMIP5 ensemble, the selected GCMs are sampled randomly from clusters of multivariate
13 characteristics regarding the changes in temperature and precipitation projected for both the Andes and
14 the Himalaya (Figure S2). We use daily precipitation and monthly temperature climate model outputs.

15 GCM downscaling: We use a stochastic approach to downscale the climate models (15, 18) to
16 generate hourly input time series until 2100. Temperature and precipitation data are downscaled to the
17 location of the closest permanent meteorological station in each region. Each station provides at least
18 nine years of data to determine the observed climate statistics for the reference period (2001-2010).
19 The statistics that are reproduced in the stochastic precipitation time series are the mean, variance, no-
20 rain probability, skewness and autocorrelation at daily scale. Disaggregation of daily to hourly
21 precipitation is performed on the basis of empirical data. Air temperature is subdivided into a
22 stochastic and a deterministic component. The temperature statistics of the deterministic part provided
23 by the climate models are the monthly mean and the monthly standard deviation. An autoregressive
24 integrated moving average (ARIMA) model is used for the stochastic generation of hourly air
25 temperatures (50). The daily cycle is provided by historical data and is added to the stationary
26 stochastic time series before destandardization with the monthly mean and standard deviation.

27 Differences between station observations and climate model outputs are first corrected through a non-

1 linear parametric bias correction method (51, 52). All climate statistics are evaluated separately for
2 each month of the year and each decade from 2001 to 2100. Temperature lapse rates and precipitation
3 gradients are used to distribute the downscaled climate data to every catchment grid cell and are
4 estimated on the basis of historical station data, taking into account their spatial and temporal
5 variability (13, 53). Daily cloud factors are sampled randomly from the available historical station
6 data, differentiating between days with strong/medium/weak or no daily precipitation, and between
7 spring-summer records and autumn-winter records in order to preserve the coupling of daily
8 cloudiness with precipitation intensity. We force the model with 20 members of the stochastic input
9 data ensemble generated for each GCM-RCP combination.

10 Model calibration and validation: The TOPKAPI-ETH setups used in this study have been established
11 and thoroughly evaluated in three companion papers (12–14). Model parameters were calibrated with
12 a multi-step, multi-objective approach (12, 14). Carefully-planned field data collection during two
13 ablation seasons at both study sites constituted the basis for the setup and validation of the model (12,
14 14). For the purpose of preventing error compensation, each parameter was estimated on the basis of
15 data which directly represent the corresponding physical process. A complete list of data sets and
16 related model components and parameters is provided by Table S1. Observed catchment runoff was
17 used for model validation. In both regions measured daily runoff from validation periods of one to
18 three years could be reproduced by the model with a Nash-Sutcliffe efficiency (NSE) higher than 0.85
19 (Table S3). In addition to the relatively short calibration and validation periods used in the companion
20 papers (12–14) we use nine years of available runoff data at either site to validate the seasonal runoff
21 cycle as simulated by the stochastic runs for the reference period (Figure S4, NSE Juncal: 0.97, NSE
22 Upper Langtang: 0.90).

23 Extreme values analysis: To assess possible changes in hydrological extremes we looked at the
24 changes in different quantiles of annual runoff (Figure S8 and S9) and at the changes in the magnitude
25 of annual peak daily runoff with different recurrence intervals (Figure 5). Twenty stochastic
26 realizations of each GCM-RCP combination lead to 200 annual values per decade that are used for this
27 analysis.

1 Units of all water balance components in this study are expressed in millimeter water equivalents
2 (volumes of water divided by catchment area). Details on the methodology and data sets can be found
3 in the Supplementary Information.

4 **Acknowledgements.** The authors are grateful to all the people who assisted with the data collection.
5 This project was funded by SNF UNCOMUN (“Understanding Contrasts in High Mountain
6 Hydrology in Asia”). Fieldwork was partially supported by a United States Agency for International
7 Development (USAID) High Mountain Glacier Watershed Programs Climber-Scientist Grant
8 (CCRDCS0010) project. This study was partially funded by the Swiss State Secretariat for Education
9 and Research (grant CJRP-1003), by the European Research Council (ERC) under the European
10 Union’s Horizon 2020 research and innovation program (grant agreement No 676819), and by the
11 South Asia Research Hub of the UK Department for International Development (DFID). The authors
12 are very grateful to the HKH Cryosphere Monitoring Project implemented by the International Centre
13 for Integrated Mountain Development (ICIMOD) and supported by the Norwegian Ministry of
14 Foreign Affairs. We thank Andrés Rivera for providing data on the ice thickness of Juncal Norte
15 Glacier and Thomas Mendlik for processing the CMIP5 model ensemble for the climate model
16 selection.

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1 **Figure Legends**

2 **Figure 1. Study catchments and glacier and debris covered area at the beginning of the century. a and b, The boxes**
3 **showing the 25th, 50th and 75th percentile of the GCM ensemble projections indicate relative changes in projected**
4 **decadal mean annual precipitation and temperature with respect to the reference period (2001-2010). i-iv, Ensemble-**
5 **median projected glacier area (RCP45 and RCP85) by the year 2100.**

6 **Figure 2. Future catchment runoff and composition of total water input. The figure shows the median of all**
7 **simulations for each decade, RCP and study area. The error bars represent the 80% confidence interval about the**
8 **climate model ensemble.**

9 **Figure 3: Future evolution of icemelt contribution to runoff, future glacier area and glacier area contributing to**
10 **icemelt. The lines show the median of all simulations for each decade, RCP and study area. Sub-debris icemelt (a) is a**
11 **component of total icemelt in the Upper Langtang catchment. The debris-covered glacier area and non-debris-**
12 **covered glacier area contributing to icemelt are shown separately (b and c). The bars represent the 80% confidence**
13 **interval about the climate model ensemble (also shown for debris area contributing to icemelt but too close to the line**
14 **to see).**

15 **Figure 4. Simulated seasonal runoff cycles and water input composition. The median results of all simulations are**
16 **represented. Error bars represent the 80% confidence interval of the climate model uncertainty. The composition of**
17 **water inputs is shown for the reference period 2001-2010 (a, d), for the period 2091-2100 and RCP45 projections (b, e)**
18 **and for the period 2091-2100 and RCP85 projections (c, f). Solid lines represent the median of all simulations for the**
19 **reference period (2001-2010) and two future periods (2051-2060, 2091-2100),**

20 **Figure 5. Simulated annual maximum daily runoff corresponding to average recurrence intervals of 2 years (ARI 2),**
21 **10 years (ARI 10) and 100 years (ARI 100) for both RCPs (RCP45 and RCP85). The figure shows the median outputs**
22 **of the stochastic runs for the reference period (2001-2010) and for two future decades (2051-2060 and 2091-2100).**