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

Silvan Ragettli, Walter W. Immerzeel, and Francesca Pellicciotti

*Institute of Environmental Engineering, ETH Zurich, 8093 Zurich, Switzerland; bDepartment of Physical Geography, Utrecht University, 3508 TC Utrecht, The Netherlands; and cDepartment of Geography, Northumbria University, Newcastle upon Tyne NE1 8ST, United Kingdom

Edited by Andrea Rinaldo, Laboratory of Ecohydrology, Ecole Polytechnique Federale Lausanne, Lausanne, Switzerland, and approved June 21, 2016 (received for review April 26, 2016)

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

significance

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

Author contributions: F.P. designed research; S.R., W.W.I., and F.P. performed research; S.R. contributed new reagents/analytic tools; S.R. analyzed data; and S.R., W.W.I., and F.P. wrote the paper.

The authors declare no conflict of interest.

This article is a PNAS Direct Submission. Freely available online through the PNAS open access option.

1To whom correspondence should be addressed. Email: ragettli@fu.baug.ethz.ch.

This article contains supporting information online at www.pnas.org/lookup/suppl/doi:10.1073/pnas.1606526113/-/DCSupplemental.

Significance

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

9222–9227 | PNAS | August 16, 2016 | vol. 113 | no. 33
(12–15), the newest climate change scenarios (SI Appendix, Table S2), and performing simulations at a spatial resolution of 100 m and hourly time steps. The high spatial resolution ensures that processes acting at relatively small scales such as gravitational snow redistribution or topographic shading are taken into account. The high temporal resolution allows to reproduce the strong subdaily variability of melt rates and to accurately determine the altitude of the freezing level and its diurnal cycles. Short-term variability of temperature has a large effect on the duration of melting episodes (16). Simulations provided here allow for an examination of future runoff extremes, which have been little investigated to date in the Himalayan (17) and Andean Mountains.

Uncertainty in model projections is quantified by taking into account a large subset of climate models, representative of the uncertainty range of the original Coupled Model Intercomparison Project Phase 5 (CMIP5) ensemble. We analyze two main scenarios [representative concentration pathways (RCPs)] that may lead to a radiative forcing of 4.5 W/m² (RCP45) and 8.5 W/m² (RCP85) in 2100. The two RCPs chosen are the climate scenarios for which the largest number of climate models are available. For each RCP and each watershed, 12 global climate models (GCMs) are considered and stochastically downscaled to the location of the study catchments. The stochastic approach allows accounting for the uncertainty due to the natural variability of the climate system (15, 18) and is particularly useful for an analysis of extremes because the multiple model runs for each GCM-RCP combination allow a statistical assessment of peak events that occur with a low frequency.

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

**Results**

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

![Fig. 1. Study catchments and glacier and debris-covered area at the beginning of the century. (A and B) The boxes showing the 25th, 50th, and 75th percentile of the GCM ensemble projections indicate relative changes in projected decadal mean annual precipitation and temperature with respect to the reference period (2001–2010). (i–iv) Ensemble-median projected glacier area (RCP45 and RCP85) by the year 2100.](image-url)
River Flow Projections. The Upper Langtang catchment exhibits increasing runoff in the first half of the 21st century for all climate scenarios (between +15% and +70% in comparison with 2001–2010; Fig. 2). Under RCP45 conditions, the simulations indicate a possible runoff decrease after the peaking in 2051–2060. Under RCP85 conditions, Langtang River runoff remains relatively stable during the second half of the 21st century, but with an increasing GCM uncertainty. In contrast to the projections for the Nepalese Himalaya, decreasing water availability is a plausible scenario for the Central Andes of Chile. The multimodel median projections for the Juncal catchment indicate stable runoff volumes until 2021–2030 and then a steadily decreasing trend. Multimodal median runoff under RCP45 conditions declines by 40% between 2001–2010 and 2091–2100, and by 62% under RCP85 conditions. Although the described trends are consistent across GCMs for the Upper Langtang catchment, the runoff trends projected for the Juncal catchment differ between climate models. Here, the projections by 5 out of 12 climate models lead to a stable runoff response until the end of the century rather than a steady decline.

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

Glacier Contributions to River Flow. Ice melt from glaciers represents roughly one-third of total simulated streamflow during the reference period (2001–2010) in Langtang, and one-fifth in Juncal (Fig. 2). We show that total ice melt is on a rising limb in Langtang at least until 2041–2050 and starts to decrease again after 2051–2060 (Fig. 3). These results confirm the findings by a previous modeling study (30). In Juncal, however, total ice melt was already beyond its tipping point at the beginning of the 21st century according to our simulations. This contrasting response to climate warming can be explained by differences in the elevation distribution of the glaciers in the two regions. In Juncal, many glaciers are melting up to the highest elevations already during the reference period. Increasing melt rates due to higher air temperatures cannot compensate the continuous loss of glacier area. In the Langtang catchment, large sections of the glaciers at high elevations are currently not exposed to melt, but will be in the future, and thus compensate for the loss of glacier area at lower elevations (SI Appendix, Fig. S11). Total glacier area contributing to melt in Langtang peaks in 2051–2060 (Fig. 3 B and C). In Juncal, glacier area contributing to melt decreases steadily (Fig. 3 E and F).

The decline in total ice melt in Langtang after 2051–2060 is compensated by a very pronounced increase in total rainfall (SI Appendix, Fig. S10C), which is due to the projected increase in precipitation by most climate models. During the second half of the century, the contribution of ice melt to total water inputs decreases by 10%, whereas the contribution of rainfall increases by 10% according to median projections (Fig. 2). Other studies have argued that, in a monsoon-dominated climate, the glacier contribution to water availability is minor (2, 31). Our results confirm that the effect of runoff decrease due to glacier decline is dampened by high relative (increasing) contributions of rain. However, if the contributions of glaciers in the Langtang catchment remained constant and were not decreasing after midcentury (Fig. 3), runoff would further increase rather than decrease or remain constant (Fig. 2 A and B). In the Juncal catchment, the decline in total ice melt until the end of the century (Fig. 3D) explains 30–40% of annual runoff declines (Fig. 2 C and D).

Future Runoff Seasonality. The changes in the hydrology of the Juncal catchment are such that peak runoff during the austral summer disappears (Fig. 4). By the end of the century, the differences between winter, spring, and summer runoff become very small (under RCP85 conditions). The climate projections by five GCMs (RCP85) lead to an earlier peak runoff shifted by 2–3 mo, whereas the simulations associated to a majority of climate models (both RCP45 and RCP85) indicate an anticipation of the seasonal runoff peak by 1 mo. The shifts in the seasonality can be explained by earlier snowmelt onset and a change of phase from liquid to solid precipitation. Total rainfall amounts will likely increase (SI Appendix, Fig. S10G)—despite mostly decreasing precipitation. In the Upper Langtang catchment, on the other hand, the changes in climate lead to almost no changes in runoff seasonality (Fig. 4), because the timing of the monsoon period and of the main melting season essentially remains unaffected by climate change. However, the hydrological regime changes to a more rainfall-dominated regime, which results in a faster transition of precipitation to runoff and a higher susceptibility to peak flows.

Future Runoff Extremes. Our simulations indicate a substantial increase in the magnitude of peak runoff events for the Upper Langtang catchment, particularly under RCP85 conditions (Fig. 5 A and SI Appendix, Figs. S8 and S9). The model runs indicate increasing annual maximum daily flows with increasing peak flows.

![Fig. 2. Future catchment runoff and composition of total water input. The figure shows the median of all simulations for each decade, RCP, and study area (A and B: Upper Langtang catchment; C and D: Juncal catchment). The error bars represent the 80% confidence interval about the climate model ensemble.](http://www.pnas.org/cgi/doi/10.1073/pnas.1606526113)
especially during the postmonsoon season. Annual maximum daily flows with recurrence intervals of 10 y are projected to increase by 100% until the end of the century (Fig. 5A). The higher peak flows can be related to precipitation state changes from solid to liquid during extreme precipitation events related to cyclonic disturbances (32, 33). This result is therefore not affected by the common limitations of climate models and bias correction techniques in predicting new precipitation extremes outside empirical ranges (34), or by the large uncertainties in precipitation projections. In Juncal, simulated future annual peak flows remain approximately constant in magnitude (Fig. 5B), due to shifts in seasonal flows (Fig. 4 and SI Appendix, Fig. S9) despite total annual runoff decreases (Fig. 2). For future dry years (cumulative probability of annual runoff less than 50%), the model projects no reduction in water availability in comparison with the present in Langtang, but 10–30% stronger decreases than for average years in Juncal (SI Appendix, Fig. S8).

**Discussion**

Our findings point to the necessity of identifying coping strategies to a reduction in water availability in the Central Andes of Chile. All simulations for the Juncal region indicate a significant decrease in summer runoff until the end of the century, and more than 90% of the runs a decrease in total annual runoff. In contrast to the Central Andes of Chile, there are no signs of decreasing water availability in the Nepalese Himalaya. Here, future research will have to focus on changes in the return periods of water-related natural hazards and assess the downstream impact of strong increases in postmonsoon peak flows from high elevation catchments as projected by our study.

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

There are also important differences in the outputs of previous detailed climate change impact modeling studies focusing on the Upper Langtang catchment. Water released by glaciers plays a more important role for future runoff from the Upper Langtang catchment than suggested by two previous studies (30, 36), and as a consequence future runoff is not entirely governed by the future precipitation trend. Our study projects stagnating (RCP85) or slightly decreasing (RCP45) runoff after midcentury, whereas refs. 30 and 36 indicate a consistent increase throughout the 21st century (both RCPs). Here, differences in model structure and between calibration strategies explain the differences in model projections. The consideration of a separate term for shortwave radiation in the melt algorithms used by
this study assures that the relationships between air temperature and melt are robust in time and thus suitable for long-term modeling (35). The higher spatiotemporal resolution and degree of process complexity allows representing the heterogeneous snowmelt and ice melt patterns controlled by solar radiation or supraglacial debris. Glaciological hydrological models in high-altitude regions are also particularly sensitive to assumptions about temperature gradients (refs. 12–15; SI Appendix, Fig. S5). While in this study temperature gradients are based on measured detailed information about air temperature distribution, previous catchment scale models in the region use substantially different approaches to determine temperature lapse rates. Parameterizations were obtained through model calibration (30, 36), which may lead to equifinality and error compensation (37). Other studies use reanalysis products (17, 38) or remotely sensed data (39), which are, however, more uncertain data sources and available only at relatively coarse resolutions.

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

Materials and Methods

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

Snowmelt and Ice Melt. Snow and ice ablation is modeled with an enhanced temperature index (ETI) approach, where melt in each grid cell is the sum of a temperature-dependent term and a shortwave radiation-dependent term (41). The approach considers therefore the fully distributed shortwave radiation balance, which is calculated from the position of the sun relative to the considered grid cell, topographic shading, cloudiness, and surface albedo. Snow albedo decreases over time if air temperatures are above the melting point (42), and constant ice albedo is used by the ETI to calculate ice melt once snow is depleted.

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

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

Glacier Dynamics. We assume a linear increase of glacier thinning rates below a given threshold elevation and a constant ice thickness above, which guarantees that ice accumulated above the ELA is redistributed to lower elevations and that declines in glacier area are delayed by flow dynamics (47). The approach is mass conserving, which implies that thinning rates depend on annually accumulated ice volumes. Snow-to-ice conversion takes place where snow remains on a glacier for longer than a year. Ice depths remain constant if the ice mass balance of a given glacier for a given year is zero. Initial ice thicknesses are estimated using the melt duration (27, 28), which is parameterized based on slope-dependent ice thickness estimation approach. The glacier dynamics algorithm is executed once a year at the end of each hydrological year.

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

GCM Data. A subset of 12 GCMs (SI Appendix, Table S2) is selected for the RCP45 and RCP85 scenarios from the latest climate model ensemble generated for the Intergovernmental Panel on Climate Change Fifth assessment report provided through CMIP5. In order for the subset to be representative of the uncertainty range of the original CMIP5 ensemble, the selected GCMs are sampled randomly from clusters of multivariate characteristics regarding the changes in temperature and precipitation projected for both the Andes and the Himalaya (SI Appendix, Fig. S2). We use daily precipitation and monthly temperature climate model outputs.

GCM Downscaling. We use a stochastic approach to downscale the climate models (15, 18) to generate hourly input time series until 2100. Temperature and precipitation data are downscaled to the location of the closest permanent meteorological station in each region. Each station provides a dataset at least 9 y of data to determine the observed climate statistics for the reference period (2001–2010). The statistics that are reproduced in the stochastic precipitation time series are the mean, variance, no-rain probability, skewness, and autocorrelation at daily scale. Disaggregation of daily to hourly precipitation is performed on the basis of empirical data. Air temperature is subdivided into a stochastic and a deterministic component. The temperature statistics of the deterministic part provided by the climate models are the monthly mean and the monthly standard deviation (SD). An autoregressive integrated moving average model is used for the stochastic generation of hourly air temperatures (50). The daily cycle is provided by historical data and is added to the stationary stochastic time series before standardization with the monthly mean and SD. Differences between station observations and climate model outputs are first corrected through a nonlinear parametric bias correction method (51, 52). All climate statistics are evaluated separately for each month of the year and each decade from 2001 to 2100. Temperature lapse rates and precipitation gradients are used to distribute the downscaled climate data to every catchment grid cell and are estimated on the basis of historical station data, taking into account their spatial and temporal variability (13, 53). Daily cloud factors are sampled randomly from the available historical station data, differentiating between days with strong/medium/weak or no daily precipitation, and between spring-summer records and autumn-winter records to preserve the coupling of daily cloudiness with precipitation intensity. We force the model with 20 members of the stochastic input data ensemble generated for each GCM-RCP combination.

Fig. 5. Simulated annual maximum daily runoff corresponding to average recurrence intervals of 2 y (ARI 2), 10 y (ARI 10), and 100 y (ARI 100) for both RCPs (RCP45 and RCP85) and study areas (A: Upper Langtang catchment; B: Juncal catchment). The figure shows the median outputs of the stochastic runs for the reference period (2001–2010) and for two future decades (2051–2060 and 2091–2100).
Model Calibration and Validation. The TOPKAPPI-ETH setups used in this study have been established and thoroughly evaluated in three companion papers (12, 14). Model parameters were calibrated with a multiobjective optimization algorithm (12, 14). Carefully planned field data collection during two ablation seasons at both study sites constituted the basis for the setup and validation of the model (12, 14). For the purpose of preventing error compensation, each parameter was estimated on the basis of data that directly represent the corresponding physical process. A complete list of datasets and related model components and parameters is provided in SI Appendix, Table S1. Observed catchment runoff was used for model validation by comparing modelled with measured daily runoff from validation periods of 1–3 years with the model with a Nash–Sutcliffe efficiency (NSE) higher than 0.85 (SI Appendix, Table S3). In addition to the relatively short calibration and validation periods used in the companion papers (12–14), we use 9 y of available runoff data at each site to validate the seasonal runoff cycle as simulated by the stochastic runs for the reference period (SI Appendix, Fig. S4; NSE Juncal: 0.97; NSE Upper Langtang: 0.90).

Extreme Values Analysis. To assess possible changes in hydrological extremes, we look at the changes in different quantiles of annual runoff with the changes in the magnitude of annual peak daily runoff with this analysis.

GCM-RCP combination lead to 200 annual values per decade that are used for the scenario analysis. Details on the methodology and datasets can be found in SI Appendix.

ACKNOWLEDGMENTS. We are grateful to all the people who assisted with the data collection. We thank Andréas Rivera for providing data on the ice thickness of Juncal Norte Glacier and Thomas Mendlik for processing the CMIP5 model ensemble for the climate model selection. We are very grateful to the Hindu Kush Himalaya Hydrology Monitoring Project implemented by the International Centre for Integrated Mountain Development and supported by the Norwegian Ministry of Foreign Affairs. Fieldwork was partially supported by the US Agency for International Development High Mountain Glacier Watershed Programs Climber-Scientist Grant CRDC50010. This study was supported by the Swiss National Science Foundation (‘Understanding Contrasts in High Mountain Hydrology in Asia’ project), with contributions from the Swiss State Secretariat for Education and Research (Grant CJP-R-1003), the European Research Council under the European Union’s Horizon 2020 Research and Innovation Program (Grant 676819) and the South Asia Research Hub of the UK Department for International Development.