

1 Reconciling high altitude precipitation in the upper Indus 2 basin with glacier mass balances and runoff

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9 10 **Abstract**

11 Mountain ranges in Asia are important water suppliers, especially if downstream climates are
12 arid, water demands are high and glaciers are abundant. In such basins, the hydrological
13 cycle depends heavily on high altitude precipitation. Yet direct observations of high altitude
14 precipitation are lacking and satellite derived products are of insufficient resolution and
15 quality to capture spatial variation and magnitude of mountain precipitation. Here we use
16 glacier mass balances to inversely infer the high altitude precipitation in the upper Indus basin
17 and show that the amount of precipitation required to sustain the observed mass balances of
18 large glacier systems is far beyond what is observed at valley stations or estimated by gridded
19 precipitation products. An independent validation with observed river flow confirms that the
20 water balance can indeed only be closed when the high altitude precipitation is up to a factor
21 ten higher than previously thought. We conclude that these findings alter the present
22 understanding of high altitude hydrology and will have an important bearing on climate
23 change impact studies, planning and design of hydropower plants and irrigation reservoirs and
24 the regional geopolitical situation in general.

25 **1 Introduction**

26 Of all Asian basins that find their headwaters in the greater Himalayas, the Indus basin
27 depends most strongly on high altitude water resources (Immerzeel et al., 2010; Lutz et al.,
28 2014) The largest glacier systems outside the polar regions are found in this area and the
29 seasonal snow cover is the most extensive of all Asian basins (Immerzeel et al., 2009). In

1 combination with a semi-arid downstream climate, a high demand for water owing to the
2 largest irrigation scheme in the world and a large and quickly growing population, the
3 importance of the upper Indus basin (UIB) is evident (Immerzeel and Bierkens, 2012).

4 The hydrology of the UIB ($4.37 \cdot 10^5 \text{ km}^2$) is, however, poorly understood. The quantification
5 of the water balance in space and time is a major challenge due the lack of measurements and
6 the inaccessibility of the terrain. The magnitude and distribution of high altitude precipitation,
7 which is the driver of the hydrological cycle, is one of its largest unknowns (Hewitt, 2005,
8 2007; Immerzeel et al., 2013; Mishra, 2015; Ragetti and Pellicciotti, 2012; Winiger et al.,
9 2005). Annual precipitation patterns in the UIB result from the intricate interplay between
10 synoptic scale circulation and valley scale topography-atmosphere interaction resulting in
11 orographic precipitation and funnelling of air movement (Barros et al., 2004; Hewitt, 2013).
12 At the synoptic scale, annual precipitation originates from two sources: the south-eastern
13 monsoon during the summer and moisture transported by the westerly jet stream over central
14 Asia (Mölg et al., 2013; Scherler et al., 2011) during winter. The relative contribution of
15 westerly disturbances to the total annual precipitation increases from south-east to north-west,
16 and the anomalous behaviour of Karakoram glaciers is commonly attributed to changes in
17 winter precipitation (Scherler et al., 2011; Yao et al., 2012).

18 At smaller scales the complex interaction between the valley topography and the atmosphere
19 dictates the spatial distribution of precipitation (Bookhagen and Burbank, 2006; Immerzeel et
20 al., 2014b). Valley bottoms, where stations are located, are generally dry and precipitation
21 increases up to a certain maximum altitude (HMAX) above which all moisture has been
22 orographically forced out of the air and precipitation decreases again. In westerly dominated
23 rainfall regimes HMAX is generally higher, which is likely related to the higher tropospheric
24 altitude of the westerly airflow (Harper, 2003; Hewitt, 2005, 2007; Scherler et al., 2011;
25 Winiger et al., 2005).

26 Gridded precipitation products are the de facto standard in hydrological assessments, and they
27 are either based on observations (e.g. APHRODITES (Yatagai et al., 2012)), remote sensing
28 (e.g. the Tropical Rainfall Monitoring Mission (Huffman et al., 2007)) or reanalysis (e.g.
29 ERA-Interim (Dee et al., 2011)) (Fig. 1, panel C to E). In most cases the station data strongly
30 influence the distribution and magnitude of the precipitation in those data products; however
31 the vast majority of the UIB is located at elevations far beyond the average station elevation
32 (Fig. 1, panel A to B). The few stations that are at elevations above 2000 m are located in dry

1 valleys and we hypothesize that the high altitude precipitation is considerably underestimated
2 (Fig. 1, panel C to E). Moreover, remote sensing based products, such as TRMM, are
3 insufficiently capable of capturing snowfall (Bookhagen and Burbank, 2006; Huffman et al.,
4 2007) and the spatial resolution (25 to 75 km^2) of most rainfall products (and the underlying
5 models) is insufficient to capture topography-atmosphere interaction at the valley scale (Fig.
6 1, panel C to E). Thus, there is a pressing need to improve the quantification of high altitude
7 precipitation, preferably at large spatial extents and at high resolution.

8 A possible way to correct mountain precipitation is to inversely close the water balance.
9 Previous studies in Sweden and Switzerland have shown that it is possible to derive vertical
10 precipitation gradients using observed runoff in a physically realistic manner (Valery et al.,
11 2009; Valéry et al., 2010). Earlier work at the small scale in high mountain Asia suggested
12 that the glacier mass balance may be used to reconstruct precipitation in its catchment area
13 (Harper, 2003; Immerzeel et al., 2012a). Fig. 1 (panel A and B) shows that UIB glaciers are
14 located at high elevations that are not represented by station data. Therefore the mass balances
15 of the glaciers may contain important information on high altitude accumulation in an area
16 that is inaccessible and ungauged, but very important from a hydrological point of view. In
17 this study we further elaborate this approach by inversely modelling average annual
18 precipitation from the mass balance of 550 large ($> 5 km^2$) glacier systems located throughout
19 the UIB. We perform a rigorous uncertainty analysis and we validate our findings using
20 independent observation of river runoff.

21 **2 Methods**

22 We estimate high altitude precipitation by using a glacier mass balance model to simulate
23 observed glacier mass balances. We use a gridded dataset from valley bottom stations as a
24 basis for our precipitation estimate and we compute a vertical precipitation gradient (PG (%
25 m^{-1})) until observed mass balances match the simulated mass balance. We repeat this process
26 for the 550 major glacier systems in the UIB, and the resulting PGs are then spatially
27 interpolated to generate a spatial field that represents the altitude dependence of precipitation.
28 We use this field to update the APHRODITE precipitation and generate a corrected
29 precipitation field that is able to reproduce the observed glacier mass balance. We validate the
30 findings independently with a water balance approach. Estimated (annual) runoff, based on
31 the corrected precipitation, actual evapotranspiration and the observed mass balance, is
32 compared with an extensive set of UIB runoff observations. A rigorous uncertainty analysis is

1 also conducted on the six most critical model parameters including potential effects of spatial
2 correlation.

3 **2.1 Datasets**

4 **2.1.1 Glacier mass balance and outlines**

5 Glacier mass balance trends based on ICESat (Kääb et al., 2012a) are recomputed for the
6 period 2003 until 2008 for the three major mountain ranges in the UIB: the Karakoram, the
7 Hindu-Kush and the Himalaya Fig. 1. For each zone the mass balance is computed including
8 a regional uncertainty estimate (Kääb et al., 2012a). From the zonal uncertainty (σ_z) we
9 estimate the standard deviation between glaciers within a zone (σ_g) as

$$10 \quad \sigma_g = \sigma_z \sqrt{n} . \quad (1)$$

11 Where n is the number of glaciers within a zone. The σ_g values used in the uncertainty
12 analysis are shown in Table 1.

13 Glacier outlines area based on the glacier inventory of the International Centre for Integrated
14 Mountain Development (Bajracharya and Shrestha, 2011).

15 **2.1.2 Precipitation and temperature**

16 The daily APHRODITE precipitation (Yatagai et al., 2012) and air temperature datasets
17 (Yasutomi et al., 2011) from 2003 until 2007 are used as reference datasets to ensure
18 maximum temporal overlap with the ICESat based glacier mass balance dataset (Kääb et al.,
19 2012a). The precipitation dataset is resampled from the nominal resolution of 25 km^2 to a
20 resolution of 1 km^2 using the nearest neighbour algorithm. The air temperature dataset is then
21 bias-corrected using monthly linear regressions with independent station data to account for
22 altitudinal and seasonal variations in air temperature lapse rates (Fig. 3).

23 **2.1.3 Runoff and evapotranspiration**

24 We use runoff data, potential (ETp) and actual evapotranspiration (ETa) data for the
25 validation of our results. For runoff we compiled all available published data, which we
26 complemented with data made available by the Pakistan Meteorological Department and the
27 Pakistan Water and Power Development Authority.

1 Evapotranspiration is notoriously difficult to monitor and there are few direct measurements
2 of ETa in the upper Indus. In earlier UIB studies, ET was estimated using an empirical
3 formulae based on air temperature but was only applied to the Siachen glacier (Bhutiyani,
4 1999; Reggiani and Rientjes, 2014). We take into account the uncertainty in ET in our stream
5 flow estimates and develop a blended product based on re-analysis datasets, a global
6 hydrological model and an energy balance model. Four gridded ETa and three gridded ETp
7 products were resampled to a 1 km^2 resolution at which we perform all our analyses:

- 8 • ERA-Interim reanalysis (Dee et al., 2011): ERA-Interim uses the HTESSEL land
9 surface scheme (Dee et al., 2011) to compute actual evapotranspiration (ETa). For
10 transpiration a distinction is made between high and low vegetation in the HTESSEL
11 scheme and these are parameterized from the Global Land Cover Characteristics
12 database at a nominal resolution of 1 km^2 .
- 13 • MERRA reanalysis (Rienecker et al., 2011): The MERRA reanalysis product of
14 NASA applies the state-of-the-art GEOS-5 data assimilation system that includes
15 many modern observing systems in a climate framework. MERRA uses the GEOS-5
16 catchment LSM land surface scheme (Koster et al., 2000) to compute actual ET. For
17 the MERRA product ETp is not available.
- 18 • ET-Look (Bastiaanssen et al., 2012): The ET-Look remote sensing model infers
19 information on ET from combined optical and passive microwave sensors, which can
20 observe the land-surface even under persistent overcast conditions. A two-layer
21 Penman–Monteith forms the basis of quantifying soil and canopy evaporation. The
22 dataset is available only for the year 2007, but it was scaled to the 2003 – 2007
23 average using the ratio between the 2003 – 2007 average and the 2007 annual ET
24 based on ERA-INTERIM.
- 25 • PCRGLOB-WB (Wada et al., 2014): The global hydrological model PCRGLOB-WB
26 computes actual evapotranspiration using potential evapotranspiration based on
27 Penman-Monteith, which is further reduced based on available soil moisture.

28 The average annual ETa for the period 2003 – 2007 for each of the four products is shown in
29 Fig. 2. The spatial patterns show good agreement, but the magnitudes differ considerably. The
30 ensemble mean ETa for the entire upper Indus equals $359 \pm 107 \text{ mm y}^{-1}$.

1 2.2 Model description

2 We use the PC-Raster spatial-temporal modelling environment (Karssenberget al., 2001) to
3 model the mass balance of the major glaciers in each zone and subsequently estimate
4 precipitation gradients required to sustain the observed mass balance. The model operates at a
5 daily time step from 2003 until 2007 and a spatial resolution of 1 km^2 . For each time step the
6 total accumulation and total melt are aggregated over the entire glacier surface. Only glaciers
7 with a surface area above 5 km^2 are included in the analysis (Karakoram = 232 glaciers,
8 Hindu-Kush = 119, Himalaya = 204 glaciers) as the ICESat measurements do not reflect
9 smaller glaciers. The model is forced by the spatial precipitation and temperature fields. The
10 precipitation fields are corrected using a precipitation gradient (PG, $\% \text{ m}^{-1}$). Precipitation is
11 positively lapsed using a PG between a reference elevation (HREF) to an elevation of
12 maximum precipitation (HMAX). At elevations above HMAX the precipitation is negatively
13 lapsed from its maximum at HMAX with the same PG according to:

$$15 \quad P_{cor} = P_{APHRO} \cdot (1 + (H - HREF)) \cdot PG \cdot 0.01 \quad \text{Eq. 1}$$

16

17 for $HREF < H \leq HMAX$, and:

18

$$19 \quad P_{cor} = P_{APHRO} \cdot (1 + (((HMAX - HREF) + (HMAX - H)) \cdot PG \cdot 0.01)) \quad \text{Eq. 2}$$

20 for $H > HMAX$

21

22 HREF and HMAX values are derived from literature (Table 1) and uncertainty is taken into
23 account in the uncertainty analysis. HMAX varies per zone and lies at a lower elevation in the
24 Himalayas than in the other two zones (Table 1). We spatially interpolate HMAX from the
25 average zonal values to cover the entire UIB.

26 The melt is modelled over the glacier area using the positive degree day (PDD) method
27 (Hock, 2005), with different degree day factors (DDF) for debris-covered (DDFd) and debris-
28 free (DDFdf) glaciers derived from literature (Table 1). To account for uncertainty in DDF,
29 the DDFd and DDFdf are taken into account separately in the uncertainty analysis. At

1 temperatures below the critical temperature of 2 °C (Immerzeel et al., 2013; Singh and
2 Bengtsson, 2004) precipitation falls in the form of snow and contributes to the accumulation.
3 Avalanche nourishment of glaciers is a key contributor for UIB glaciers (Hewitt, 2005, 2011)
4 and to take this process into account, we extend the glacier area with steep areas directly
5 adjacent to the glacier with a slope over an average threshold slope (TS) of 0.2 m m^{-1} . This
6 average threshold slope is derived by analysing the slopes of all glacierized pixels in the basin
7 (Fig.4). To account for uncertainty in TS this parameter is taken into account in the
8 uncertainty analysis.

9 For each glacier system we execute transient model runs from 2003 until 2007 and we
10 compute the average annual mass balance from the total accumulation and melt over this
11 period. We make two different runs for each glacier system with two different PGs (0.3%
12 m^{-1} and 0.6% m^{-1}) and we use the simulated mass balances of these two runs and the
13 observed mass balances based on ICESat to optimize the PG per glacier, such that the
14 simulated mass balance matches the observed.

15 To interpolate the glacier-specific PG-values to PG spatial fields over the entire domain we
16 use geostatistical conditional simulation (Goovaerts, 1997). Simulated spatial fields of PG are
17 thus conditioned on the PGs determined at the glaciers centroid. The semi-variogram has the
18 following parameters: nugget = 0, the range = 120 km, sill = variance of PGs.

19 **2.3 Uncertainty analysis**

20 A rigorous uncertainty analysis is performed to take into account the uncertainty in parameter
21 values and uncertainty in regional patterns. To account for parameter uncertainty we perform
22 a 10,000 member Monte Carlo simulation on the parameters given in Table 1. For each run
23 we randomly sample the parameter space based on the average (μ) and the standard deviation
24 (σ), which are all based on literature values. For the positively-valued parameters we use a
25 log-Gaussian distribution and a Gaussian distribution in case parameter values can be
26 negative. We take into account uncertainty in the following key parameters (HREF, HMAX,
27 DDFd, DDFdg, TS) for the PG as well as uncertainty in the mass balance against which the
28 PG is optimized (MB). . We randomly vary the 5 parameters (HREF, HMAX, DDFd, DDFdg,
29 TS) 10,000 times and calculate the PG for each glacier for each random parameter set drawn,
30 thus resulting in 10,000 PG-sets for each glacier considered. For each of the 10,000 PG-sets
31 we then use conditional simulation (see above) to arrive at 10,000 equally probable spatial

1 PG-fields, taking account of parameters uncertainty, mass-balance uncertainty and the
2 interpolation error. Note that for each of the 10,000 sets the variogram is scaled with the
3 variance of the PGs associated with the specific parameter combination drawn. Finally, based
4 on the results of the 10,000 simulations we derive the average corrected precipitation field
5 and the associated uncertainty in the estimates

6 Using the 10,000 combinations of parameters and associated PGs we ran a multi-variate linear
7 regression analysis to determine relative contribution of each parameter to the spread in the
8 PG to understand which parameter has the largest influence on the PG.

9 It is possible that certain parameters used in the model are spatially correlated. To account for
10 uncertainty in this spatial correlation and the presence of spatial patterns in the parameters we
11 perform a sensitivity analysis where we consider three cases:

- 12 • Fully correlated: we assume the parameters are spatially fully correlated within a zone,
13 e.g. for each of the 10,000 simulations a parameter has the same value within a zone
- 14 • Uncorrelated: we assume the parameters are spatially uncorrelated and within each
15 zone each glacier system is assigned a random value
- 16 • Intermediate case: we use geostatistical unconditional simulation (Goovaerts, 1997)
17 with a standardized semi-variogram (nugget = 0, sill = variance of parameter, range =
18 120 km) to simulate parameter values for each glacier system.

19 **2.4 Validation**

20 We estimate the average annual runoff (Q) for sub-basins in the UIB from

$$21 \quad Q = P_{cor} - ET + MB \quad (3)$$

22 Where P_{cor} is the corrected average precipitation, ET is the average annual evapotranspiration
23 based on the four products described above and MB is the glacier mass balance expressed
24 over the sub-basin area in $mm\ y^{-1}$. We then compare the estimated runoff values to the
25 observed time series (Table 2).

26 For the three zones (Himalaya, Karakoram and Hindu-Kush) we also perform a water balance
27 analysis to verify whether the use of the corrected precipitation product results in a more
28 realistic closure of the water balance. Finally we test the physical realism of the corrected
29 precipitation product using a non-dimensional Turc-Budyko plot as described in Valéry et al.

1 (2010). This plot is based on two assumptions: (i) the mean annual runoff should not exceed
2 the mean annual precipitation and (ii) the mean annual runoff should be larger than or equal to
3 the difference between precipitation and potential evapotranspiration. By plotting P/ET_p
4 versus Q/P on catchment basis it is tested whether the use of corrected precipitation results in
5 more physically-realistic values.

6

7 **3 Results and discussion**

8 **3.1 Corrected precipitation**

9 The average annual precipitation based on 10,000 conditionally simulated fields reveals a
10 striking pattern of high altitude precipitation. The amount of precipitation required to sustain
11 the large glacier systems is much higher than either the station observations or the gridded
12 precipitation products imply. For the entire UIB the uncorrected average annual precipitation
13 (Yatagai et al., 2012) for 2003-2007 is 437 mm y^{-1} ($191 \text{ km}^3 \text{ y}^{-1}$), an underestimation of more
14 than 200% compared with our corrected precipitation estimate of $913 \pm 323 \text{ mm y}^{-1}$ ($399 \pm$
15 $141 \text{ km}^3 \text{ y}^{-1}$ (Fig. 5)). The greatest corrected annual precipitation totals in the UIB (1271 mm
16 y^{-1}) are observed in the elevation belt between 3750 m to 4250 m (compared to 403 mm y^{-1} for
17 the uncorrected case). In absolute terms the main water-producing region is located in the
18 elevation belt between 4250 m and 4750 m where approximately 78 km^3 of rain and snow
19 precipitates annually.

20 In the most extreme case, precipitation is underestimated by a factor 5 to 10 in the region
21 where the Pamir, Karakorum and Hindu-Kush ranges intersect (Fig. 5). Our inverse modelling
22 shows that the large glacier systems in the region can only be sustained if snowfall in their
23 accumulation areas totals around 2000 mm y^{-1} (Hewitt, 2007). This is in sharp contrast to
24 precipitation amounts between 200 and 300 mm y^{-1} that are reported by the gridded
25 precipitation products (Fig. 1). Our results match well with the few studies on high-altitude
26 precipitation that are available. Annual accumulation values between 1000 and 3000 mm have
27 been reported for accumulation pits above 4000 m in the Karakoram (The Batura Glacier
28 Investigation Group, 1979; Wake, 1989; Winiger et al., 2005). Our results show that the
29 highest precipitation amounts are found along the monsoon-influenced southern Himalayan
30 arc with values up to 3000 mm y^{-1} , while north of the Himalayan range the precipitation
31 decreases quickly towards a vast dry area in the north-eastern part of the UIB (Shyok sub-

1 basin). In the north-western part of the UIB, westerly storm systems are expected to generate
2 considerable amounts of precipitation at high altitude.

3 Our results reveal a strong relation between elevation and precipitation with a median PG for
4 the entire upper Indus of $0.0989 \% m^{-1}$, but with large regional differences. Median
5 precipitation gradients in the Hindu-Kush and Karakoram ranges ($0.260 \% m^{-1}$ and 0.119%
6 m^{-1} respectively) are significantly larger than those observed in the Himalayan range, e.g
7 $0.044 \% m^{-1}$ (Fig. 6). In the Hindu-Kush, for example, for every 1000 *m* elevation rise,
8 precipitation increases by 260% with respect to APHRODITE, which is based on valley floor
9 precipitation. In combination with a higher HMAX in the Hindu-Kush and the Karakoram
10 (e.g. 5500 m versus 4500 m in the Himalayas, (Hewitt, 2007; Immerzeel et al., 2014a;
11 Putkonen, 2004; Seko, 1987; Winiger et al., 2005) this suggests that westerly airflow indeed
12 has a higher tropospheric altitude and that the interplay between elevation and precipitation is
13 stronger for this type of precipitation. Further research should thus focus on the use of high
14 resolution cloud-resolving weather models (Collier et al., 2014; Mölg et al., 2013) for this
15 region to further resolve seasonal topography-precipitation interaction at both synoptic and
16 valley scales.

17 The estimated precipitation is considerably higher than what was reported in previous studies.
18 Several studies have used TRMM products to quantify UIB precipitation (Bookhagen and
19 Burbank, 2010; Immerzeel et al., 2009, 2010) and they show average annual precipitation
20 values around 300 *mm*. It was also noted that the water balance was not closing and average
21 annual river runoff at Tarbela exceeded the TRMM precipitation (Immerzeel et al., 2009).
22 Two possible reasons have been suggested to explain this gap: (i) the high altitude
23 precipitation is underestimated, (ii) the glaciers are in a significant negative imbalance
24 (Immerzeel et al., 2009). Since the ICESat study and several other geodetic mass balance
25 studies (Gardelle et al., 2013; Kääb et al., 2012b) it has become clear that the glaciers in this
26 region are not experiencing a significant ice loss and that this cannot be the explanation for
27 the missing water in the water balance. This supports our conclusion that it is the high altitude
28 precipitation that has been underestimated. A study based on long term observations of
29 Tarbela inflow also confirm our results (Reggiani and Rientjes, 2014). In this study the total
30 UIB precipitation above Tarbela is estimated to be $675 \pm 100 mm y^{-1}$ and the difference
31 remaining between our results may stem from the fact that the UIB we consider is twice the
32 size of the area above the Tarbela, the different approach used to estimate actual ET, the

1 different period considered and their assumption that ice storage has not changed between
2 1961 and 2009.

3 **3.2 Uncertainty**

4 We estimated the uncertainty in the modelled precipitation field with the standard deviation
5 (σ) of the 10,000 realizations (Fig. 5c). The signal-to-noise ratio is satisfactory in the entire
6 domain, e.g. the σ is always considerably smaller than the average precipitation with an
7 average coefficient of variation of 0.35. The largest absolute uncertainty is found along the
8 Himalayan arc and this coincides with the precipitation pattern found here. Strikingly, the
9 region where the underestimation of precipitation is largest, at the intersection of the three
10 mountain ranges in the northern UIB, is also an area where the uncertainty is small even
11 though precipitation gradients are large.

12 By running a multiple regression analysis after optimizing the PGs we quantify the
13 contribution of each parameter to the total uncertainty. The largest source of uncertainty in
14 our estimate of UIB high altitude precipitation stems from the MB estimates, followed by the
15 DDFdf, DDFd, TS, HREF and HMAX, although regional differences are considerable (Fig.
16 7). The MB constrains the precipitation gradients and thereby exerts a strong control on the
17 corrected precipitation fields, in particular because the intra-zonal variation in MB is
18 relatively large (Table 1). Thus, improved spatial monitoring techniques of the MB are
19 expected to greatly improve precipitation estimates.

20 Fig. 8 shows the result of uncertainty analysis associated to the spatial correlation of the
21 parameters, which reveals that the impact on the average corrected precipitation is limited.
22 Locally there are minor differences in the corrected precipitation amounts, but overall the
23 magnitude and spatial patterns are similar. However, there are considerable differences in the
24 uncertainty. The lowest uncertainty is found for the fully uncorrelated case, the fully
25 correlated case has the highest uncertainty whereas the intermediate case is in between both.
26 For the fully correlated case all glacier systems have the same parameter set for the specific
27 realization and this results in a larger final uncertainty. In the uncorrelated case each glacier
28 system has a different randomly sampled parameter set and this reduces the overall
29 uncertainty as it spatially attenuates the variation in precipitation gradients.

1 3.3 Validation

2 The corrected precipitation is validated independently by a comparison to published average
3 annual runoff data of 27 stations (Table 2). Runoff estimates based on the corrected
4 precipitation agree well with the average observed annual runoff (Fig. 9, top panel). The
5 runoff estimated for the uncorrected APHRODITE is consistently lower than the observed
6 runoff, and in some occasions even negative. Runoff estimates were also made based on the
7 ERA-INTERIM and TRMM precipitation products. The TRMM results yield a similar
8 underestimation as the uncorrected APHRODITES product, but the runoff estimates based on
9 the ERA-INTERIM precipitation agree reasonably well with the observations. However the
10 coarse resolution ($\sim 70 \text{ km}^2$) (Fig. 1) is problematic and cannot be used to reproduce the mass
11 balance (Fig. 11). Averaged over large catchments the precipitation may be applied for
12 hydrological modeling, but at smaller scales there are likely very large biases. As a result, the
13 observed glacier mass balances cannot be reproduced when the ERA-INTERIM dataset is
14 used.

15 The zonal water balance analysis (Fig. 9, bottom panels) reveals that the water balance is
16 much more realistic when the corrected precipitation is used. Although the uncertainties are
17 considerable, our analysis shows that the Himalaya and Hindu-Kush zones are about twice as
18 wet as the Karakoram zone. For all three zones the glacier mass imbalance only plays a
19 marginal role in the overall water balance and about 60% of the total precipitation runs off
20 while 40% is lost through evapotranspiration. Notable are the values for *Corg*, which
21 represents the water balance gap in case the uncorrected precipitation is used, are
22 approximately 500 mm y^{-1} in all three zones. Our validation does not take into account
23 groundwater fluxes and we have assumed that over the observed period from 2003 until 2007
24 there is no net loss or gain of groundwater in the upper Indus basin. We do acknowledge that
25 groundwater may play an important role in the hydrology. A study in the Nepal Himalaya
26 shows that fractured basement aquifers fill during the monsoon and they purge in the post-
27 monsoon thus causing a natural delay in runoff (Andermann et al., 2012). However this does
28 not imply significant net gains or losses over multiple year periods, which is what we
29 consider. A second component that we have not considered and that may play a role in the
30 high altitude water cycle is sublimation. There are some indications that wind redistribution
31 and sublimation may play a considerable role in the high altitude water balance (Wagnon et
32 al., 2013). However our PGs are constrained on the observed mass balance, hence our

1 precipitation can be considered as a net precipitation and sublimation losses are thus
2 accounted for.

3 In Fig. 10 the Budyko-Turc plot is shown to confirm the physical realism of our results. Those
4 dots located in the hatched part of the graph are physically less realistic. For the uncorrected
5 case almost all dots (open dots) are above the $Q/P = 1$ line. For the corrected case the Q/P
6 values are much more plausible, however there many catchments which are located slightly to
7 the right side of the theoretical Budyko line, meaning that the Q is smaller than the difference
8 between P and ET_p . Possible deviation can potentially be explained by uncertainties in
9 observed flows and ET_p estimates, the fact the in glacierized catchments the theoretical
10 Budyko curve may be different because a glacier imbalance can be an additional water
11 balance term that is unaccounted for, a too short time period that is used to construct the water
12 balance and finally that some of the discharge observations do not align in time with the rest
13 of the water balance terms. Overall we conclude though that the use of the corrected
14 precipitation results in physically more realistic results, where the water balance could be
15 closed and no significant amount of precipitation input is missing.

16 Fig 11 shows how the average simulated zonal glacier mass balance using the corrected, the
17 APHRODITES, the ERA-Interim and the TRMM precipitation datasets. It shows that none of
18 the precipitation products can reproduce the observed mass balance. Mostly the mass balances
19 are underestimated which is consistent with and underestimation of the precipitation. The
20 ERA-Interim dataset overestimates the mass balance in the Himalaya, but underestimates the
21 mass balances in the other two zones as result of the coarse resolution.

22 **4 Conclusions**

23 In this study we inversely model high altitude precipitation in the upper Indus basins from
24 glacier mass balance trends derived by remote sensing. Although there are significant
25 uncertainties, our results unambiguously show that high altitude precipitation in this region is
26 underestimated and that the large glaciers here can only be sustained if high altitude
27 accumulation is much higher than most commonly used gridded data products.

28 Our results have an important bearing on water resources management studies in the region.
29 The observed gap between precipitation and streamflow (Immerzeel et al., 2009) (with stream
30 flow being larger) cannot be attributed to the observed glacier mass balance (Kääb et al.,
31 2012a), but is most likely the result of an underestimation of precipitation, as also follows
32 from this study. With no apparent decreasing trends in precipitation (Archer and Fowler,

1 2004) the observed negative trends in stream flow in the glacierised parts of the UIB should
2 thus be primarily attributed to decreased glacier and snow melt (Sharif et al., 2013) and
3 increased glacier storage (Gardelle et al., 2012). In a recent study the notion of negative trends
4 in UIB runoff was contested and based on a recent analysis (1985 - 2010) it was concluded
5 that runoff of Karakoram rivers is increasing (Mukhopadhyay and Khan, 2014b). The study
6 suggests that increase glacier melt during summer is the underlying reason, which in
7 combination with positive precipitation trends in summer does not contradict the neutral
8 glacier mass balances in the region. From all of these studies it becomes apparent that
9 precipitation is the key to understanding behaviour of glacier and hydrology at large in the
10 UIB. The precipitation we estimate in this study differs considerably, in magnitude and spatial
11 distribution, from datasets that are commonly used in design of reservoirs for hydropower and
12 irrigation and as such it may have a significant impact and improve such planning processes.

13 The water resources of the Indus River play an important geopolitical role in the region, and
14 our results could contribute to the provision of independent estimates of UIB precipitation.
15 We conclude that the water resources in the UIB are even more important and abundant than
16 previously thought. Most precipitation at high altitude is now stored in the glaciers, but when
17 global warming persists and the runoff regime becomes more rain dominated, the downstream
18 impacts of our findings will become more evident.

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- 17

1 Table 1. Averages (μ) and standard deviations (σ) of predictors for the precipitation gradient.
2 Values and ranges are based on literature as follows: HREF, HMAX: (Hewitt, 2007, 2011;
3 Immerzeel et al., 2012b, 2014b; Putkonen, 2004; Seko, 1987; Winiger et al., 2005), DDFd,
4 DDFdf: (Azam et al., 2012; Hagg et al., 2008; Immerzeel et al., 2013; Mihalcea et al., 2006;
5 Nicholson and Benn, 2006), MB: (Kääb et al., 2012a)

Variable	Acronym	Distribution	μ	σ
Reference elevation (<i>m</i>)	HREF	log-Gaussian	2500	500
Maximum elevation (<i>m</i>)	HMAX	log-Gaussian		
<i>Himalaya</i>			4500	500
<i>Karakoram</i>			5500	500
<i>Hindu-Kush</i>			5500	500
Degree day factor debris covered glaciers ($mm\ ^\circ C^{-1}\ d^{-1}$)	DDFd	log-Gaussian	2	2
Degree day factor debris free glaciers ($mm\ ^\circ C^{-1}\ d^{-1}$)	DDFdf	log-Gaussian	7	2
Threshold slope ($m\ m^{-1}$)	TS	log-Gaussian	0.2	0.05
Mass balance ($m\ w.e.\ y^{-1}$)	MB	Gaussian		
<i>Himalaya</i>			-0.49	0.57
<i>Hindu-Kush</i>			-0.21	0.76
<i>Karakoram</i>			-0.07	0.61

6

7

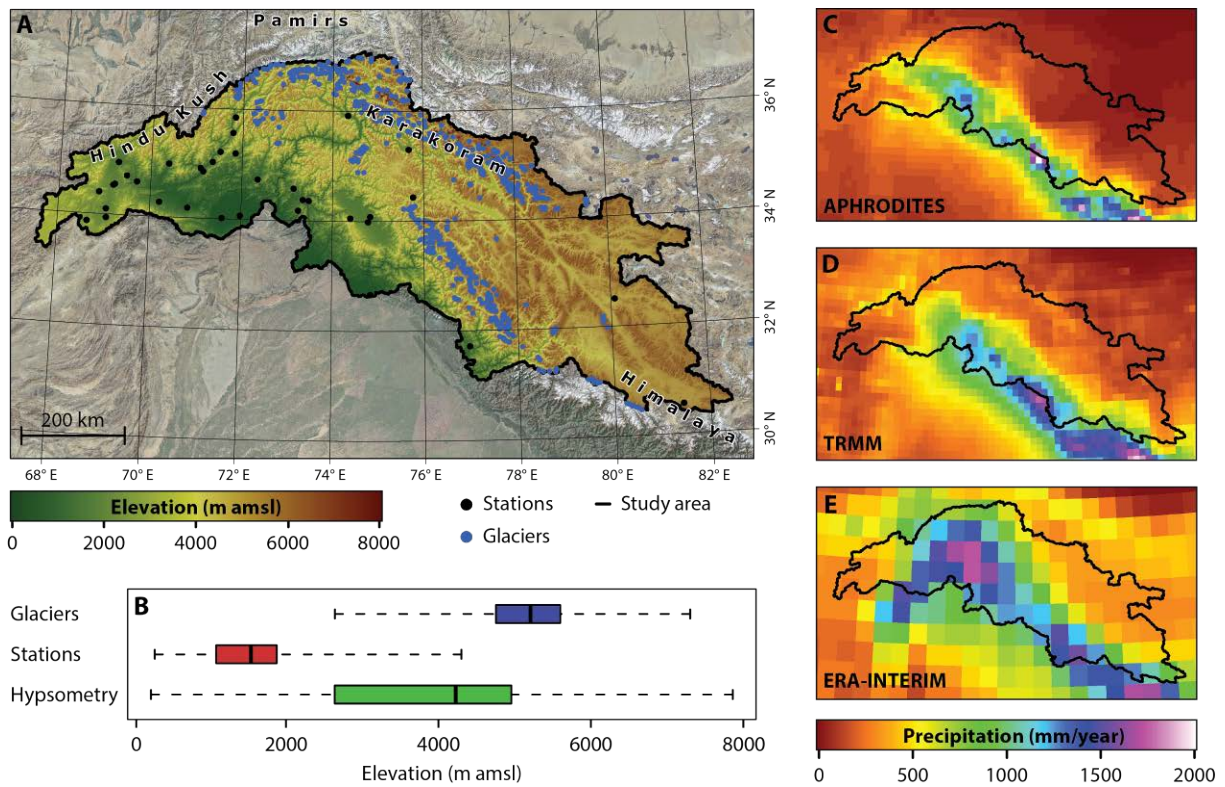
1 Table 2. Runoff stations used for validation. Catchment areas are delineated based on SRTM
2 DEM. * = calculated based on discharge provided by the Pakistan Water and Power
3 Development (WAPDA), ** = based on (Mukhopadhyay and Khan, 2014a), *** = based on
4 (Sharif et al., 2013), **** = based on (Archer, 2003), ***** = based on (Khattak et al., 2011).

Station	LAT	LON	Area (km²)	Observed Q (m³ s⁻¹)	Period
Besham Qila*	34.906	72.866	198741	2372.2	2000-2007
Tarbela inflow*	34.329	72.856	203014	2370.3	1998-2007
Mangla inflow*	33.200	73.650	29966	831.8	1998-2007
Marala inflow*	32.670	74.460	29611	956.5	1998-2007
Dainyor bridge*	35.925	74.372	14147	331.8	1998-2004
Skardu - Kachura**	35.435	75.468	146200	1074.2	1970-1997
Partab Bridge**	35.767	74.597	177622	1787.9	1962-1996
Yogo**	35.183	76.100	64240	359.4	1973-1997
Kharmong**	34.933	76.217	70875	452.3	1982-1997
Gilgit**	35.933	74.300	13174	286.7	1960-1998
Doyian**	35.550	74.700	4000	135.7	1974-1997
Chitral**	35.867	71.783	12824	271.9	1964-1996
Kalam**	35.467	72.600	2151	89.6	1961-1997
Naran**	34.900	73.650	1181	48.1	1960-1998
Alam bridge**	35.767	74.600	28201	644	1966-1997
Chakdara**	34.650	72.017	5990	178.9	1961-1997
Karora**	34.900	72.767	586	21.2	1975-1996
Garhi Habibullah**	34.450	73.367	2493	101.8	1960-1998
Muzafferabad***	34.430	73.486	7604	357	1963-1995
Chinari***	34.158	73.831	14248	330	1970-1995
Kohala***	34.095	73.499	25820	828	1965-1995

Kotli***	33.525	73.890	2907	134	1960-1995
Shigar**	35.422	75.732	6681	202.6	1985-1997
Phulra**	34.317	73.083	1106	19.2	1969-1996
Daggar**	34.500	72.467	534	6.9	1969-1996
Warsak****	34.100	71.300	74092	593	1967-2005
Shatial Bridge**	35.533	73.567	189263	2083.2	1983-1997

1

2



1

2

3 Figure 1. Overview of the UIB, basin hypsometry and three gridded precipitation products.

4 Panel A shows the digital elevation model and the location of the major glacier systems (area

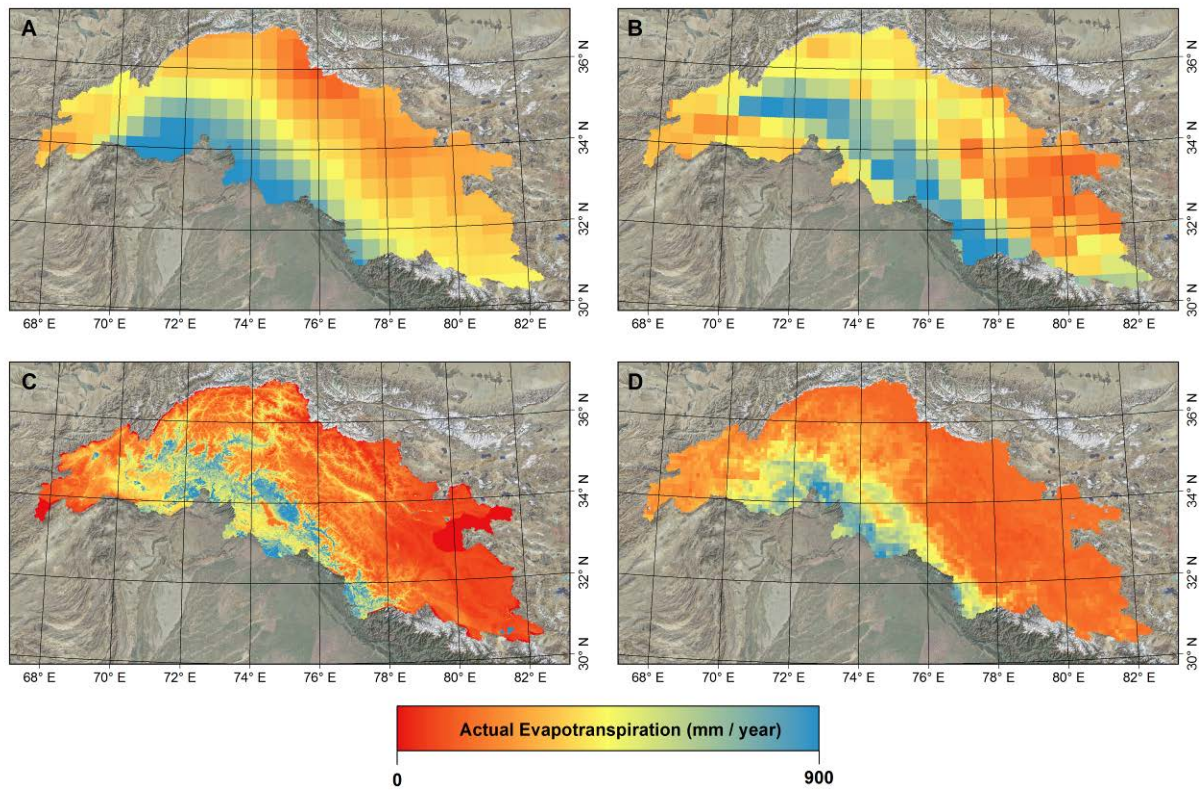
5 $> 5 \text{ km}^2$) and the available stations in the Global Summary of the Day (GSOD) of the World

6 Meteorological Organization (WMO). Panel B shows boxplots of the elevation distribution of

7 the basin, the large glacier systems and the GSOD stations. Panel C to E show the average

8 gridded annual precipitation between 1998-2007 for the APHRODITE(Yatagai et al., 2012),

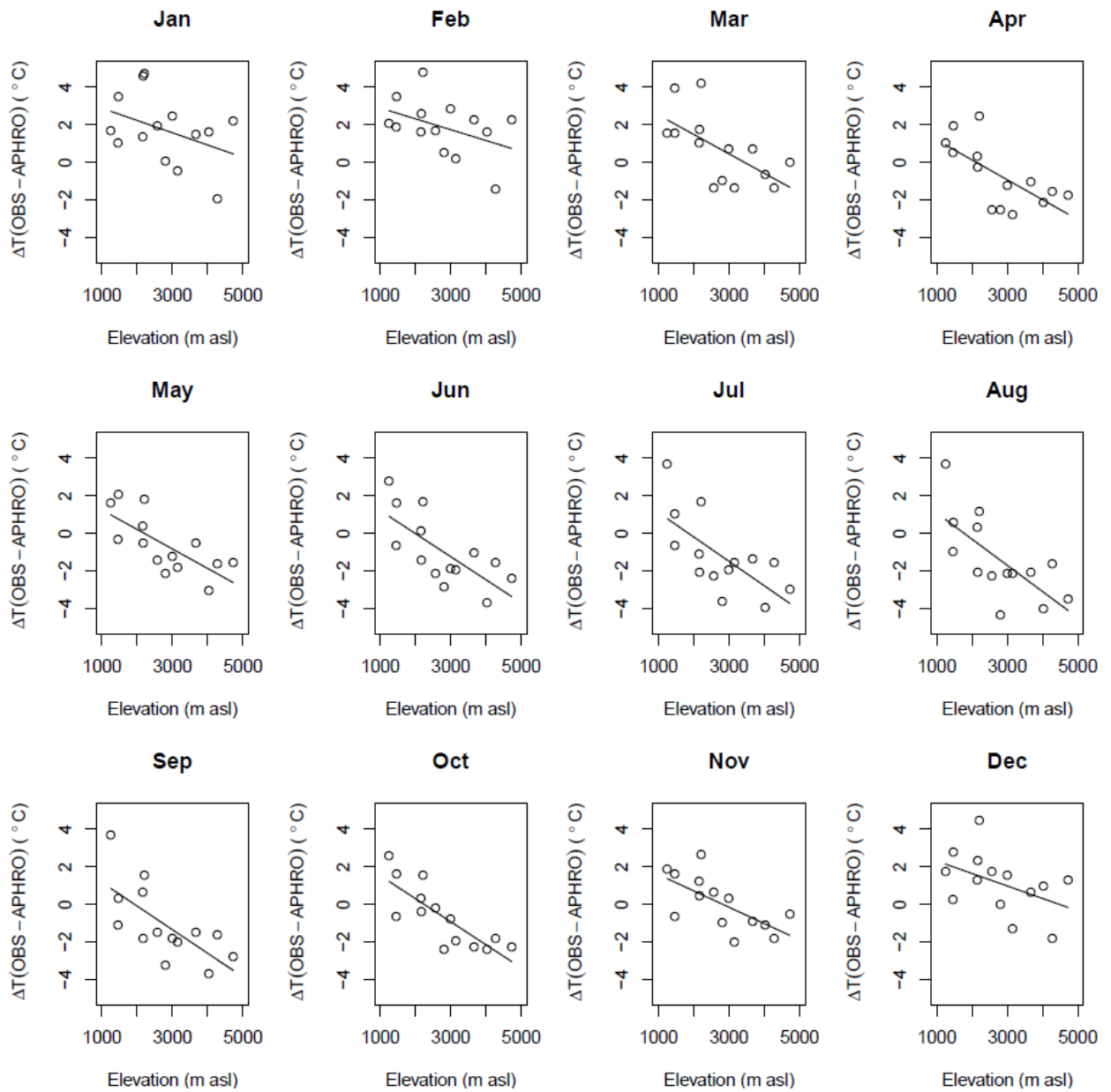
9 TRMM (Huffman et al., 2007) and ERA-INTERIM (Dee et al., 2011) datasets.



1

2 Figure 2. Average annual actual evapotranspiration between 2003 and 2007 for ERA-Interim
 3 (A), MERRA (B), ET-Look (C) and PCRGLOB-WB (D).

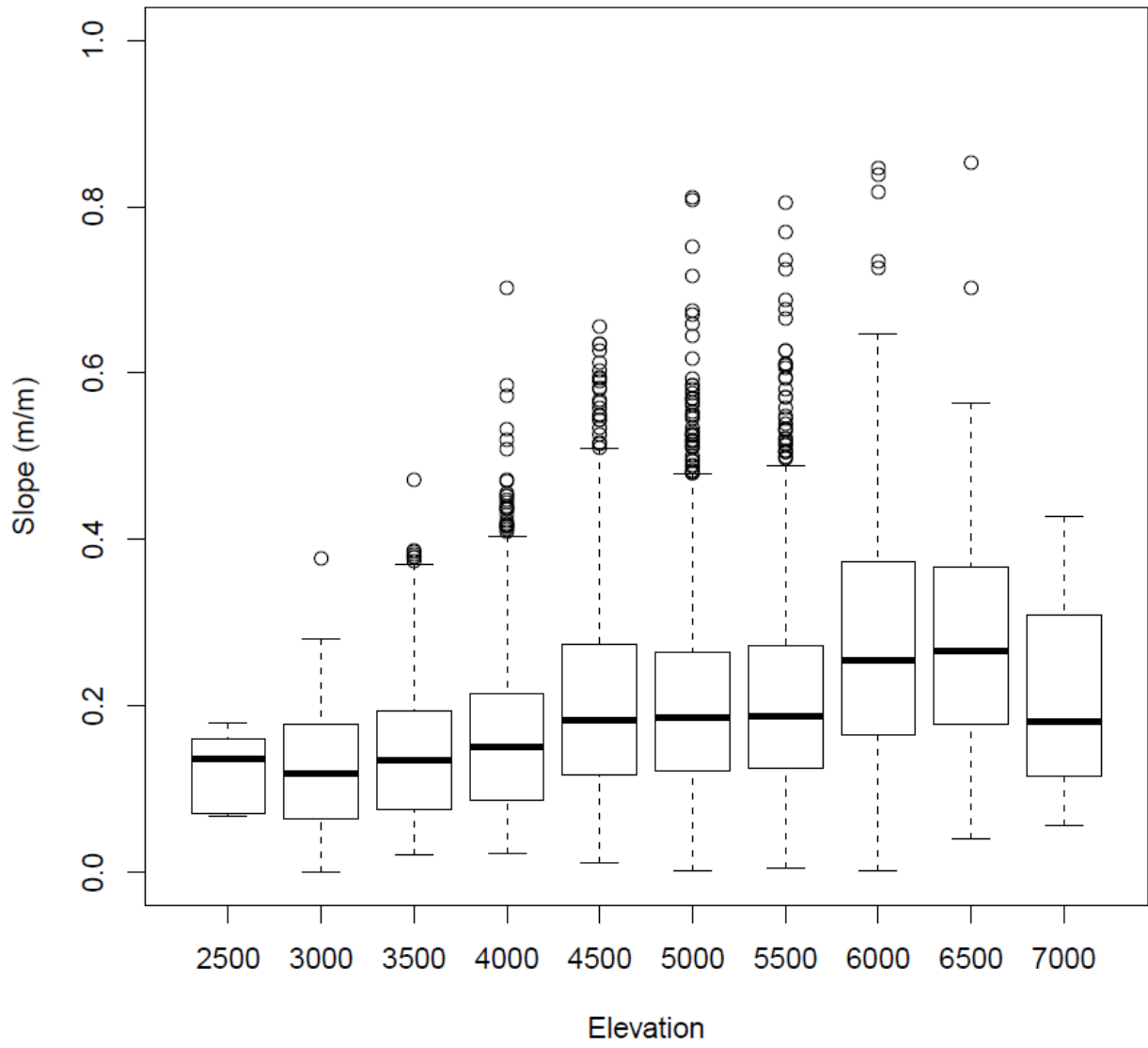
4



1

2 Figure 3. Monthly relation between observed temperatures at meteorological stations (OBS)
 3 and the APHRODITE temperature fields (APHRO) (Yasutomi et al., 2011).

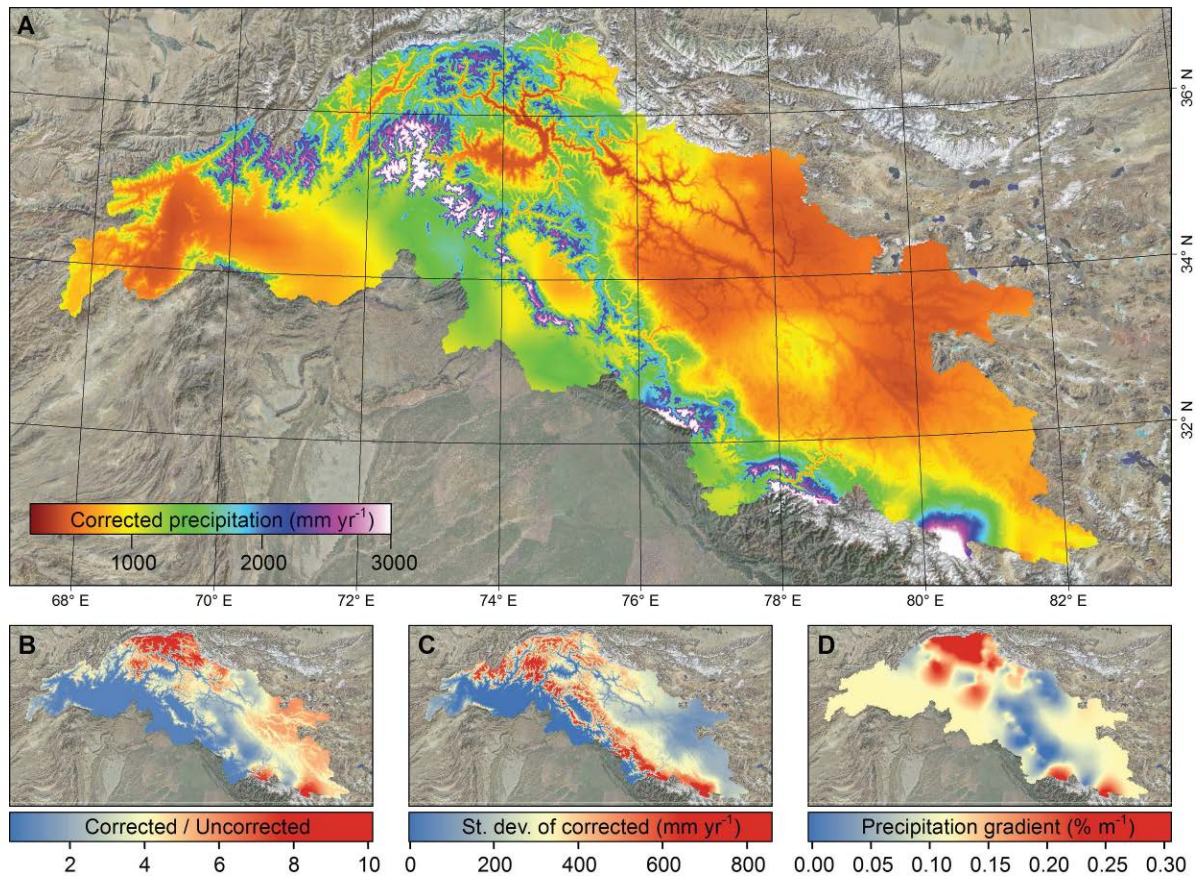
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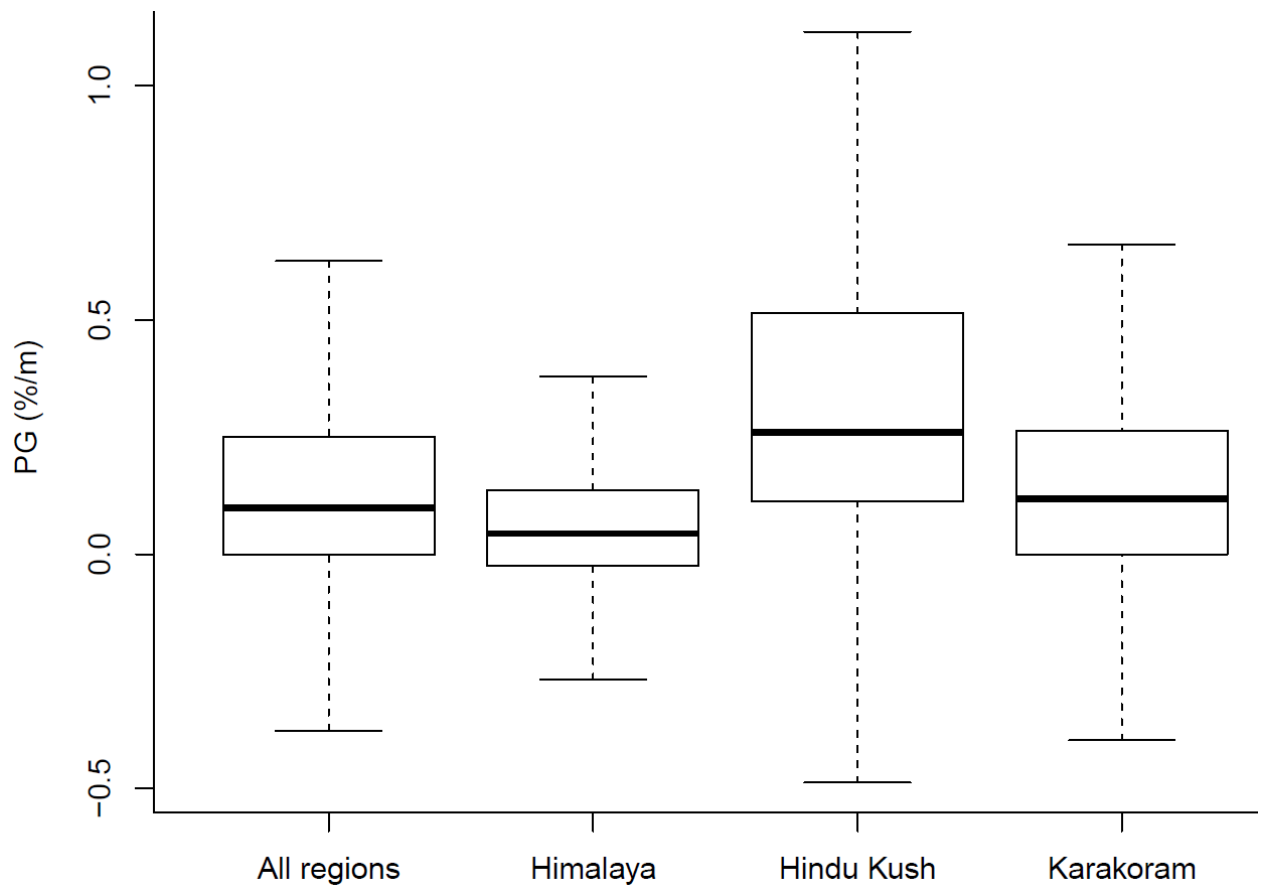
1

2 Figure 4. Boxplots of slopes of glacierised areas per elevation bin.

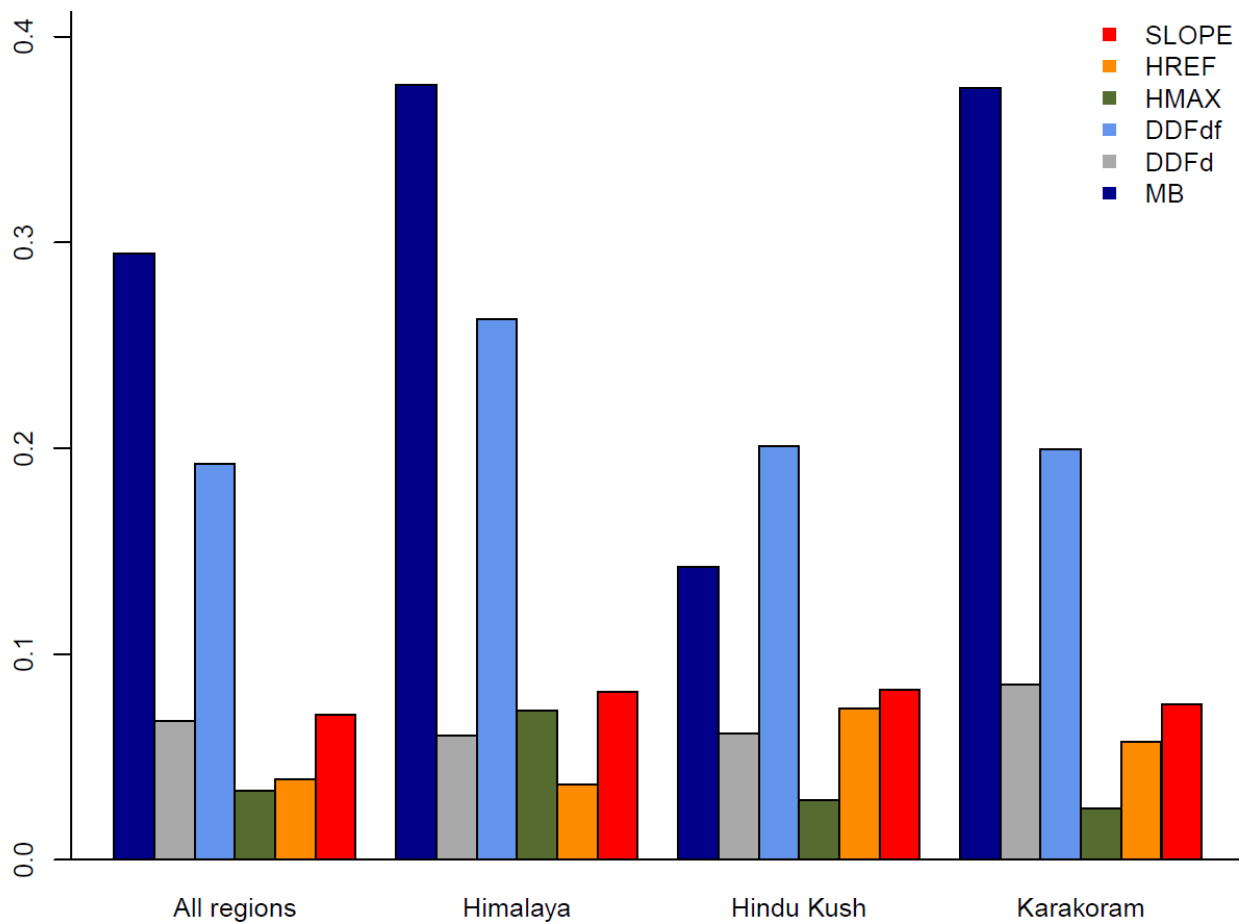
3



1
 2 Figure 5. Corrected precipitation and estimated uncertainty for the UIB for the case with
 3 intermediate spatial correlation between model parameters. Panel A shows the average
 4 modelled precipitation field based on 10,000 simulations for the period 2003-2007, Panel B
 5 shows the ratio of corrected precipitation to the uncorrected APHRODITE precipitation for
 6 the same period, Panel C shows the standard deviation of the 10,000 simulations and panel D
 7 shows the average precipitation gradient.
 8



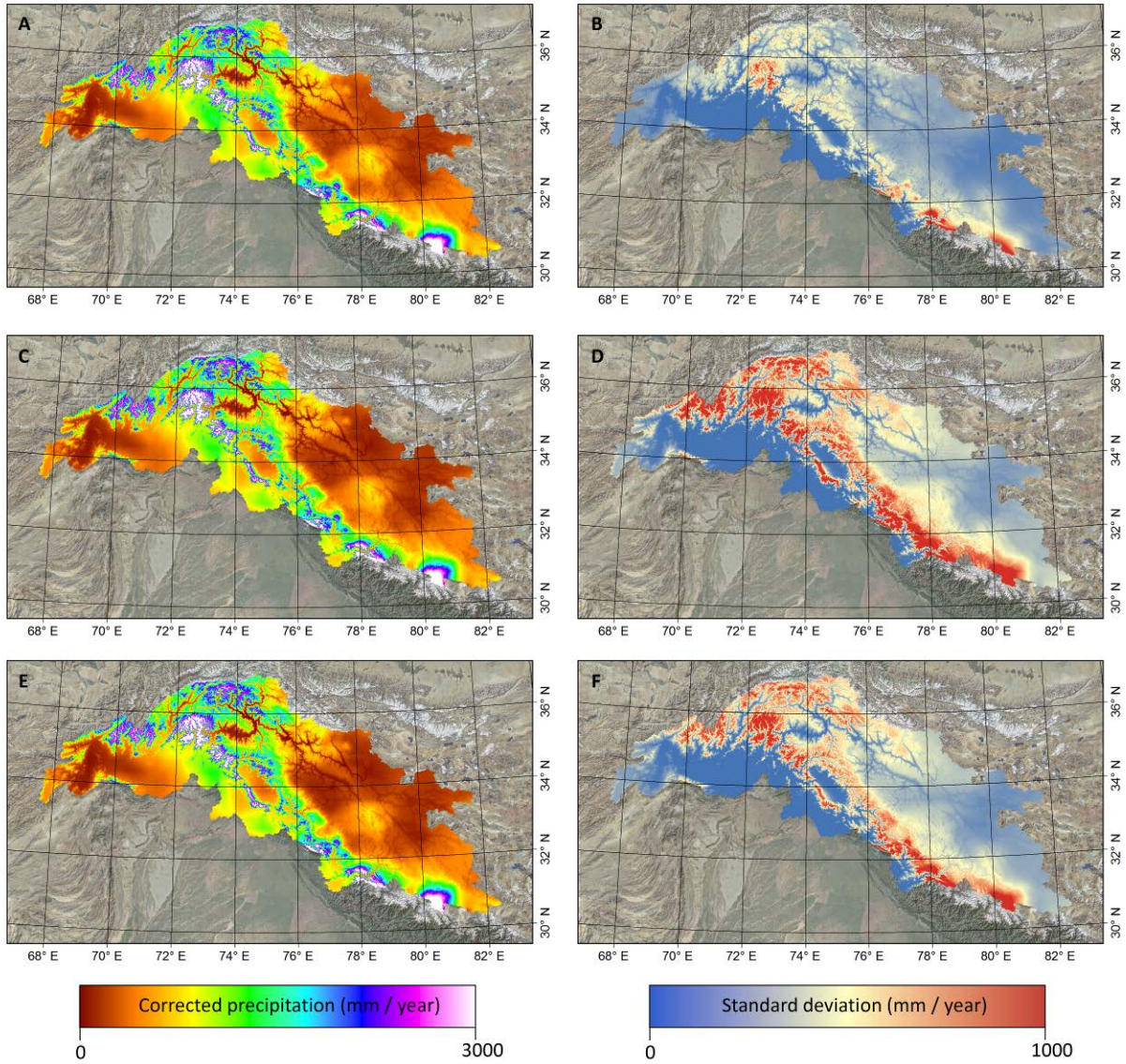
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 2 Figure 6. Box plots of precipitation gradients for the entire UIB and for the three regions
 3 separately.
 4



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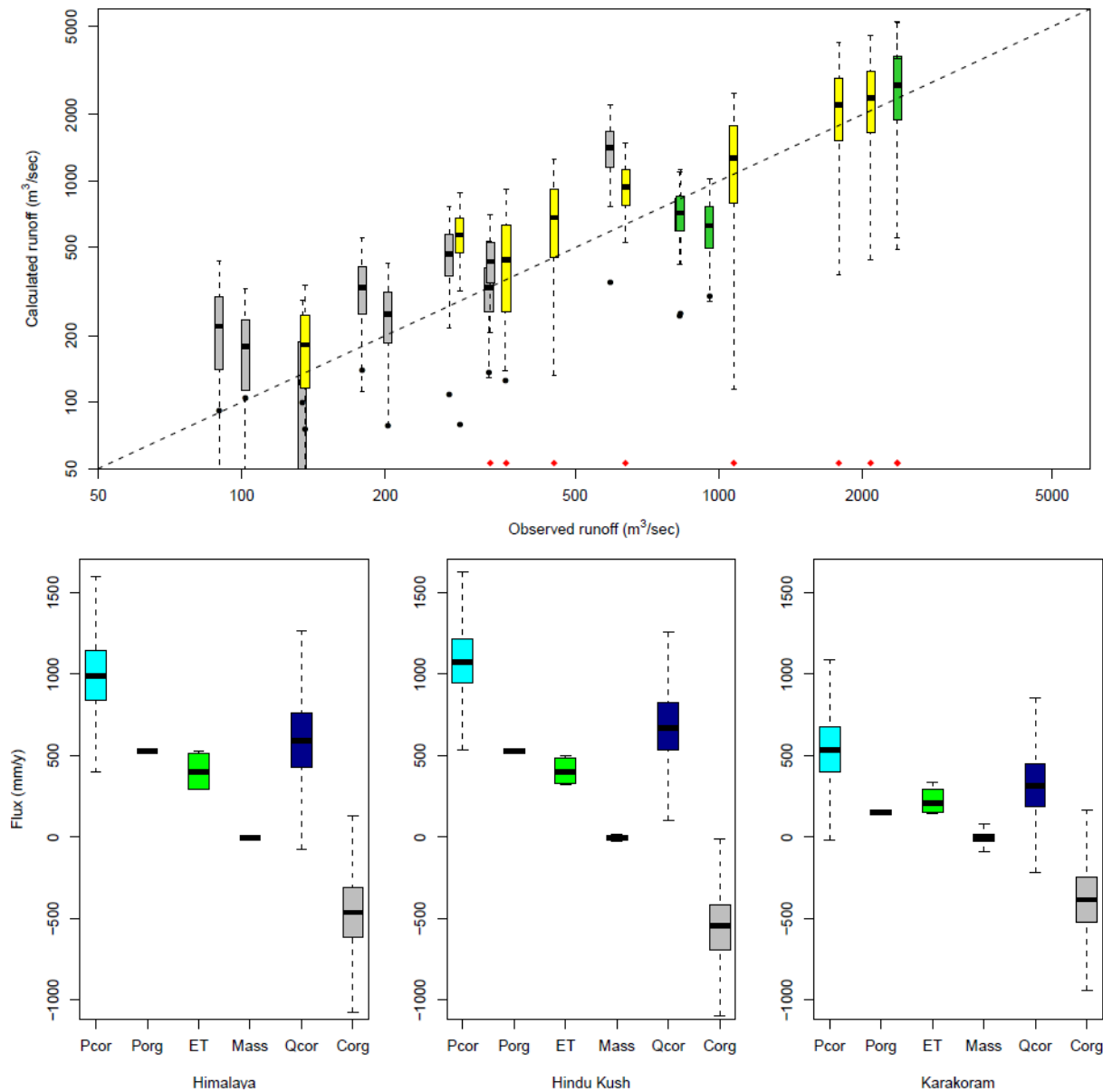
2 Figure 7. Normalized weights of multiple regression of the precipitation gradients by the
 3 predictors slope (slope threshold for avalanching to contribute to accumulation), HREF (base
 4 elevation from which lapsing starts), HMAX (elevation with peak precipitation), DDFd
 5 (degree day factor for debris covered glaciers), DDFdf (degree day factor for debris free
 6 glaciers) and the MB (mass balance of the glacier).

7

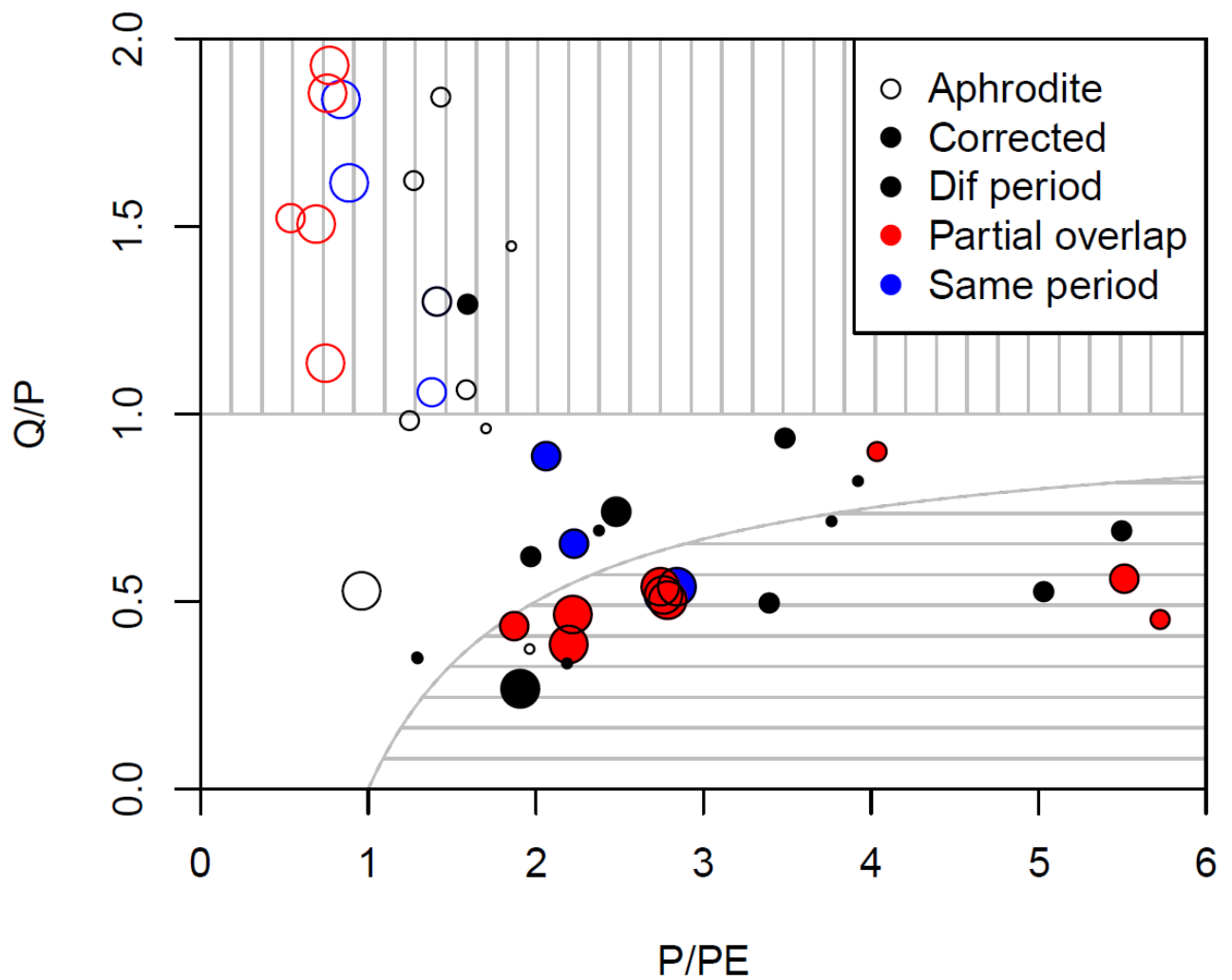


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Figure 8. Impact of spatial correlation of parameters on the corrected precipitation field and associated uncertainty. The top panels show the corrected precipitation field (panel A) and uncertainty (panel B) for the fully uncorrelated case. The middle panels (D,E) for the fully correlated case and the bottom panels (E,F) for the intermediate case.

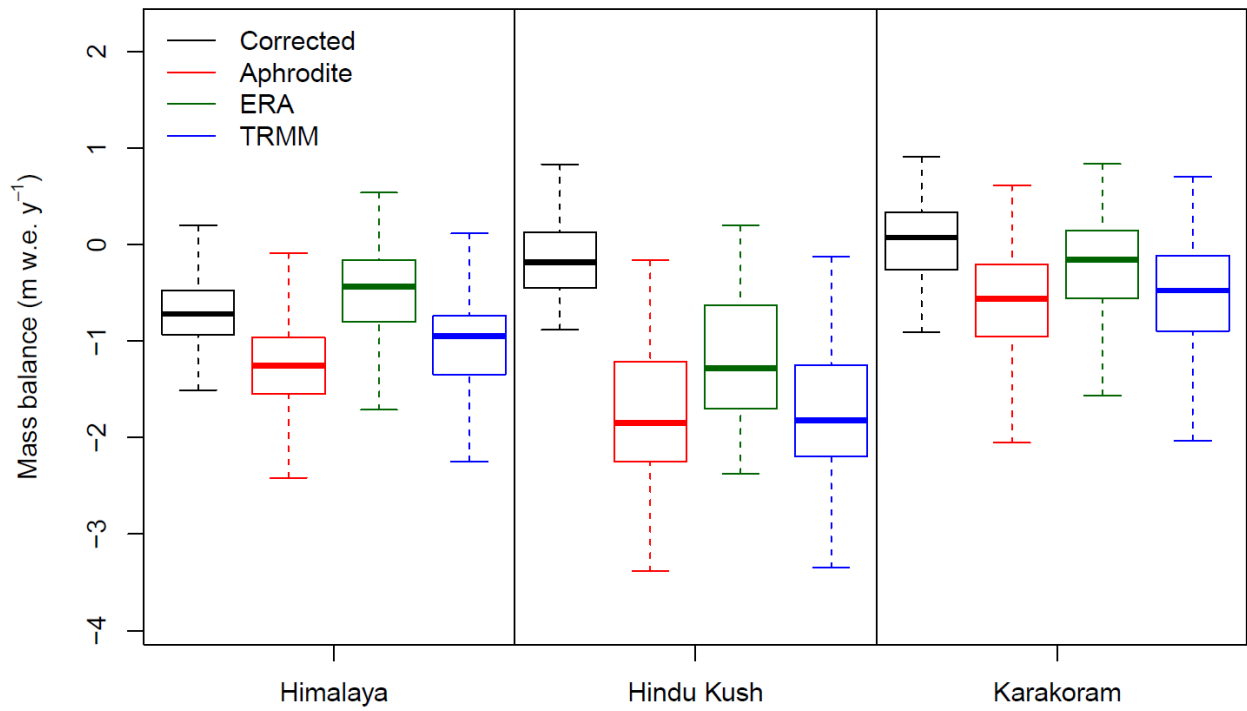


1
 2 Figure 9. Validation of the precipitation correction using observed discharge (Table 2). Top
 3 panel: The box plots are based on the runoff estimate based on 10,000 corrected precipitation
 4 fields (grey: stations for which the observed record does not coincide with the 2003-2007
 5 period, yellow: stations for which the 2003 – 2007 period is part of the observational record,
 6 green: stations for which the observations are based precisely on the 2003 – 2007 period. The
 7 black dots and red diamonds (estimated runoff below $50 \text{ m}^3 \text{ s}^{-1}$) show the estimated runoff
 8 based on the uncorrected precipitation. Bottom panels: Water balance components of each
 9 zone (Pcor = corrected precipitation, Porg = uncorrected APHRODITES precipitation, ET =
 10 actual evapotranspiration, Mass = glacier mass balance, Qcor = estimated runoff, Corg =
 11 water balance gap in case the Porg is used).



1

2 Figure 10. Non-dimensional graphical representation of catchments using their mean runoff,
 3 Q, precipitation, P, and potential evapotranspiration, PE. The grey line in the empty centre
 4 area represents the theoretical Budyko relationship in the non-dimensional graph. The size of
 5 the dots is scaled to the catchment area.



1

2 Figure 11. Reconstructed mass balances based on the corrected, APHRODITE, ERA-
 3 INTERIM and TRMM datasets. The black horizontal dotted line shows the observed mass
 4 balance for each zone.

5

