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Comparison of Cenozoic surface uplift and glacial-interglacial cycles on Himalaya-Tibet paleo-climate: Insights from a regional climate model

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44 **Abstract**

45

46 Assessing paleo-climatic changes across the Tibetan Plateau and the underlying
47 driving mechanisms provides insights for the natural variability in the Earth's climate
48 system in response to tectonic processes and global climate change. In this study, we
49 use a high-resolution regional climate model to investigate various episodes of distinct
50 climate states over the Tibetan Plateau region during the Cenozoic rise of the Plateau
51 and Quaternary glacial/interglacial cycles. The main objective is to compare climate
52 changes during the Miocene-Pliocene uplift period with climate anomalies during the
53 last glacial maximum and the mid-Holocene optimum, based on a consistent modeling
54 framework.

55 Reduced plateau elevation leads to regionally differentiated patterns of higher
56 temperature and lower precipitation amount on the plateau itself, whereas surrounding
57 regions are subject to colder conditions. In particular, Central Asia receives much more
58 precipitation prior to the uplift, mainly due to a shift of the stationary wave train over
59 Eurasia. Cluster analysis indicates that the continental-desert type climate, which is
60 widespread over Central Asia today, appears with the Tibetan Plateau reaching 50 %
61 of its present-day elevation.

62 The mid-Holocene is characterized by slightly colder temperatures, and the last glacial
63 maximum by considerably colder conditions over most of central and southern Asia.
64 Precipitation anomalies during these episodes are less pronounced and spatially
65 heterogeneous over the Tibetan Plateau. The simulated changes are in good
66 agreement with available paleo-climatic reconstructions from proxy data. The present-
67 day climate classification is only slightly sensitive to the changed boundary conditions
68 in the Quaternary Quaternary. It is shown that in some regions of the Tibetan Plateau



69 the climate anomalies during the Quaternary Quaternary have been as strong as the
70 changes occurring during the uplift period.

71

72 **Keywords:** Tibetan Plateau, dynamical downscaling, paleo-climate, surface uplift,
73 Last Glacial Maximum, mid-Holocene

74



75 **1 Introduction**

76

77 The Tibetan Plateau (TP) is the highest and largest modern elevated plateau and a
78 primary feature of Earth's climate system. Over the last millions of years it has
79 undergone remarkable climate changes, especially during the Miocene-Pliocene uplift
80 period (Clift and Plumb 2008, Mielke et al. 2013, Clift et al. 2014) and in the interplay
81 of glacial and interglacial epochs (Braconnot et al. 2012, Ruddiman 2014). The TP
82 represents a dominant driver of climate variability all around the globe (Molnar et al.
83 2005). In particular, the thermodynamic and mechanical effect of the TP on the strength,
84 spatial extent and seasonal evolution of the South and East Asian monsoon systems
85 makes it an outstanding factor for low-frequency climate variations (Molnar and
86 Rajagopalan 2012, Park et al. 2012, Tang et al. 2013a). The TP is also a key region of
87 investigation for the current debate on anthropogenic climate change. Climate
88 reconstructions and meteorological data attest tremendous warming rates and record
89 high temperatures during recent decades to the TP region (Guo and Wang 2012,
90 Guangliang et al. 2013) and climate model projections identify one of the hot spots of
91 future climate change in the mountainous areas of Central Asia (Yang et al. 2012,
92 IPCC 2013, Mannig et al. 2013).

93 The assessment of the anthropogenic interference with climate and its relevance for
94 ecosystems and society requires that the magnitude of natural climate variations and
95 their underlying mechanisms are well understood (e.g. Ruddiman 2014). Given this,
96 this study is dedicated to several outstanding paleo-climatic episodes that we compare
97 in a consistent modeling framework. A number of sensitivity studies with a high-
98 resolution regional climate model (RCM) are presented to simulate climate equilibrium
99 states: (1) during different stages of surface uplift of the TP, (2) during the last glacial



100 maximum (LGM), and (3) during the mid-Holocene (MH) climate optimum. These
101 episodes have been identified as major time slices of paleo-climatic variability
102 worldwide and in the TP region, in particular (e.g. Braconnot et al. 2012, Mielke et al.
103 2013, Clift et al. 2014, Ruddiman 2014).

104 In this study, we complement previous work with a focus on three objectives. The first
105 objective of our study addresses the former hypothesis from Prell and Kutzbach (1992),
106 suggesting that glacial versus interglacial boundary conditions during the Quaternary
107 can produce climate anomalies across the TP that are in the same order of magnitude
108 as during Cenozoic (last ~55 Ma) surface uplift. For this purpose, we compare these
109 different episodes for the first time by means of state-of-the-art climate models using a
110 coherent experimental design.

111 The second objective pertains to the generation of high-resolution patterns of paleo-
112 climate for the TP and surrounding regions. Typically, paleo-climatic episodes are
113 derived from the interpretation of proxy-data based on tree rings, sedimentary archives,
114 and stable isotope methods (Bartlein et al. 2011, Griesinger et al. 2011, Bershaw et al.
115 2012, Mischke et al. 2013, Saraswat et al. 2013). Here, we produce simulated climate
116 equilibrium states, prescribing the specific boundary conditions for the various paleo-
117 climatic episodes, which serve as a theoretical basis and benchmark for the
118 interpretation of proxy-data. In addition, some episodes during the early uplift period
119 are addressed where proxy-data are rarely available (Clift and Plumb 2008).

120 Our third objective aims at improving the spatiotemporal representation of paleo-
121 climate across Asia and the TP by relying on a high-resolution RCM.

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125 **2 Background and previous related work**

126

127 The effect of the TP uplift on local and regional climate, with particular focus on the
128 Indian summer monsoon, was investigated in earlier work (Hahn and Manabe 1975),
129 revealing the general relationship between uplift and cooling and drying on the TP as
130 well as a strengthening of the South and East Asian monsoons (cf. Kutzbach et al.
131 1989, 1993). Since then, these basic mechanisms and climate responses have been
132 corroborated by a number of modeling studies with global models using increasing
133 resolution and complexity (e.g. Kitoh et al. 2010, Molnar and Rajagopalan 2012, Park
134 et al. 2012, Wu et al. 2012, Botsyun et al. 2016). Recently, Ma et al. (2014) have
135 demonstrated that the elevation of the Himalaya is more relevant to the changing
136 climate patterns in the Asian monsoon regions than the height of the TP itself. Zhang
137 et al. (2014) have prescribed several regionally differentiated uplift scenarios and draw
138 a rather heterogeneous picture of the climate response to surface uplift. Common to
139 these studies is that they are based on coarse-grid climate models which, probably,
140 cannot account for the finer-scale characteristics of lower atmospheric circulation,
141 precipitation and height-dependent temperature (Braconnot et al. 2012). The deficits
142 of state-of-the-art global climate models for the assessment of present-day climate on
143 the TP have been highlighted by Su et al. (2013). Harrison et al. (2014) have compared
144 paleo-climatological simulations from various climate models with available proxy-data
145 and pointed to a number of substantial biases over Asia. Kitoh et al. (2010) reported
146 some added value when they enhanced the model horizontal resolution of their uplift
147 experiments to 120 km. Finally, work by Botsyun et al. (2016) applied an isotope
148 tracking global general circulation model (LMDZ-iso) to investigate the effect of surface
149 uplift on precipitation $\delta^{18}\text{O}$. They document significant change in the isotopic



150 composition of precipitation during surface uplift, and estimate only ~40% of existing
151 stable isotope based paleoaltimetry sample locations contain a signal of the plateau
152 surface uplift history. The work of Botsyun et al. (2016) highlights the importance of
153 paleoclimate studies for understanding the tectonic, surface uplift, history of large
154 orogenic plateaus.

155 As a consequence, RCMs have also been applied to paleo-climatological issues in the
156 TP and Asian monsoon regions. Zheng et al. (2004) have addressed the MH period at
157 a rather low resolution of 120 km. The LGM and MH have been compared by Ju et al.
158 (2007) who revealed an added value of their RCM at 60 km resolution with respect to
159 the driving global model. A similar result was obtained by Polanski et al. (2012) at 50
160 km resolution for the MH episode. Liu et al. (2008a,b) have used an RCM with a lower
161 resolution of 90 km but accounted for large number of individual and combined forcings,
162 indicating that changing vegetation, sea level and large-scale atmospheric circulation
163 play a more important role in glacial climate than greenhouse gas concentrations and
164 orbital parameters. In terms of the surface uplift process, Song et al. (2009) have
165 carried out a sequence of experiments with stepwise elevation increase but restricted
166 their simulation period to several months only. More recent RCM simulations with low
167 resolution but spatially differentiated uplift scenarios highlighted the crucial importance
168 of the Himalayan barrier effect for the Indian monsoon and the predominating impact
169 of the TP elevation on the East Asian monsoon system (Tang et al. 2013a), whereas
170 interannual monsoon variability is hardly affected by the uplift stages (Tang et al.
171 2013b).

172 In order to meet our main objectives, we pursue a dynamical downscaling approach
173 nesting the RCM REMO at 0.5° (~55 km) resolution into the global climate model
174 ECHAM5 (Li et al. 2016a,b; Mutz et al., 2016). For all paleo-climatic episodes the same



175 model setup is used in terms of domain, resolution, model physics and driving model,
176 albeit with different boundary conditions for surface uplift magnitude, LGM and MH
177 (see section 2). This approach allows for direct comparison of the resulting climate
178 anomalies with each other. Sensitivity studies over 12 to 40 years are realized which
179 ensures climate equilibrium states of sufficient length for a thorough statistical (cluster)
180 analyses. REMO has been successfully applied and evaluated for climatological issues
181 in Europe (Jacob et al. 2001), Africa (Paeth et al. 2005, 2009), South East Asia (Saeed
182 et al. 2011) and Central Asia (Mannig et al. 2013). In the paleo-climatological context,
183 it has been used in a former version with stable isotopes for uplift scenarios in the
184 Andean region (Insel et al. 2012). Paxian et al. (2016) have recently identified an added
185 value of REMO and other RCMs when it comes to decadal predictability in the West
186 African monsoon system. In the present study, REMO is applied for the first time to a
187 paleo-climatological problem in Asia, assuming that it represents a valuable and
188 efficient tool to generate high-resolution patterns of paleo-climatic episodes to be
189 compared with proxy- data.

190 The subsequent section is dedicated to the description of the model, the experimental
191 design, the validation data and the statistical methods. The results of our analysis are
192 presented in section 4 and discussed in section 5. The last section comprises the main
193 conclusions and a brief outlook.

194

195 **3 Modeling design, data and methods**

196

197 The dynamical downscaling is undertaken with the hydrostatic RCM REMO (Jacob
198 2001). It is nested in the atmospheric global climate model ECHAM5 as atmospheric
199 part of the Max-Planck Institute for Meteorology coupled general circulation model



200 ECHAM5/MPI-OM (Jungclaus et al. 2006). For all paleo-climatic episodes REMO is
201 run over the same model domain, extending from 50°E to 130°E and from the equator
202 to 60°N (see Fig. 1). In addition, the same horizontal and vertical resolution is used,
203 amounting to 0.5° and 27 levels, respectively. The current REMO version is based on
204 the improved land surface scheme from Hagemann and Dümenil (2003) and has been
205 successfully employed in various regions of the globe (e.g. Paeth et al. 2009, Saeed
206 et al. 2011, Insel et al. 2012, Mannig et al. 2013). The REMO setup used here has
207 been extensively evaluated by means of a number of sensitivity studies. Basically two
208 types of experiments have been performed which are described below.

209 The first type of REMO experiments addresses the uplift process of the TP. For this
210 purpose, the RCM in 0.5° resolution is nested into AMIP-type global simulations with
211 ECHAM5 using a horizontal spectral resolution of T63 (~1.875°), 19 vertical levels and
212 observed sea surface temperature and sea ice data from GISST (Rayner et al. 1996).
213 The uplift is accounted for by a stepwise approach, setting the model topography in
214 ECHAM5 and REMO to 100%, 75%, 50%, 25% and 0% of the present-day elevation.
215 According to current geological knowledge (e.g. Molnar et al. 2005, Mielke et al. 2013,
216 Clift et al. 2014), the topography is modified in the domain indicated by the red line in
217 the top panel of Fig. 2, comprising the TP itself, the Himalaya, the Pamir, the Hindukush,
218 the Tian Shan and the Southeast Asian highlands. In contrast, the Altai region in the
219 northern part of our model domain was built prior to the TP uplift and, hence, is not
220 modified (cf. Wang et al. 2006). In order to avoid an artificial trough in Central Asia, a
221 minimum height of 500 m above sea level is prescribed when reduced topography falls
222 below this threshold. This is in agreement with Le Pichon et al. (1992) who suggested
223 a base height of at least 500 m for the TP area prior to plateau uplift in the Cenozoic.
224 Along the borders of the region with modified elevation, the discontinuous transition



225 between modified and unmodified topography is smoothed by interpolation. The
226 resulting model topographies for the control run (100% elevation) and the 4 sensitivity
227 studies with stepwise reduction of elevation are displayed in Fig. 2. It is obvious that
228 for each simulation a continuous land surface without unnatural cracks and slopes is
229 given. In addition, a part of the present-day relative topography across the TP region
230 is retained. Note that we use a rather idealized uplift scenario without regionally
231 differentiated uplift rates (cf. Tapponnier et al. 2001, Tang et al. 2013a, Zhang et al.
232 2014) nor a modification of the land-sea mask (cf. Fluteau et al. 1999) in order to
233 facilitate the physical interpretation of the climate response to the stepwise uplift
234 process. The control run with present-day elevation covers the period 1961-2000, the
235 other time slices are also based on 40-year model periods. For validation and statistical
236 analysis the first 10 model years are skipped in order to remove some slight spin-up
237 effects during the first model years, as found in temperature and moisture of the
238 deepest soil layer.

239 The second type of REMO experiments pertains to the LGM and MH episodes. In this
240 case, REMO is nested in a higher-resolution ECHAM5 version with T106 (~1.125°)
241 and 31 vertical levels (Werner et al. 2011). This ECHAM5 version includes stable
242 isotopes and has been analyzed in terms of $\delta^{18}\text{O}$ trajectories (Li et al. 2016a,b) and
243 governing factors of $\delta^{18}\text{O}$ (Mutz et al. 2016). The boundary conditions for the LGM
244 and MH periods have been prescribed equally in ECHAM5 and REMO. This includes
245 greenhouse gas concentrations, land surface characteristics, land-sea mask due to
246 sea level changes (in ECHAM5), and orbital parameters (cf. Ruddiman 2014). The
247 specific boundary conditions for LGM and MH are described in detail by Dietrich et al.
248 (2013). They are in agreement with the guidelines of the Paleoclimate Modelling
249 Intercomparison Project Phase 3 (PMIP3, Arnold et al. 2009). Oceanic boundary



250 conditions and sea ice coverage for present-day, LGM and MH are taken from a
251 coupled ECHO-G model experiment (Lorenz and Lohmann 2004). In REMO, the
252 modified land cover classes are translated into monthly varying seasonal cycles of
253 background surface albedo, fractional vegetation cover and leaf area index in order to
254 account for the seasonal evolution of the Asian monsoon systems in a more
255 appropriate way. Given the modified ECHAM5 resolution and model settings in
256 ECHAM5 and REMO, an additional control run is realized covering the period 1978-
257 1989. 12-year model periods are also applied to the LGM and MH episodes, retaining
258 the last 10 years for model validation and statistical analysis of climate anomalies.
259 For the validation of the present-day climate simulated by ECHAM5 and REMO, we
260 rely on monthly temperature and precipitation data from the CRU data set, a gridded
261 data product based on station observations over land (Mitchell and Jones 2005), from
262 the NCEP (Kalnay et al. 1996) and ERA40 (Uppala et al. 2005) reanalyses, and from
263 the High Asia Reanalysis (HAR), a regional reanalysis with the RCM WRF at a
264 resolution of 0.3° covering the period 2000-2012 (Maussion et al. 2014) which
265 represents a benchmark for state-of-the-art high-resolution data sets for the TP region.
266 Satellite-based precipitation products such as TRMM have not been considered due
267 to distinct biases over the Himalaya and TP (cf. Barhi and Singh 2015). The LGM and
268 MH paleo-climatic episodes are compared with quantitative reconstructions of
269 temperature and precipitation from pollen-based proxy-data available over several 2°
270 $\times 2^\circ$ grid boxes across the TP and eastern Asia (Bartlein et al. 2011). This data set is
271 also used as a reference for PMIP3 paleo-simulations.
272 The statistical significance of climate changes due to the uplift process and to the LGM
273 and MH boundary conditions, respectively, is assessed by means of a one-way
274 analysis of variance (Wilks 2006), taking the offset of the means due to the changing



275 boundary conditions as treatment effect and the year-to-year changes within the
276 simulated time series as internal variability.

277 A special focus of our study is on the TP region and on the question whether climate
278 types have noticeably changed due to the various forcings. Therefore, a classification
279 of the Tibetan climate is carried out by a hierarchical cluster analysis combined with a
280 subsequent non-hierarchical k-means correction approach, dedicated to achieve
281 homogeneous clusters which are sharply distinguishable from each other (cf. Paeth et
282 al. 2005, Wilks 2006, Geng et al. 2014, Netzel and Stepinski 2016). The cluster
283 analysis is applied to the domain indicated by the rectangle in Fig. 1. It is based on
284 seven atmospheric variables, including annual-mean 2 meter temperature, annual
285 precipitation amount, annual temperature range, 10 meter zonal and meridional wind
286 components in January and July. Eight clusters have been found to provide the best
287 result in terms of interpretable and discriminable climate types across the TP and
288 surrounding regions.

289

290 **4 Results**

291

292 **4.1 Model validation**

293 A comparison of annual temperature and precipitation averaged over the cluster
294 domain (see Fig. 1) between REMO, ECHAM5 and various observational data sets is
295 depicted in Fig. 3. The validation period is 1978-1989. The respective control run from
296 REMO marks the origin of the graph and all other data sets are indicated as anomalies
297 with respect to this control run. Obviously, the considered gridded data and reanalyses
298 differ substantially. Especially, CRU is much drier in the TP region and NCEP
299 noticeably cooler than ERA40 and HAR. The uncertainty on behalf of the observational



300 data is at least as pronounced as the potential bias of the RCM REMO. In fact, REMO
301 is quite close to the ERA40 reanalysis and the HAR data set. Note that the latter
302 certainly represents a benchmark for present-day climate on the TP, but the available
303 data period 2000-2012 is rather short and does not coincide with the simulation periods
304 of ECHAM5 and REMO. The driving global climate model ECHAM5 is characterized
305 by a strong wet bias on the TP. This bias is considerably reduced by dynamical
306 downscaling and points to a clear added value of REMO compared with ECHAM5.
307 This is in line with the study by Mannig et al. (2013) who have validated REMO
308 simulations driven by ERA40 reanalyses and ECHAM5/MPI-OM coupled climate
309 model simulations over entire Asia and concluded that the RCM performs well over the
310 continent and exhibits an added value with respect to the coarser-grid global climate
311 model.

312 A more detailed validation of REMO in the model domain and version presented here
313 has been conducted in the framework of a PhD project (unpublished PhD thesis in
314 2015 by Christian Steger). It has been demonstrated that REMO is able to reproduce
315 the observed main characteristics of climate on the TP and in surrounding regions.
316 Especially temperature at high-elevation sites is well captured and clusters of distinct
317 climate types agree almost perfectly between REMO, ERA40 and HAR. Therefore, it
318 is assumed that REMO provides a reliable basis to simulate paleo-climatic changes.

319

320 **4.2 Surface uplift experiments**

321 The temperature changes associated with the surface uplift of the TP and surrounding
322 mountain areas are illustrated in Fig. 4. As the plateau itself and the Asian monsoon
323 systems are subject to a distinct seasonal regime, results are shown for annual means,
324 winter means (DJF) and summer means (JJA). Indeed, the annual temperature range



325 increases from south to north and amounts to more than 40°C between winter and
326 summer in Central Asia and in parts of the TP (Fig. 4, top panels). It is also obvious
327 that near-surface temperature is a stringent function of elevation. Thus, reduced
328 elevation leads to higher temperature over the TP, simply due to the thermal
329 stratification of the atmosphere (Fig. 4, panels d-o). The warming amounts to more
330 than 10°C in the experiment with 0% of present-day elevation (with a minimum height
331 of 500 m on the TP). Interestingly, it is a nonlinear function of reduced topography
332 since the warming exhibits larger magnitudes in the 75% and 50% experiments than
333 in the 25% and 0% experiments. On the TP itself, warming rates in winter and summer
334 are virtually the same. Central Asia is marked by a slight cooling and the continental
335 desert region north of the TP by a noticeable cooling in both seasons, amounting to
336 more than 7°C, especially in winter. Later on, this will be explained by precipitation and
337 circulation changes. Over South Asia, reduced elevation implies lower temperature in
338 winter and slightly warmer conditions in summer. While the first effect is associated
339 with enhanced cold air advection from the continent due to the missing barrier of the
340 Himalaya and TP, the latter effect is linked to reduced precipitation amount (see below).
341 Thanks to the high resolution of REMO, the resulting climate change patterns are
342 characterized by numerous regional details such as the basin structure on the
343 northeastern TP, the fine-scale changes in the Tian Shan mountain range and the
344 desert basins in western China.

345 The impact of the stepwise uplift on annual, winter and summer precipitation is
346 portrayed in Fig. 5. In the South and East Asian monsoon regions precipitation is
347 mainly a summer phenomenon, whereas the mid-latitudes in Asia follow a more
348 equalized seasonal regime (Fig. 5, top panels). The precipitation response to reduced
349 elevation is marked by a spatially heterogeneous picture (Fig. 5, panels d-o). The TP,



350 South Asia, eastern China and the southern part of Central Asia exhibit drier conditions.
351 In the monsoon areas this amounts to less than 50%. In Central Asia the relative drying
352 is more important. However, the most striking feature in the precipitation response is
353 simulated over the continental arid regions of western China and northwestern Central
354 Asia. Indeed, precipitation increases by up to 250%, when the plateau height is
355 reduced to 25%. This rises to up to four times more than the present-day climatology
356 in some arid regions and is related to circulation changes (see below). Again, the major
357 changes occur up to a reduction to 25% of present-day elevation, while the 0%
358 experiment only increases the precipitation signal on the TP itself. An interesting
359 regional detail is given by the wetter conditions in winter along southeastern India and
360 Southeast Asia, where the missing barrier effect of the Himalaya and TP leads to more
361 relief rainfall from the continental winter monsoon flow.

362 To assess the importance and statistical significance of temperature and precipitation
363 changes due to the uplift process, a one-way analysis of variance has been applied
364 using pairs of uplift experiments or all uplift experiments at once. The analysis of
365 variance reveals that the stepwise uplift scenario accounts for more than 90% of total
366 annual temperature variability on the TP (Fig. 6, top panels). This implies that the
367 topographic forcing between the 30-year time slices dominates by far the internal year-
368 to-year fluctuations of temperature within the time slices. This effect is already
369 achieved in the 75% experiment. In the surrounding regions, the uplift effect is less
370 pronounced, ranging between 20% and 80% of explained variance. Nevertheless, the
371 temperature changes are statistically significant in almost the entire model domain. For
372 precipitation the picture is more dispersed and the uplift effect is much less pronounced
373 in the 75% experiment: Precipitation is characterized by substantially stronger internal
374 variability than temperature (e.g. IPCC 2013) and, hence, the uplift process accounts



375 for a smaller portion of total variance. Nonetheless, the explained variance reaches up
376 to 90% in the arid regions of western China, where the uplift comes along with a
377 multiplication of present-day precipitation amount.

378 In the following, the substantial drying of the desert basins of western and northern
379 China during the uplift phase is analyzed in more detail. At first sight, this may be
380 interpreted as an increasing barrier effect of the topography, hindering the moist
381 summer monsoon flow from the northern Indian Ocean to penetrate far into the Asian
382 continent. However, a deeper look on the wind components in various atmospheric
383 levels before and after the uplift reveals that in none of the uplift experiments the
384 monsoon air masses propagate into the arid regions of western China, rather these
385 are characterized by extratropical westerlies independent of the topography (not
386 shown). The clue to the problem is given by changes in the extratropical wave train.

387 Fig. 7 displays the mean position of troughs and ridges in the 500 hPa geopotential
388 height field under present-day conditions (top panel) and the respective changes in the
389 50% and 0% experiments (bottom panels). Currently, long-term mean geopotential
390 height is higher over Central Asia and lower over northeastern Asia, indicating the
391 typical position of ridges and troughs, respectively, in the extratropical Rossby waves.

392 Apparently, reduced elevation of the TP implies a substantial large-scale decrease of
393 geopotential height over Central Asia, leading to more advection of moist air masses
394 at the forefront of the trough which is located exactly over the region where the
395 precipitation increase peaks in northwestern China. This appears to be the key
396 mechanism of how the uplift process in the TP regions affects extratropical climate all
397 around the globe (cf. Molnar et al. 2005).

398 The last aspect addressed from the uplift experiments pertains to the identification of
399 distinct climate types on the TP and adjacent regions and their sensitivity to the uplift



400 process. In the high-dimensional climate system, climate types often allow for a more
401 comprehensive interpretation of climate changes than individual grid boxes (e.g. Paeth
402 et al. 2005, Geng et al. 2014, Netzel and Stepinski 2016). Fig. 8 presents a climate
403 classification based on eight clusters derived from the 30-year REMO control run (top
404 panel). The clusters are arranged from warm climates (red) to cold climates (blue). The
405 remaining panels refer to the experiments with reduced elevation, but use the cluster
406 centroids from the control run. Under present-day conditions, the desert Tharr in
407 western India occupies cluster 1 which is characterized by very high temperature and
408 low annual rainfall amount. Northwestern India is marked by cluster 2 with still rather
409 high temperature but more humid conditions. Cluster 3 represents the southern slopes
410 of the Himalaya where climate is cooler but annual precipitation exceeds 3000 mm/yr.
411 Cluster 4 is characterized by intermediate temperature and very dry conditions and
412 covers the continental deserts with less than 100 mm of annual precipitation,
413 particularly in the Tarim basin. Cluster 5 extends along the foothills on the northern
414 side of the TP, the Tian Shan and the Pamir as well as over the less arid regions of
415 western China. It denotes a colder and more humid climate than cluster 4. The
416 southern and eastern part of the TP is given by cluster 6. Annual mean temperature is
417 below 0°C and rainfall exceeds 1500 mm per year. This cluster denotes those parts of
418 the TP on the wind-ward side that are more influenced by moist air advection from the
419 South Asian summer monsoon. In contrast, cluster 8 stands for the much drier and
420 even colder northern part of the TP. Cluster 7 is rather peculiar: it spreads over those
421 areas of the plateau that are marked by a large number of lakes. In the vicinity of these
422 lakes, REMO simulates a climate that is as cold as the one in cluster 8 but with
423 precipitation totals of up to 2000 mm/year. Later on, this cluster will be identified as a
424 model artefact.



425 The experiments with reduced topography manifest a stepwise shift of these climate
426 types towards a setting where the TP exhibits a much warmer and drier climate. North
427 of the TP, climate switches from the continental desert type to a much colder and more
428 humid type which is confined today to the lee side of the TP. Altogether, the climate
429 classification of the 0% experiment leads to a pattern which fundamentally differs from
430 the one under present-day conditions, highlighting the tremendous impact of the uplift
431 process on climate types on and around the TP. It is also intriguing that the shift of the
432 clusters mainly takes place when the elevation is reduced to 25% and 0% of the current
433 topography.

434

435 **4.3 Mid-Holocene and LGM experiments**

436 The simulated temperature anomalies during the LGM and MH episodes are illustrated
437 in Fig. 9. Again, it is differentiated between annual, winter and summer means. Note
438 that the control run for these time slices is slightly different from the one for the uplift
439 experiment and only 10 years per run are available for analysis because the ECHAM5
440 simulations used for the lateral boundary conditions have much higher resolution and,
441 thus, have been computationally more expensive. It is obvious that the temperature
442 response during the LGM period is much more pronounced than during the MH time
443 slice. The LGM exhibits lower temperatures in virtually the entire model domain. The
444 cooling amounts to 3-8°C with largest amplitude on the TP.

445 During the MH, the picture is more heterogeneous and temperature changes do not
446 exceed 3°C. Most regions undergo slightly colder conditions, except for the
447 extratropical part of Asia in summer. Again, the cooling effect peaks over the TP and,
448 in particular, along the southern foothills of the Himalaya.



449 The precipitation response during the LGM and MH periods is spatially rather
450 incoherent (Fig. 10). During the LGM, drier conditions prevail in most parts of Asia,
451 especially in Central Asia and the Middle East. Higher-elevation sites with southward
452 exposition along the Himalaya, the Pamir, the Tian Shan and the Altai may have
453 experienced more precipitation. The most striking feature, however, is the enormous
454 rainfall increase in winter by more than 100% over eastern India, Bangladesh and parts
455 of Southeast Asia. This increase during the northeasterly winter monsoon flow is
456 contradictory to the present-day seasonality. We have no reasonable physical
457 explanation for such a process and assume it to be a model artefact, possibly arising
458 from a misrepresentation of the prescribed oceanic boundary conditions in the Bay of
459 Bengal that produces a prominent wet bias inland (cf. Paeth et al. 2005).

460 The MH episode appears to be subject to slightly more annual precipitation amount in
461 most parts of the model domain, except for Central India and the Middle East. This
462 increase mainly takes place during the summer season, initiated by a stronger
463 monsoon flow in East Asia and more advection of moist air masses by westerlies in
464 the extratropics. The pattern of summer precipitation changes is consistent with the
465 temperature response in Fig. 9: higher temperatures north of the TP provoke a
466 northward shift of the summer monsoon flow with more rainfall along the Himalaya and
467 the TP. Vice versa, higher rainfall in northern India and over the foothills of the
468 Himalaya implies a cooling in these areas.

469 The analysis of variance demonstrates that the impact of LGM and MH boundary
470 conditions on temperature and precipitation does not everywhere stand out from
471 internal variability (Fig. 11). During the LGM period, the effect on near-surface
472 temperature is outstanding with more than 90% of explained variance, especially on
473 the TP, in Central Asia and the Middle East. In contrast, the MH episode comes along



474 with a weaker temperature response, which peaks over South and Southeastern Asia.
475 Only a part of the TP region is significantly affected by the MH boundary conditions.
476 Typically, the forcing accounts for 20% to 70% of total temperature variance. For
477 precipitation, the picture is more dispersed: the MH is barely characterized by a
478 coherent rainfall response pattern and, thus, the changes denoted in Fig. 10 should
479 not be over-interpreted, except for the wetter conditions in the southern and western
480 part of the TP. The LGM only leaves a mark in western and southeastern Asia where
481 precipitation is below present-day totals. However, the explained variance ranges
482 mostly below 50%.

483 In Fig. 12 the simulated annual temperature and precipitation changes during the LGM
484 and MH episodes are compared with available quantitative reconstructions from
485 pollen-based proxy-data (Bartlein et al. 2011). Note that in contrast to Fig. 10 the
486 precipitation response is expressed in mm per year and not in percentage. The MH
487 period is much better represented by proxy-data than the LGM (cf. Clift and Plumb
488 2008, Bartlein et al. 2011). During the LGM episode, there is a good agreement
489 between simulated and reconstructed temperature and precipitation, although this
490 comparison is based on a small number of grid boxes. Colder and drier conditions over
491 China and Central Asia are likewise suggested by REMO and proxy-data. The two
492 neighboring reconstructed pixels over the western Tian Shan with opposite
493 temperature changes obviously reflect a data artefact. Unfortunately, the simulated
494 phenomenon of substantial rainfall increase north of the Bay of Bengal (see Fig, 10)
495 can neither be confirmed nor falsified by proxy-data by reason that they are not
496 available in this region.

497 During the MH, the pattern is more complex: reconstructions indicate a small-scale
498 coexistence of warmer and colder conditions over the TP, China and Central Asia,



499 whereas in REMO a slight cooling prevails over most of the model domain. On the part
500 of the reconstructions, the temperature response varies by up to 10°C, partly with
501 opposite sign, over a distance of some hundreds of kilometers. This reflects a large
502 amount of uncertainty in the proxy-data, while the RCM produces a spatially coherent
503 response pattern. In terms of precipitation, there is a satisfactory agreement between
504 model and reconstructions: wetter conditions prevail over the TP, China and parts of
505 Central Asia.

506 Finally, Fig. 13 clearly reveals that the spatial distribution of main climate types over
507 the TP and surrounding regions is more or less insensitive to the modified boundary
508 conditions during the LGM and MH periods. Note that in this case the reference
509 clusters for present-day climate have been derived from a shorter period and, hence,
510 the pattern differs slightly from the one in Fig. 8. Under LGM and MH conditions the
511 climate classification barely changes. This is somewhat contradictory to the simulated
512 temperature and precipitation changes presented in Figs. 9 and 10, especially with
513 respect to the considerably colder conditions during the LGM. The fact that the clusters
514 still remain stationary is a clear indication that the other atmospheric variables entering
515 the cluster analysis – i.e. wind components, annual temperature range and, to a certain
516 extent, precipitation – are hardly affected by the modified boundary conditions during
517 LGM and MH. This is a major difference from the uplift experiments (see Fig. 8).

518

519 **4.4 Comparison of episodes**

520 In the last step of our analysis the annual temperature and precipitation changes under
521 uplift, LGM and MH conditions are directly compared with each other, once as regional
522 means over the entire cluster domain and once as averages over each individual
523 cluster (Fig. 14). In addition, the results from REMO and ECHAM5 are compared with



524 each other in order to assess which model type is associated with a higher sensitivity.
525 Over the cluster domain, the strongest temperature and precipitation response is found
526 for the uplift experiments, denoting a noticeably drier and warmer climate on and
527 around the TP (top left panel). There is an offset between ECHAM5 and REMO (cf.
528 Fig., 3) but the response is very similar. In both models, the largest changes occur
529 between the 100%, 75% and 50% experiments, whereas the uplift from 0% to the 25%
530 simulation plays a minor role. The LGM and MH affect the mean climate in the cluster
531 domain to a much lower extent. The response in REMO and ECHAM5 is again similar,
532 with one exception: REMO simulates slightly wetter conditions, mainly due to the
533 outstanding precipitation increase over the southeastern TP (see Fig. 10), while
534 ECHAM5 suggests a drying.

535 Fig. 14 also indicates that the striking climate changes during the uplift process are
536 mainly performed by clusters 3, 6, 7 and 8. These clusters represent the higher-
537 elevation sites and southward slopes of the TP and Himalaya. The changes in cluster
538 3 and 8 even lie outside the illustrated precipitation and temperature range,
539 respectively, highlighting the enormous drying along the southern slope and warming
540 on the northern TP at reduced elevation. In addition, clusters 4 and 5 experience a
541 substantial precipitation increase in the 25% and 0% experiments, suggesting that the
542 continental dry climate north of the TP has not yet existed during the early uplift phase
543 (cf. Fig. 5). This effect is only simulated by REMO at higher spatial resolution. LGM
544 and MH have mainly affected the clusters 3, 4, 5 and 6. Clusters 4 and 5 are
545 representative of the arid and semi-arid climate types north of the TP. Nonetheless,
546 temperature over the TP (clusters 7 and 8) also decreases considerably under LGM
547 conditions. In most cluster regions around the TP, the climate changes during LGM
548 and MH range are of the same order of magnitude as during the uplift process. In



549 contrast, the TP itself is much more sensitive to the plateau uplift and, hence, stronger
550 signals should be expected in proxy-data for that earlier period.

551

552 **5 Discussion**

553

554 This study provides new insight into prominent Cenozoic and Quaternary paleo-
555 climatic variations in Asia and, more specifically, in the TP region. The comparative
556 validation of the RCM REMO and the driving global climate model ECHAM5 has shown
557 that the higher-resolution model is closer to the available observational data and
558 reanalyses, particularly with respect to precipitation. This is consistent with other
559 studies highlighting the added value of higher-resolution climate models in Asia and
560 other regions of the globe (e.g. Kitoh et al. 2010, Paeth et al. 2011, Mannig et al. 2013,
561 Paxian et al. 2016).

562 The uplift experiments with stepwise reduction of topography describe a continuous
563 transition towards a warmer and drier climate on the TP when elevation decreases.

564 This general tendency has been found in virtually all uplift experiments since the
565 pioneering work by Hahn and Manabe (1975) and Kutzbach et al. (1989, 1993).

566 Another earlier finding, which we could corroborate with this study, is the fact that the
567 later uplift process has played a more important role for the TP climate and surrounding
568 monsoon systems than the early phase. In our study, the changes are stronger
569 between the 100%, 75% and 50% experiments than between the 25% and 0%
570 simulations. Kitoh et al. (2010) have reported that their 60% experiment represented a
571 threshold for accelerated climate change. A novel aspect of our modeling design
572 pertains to the high spatial resolution and, at the same time, the relatively long
573 simulation periods for the time slices, which allow for a more sophisticated statistical



574 assessment of changes and their significance. A new finding arising from the high
575 resolution of the REMO model concerns the appearance of the continental-desert type
576 of climate that first occurred in the transition from the 25% to the 50% experiment. This
577 process is probably related to dynamical changes in the extratropical Rossby wave
578 configuration. The identification of climate types by means of cluster analysis has
579 proven to be an instructive tool to illustrate changes at the scale of regional climates
580 (cf. Geng et al. 2014, Netzel and Stepinski 2016). In fact, it could be shown that the
581 uplift process induces a fundamental rearrangement of present-day climate types on
582 the TP and in surrounding regions, especially to the north.

583 During the LGM period, REMO simulates colder conditions in most regions of Asia with
584 the largest temperature decrease in winter over the TP and in the northernmost part of
585 the model domain along 60°N. This is in agreement with previous modeling studies
586 (e.g. Liu et al. 2008a,b, Harrison et al. 2014) and reconstructions (e.g. Mischke et al.
587 2013, Saraswat et al. 2013). The cooling in our model ranges between 3°C and 8°C
588 on the TP which agrees well with proxy-data from Bartlein et al. (2011). In contrast,
589 other authors have stated that the LGM cooling rate is underestimated by their global
590 (Harrison et al. 2014) and regional (Zheng et al. 2004, Ju et al. 2007) climate models.
591 In terms of precipitation, the pattern is dominated by drier conditions in most of the
592 model domain. However, precipitation amount over the TP itself is mainly unaffected,
593 except for an extensive increase in the southeastern part, which we assign to a model
594 artefact. In general, precipitation changes are strongly superimposed by internal
595 variability within the time slices. Despite the noticeable cooling tendency, the spatial
596 arrangement of climate types during the LGM episode can hardly be distinguished from
597 the present-day situation, mainly because atmospheric circulation is more or less
598 unchanged.



599 The temperature effect during the MH period is minor compared with the other paleo-
600 climatic time slices. A slight cooling of less than 1°C prevails over most of Asia. Over
601 the TP and northern India it reaches up to 3°C. During summer, the extratropical part
602 of the model domain experiences some warming by 1-2°C. While this minor cooling
603 tendency in REMO is consistent with the majority of temperature reconstructions in
604 central and eastern Asia (cf. Bartlein et al. 2011, Braconnot et al. 2012), it disagrees
605 with former modeling approaches for the MH period which suggest a slight warming
606 (Zheng et al. 2004, Polanski et al. 2012, Harrison et al. 2014). Concerning precipitation,
607 our model confirms the wetter conditions over the TP, northern India, eastern Asia and
608 northern China as suggested by other authors (e.g. Zheng et al. 2004, Polanski et al.
609 2012, Saraswat et al. 2013). Yet, the precipitation changes hardly stand out from the
610 internal variability within the MH time slice. The climate classification during the MH is
611 also largely congruent with the one under present-day conditions.

612 The application of a uniform downscaling approach with an RCM using the same model
613 domain, resolution and model physics allows for a direct comparison of climate change
614 signals among all considered time slices, which is unprecedented for uplift, LGM and
615 MH episodes. Not surprisingly, the uplift process is associated with the largest
616 temperature and precipitation response and shift of climate types, at least on the basis
617 of the entire TP region. These regional-mean changes hardly differ between REMO
618 and ECHAM5. In both models, the uplift mainly affects the high-elevation clusters over
619 the TP. However, the disappearance of the arid climate types north of the TP in the
620 25% and 0% experiments is a feature that only occurs at higher resolution in REMO.
621 Another interesting finding is that at the scale of individual clusters, i.e. regional
622 climates, glacial and, to a lower extent, MH conditions can provoke climate anomalies
623 in the same order of magnitude as the uplift process. This mainly pertains to the climate



624 types in northern India and along the southern slopes of the Himalaya. Thus, the former
625 hypothesis by Prell and Kutzbach (1992) can be supported by our model results.

626

627 **6 Conclusions**

628

629 The main objectives of the present study were to assess and compare paleo-climatic
630 changes in Asia and, especially, in the TP region. For this purpose, we have carried
631 out a series of RCM experiments that are characterized by high spatial resolution and
632 sufficiently long simulation periods to account for the impacts of uplift, LGM and MH
633 boundary conditions on the regional climate and to allow for a comparison with
634 available climate reconstructions from proxy-data. The modeling design was such that
635 a direct comparison of all considered paleo-climatic episodes is ensured.

636 Our first objective pertained to the former hypothesis by Prell and Kutzbach (1992),
637 suggesting that glacial boundary conditions can produce similar climate changes as
638 the uplift process during the Miocene-Pliocene period. This hypothesis could be
639 confirmed on the basis of regionally confined climate types, especially to the south of
640 the TP. However, at the scale of the entire region of the TP and adjacent areas, climate
641 is clearly more sensitive to the uplift than to Quaternary forcings during the LGM and
642 MH periods.

643 The second main objective referred to the generation of high-resolution patterns of
644 paleo-climatic states that are suitable for a quantitative comparison with climate
645 reconstructions. For the LGM period, this comparison is successful but based on a
646 very small number of grid boxes. The proxy-data for the MH episode are more
647 numerous and wide-spread over Asia. However, they exhibit substantial uncertainty,
648 drawing a picture of substantial temperature rise and decrease at neighboring locations.



649 In contrast, our model produces patterns of paleo-climatic anomalies that are spatially
650 more coherent and physically consistent. Therefore, we suggest that RCMs are able
651 to provide a valuable theoretical basis for the assessment of paleo-climate and the
652 climatological interpretation of proxy-data.

653 The third objective aimed at improving the spatiotemporal representation of paleo-
654 climatic episodes in Asia. Indeed, the simulation periods of 12 to 40 years per time
655 slice allowed for a sophisticated statistical analysis of climate changes in order to
656 distinguish between real climate change signals and climate anomalies that do not
657 stand out from internal variability. The horizontal resolution of 0.5° used in the RCM
658 REMO has proven to add value to the representation of climate across the TP,
659 compared with the driving global climate model. This relates to the model bias in terms
660 of present-day climate, the regional details of simulated climate patterns – especially
661 for the cluster analysis – and some new insights into climate changes north of the TP,
662 i.e. the appearance of the arid climate types in western China when the TP reaches at
663 least 50% of its present-day elevation.

664 There is still room for improvement of the model itself, the modified boundary
665 conditions and the number of experiments. First of all, the matrix of paleo-climatic
666 simulations could be extended, accounting for spatially differentiated uplift scenarios
667 (cf. Tapponnier et al. 2001, Tang et al. 2013a, Zhang et al. 2014), modified paleo-
668 geography (Fluteau et al. 1999), and a decomposition of forcings during the LGM and
669 MH periods (Liu et al. 2008a,b, Polanski et al. 2012). This would lead to a more
670 comprehensive understanding of the processes leading to paleo-climatic changes but,
671 at the same time, require enormous computing resources beyond the scope of our
672 project. In addition, simulations over several hundreds or thousands of years would
673 help to assess the superposition of time scales of climate fluctuations and changes



674 arising from internal and external sources of variability, respectively (cf. Lorenz and
675 Lohmann 2004).

676 The RCM REMO has been shown to have a good performance in terms of the
677 prominent features of regional climate in Asia and on the TP. Some systematic biases
678 can still be removed by statistical post-processing (e.g. Paeth 2011). However, a major
679 deficit is the simulation of precipitation over high-elevation lakes on the TP where
680 surface water temperature in winter is substantially overestimated and precipitation
681 amount by far too high. In the future, this problem will be tackled by implementing the
682 lake model FLake (Mironov et al. 2010) in REMO. Some further improvements can be
683 expected when REMO is used in a version that is fully coupled to an ocean model
684 enhancing the physical consistency of the climate simulations (cf. Sein et al. 2015).
685 Recently, Gao et al. (2015) have implemented a more realistic land cover classification
686 in REMO and reported an enhanced skill of the model. Finally, the aspect of model
687 uncertainty must be addressed also for the assessment of paleo-climatic changes, e.g.
688 by running a multi-model ensemble of RCMs nested into different global climate
689 models from the PMIP3 initiative. Combining this with the matrix of more sensitivity
690 studies implies an enormous computational effort that could be achieved in the
691 framework of a potential future paleo-CORDEX initiative.

692

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694

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701

702



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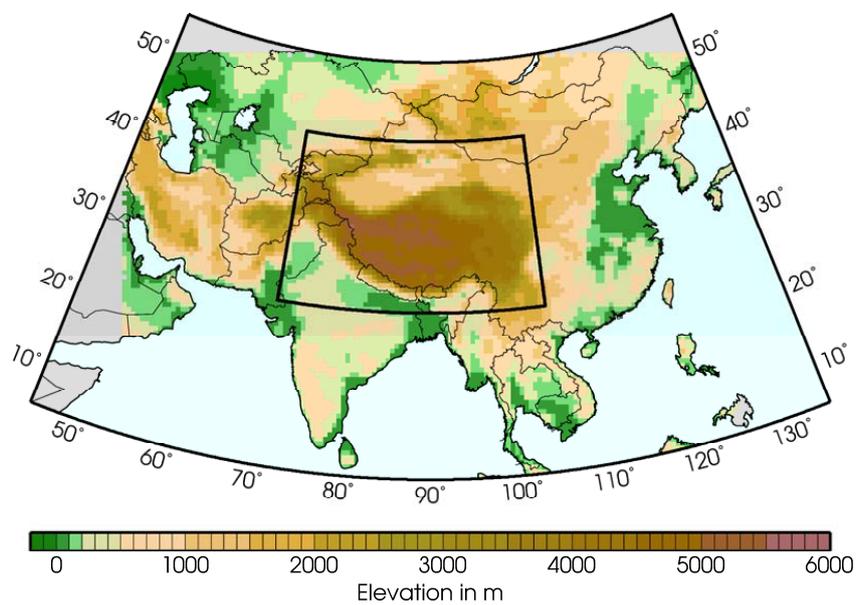
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904 **Figures**

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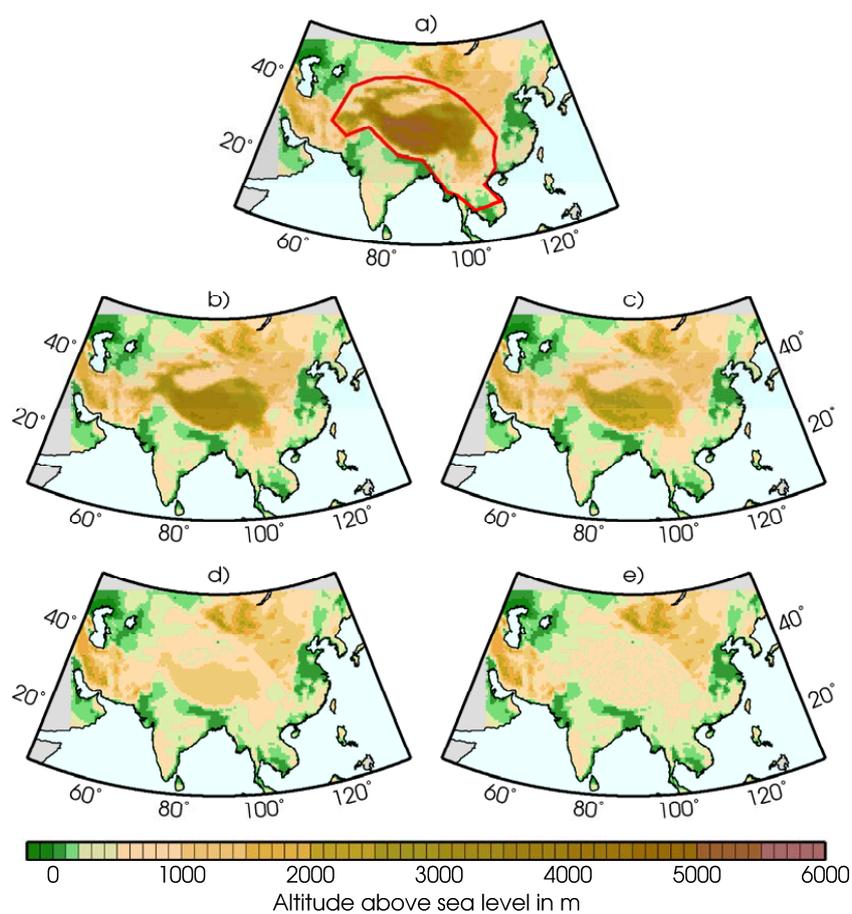
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909 Figure 1: Domain for the dynamical downscaling with REMO (outer region) and for the
910 cluster analysis centered over the Tibetan Plateau (inner region).

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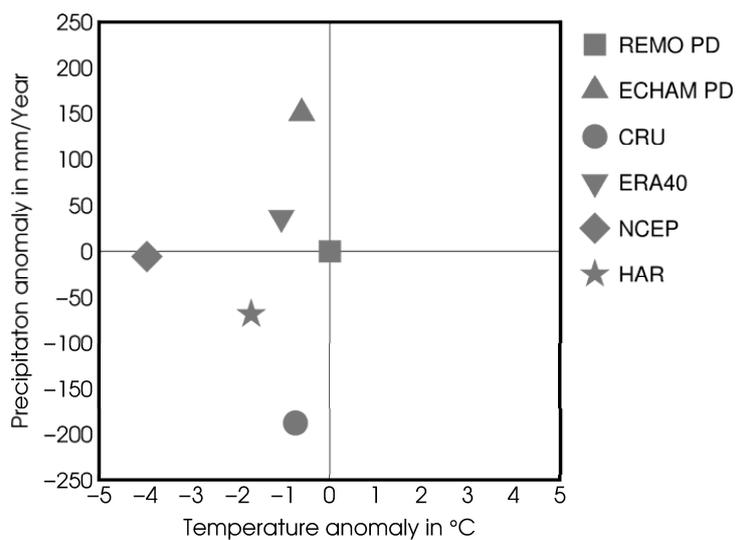
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915 Figure 2: Topography over Central Asia for the control run with present-day elevation

916 (a) and the stepwise reduction to 75% (b), 50% (c), 25% (d) and complete leveling (e)

917 compared with present-day elevation.

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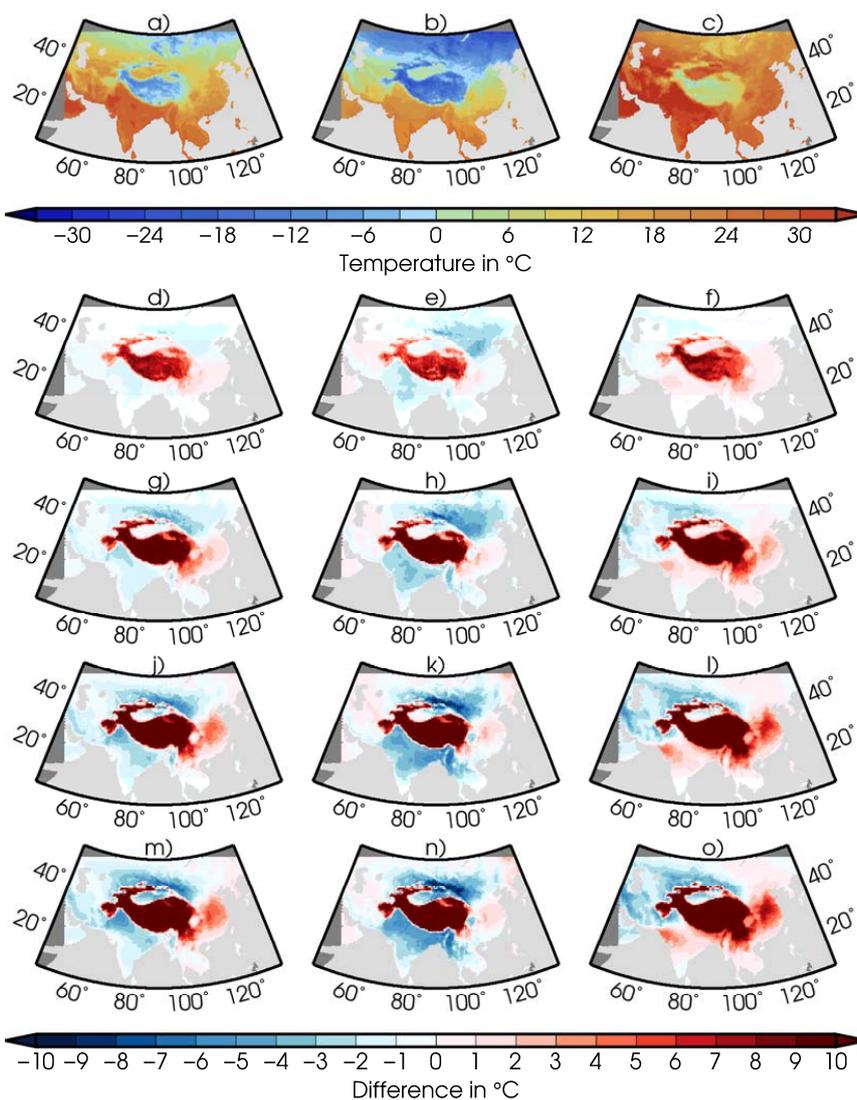


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921 Figure 3: Comparison of mean temperature and precipitation averaged over the cluster
922 domain (cf. Fig. 1) and the time period 1971-2000 from the regional climate model
923 REMO (PD = present-day), the driving global climate model ECHAM, the regional re-
924 analysis HAR, the global reanalysis NCEP, and the CRU observational data set.

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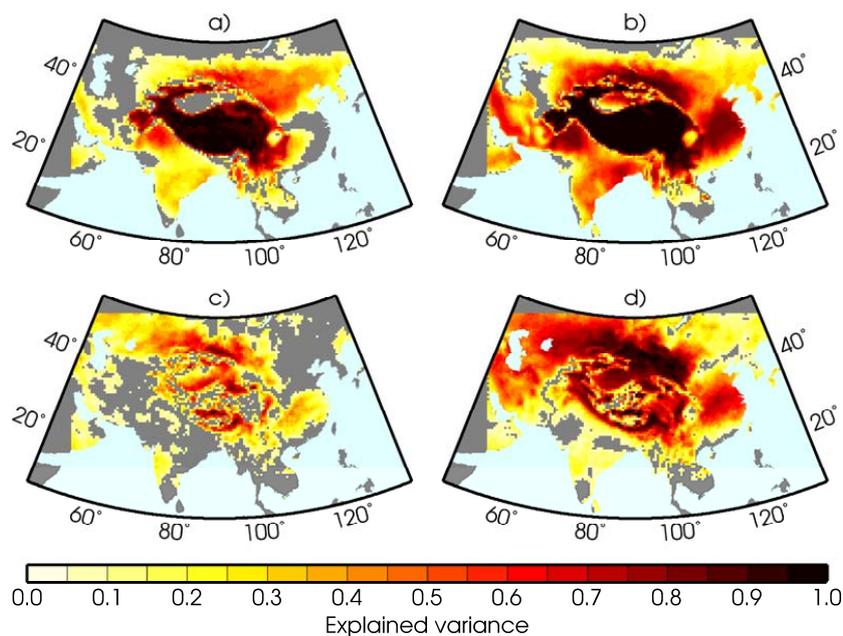


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928 Figure 4: Near-surface temperature averaged over the control period 1971-2000 (top
929 row) for annual means (left column), winter means (middle column) and summer
930 means (right column) as well as temperature changes due to a stepwise reduction of
931 elevation (d-o) from T075 (upper panels) to T000 (lower panels).

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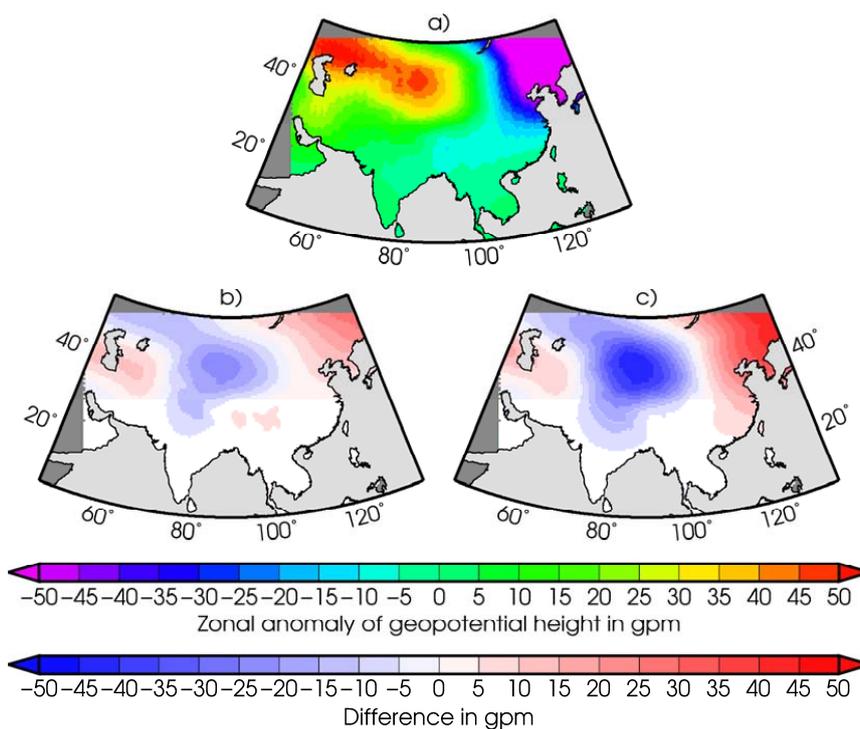


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942 Figure 6: Portion of variance explained by the uplift of the Tibetan Plateau as inferred
943 from a one-way analysis of variance for near-surface temperature (top panels) and
944 total precipitation (bottom panels) assessed between the T100 and T075 experiments
945 (left column) and between all uplift experiments T100 – T000 (right column). Only
946 values significant at the 5% level are plotted.

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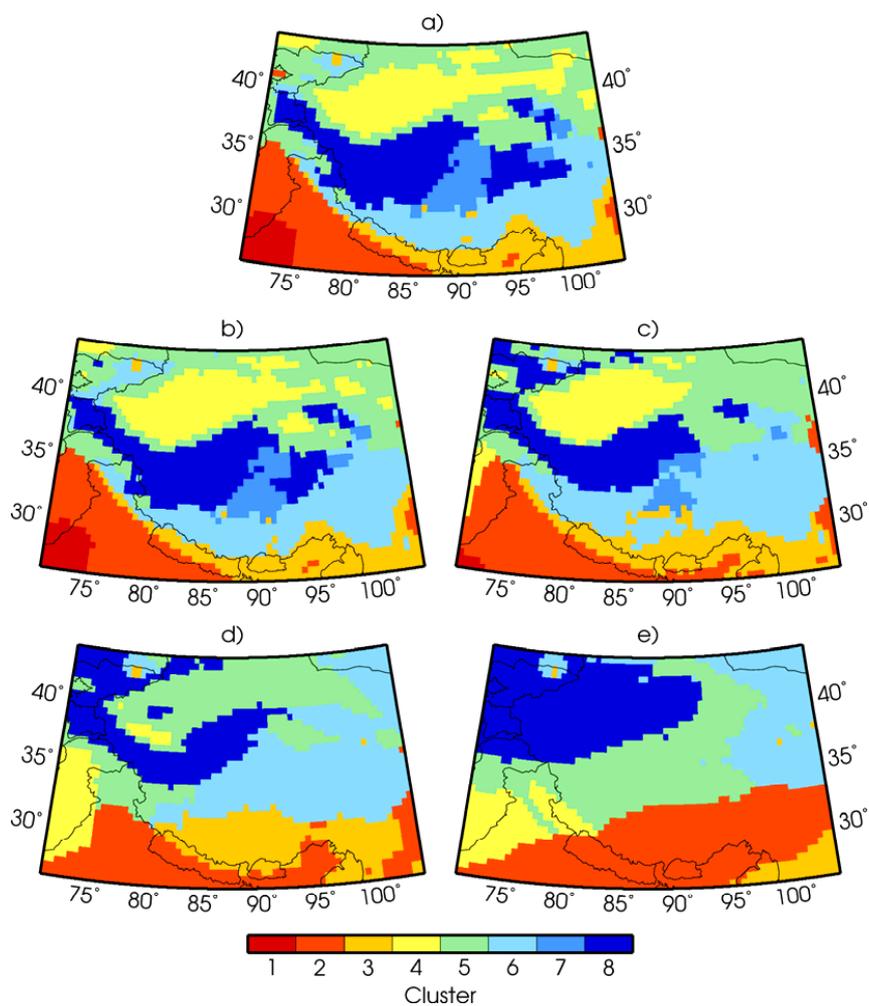


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950 Figure 7: Anomalies of 500 hPa geopotential height from the zonal mean averaged
951 over the control period 1971-2000 (a) and changes due to a stepwise reduction of
952 elevation in the T050 (b) and T000 (c) experiments.

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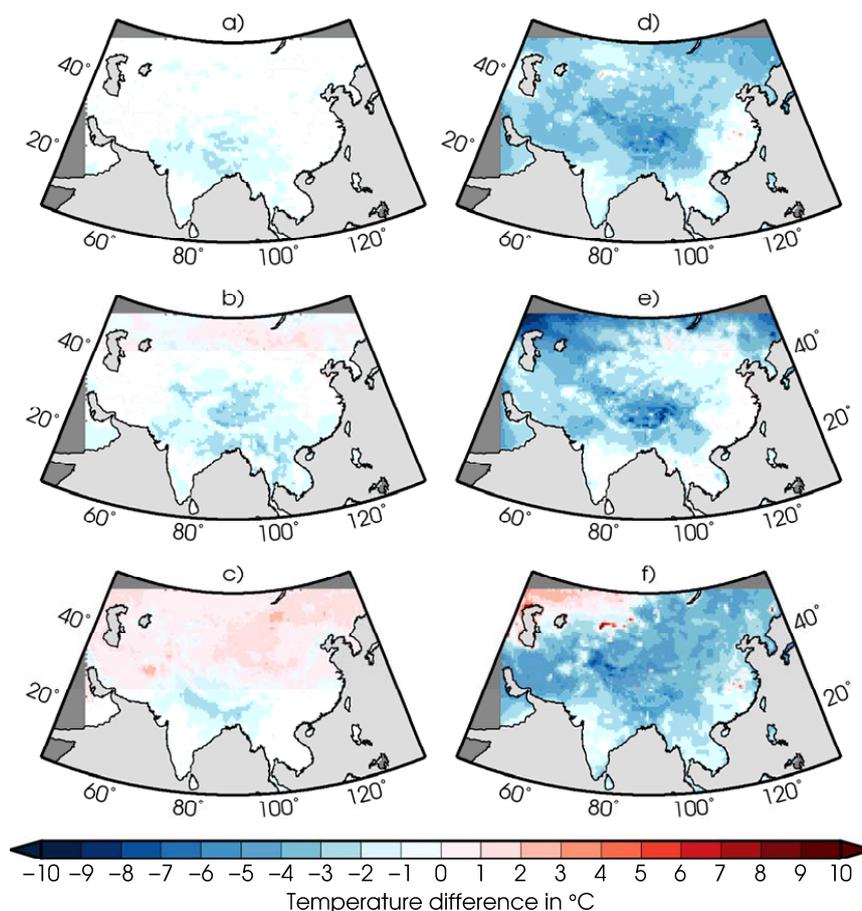


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956 Figure 8: Results of the cluster analysis prescribing 8 clusters in the Tibetan Plateau
957 region for the control period 1971-2000 (a) and projection of the uplift experiments
958 T075 (b) to T000 (e) onto the same centroids from the control period.

959



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962 Figure 9: Changes in near-surface temperature for annual means (top row), winter

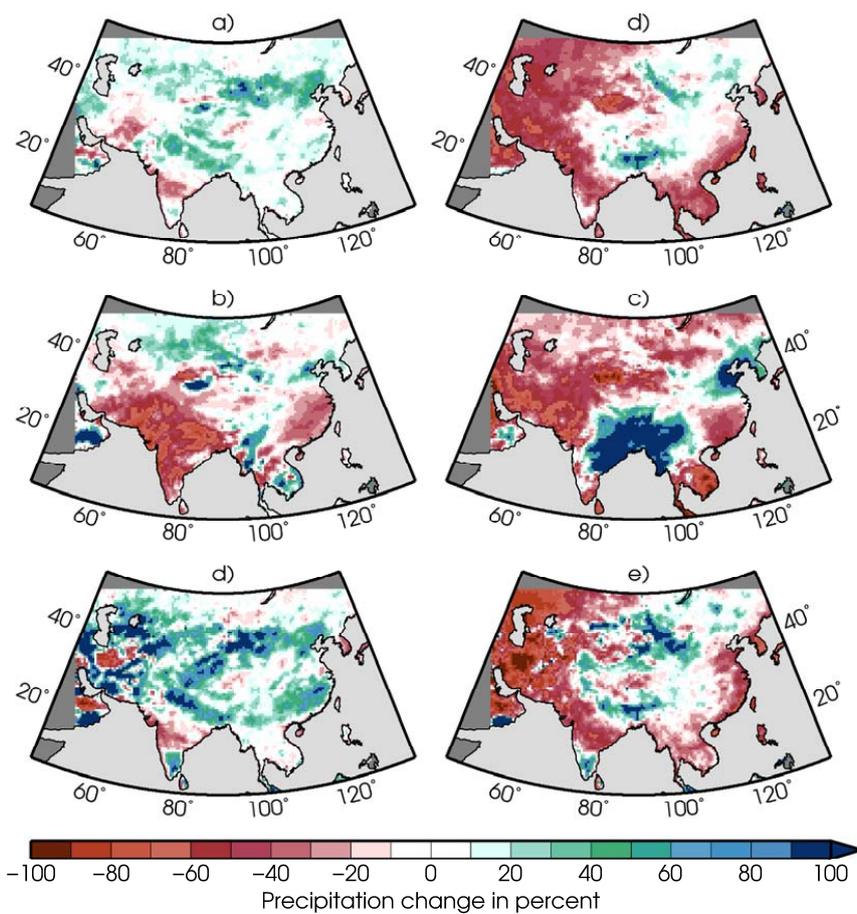
963 means (middle row) and summer means (bottom row) as simulated by the MH (left

964 column) and LGM (right column) experiment compared with the 1980-1989 control pe-

965 riod.

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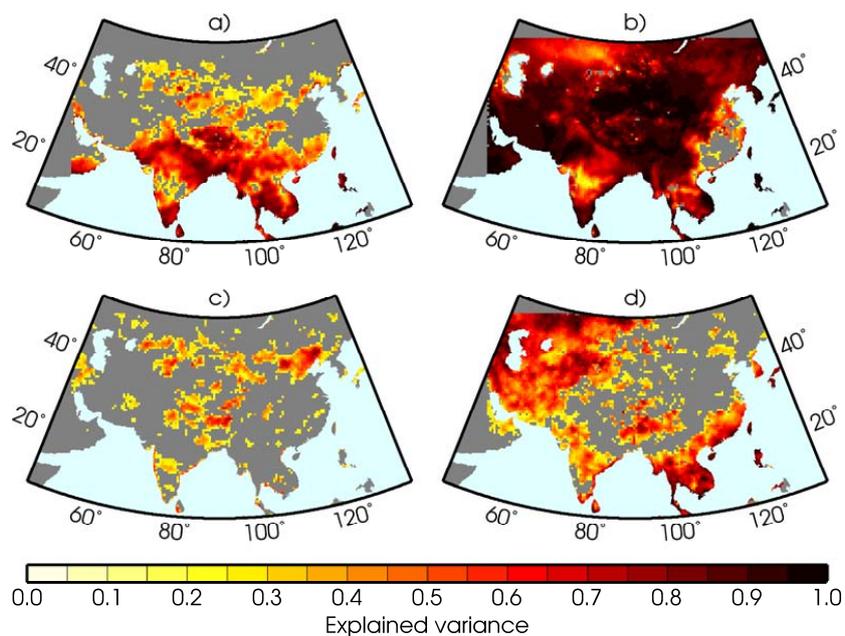


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970 Figure 10: Same as Fig. 9 but for total precipitation.

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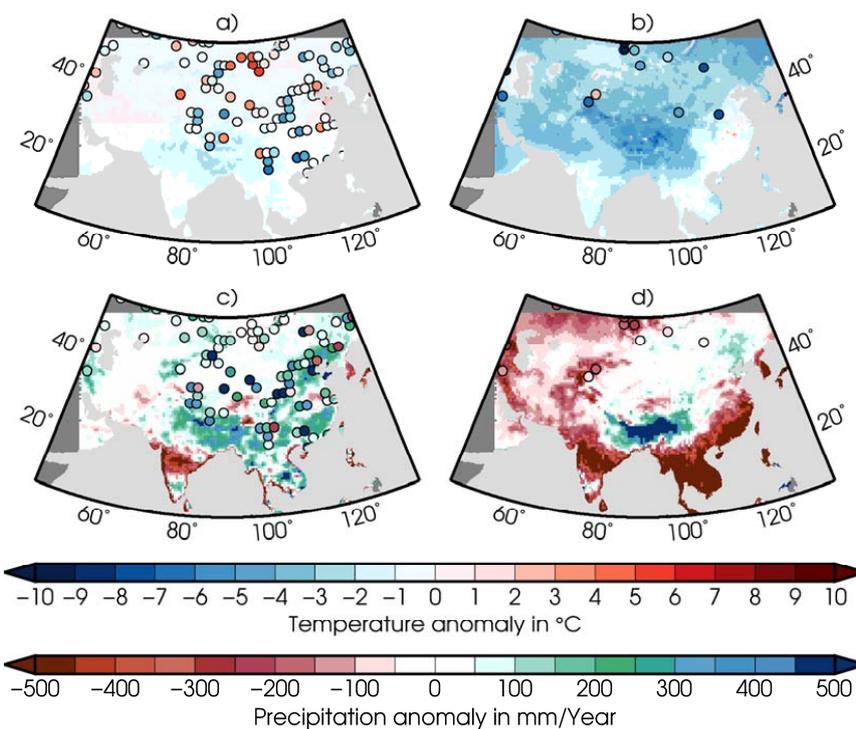


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974 Figure 11: Portion of variance explained by the MH (left column) and LGM (right col-
975 umn) experiment as inferred from a one-way analysis of variance for near-surface tem-
976 perature (top panels) and total precipitation (bottom panels). Only values significant at
977 the 5% level are plotted.

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981 Figure 12: Comparison of simulated mean annual temperature (top row) and total pre-

982 cipitation (bottom row) changes in the MH (left column) and LGM (right column) period

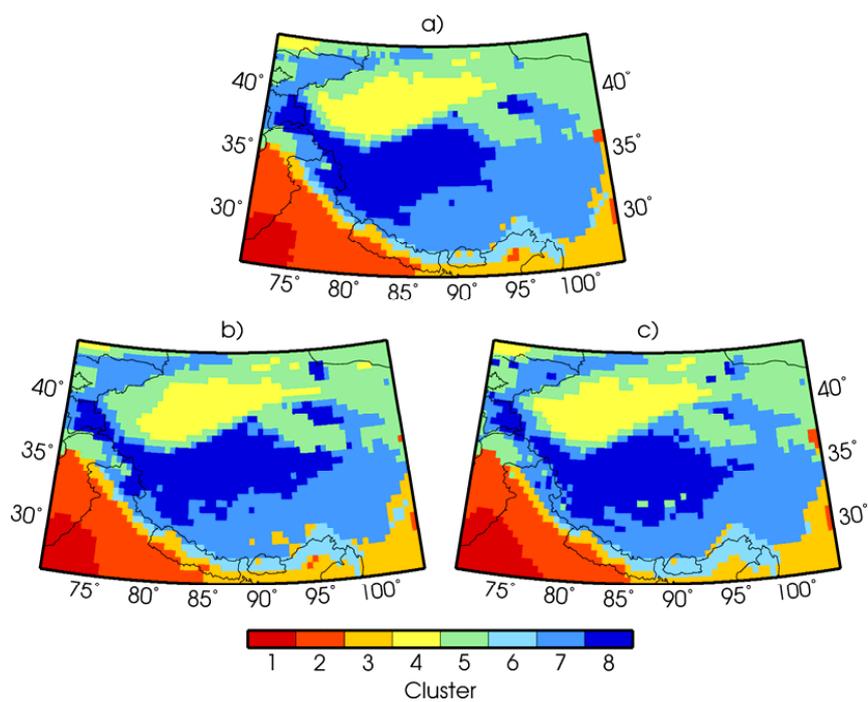
983 with available paleo-climatic reconstructions (filled circles) from proxy data based on

984 pollen analysis (Bartlein et al. 2011).

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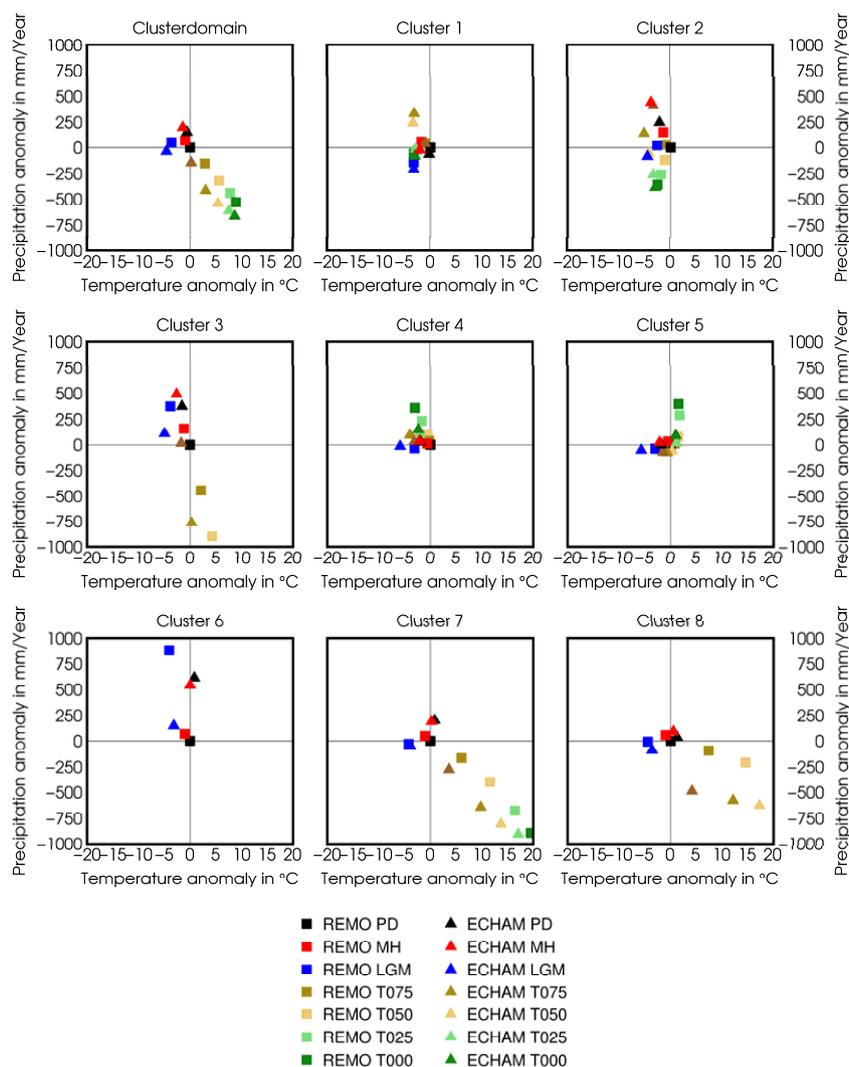


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989 Figure 13: Results of the cluster analysis prescribing 8 clusters in the Tibetan Plateau
990 region for the control period 1980-1989 (a) and projection of the MH (b) and LGM (c)
991 experiments onto the same centroids from the control period.

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993

994 Figure 14: Synthesis of changes in annual mean near-surface temperature (x-axis) and
 995 total precipitation (y-axis) averaged over the entire cluster domain (top left) and over
 996 the individual 8 cluster regions (remaining panels) from the uplift, MH and LGM exper-
 997 iments compared with the respective control periods, once for the regional climate
 998 model REMO and once for the driving global climate model ECHAM.