



1 **Changing seasonality of moderate and extreme precipitation events in the Alps**

2 Stefan Brönnimann^{1,2}, Jan Rajczak³, Erich Fischer³, Christoph C. Raible^{1,4}, Marco Rohrer^{1,2}, Christoph
3 Schär³

4

5 ¹ Oeschger Centre for Climate Change Research, University of Bern, Switzerland

6 ² Institute of Geography, University of Bern, Switzerland

7 ³ Institute for Atmospheric and Climate Science, ETH Zurich, Switzerland

8 ⁴ Climate and Environmental Physics, Physics Institute, University of Bern, Switzerland

9

10 **Abstract.** The intensity of precipitation events is expected to increase in the future. The rate of
11 increase depends on the strength or rarity of the events; very strong and rare events tend to follow the
12 Clausius-Clapeyron relation, whereas weaker events or precipitation averages do not. An often
13 overlooked aspect is seasonal occurrence of such events, which might change in the future. To address
14 the impact of seasonality, we use a large ensemble of regional and global climate model simulations,
15 comprising tens of thousands of model years of daily temperature and precipitation for the past,
16 present and future. In order to make the data comparable, they are quantile-mapped to observation-
17 based time series representative of the Aare catchment in Switzerland. Model simulations show no
18 increase in annual maximum 1-day precipitation events (Rx1day) over the last 400 yrs and an increase
19 of 10-20% until the end of the century for a strong (RCP8.5) forcing scenario. This fits with a
20 Clausius-Clapeyron scaling of temperature at the event day, which increases less than annual mean
21 temperature. An important reason for this is a shift in seasonality. Rx1day events become less frequent
22 in late summer and more frequent in early summer and early fall, when it is cooler. The seasonality
23 shift is shown to be related to summer drying. Models with decreasing annual mean or summer mean
24 precipitation show this behavior more strongly. The highest Rx1day per decade, in contrast, shows no
25 change in seasonality in the future. This discrepancy implies that decadal-scale extremes are
26 thermodynamically limited; conditions conducive to strong events still occur during hottest time of the
27 year on a decadal scale. In contrast, Rx1day events are also limited by other factors. Conducive
28 conditions are not reached every summer in the present, and even less so in the future. Results suggest
29 that changes in the seasonal cycle need to be accounted for when preparing for moderately extreme
30 precipitation events and assessing their socio-economic impacts.



31 1. Introduction

32 Heavy precipitation extremes in the Alps may trigger flood events or landslides that lead to loss of
33 lives, cause large monetary losses and threaten important infrastructure. While the most extreme
34 events are the most costly, even moderate extreme events such as the annual maximum rainfall
35 (Rx1day) or events with a 10-yr return period may be relevant for climate change adaptation. Even for
36 these events, the observation records at a single location may not be long enough to capture trends.
37 Still, aggregating weather station records from Switzerland, Scherrer et al. (2016) found an increase in
38 intensity as measured by the magnitude of the annual maximum 1-day precipitation extreme (Rx1day)
39 or exceedances of the 99th percentile.

40 Heavy precipitation requires moisture convergence and convection or synoptic-scale uplift. With
41 increasing temperatures, saturation specific humidity increases and as a consequence an increased
42 precipitation intensity is expected. At the global scale, precipitation is limited by radiation that has to
43 balance the latent heat release (e.g. Allen and Ingram 2002). For extreme precipitation events,
44 however, Scherrer et al. (2016) found that the increase closely follows a Clausius-Clapeyron scaling of
45 the annual mean temperature trend. Fischer and Knutti (2016) also found a close to Clausius-
46 Clapeyron scaling to regional temperature changes both in global models and observations. However,
47 the scaling may not hold exactly for various reasons.

48 Firstly, global models do not resolve changes in convection (Prein et al., 2015; Zhang et al., 2017),
49 which is important for the case of Alpine precipitation (Giorgi et al., 2016). For instance, trends in
50 Alpine precipitation are different during summer in convection-resolving regional models (Ban et al.,
51 2014). Secondly, the Clausius-Clapeyron scaling of extreme precipitation with temperatures at day-to-
52 day time scales cannot be necessarily extrapolated to the future based on annual or seasonal mean
53 temperatures (e.g., Zhang et al. 2017). One possible cause for a discrepancy between scaling at day-to-
54 day time scales and at time scales of long-term warming is a potential change in seasonality. Pfahl et
55 al. (2017) in an analysis of CMIP5 data (Taylor et al., 2012), found a shift in the future Rx1day
56 towards smaller saturation specific humidities over most of the northern extratropical land areas (their
57 Fig. S6), indicative of lower temperatures and thus a shift towards the colder seasons. Recently,
58 Messmer et al. (submitted) focused on one dynamical feature, which is responsible for extreme
59 precipitation events in the Alps, the so-called Vb cyclone (van Bebber, 1898; Messmer et al., 2015).
60 They found in dynamically downscaled scenario simulations (RCP 8.5) that the occurrence of extreme
61 Vb cyclones is shifted from the midsummer to May and September in the future.

62 In this paper we focus on seasonality changes of heavy precipitation events (Rx1day) in the Alps using
63 a comprehensive set of global and regional model simulations comprising tens of thousands of years
64 of daily temperature and precipitation quantile mapped to observations representative of a large
65 catchment in Switzerland. Based on this comprehensive data set, we analyse (1) trends in Rx1day and



66 their relation to temperature trends, (2) changes in the seasonal cycle of temperature on Rx1day events
67 and (3) changes in the seasonal frequency of occurrence of Rx1day.

68 The paper is organised as follows. Section 2 gives a brief overview of the data and the quantile
69 mapping approach (Rajczak et al., 2018) and outlines the analyses performed. Section 3 presents the
70 results for Rx1day. A discussion in terms of underlying mechanisms and differences between models
71 and model set-ups follows in Section 4. Conclusions are drawn in Section 5.

72

73 2. Data and Methods

74 The study focuses on the Aare river catchment in Switzerland, an area of approximately 17 000 km²
75 (see Fig. 1). We extract daily temperature and precipitation data over this domain from a large data set,
76 comprising simulations of the past, present and the future from different set-ups (coupled and
77 uncoupled simulations, global and regional simulations, single member or ensemble simulations),
78 reanalyses, dynamically downscaled reanalyses, and observations. A total of 55,000 simulation years
79 are available (Rajczak et al., 2018, the data are available from this website:
80 <https://doi.pangaea.de/10.1594/PANGAEA.886881>). In this study we focus on experiments with
81 regional or global models. The following ensembles are used:

- 82 • CCC400: An ensemble of 30 AMIP-type global GCM simulations using ECHAM5.4 at T63
83 and covering the period 1600–2005
- 84 • the CMIP5 ensemble of 25 simulations (historical and RCP8.5) covering the period 1900–
85 2100
- 86 • initial condition ensemble of 21 COSMO simulations (0.44°) using the RCP8.5 scenario and
87 covering the years 1950–2100
- 88 • a set of 13 ENSEMBLES simulations using the A1B scenario and covering the period 1971–
89 2100
- 90 • a set of 15 CORDEX-11 simulations at 0.11° resolution (RCP8.5 and RCP4.5) covering the
91 period 1970–2099
- 92 • a set of 17 (RCP4.5) and 19 (RCP8.5) CORDEX-44 simulations at 0.44° resolution covering
93 the period 1970–2099

94 Table 1 gives an overview of all experiments used in our study, including references. The fact that the
95 data sets differ in several key aspects (e.g., resolution, time period covered, scenario used, complexity
96 of the model) allows a comprehensive view of possible changes and sources of uncertainties.

97 A consistent analysis of the available data sources is hardly feasible, as the data stems from models
98 with vastly different resolution and characteristics. All models suffer to some extent from biases, and



99 the variety of modeling approaches thereby implies significant model-to-model differences. For this
100 reason, we focus on a fairly large catchment (where all models should arguably have at least some
101 potential), and use a statistical approach (quantile mapping) to calibrate the simulation results against
102 observations.

103 More specifically, the data are quantile-mapped (Themessl et al., 2011; Gudmundsson et al., 2012;
104 Teutschbein and Seibert, 2012; Rätty et al., 2014) to observation-based time series representative of the
105 average over the Aare catchment in Switzerland (see Fig. 1 for the stations; Fig. S1 shows the
106 observation-based, annual time series). Both data sets are used with daily resolution. The data and
107 downscaling are described in detail in Rajczak et al. (2018). The method has recently been used in
108 other Swiss climate impact studies (Rajczak et al., 2016a), and is reasonably skillful in daily and
109 multi-day precipitation diagnostics (Rajczak et al., 2016b). The transfer-function depends on the
110 season and is based on a 91-day moving window, centered over the day of the year (Themessl et al.,
111 2012; Wilcke et al., 2013; Rajczak et al., 2016a, 2016b). Values above the observed range of values
112 are corrected according to the 99.9th percentile (p99.9) in a constant manner (p1 and p99 for
113 temperature). Studies recommend such an implementation opposed to complex extrapolation methods
114 (Gutjahr and Heinemann, 2013; Ivanov and Kotlarski, 2017; Themessl et al., 2012). Comparisons of
115 raw and quantile-mapped data are shown in Fig. S2. Note that quantile mapping does affect mean
116 temperature as well as temperature at the event day, which implies changes in saturation specific
117 humidity. It is therefore important to perform all analyses also for the raw data.

118 The analysis focuses on Rx1day, the maximum Rx1day per decade, annual mean temperature,
119 temperature at the event day, and the day of year of the event. All analyses are applied to each
120 individual simulation; only then the ensemble statistics are formed. No further weighting is performed
121 for multi-model data sets, but for some analyses the simulations are separated into those exhibiting a
122 drying or a wetting. We have performed all analyses for both the raw and quantile-mapped data. We
123 refer to the raw data occasionally in the text, more comprehensive material is added to the
124 supplementary materials. For our analyses the following approach is used:

125 (i) We analyse long-term changes in Rx1day in each data set and compare this to the long-term
126 changes in annual mean temperature and as well as the trend in temperature at the event day. This
127 gives an indication of the proximity of trends in precipitation changes to a Clausius-Clapeyron scaling.

128 (ii) Changes in the seasonal cycle of temperature and event frequency (see below) are then addressed
129 by partitioning the data sets into sub-periods. For simulations of past climate (CCC400, 1600–2000),
130 we partition the data into centuries. For the present and future, for reasons of consistency, we show
131 results only for the two 35-yr periods 1971–2005 and 2065–2099, which are common to all model
132 experiments. In the electronic supplement we also show results for all future simulations partitioned
133 into three periods of equal length (four periods for the longer simulations CMIP5 and CCC400) to
134 make full use of all data.



135 (iii) Within the periods, we analyse the seasonal cycle of the relative frequency of Rx1day events as
136 well as the seasonal cycle of temperature at the event day. The former indicates whether the
137 seasonality of Rx1day events changes, which is likely to affect the temperature trend on event days.
138 The latter takes this seasonality shift into account and indicates whether for a given calendar day,
139 temperature trends on event days differ from trends on all days. The annual cycle of temperature for
140 all days and for event days is estimated by fitting the first two harmonics of the seasonal cycle.

141 (iv) Finally, for multi-model ensembles, we stratify the simulations within an ensemble according to
142 their linear trend (obtained with least-squares regression) in annual or summer mean precipitation over
143 the period 1971–2099. This allows addressing common signatures, *e.g.*, whether drying models tend to
144 show stronger changes in seasonality than wetting models.

145

146 3. Results

147 Time series of Rx1day for the ensembles CCC400 (1600–2005), CMIP5 (historical and RCP8.5,
148 1900–2100), the COSMO initial condition ensemble (RCP8.5, 1950–2100), the ENSEMBLES
149 simulations (A1B, 1971–2100), CORDEX-11 (RCP8.5, 1970–2099) and CORDEX-44 (RCP8.5,
150 1970–2099) are shown in Fig. 2 (top). Over the past 400 years, no change in Rx1day is found (the
151 same is true for past millennium simulations (Lehner et al., 2015) or the Twentieth Century Reanalysis
152 20CRv2c (Compo et al., 2011) included in Rajczak et al. (2018); not shown). A slight increase in
153 Rx1day over the past 50 years appears in CMIP5 (historical). Note, that there are also indications at
154 the continental to global scale that the majority of RCMs and GCMs tend to underestimate the
155 observed intensification in heavy rainfall (Fischer and Knutti, 2016; Borodina et al., 2017).
156 Simulations for the future show a clear increase in Rx1day of around 10-20%. However, in the
157 ensemble mean this increase remains relatively modest. In the CMIP5 ensemble, the upper range of
158 the ensemble shows a stronger increase than the mean.

159 Annual mean temperature (Fig.2, bottom, red line) shows a pronounced increase of 4-6 °C in the 21st
160 century. According to the Clausius-Clapeyron relation, this would correspond to a 25-40% increase in
161 precipitation extremes, which is not the case. However, if we consider only the temperature on the day
162 of the Rx1day precipitation events (Fig. 2, bottom, full blue line, note that this time series is
163 smoothed), we find a smaller temperature increase for most of the simulations. For the 21st century, the
164 temperature increase during Rx1day events amounts to only 3 °C (CMIP5) or even just 1 °C
165 (COMSO), but with considerable differences between experiments. Trends in Rx1day of 10-20% thus
166 approximately follow the Clausius-Clapeyron scaling. The same analyses for raw data gives very
167 similar results (shown in Fig. S2).

168 As already stated above, the Clausius-Clapeyron argument is expected to apply for the largest events,
169 but not for intermediate events, and may be not even for Rx1day events. Ban et al. (2015) analyzed



170 convection-resolving climate simulations at a resolution of 2 km, and calculated the scaling rates as a
171 function of all-day percentiles in the Alpine region for summer (see their Fig. 4e). In terms of
172 percentiles, our Rx1day event class would correspond to roughly the 99.7th percentile, and at this event
173 category the scaling estimate of Ban et al. (2015) amounts to about 1-3%/°C. This is roughly
174 consistent with the results in Fig.2 (when using mean temperature changes). Nevertheless, Ban et al.
175 (2015) found that increases in precipitation intensity would approximately asymptote towards the
176 Clausius-Clapeyron rate at very high percentiles.

177 The lower part of Fig. 2 also shows the same analysis for the temperature taken at the highest Rx1day
178 per decade (dashed lines). Note that sample size here is very small except for CCC400 and CMIP.
179 Interestingly, in CMIP5 temperatures during these events increase more strongly than for all Rx1day
180 events. The trend resembles that of the annual mean.

181 Two factors can contribute to the fact that the temperature increase on Rx1day events is smaller than
182 the increase in the annual mean: (1) a change in the occurrence frequency of Rx1day events towards a
183 colder season or (2) a different change of the temperature on Rx1day events even for unchanged
184 seasonality. We first analyse the latter. Considering the seasonal cycle of temperature during Rx1day
185 events (Fig. 3, top row), we find that in summer, Rx1day events in the present occur on days that are
186 slightly colder than average, while those few Rx1day events that occur during the cold season occur
187 with warmer than average conditions. In the future, Rx1day events tend to occur on days that are even
188 cooler than average, *i.e.*, the trend on Rx1day events (the difference between 2065–2099 and 1971–
189 2005 is shown on the right) is ca. 1 °C smaller than the trend on all days. This holds for all model
190 ensembles, and the difference has no obvious seasonal cycle. Thus, part of the trend difference
191 between Rx1day and all days is unrelated to the seasonality of Rx1day.

192 The day-of-occurrence of Rx1day (Fig. 3, second row) shows two peaks, in early summer (June) and
193 in early fall (September). Both in the past (CCC400) and in the future, there is a trend towards fewer
194 occurrences during June and July and more during the neighbouring months, *i.e.*, both the early
195 summer and the early fall peaks shift towards the cold season. This becomes particularly clear when
196 plotting the difference in relative frequency as a function of day of year between 2065–2099 and
197 1971–2005 (Fig. 3, middle right) for each experiments. The mid-summer events become rare; their
198 occurrence already decreased by 10% during the past 400 years and might further decrease by 30% in
199 the future. This shift in seasonality explains the remaining difference in temperature trends between
200 Rx1day event days and all days.

201 As we use the periods 1971–2005 and 2065–2099, Figure 3 (second row) is based on 100 (left) or 35
202 (middle) values per ensemble member and results were then smoothed with a Gaussian kernel with a
203 bandwidth of 15 days. In order to exploit the full data set, we also partitioned each data set into equal-
204 sized bins (Fig. S3) and found very similar results. Note, also, that the change in seasonality is not due
205 to the quantile mapping but appears also in the raw data (see Fig. S2).



206 However, the above results depend on the rarity of events. When analyzing more rare events, such as
207 the highest Rx1day events per decade (Fig. 3, bottom; note that samples are large enough only for
208 CCC400 and CMIP5, results for other experiments are shown in Fig. S3), we find a different result.
209 These events are even more concentrated to mid-summer than Rx1day in the present climate. In
210 CCC400, their occurrence is high throughout the summer whereas in CMIP5 these are mostly early
211 fall events. Interestingly, there is no change in the frequency of occurrence over time. Hence, only
212 moderately extreme events (with a frequency of once per year) are affected by a change in seasonality
213 and not the more extreme events.

214 What causes this change in seasonality? Given the limited number of variables at our disposition
215 (temperature, precipitation), it is not possible to properly disentangle dynamical and thermodynamical
216 contributions. However, analyzing differences between the ensemble members gives important
217 indications. Within ENSEMBLES, CORDEX-11, CORDEX-44, and CMIP5 the contributing
218 ensemble members are grouped according to their trend in annual or summer mean precipitation (Fig.
219 4) and the analysis of the frequency of occurrence is repeated. Clear differences are found between
220 ensemble simulations, which show a drying and ensemble members with a positive trend in
221 precipitation in summer and in the annual mean (Fig. 4), particularly in CMIP5 and CORDEX-44, less
222 so in ENSEMBLES and CORDEX-11. The wetting ensemble members show hardly any change in
223 seasonality or even an increased frequency in summer, while the drying members show a decrease in
224 the frequency of occurrence of Rx1day events in mid-summer. The inter-model variability is large as
225 the samples get smaller, but the signature is robust across the entire data set (see Fig. S4).

226 Drying might also explain the changing seasonality in CCC400, as the ensemble mean shows
227 decreasing precipitation over the past 400 years (Fig. S6). Note that a misspecified trend in land-
228 surface parameters might affect CCC400 (Rohrer et al., 2018). However, a simulation with corrected
229 land surface shows the same change in seasonality (Fig. S3).

230

231 4. Discussion

232 Heavy precipitation events (Rx1day) in the Aare catchment in the present climate are most frequent
233 during the warm season. If all preconditions (atmospheric circulation, stability, soil moisture, etc.) are
234 conducive to a heavy precipitation event, the intensity of the event is then expected to depend on
235 mainly on thermodynamics, *i.e.*, on saturation specific humidity. It will thus follow a Clausius-
236 Clapeyron scaling as a first order approximation, often summarized as thermodynamic changes. In
237 fact, trends in Rx1day over the Aare catchment do follow a Clausius-Clapeyron scaling, however not
238 of annual mean temperature but rather of event-day temperature. This is because Rx1day may also
239 change due to dynamic changes *e.g.*, conducive conditions become less frequent and are not reached
240 each summer. Such changes may relate to changes in the large-scale circulation patterns or enhanced
241 atmospheric stability (*e.g.* Kröner et al., 2017) leading to reduced vertical updrafts (Pfahl et al., 2017).



242 In this respect, Rx1day and decadal-scale events exhibit a different behaviour in that the former show
243 a change in seasonality while the later do not. This discrepancy may be interpreted in the following
244 way: Decadal-scale precipitation extremes are mainly thermodynamically limited, *i.e.*, chances are
245 high that within a decade, conducive conditions occur on a late summer day, when temperatures and
246 thus saturation specific humidity are highest. In the future, such conditions can still be reached in late
247 summer once per decade, and thus the increase in temperature leads to an intensification of the most
248 extreme events, with no apparent change in seasonality, consistent with Ban et al. (2015). However, on
249 an annual scale such conditions are not reached every summer. Even in the present climate, Rx1day
250 events do not occur most frequently at the warmest time of the year. This indicates that other factors
251 (which we termed preconditions above) matter and Rx1day events are not fully thermodynamically
252 limited. This tendency will strengthen in the future as preconditions change, leading to a change in
253 seasonality. This result is consistent though more pronounced than identified by Pfahl et al. (2017)
254 who found a change of Rx1day events towards lower saturation specific humidities over the northern
255 extratropical land areas, indicating that the behaviour found in our analysis may be true for other
256 extratropical land regions as well.

257 While we cannot disentangle the contributions of individual factors to the preconditions, we can
258 identify drying as one of the important factors in models. Models that show a drying trend tend to
259 exhibit a shift in the seasonal cycle of moderate extremes (Rx1day) away from late-summer towards
260 early summer and fall, whereas the others do not.

261 A further result of our study is that even for a given calendar day, the future temperature trend in the
262 models is smaller for Rx1day than for all days. A possible explanation, particularly for summer days,
263 is model drying and thus excessive heating on non-precipitation days. In fact, stratifying the
264 simulations according to their summer-mean precipitation trends indicates that drying simulations
265 have a stronger cooling of event days relative to all days (Fig. S5).

266 The findings of our study - intensification of the most extreme events and change of seasonality for
267 moderate extremes - is relevant for adapting to future climatic changes. This is particularly the case as
268 observational coverage of the highest Rx1day per decade is limited. Whether or not the seasonality
269 shift is a real effect or occurs due to artificial summer drying of the models remains to be studied.
270 More detailed studies of the underlying processes, including stability and atmospheric circulation,
271 during future extreme events are also required (*e.g.*, Messmer et al. submitted).

272

273 5. Conclusions

274 Over the Alpine region moderate precipitation extremes such as characterized by Rx1day may not
275 increase as much as expected from applying a Clausius-Clapeyron to the change in annual mean
276 temperature, due to a change in seasonality of Rx1day events, and due to smaller temperature trends



277 during event days (regardless of the season). Both reasons are due to the fact that moderate extreme
278 events are not purely thermodynamically limited. In our study, we find that $Rx1day$ events in a future
279 climate tend to occur less frequently in mid-summer, but more often in spring and autumn. A similar
280 change is also found over the past 400 years in model simulations. Further analyses show that mostly
281 those models are concerned whose annual mean or summer mean precipitation decreases, i.e., the
282 changing seasonality is in part due to drying. Conversely, 10-yr events do exhibit their highest
283 frequency in mid-summer also in the future, with no apparent change in seasonality.

284 For flood protection this means that moderate events might shift towards the cold season with only a
285 small change in intensity, but the more relevant extreme events such as those with 10-yr return period
286 remain in summer and increase strongly in intensity.

287

288 **Acknowledgements.** This work was funded by the Swiss Federal Office for the Environment (FOEN),
289 the Swiss Federal Office of Energy (SFOE), and the Swiss Federal Nuclear Safety Inspectorate (ENSI)
290 in the framework the project EXAR: Understanding Extreme Flooding Events Aare-Rhein in
291 Switzerland as well as by the Swiss National Science Foundation (project 200021_143219 “EXTRA-
292 LARGE”). We acknowledge the World Climate Research Programme’s Working Group on Regional
293 Climate, and the Working Group on Coupled Modelling, former coordinating body of CORDEX and
294 responsible panel for CMIP5. We also thank the climate modelling groups (listed in Table S1 of this
295 paper) for producing and making available their model output. We also acknowledge the Earth System
296 Grid Federation infrastructure an international effort led by the U.S. Department of Energy’s Program
297 for Climate Model Diagnosis and Intercomparison, the European Network for Earth System Modelling
298 and other partners in the Global Organisation for Earth System Science Portals (GO-ESSP).

299

300 **References**

301 Allen, M. R., and Ingram, W. J.: Constraints on future changes in climate and the hydrologic cycle,
302 *Nature*, 419, 224–232, 2002.

303 Ban, N., Schmidli, J., and Schär, C.: Evaluation of the convection-resolving regional climate
304 modelling approach in decade-long simulations, *J. Geophys. Res. Atmos.*, 119, 7889–7907, doi:
305 10.1002/2014JD021478, 2014.

306 Ban, N., Schmidli, J., and Schär, C.: Heavy precipitation in a changing climate: Does short-term
307 summer precipitation increase faster? *Geophys. Res. Lett.*, 42, 1165–1172,
308 doi:10.1002/2014GL062588, 2015.

309 Begert, M.: Die Repräsentativität der Stationen im Swiss National Basic Climatological Network
310 (Swiss NBCN), *Arbeitsberichte der MeteoSchweiz* 217, 2008.



- 311 Bhend, J., Franke, J., Folini, D., Wild, M., and Brönnimann, S.: An ensemble-based approach to
312 climate reconstructions, *Clim. Past*, 8, 963–976, doi:10.5194/cp-8-963-2012, 2012.
- 313 Borodina, A., Fischer, E. M., and Knutti, R.: Models are likely to underestimate increase in heavy
314 rainfall in the extratropical regions with high rainfall intensity, *Geophys. Res. Lett.*, 44, 7401–7409,
315 2017.
- 316 Compo, G. P., Whitaker, J. S., Sardeshmukh, P. D., Matsui, N., Allan, R. J., Yin, X., Gleason, B. E.,
317 Vose, R. S., Rutledge, G., Bessemoulin, P., Brönnimann, S., Brunet, M., Crouthamel, R. I., 1035
318 Grant, A. N., Groisman, P. Y., Jones, P. D., Kruk, M. C., Kruger, A. C., Marshall, G. J., Maugeri,
319 M., Mok, H. Y., Nordli, Ø., Ross, T. F., Trigo, R. M., Wang, X. L., Woodruff, S. D., and Worley,
320 S. J.: The Twentieth Century Reanalysis Project, *Q. J. Roy. Meteor. Soc.*, 137, 1–28, 2011.
- 321 Fischer, E. M., and Knutti, R.: Anthropogenic contribution to global occurrence of heavy-precipitation
322 and high-temperature extremes, *Nature Climate Change*, 5, 560–564, 2015.
- 323 Fischer, E. M., and Knutti, R.: Observed heavy precipitation increase confirms theory and early
324 models, *Nature Climate Change*, 6, 986–991, 2016.
- 325 Giorgi, F., Torma, C., Coppola, E., Ban, N., Schär, C., and Somot, S.: Enhanced summer convective
326 rainfall at Alpine high elevations in response to climate warming, *Nature Geoscience* 9, 584–589,
327 2016.
- 328 Gudmundsson, L., Bremnes, J. B., Haugen, J. E., and Engen-Skaugen, T.: Technical Note:
329 Downscaling RCM precipitation to the station scale using statistical transformations - a comparison
330 of methods, *Hydrol. Earth Syst. Sci.*, 16, 3383–3390, 2012.
- 331 Gutjahr O., and Heinemann, G.: Comparing precipitation bias correction methods for high-resolution
332 regional climate simulations using COSMO-CLM, *Theor. Appl. Climatol.*, 114, 511–529, doi:
333 10.1007/s00704-013-0834-z, 2013.
- 334 Ivanov, M. and Kotlarski, S.: Assessing distribution-based climate model bias correction methods over
335 an alpine domain: added value and limitations, *Int. J. Climatol.*, 37, 2633–2653, doi:
336 10.1002/joc.4870, 2016.
- 337 Jacob, D., et al.: EURO-CORDEX: new high-resolution climate change projections for European
338 impact research, *Regional Environmental Change*, 14, 563–578, 2014.
- 339 Kotlarski, S., Keuler, K., Christensen, O. B., Colette, A., Déqué, M., Gobiet, A., Goergen, K., Jacob,
340 D., Lüthi, D., van Meijgaard, E., Nikulin, G., Schär, C., Teichmann, C., Vautard, R., Warrach-Sagi,
341 K., and Wulfmeyer, V.: Regional climate modeling on European scales: a joint standard evaluation
342 of the EURO-CORDEX RCM ensemble, *Geosci. Model Dev.*, 7, 1297–1333,
343 <https://doi.org/10.5194/gmd-7-1297-2014>, 2014.
- 344 Kröner, N., Kotlarski, S., Fischer, E., Lüthi, D., Zubler, E., and Schär, C.: Separating climate change
345 signals into thermodynamic, lapse-rate and circulation effects: theory and application to the
346 European summer climate, *Clim. Dyn.* 48, 3425–3440, 2017



- 347 Lehner, F., Keller, K., Raible, C. C., Mignot, J., Born, A., Joos, F., and Stocker, T. F.: Climate and
348 carbon cycle dynamics in a CESM simulation from 850 - 2100CE, *Earth System Dynamics*, 6,
349 411 - 434, 2015.
- 350 Messmer, M., Gomez-Navarro, J. J., and Raible, C. C.: Climatology of Vb-cyclones, physical
351 mechanisms and their impact on extreme precipitation over Central Europe, *Earth System*
352 *Dynamics*, 6, 541–553, 2015.
- 353 Messmer, M., Gomez-Navarro, J. J., and Raible, C. C.: The Impact of Climate Change on the
354 Climatology of Vb-Cyclones, *Tellus*, submitted.
- 355 Pfahl S.: Characterising the relationship between weather extremes in Europe and synoptic circulation
356 features, *Nat. Hazards Earth Syst. Sci.* 14, 1461–1475, 2014.
- 357 Pfahl, S., O’Gormann, P. A., and Fischer, E. M.: Understanding the regional pattern of projected
358 future changes in extreme precipitation, *Nature Geoscience*, 7, 423–427, 2017.
- 359 Prein, A. F., Langhans, W., Fossier, G., Ferrone, A., Ban, N., Goergen, K., Keller, M., Tölle, M.,
360 Gutjahr, O., Feser, F., Brisson, E., Kollet, S., Schmidli, J., van Lipzig, N. P. M., and Leung, R.: A
361 review on regional convection-permitting climate modeling: Demonstrations, prospects, and
362 challenges, *Rev. Geophys.*, 53, 323–361, doi:10.1002/2014RG000475, 2015.
- 363 Rajczak J., Kotlarski S., and Schär, C.: Does quantile mapping of simulated precipitation correct for
364 biases in transition probabilities? *J. Clim.*, 29, 1605–1615, doi: 10.1175/JCLI-D-15-0162.1, 2016a.
- 365 Rajczak J., Kotlarski, S., Salzmann N., and Schär, C.: Robust climate scenarios for sites with sparse
366 observations: A two-step bias correction approach, *Int. J. Climatol.*, 36, 1226–1243. doi:
367 10.1002/joc.4417, 2016b.
- 368 Rajczak, J., Brönniman, S., Fischer, E. M., Raible, C. C., Rohrer, M., Sørland, S. L., and Schär, C.:
369 Daily precipitation and temperature series from multiple climate model simulations for the Aare
370 river catchment in Switzerland, *Earth System Science Data*, submitted, 2018 (paper: essd-2018-26).
- 371 Rätty, O., Räisänen, J., and Ylhäisi, J.: Evaluation of delta change and bias correction methods for
372 future daily precipitation: intermodel cross-validation using ENSEMBLES simulations, *Climate*
373 *Dynamics* 42, 2287–2303, doi:10.1007/s00382-014-2130-8, 2014.
- 374 Rohrer, M., Brönnimann, S., Martius, O., Raible, C. C., Wild, M., and Compo, C. P. (2018)
375 Representation of extratropical cyclones, blocking anticyclones, and Alpine circulation types in
376 multiple reanalyses and model simulations, *J. Clim.* (accepted).
- 377 Scherrer, S. C., Fischer, E. M., Posselt, R., Liniger, M. A., Croci-Maspoli, M., and Knutti, R.:
378 Emerging trends in heavy precipitation and hot temperature extremes in Switzerland, *J. Geophys.*
379 *Res. Atmos.*, 121, doi:10.1002/2015JD024634, 2016.
- 380 Taylor, K. E., Stouffer, R. J., and Meehl, G. A.: An Overview of CMIP5 and the Experiment Design,
381 *B. Am. Meteorol. Soc.*, 93, 485–498. doi:10.1175/BAMS-D-11-00094.1, 2012.



- 382 Teutschbein, C., and Seibert, J.: Bias correction of regional climate model simulations for hydrological
383 climate-change impact studies: Review and evaluation of different methods. *Journal of Hydrology*,
384 456–457, 12-29, 2012.
- 385 Themessl, M., Gobiet, A., and Leuprecht, A.: Empirical-statistical downscaling and error correction of
386 daily precipitation from regional climate models, *Int. J. Climatol.*, 31, 1530–1544, 2011.
- 387 Themessl, M., Gobiet, A., and Leuprecht, A.: Empirical-statistical downscaling and error correction of
388 regional climate models and its impact on the climate change signal, *Climatic Change*, 112, 449–
389 486, doi: 10.1007/s10584-011-0224-4, 2012.
- 390 Van Bebber, W.: Die Zugstrassen der barometrischen Minima nach den Bahnenkarten der deutschen
391 Seewarte für den Zeitraum 1875–1890, *Meteorol. Z.*, 8, 361–366, 1891.
- 392 Van der Linden, P. and Mitchell, J. F. B.: ENSEMBLES: Climate Change and Its Impacts: Summary
393 of Research and Results From the ENSEMBLES Project, MetOffice Hadley Centre, Exeter, UK,
394 2009.
- 395 Van Vuuren, D. P., Edmonds, J., Kainuma, M., Riahi, K., Thomson, A., Hibbard, K., Hurtt, G. C.,
396 Kram, T., Krey, V., Lamarque, J.-F., Masui, T., Meinshausen, M., Nakicenovic, N., Smith, S. J.,
397 Rose, S. K.: The representative concentration pathways: an overview, *Climatic Change*, 109, 5–31,
398 doi:10.1007/s10584-011-0148-z, 2011.
- 399 Wilcke, R. A. I., Mendlik, T., and Gobiet, A.: Multi-variable error correction of regional climate
400 models. *Climatic Change*, 120, 871–887, doi: 10.1007/s10584-013-0845-x, 2013.
- 401 Zhang, X., Zwiers, F. W., Li, G., Wan, H, and Cannon, A. J.: Complexity in estimating past and future
402 extreme short-duration rainfall, *Nature Geoscience* 10, 255–259, 2017.



403 **Tables**

404 **Table 1.** Overview of model experiments used in our study (see Table S1 for further details)

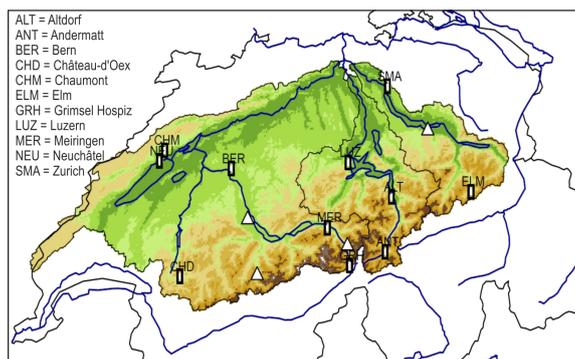
Experiment	Type	Resol.	<i>n</i>	Period	Scenario	Reference
CCC400	Global GCM	2°	30+1*	1600–2005	Historical	Bhend et al. (2012) Rohrer et al. (2018)
CMIP5	Global AOGCM	various	25	1901–2100	Historical, RCP8.5	Taylor et al. (2012)
ENSEMBLES	Regional	25 km	13	1970–2099	A1B	van der Linden and Mitchell (2009)
COSMO	Regional	0.44°	21	1950–2100	RCP8.5	
CORDEX-11	Regional	0.11°	15	1971–2099	RCP8.5	
CORDEX-44	Regional	0.44°	19	1971–2099	RCP8.5	
CORDEX-11	Regional	0.11°	15	1971–2099	RCP4.5	Jacob et al. (2014); Kotlarski et al. (2014)
CORDEX-44	Regional	0.44°	17	1971–2099	RCP4.5	

405 * Trends in some land-surface properties were mis-specified in the 30-member ensemble, an additional member with
406 corrected land-surface properties was performed to assess the errors (Rohrer et al., 2018). The figures in this paper show the
407 30 member ensemble; results for the land-surface corrected member share the same features and are shown in Fig. S3.



408 **Figures**

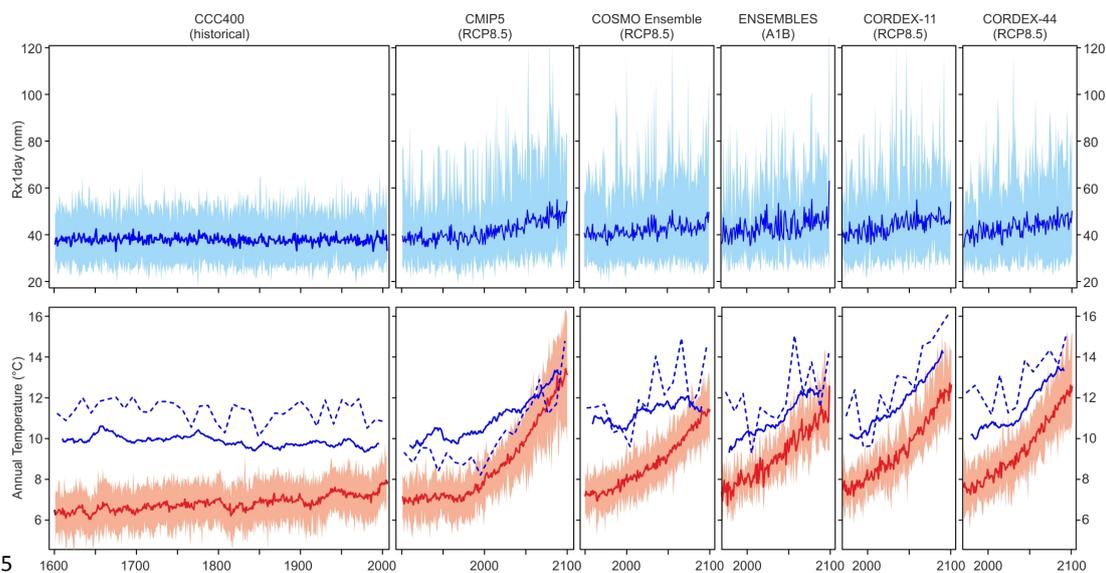
409



410

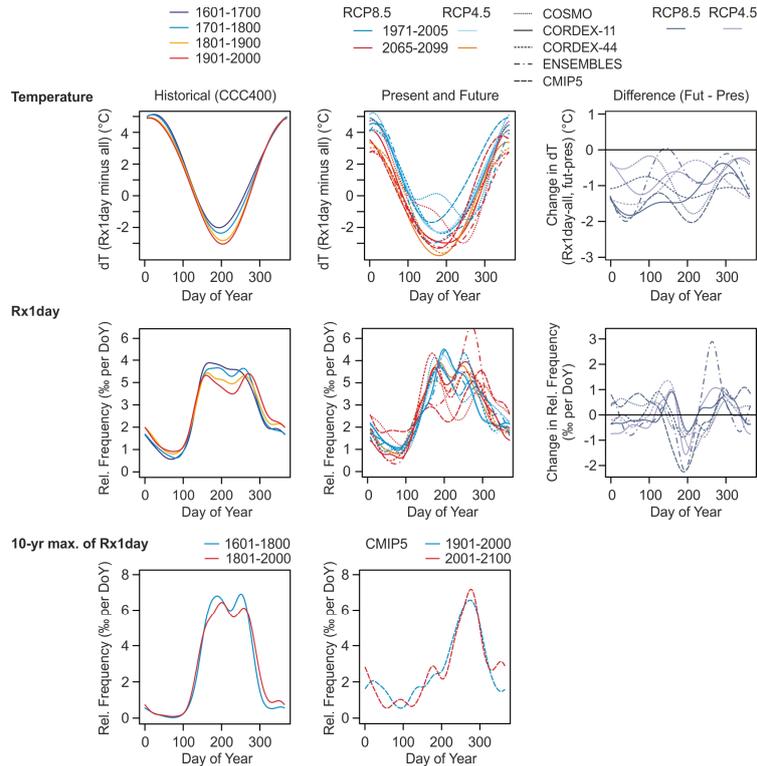
411 **Fig. 1:** Catchment of the Aare river. All data used in this study represent averages over this region.
 412 The representative weather stations are shown as squares (providing both temperature and
 413 precipitation) and triangles (only precipitation).

414



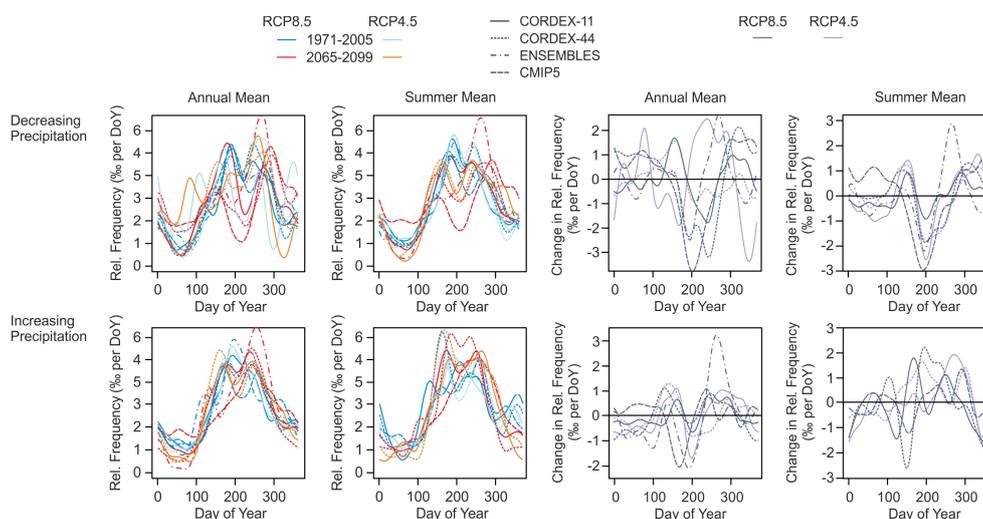
415

416 **Fig. 2:** Annual time series (ensemble mean as line and ensemble range as shading) of (top) Rx1day
 417 and (bottom) annual mean temperature as well as the temperature during the Rx1day event (blue solid
 418 line, smoothed with a 20-yr running average) in CCC400, CMIP5 and COSMO (quantile mapped).
 419 Also shown is the temperature at the highest Rx1day event per decade (blue dashed; ensemble mean).



420

421 **Fig. 3:** (top row) Temperature difference as a function of calendar day between event days (Rx1day)
 422 and all days for different time periods in CCC400 (left) and in model simulations of the present
 423 (1971–2005) and future (2065–2099) (middle; the right figure shows the difference between the two
 424 periods). (middle row) Density plot of the day of occurrence of Rx1day events for different time
 425 periods in CCC400 (left) and in model simulations of the present (1971–2005) and future (2065–2099)
 426 (middle; the right figure shows the difference between the two periods). (bottom row) Same as second
 427 row, but only for the highest Rx1day per decade. Note that only CCC400 and CMIP5 have sufficiently
 428 large ensembles, and longer time periods were chosen. All analyses are based on quantile-mapped data
 429 (see Fig. S3 for additional plots and Fig. S2 for the same analysis based on raw data). A Gaussian
 430 Kernel smoother with a bandwidth of 15 days was used for plotting.



431

432 **Fig. 4:** Density plot of the day of occurrence of Rx1 day for the present (1971–2005) and future (2065–
 433 2099) in different the multi-model ensembles (left two columns). The right two columns show the
 434 difference between future and present climate. Ensemble members are separated into those that show a
 435 positive or negative trend in annual mean (first and third column) or summer mean (second and fourth
 436 column) precipitation over the 1971–2099 period (see Fig. S4 for additional plots from other scenarios
 437 and different time periods).