



AMERICAN METEOROLOGICAL SOCIETY

Bulletin of the American Meteorological Society

EARLY ONLINE RELEASE

This is a preliminary PDF of the author-produced manuscript that has been peer-reviewed and accepted for publication. Since it is being posted so soon after acceptance, it has not yet been copyedited, formatted, or processed by AMS Publications. This preliminary version of the manuscript may be downloaded, distributed, and cited, but please be aware that there will be visual differences and possibly some content differences between this version and the final published version.

The DOI for this manuscript is doi: 10.1175/BAMS-D-15-0004.1

The final published version of this manuscript will replace the preliminary version at the above DOI once it is available.

If you would like to cite this EOR in a separate work, please use the following full citation:

Kendon, E., N. Ban, N. Roberts, H. Fowler, M. Roberts, S. Chan, J. Evans, G. Fosser, and J. Wilkinson, 2016: Do convection-permitting regional climate models improve projections of future precipitation change? Bull. Amer. Meteor. Soc. doi:10.1175/BAMS-D-15-0004.1, in press.



**Do convection-permitting regional climate models improve projections of
future precipitation change?**

Elizabeth J. Kendon*

Met Office Hadley Centre, Exeter, UK

Nikolina Ban

Institute for Atmospheric and Climate Science, ETH Zurich, Switzerland

Nigel M. Roberts

MetOffice@Reading, Reading, UK

Hayley J. Fowler

School of Civil Engineering and Geosciences, Newcastle University, UK

Malcolm J. Roberts

Met Office Hadley Centre, Exeter, UK

Steven C. Chan

School of Civil Engineering and Geosciences, Newcastle University, UK

Jason P. Evans

University of New South Wales, Australia

17

Giorgia Fosser

18

CNRM-GAME, CNRS and Météo-France, Toulouse, France, now at Met Office Hadley Centre,

19

Exeter, UK.

20

Jonathan M. Wilkinson

21

Met Office, Exeter, UK

22 *Corresponding author address: Met Office Hadley Centre, Fitzroy Road, Exeter, EX1 3PB, UK.

23 E-mail: elizabeth.kendon@metoffice.gov.uk

ABSTRACT

24 Regional climate projections are used in a wide range of impact studies,
25 from assessing future flood risk to climate change impacts on food and en-
26 ergy production. These model projections are typically at 12-50km resolution,
27 providing valuable regional detail but with inherent limitations, in part due to
28 the need to parameterise convection. The first climate change experiments
29 at convection-permitting resolution (kilometre-scale grid spacing) are now
30 available for the UK, Alps, Germany, Sydney Australia and western USA.
31 These models give a more realistic representation of convection, and are bet-
32 ter able to simulate hourly precipitation characteristics that are poorly rep-
33 resented in coarser resolution climate models. Here we examine these new
34 experiments to determine whether future mid-latitude precipitation projec-
35 tions are robust from coarse to higher resolutions, with implications also for
36 the tropics. We find that the explicit representation of the convective storms
37 themselves, only possible in convection-permitting models, is necessary for
38 capturing changes in the intensity and duration of summertime rain on daily
39 and shorter timescales. Other aspects of rainfall change, including changes in
40 seasonal mean precipitation and event occurrence, appear robust across res-
41 olution and therefore coarse resolution regional climate models are likely to
42 provide reliable future projections, provided that large-scale changes from the
43 global climate model are reliable. The improved representation of convective
44 storms also has implications for projections of wind, hail, fog and lightning.
45 We identify a number of impact areas, especially flooding, but also transport
46 and wind energy, for which very high resolution models may be needed for
47 reliable future assessments.

(Capsule Summary) **Climate change experiments at very high (kilometre-scale) resolution are now available which provide potential added value to future projections for convective precipitation, wind gusts, hail, fog and lightning.**

1. Introduction

A recent step change in climate modelling capability has allowed a number of international groups to carry out very high resolution (kilometre-scale) regional climate model experiments. This provides an opportunity to review the extent to which currently available regional climate projections from coarser resolution models are reliable. In this paper, we examine whether very high resolution models provide new or different precipitation projections from traditional coarser resolution climate models, and equally importantly where results are robust across resolutions. Our aim is to provide guidance on the extent to which current regional and national climate scenarios, all of which are based on coarse resolution model output, provide reliable information with which to inform policy decisions and impacts assessments, and which information must be updated using very high resolution projections.

Global climate models (GCMs) are our primary tool for understanding how climate may change in the future with increasing greenhouse gases. These typically have coarse resolutions with grid spacings of 60-300km. To provide regional detail, higher resolution regional climate models (RCMs; 12-50km grid spacing) are often used, which only span a limited area. Since different models typically represent key small-scale processes in the climate system in slightly different ways, an ensemble of multiple GCMs and RCMs is often used to give an estimate of modelling uncertainty in regional climate projections. Recent examples of coordinated regional climate modelling experiments are the ENSEMBLES (Hewitt and Griggs 2004), NARCCAP (Mearns et al. 2009), NARCLiM (Evans et al. 2014) and CORDEX (Giorgi et al. 2009) projects, which use a

71 multi-model approach to downscale to 12-50km resolution. An alternative approach was taken
72 by the UK climate projections (UKCP09, Murphy et al. 2009) which used a perturbed physics
73 ensemble, whereby uncertain model parameters were varied within a single GCM to give prob-
74 abilistic projections which sampled both parameter uncertainty and internal climate variability.
75 Information from a multi-model ensemble was also incorporated, as well as information from ob-
76 servations to weight different model versions depending on their ability to accurately simulate the
77 historical climate. In UKCP09, the GCM simulations were run at a resolution of 300km and were
78 then downscaled using an ensemble of RCMs at 25km resolution.

79 Although these ensemble based approaches give some estimate of modelling uncertainty, they
80 do not reveal uncertainties that arise from limitations inherent in all models at those resolutions.
81 At typical climate model resolutions (10-100 km) many important processes (such as those in
82 clouds) occur on spatial scales too small to be resolved explicitly on the model grid and are there-
83 fore represented using parameterisations. A good example of this is convection parameterisation,
84 which aims to describe the average properties of convection over a model grid box but leads to
85 deficiencies in the diurnal cycle of convection, precipitation occurrence and extremes (Dai 2006;
86 Hohenegger et al. 2008; Stephens et al. 2010).

87 The first climate change experiments at very high resolution (<5 km grid spacing) have recently
88 been completed for the UK (Kendon et al. 2014), the Alpine region (Ban et al. 2015), central
89 Germany (Tolle et al. 2014), southwestern Germany (Fosser et al. in press), Sydney Australia (Ar-
90 gueso et al. 2014), the Colorado headwaters (Rasmussen et al. 2014) and the western USA (Pan
91 et al. 2011). At these resolutions, the deep convective parameterisation scheme can be switched
92 off, with the use of shallow convective parameterisation varying between studies. Deep convec-
93 tion schemes are not designed to operate in km-scale models, and many of the assumptions of
94 these schemes (e.g. that the cloud coverage is small compared to the grid square) are violated

at these resolutions. Shallow plumes typically have smaller horizontal length scales and without some kind of parameterisation there is a missing process (small cumulus clouds are too small to be explicitly represented on the grid), but the use of shallow convective parameterisation is only appropriate if the scheme has been designed to operate at km-scale. At this scale, the model is termed ‘convection-permitting’ because larger convective storms are ‘permitted’ but convective plumes and smaller showers are still not resolved. Such convection-permitting models are commonly used in short-range weather forecasting, where they have been shown to give a much more realistic representation of convection and can be used to forecast the possibility of localised high-impact rainfall not captured at coarser resolutions (Done et al. 2004; Lean et al. 2008; Weisman et al. 2008; Weusthoff et al. 2010; Schwartz 2014). However, due to their high computational cost, they have not commonly been applied at climate-time scales. Convection-permitting models do not necessarily better represent daily mean precipitation (Chan et al. 2013), but have significantly better sub-daily rainfall characteristics with improved representation of the diurnal cycle of convection (Ban et al. 2014), the spatial structure of rainfall and its duration-intensity characteristics (Kendon et al. 2012), and the intensity of hourly precipitation extremes (Chan et al. 2014b; Ban et al. 2014; Fosser et al. 2015), which are typically poorly represented in climate models. Prein et al. (2015) provide an excellent review of the added value of convection permitting climate modelling, showing added value emerging when and where deep convection is dominant and in regions of strong spatial heterogeneities (mountains and urban areas). Although convection-permitting models still contain errors, e.g. smaller showers are not properly resolved leading to a tendency for heavy rainfall to be too intense (Kendon et al. 2012; Fosser et al. 2015), they provide a step change in our ability to represent convection and an opportunity to examine the importance of representing local storms for future climate projections.

118 In this paper, we bring together results from climate change experiments performed at 1.5km
119 resolution over the southern UK (Kendon et al. 2014) and 2.2km over the Alpine region (Ban et al.
120 2015) to explore commonalities, and also refer to findings from Numerical Weather Prediction
121 (NWP) and convection-permitting climate studies recently completed for other regions. The UK
122 simulations span the southern UK and are driven by a 12km RCM which spans Europe and is
123 in turn driven by a 60km GCM, with all models being configurations of the Met Office Unified
124 Model (Walters et al. 2011). The Alps simulations span the greater Alpine region and are driven by
125 a 12km European RCM, which is in turn driven by a T63 spectral (~ 200 km) resolution GCM. In
126 this case the RCM is the COSMO-CLM model, whilst the GCM is the Earth System Model of the
127 Max Planck Institute (MPI-ESM-LR) (Stevens et al. 2013). In both cases the simulations are for
128 two ~ 10 -year periods, one corresponding to the present-day and the other to the end of the century
129 under RCP8.5 (the exact period and run length varies between experiments, see Supplementary
130 Material for details). Given in each case we are restricted to only one model realisation, we are
131 unable to assess uncertainty in the climate change signal. However where commonalities in the
132 effect of resolution are identified across different regions and different climate models, we have
133 greater confidence in the result.

134 This paper complements the review of Prein et al. (2015), which mostly focussed on the
135 added value of convection-permitting simulations for present-day climate variability, by assess-
136 ing whether such very high resolution models are needed for reliable future projections. We focus
137 on rainfall projections, since these are expected to be most sensitive to model resolution and the
138 explicit representation of convection. We also consider implications for other climate variables,
139 including wind, hail, fog, lightning and soil moisture.

2. To what extent do precipitation projections agree between coarse and high resolution RCMs?

a. Coarse resolution RCMs are likely to provide robust projections of changes in seasonal mean rainfall, providing that large-scale changes from the GCM are reliable.

We find that projected changes in seasonal mean rainfall are in close agreement between high resolution (convection-permitting) and coarser resolution (convection-parameterised) regional climate models, driven by the same large scale conditions. For example, Figures 1 and 2 show projected changes to winter and summer rainfall over the southern UK for a 1.5km and 12km RCM (described in Kendon et al. 2014). Similar agreement in projected changes in summer mean precipitation is found in Ban et al. (2015) for 2.2km and 12km RCMs over the Alps and in Fosser et al. (in press) for 2.8km and 7km models over SW Germany. For summer, this agreement is probably because the convection scheme is able to (and indeed designed to) simulate time-average precipitation, even if the diurnal cycle is incorrect (too strong midday peak) and short-timescale intensity is too weak; and for winter, because mean rainfall is dominated by larger-scale dynamical rainbands that are well captured at 12km. We note that Rasmussen et al. (2014) found reduced biases in the representation of mean summer rainfall in a convection-permitting model over the Colorado headwaters, which may impact on future projected changes, but the improvement here is likely to be linked specifically to the improved representation of complex topography at high resolution.

Changes in daily and hourly rainfall are, however, more important than changes in seasonal averages for many climate impacts. In particular, changes in the intensity and duration of rainfall are crucial for understanding future flood risk and changes in the occurrence of rainfall are important

162 for water resource planning and agriculture. Hence we now turn our attention to examine whether
163 these aspects of rainfall change are robust from coarser to higher model resolution.

164 *b. Changes in rainfall occurrence are largely consistent between convection-permitting and*
165 *coarser resolutions in summer and winter*

166 We find that changes in rainfall occurrence are largely consistent between the convection-
167 permitting and parent 12km RCM, in winter and summer, for both the UK and Alps simulations
168 (Figures 3 and 4) with large decreases in rainfall occurrence in summer at both resolutions. For the
169 UK, this is despite present-day biases being quite different between the two different resolution
170 models, with the 12km RCM tending to have too many wet hours and wet days, whilst the 1.5km
171 RCM has too few especially in summer. Fosser et al. (in press) also show a similar decrease in
172 summer rainfall occurrence in 2.8km and 7km models, despite quite different biases across model
173 resolution for the present-day. Large RCM biases for the Alps simulations are largely inherited
174 from the driving GCM, with biases for European Reanalysis (ERA) Interim (Dee et al. 2011)
175 driven simulations being much smaller (see Supplementary Figures 1 and 2). Known problems
176 with radar over mountainous regions, and sampling errors associated with gauge observations,
177 may also contribute to the apparently larger biases over the Alps compared to the UK. The ten-
178 dency for too much low intensity precipitation in the Alps simulations in winter is most apparent
179 on the hourly timescale, since when averaging over longer periods, multiple hourly occurrences
180 still give the same daily count as a single occurrence on that day. We note that although the differ-
181 ences between the Alps and UK results are probably in part due to the driving model differences
182 and/or verification data, there are also likely to be some differences that come from the contrast
183 in topography and nature of the rainfall between the two regions (Alps drier in winter and more
184 convective in summer in comparison to the UK). We expect high resolution to be beneficial for

185 both frontal and convective rain over regions of steep orography, with the occurrence and spatial
186 patterns of rainfall better captured over mountainous regions (Prein et al. 2015), whereas in flatter
187 regions frontal rain (especially in winter) is sufficiently well captured at coarser resolutions. In
188 general, the improved representation of orography at high resolution is expected to impact rainfall
189 on longer temporal scales, with model biases focussed on the hourly timescale expected to be more
190 related to the representation of convection.

191 Chan et al. (2014a) show that over the UK the large-scale conditions from the driving GCM
192 control whether precipitation is triggered or not, but how long the precipitation lasts once trig-
193 gered is sensitive to the RCM model physics (and is different between convection-permitting and
194 convection-parameterised models). The former (i.e. changes in triggering) dominates the future
195 decrease in summer rainfall occurrence, whereas the latter (i.e. how long it lasts) leads to the
196 resolution dependence of present-day biases. Thus coarser resolution RCMs are likely to be suf-
197 ficient for projecting changes in the occurrence of rainfall where this is dominated by changes to
198 the triggering of events. Since dry spell length is determined by the sequence of rainfall events
199 (synonymous with triggering), we expect changes in dry spell length to also be robustly captured
200 at coarser resolution. We note this result may not extend to those tropical regions where large-scale
201 forcing is less dominant in triggering events, although may apply in some tropical areas where the
202 large-scale matters more (e.g. African Easterly Waves, Indian Monsoon).

203 *c. Changes in rainfall intensity are robust across model resolution in winter, but show significant*
204 *differences in summer*

205 We find good agreement between convection permitting and 12km RCM results, for both the
206 UK and Alps simulations, for change to winter precipitation intensities (Figure 5). This is despite
207 resolution-dependent model biases for the present-day in some cases. The agreement in changes

208 across resolution can be explained by the fact that changes in intensity over Europe in winter
209 come predominantly from frontal rain from mid-latitude weather systems with greater moisture
210 availability (Kendon et al. 2014) - a process well captured by coarse resolution climate models. In
211 summer, by contrast, convection-permitting models show significant increases in rainfall intensity
212 over the southern UK and the Alps (Figure 6), that are not captured by the parent 12km RCM, for
213 a comparison made at the 12km scale (Kendon et al. 2014; Ban et al. 2015). This can be explained
214 by the 12km RCM being constrained by the behaviour of the convection parameterisation. These
215 increases become larger in relative terms at finer spatial and temporal scales. It is notable that
216 the pattern of future changes across space and time-scales, and across resolution, is remarkably
217 similar between the southern UK and the Alps simulations. The convection-permitting model
218 tends to have reduced biases in summer, although in the case of the Alps simulations, there are
219 large RCM biases inherited from the driving GCM (see Supplementary Figures 3 and 4). Fosser
220 et al. (in press) found contrasting results with similar increases in intensity for 2.8km convection-
221 permitting and 7km convection-parameterised models over SW Germany, although with different
222 spatial patterns. The agreement across resolution in this latter study may be due to the relatively
223 high resolution of the convection-parameterised model (7km compared to 12km in Kendon et al.
224 (2014) and Ban et al. (2015)).

225 We note that Ban et al. (2015) found greater consistency between the 2.2km and 12km RCMs
226 for a metric of heavy rainfall (specifically percentiles of all values instead of only wet values, and
227 hence a measure of rainfall intensity and frequency combined). However further analysis for the
228 UK reveals such agreement does not extend across a range of percentiles. Thus the finding that
229 changes in summertime rainfall intensity are resolution-dependent does not appear to simply be
230 an artefact of the chosen metric.

231 *d. Convection-permitting resolution is needed to capture changes in the duration of summertime*
232 *rainfall*

233 Projected changes in the duration of summertime rain are quite different between convection-
234 permitting and coarser resolution models over the southern UK (Kendon et al. 2014, and figures
235 therein). In the 12km RCM precipitation tends to be too low-intensity and too long-duration,
236 with these biases largely eliminated in the 1.5km model due to the better representation of con-
237 vective rain (Kendon et al. 2012). Fosser et al. (in press) found contrasting results with sim-
238 ilar changes in the intensity-duration characteristics of rainfall in 2.8km and 7km models over
239 SW Germany. This better agreement may be due to the relatively high resolution (7km) of the
240 convection-parameterised model or the small size of the 2.8km model domain, which may limit
241 the RCM’s ability to generate its own small-scale features.

242 *e. Changes in daily and hourly rainfall extremes are not robust across model resolution, particu-*
243 *larly in summer*

244 For the UK, projected changes to extreme hourly rainfall differ markedly between 1.5km and
245 12km resolution in summer, to the extent that they differ in sign (Chan et al. 2014a). This dis-
246 crepancy is also found for daily extremes, although to a lesser extent. Extreme events (with return
247 periods of greater than 20 years) in the 12km RCM are linked to single grid-point storms or storms
248 with unphysically large updraught regions, providing low confidence in projections. Results for
249 the Alps are consistent with results for the UK: Ban et al. (2015) find considerable discrepancies
250 in projected changes in extreme hourly intensities in summer between 2.2km and 12km resolu-
251 tion. In winter, there is some suggestion that projected changes in UK extreme daily rainfall over
252 orography may be greater at convection-permitting resolution (Chan et al. 2014a). This may be

253 explained by a stronger response associated with the more accurate representation of orography
254 in the high resolution model, which may affect daily rainfall accumulations more than 1-hour ac-
255 cumulations. We may expect the benefit of high resolution to be different over orography, with
256 improved representation of topography as well as improved representation of convective storms
257 impacting on precipitation. On hourly timescales, the most intense rainfall is from convective
258 storms rather than orographic rain, and hence on shorter timescales the improved representation
259 of convective processes should dominate as realised by commonalities between the Alps and UK
260 results.

261 **3. Implications for other climate variables**

262 In this section we consider how the improved representation of convection at the km-scale has
263 implications for other climate variables. In particular we consider wind, hail, fog, lightning and
264 soil moisture that in turn impacts local temperature. We note that the improved representation of
265 local topography and surface heterogeneities in convection-permitting models will also have sig-
266 nificant impact, for example for projections of urban temperature (Trusilova et al. 2013; Argueso
267 et al. 2014, 2015) and climate over mountainous regions (Knote et al. 2010; Rasmussen et al.
268 2014). However, a discussion of this is beyond the scope of the current paper, which focuses on
269 the implications of the improved representation of the local storm dynamics.

270 *a. Convection-permitting resolution is needed to capture changes in severe wind gusts*

271 Kilometre-scale models allow a more accurate representation of local wind, in part through the
272 better representation of topography (e.g. sea-breezes, mountain effects), but also as a consequence
273 of capturing convective storms. Modelling convective wind gusts requires a convection-permitting
274 model, for example derechos are severe convective squall lines with intense winds which can only

275 be represented at convection-permitting scales. Over the UK, such systems are rare and only likely
276 to occur in summer. In winter, the highest winds on sub-daily timescales are typically associated
277 with cyclonic storms, and coarse resolution models ($\sim 12\text{-}25\text{km}$ grid spacing) with appropriate gust
278 diagnostics are able to reasonably well represent these. However, there will still be some situations
279 when local processes dominate in the smallest mesoscale systems (e.g. embedded convection or
280 small ‘sting-jets’ (Browning 2004)).

281 *b. Convection-permitting resolution is likely to give more reliable projections of future changes in*
282 *hail, but further research is needed*

283 Hail is of particular interest, being responsible for increasingly significant economic damages
284 to buildings, crops, cars and other infrastructure. Two events in central and southern Germany
285 on 27 and 28 July 2013 caused the highest insured loss from a natural hazard in Germany to
286 date. A number of studies have shown that convection-permitting climate models are able to
287 provide useful guidance of the occurrence of hail (Trapp et al. 2011; Gensini and Mote 2014).
288 Mahoney et al. (2012) investigated future changes in hail in convection-permitting simulations
289 using a case study approach and found that although more hail was generated within the cloud with
290 storm intensification, little reached the surface due to enhanced melting. It is unclear, however,
291 whether such high resolutions are required to provide reliable projections of future changes in
292 hail and whether the hail produced by microphysics schemes and post-processing algorithms is
293 sufficiently good. In particular, coarser resolution models may be sufficient, if post-processed hail
294 diagnostics (e.g. Hand and Cappelluti 2011) are sufficiently accurate, although hail diagnosis from
295 convection-permitting models should still give a better physical representation and is less reliant
296 on arguably tenuous links between large-scale environmental conditions and small-scale weather
297 extremes (Mahoney et al. 2012).

298 *c. Further research is required to assess whether very high resolution is needed for projecting*
299 *future changes in fog*

300 The UKCP09 25km models suggest the frequency of fog will be reduced in future across the
301 UK in many regions and seasons, but with considerable uncertainties. Where changes in fog
302 are driven by large-scale variables such as temperature and humidity, we may be able to make
303 confident statements about large-scale changes in fog. However on the local scale, fog depends on
304 many different variables including topography, aerosol amounts and local turbulent fluctuations.
305 These processes may be better represented in a convection-permitting model, although it is unclear
306 whether km-scale models provide more accurate future projections compared to post-processed
307 coarse model output. Boundary layer clouds may require resolution of turbulent eddies in some
308 cases, which would require grid scales of ‘Large Eddy Simulation’ models (tens of metres).

309 *d. Convection-permitting resolution has the potential to provide more accurate lightning projec-*
310 *tions, but further research is needed*

311 A lightning diagnostic has been developed for convection-permitting configurations of the Met
312 Office Unified Model which uses a physically-based link between cloud properties and lightning
313 flash rate (Wilkinson and Bornemann 2014). In particular lightning flash rate is determined from
314 the upward flux of graupel (McCaul et al. 2009), which in convection-parameterised models tends
315 to be too low. An alternative commonly used by convection-parameterised models is to link the
316 flash rate to bulk cloud properties (e.g. Price and Rind 1992). However, this statistical approach re-
317 moves the link between the cloud microphysical processes. It is only in the convection-permitting
318 model that we are able to use actual physical processes in the cloud to determine lightning. This

latter approach shows good skill in forecasting the timing and occurrence of lightning but overpredicts its extent, although its behaviour is yet to be fully determined over a long period.

e. Convection-permitting models may give different soil moisture conditions and feedbacks

The more realistic representation of rainfall in convection-permitting models impacts soil moisture conditions. In particular, in the present-climate, the southern UK 1.5km RCM has drier soils than the 12km RCM due to the more sporadic nature of rainfall, which is less effective at wetting the soils. Soil moisture conditions and surface evapotranspiration may in turn have a considerable impact on local temperature changes (Tolle et al. 2014). If soils become dry enough they can limit evaporation, leading to hotter temperatures. Soil-moisture precipitation feedbacks were also found to be very different in a convection-permitting model compared to a convection-parameterised model over the Alps (Hohenegger et al. 2009). The extent to which future projected changes in soil moisture are resolution-dependent is currently unexplored. However, the fact that soil moisture conditions and potentially their future change may be quite different in convection-permitting models has important implications for changes in temperature extremes and also for climate change impacts for example on agriculture.

4. Discussion

National climate change scenarios are currently available for many countries worldwide, for example, for the UK (UKCP09, Murphy et al. 2009), Netherlands (KNMI 2015) Switzerland (CH 2011) and U.S. (Melillo et al. 2014). These inform adaptation planning and often feed into downstream impact assessments to inform decisions in sectors such as transport, healthcare, water resources and flood protection. However, the quality of these national climate scenarios depends on the ability of the underlying model experiments to capture key processes and all are based

341 on coarse resolution climate model simulations. In this paper we have demonstrated how new,
342 very high resolution, regional climate models allow us to assess the robustness of current national
343 climate scenarios. We conclude by providing users information on where currently available pro-
344 jections are reliable and, conversely, where it is necessary to use results emerging from very high
345 resolution model experiments.

346 We have identified a number of aspects of mid-latitude precipitation change which disagree
347 significantly between convection-permitting and coarser resolution RCMs, and for which very high
348 resolution (km-scale) models are needed for accurate future projections. These include changes
349 in:

- 350 ● summertime rainfall intensity and duration
- 351 ● hourly and daily rainfall extremes in summer
- 352 ● daily precipitation extremes over mountains in winter

353 Other aspects of precipitation change appear to be reliably captured in currently available pro-
354 jections from regional climate models, providing large-scale changes from the driving GCM are
355 reliable. These include changes in:

- 356 ● seasonal mean precipitation
- 357 ● rainfall event occurrence
- 358 ● precipitation intensity in winter

359 In addition, we have identified other climate variables for which km-scale climate models are
360 likely to be needed for future projections, e.g. severe wind gusts; whilst for others, including hail,
361 fog and lightning, further research is required. A summary of these conclusions is presented in
362 Figure 7.

363 It is encouraging that the resolution dependence of projections for simulations over the UK and
364 the Alps show considerable consistency. This suggests firstly, that with increases in model resolu-
365 tion climate projections from different models may show convergence. Secondly, it indicates that
366 the conclusion that the local storm dynamics (only represented in convection-permitting models)
367 are needed for future projections of short-duration extremes is more widely applicable in other re-
368 gions, for convectively-dominated regimes and seasons. This is supported by some observational
369 evidence (Lenderink and van Meijgaard 2008). Although these results are for mid-latitude summer
370 precipitation, they are likely to be similar throughout the tropics, where much more precipitation
371 arises from smaller scale motions.

372 A move towards very high resolution climate modelling seems necessary for quantification of
373 certain impacts, particularly around extremes. Reliable projections of rainfall extremes are im-
374 portant for understanding future flood risk, and hence for informing decisions regarding urban
375 planning, flood protection and the design of resilient infrastructure. Surface water flooding and
376 flooding from drainage networks is the predominant source of flooding in urban areas as a direct
377 response to high intensity short-duration rainfall. In a recent pilot study, Dale et al. (2015) showed
378 that estimates of rainfall intensity change from the UK 1.5km climate model were higher than
379 existing UK climate change allowances, leading to projections of more frequent sewer overflow
380 spills in future. This is only one example but is indicative of the fact that current climate change
381 guidance based on national climate scenarios, e.g. current allowances for changes in peak rainfall
382 intensity in the UK, may not be adequate. There is a need to revisit current guidance in the light of
383 emerging results from very high resolution climate models. New information from these models
384 may also help to inform redesign of existing critical infrastructure systems. A number of recent
385 major flood events in the UK has brought this starkly into focus.

386 Our findings also have implications for a number of other impact areas (Table 1).

- a. Extreme winds can cause significant disruption to electricity infrastructure and transportation, and an understanding of future wind risk will inform prioritisation of adaptation investments. Km-scale models are needed to give improved projections of future change to severe wind gusts. These are also needed by the wind energy industry for the planning, design and operation of wind turbines.
- b. Lightning strikes can cause significant disruption to electricity infrastructure, and the potential for more accurate lightning predictions from convection-permitting climate models is likely to be of considerable interest to utility companies.

Multi-model ensemble experiments such as ENSEMBLES (Hewitt and Griggs 2004), NARC-CAP (Mearns et al. 2009), NARCLiM (Evans et al. 2014) and CORDEX (Giorgi et al. 2009) are limited by the inherent deficiencies in traditional convection-parameterised models. Although they provide valuable information about uncertainties in projections of seasonal mean rainfall and dynamically driven frontal systems, they cannot provide reliable estimates of changes in summertime rainfall intensity and duration or other convectively driven phenomena such as severe wind gusts. For this, explicit representation of convective storms, only possible in convection-permitting models, is crucial. However, currently only single model realisations over small domains are available at these resolutions. There is a need for an international effort, e.g. CORDEX, to provide coordinated multi-model experiments at convection-permitting resolutions over a series of common domains to estimate modelling uncertainty at these scales. This would allow a comprehensive evaluation of the potential for improving simulations of not only precipitation characteristics, but other aspects of climate such as land-atmosphere interactions, convective systems, and mountain or urban effects that are impacted by the improved representation of surface heterogeneities.

409 In the UK, an update to the UKCP09 projections is currently underway. This will include a
410 downscaling component, running an ensemble of km-scale models over the UK. This will hope-
411 fully allow projections to be provided on sub-daily and local scales, with some estimate of uncer-
412 tainty, which was beyond the UKCP09 modelling capability. We note however the very high cost
413 of convection-permitting simulations means it is not possible to run simulations at these scales for
414 many regions and for large ensembles of driving data. This impacts our ability to fully explore
415 uncertainties at these scales. This is of critical importance for climate change projections, which
416 are necessarily probabilistic and need to incorporate uncertainty due to natural climate variabil-
417 ity as well as modelling uncertainty. A potential approach for future progress could be through
418 combined statistical and dynamical downscaling. For example, it may be possible to sub-select
419 periods and ensemble members for dynamical downscaling from the large-scale conditions, and
420 thus achieve effective targeting of km-scale simulations. In particular, early work suggests statis-
421 tical regression relationships based on the large-scale climate state may be skilful in identifying
422 when local precipitation extremes may occur (although not their intensity or duration) and hence
423 when information from a convection-permitting model is needed. It should be noted although the
424 signal coming from the convection-permitting model is conditional on the larger-scale environ-
425 ment, it still provides a considerable improvement in local projections compared to solely relying
426 on coarser resolutions.

427 Despite the advent of km-scale regional climate models our confidence in projections is strongly
428 controlled by the ability of global climate models to represent relevant large-scale processes and
429 changes in large-scale circulation patterns. In parallel with developments in convection-permitting
430 regional climate modelling, there are now a number of examples of high resolution experiments
431 with global climate models. Long climate experiments have been performed within the UPSCALE
432 project (Mizielinski et al. 2014) with GCMs at 25km grid spacing. At resolutions of 60km or

433 higher, large-scale moisture transport seems to have converged in the mid-latitudes (Demory et al.
434 2014). The representation of many other processes is also found to improve with increasing model
435 resolution, e.g. atmospheric blocking (Scaife et al. 2011; Berckmans et al. 2013) and regional
436 modes of variability (MacLachlan et al. 2014), however, the convergence of these processes with
437 model resolution is not yet robustly established. The two simulations in the main examined here
438 used very different resolution driving GCMs. The Alps simulations showed significant biases
439 in large-scale conditions inherited from a $\sim 200\text{km}$ resolution GCM. These were found to im-
440 pact rainfall frequency and intensity over Switzerland in downscaling simulations (Supplementary
441 figures 1-4). By comparison, 60km GCM-driven downscaling simulations over the UK gave com-
442 parable biases to ERA-Interim driven simulations (Supplementary figures 5-8), suggesting that the
443 60km GCM is able to capture the synoptic and mesoscale variability important for constraining
444 local rainfall in this region. In general, present-day biases will only be a good guide of the reli-
445 ability of future projections, where the biases relate to the same processes which are controlling
446 the future change. It is essential that GCM ability to capture the key large-scale processes driving
447 future changes, which will vary between regions, is established before any regional downscaling
448 is attempted.

449 This study is not a comprehensive assessment of the reliability of regional projections from
450 coarse resolution models. It only gives an indication of the robustness of projections going from
451 coarser to higher resolution, and in particular to convection-permitting scales. The experiments
452 available to date are limited to single model realisations for different mid-latitude regions, and so
453 no estimate of modelling uncertainty in the climate change signal is possible. There are also many
454 other potential deficiencies and in some cases relating to limitations that are common to all current
455 climate models. For example, there are large uncertainties in earth system processes that impact on
456 atmospheric CO_2 concentrations for given CO_2 emissions, as well as provide feedbacks on climate

change; and uncertainties in processes occurring on very small scales, such as cloud microphysics. One important question is the extent to which we may expect changes to be robust on going to even higher resolutions. In the foreseeable future, it will likely become possible to run climate simulations at convection-resolving scales (grid scales of order 10m) for very small domains. Initial indications from numerical weather prediction are that the representation of convection does not necessarily improve with further increases in resolution once in the regime of being able to explicitly represent convection (Hanley et al. 2015), given our current knowledge about the representation of turbulence and microphysical processes. However, what we can say is that even at km-scale resolution, these models generate new climate projections with potentially profound implications for a number of sectors.

Acknowledgments. E.J. Kendon and M.J. Roberts gratefully acknowledge funding from the Joint Department of Energy and Climate Change (DECC) and Department for Environment Food and Rural Affairs (Defra) Met Office Hadley Centre Climate Programme (GA01101). This work also forms part of a joint UK Met Office and Natural Environment Research Council (UKMO-NERC) funded project on Convective Extremes (CONVEX, NE/1006680/1) and the European Research Council funded INTENSE project (ERC-2013-CoG). N. Ban is funded by the Swiss National Science Foundation through the grant 200021_132614 and through the Sinergia grant CRSII2_154486 crCLIM. H.J. Fowler is funded by the Wolfson Foundation and the Royal Society as a Royal Society Wolfson Research Merit Award (WM140025) holder. G. Fosser acknowledges funding from the French National Research Agency (ANR) within the project REMEMBER (contract ANR-12-SENV-001). J.P. Evans is funded through the Australian Research Council as part of the Future Fellowship FT110100576.

References

- Argueso, D., J. P. Evans, L. Fita, and K. J. Bormann, 2014: Temperature response to future urbanization and climate change. *Clim. Dyn.*, **42**, 2183–2199, doi:10.1007/s00382-013-1789-6.
- Argueso, D., J. P. Evans, A. J. Pitman, and A. Di Luca, 2015: Effects of city expansion on heat stress under climate change conditions. PLOS ONE.
- Ban, N., J. Schmidli, and C. Schar, 2014: Evaluation of the convection-resolving regional climate modeling approach in decade-long simulations. *J. Geophys. Res.*, **119** (13), 7889–7907, doi:10.1002/2014JD021478.
- Ban, N., J. Schmidli, and C. Schar, 2015: Heavy precipitation in a changing climate: Does short-term summer precipitation increase faster? *Geophys. Res. Lett.*, **42** (), 1165–1172, doi:10.1002/2014GL062588.
- Berckmans, J., T. Woollings, M.-E. Demory, P.-L. Vidale, and M. J. Roberts, 2013: Atmospheric blocking in a high resolution climate model: influences of mean state, orography and eddy forcing. *Atmos. Sci. Lett.*, **14**, 34–40, doi:10.1002/asl2.412.
- Browning, K. A., 2004: The sting at the end of the tail: Damaging winds associated with extratropical cyclones. *Q. J. R. Meteorol. Soc.*, **130** (597), 375–399, doi:10.1256/qj.02.143.
- CH, 2011: *Swiss climate change scenarios CH2011*. C2SM, MeteoSwiss, ETH, NCCR Climate and OcCC, Zurich, Switzerland, 88 pp. ISBN: 9783033030657
- Chan, S. C., E. J. Kendon, H. J. Fowler, S. Blenkinsop, C. A. T. Ferro, and D. B. Stephenson, 2013: Does increasing the spatial resolution of a regional climate model improve the simulated daily precipitation? *Clim. Dyn.*, **41** (5), 1475–1495, doi:10.1007/s00382-012-1568-9.

500 Chan, S. C., E. J. Kendon, H. J. Fowler, S. Blenkinsop, and N. M. Roberts, 2014a: Projected in-
501 creases in summer and winter UK sub-daily precipitation extremes from high resolution regional
502 climate models. *Environ. Res. Lett.*, **9** (), 084 019, doi:10.1088/1748-9326/9/8/084019.

503 Chan, S. C., E. J. Kendon, H. J. Fowler, S. Blenkinsop, N. M. Roberts, and C. A. T. Ferro, 2014b:
504 The value of high-resolution Met Office regional climate models in the simulation of multi-
505 hourly precipitation extremes. *J. Climate*, **27** (16), 6155–6174, doi:10.1175/JCLI-D-13-00723.
506 1.

507 Dai, A., 2006: Precipitation characteristics in eighteen coupled climate models. *J. Climate*, **19**,
508 4605–4630.

509 Dale, M., B. Luck, H. J. Fowler, S. Blenkinsop, E. Gill, J. Bennett, E. J. Kendon, and S. C.
510 Chan, 2015: New climate change rainfall estimates for sustainable drainage. *Proceedings of*
511 *ICE Engineering Sustainability*.

512 Dee, D. P., and Coauthors, 2011: The ERA-Interim reanalysis: configuration and performance of
513 the data assimilation system. *Q. J. R. Meteorol. Soc.*, **137** (656), 553–597, doi:10.1002/qj.828.

514 Demory, M.-E., P. L. Vidale, M. Roberts, P. Berrisford, J. Strachan, R. Schiemann, and
515 M. Mizieliński, 2014: The role of horizontal resolution in simulating drivers of the global hy-
516 drological cycle. *Clim. Dyn.*, **42**, 2201–2225, doi:10.1007/s00382-013-1924-4.

517 Done, J., C. A. Davis, and M. L. Weisman, 2004: The next generation of NWP: explicit forecasts
518 of convection using the weather research and forecasting (WRF) model. *Atmos. Sci. Lett.*, **5** (),
519 110–117.

520 Evans, J. P., F. Ji, C. Lee, P. Smith, D. Argueso, and L. Fita, 2014: Design of a regional climate
 521 modelling projection ensemble experiment - NARClIM. *Geosci. Model Devel.*, **7** (), 621–629,
 522 doi:10.5194/gmd-7-621-2014.

523 Fosser, G., S. Khodayar, and P. Berg, 2015: Benefit of convection permitting climate model sim-
 524 ulations in the representation of convective precipitation. *Clim. Dyn.*, **44** (1-2), 45–60, doi:
 525 DOI10.1007/s00382-014-2242-1.

526 Fosser, G., S. Khodayar, and P. Berg, in press: Climate change in the next 30 years: what can a
 527 convection-permitting model tell us that we did not already know? *Clim. Dyn.*, doi:10.1007/
 528 s00382-016-3186-4.

529 Gensini, V. A., and T. L. Mote, 2014: Estimations of hazardous convective weather in
 530 the United States using dynamical downscaling. *J. Climate*, **27**, 6581–6589, doi:10.1175/
 531 JCLI-D-13-00777.1.

532 Giorgi, F., C. Jones, and G. R. Asrar, 2009: Addressing climate information needs at the regional
 533 level: the CORDEX framework. *WMO Bull.*, **58** (3), 175–183.

534 Golding, B. W., 1998: Nimrod: A system for generating automated very short range forecasts.
 535 *Meteorol. Appl.*, **5**, 1–16, doi:10.1017/S1350482798000577.

536 Hand, W. H., and G. Cappelluti, 2011: A global hail climatology using the UK Met Office con-
 537 vection diagnosis procedure (CDP) and model analyses. *Meteorol. Appl.*, **18** (4), 446–458, doi:
 538 10.1002/met.236.

539 Hanley, K. E., R. S. Plant, T. H. M. Stein, R. J. Hogan, J. C. Nicol, H. W. Lean, C. Halliwell, and
 540 P. A. Clark, 2015: Mixing length controls on high resolution simulations of convective storms.
 541 *Q. J. R. Meteorol. Soc.*, **141** (686), 272–284, doi:10.1002/qj.2356.

542 Hewitt, C. D., and D. J. Griggs, 2004: Ensembles-based predictions of climate changes and their
543 impacts. *Eos, Trans. Am. Geophys. Union*, **85** (52), 566, doi:10.1029/2004EO520005.

544 Hohenegger, C., P. Brockhaus, C. S. Bretherton, and C. Schär, 2009: The soil moisture-
545 precipitation feedback in simulations with explicit and parameterized convection. *J. Climate*,
546 **22**, 5003–5020.

547 Hohenegger, C., P. Brockhaus, and C. Schär, 2008: Towards climate simulations at cloud-resolving
548 scales. *Meteorol. Z.*, **17** (4), 383–394, doi:10.1127/0941-2948/2008/0303.

549 Kendon, E. J., N. M. Roberts, H. J. Fowler, M. J. Roberts, S. C. Chan, and C. A. Senior, 2014:
550 Heavier summer downpours with climate change revealed by weather forecast resolution model.
551 *Nature Climate Change*, **4**, 570–576, doi:10.1038/nclimate2258.

552 Kendon, E. J., N. M. Roberts, C. A. Senior, and M. J. Roberts, 2012: Realism of rainfall
553 in a very high resolution regional climate model. *J. Climate*, **25**, 5791–5806, doi:10.1175/
554 JCLI-D-11-00562.1.

555 KNMI, 2015: *KNMI'14 climate scenarios for the Netherlands; A guide for professionals in cli-*
556 *mate adaptation*. De Bilt, The Netherlands, KNMI, 34 pp.

557 Knote, C., G. Heinemann, and B. Rockel, 2010: Changes in weather extremes: Assessment of
558 return values using high resolution climate simulations at convection-resolving scale. *Meteorol.*
559 *Z.*, **19** (1), 11–23, doi:10.1127/0941-2948/2010/0424.

560 Lean, H. W., P. A. Clark, M. Dixon, N. M. Roberts, A. Fitch, R. Forbes, and C. Halli-
561 well, 2008: Characteristics of high-resolution versions of the Met Office Unified Model for
562 forecasting convection over the United Kingdom. *Mon. Weather Rev.*, **136**, 3408–3424, doi:
563 10.1175/2008MWR2332.1.

- 564 Lenderink, G., and E. van Meijgaard, 2008: Increase in hourly precipitation extremes beyond
565 expectations from temperature changes. *Nature Geosci.*, **1**, 511–514, doi:10.1038/ngeo262.
- 566 MacLachlan, C., and Coauthors, 2014: Global Seasonal forecast system version 5 (GloSea5): a
567 high-resolution seasonal forecast system. *Q. J. R. Meteorol. Soc.*, doi:10.1002/qj.2396.
- 568 Mahoney, K., M. A. Alexander, G. Thompson, J. J. Barsugli, and J. D. Scott, 2012: Changes in
569 hail and flood risk in high-resolution simulations over Colorado’s mountains. *Nature Climate
570 Change*, **2** (2), 125–131, doi:10.1038/NCLIMATE1344.
- 571 McCaul, E. W., S. J. Goodman, K. M. LaCasse, and D. J. Cecil, 2009: Forecasting lightning
572 threat using cloud-resolving model simulations. *Weather and Forecasting*, **24**, 709–729, doi:
573 10.1175/2008WAF2222152.1.
- 574 Mearns, L. O., W. J. Gutowski, R. Jones, L.-Y. Leung, S. McGinnis, A. M. B. Nunes, and Y. Qian,
575 2009: A regional climate change assessment program for North America. *Eos, Trans. Am. Geo-
576 phys. Union*, **90** (32), 311–312.
- 577 Melillo, J. M., T. C. Richmond, and G. W. Yohe, Eds., 2014: *Climate Change Impacts in the
578 United States: The Third National Climate Assessment*, U.S. Global Change Research Program,
579 841 pp. doi:10.7930/J0Z31WJ2.
- 580 Mizielinski, M. S., and Coauthors, 2014: High resolution global climate modelling; the UP-
581 SCALE project, a large simulation campaign. *Geosci. Model Devel.*, **7** (), 1629–1640, doi:
582 10.5194/gmd-7-1629-2014.
- 583 Murphy, J. M., and Coauthors, 2009: *UK climate projections science report: Climate change
584 projections*. Met Office Hadley Centre, Exeter, UK.

585 Pan, L.-L., S.-H. Chen, D. Cayan, M.-Y. Lin, Q. Hart, M.-H. Zhang, Y. Liu, and J. Wang, 2011:
 586 Influences of climate change on California and Nevada regions revealed by a high-resolution
 587 dynamical downscaling study. *Clim. Dyn.*, **37**, 2005–2020, doi:10.1007/s00382-010-0961-5.

588 Prein, A. F., and Coauthors, 2015: A review on regional convection-permitting climate model-
 589 ing: demonstrations, prospects and challenges. *Rev. Geophys.*, **53** (2), 323–361, doi:10.1002/
 590 2014RG000475.

591 Price, C., and D. Rind, 1992: A simple lightning parameterization for calculating global lightning
 592 distributions. *J. Geophys. Res.*, **97** (D9), 9919–9933.

593 Rasmussen, R., and Coauthors, 2014: Climate change impacts on the water balance of the Col-
 594 orado headwaters: High-resolution regional climate model simulations. *J. Hydrometeorol.*,
 595 **15** (), 1091–1116, doi:10.1175/JHM-D-13-0118.1.

596 Scaife, A. A., and Coauthors, 2011: Improved Atlantic winter blocking in a climate model. *Geo-
 597 phys. Res. Lett.*, **38** (L23703), doi:10.1029/2011GL049573.

598 Schwartz, C. S., 2014: Reproducing the September 2013 record-breaking rainfall over the Col-
 599 orado Front Range with high-resolution WRF forecasts. *Wea. Forecasting*, **29** (), 393–402, doi:
 600 <http://dx.doi.org/10.1175/WAF-D-13-00136.1>.

601 Stephens, G. L., and Coauthors, 2010: Dreary state of precipitation in global models. *J. Geophys.*
 602 *Res.*, **115**, D24 211, doi:10.1029/2010JD014532.

603 Stevens, B., and Coauthors, 2013: Atmospheric component of the MPI-M Earth System Model:
 604 ECHAM6. *J. Adv. Model Earth Syst.*, **5** (2), 146–172, doi:10.1002/jame.20015.

605 Tolle, M. H., O. Gutjahr, G. Busch, and J. C. Thiele, 2014: Increasing bioenergy production on
606 arable land: Does the regional and local climate respond? Germany as a case study. *J. Geophys.*
607 *Res.*, **119** (6), 2711–2724, doi:10.1002/2013JD020877.

608 Trapp, R. J., E. D. Robinson, M. E. Baldwin, N. S. Diffenbaugh, and B. R. J. Schwedler, 2011:
609 Regional climate of hazardous convective weather through high-resolution dynamical down-
610 scaling. *Clim. Dyn.*, **37**, 677–688, doi:10.1007/s00382-010-0826-y.

611 Trusilova, K., B. Fruh, S. Brienens, A. Walter, V. Masson, G. Pigeon, and P. Becker, 2013: Im-
612 plementation of an urban parameterization scheme into the Regional Climate Model COSMO-
613 CLM. *J. Appl. Meteorol. Clim.*, **52** (10), 2296–2311, doi:10.1175/JAMC-D-12-0209.1.

614 Walters, D. N., and Coauthors, 2011: The Met Office Unified Model global atmosphere 3.0/3.1
615 and JULES global land 3.0/3.1 configurations. *Geosci. Model Devel.*, **4**, 919–941, doi:10.5194/
616 gmd-4-919-2011.

617 Weisman, M. L., C. Davis, W. Wang, K. W. Manning, and J. B. Klemp, 2008: Experiences with 0-
618 36-h explicit convective forecasts with the WRF-ARW model. *Weather and Forecasting*, **23** (3),
619 407–437.

620 Weusthoff, T., F. Ament, M. Arpagaus, and M. W. Rotach, 2010: Assessing the benefits of
621 convection-permitting models by neighbourhood verification: Examples from MAP D-PHASE.
622 *Mon. Weather Rev.*, **138**, 3418–3433.

623 Wilkinson, J. M., and F. J. Bornemann, 2014: A lightning forecast for the London 2012 Olympics
624 opening ceremony. *Weather*, **69** (1), 16–19, doi:10.1002/wea.2176.

625 Wuest, M., C. Frei, A. Altenhoff, M. Hagen, M. Litschi, and C. Schär, 2010: A gridded hourly
626 precipitation dataset for Switzerland using rain-gauge analysis and radar-based disaggregation.
627 *Int. J. Climatol.*, **30** (), 1764–1775.

628 **LIST OF TABLES**

629	Table 1.	Summary of currently available future climate projections and new information	
630		from high resolution climate modelling, for selected impact areas	33

Impact area	Currently available future climate projection	New information from high resolution climate modelling
Flash flooding (important in urban areas and small steep catchments)	Heavy daily rainfall is expected to increase globally, with projected increases across northern Europe and the UK in winter. Coarse resolution climate models are unable to provide reliable projections of future changes in short duration intense rainfall.	First evidence that intense rainfall events, associated with severe flash flooding (30mm in an hour), could become several times more frequent by the end of the century (Kendon et al. 2014).
Renewable energy (wind energy)	Future changes in wind are uncertain. 12-25km resolution models with appropriate gust diagnostics can represent cyclonic storms and their associated winds, but not the most severe convective wind gusts.	Kilometre-scale models are needed to represent severe wind gusts, associated with convective squall lines.
Transport (flooding, visibility, strong winds and snow)	Heavy daily rainfall is expected to increase (depending on region and season) with an associated increase in large scale flooding, but see above for flash flooding. There is large uncertainty in fog projections at 25km. Coarse resolution models should be sufficient for projecting changes in cyclonic storms and temperature-driven changes in snow.	See above for flash flooding. High resolution models are needed to adequately represent severe wind gusts and convective snow storms. High resolutions may be required for accurate projections of local fog and snow over mountains.
Electrical distribution (lightning)	25km models suggest increases in the number of lightning days in future across the UK, but there is considerable uncertainty in the accuracy of coarse model lightning diagnostics.	New lightning diagnostics, developed for kilometre scale models, have the potential for more accurate lightning predictions.

TABLE 1. Summary of currently available future climate projections and new information from high resolution climate modelling, for selected impact areas

633	LIST OF FIGURES	
634	Fig. 1.	Seasonal mean rainfall (mm/d) over the southern UK for winter (DJF) in the (a) radar, and
635		(b,d) model-radar differences (%) and (c,e) future changes (%) for the 1.5km and 12km
636		RCMs. The RCM simulations are for 13-year present-day (1996-2009) and 13-year future
637		(2100, under RCP8.5 scenario) periods. Radar data are for the period 2003-2012. Results
638		are only shown over UK land points. 35
639	Fig. 2.	As Figure 1 but for summer (JJA). 36
640	Fig. 3.	Biases and changes (%) in rainfall occurrence across space and time scales averaged over
641		the (top) southern UK and (bottom) Switzerland, in winter (DJF). Biases are calculated as
642		model-observation differences, using hourly 5km Nimrod radar data for the UK for 2003-
643		2012 (Golding 1998) and 1km Rdisaggh combined radar-gauge observations for Switzer-
644		land for 2004-2010 (Wuest et al. 2010). Changes correspond to differences between decadal-
645		length present-day and future (end of century, under RCP8.5) simulations. Results for the
646		convection-permitting (1.5km or 2.2km) RCM are shown in the upper left triangle and for
647		the 12km RCM in the lower right triangle. Grey indicates where results are not available. A
648		threshold of 0.1mm per accumulation period is used to define rainfall occurrence. 37
649	Fig. 4.	As Figure 3 but for summer (JJA). 38
650	Fig. 5.	Biases and changes (%) in rainfall intensity across space and time scales averaged over the
651		(top) southern UK and (bottom) Switzerland, in winter (DJF). Rainfall intensity is defined
652		as the mean of wet values (>0.1mm per accumulation period). Definitions are as in Figure 3. 39
653	Fig. 6.	As Figure 5 but for summer (JJA). 40
654	Fig. 7.	Schematic summarising where we have confidence in coarse resolution regional climate
655		model projections, and where very high resolution (km-scale) models are needed for accu-
656		rate projections 41

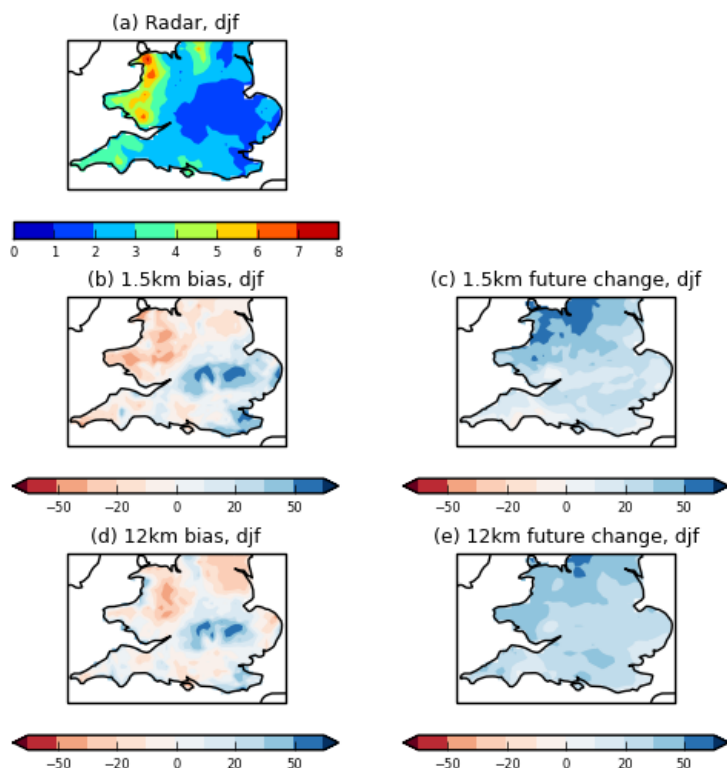


FIG. 1. Seasonal mean rainfall (mm/d) over the southern UK for winter (DJF) in the (a) radar, and (b,d) model-radar differences (%) and (c,e) future changes (%) for the 1.5km and 12km RCMs. The RCM simulations are for 13-year present-day (1996-2009) and 13-year future (2100, under RCP8.5 scenario) periods. Radar data are for the period 2003-2012. Results are only shown over UK land points.

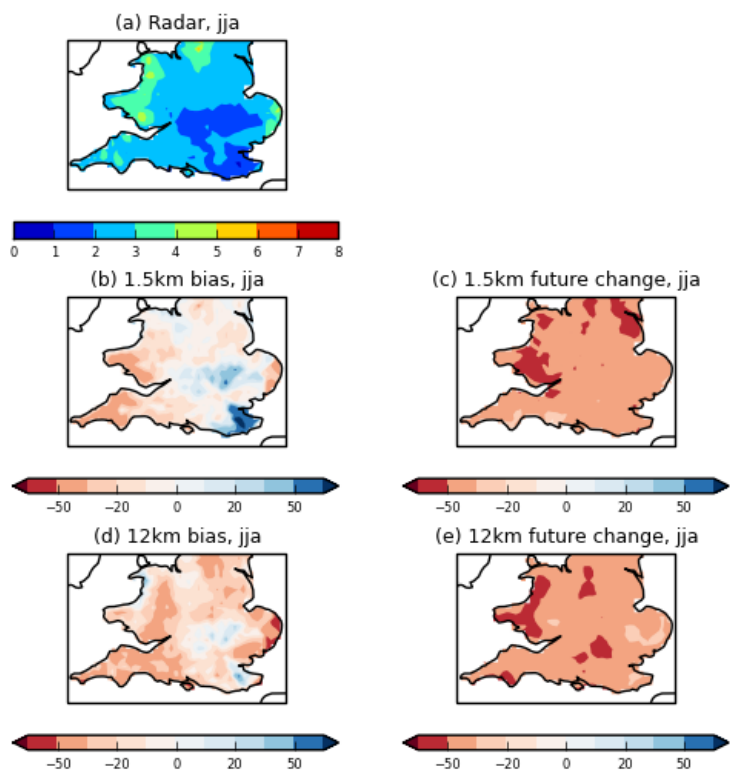


FIG. 2. As Figure 1 but for summer (JJA).

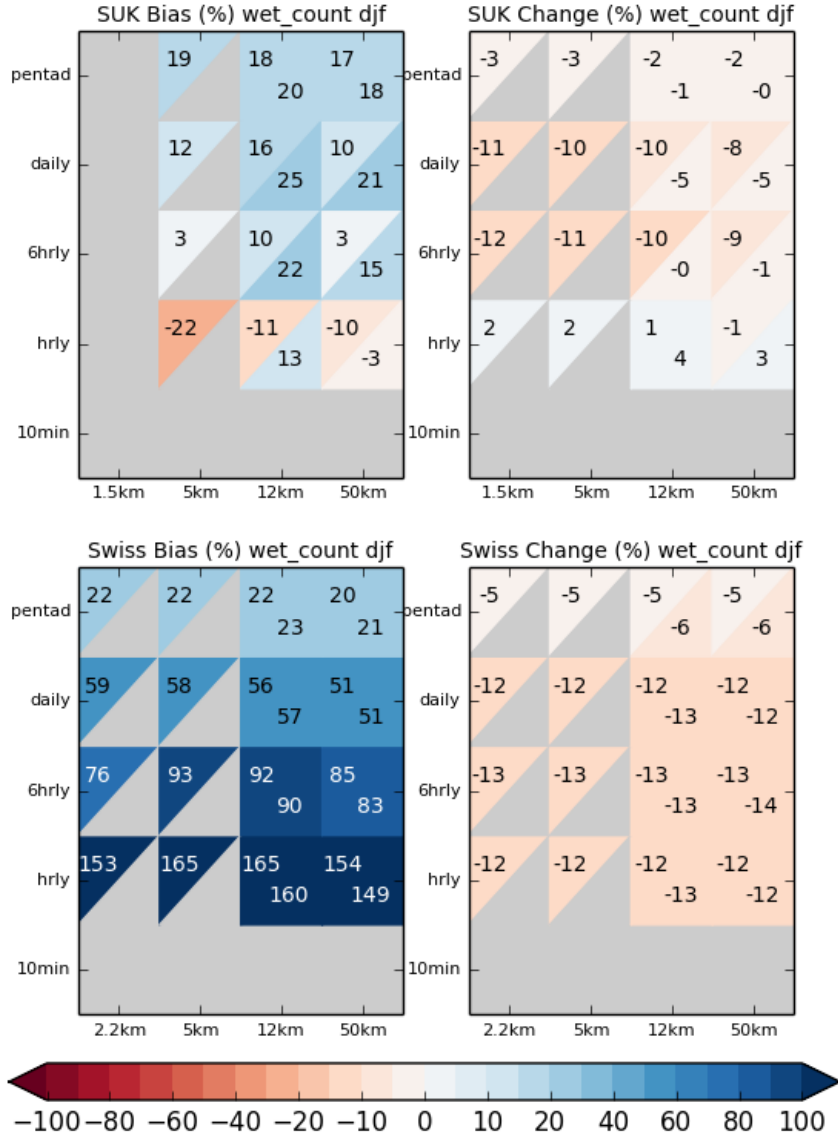


FIG. 3. Biases and changes (%) in rainfall occurrence across space and time scales averaged over the (top) southern UK and (bottom) Switzerland, in winter (DJF). Biases are calculated as model-observation differences, using hourly 5km Nimrod radar data for the UK for 2003-2012 (Golding 1998) and 1km RdisaggH combined radar-gauge observations for Switzerland for 2004-2010 (Wuest et al. 2010). Changes correspond to differences between decadal-length present-day and future (end of century, under RCP8.5) simulations. Results for the convection-permitting (1.5km or 2.2km) RCM are shown in the upper left triangle and for the 12km RCM in the lower right triangle. Grey indicates where results are not available. A threshold of 0.1mm per accumulation period is used to define rainfall occurrence.

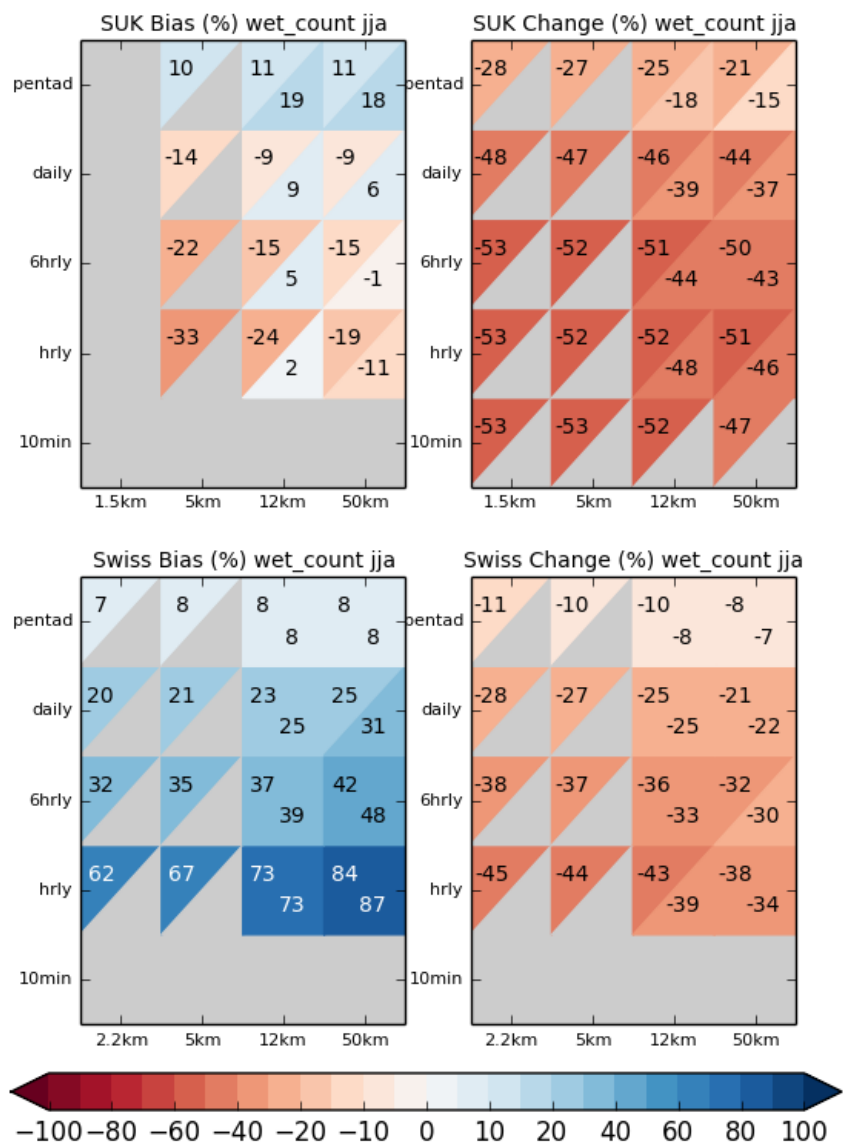


FIG. 4. As Figure 3 but for summer (JJA).

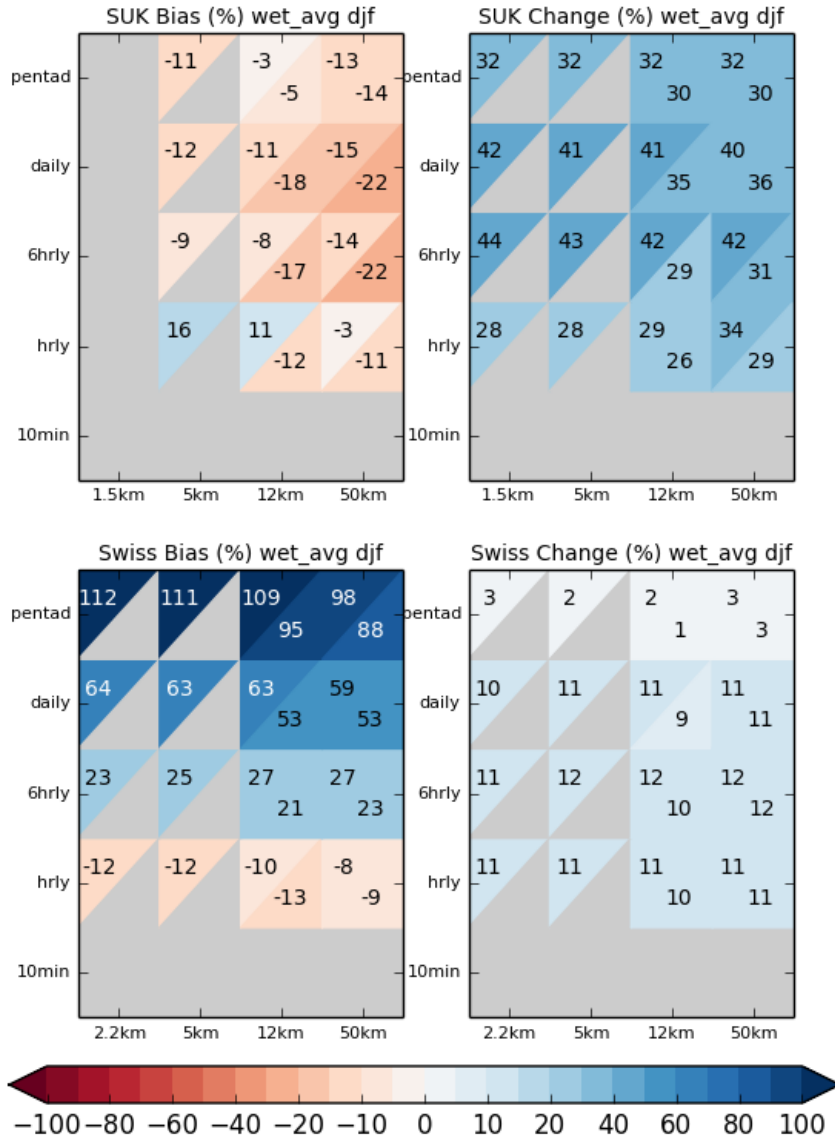


FIG. 5. Biases and changes (%) in rainfall intensity across space and time scales averaged over the (top) southern UK and (bottom) Switzerland, in winter (DJF). Rainfall intensity is defined as the mean of wet values ($>0.1\text{mm}$ per accumulation period). Definitions are as in Figure 3.

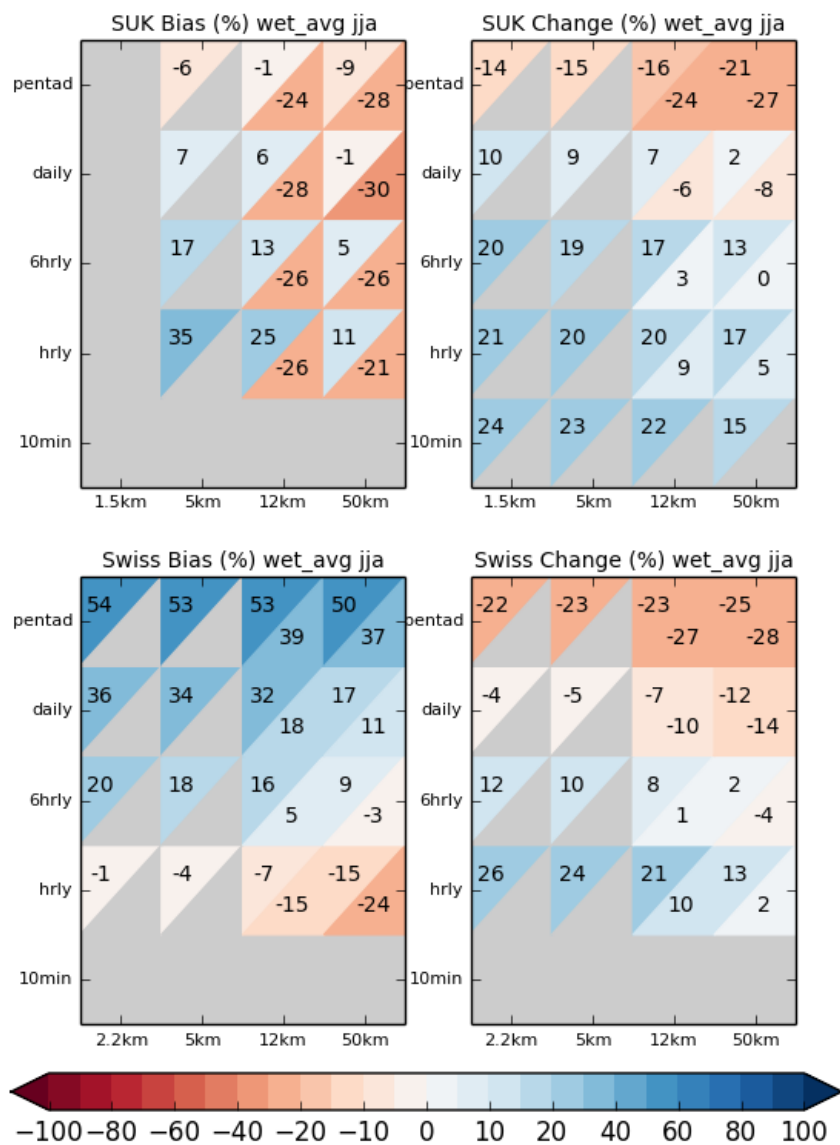


FIG. 6. As Figure 5 but for summer (JJA).

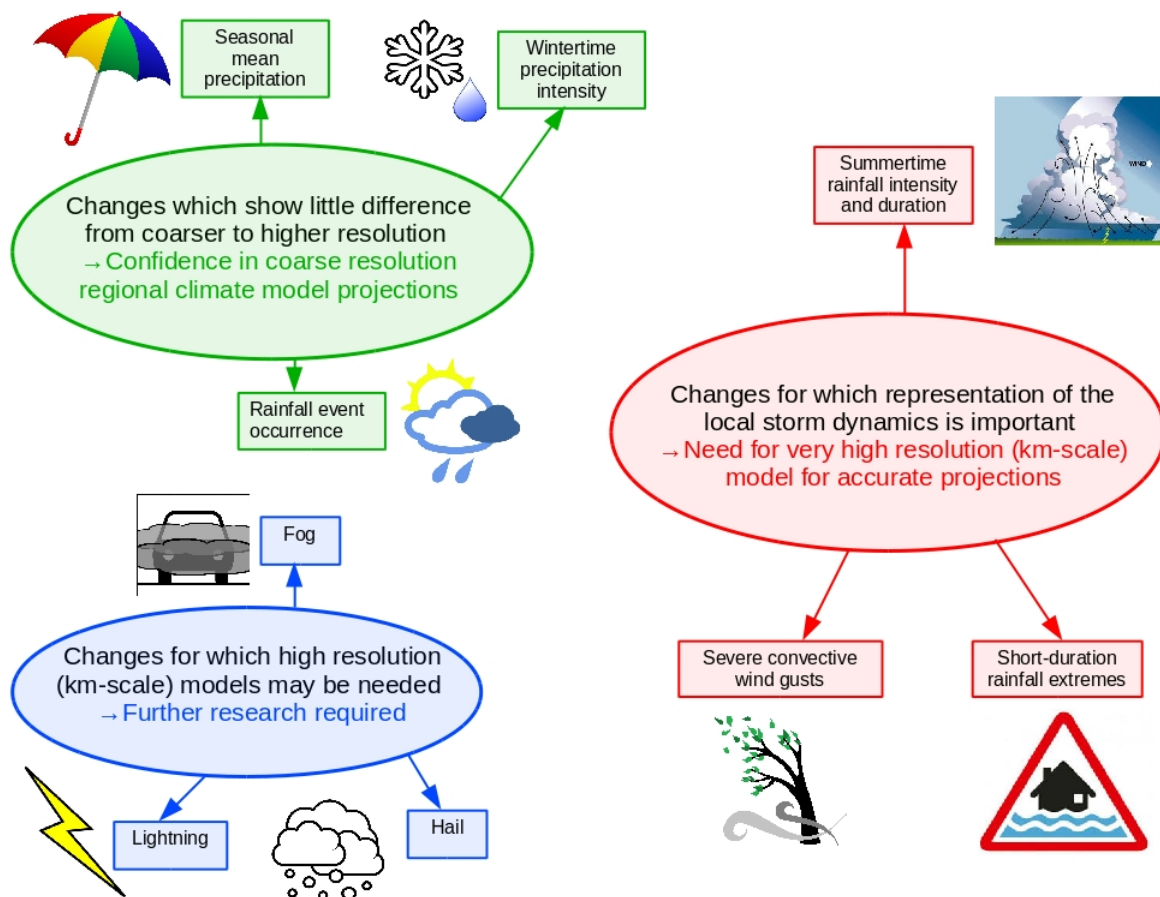


FIG. 7. Schematic summarising where we have confidence in coarse resolution regional climate model projections, and where very high resolution (km-scale) models are needed for accurate projections

Do convection-permitting regional climate models improve projections of future precipitation change?

Supplementary Material

The UK experiments used here are described in Kendon et al. (2014). They are configurations of the Met Office Unified Model (UM). The 1.5km model spans the southern UK and is driven by a 12km RCM which spans Europe and is in turn driven by a 60km GCM. The 12km RCM and 60km GCM both have the UM Global Atmosphere 3.0 (GA3) configuration (Walters et al. 2011). The 1.5km model has similar model physics, except that the convection scheme has been switched off and Smagorinsky-Lilly turbulence diffusion is applied. The simulations examined here are for 13-year present-day (1996-2009) and 13-year future (2100, under RCP8.5 scenario) periods. Model biases for the present-day have been assessed by comparison with 5km gridded hourly observations from radar, available for the UK for 2003-2012 (Golding 1998).

The Alps experiments used here are described in Ban et al. (2015). The 2.2km model spans the greater Alpine region and is driven by a 12km RCM which spans Europe and is in turn driven by a T63 spectral (~ 200 km) resolution GCM. The RCMs are the COSMO-CLM model (Steppeler et al. 2003), whilst the GCM is the Earth System Model of the Max Planck Institute (MPI-ESM-LR) (Stevens et al. 2013). In the 2.2km RCM the deep convection scheme is switched off, whilst the shallow convection scheme remains active. The models are run for 10-year present-day (1991-2000) and 10-year future (2081-2090, under RCP8.5 scenario) periods. Model biases for the present-day have been assessed by comparison with 1km RdisaggH combined radar-gauge hourly observations, available for Switzerland for 2004-2010 (Wuest et al. 2010).

References

- Ban, N., J. Schmidli, and C. Schar, 2015: Heavy precipitation in a changing climate: Does short-term summer precipitation increase faster? *Geophys. Res. Lett.*, **42** (), 1165–1172, doi:10.1002/2014GL062588.
- Golding, B. W., 1998: Nimrod: A system for generating automated very short range forecasts. *Meteorol. Appl.*, **5**, 1–16, doi:10.1017/S1350482798000577.
- Kendon, E. J., N. M. Roberts, H. J. Fowler, M. J. Roberts, S. C. Chan, and C. A. Senior, 2014: Heavier summer downpours with climate change revealed by

- weather forecast resolution model. *Nature Climate Change*, **4**, 570–576, doi:10.1038/nclimate2258.
- Steppeler, J., G. Doms, U. Schattler, H. W. Bitzer, A. Gassmann, U. Damrath, and G. Gregoric, 2003: Meso-gamma scale forecasts using the nonhydrostatic model LM. *Meteorol. Atmos. Phys.*, **82** (), 75–96, doi:10.1007/s00703-001-0592-9.
- Stevens, B., M. Giorgetta, M. Esch, T. Mauritsen, T. Crueger, S. Rast, M. Salzmann, H. Schmidt, J. Bader, K. Block, R. Brokopf, I. Fast, S. Kinne, L. Kornbluh, U. Lohmann, R. Pincus, T. Reichler, and E. Roeckner, 2013: Atmospheric component of the MPI-M Earth System Model: ECHAM6. *J. Adv. Model Earth Syst.*, **5** (2), 146–172, doi:10.1002/jame.20015.
- Walters, D. N., M. J. Best, A. C. Bushell, D. Copsey, J. M. Edwards, P. D. Falloon, C. M. Harris, A. P. Lock, J. C. Manners, C. J. Morcrette, M. J. Roberts, R. A. Stratton, S. Webster, J. M. Wilkinson, M. R. Willett, I. A. Boutle, P. D. Earnshaw, P. G. Hill, C. MacLachlan, G. M. Martin, W. Moufouma-Okia, M. D. Palmer, J. C. Petch, G. G. Rooney, A. A. Scaife, and K. D. Williams, 2011: The Met Office Unified Model global atmosphere 3.0/3.1 and JULES global land 3.0/3.1 configurations. *Geosci. Model Devel.*, **4**, 919–941, doi:10.5194/gmd-4-919-2011.
- Wuest, M., C. Frei, A. Altenhoff, M. Hagen, M. Litschi, and C. Schär, 2010: A gridded hourly precipitation dataset for Switzerland using rain-gauge analysis and radar-based disaggregation. *Int. J. Climatol.*, **30** (), 1764–1775.

List of Figures

1	Biases (%) in rainfall occurrence across space and time scales averaged over Switzerland for RCMs driven by MPI-ESM GCM (top) and ERA-Interim (bottom), in winter (DJF). Biases are calculated as model-observation differences, using 1km RdisaggH combined radar-gauge observations for Switzerland. Results for the convection-permitting (2.2km) RCM are shown in the upper left triangle and for the 12km RCM in the lower right triangle. Grey indicates where results are not available. A threshold of 0.1mm per accumulation period is used to define rainfall occurrence.	4
2	As Supplementary Figure 1 but for summer (JJA).	5
3	Biases (%) in rainfall intensity across space and time scales averaged over Switzerland for RCMs driven by MPI-ESM GCM (top) and ERA-Interim (bottom), in winter (DJF). Rainfall intensity is defined as the mean of wet values (>0.1mm per accumulation period). Definitions are as in Supplementary Figure 1.	6
4	As Supplementary Figure 3 but for summer (JJA).	7
5	Biases (%) in rainfall occurrence across space and time scales averaged over the southern UK for RCMs driven by the Met Office UM-GA3 GCM (top) and ERA-Interim (bottom), in winter (DJF). Biases are calculated as model-observation differences, using hourly 5km Nimrod radar data for the UK. Results for the convection-permitting (1.5km) RCM are shown in the upper left triangle and for the 12km RCM in the lower right triangle. Grey indicates where results are not available. A threshold of 0.1mm per accumulation period is used to define rainfall occurrence. . . .	8
6	As Supplementary Figure 5 but for summer (JJA).	9
7	Biases (%) in rainfall intensity across space and time scales averaged over the southern UK for RCMs driven by the Met Office UM-GA3 GCM (top) and ERA-Interim (bottom), in winter (DJF). Rainfall intensity is defined as the mean of wet values (>0.1mm per accumulation period). Definitions are as in Supplementary Figure 5.	10
8	As Supplementary Figure 7 but for summer (JJA).	11

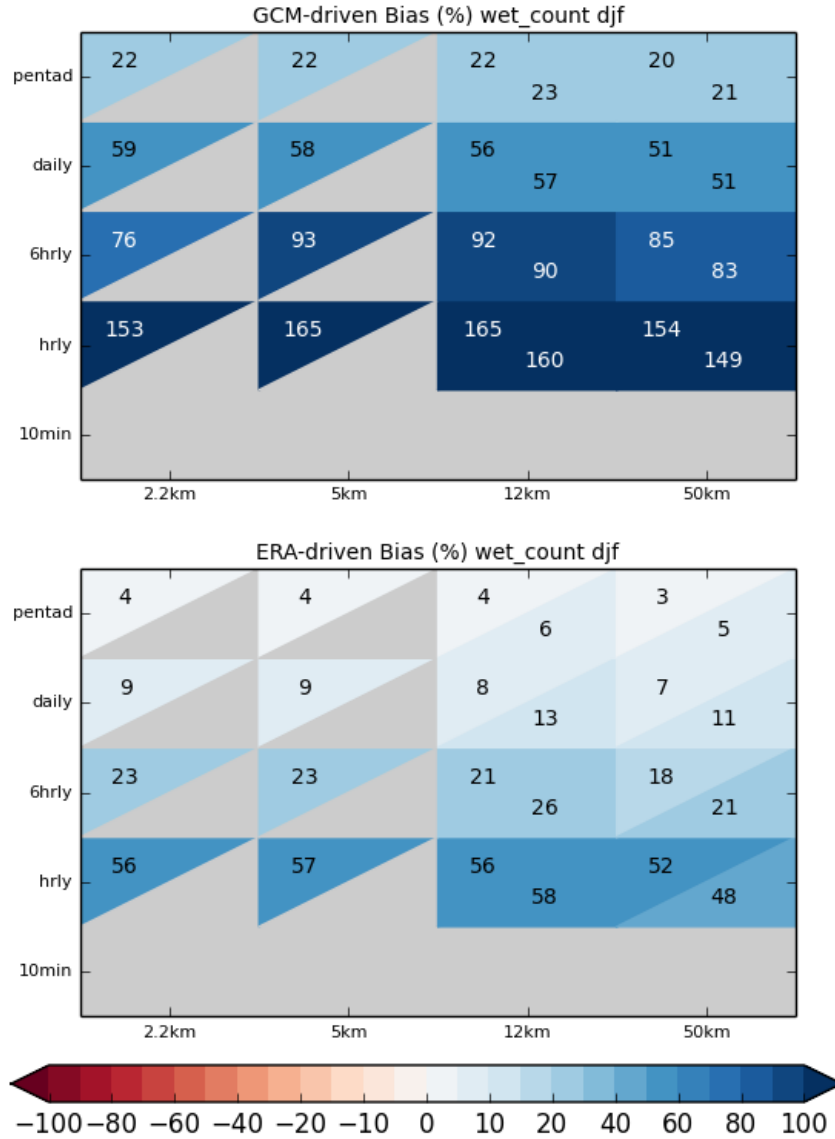


FIG. 1. Biases (%) in rainfall occurrence across space and time scales averaged over Switzerland for RCMs driven by MPI-ESM GCM (top) and ERA-Interim (bottom), in winter (DJF). Biases are calculated as model-observation differences, using 1km RdisaggH combined radar-gauge observations for Switzerland. Results for the convection-permitting (2.2km) RCM are shown in the upper left triangle and for the 12km RCM in the lower right triangle. Grey indicates where results are not available. A threshold of 0.1mm per accumulation period is used to define rainfall occurrence.

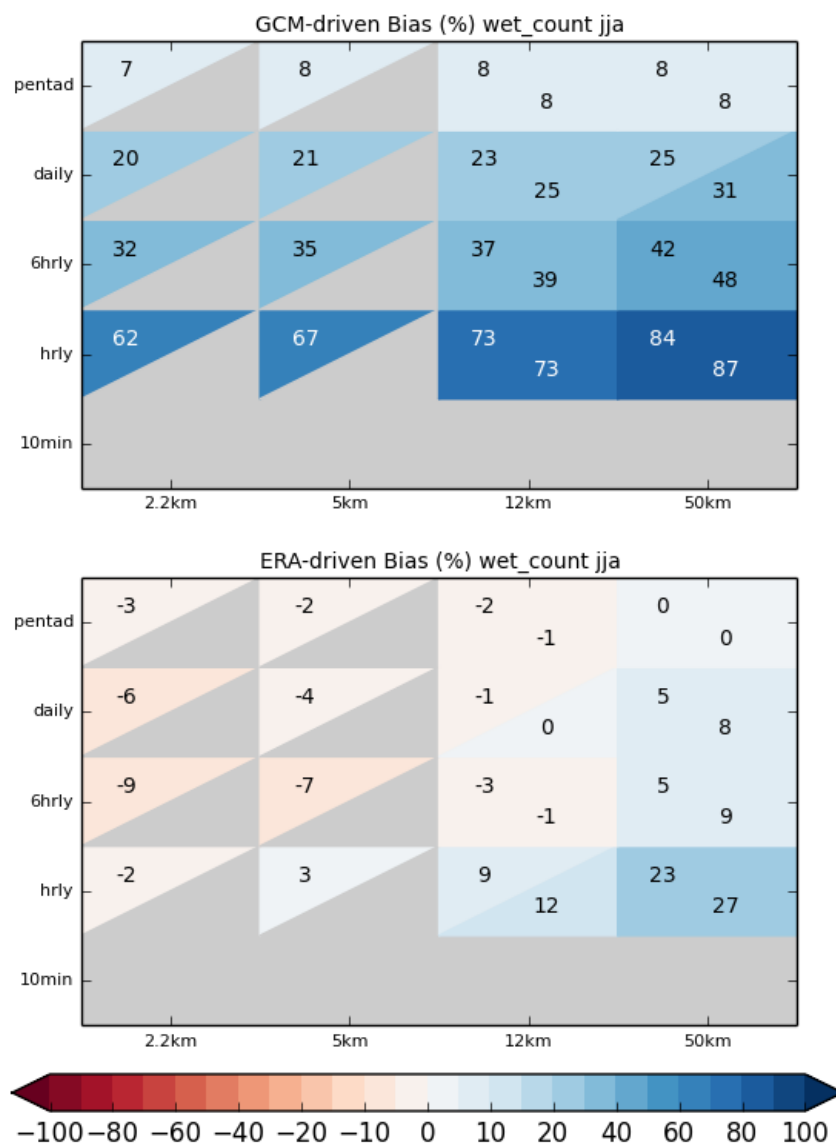


FIG. 2. As Supplementary Figure 1 but for summer (JJA).

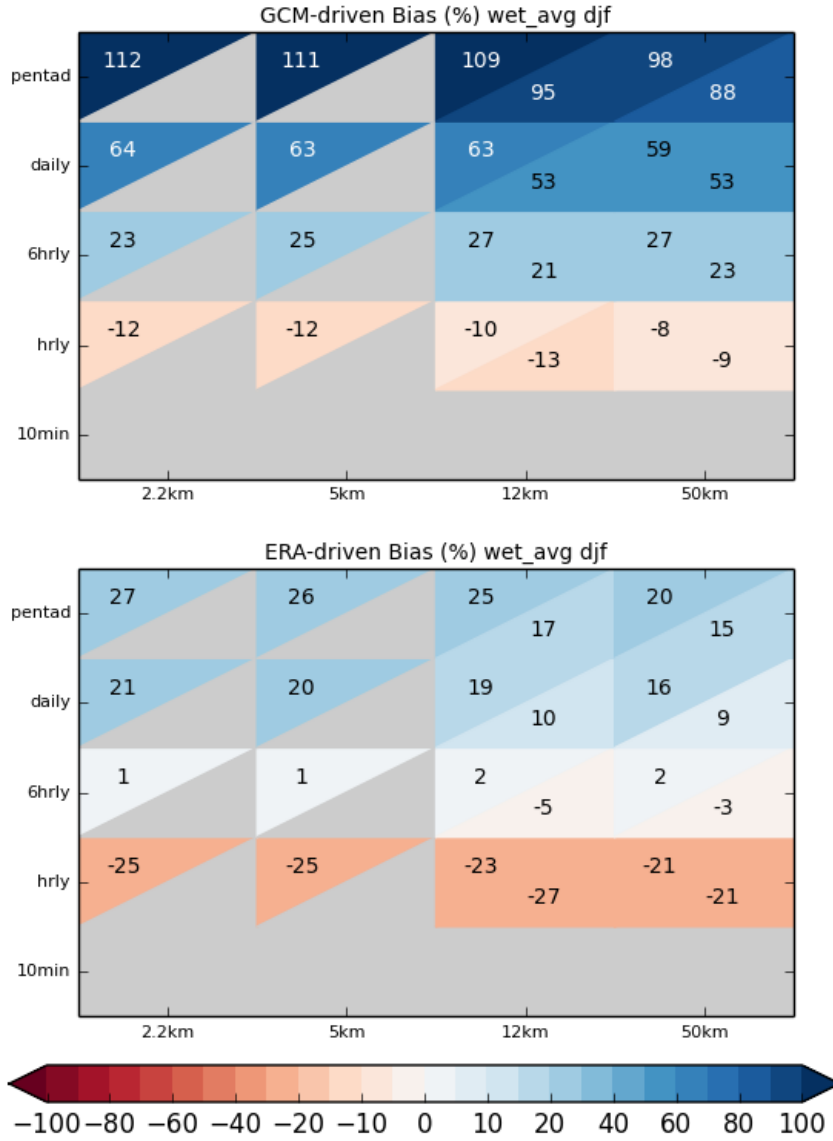


FIG. 3. Biases (%) in rainfall intensity across space and time scales averaged over Switzerland for RCMs driven by MPI-ESM GCM (top) and ERA-Interim (bottom), in winter (DJF). Rainfall intensity is defined as the mean of wet values ($>0.1\text{mm}$ per accumulation period). Definitions are as in Supplementary Figure 1.

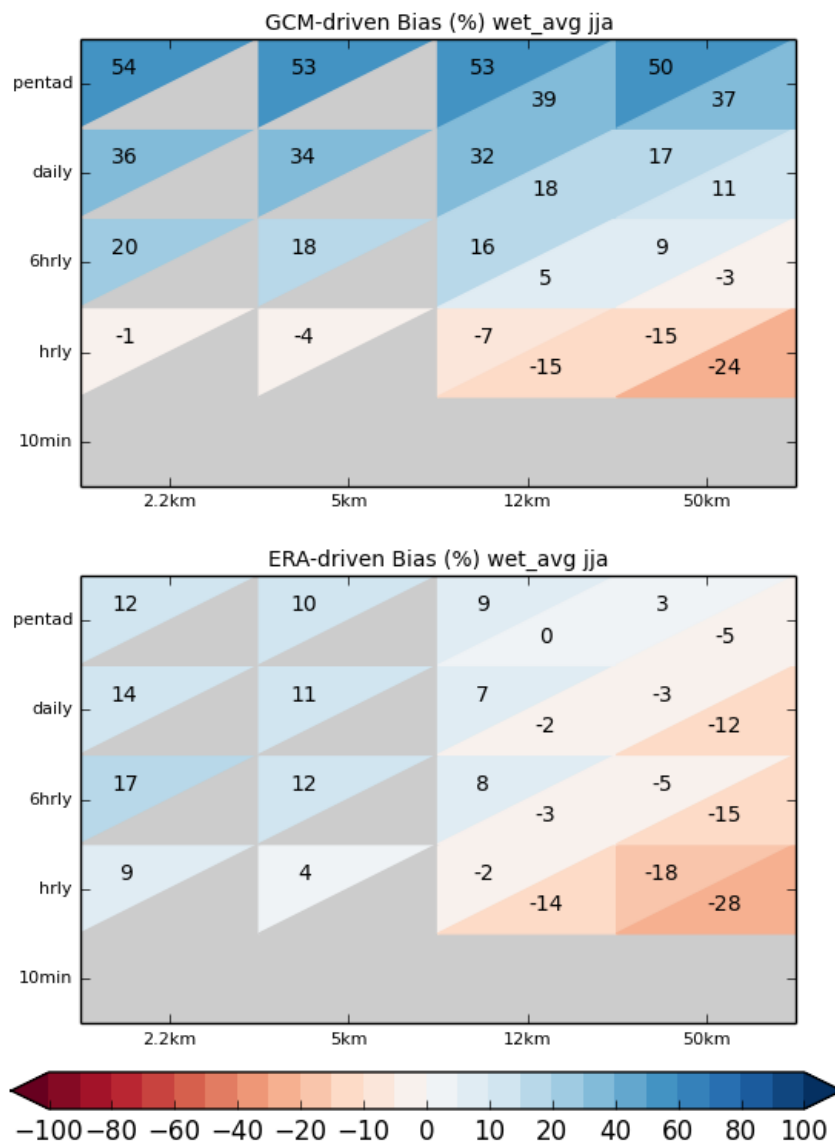


FIG. 4. As Supplementary Figure 3 but for summer (JJA).

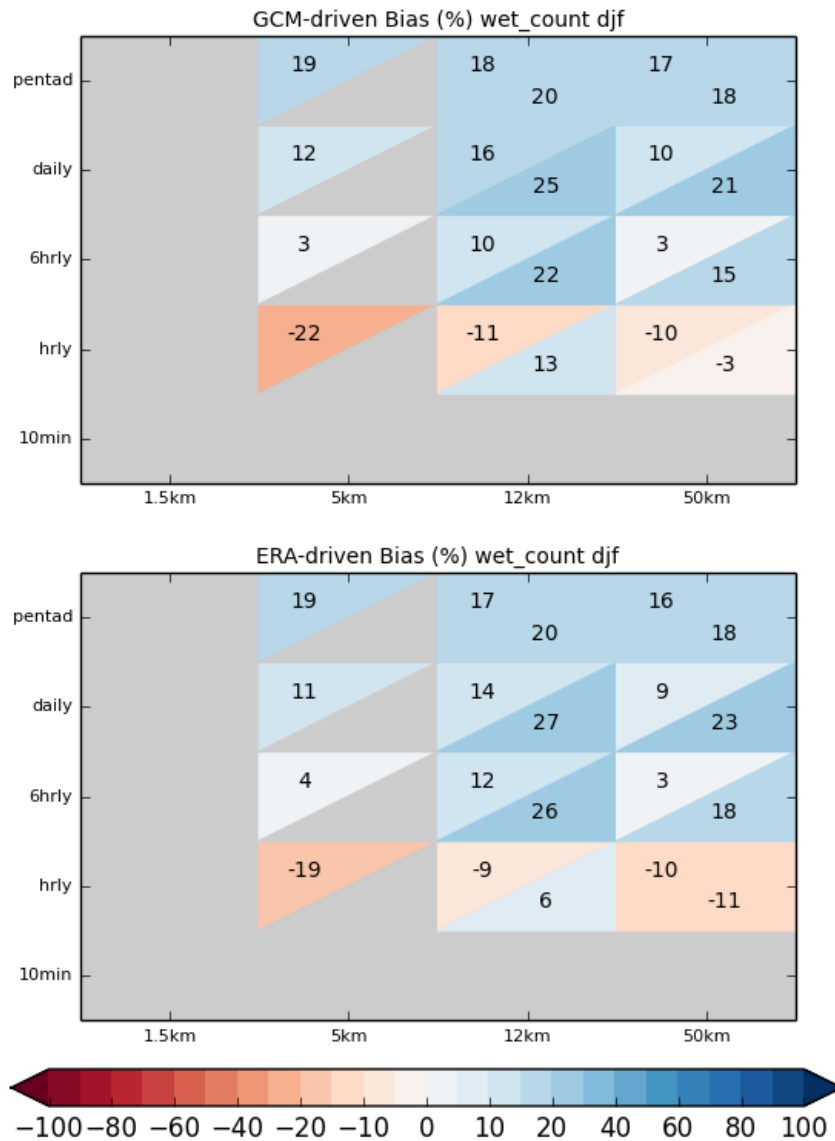


FIG. 5. Biases (%) in rainfall occurrence across space and time scales averaged over the southern UK for RCMs driven by the Met Office UM-GA3 GCM (top) and ERA-Interim (bottom), in winter (DJF). Biases are calculated as model-observation differences, using hourly 5km Nimrod radar data for the UK. Results for the convection-permitting (1.5km) RCM are shown in the upper left triangle and for the 12km RCM in the lower right triangle. Grey indicates where results are not available. A threshold of 0.1mm per accumulation period is used to define rainfall occurrence.

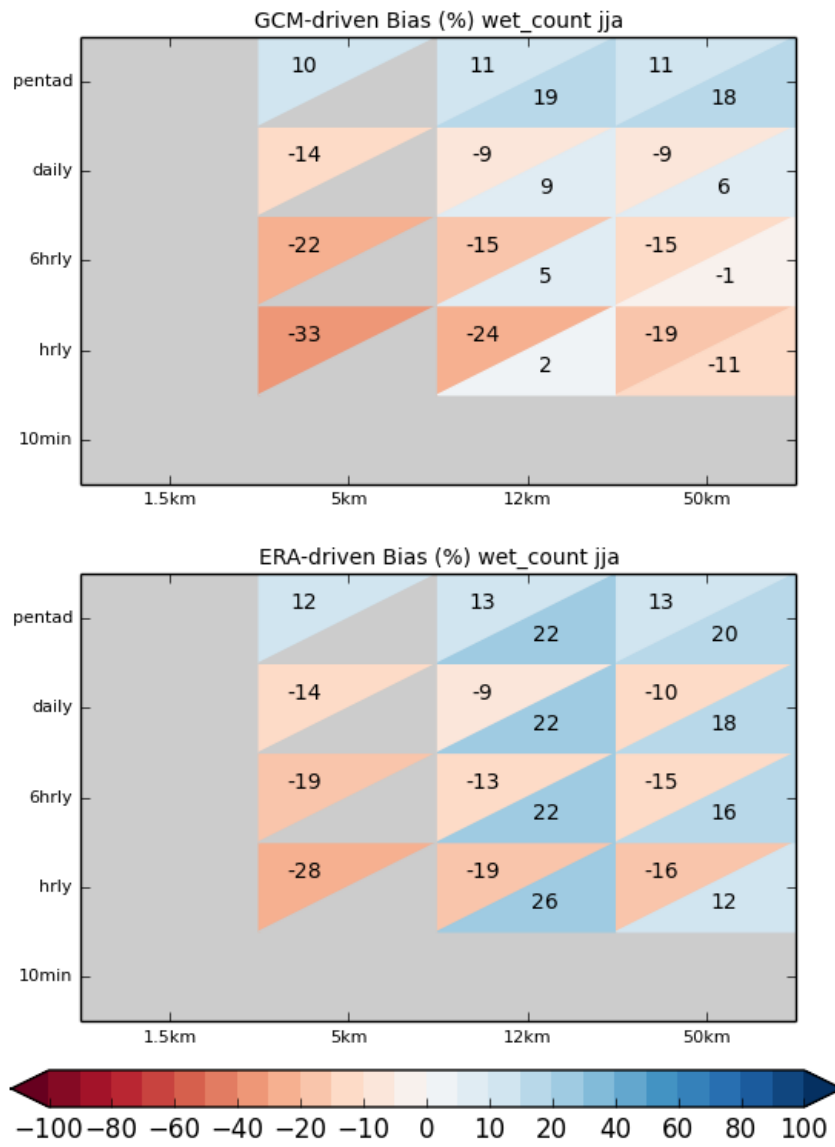


FIG. 6. As Supplementary Figure 5 but for summer (JJA).

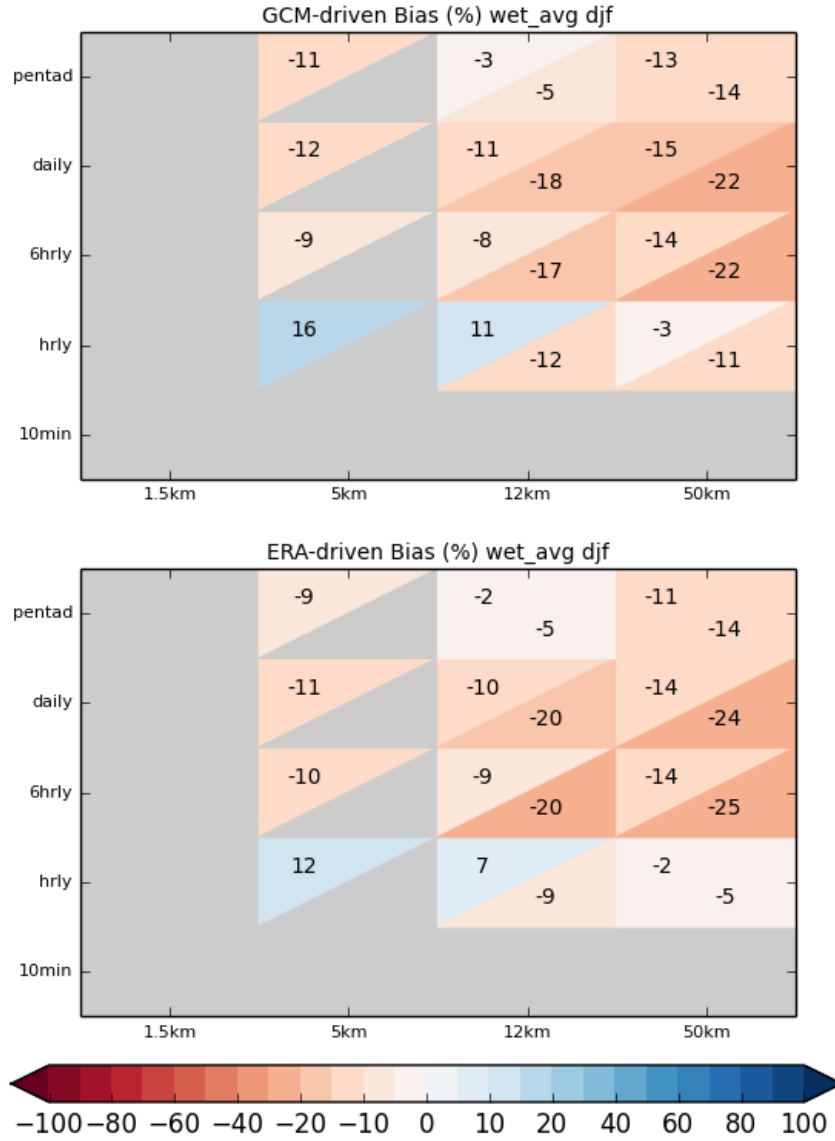


FIG. 7. Biases (%) in rainfall intensity across space and time scales averaged over the southern UK for RCMs driven by the Met Office UM-GA3 GCM (top) and ERA-Interim (bottom), in winter (DJF). Rainfall intensity is defined as the mean of wet values ($>0.1\text{mm}$ per accumulation period). Definitions are as in Supplementary Figure 5.

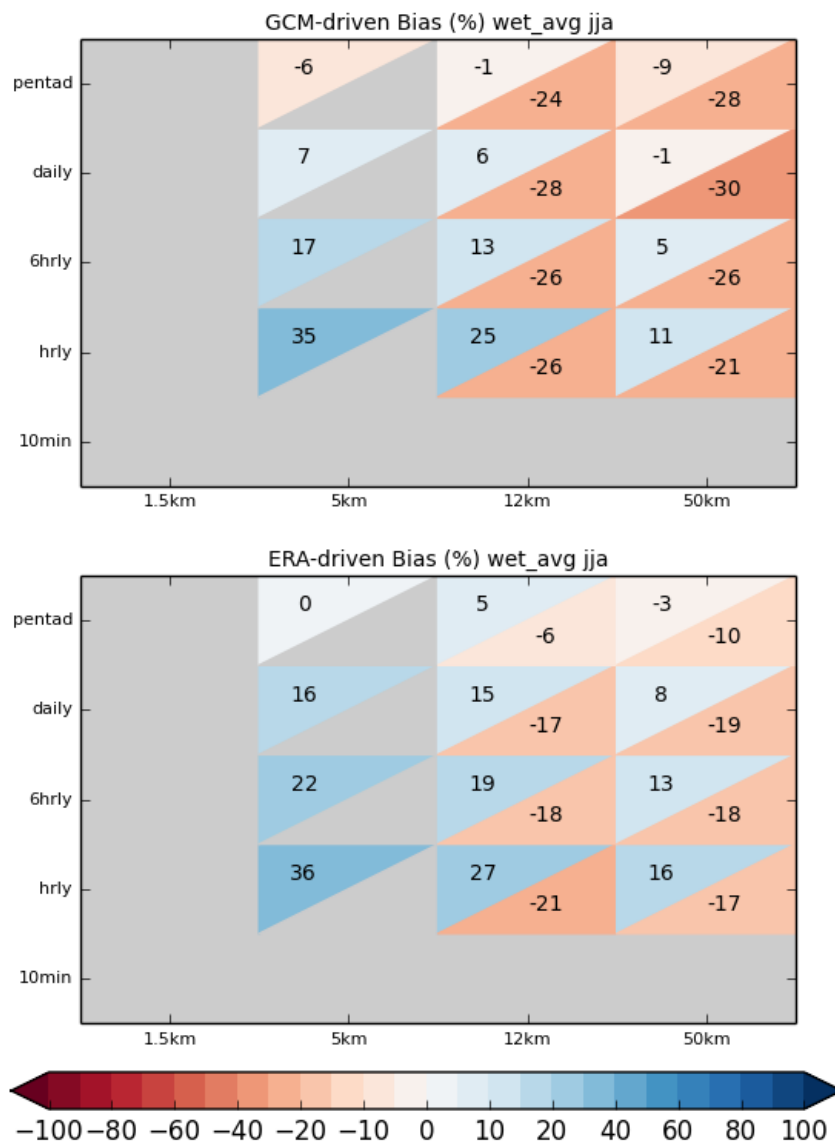


FIG. 8. As Supplementary Figure 7 but for summer (JJA).