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**Extreme precipitation in 150-year continuous simulations by 20-km and 60-km atmospheric general circulation models with dynamical downscaling over Japan by a 20-km regional climate model**

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22 **Abstract**

23 Continuous simulations from the middle of the 20th century to the end of the 21st  
24 century were performed using a 20-km atmospheric general circulation model (AGCM),  
25 and a 60-km AGCM with dynamical downscaling via a 20-km regional climate model  
26 (RCM), to explore the transitional changes in regional extreme events. The representative  
27 scenario simulations by the AGCMs followed the protocol of the High Resolution Model  
28 Intercomparison Project experiments. In addition, ensemble simulations using four  
29 emission scenarios were conducted using the 60-km AGCM with 20-km RCM  
30 downscaling.

31 Regardless of the emission scenario used, the global-mean relative increase in annual  
32 maximum daily precipitation (Rx1d) was roughly proportional to the increase in the  
33 global-mean surface air temperature (SAT), consistent with previous results from coarser-  
34 resolution climate models. It means that the relationship is also valid for a higher-  
35 resolution model. A similar correlation between Rx1d and SAT was seen also in the  
36 values averaged over the Japanese land area in the 20-km AGCM and the 20-km RCM  
37 simulations after applying a 10-year running mean. However, it was not so clear in the  
38 60-km AGCM, mainly due to insufficient grid points over land in Japan in the 60-km  
39 AGCM due to too large noise. It suggests that transitional changes in Rx1d at regional  
40 scales such as the Japanese land area can only be represented by using a model resolution  
41 as high as 20 km, unless using ensemble simulations.

42

## 43 **1. Introduction**

44 Changes in extreme events, caused by global warming, are having a significant effect  
45 on the magnitude and frequency of natural disasters, as well as on agricultural activity  
46 and water resource management, and there is a need for more detailed forecasts of such  
47 changes at the regional scale. To study extreme phenomena such as heavy rainfall, long-  
48 term simulations that cover several decades or more are required at a resolution that is  
49 high enough to accurately represent heavy rainfall. Therefore, we have used our  
50 atmospheric general circulation model (AGCM) with horizontal resolutions of 60 and 20  
51 km, together with dynamical downscaling via our regional climate model (RCM) at  
52 horizontal resolutions of 20, 5, 2, and 1 km. Previous studies have carried out time-slice  
53 simulations of future specific years (e.g., 2075–2099; Kitoh et al. 2016), and future years  
54 at specific warming levels (e.g., 4°C, 2°C, and 1.5°C warming; Mizuta et al. 2017; Fujita  
55 et al. 2019; Nosaka et al. 2020; Ishii and Mori 2020), with prescribed sea surface  
56 temperature (SST) warming derived from the models used in the Coupled Model  
57 Intercomparison Project (CMIP).

58 It has been shown that some recent heavy rainfall events, as well as extreme  
59 temperature events were influenced by global warming (Imada et al. 2020). There is an  
60 increasing need for better information regarding the temporal evolution of these extreme  
61 events from now to the end of this century, and on the timing of the emergence of the  
62 signal associated with any changes above the noise of natural variability (Hawkins et al.  
63 2020). Large changes that fall outside the range of past experience could have a  
64 significant impact on natural disasters, water resources, agriculture, ecosystems, and  
65 human health. We also need information on how such changes depend on the emissions  
66 scenarios. This information is becoming increasingly important, especially for the  
67 development of adaptation policies for global warming.

68 Therefore, in this study, we performed high-resolution continuous simulations from  
69 1950 to 2099 using the same model as our previous time-slice simulations. Continuous

70 simulations with a 60-km resolution AGCM have already been shown to be useful for the  
71 analysis of the temporal evolution of changes in extreme rainfall events such as tropical  
72 cyclones (Sugi and Yoshimura 2012) and precipitation intensity over East Asia (Kusunoki  
73 and Mizuta 2013). Here, we used a 20-km AGCM and 20-km RCM, as well as a 60-km  
74 AGCM to facilitate comparisons among different resolutions. The representative scenario  
75 simulations with the AGCMs are conducted as the High Resolution Model  
76 Intercomparison Project (HighResMIP; Haarsma et al. 2016) experiments. HighResMIP  
77 is one of the CMIP6 (Eyring et al. 2016) experiments to compare the impact of increasing  
78 the horizontal resolution of climate models on climate reproduction among climate  
79 models from various organizations around the world. In addition, ensemble simulations  
80 based on four emissions scenarios were conducted using the 60-km AGCM with 20-km  
81 RCM downscaling.

82 Here, we will document the detailed experimental design and temporal changes in the  
83 surface temperature and precipitation, then show the relationships between the warming  
84 and the extreme precipitation increase in the global and Japanese regions, and their  
85 differences among the experiments.

86

## 87 **2. Models and experimental design**

88 We used the MRI-AGCM3.2 model (Mizuta et al. 2012) with horizontal resolutions of  
89 20 and 60 km for the global simulations (hereafter AGCM20 and AGCM60, respectively),  
90 and the NHRCM model (Sasaki et al. 2008; Murata et al. 2015) with a horizontal  
91 resolution of 20 km for the dynamical downscaling around Japan (hereafter RCM20).  
92 AGCM60 and RCM20 were the same models and settings as those used in the Database  
93 for Policy Decision-Making for Future Climate Change (d4PDF; Mizuta et al. 2017),  
94 except that the RCM20 has been modified to produce sea ice depending on the SST, since  
95 the sea ice over Okhotsk Sea is not represented in the RCM20 of d4PDF. AGCM20 has  
96 the same parameter settings as AGCM60 except for the integration time step (10 minutes

97 for AGCM20 and 20 minutes for AGCM60), which are also the same ones as the model  
98 used in previous studies (e.g., Murakami et al. 2012; Kitoh and Endo 2016).

99 The future scenarios used in this study were the CMIP5 RCP8.5 scenario for AGCM20  
100 and the RCP8.5, RCP6.0, RCP4.5, and RCP2.6 scenarios for AGCM60 with the RCM20  
101 downscaling. The RCP8.5 is the highest emission scenario, where greenhouse gas  
102 emissions continue to grow unmitigated, and the RCP2.6 is a mitigation scenario aiming  
103 to limit the increase of global mean temperature to 2°C. Each simulation was conducted  
104 between 1950 and 2099. Simulations of the historical climate up to the present time  
105 (1950–2014) were performed based on the observational boundary conditions with  
106 different initial values and connected to the respective scenario experiments (2015–2099).  
107 The multiple historical simulations are used to evaluate the internal variability of the  
108 atmosphere as the ensemble spread.

109 For the boundary conditions settings, we followed the HighResMIP (Haarsma et al.  
110 2016) protocol. The settings are listed in the right column of Table S1. For the historical  
111 simulation, we used the 0.25° daily gridded SST and sea ice concentrations from  
112 HadISST2.2 dataset (Kennedy et al. 2017), the same SST as in the *highresSST-present*  
113 experiment of the HighResMIP. As specified in the HighResMIP, the same settings as in  
114 the CMIP6 *historical* experiments (O'Neill et al. 2016) are used for ozone, volcanic  
115 aerosols, greenhouse gases, and solar activity. For non-volcanic aerosols, we used the  
116 monthly mean three-dimensional output from the *historical* experiments by MRI-ESM2  
117 (Yukimoto et al. 2019), instead of implementing the recommended HighResMIP protocol  
118 into MRI-AGCM.

119 For the future climate simulations, the CMIP5 model-averaged temperature increase is  
120 added to the observed SST and sea ice concentrations (see below). For ozone, volcanic  
121 aerosols, greenhouse gases, and solar activity, the protocols of the CMIP6 *ssp585*, *ssp460*,  
122 *ssp245*, and *ssp126* experiments are used, and the monthly mean outputs from the *ssp585*,  
123 *ssp460*, *ssp245*, and *ssp126* experiments by MRI-ESM2 are used for the non-volcanic

124 aerosols.

125 For the RCP8.5 scenario experiments, we used the same SST as in the *highresSST-*  
126 *future* experiment of the HighResMIP, in which the SST increase averaged over eight  
127 CMIP5 models (ACCESS1-0, ACCESS1-3, GFDL-CM3, IPSL-CM5A-LR, IPSL-  
128 CM5A-MR, MPI-ESM-MR, CNRM-CM5, and HadGEM2-ES), that were selected based  
129 on their representation of Arctic sea ice variability, are used. As this is a continuous  
130 simulation, consideration was given to avoid major discontinuities around 2015. The SST  
131 for 2015 was based on the observed HadISST data. After 2016, monthly deviations from  
132 the 2005–2025 average were calculated for each model. Nine-year running means of the  
133 monthly deviations were averaged across all models and added to the 2007–2015 average  
134 of HadISST. This defines the SST without interannual variability after 2016 (thick black  
135 dashed line behind red lines in Fig. 1a). Next, as a component of interannual variability,  
136 daily deviations from the 9-year running mean of HadISST were calculated for 1980–  
137 2015, and the time series were added to it repeatedly over the periods 2016–2051, 2052–  
138 2087, and 2088–2099 (red lines in Fig. 1a). Note that, for smoother transitions, the values  
139 on January of 2016, 2052, and 2088 are linearly interpolated with those at the end of the  
140 previous year. For the sea ice concentration, equations expressing sea ice concentration  
141 as a function of SST are constructed for HadISST and CMIP5 models, respectively. The  
142 sea ice concentration in the future simulations is estimated by inputting the generated  
143 future SST into it. Linear interpolation of the two for 2016–2030 is used, and the latter  
144 equation is used for 2031–2099. More details can be found in the document referenced in  
145 <https://github.com/PRIMAVERA-H2020/HighResMIP-futureSSTSeaice>.

146 For the other RCP scenarios, we used the same setup as for the *highresSST-future*  
147 experiment, but with the SST changes from the RCP6.0, RCP4.5, and RCP2.6 scenarios.  
148 However, because not all of the eight models used for the RCP8.5 SST provide the other  
149 scenario experiment results, the same 28 CMIP5 models (21 models for RCP6.0) as  
150 Mizuta et al. (2014) are used for the other scenario SST future changes. Figure 1(a) shows

151 the time series of the SSTs for the four RCP scenarios. As RCP6.0 is radiatively forced  
152 below RCP4.5 until around 2070, the SSTs are also below those of RCP4.5. Figure S1  
153 shows the SST warming patterns from the end of the 20th century to the end of the 21th  
154 century. Although the large-scale distributions for the RCP8.5 scenario (Fig. S1a) are  
155 similar to those from the previous 20-km AGCM time-slice simulations (Kitoh and Endo  
156 2016), the detailed patterns differ because of the different methods and number of CMIP5  
157 models used.

158 The differences in the boundary conditions compared with our previous time-slice  
159 simulations and d4PDF are shown in Table S1. We confirmed that, even with these  
160 differences, the simulations covering the historical period generated a similar climate  
161 representation performance. Figure S2 shows the climatological seasonal precipitation  
162 from 1979 to 2003, and demonstrates that the AGCM20 and AGCM60 results from this  
163 study (Fig. S2c, d, g, h) have a similar precipitation distribution to the previous 20-km  
164 AGCM time-slice (Fig. S2a, b) and d4PDF (Fig. S2e, f) results.

165 To evaluate the model representation around Japan, we used observational data of  
166 Japan Meteorological Agency. We used data at all stations in Japan that have long-term  
167 observation from 1950 to 2020 (134 stations) to calculate the annual maximum daily  
168 precipitation (Rx1d), and selected non-urban 15 stations (Abashiri, Nemuro, Suttsu,  
169 Yamagata, Ishinomaki, Fushiki, Iida, Choshi, Sakai, Hamada, Hikone, Tadotsu, Miyazaki,  
170 Naze, and Ishigakijima), the same as those in Japan Meteorological Agency (2021), for  
171 the surface air temperature (SAT) to avoid including the effects of urbanization.

172

### 173 **3. Changes in surface air temperature and precipitation**

174 The time series of global-mean SAT in the simulations are shown in Fig. 1b. The SAT  
175 increases almost in proportion to the increase in the given SST (Fig. 1a). The warming  
176 above the level of 1950–1979 at 2070–2099 is 1.3°C, 2.0°C, 2.4°C, and 3.9°C for the  
177 RCP2.6, RCP4.5, RCP6.0, and RCP8.5 scenario, respectively. These temperature

178 increases are close to the changes in the multi-model averages of the CMIP5 experiments  
179 (Collins et al. 2013). Figure 1c shows the time series of the 10-year running mean of SAT  
180 averaged over the Japanese land grids from both the global and regional models,  
181 compared with the average of the observed data. Although the same SST was prescribed  
182 for all simulations until 2014 (Fig. 1a), these narrow regional averages show a spread  
183 among the members even in the 10-year running mean. Therefore, we can assume that  
184 the SAT in the future simulations also contains such uncertainty. This spread is about  
185 0.5°C both in the AGCM60 simulations and in the RCM20 simulations. The average SAT  
186 for the period 2011–2020 in the observations is 0.97°C higher than the average for 1950–  
187 1979. The warming over Japan at 2070–2099 in RCM20 is 1.3°C, 2.6°C, 3.1°C, and 5.1°C  
188 for the RCP2.6, RCP4.5, RCP6.0, and RCP8.5 scenario, respectively. It is about 30%  
189 larger than the global-mean warming, except for the RCP2.6 scenario.

190 Figure 2 compares the changes in seasonal mean precipitation and Rx1d for the RCP8.5  
191 simulations of AGCM20, and for the RCP8.5 and RCP2.6 simulations of AGCM60. In  
192 the same RCP8.5 simulations, there are no noticeable differences between AGCM20 and  
193 AGCM60 in terms of mean precipitation and annual maximum daily precipitation (Fig.  
194 2a–f). The general patterns of change in the AGCMs are consistent with the results from  
195 our previous AGCM time-slice simulations (Kitoh and Endo 2016) and d4PDF (Mizuta  
196 et al. 2017), although there is a difference near the tropics due to the slight difference in  
197 the patterns of SST change (Fig. S3). Comparing the RCP8.5 and RCP2.6 simulations of  
198 AGCM60 (Fig. 2d–i), while the change is proportional to the amount of temperature  
199 increase over a large part of the world, there are some places where signs of this change  
200 differ depending on the scenario, especially in the summer western Pacific region  
201 including Japan (Fig. 2e, h). Such differences are more significant for changes in annual  
202 maximum daily precipitation (Fig. 2f, i).

203 Those changes in the RCP8.5 for the area around Japan are shown in Fig. 3, which  
204 compares the results from RCM20 in addition to those from the AGCM simulations. The

205 differences between the resolutions and the models are small in winter (Fig. 3a, d, g), with  
206 a decrease over the southern coast of Japan and an increase over northern Japan. These  
207 changes correspond to the northward shift of the storm track (Kawase et al. 2021). There  
208 is a difference on the Sea of Japan side of eastern Japan, partly due to the differences in  
209 the topography between the model resolutions and in the representation of topographic  
210 precipitation between the AGCM and RCM. While the water vapor increases due to the  
211 warming contribute to the precipitation increase, weakened northwesterly wind contribute  
212 to the precipitation decrease. Such a cancellation makes large uncertainty of the  
213 precipitation change over this area. In summer, difference between the simulations is large  
214 over western Japan. An increasing trend over northern Japan and a decreasing over the  
215 southern coast of eastern Japan are common to the three simulations. It is also common  
216 to the simulations using MRI-AGCM (Fig.S3b,d), but is not consistent with the majority  
217 of CMIP models, which show slight increase all over Japan (Ose 2019). There is a  
218 decrease in the summer mean precipitation and Rx1d only for AGCM20 over the ocean  
219 south of Japan (Fig. 3c). As the value of Rx1d in this region is controlled by typhoon  
220 activity (Kitoh and Endo 2019), differences in typhoon representation related to model  
221 resolution and model physics, as well as uncertainty due to insufficient sample size, could  
222 be the cause of the variations seen in the spatial patterns of seasonal and extreme  
223 precipitation around Japan.

224

#### 225 **4. Relationships of changes in extreme precipitation and temperature**

226 Figure 4a shows the time series of the global mean of Rx1d. The percentage increase  
227 relative to the 30-year average for the period 1950–1979 is shown. The thin and thick  
228 lines show the value for each year and the 10-year running mean, respectively. The pattern  
229 of interannual variability is similar among the four simulations. For instance, there are  
230 positive peaks in 1998, 2034, and 2070 for all simulations, which correspond to the  
231 interannual variability of the prescribed SST. Some decadal variations remain even in the

232 10-year running mean. For example, there is a large increase from 1980 to 2000, but a  
233 small increase from 2000 to 2020. Such variations can be seen even in the latter half of  
234 the 21st century, except for RCP8.5, while the large increasing trend in the RCP8.5  
235 simulation dominates the variation. In the RCP8.5 simulations, the increase is slightly  
236 smaller in AGCM60 than in AGCM20. This is consistent with the resolution dependence  
237 shown in Mizuta and Endo (2020), in which, using the same 60km model, the relative  
238 change becomes slightly smaller when Rx1d is calculated after daily precipitation is  
239 regrided to 1.25° or 2.5° grid.

240 Figure 4b–c shows the time series of the 10-year running means of Rx1d averaged over  
241 land in Japan. The four AGCM60 (Fig. 4b) and four RCM20 (Fig. 4c) results are  
242 compared with the AGCM20 result and the average of the observed data. The  
243 observations show no obvious trend until around 1980, but there is an obvious increase  
244 after 1995. The increase in Rx1d averaged over 1991–2020 from 1950–1979 is 7.2%.  
245 While every single member of AGCM20 and RCM20 simulates the increasing trend in  
246 the historical simulation, AGCM60 does not show the increasing trend. Even for the 10-  
247 year running means, the time series are noisy, especially for AGCM60 (Fig. 4b), showing  
248 large variability on a decadal scale. The difference between the scenarios is not clear even  
249 around 2050. The change in the RCP8.5 scenario does not exceed that of all other  
250 scenarios until after 2070.

251 Next, we plotted the relationship between the relative increase in Rx1d and the SAT  
252 increase (Fig. 5). For the global average, there is a good correlation between the two  
253 values for each year (Fig. 5a), and a much stronger linear correlation for the 10-year  
254 running mean (Fig. 5b). The scenario dependence is very small, consistent with the results  
255 from CMIP multi-model ensemble mean (Li et al. 2021), with a slope of ~7.5% per degree,  
256 which is close to the Clausius–Clapeyron rate of change. It means that the relationship  
257 found in the CMIP climate models is valid also for a higher resolution model. Note that,  
258 however, this rate of increase varies greatly depending on the timescale of the

259 precipitation and the return period (Mizuta and Endo 2020).

260 Figure 5c–d shows the same relationship over land in Japan. The results for the 10-year  
261 running mean are shown here. AGCM20 and RCM20 (Fig. 5d) show a correlation similar  
262 to Fig. 5a. The scenario dependence is small, with a slope of 6% per degree. For AGCM60,  
263 on the other hand, the correlation is lower and the slope is smaller, ~4% per degree. It is  
264 associated with the smaller increase over northern Japan (Fig. 3f) than those in AGCM20  
265 (Fig. 3c) and RCM20 (Fig. 3i). The smaller slope is partly due to the slight resolution  
266 dependence as seen in the global mean (Fig. 4a). The low correlation is mainly due to the  
267 result of insufficient grid points over land in Japan in AGCM60 (123 points). Figure S4  
268 shows the relationship over the region of 128–147°E, 30–47°N, which includes 10× larger  
269 area than the land in Japan. The relationship in AGCM60 becomes similar to AGCM20  
270 and RCM20, suggesting that it is difficult to evaluate the Rx1d change over land in Japan  
271 by a single member of 60km resolution simulation due to too large noise.

272

## 273 **5. Summary and concluding remarks**

274 We performed 150-year continuous simulations using AGCM20, and AGCM60 with  
275 RCM20 downscaling. The global-mean relative increase in Rx1d was roughly  
276 proportional to the increase in the global-mean SAT, especially when viewed as a 10-year  
277 running mean, regardless of the emission scenario used. Such a proportional relationship  
278 is consistent with previous CMIP multi-model results (Li et al. 2021). Our study shows  
279 that the scaling law is valid also for models with a higher resolution than the CMIP climate  
280 models. A similar correlation between Rx1d and SAT was seen also in the values averaged  
281 over the Japanese land area in AGCM20 and RCM20 after applying a 10-year running  
282 mean. Although such a proportional relationship at the regional scale has been suggested  
283 by a comparison of +2K and +4K time-slice simulations (Fujita et al. 2019), it becomes  
284 clearer by our high-resolution continuous experiment in this study.

285 These simulations allow us to analyze transitional changes caused by global warming

286 in phenomena that require high-resolution simulations, such as extreme precipitation  
287 events. More detailed studies using these simulations are now underway that consider: 1)  
288 phenomena that we do not know whether the scaling law with respect to temperature is  
289 valid or not (e.g., snow cover, droughts and coastal conservation); 2) areas where the  
290 multi-year history of temperature and precipitation is important (e.g., water resources and  
291 agriculture); 3) analysis of the timing of the emergence of the change signal from the  
292 natural variability; and 4) the scheduling of adaptation policy making.

293 The proportional relationship between the rate of increase in Rx1d and the SAT  
294 increase over the Japanese land area can be seen in AGCM20 and RCM20, but it was not  
295 so clear in AGCM60. This is mainly due to the small number of sample grids. When the  
296 target spatial and temporal scales become smaller, the effect of internal variability  
297 becomes critically larger, which is consistent with previous studies (Hawkins and Sutton  
298 2009). An initial ensemble experiment will be required if we wish to investigate regional-  
299 scale extreme precipitation changes at the decadal timescale over the Japan region. In  
300 addition, the results presented here are from a single model. Intercomparisons with other  
301 HighResMIP simulations, already analyzed for tropical cyclones (Roberts et al. 2020;  
302 Yamada et al. 2021), would assist our evaluation of the uncertainty associated with the  
303 climate models.

304

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313

## 314 **Data Availability Statement**

315 Data of the RCP8.5 experiments with AGCM20 and AGCM60 are publicly available  
316 as the CMIP6 HighResMIP through the Earth System Grid Federation ([https://esgf-](https://esgf-node.llnl.gov/projects/cmip6/)  
317 [node.llnl.gov/projects/cmip6/](https://esgf-node.llnl.gov/projects/cmip6/)). AGCM60 and RCM20 data will be available from the  
318 Data Integration and Analysis System (DIAS) website (<https://diasjp.net/>).

319

## 320 **Supplements**

321 Supplement 1: One table and four supplemental figures are included.

322

## 323 **References**

324 Collins, M., et al. (2013), Long-term climate change: Projections, commitments and  
325 irreversibility, in Climate Change 2013: The Physical Science Basis. Contribution of  
326 Working Group I to the Fifth Assessment Report of the Intergovernmental Panel on  
327 Climate Change, edited by T. F. Stocker et al. Cambridge Univ. Press, Cambridge, U.  
328 K., and New York.

329 Eyring, V., S. Bony, G. A. Meehl, C. Senior, B. Stevens, R. J. Stouffer, and K. E. Taylor,  
330 2016: Overview of the Coupled Model Intercomparison Project Phase 6 (CMIP6)  
331 experimental design and organization. *Geosci. Model Dev.*, 9, 1937–1958.

332 Fujita, M., R. Mizuta, M. Ishii, H. Endo, T. Sato, Y. Okada, S. Kawazoe, S. Sugimoto, K.  
333 Ishihara, and S. Watanabe, 2019: Precipitation changes in a climate with 2-K surface  
334 warming from large ensemble simulations using 60-km global and 20-km regional  
335 atmospheric models. *Geophys. Res. Lett.*, 46, 435-442, doi:10.1029/2018GL079885.

336 Haarsma, R. J., M. Roberts, P. L. Vidale, C. A. Senior, A. Bellucci, Q. Bao, P. Chang, S.  
337 Corti, N. S. Fučkar, V. Guemas, J. von Hardenberg, W. Hazeleger, C. Kodama, T.  
338 Koenigk, L.-Y. R. Leung, J. Lu, J.-J. Luo, J. Mao, M. S. Mizieliński, R. Mizuta, P.  
339 Nobre, M. Satoh, E. Scoccimarro, T. Semmler, J. Small, and J.-S. von Storch, 2016:

340 High Resolution Model Intercomparison Project (HighResMIP v1.0) for CMIP6.  
341 *Geosci. Model Dev.*, 9, 4185–4208, doi:10.5194/gmd-9-4185-2016.

342 Hawkins, E., and R. Sutton, 2009: The potential to narrow uncertainty in regional climate  
343 predictions. *Bull. Amer. Meteor. Soc.*, 90, 1095–1107. doi:10.1175/2009BAMS2607.1

344 Hawkins, E., Frame, D., Harrington, L., Joshi, M., King, A., Rojas, M., and Sutton, R.  
345 (2020). Observed emergence of the climate change signal: From the familiar to the  
346 unknown. *Geophys. Res. Lett.*, 47, e2019GL086259. doi:10.1029/2019GL086259

347 Imada, Y., H. Kawase, M. Watanabe, M. Arai, H. Shiogama, and I. Takayabu, 2020:  
348 Advanced risk-based event attribution for heavy regional rainfall events. *npj Clim*  
349 *Atmos Sci*, 3, 37 (2020). doi:10.1038/s41612-020-00141-y

350 Ishii, M., and N. Mori, 2020: d4PDF: large-ensemble and high-resolution climate  
351 simulations for global warming risk assessment. *Prog. Earth Planet. Sci.*, 7,  
352 doi:10.1186/s40645-020-00367-7.

353 Japan Meteorological Agency, 2021: Climate change monitoring report 2020.  
354 <https://www.jma.go.jp/jma/en/NMHS/ccmr/ccmr2020.pdf>.

355 Kawase, H., A. Murata, K. Yamada, T. Nakaegawa, R. Ito, R. Mizuta, M. Nosaka, S.  
356 Watanabe, and H. Sasaki, 2021: Regional characteristics of future changes in snowfall  
357 in Japan under RCP2.6 and RCP8.5 scenarios. *SOLA*, 17, 1–7, doi:10.2151/sola.2021-  
358 001.

359 Kennedy, J., H. Titchner, N. Rayner, and M. Roberts, 2017: input4MIPs. MOHC.  
360 SSTsAndSeaIce. HighResMIP. MOHC-HadISST-2-2-0-0-0. Earth System Grid  
361 Federation. doi:10.22033/ESGF/input4MIPs.1221.

362 Kitoh, A., and H. Endo, 2016: Changes in precipitation extremes projected by a 20-km  
363 mesh global atmospheric model. *Weather and Climate Extremes*, 11, 41–52.

364 Kitoh, A., T. Ose, and I. Takayabu, 2016: Dynamical downscaling for climate projection  
365 with high-resolution MRI AGCM-RCM. *J. Meteor. Soc. Japan*, 94A, 1–16,  
366 doi:10.2151/jmsj.2015-022.

367 Kitoh, A., and H. Endo (2019). Future changes in precipitation extremes associated with  
368 tropical cyclones projected by large-ensemble simulations. *J. Meteor. Soc. Japan*, 97,  
369 141–152.

370 Kusunoki, S., and R. Mizuta, 2013: Changes in precipitation intensity over East Asia  
371 during the 20th and 21st centuries simulated by a global atmospheric model with a  
372 60 km grid size. *J. Geophys. Res.*, 118, 11007-11016, doi:10.1002/jgrd.50877.

373 Li, C., Zwiers, F., Zhang, X., Li, G., Sun, Y., and Wehner, M. (2021). Changes in Annual  
374 Extremes of Daily Temperature and Precipitation in CMIP6 Models, *J. Climate*, 34(9),  
375 3441-3460, doi:10.1175/JCLI-D-19-1013.1.

376 Mizuta, R., H. Yoshimura, H. Murakami, M. Matsueda, H. Endo, T. Ose, K. Kamiguchi,  
377 M. Hosaka, M. Sugi, S. Yukimoto, S. Kusunoki, and A. Kitoh, 2012: Climate  
378 simulations using MRI-AGCM3.2 with 20-km grid. *J. Meteor. Soc. Japan*, 90A, 233–  
379 258.

380 Mizuta, R., O. Arakawa, T. Ose, S. Kusunoki, H. Endo, and A. Kitoh, 2014: Classification  
381 of CMIP5 future climate responses by the tropical sea surface temperature changes.  
382 *SOLA*, 10, 167–171.

383 Mizuta, R., A. Murata, M. Ishii, H. Shiogama, K. Hibino, N. Mori, O. Arakawa, Y. Imada,  
384 K. Yoshida, T. Aoyagi, H. Kawase, M. Mori, Y. Okada, T. Shimura, T. Nagatomo, M.  
385 Ikeda, H. Endo, M. Nosaka, M. Arai, C. Takahashi, K. Tanaka, T. Takemi, Y. Tachikawa,  
386 K. Temur, Y. Kamae, M. Watanabe, H. Sasaki, A. Kitoh, I. Takayabu, E. Nakakita, and  
387 M. Kimoto, 2017: Over 5000 years of ensemble future climate simulations by 60 km  
388 global and 20 km regional atmospheric models. *Bull. Amer. Meteor. Soc.*, 98, 1383–  
389 1398, doi:10.1175/BAMS-D-16-0099.1.

390 Mizuta, R., and H. Endo, 2020: Projected changes in extreme precipitation in a 60-km  
391 AGCM large ensemble and their dependence on return periods, *Geophys. Res. Lett.*,  
392 47, e2019GL086855. doi:10.1029/2019GL086855.

393 Murata, A., H. Sasaki, H. Kawase, M. Nosaka, M. Oh'izumi, T. Kato, T. Aoyagi, F. Shido,

394 K. Hibino, S. Kanada, A. SuzukiParker, and T. Nagatomo, 2015: Projection of future  
395 climate change over Japan in ensemble simulations with a highresolution regional  
396 climate model. *SOLA*, 11, 90–94.

397 Nosaka, M., M. Ishii, H. Shiogama, R. Mizuta, A. Murata, H. Kawase, and H. Sasaki,  
398 2020: Scalability of future climate changes across Japan examined with large-ensemble  
399 simulations at +1.5 K, +2 K, and +4 K global warming levels. *Prog. Earth Planet. Sci.*,  
400 7, doi:10.1186/s40645-020-00341-3.

401 O'Neill, B. C., Tebaldi, C., van Vuuren, D. P., Eyring, V., Friedlingstein, P., Hurtt, G.,  
402 Knutti, R., Kriegler, E., Lamarque, J.-F., Lowe, J., Meehl, G. A., Moss, R., Riahi, K.,  
403 and Sanderson, B. M.: The Scenario Model Intercomparison Project (ScenarioMIP) for  
404 CMIP6, *Geosci. Model Dev.*, 9, 3461-3482, doi:10.5194/gmd-9-3461-2016, 2016

405 Ose, T., 2019: Future changes in summertime East Asian monthly precipitation in CMIP5  
406 and their dependence on present-day model climatology. *J. Meteor. Soc. Japan*, 97,  
407 1041–1053, doi:10.2151/jmsj.2019-055.

408 Roberts, M. J., J. Camp, J. Seddon, P. L. Vidale, K. Hodges, B. Vannière, J. Mecking, R.  
409 Haarsma, A. Bellucci, E. Scoccimarro, L.-P. Caron, F. Chauvin, L. Terray, S. Valcke,  
410 M.-P. Moine, D. Putrasahan, C. D. Roberts, R. Senan, C. Zarzycki, P. Ullrich, Y.  
411 Yamada, R. Mizuta, C. Kodama, D. Fu, Q. Zhang, G. Danabasoglu, N. Rosenbloom,  
412 H. Wang, and L. Wu, 2020: Projected future changes in tropical cyclones using the  
413 CMIP6 HighResMIP multimodel ensemble. *Geophys. Res. Lett.*, 47, e2020GL088662.  
414 doi:10.1029/2020GL088662.

415 Sasaki, H., K. Kurihara, I. Takayabu, and T. Uchiyama, 2008: Preliminary experiments  
416 of reproducing the present climate using the non-hydrostatic regional climate model,  
417 *SOLA*, 4, 25–28.

418 Sugi, M., and J. Yoshimura, 2012: Decreasing Trend of Tropical Cyclone Frequency in  
419 228-year High-resolution AGCM Simulations. *Geophys. Res. Lett.*, 39, L19805.

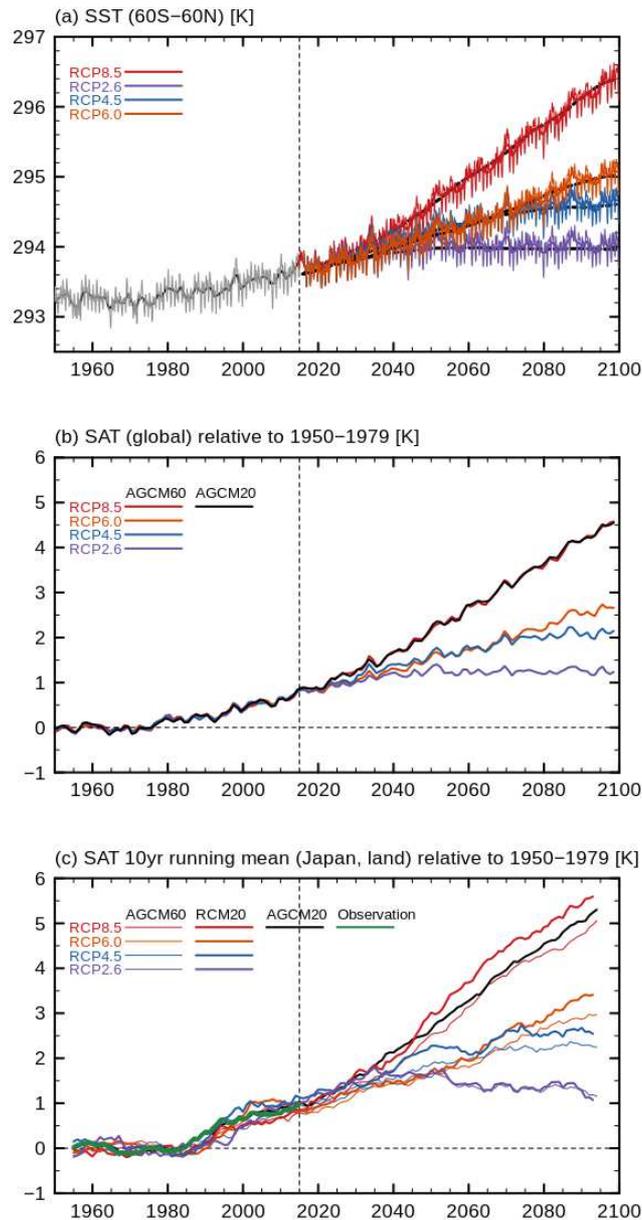
420 Yamada, Y., C. Kodama, M. Satoh, M. Sugi, M. J. Roberts, R. Mizuta, A. T. Noda, T.

421 Nasuno, M. Nakano, and P. L. Vidale, 2021: Evaluation of the contribution of tropical  
422 cyclone seeds to changes in tropical cyclone frequency due to global warming in high-  
423 resolution multi-model ensemble simulations. *Prog. Earth Planet Sci.*, 8, 11.  
424 doi:10.1186/s40645-020-00397-1

425 Yukimoto, S., H. Kawai, T. Koshiro, N. Oshima, K. Yoshida, S. Urakawa, H. Tsujino, M.  
426 Deushi, T. Tanaka, M. Hosaka, S. Yabu, H. Yoshimura, E. Shindo, R. Mizuta, A. Obata,  
427 Y. Adachi, and M. Ishii, 2019: The Meteorological Research Institute Earth System  
428 Model Version 2.0, MRI-ESM2.0: Description and Basic Evaluation of the Physical  
429 Component. *J. Meteor. Soc. Japan*, 97, 931–965, doi:10.2151/jmsj.2019-051.

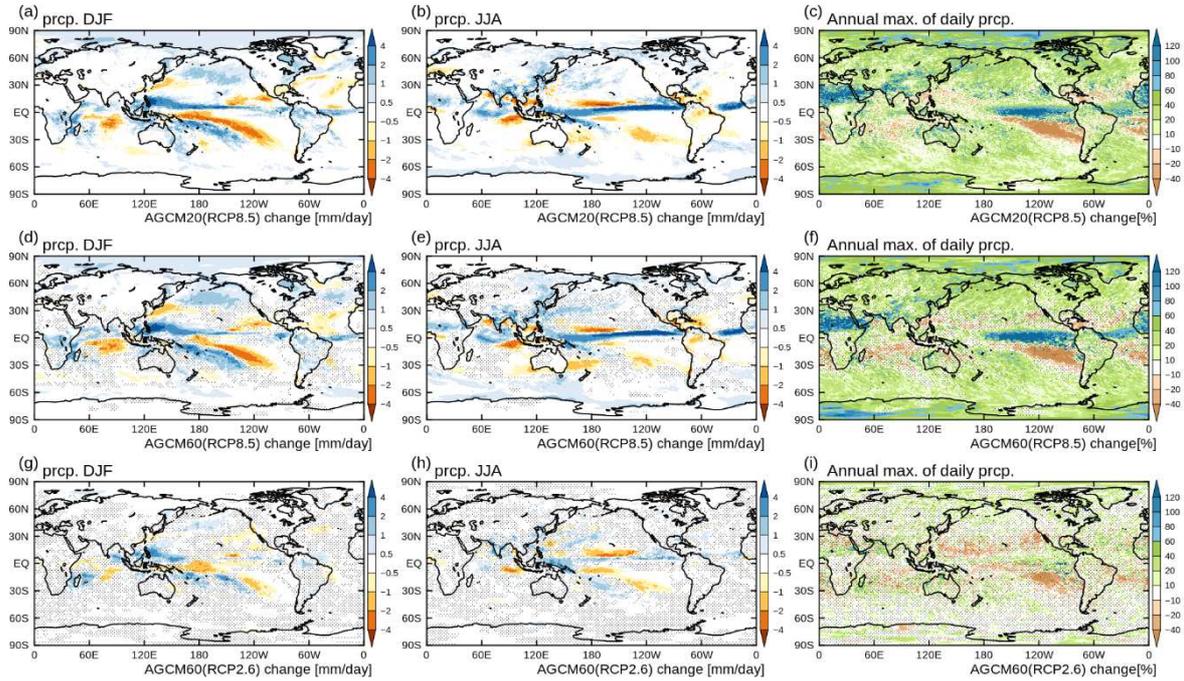
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433 Fig. 1. (a) Time series of SSTs prescribed for the AGCM simulations averaged over 60°S–  
 434 60°N. Thin lines are monthly-mean values and thick lines are annual-mean values. Thick  
 435 black dashed lines after 2015 show the warming trends without interannual variability  
 436 calculated from the CMIP5 model experiments. (b) Time series of the global-mean  
 437 annual-mean SAT change from the 30-year average of 1950–1979. (c) As (b), but 10-year  
 438 running-mean values averaged over the Japanese land grids, including the average of  
 439 observed data.

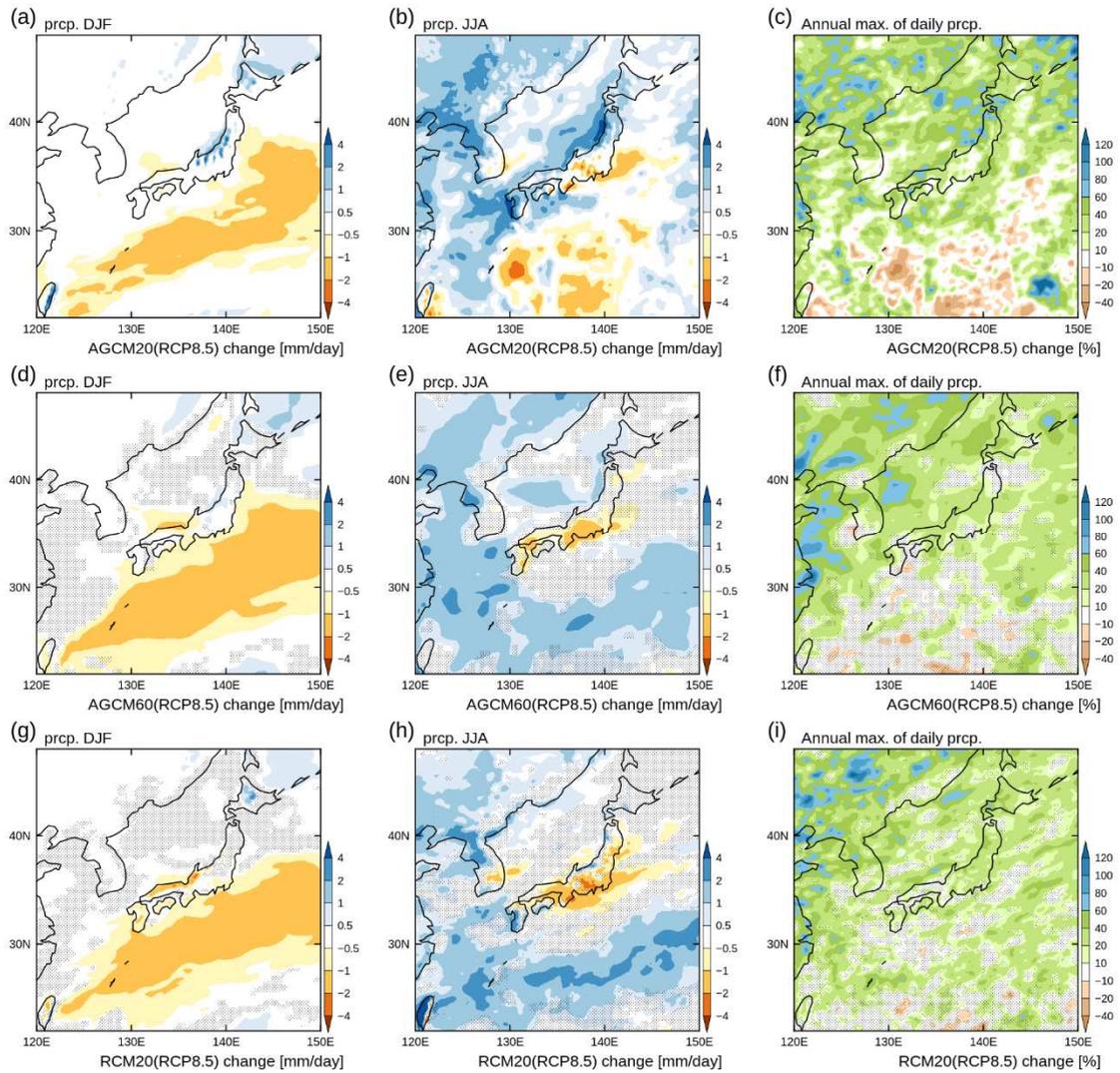


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441 Fig. 2. Horizontal distribution of the changes from 1979–2003 to 2075–2099 for (a, d, g)  
 442 mean precipitation from December to February (mm/day), (b, e, h) mean precipitation  
 443 from June to August (mm/day), and (c, f, i) relative changes in Rx1d (%), using (a–c)  
 444 RCP8.5 with AGCM20, (d–f) RCP8.5 with AGCM60, and (g–i) RCP2.6 with AGCM60.

445 The hatches in (d–i) indicate that the change is not statistically significant at the 95 %  
 446 level against the ensemble spread, calculated from the four-member historical simulations.

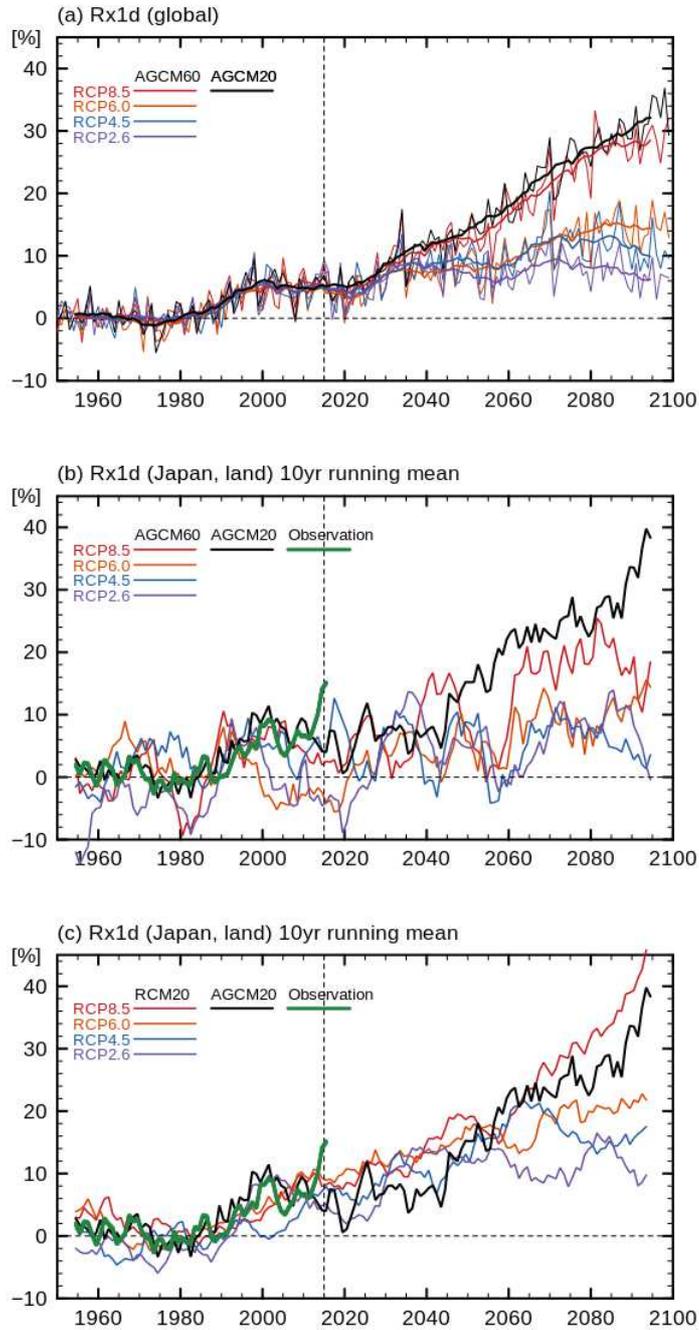
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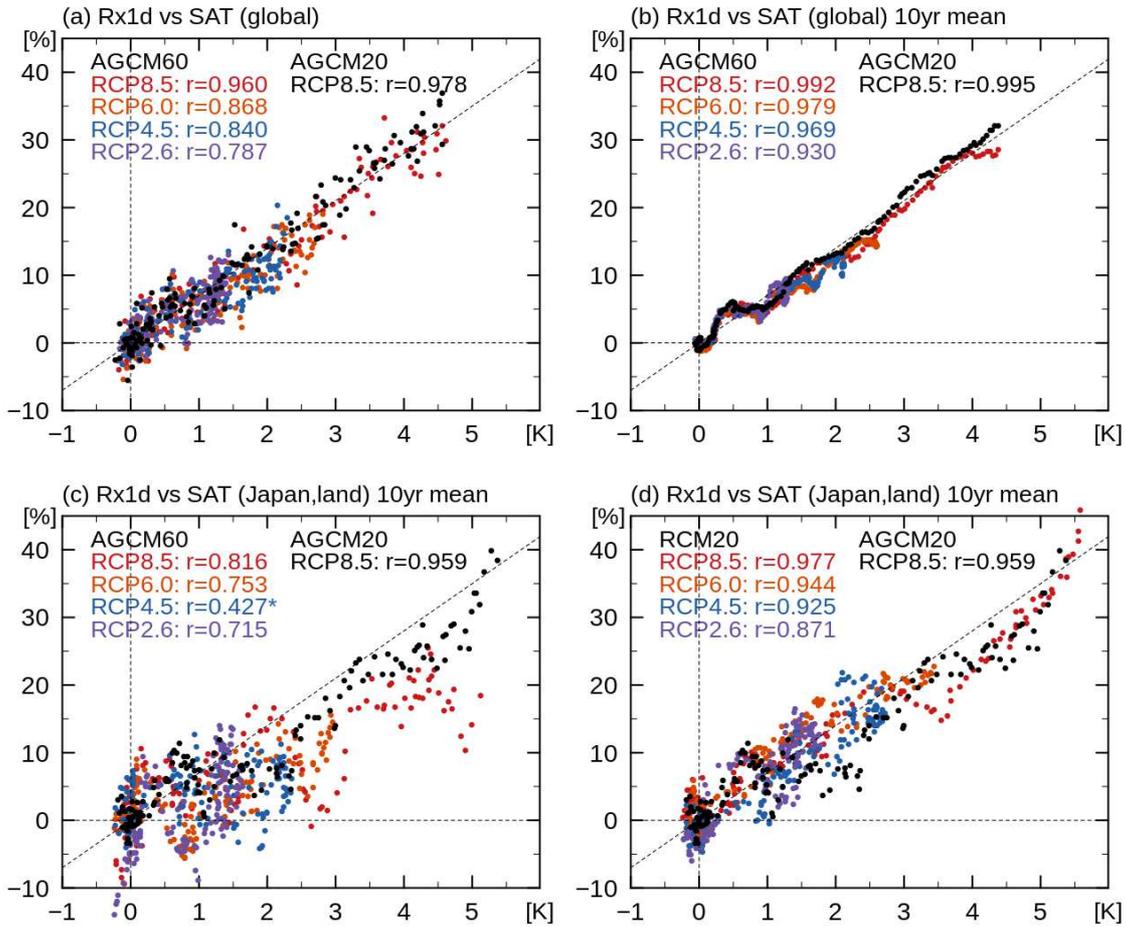
449 Fig. 3. As Fig. 2, but around Japan, using RCP8.5 with (a–c) AGCM20, (d–f) AGCM60,  
 450 and (g–i) RCM20.

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453 Fig. 4. (a) Time series of the global mean of Rx1d change relative to the 30-year average  
 454 over the period 1950–1979. Thin lines are the value for each year, and the thick lines are  
 455 the 10-year running mean. (b, c) As (a), but showing the 10-year running-mean values  
 456 averaged over the Japanese land grids, from (b) AGCM60 and (c) RCM20, compared  
 457 with AGCM20 and the average of the observed data.



458

459 Fig. 5. Scatter plots of relative change in Rx1d (%) and SAT change (K), for (a) global-  
 460 mean values for each year, (b) global-mean values for 10-year running mean,  
 461 10-year running mean values averaged over the Japanese land grids, from (b) AGCM60 and  
 462 (c) RCM20, compared with AGCM20. The changes are relative to the 30-year average  
 463 over the period 1950–1979. The dashed line indicates a slope of 7% per degree. The  
 464 correlation coefficients are calculated using 10-year means of every 10 years. Asterisks  
 465 denote that the correlation is not statistically significant at the 99% significance level.

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