# Evaluation of extreme precipitation in the Yangtze River Delta Region of China using a 1.5 km mesh convection-permitting regional climate model

Guangtao Dong<sup>1</sup>, Zhiyu Jiang<sup>2</sup>, Ya Wang<sup>3</sup>, Zhan Tian<sup>4</sup>, and Junguo Liu<sup>5</sup>

 <sup>1</sup>Shanghai Climate Center
 <sup>2</sup>School of Atmospheric Sciences, Nanjing University
 <sup>3</sup>Shanghai meteorological disaster prevention technology center
 <sup>4</sup>School of Environmental Science and Engineering, Southern University of Science and Technology
 <sup>5</sup>Southern University of Science and Technology

November 24, 2022

#### Abstract

Realistic representation of detailed rainfall characteristics on local scale by current state-of-the-art climate models remains a key challenge, especially on sub-daily timescales. In this research, the convection-permitting Weather Research and Forecasting (WRF) model configured with 1.5 km grid spacing is used to simulate precipitation on sub-daily timescales over the Yangtze River Delta Region of China for continuous 10 years (2005–2014). The simulations are compared to rain gauge observations, reanalysis data, and the simulations of a lower resolution WRF with 9 km grid spacing that has a parameterization of convection. The results show that precipitation over the region can be well captured by using the convection-permitting model (CPM). Furthermore, the intensity, duration and coverage of these precipitation events can be more accurately described by the CPM. On the convection timescales of 1–4 hours, especially for heavy rainfall events, the CPM is more accurate than the convection-parameterized model in capturing the short-duration events, which may be related to its better account of physical processes related with the convection on the convection-permitting scale. In addition, the extreme events which are more localized and with short-duration can be represented better by the CPM while the convection-parameterized model tends to produce widespread precipitation events covering more grid cells than observed. Biases of the simulation by the 9-km mesh convection-parameterized mode appear to be related to the deficiencies in the representation of convections.

1	Evaluation of extreme precipitationin the Yangtze River Delta Region of China using
2	a 1.5 km mesh convection-permitting regional climate model
3	
4	Guangtao Dong <sup>1,2</sup> , Zhiyu Jiang <sup>3</sup> , Ya Wang <sup>4</sup> , Zhan Tian <sup>5</sup> *, Junguo Liu <sup>5</sup>
5	<sup>1</sup> Shanghai Climate Center, Shanghai, 200030, China
6 7	<sup>2</sup> Key Laboratory of Cities' Mitigation and Adaptation to Climate Change in Shanghai, China Meteorological Administration, Shanghai, 200030, China
8	<sup>3</sup> School of Atmospheric Sciences, Nanjing University, Nanjing, China, 201400, China
9	<sup>4</sup> Shanghai Meteorological Disaster Prevention Technology Center, Shanghai, 200030, China
10 11	<sup>5</sup> School of Environmental Science and Engineering, Southern University of Science and Technology, Shenzhen, 518055, China
12	
13	Correspondingauthor: Zhan Tian (tianz@sustech.edu.cn)
14	Key Points:
15	• We employ a 1.5 km mesh convection-permitting regional climate model (CPM) and a 9
16 17	km regional climate model (RCM) to run and evaluate decadal continuous regional climate simulations on sub-daily timescales.
18	• The 9-km RCM tends to under-estimate the intensity of heavy rainfall events, and over-
19	estimate their persistence and spreading extent. By contrast, the 1.5-km CPM gives a
20	much better representation of their intensity, duration and spatial extent.
21	• The extreme events which are more localized and with short-duration can be represented
22	better by the 1.5 km CPM due to its better account of physical processes related with
23 24	convection.
24	

#### 25 Abstract

Realistic representation of rainfall characteristics on local scales by state-of-the-art climate 26 models remains a key challenge, especially on sub-daily timescales. In this study, the 27 convection-permitting Weather Research and Forecasting (WRF) model configured with 1.5 km 28 29 grid spacing is used to simulate precipitation on sub-daily timescales over the Yangtze River Delta Region of China for continuous 10 years (2005–2014). The simulations are compared 30 withrain gauge observations, reanalysis data, and the simulations of a lower resolution WRF with 31 32 9 km grid spacing that has a parameterization of convection. The results show that precipitation 33 over the region can be well captured by using the convection-permitting model (CPM). Furthermore, the intensity, duration and coverage of these precipitation events can be more 34 accurately described by the CPM. On the convection timescales of 1-4 hours, especially for 35 heavy rainfall events, the CPM is more accurate than the convection-parameterized model in 36 capturing the short-duration events, which may be due to its better account of physical processes 37 related to the convection on the convection-permitting scale. In addition, the extreme events 38 39 which are more localized and with short-duration can be represented better by the CPM while the convection-parameterized model tends to produce widespread precipitation events covering more 40 grid cells than observations Biases of the simulation by the 9-km mesh convection-parameterized 41 mode appear to be related to the deficiencies in the representation of convections. 42

#### 43 **1 Introduction**

Changes in the intensity and temporal distribution of sub-daily precipitation are of great 44 45 interest due to their potentially important hydrological impacts (Ban et al., 2015). They are often responsible for flash floods and can cause significant damages in small and steep catchments as 46 well as urban areas (Kendon et al., 2018). Changes in extreme and intensive precipitation are of 47 particular concern because of their great impacts on the society through the generation of floods, 48 49 leading to infrastructural damages and even human casualties (Helsen et al., 2019). Therefore, better understanding of the extreme sub-daily precipitation changes is crucial for decision makers 50 to make more sustainable and responsive management plans. Precipitation data at sub-daily 51 scales and spatial scales of 1–10 km<sup>2</sup> are needed for the urban drainage design (Arnbjerg-Nielsen 52 et al., 2013). Therefore, it is of great significance to understand the processes influencing daily-53 scale or hourly-scale precipitation. 54

Regional climate models (RCMs) are widely used in regional climate researches and 55 future climate simulations all over the world (Feser et al., 2011; Donner et al., 2011), and it is a 56 powerful tool for studying regional precipitation characteristics. The RCMs allow a better 57 representation of details of orography and coastal zones, characteristics of vegetation and soil, 58 and descriptions of atmospheric processes at smaller scales (Fosser et al., 2015). These 59 characteristics enable RCMs to produce more realistic simulation results than global circulation 60 models (GCMs), and to better capture the temporal and spatial characteristics of the regional 61 climate (Feser et al., 2011). Generally, RCMs are able to capture the average statistics of daily 62 and monthly precipitation. They show a good agreement with observations for moderate 63 precipitation, but their simulation biases increase for heavier precipitation events (Kendon et al., 64 2012). RCMs also usually underestimate non-precipitation days (Fowler et al., 2007) and 65 overestimate continuous light-rain days (Boberg et al., 2009; Kendon et al., 2019). When the 66 horizontal resolution is coarser than 10 km, the convective parameterization is adopted to 67 represent convections. The parameterization is known as a major source of the error and 68 uncertainty in climate simulations (Prein et al., 2015; Brisson et al., 2016), usually leading to the 69 70 misrepresentation of the diurnal cycle of convective precipitation (Kendon et al., 2012, 2014; Prein et al., 2015; Maraun et al., 2017) and the underestimation of hourly precipitation intensities 71 72 (Prein et al., 2015).

Previous studies have shown that model resolution is an important factor affecting the 73 simulation ability of RCMs (Lucas-Picher et al., 2017). Generally speaking, the higher the 74 resolution is, the better the temporal and spatial characteristics of the regional climate can be 75 captured. With the horizontal resolution increasing until reaching the convection-permitting (CP) 76 scales (~4 km), the cumulus convective parameterizations could be switched off and it is 77 possible to explicitly represent the convection when simulating the regional climate (Guo et al., 78 79 2019b). In these convection-permitting models (CPMs), the convection parameterization can be switched off, and the representation of surface and boundary layer processes can be greatly 80 enhanced (Guo et al., 2019a). To date, there have been increasing number of studies on regional 81 climate simulations using CPMs. Kendon et al. (2012) showed that rainfall simulated by the 1.5-82 km-resolution RCM is much more realistic than that simulated by the 12-km-resolution RCM, 83 due to that much continueslight rain and pretty reduced errors in the diurnal cycle by the 1.5-km-84 resolution RCM. Stratton et al. (2018) also emphasized the CPMs' improvement over coarse 85

RCMs on extreme precipitation through a case study over pan Africa. Long-term (more than ten 86 years) continuous simulations in Europe and the United States show that CPMs can significantly 87 improve the representation of both hourly intensity distribution and diurnal cycle of precipitation 88 (Prein et al., 2015; Kendon et al., 2015). Some studies have found that there are some large 89 errors in the hourly precipitation simulated by the coarse-resolution model compared with the 90 observation—the duration of precipitation is significantly longer than the observation (Berthou et 91 al., 2019), the intensity of hourly precipitation is weaker than the observation (Knist et al., 2018), 92 and the spatial coverage of strong hourly precipitation is larger than the observed one (Chang et 93 al., 2018).By contrast, the CPMs have exhibited an obviously better performance on the sub-94 daily scale, and can well capture the precipitation with the short duration, high intensity and 95 concentrated range (Chan et al., 2014; Ban et al., 2014; Fosser et al., 2015; Yang et al., 2019). 96 97 The improvement in precipitation characteristics simulated by CPMs is attributed to the fact that the CPMs can generate realistic showers by better representating the organization of convections, 98 99 and forecast local extreme events, which is hardly captured when the resolution is not fine enough(Kendon et al., 2012). The convection-permitting scale has added values in the 100 101 representation of the precipitation field and the corresponding atmospheric fields (Fosser et al., 2015). 102

In China, Li et al. (2018) simulated precipitation during the warm season (April-103 September) in China, showing that CPMs can improve the simulated spatial distribution of 104 precipitation in summer. Xiong et al. (2013) showed that the Regional Integrated Environmental 105 Model System 2.0 (RCMs without the convection parameterization) with the resolution of 3 km 106 had a good representation of the seasonal cycle of precipitation during January 1 to December 31, 107 2000 in the Heihe River Basin. In the eastern China, the effects of microphysics parameterization 108 (MP) schemes on precipitation characteristics were investigated by Guo et al. (2019b) with a six-109 year warm-season climate simulation, and Guo et al. (2019a) with a 13-year continuous 110 simulation. Their results indicated that CPMs can also capture the diurnal characteristics of the 111 precipitation amount and frequency. Disastrous weather such as short-term heavy precipitation 112 and mesoscale convective systems frequently occur in the eastern China (Yang et al., 2015), 113 which can be explicitly represented in CPMs. Nevertheless, even in convection-committing scale, 114 115 higher benefit can still be gained when resolution is enhanced from 4 km to 1.5 km due to insufficient representation of convective downdrafts in 4 km convection-permitting simulations 116

(Murata et al., 2017). So far, there have been no studies dealing with decadal long continuous convection-permitting simulations at 1.5 km resolution over eastern China. Thus there is an urgent need to conduct long-term continuous simulations by CPMs at 1.5 km resolution and to analyze the added values of such simulations over traditional RCMs in terms of hourly precipitation characteristics, such as the intensity, duration and coverage of extremely hourly precipitation as well as diurnal cyle.

By evaluating the performance of CPMs in reproducing the spatial-temporal distribution 123 124 of precipitation in the Yangtze River Delta, this paper aims to explore what improvements CPMs can make in simulating the frequency, intensity, and duration of precipitation in the region, 125 compared with the low-resolution model with convection being parameterized. The manuscript is 126 organized as follows. Section 2 introduces the RCM model and experimental design, describes 127 the observation and reanalysis datasets used in the analysis. The performance of CPM in 128 129 reproducing extreme precipitation characteristics is assessed in Section 3, including the simulation effects of CPM on the precipitation duration, intensity and coverage. Finally, the 130 discussion and conclusions from this study are given in section 4. 131

#### 132 2 Methods and study area

### 133 2.1 Observation data

In this paper, three observation datasets are applied in evaluating the capacity of the 134 model in simulating precipitation over the eastern China. Hourly precipitation during 2003–2012 135 136 and daily precipitation during 2005–2014 and are used in the analysis. These data were obtained from 70 automatic weather stations in the Yangtze River Delta region, are provided by the 137 National Meteorological Information Center (NMIC) of China Meteorological Administration 138 (CMA). The gauge-satellite merged dataset (AWSCMORPH; Shen et al., 2014), which is 139 obtained by merging the observations at over 30,000 national and regional automatic weather 140 141 stations with the precipitation product (Joyce et al. 2004), is from the NOAA Climate Precipitation Center Morphing Technique (CMORPH). This merged dataset has a temporal 142 resolution of one hour and a spatial resolution of  $0.1^{\circ} \times 0.1^{\circ}$  for the observational period of 143 2008–2014. The CMORPH precipitation dataset has been applied widely to evaluate the features 144 145 of rainfall over China, and it has been proven suitable for examining the detailed spatial patterns

and temporal variations of precipitation over the eastern China (Qian et al., 2015; Chen et al.,
2016). In this study, the simulated rainfall spatial coverage is evaluated by CMORPH hourly
precipitation.



149

Figure 1. The domains of the WRF model domain configuration in this study: the boundary of the 9-km-resolution outer domain (D01, blue) and the 1.5-km-resolution inner domain (D02, red). Black dots are the locations of automatic weather stations used in the model evaluation.

154 2.2 Model configuration

The climate model used in this paper is the two-domain nested atmospheric model WRF 155 (Weather Research and Forecasting model) version 3.7.1. The WRF driven by the ERA-Interim 156 reanalysis data has been run with an outer resolution of 9 km (hereinafter referred to as 9-km 157 models or W9k or LSM) in this study (Figure 1). In this configuration, deep convection is 158 simulated by Kain-Fritsch convective parameterization scheme (Kain et al., 2004). The inner 159 160 domain with a resolution of 1.5 km (hereinafter referred to as 1.5-km models or W1p5k or CPM) makes it possible to explicitly represent the deep convection without the convection 161 parameterization scheme. The simulation was performed from December 1, 2004 to January 1, 162 2014. The data from the first month (i.e., December 2004) of the simulation was discarded as the 163 spin-up. The simulated meteorological variable values at the grids closest to the weather stations 164

are used for the comparison with the observation. The locations of the automatic weather stationsare shown in Figure 1.

167

**Table 1.** Experiment design for the simulations.

Model property	W1p5k	W9k	
Horizontal resolution	1.5 km	9 km	
Vertical resolution	51 levels	51 levels	
Time step	5 s	30 s	
Convection scheme	None	Kain-fritsch(Kain et al., 2004)	
Microphysics	WSM5	WSM5(Hong et al., 2004)	
Radiation	Rrtmg	Rrtmg(Iacono etal., 2008)	
Land surface	NOAH	NOAH(Chen and Dudhia,	
		2001)	

168

#### 169 2.3 Analysis methods

To evaluate the rainfall simulated by the W1p5k and W9k, the duration and intensity of 170 rainfall events are examined in this paper. The analysis includes all the stations within the 171 W1p5k inner domain (red rectangle in Figure 1). 10 years of hourly rainfall data from each 172 173 dataset, corresponding to the period of 2005–2014 (hourly observation 2003-2012) are analyzed in this paper. For each station in the D02 domain, precipitation occurrences or "events" are 174 175 defined as the precipitation amount/duration exceeds a given threshold at a given location. For each precipitation event, the duration of precipitation above the threshold and total precipitation 176 177 are also recorded.

As our focus is on heavy precipitation, the intensity within a threshold range at the tail end of the precipitation distribution has been analyzed. Considering the uncertainty of observations and the possibility of underestimating heavy precipitation, percentile thresholds are used in this study. In particular, the p-th percentile of the hourly rainfall distribution for all ten years in each dataset is considered as the percentile threshold of p. Through the study on the temporal characteristics of rainfall over this percentile, we can see that the absolute values vary among different datasets. The advantage of using percentile values is that the effect of precipitation biases between simulations and observations could be eliminated, while the reliable information on precipitation patterns and behaviors could still be retained (Kendon et al., 2012).

In order to analyze the relationship between precipitation intensity and the precipitation duration, the joint probability distribution is analyzed in this paper. Let  $\{a_1, a_2, ..., a_n\}$  denote the rainfall amount of all precipitation events,  $\{d_1, d_2, ..., d_n\}$  the corresponding durations, then the rainfall and durations are further divided into predefined ranges, respectively, and are recorded as  $\{A_1, A_2, ..., A_p\}$ ,  $\{D_1, D_2, ..., D_q\}$ . The joint probability is then defined as follows.

192 
$$p_{ij} = P\{X = A_i, Y = D_j\}$$
 (1),

#### 193 where

194  $p_{ij} \ge 0, \sum_i \sum_j p_{ij} = 1$  (2).

#### 195 **3 Results**

In this section, the WRF simulation results are compared with the observations to assess the performance of the CPM. To facilitate direct comparison between W9k and W1p5k, the W9k provides lateral boundary conditions to the W1p5k.

## 199 3.1 The distribution of mean and extreme precipitation

To give a broad picture of precipitation in the study area, Figure 2 shows the monthly 200 mean precipitation, wet days and wet hours. Table 2 further lists the statistical results including 201 the correlation coefficients (CORs) and root mean square errors (RMSEs) between the 202 observations (records at the automatic weather stations) and simulations. Figure 2a shows that 203 precipitation in this region occurs mostly in the summer (June-August, JJA), with much lower 204 level and relatively little variation in other months. The simulated rainfall exhibits significant 205 peaks in June and July for W1p5k and W9k, respectively. In the summer, the W9k overestimates 206 the precipitation amount, which is almost twice the observation in July. W1p5k has a much 207 better performance than W9K, although it underestimates monthly rainfall in July and August. 208 Both of themperform well in other months. In general, the simulation results of the CPM are 209 close to those of the LSM in terms of monthly average precipitation in dry seasons. Major 210

211 improvements of the CPM occur in the summer when precipitation is dominated by deep

212 convections.



Figure 2.Seasonal cycles of monthly mean (a) precipitation, (b) wet days and (c) wet hours. Wet days/hours are define as precipitation  $\geq 1 \text{ mmd}^{-1}$  or  $0.1 \text{mmh}^{-1}$  respectively.

216

213

Table 2. Statistics between the simulations and observations for the seasonal cycle of
 precipitation.

Statistics		W9k		W1p5k	
Statistics	COR	RMSE	COR	RMSE	
PRCPTOT	0.99	67.66 mm	0.77	32.36 mm	
Wet days	0.80	3.85 day	0.91	1.40 day	
Wet hours	0.40	23.82 h	0.80	11.29 h	

219



Figure 3. Biases of the RCM-simulated precipitation intensity compared with the observations on (a–f)daily and (g-l) hourly scales.

223

As shown in Figure 3, the precipitation intensity simulated by W1p5k is more consistent with that from the observation than W9k. The observed daily precipitation intensity is largely between 12.0 and 13.5 mmday<sup>-1</sup>, and the simulated values at most stations by W1p5k are also within this range. The differences between the two are mostly below 10%. The simulated precipitation intensities at individual stations such as those located in Shanghai area, are up to 15

mmday<sup>-1</sup>. In contrast, the daily precipitation intensity simulated by W9k is about 20% higher 229 than the observations. Forhourly precipitation intensity, the simulations by both W9k and W1p5k 230 are close to the observations, with the absolute difference below 30%. Although the annual-mean 231 hourly precipitation intensity (defined as the annual total precipitation divided by the annual total 232 rainy hours) simulated by W9k is close to the observations, the annual mean precipitation and 233 rainy hours are over 20% higher than the observation. The results are much improved by W1p5k 234 simulations which show good skills in simulating annual mean precipitation, rainy hours and 235 hourly precipitation. Furthermore, the probability density distribution (PDF) of hourly 236 precipitation is also better simulated by W1p5k. The occurrence frequency of precipitation 237 within 0.1–15 mmh<sup>-1</sup> simulated by W9k is significantly higher than the observed values, while 238 W9k underestimates the occurrence frequency of above 15 mmh<sup>-1</sup>. On the other hand, the PDF 239 of the hourly precipitation simulated by W1p5k agrees well with that of the observations. 240

The precipitation amount simulated by W9k is much higher than the observations (Figure 4), especially for JJA (June, July and August), while that simulated by W1p5k is slightly lower than the observations in JJA. For the boreal winter (DJF), precipitation amount is well simulated by both W9k and W1p5k. Note that as precipitation amount in JJA is much more than that in DJF, the model errors in representing the annual (ANN) precipitation is largely caused bythe errors in JJA.

In order to investigate the reasons leading to the overestimation of JJA precipitation by 9-247 km RCM. The difference of the large-scale atmospheric circulationsbetween the 9-km RCM and 248 ERA-interim Reanalysis are presented in Supplementary Materials (SM). Figures and discussions 249 in SM show that 9-km RCM lacks the skill for realistically simulating the position and intensity 250 of the western Pacific subtropical high which leads to much of the moisture transport 251 concentrates near Zhejiang province. The intensities of 850-hPa wind speed are also 252 overestimated to some extent, which could be responsible for the overestimation of the summer 253 precipitation over the eastern China. 254



255

Figure 4. Spatial distribution patterns of annual mean (2005–2014) and seasonal mean precipitation in the observations (top) and the simulations by the RCMs (middle: W9k; bottom: W1p5k) (unit: mm).

Figure5a further shows the frequency distributions of daily precipitation intensity 260 averaged over all the 70 stations from both the observations and simulations (W9k and W1p5k) 261 during 2005–2014. Particular attention has been paid to the heavy or extreme precipitation with 262 the intensity over 30 mm  $d^{-1}$ . For daily precipitation intensity within the range of 30–150 mm, 263 the W1p5k simulation shows lower frequency than that by W9k, which agrees better with the 264 observations. It can be clearly seen that W1p5k well simulates the rainfall frequency below the 265 99th percentile (Figure 6b). It is obvious that the high-resolution CPM makes better performance 266 in simulating extreme precipitation, compared with coarser-resolution experiments using a 267 convection parameterization which may overestimate the precipitation with a longer rainy season. 268



Figure 5. Observed and two RCM simulated frequency distributions of the (a) daily and (b) hourly precipitation intensity .



269

Figure 6. (a) Daily precipitation intensity, (b) hourly intensity for cases with precipitation > 0.1 mmh<sup>-1</sup> and (c) hourly intensity in all hours corresponding to a given percentile threshold in the observations and simulations by W1p5k and W9k during 2005–2014. The horizontal axis means the percentile threshold of hourly precipitation across the evaluation period (10 years).



Figure 7. Difference of the RCM-simulated frequencies with the observation at different precipitation grades, which are (a–b) wet (> 0.1 mm h<sup>-1</sup>), (c–d) light rain (0.1–5.0 mm h<sup>-1</sup>), (e–f) moderate rain (5.0–20.0 mm h<sup>-1</sup>) and (g–h) heavy rain (>= 20.0 mm h<sup>-1</sup>), respectively.

277

The performances of W9k and W1p5k for simulating precipitation with different 282 intensity thresholds are assessed by analyzing their corresponding frequencies (Figure 7). W1p5k 283 shows a good representation of the total precipitation hours and light rain hours (Figures 7a–7b) 284 285 with the model biases at most stations within 10%. While the simulations by W9k are over 30% larger than the observations. The results for light rain are consistent with that of total rainy hours 286 which lead to the large amount of light rain in the study area. Furthermore, the precipitation 287 occurrence within the range of 5–20 mm  $h^{-1}$  (Figures 7e–7f) simulated by W9k is significantly 288 overestimated by more than 70% above the observations, while that simulated by W1p5k at the 289 stations nearby Shanghai is overestimated by around 30%. Nevertheless, there are little 290 differences between the simulations of precipitation above 20 mm  $h^{-1}$  by W9k and W1p5k, with 291 both of them exhibiting an underestimation in these extreme precipitation events. 292

293

### 3.2 Duration of heavy precipitation

It is found that precipitation in W9k peaks around 12:00pm (Figure8b), and the event with this peak lasts for 3–5 hours. This significantly differs from the observations which show

that precipitation reaches its peak around 15:00-18:00pm and with duration of 1-3 hours. The 297 precipitation diurnal cycle is improved in W1p5k which shows that more intensive precipitation 298 occurs in the afternoon around 12:00-18:00pm, with duration of 1-3 hours. The observed 299 precipitation generally lasts for no more than 8 hours, however the durations of events which 300 start during 0:00-6:00 could last for about 10 hours in some cases. Comparing with W9k, more 301 precipitation is produced by W1p5K during the period of 20:00-03:00, and the associated 302 duration is longer than the observations. In addition, it is noted that the observed short-term 303 precipitation during the period of 15:00-18:00 accounts for the largest ratio to the total 304 precipitation. Generally, W1p5k performs better than W9k in both duration and distribution of 305 heavy precipitation with short durations. 306



307

**Figure8.** The diurnal cycles of different durations of the intensity of precipitation to its total precipitation for the (a) observation, (b) W9k and (c) W1p5K (unit: %) in JJA. The horizontal axis represents the Beijing time, and the vertical axis indicates the duration of the precipitation events and the shaded colors are the corresponding ratios to total precipitation. Note the events longer than 24 hours are not shown.

313

Previous studies have revealed that the convective precipitation generally lasts for a short time (1–4 hours) and tends to be more local (Chan et al., 2014). Thus, the precipitation structure in the study area is further illustrated (Figure 9). It is found that short-duration precipitation in summer is one of the principal precipitation types in the study domain. Observations show that precipitation with a duration of 2–5 hours accounts for nearly 30% of the total precipitation. After that, the proportion decreases with the increase of the duration, and tends to be flat for the events longer than 20 hours. The maximum precipitation ratio simulated by W9k reaches about 8.5% for the rainfall events with durations of 4 hours, and then the ratio decreases with the increase of the duration. W9k underestimates the precipitation during the events which last less than 3 hours. This means that convective precipitation may be underestimated in W9k. In contrast, the W1p5k shows a more realistic peak duration which corresponds well with the observation.

In order to investigate whether W1p5k can represent the rainstorm with more realistic 326 327 characteristics in the study domain, special attention has been paid to the rainstorm duration. Figure 10 shows the frequency distribution of precipitation events at different duration thresholds. 328 The horizontal rows correspond to the frequency distribution of precipitation events that exceed 329 a specific precipitation threshold. Each box in the row represents the probability of precipitation 330 events for a certain duration. Observations show that high threshold precipitation usually has 331 short duration. For the precipitation threshold value of 99.9% or above, the proportion of 332 precipitation events lasting for 4 hours or above is very low. In the W9k, rainfall persists too long 333 334 compared with the observations. The 9-km RCM lacks one-hour precipitation events while abounds two or more hours events. Indeed, the comparison between the 9-km RCM & 335 observations and the 1.5-km & 9-km RCMs look like the inverse of each other (cf. Figure 10d 336 and 10f). Heavy rain in the 1.5-km RCM is shorter-persistent than in the 9-km RCM, the 337 probability of rainfall lasting for just one hour is significantly higher for all percentile thresholds 338  $\leq$  99.99%, and also with a remarkably fewer probability of rain lasting 2~5 hours for all 339 percentile thresholds  $\leq$  99.5%. The 1.5-km RCM shows great added value in terms of rainstorm 340 duration characteristics. 341





**Figure9.** Ratio of the precipitation with different durations to the total precipitation (unit: %, events with durations longer than 24 hours are not shown) in the observation (black bar) and the two simulations by W9k (blue solid line) and W1p5k (red solid line) in JJA. The horizontal axis represents the duration of the precipitation event (> 0.1 mm, unit: hour), and the vertical axis represents the proportion of precipitation with different durations to the total precipitation (unit: %).

#### 349 3.3 Size of heavy rain cells

To assess the spatial structure, the coverage of rainfall cells exceeding certain intensity thresholds is shown in Figure 11. Observations show that heavy precipitation events tendto be more local. The probability of large precipitation cells decreases with the increasing threshold. Heavy precipitation (exceeding the percentile threshold of 99.5) covers a widespread area in the simulation by W9k, while there are too many moderate rainfall cells covering 50–100 grids and extreme rainfall covering a single grid. At the same time, there is a lack of rainfall events covering a single grid in W9k simulations.

In contrast, heavy precipitation simulated by W1p5k is more local than that simulated by W9k, leading to a better agreement with the observation. W1p5k produces more widespread light rainfall cells while W9k tends to produce more widespread heavy rainfall cells. The precipitation simulated by W1p5k seems to be more concentrated, with numerous one-grid light or moderate rainfall events. Considering that the difference between W1p5k and W9k is almost opposite to that of W9k against the observations, it can be concluded that the W1p5k has a better skill in simulating the size of heavy rain cells.

364 3.4 Rainfall durations versus peak intensity

Considering only some of the rainfall events exceeding a given threshold, there is a 365 limitation that the heaviest rainfall, which usually short-lived and local, will be embedded in a 366 larger scale or longer duration precipitation event with lower intensity (Kendon et al., 2012). To 367 assess this characteristic, Figure 12a shows that precipitation events with a lower peak intensity 368 are short-lived, and moderate-intensity precipitation is more likely to be embedded within 369 relatively longer precipitation events (2–5 hours). W9k produces too few short (1–2 hours) 370 medium-intensity (10-20 mm) and high-intensity (> 30 mm) events and too many long (2-6 371 hours) low-intensity events. On the other hand, W1p5k tends to agree with the observations quite 372 well when the rainfall intensity exceeds 5 mm h-1 (Figure 12c). The differences of the W9k 373 simulation against the observations and that of the W1p5k simulation against the W9k simulation 374 look opposite to each other, indicating that W1p5k is in a better agreement with the observations. 375 Though both W9k and W1p5k underestimate the rainfall with the intensity of 20-30 mm or 376 heavier and lasting 2–6 hours, W1p5k produces a higher precipitation probability than W9k at 377 this condition. It means W1p5k is more skillful in representing short-lived and high-peak-378 intensity events. Similar improvements can be found for the events with the intensity of 10-20 379 380 mm and lasting 1–2 hours.



Figure 10. Probability distribution of rainfall duration exceeding intensity percentile thresholds(unit: %) for (a) OBS, (b) W9k, (c) W1p5k, as well as the difference for (d) W9k–OBS, (e) W1p5k–OBS and (f) W1p5k–W9k. The threshold used here is defined in Figure6c (calculated for all the hours).

381



Figure 11. Probability distribution of rain cell size where rainfall over given percentile thresholds (unit: %). The data are interpolated to the gridswith the resolution of 10 km×10 km in the same region. Results are shown for (a) OBS, (b) W9k, (c) W1p5k, as well as the difference for (d) W9k–OBS, (e) W1p5k–OBS and (f) W1p5k–W9k. The threshold used here is defined in Figure6c (calculated for all the hours).

387



394

Figure 12. Joint probability distribution of duration versus peak intensity for rainfall  $\geq$  0.1 mmh<sup>-1</sup> (unit: %). Results are shown for (a) OBS, (b) W9k–OBS, (c) W1p5k–OBS and (d) W1p5k–W9k.

As W9k and W1p5k are able to show some different characteristics of the simulated 399 precipitation events, it is informative for us to further explore how precipitation in these 400 401 simulations is related to the convection properties in these phenomena. The convection fraction (CF) is used to distinguish these wet days with different convection contributions. The CF is 402 calculated in W9k at each station, and then the duration and characteristics of peak intensity are 403 examined at a given range of CF. As the convection scheme is switched off in W1p5k, in this 404 part of the analysis, we assume that CF in W9k is also applicable for W1p5k given the fact that 405 large-scale factors affecting convection should be largely similar in these two resolutions. 406

407 Comparing high-CF days (CF > 0.7) with low-CF days (CF < 0.1) in W1p5k, there are 408 much more short-duration high-peak-intensity rain when CF is high. High-CF days are often

associated with more rainfall events with 1-2 hours durations, while longer rainfall events can be 409 seen in other CF categories (without taking the intensity into account). Results are rather 410 different for W9k. High-CF events do not show a different pattern as to other CF category 411 eventsfor high-peak-intensity rainfall. However, longer moderate-intensity precipitation is 412 clearly presented in all the three sets of data (Figures 13 b, 13d and 13f). The differences 413 between W1p5k simulation for high CF events and that for middle CF ones in Figure 13 are 414 similar to the differences between W1p5kand W9k simulations as shown in Figure 12. This 415 suggests that the differences in the model-simulated rainfall duration and intensity characteristics 416 in these two sets of model configurations are mainly due to the improved representation of 417 convections in W1p5k. 418



Figure 13. Differences in the joint probability distribution of rain duration versus peak intensity for different CF ranges. Results are high convective fraction ( $CF \ge 0.7$ ) days minus lower level convective fraction days.

423 **4 Discussion and conclusions** 

424 Many applications of risk assessment and environmental impact assessment (EIA) require 425 detailed information on the local climate at a very fine spatial scale (Wang et al., 2013). There is

an increasing need to use high-resolution regional climate models to make more reliable 426 assessments on possible changes of small-scale and high-impact extreme weather events under 427 global warming. Our study is among the first in China to use a convection permitting model at 428 1.5km resolution to produce a 10-yr simulation so that we can properly explore the sensitivity of 429 modelling rainfall characteristics by considering the model resolutions and physics 430 representations. The knowledge obtained from this study will provide valuable insights to the 431 efforts, and increase our confidence, in providing more reliable and detailed projections of future 432 rainfall at local and regional scales under the condition of global warming. 433

This paper aims at evaluating the skill of two selected WRF high-resolution experiments (9-km and 1.5-km RCMs) in simulating precipitation extremes over a region in eastern China for the 10-yr period of 2005 to 2014. We have been focused on assessing whether there are added values in the convection-permitting configuration of W1p5k for simulating precipitation extremes. Our results show that the 1.5-km convection-permitting RCM produces more realistic extremes than the W9k.

In the simulations, we found that the incorporation of explicit convection parametrization 440 produces not only smaller wet bias of mean precipitation and a more realistic rainfall diurnal 441 cycle, but alsomore accurate short-duration individual rainfall events which are more consistent 442 with the observations (Chang et al., 2018). W1p5k has a more realistic performance for the 443 precipitation duration simulation, including the proportion of precipitation to total precipitation 444 with different intensities or durations. The observations show that there are many short-duration 445 (1-2 hours) heavy precipitation events, and in a continuous precipitation event precipitation 446 distribution has a significant peak. The short-duration (1-2 hours) precipitation and its peak can 447 be better represented in the W1p5k simulation. In contrast, the heavy-precipitation duration 448 simulated by W9k is longer (2–4 hours) than the observed, and the simulated precipitation tends 449 to be more evenly distributed than the observations during the rainfall period. The average 450 temporal and spatial characteristics of convections can be simulated, but individual rainfall 451 events are not well represented. 452

453 Comparing with W9k, W1p5k also has a better performance on the spatial coverage of 454 precipitation. As the convective parameterization scheme is used for dedcribing the average 455 properties of convections over a grid cell (Kendon et al., 2012), it has a limitation in representing

extreme precipitation or concentrated precipitation. Our results show that the precipitation in 456 W9k tend to have a spatially more widespread and temporally long-lasting tendency, while the 457 model deficiency is much improved in W1p5k. The convection-permitting simulation can 458 represent many important processes that control convections, including local dynamics and 459 topographic forcing (such as convergence lines) that lead to the triggering and inhibition of 460 convections, as well as the environmental mixing (Roberts, 2007, cited in Kendon et al., 2012). 461 Therefore, the CPM can represent the spatial-temporal structure of rainfall well. However, the 462 463 convection is still not properly resolved at the kilometer scale, and the grid spacing is still too coarse to resolve updrafts (Prein et al., 2015). This means that the updraft of some showers will 464 appear on wrong scales, as the turbulent mixing is inherently insufficient. This explains the trend 465 of the vertical velocity, so the intensity of convective showers is too strong in some cases 466 467 simulated by the kilometer-scale model (Lean et al., 2008).

Our results showed that W9k produces stronger extreme rainfall than W1p5k, as the 468 precipitation simulated by W9k is overestimated to a large extent. Generally, precipitation 469 intensity by coarse-resolution RCMs may be somewhat underestimated (Li et al., 2018; Kendon 470 et al., 2012), which is mainly due to the overestimation of the frequency of light rains. The high 471 intensity presented by W9k in this study is due to the overestimation of rainfall, especially in 472 summer. This could be attributed to the fact that the resolution of 9 km is finer than that of 473 common large-scale models. Previous studies (e.g., Xiong, 2013) found that for the rainfall 474 simulation in the upstream area of the Heihe River Basin in China with concentrated rainfall 475 from May to September at the resolution of 3 km, both the convective parameterization scheme 476 and the convection-permitting scheme overestimate the precipitation. There is a possibility that 477 the 9-km model with the convective parameterization scheme leads to the overestimation of 478 precipitation. As the CPM is more sensitive to produce convective precipitation, theoretically 479 there should be more precipitation produced in CPM, while less precipitation in coarse-480 resolution RCMs. Such biases in the model may be related to the uncertainty of the study domain, 481 the model or the observation (Isotta et al., 2014). These aspects of the model performance will be 482 further explored in our future analysis based on our 10-yr simulations. 483

This study is the first decadal long continuous convection-permitting simulations at 1.5 km resolution over eastern China. And it is also a first attempt to look at the added value of CPMs in terms of the temporal and spatial structure of hourly precipitation over eastern China.

Results show that the 1.5-km RCM tend toslightly overestimate the daily rainfall intensity, 487 however, it presents a much better agreement in terms of hourly rainfall frequency and intensity, 488 and also shows better agreements with observations in terms of temporal and spatial structure of 489 hourly rainfall. In the 9-km RCM, the short-duration high peak intensity events are insufficient 490 representation overall, and rainfall tends to be too widespread and persistent. The bias is 491 remarkably reduced by the 1.5 km CPM due to its better account of physical processes related 492 with convection. We note, however, the grid spacing is still too coarse to resolve updrafts that 493 are narrower than several kilometers (Murata et al., 2017; Lean et al., 2008), which suggests 494 higher resolution and more accurate microphysical parameterizations in future. Meanwhile, 495 short-duration high peak intensity events are usually related to mesoscale convective system 496 (MCS) in Eastern China (Luo and Chen, 2015; Luo et al., 2016; He et al., 2017). Whether the 497 498 better representation of spatial and temporal structure of hourly rainfall of 1.5-km RCM is caused by better representation of MCS also need be further investigated. 499

#### 500 Acknowledgments

The authors would like to thank the sponsors of this work: the National Key R&D Program of China (Grant no. 2019YFE0124800), National Natural Science Foundation of China (51761135024 and 41730959), the High-level Special Funding of the Southern University of Science and Technology (Grant No. G02296302, G02296402), and also Social Development Projects of STCSM (19DZ1201402 and 19DZ1201500).

#### 506 Data Availability Statement

507 The WRF simulation data in this study is publicly available at https://osf.io/nfqt2/files/

508

#### 509 **References**

510 Arnbjerg-Nielsen, K., Willems, P., Olsson, J., Beecham, S., Pathirana, A., Bülow Gregersen,

- I., . . . Nguyen, V.-T.-V. (2013), Impacts of climate change on rainfall extremes and urban
  drainage systems: a review. *Water Science and Technology*, 68(1), 16-28.
- Ban, N., Schmidli, J., & Schär, C. (2014), Evaluation of the convection-resolving regional
  climate modeling approach in decade-long simulations. *Journal of Geophysical Research:*

- 515 *Atmospheres*, *119*(13), 7889-7907.
- Ban, N., Schmidli, J., & Schär, C. (2015), Heavy precipitation in a changing climate: Does
  short-term summer precipitation increase faster? *Geophysical Research Letters*, 42(4),
  1165-1172.
- Berthou, S., Rowell, D. P., Kendon, E. J., Roberts, M. J., Stratton, R. A., Crook, J. A., & Wilcox,
   C. (2019), Improved climatological precipitation characteristics over West Africa at
   convection-permitting scales. *Climate Dynamics*, *53*, 1991-2011.
- Boberg, F., Berg, P., Thejll, P., Gutowski, W. J., & Christensen, J. H. (2009), Improved
  confidence in climate change projections of precipitation evaluated using daily statistics
  from the prudence ensemble. *Climate Dynamics*, *32*(7-8), 1097-1106.
- Brisson, E., Van Weverberg, K., Demuzere, M., Devis, A., Saeed, S., Stengel, M., & van Lipzig,
  N. P. (2016), How well can a convection-permitting climate model reproduce decadal
  statistics of precipitation, temperature and cloud characteristics? *Climate Dynamics*,
  47(9-10), 3043-3061.
- Chan, S. C., Kendon, E. J., Fowler, H. J., Blenkinsop, S., Roberts, N. M., & Ferro, C. A. (2014),
  The value of high-resolution Met Office regional climate models in the simulation of
  multihourly precipitation extremes. *Journal of Climate*, 27(16), 6155-6174.
- Chang, W., Wang, J., Marohnic, J., Kotamarthi, V. R., & Moyer, E. J. (2018), Diagnosing added
  value of convection-permitting regional models using precipitation event identification
  and tracking. *Climate Dynamics*, 55, 175-192.
- Chen, F., & Dudhia, J. (2001), Coupling and advanced land surface-hydrology model with the
   Penn State-NCAR MM5 modeling system. Part I: model implementation and sensitivity.
   *Monthly Weather Review, 129*, 569–585
- Chen, X., Zhang, F., Zhao, K. (2016), Diurnal variations of land/sea breeze and its related
  precipitation over South China. *Journal of the Atmospheric Sciences*, 73(12).
  https://doi.org/10.1175/JAS-D-16-0106.1
- Donner, L. J., Wyman, B. L., Hemler, R. S., Horowitz, L. W., Ming, Y., Zhao, M., . . .
  Schwarzkopf, M. D. (2011), The dynamical core, physical parameterizations, and basic
  simulation characteristics of the atmospheric component AM3 of the GFDL global
  coupled model CM3. *Journal of Climate*, 24(13), 3484-3519.
- 545 Feser, F., Rockel, B., von Storch, H., Winterfeldt, J., & Zahn, M. (2011), Regional climate

- models add value to global model data: a review and selected examples. *Bulletin of the American Meteorological Society*, 92(9), 1181-1192.
- Fosser, G., Khodayar, S., & Berg, P. (2015), Benefit of convection permitting climate model
  simulations in the representation of convective precipitation. *Climate Dynamics*, 44(1-2),
  45-60.
- Fowler, H. J., and M. Ekstrom. (2009), Multi-model ensemble estimates of climate change
   impacts on UK seasonal precipitation extremes. *International Journal of Climatology*, 29,
   385–416.
- He, Z., Zhang, Q., Bai, L., et al. (2016), Characteristics of mesoscale convective systems in
   centralEast China and their reliance on atmospheric circulation patterns. *International Journal of Climatology*, *37*(7), 3276-3290.
- Guo, Z., Fang, J., Sun, X., Tang, J., Yang, Y., & Tang, J. (2019a), Decadal long convection permitting regional climate simulations over eastern China: evaluation of diurnal cycle of
   precipitation. *Climate Dynamics*, 54(3), 1329-1349.
- Guo, Z., Fang, J., Sun, X., Yang, Y., & Tang, J. (2019b), Sensitivity of Summer Precipitation
  Simulation to Microphysics Parameterization Over Eastern China: Convection-Permitting
  Regional Climate Simulation. *Journal of Geophysical Research: Atmospheres, 124*(16),
  9183-9204.
- Helsen, S., van Lipzig, N. P., Demuzere, M., Broucke, S. V., Caluwaerts, S., De Cruz, L., . . . Van
  Schaeybroeck, B. (2019), Consistent scale-dependency of future increases in hourly
  extreme precipitation in two convection-permitting climate models. *Climate Dynamics*,
  54, 1267-1280.
- Hong, S.Y., Dudhia, J., & Chen, S.H. (2004), A revised approach to ice microphysical processes
  for the bulk parameterization of clouds and precipitation. *Monthly Weather Review*, *132*(1), 103-120.
- Iacono, M. J., Delamere, J. S., Mlawer, E. J., Shephard, M. W., Clough, S. A., and Collins, W. D.
  (2008), Radiative forcing by long-lived greenhouse gases: Calculations with the AER
  radiative transfer models, *Journal of Geophysical Research Atmospheres*, *113*, D13103
- Isotta, F. A., Frei, C., Weilguni, V., Perčec Tadić, M., Lassegues, P., Rudolf, B., ... Ratto, S. M.
  (2014), The climate of daily precipitation in the Alps: development and analysis of a
  high-resolution grid dataset from pan-Alpine rain-gauge data. *International Journal of*

- 577 *Climatology*, *34*(5), 1657-1675.
- Kain JS. (2004), The Kain–Fritsch convective parameterization: an update. *Journal of Applied Meteorology*, 43(1), 170–181. https://doi.org/10.1175/1520-0450(2004)043<0170:TKCPAU>2.0.CO;2.
- Kendon, E. J., Roberts, N. M., Senior, C. A., & Roberts, M. J. (2012), Realism of rainfall in a
  very high-resolution regional climate model. *Journal of Climate*, 25(17), 5791-5806.
- Kendon, E. J., Roberts, N. M., Fowler, H. J., Roberts, M. J., Chan, S. C., & Senior, C. A. (2014),
  Heavier summer downpours with climate change revealed by weather forecast resolution
  model. *Nature Climate Change*, 4(7), 570.
- Kendon, E. J., Stratton, R. A., Tucker, S., Marsham, J. H., Berthou, S., Rowell, D. P., & Senior, C.
   A. (2019), Enhanced future changes in wet and dry extremes over Africa at convection permitting scale. *Nature communications*, *10*(1), 1794.
- Knist, S., Goergen, K., & Simmer, C. (2018), Evaluation and projected changes of precipitation
   statistics in convection-permitting WRF climate simulations over Central Europe.
   *Climate Dynamics*, 55, 325-341.
- Lean, H. W., Clark, P. A., Dixon, M., Roberts, N. M., Fitch, A., Forbes, R., & Halliwell, C.
  (2008), Characteristics of high-resolution versions of the Met Office Unified Model for
  forecasting convection over the United Kingdom. *Monthly Weather Review*, 136(9),
  3408-3424.
- Lucas-Picher, P., Laprise, R., & Winger, K. (2017), Evidence of added value in North American
   regional climate model hindcast simulations using ever-increasing horizontal
   resolutions. *Climate Dynamics*, 48(7-8), 2611-2633.
- Luo, Y., Chen, Y. (2015), Investigation of the predictability andphysical mechanisms of an
  extreme-rainfall-producing mesoscaleconvective system along the Meiyu front in East
  China:an ensemble approach. *Journal of Geophysical Research: Atmospheres, 120*(20),
  10593–10618. https://doi.org/10.1002/2015jd023584
- Luo, Y., Wu, M., Ren, F., Li, J., Wong, W. (2016), Synoptic situations of extreme hourly
   precipitationover China. *J Climate*, 29(24), 8703–8719. https://doi.org/10.1175/jcli-d-16 0057.1
- Li, P., Furtado, K., Zhou, T., Chen, H., Li, J., Guo, Z., & Xiao, C. (2018), The diurnal cycle of East Asian summer monsoon precipitation simulated by the Met Office Unified Model at

608 convection-permitting scales. *Climate Dynamics*, 55, 131-151.

- Murata, A., Sasaki, H., Kawase, H., Nosaka, M. (2017), Evaluation of precipitation over an
   oceanic region of Japan in convection-permitting regional climate model simulations.
   *Climate dynamics*, 48, 1779-1792. https://doi.org/10.1007/s00382-016-3172-x
- Prein, A. F., Langhans, W., Fosser, G., Ferrone, A., Ban, N., Goergen, K., . . . Feser, F. (2015), A
  review on regional convection-permitting climate modeling: Demonstrations, prospects,
  and challenges. *Reviews of Geophysics*, *53*(2), 323-361.
- Qian, T., Zhao, P., Zhang, F., Bao, X. (2015), Rainy-season precipitation over the Sichuan basin
  and adjacent regions in southwestern China. *Monthly Weather Review*, *143*(1), 383–394
- Roberts, N. (2007), Meteorological components in forecasts of extreme convective rainfall using
   12-km and 1-km NWP models: A tale of two storms. *Met Office Technical Report no. 520.*
- Stratton, R. A., Senior, C. A., Vosper, S. B., Folwell, S. S., Boutle, I. A., Earnshaw, P. D., ...
  Manners, J. (2018), A pan-African convection-permitting regional climate simulation
  with the Met Office Unified Model: CP4-Africa. *Journal of Climate*, *31*(9), 3485-3508.
- Wang, C. (2013), Using an ultrahigh-resolution regional climate model to predict local
  climatology. *Quarterly Journal of the Royal Meteorological Society*, *139*(677), 19641976.
- Xiong, Z., & Yan, X. (2013), Building a high-resolution regional climate model for the Heihe
  River Basin and simulating precipitation over this region. *Chinese Science Bulletin*,
  58(36), 4670-4678.
- Yang, Q., Houze Jr, R. A., Leung, L. R., & Feng, Z. (2017), Environments of long-lived
   mesoscale convective systems over the Central United States in convection permitting
   climate simulations. *Journal of Geophysical Research: Atmospheres*, *122*(24), 13-288.
- Yun, Y., Liu, C., Luo, Y., Liang, X., Huang, L., Chen, F., &Rasmmusen, R. (2019), Convection permitting regional climate simulation of warm-season precipitation over Eastern China.
   *Climate Dynamics*, 54, 1469-1489.
- 634
- 635
- 636



[Journal of Geophysical Research - Atmospheres]

Supporting Information for

## [Evaluation of extreme precipitation in the Yangtze River Delta Region of China using a 1.5 km mesh convection-permitting regional climate model]

[Guangtao Dong<sup>1,2</sup>, Zhiyu Jiang<sup>3</sup>, Ya Wang<sup>4</sup>, Zhan Tian<sup>5\*</sup>, Junguo Liu<sup>5</sup>]

[1 Shanghai Climate Center, Shanghai, 200030, China

<sup>2</sup> Key Laboratory of Cities' Mitigation and Adaptation to Climate Change in Shanghai, China Meteorological Administration, Shanghai, 200030, China

<sup>3</sup> School of Atmospheric Sciences, Nanjing University, Nanjing, China, 201400, China

<sup>4</sup> Shanghai Meteorological Disaster Prevention Technology Center, Shanghai, 200030, China

<sup>5</sup> School of Environmental Science and Engineering, Southern University of Science and Technology, Shenzhen, 518055, China]

## **Contents of this file**

Figures S1

## Introduction

In order to investigate the reason leading to the overestimation of JJA precipitation by 9km RCM, we examine the ability of 9-km RCM to reproduce JJA large-scale atmospheric circulations in this Supplementary Material.

Fig. S1 shows the spatial distributions of the ten-year (2005–2014) averaged JJA largescale atmospheric circulations, including 500 hPa geopotential height, 850 hPa wind field, specific humidity and the moisture flux divergence. As a dominant factor controlling the summer precipitation over the eastern China, the observed western Pacific subtropical high in boreal summer at 500 hPa dominates the southeastern China (Fig. S1a). However, the WRF lacks the skill for realistically simulating the position and intensity of the western Pacific subtropical high which leads to much of the moisture transport concentrates near Zhejiang province (Fig. S1b). The intensities of 850-hPa wind speed are also overestimated to some extent, which could be responsible for the overestimation of the summer precipitation over the eastern China (Fig. S1d).



**Figure S1.** The averaged summer-mean moisture flux divergence (shaded, unit:  $10^{-7}$  g cm<sup>-2</sup> hPa<sup>-1</sup> s<sup>-1</sup>) and 500-hPa geopotential height (contour, unit: dagpm) from (a) ERA-Interim reanalysis data and (b) W9k simulation for the period of 2005-2014, with summer-mean 850-hPa wind filed (vector, unit: m s<sup>-1</sup>), specific humidity (shaded, unit: kg kg<sup>-1</sup>) and 200-hPa geopotential height (contour, unit: dagpm) from (c) ERA-Interim reanalysis data and (d) W9k simulation. The spatial correlation of the mean moisture flux divergence between the model simulation and the observation is 0.18, and that of the specific humidity is 0.98. The RMSE of mean moisture flux divergence is 0.59, and that of the specific humidity is 2.6 g/kg.



[Journal of Geophysical Research - Atmospheres]

Supporting Information for

## [Evaluation of extreme precipitation in Yangtze River Delta Region of China using a 1.5 km mesh convection-permitting regional climate model]

[Guangtao Dong<sup>1,2</sup>, Zhiyu Jiang<sup>3</sup>, Ya Wang<sup>4</sup>, Zhan Tian<sup>5\*</sup>]

[<sup>1</sup> Shanghai Climate Center, Shanghai, 200030, China

<sup>2</sup> Key Laboratory of Cities' Mitigation and Adaptation to Climate Change in Shanghai, China Meteorological Administration, Shanghai, 200030, China

<sup>3</sup> School of Atmospheric Sciences, Nanjing University, Nanjing, China, 201400, China

<sup>4</sup> Shanghai meteorological disaster prevention technology center, Shanghai, 200030, China

<sup>5</sup> School of Environmental Science and Engineering, Southern University of Science and Technology, Shenzhen, 518055, China]

## **Contents of this file**

Figures S1

## Introduction

In order to investigate the reason leading to the overestimation of JJA precipitation by 9km RCM, we examine the ability of 9-km RCM to reproduce JJA large-scale atmospheric circulations in this Supplementary Material.

Fig. S1 shows the spatial distributions of the ten-year (2005–2014) averaged JJA largescale atmospheric circulations, including 500 hPa geopotential height, 850 hPa wind field, specific humidity and the moisture flux divergence. As a dominant factor controlling the summer precipitation over the eastern China, the observed western Pacific subtropical high in boreal summer at 500 hPa dominates the southeastern China (Fig. S1a). However, the WRF lacks the skill for realistically simulating the position and intensity of the western Pacific subtropical high which leads to much of the moisture transport concentrates near Zhejiang province (Fig. S1b). The intensities of 850-hPa wind speed are also overestimated to some extent, which could be responsible for the overestimation of the summer precipitation over the eastern China (Fig. S1d).



**Figure S1.** The averaged summer-mean moisture flux divergence (shaded, unit:  $10^{-7}$  g cm<sup>-2</sup> hPa<sup>-1</sup> s<sup>-1</sup>) and 500-hPa geopotential height (contour, unit: dagpm) from (a) ERA-Interim reanalysis data and (b) W9k simulation for the period of 2005-2014, with summer-mean 850-hPa wind filed (vector, unit: m s<sup>-1</sup>), specific humidity (shaded, unit: kg kg<sup>-1</sup>) and 200-hPa geopotential height (contour, unit: dagpm) from (c) ERA-Interim reanalysis data and (d) W9k simulation. The spatial correlation of the mean moisture flux divergence between the model simulation and the observation is 0.18, and that of the specific humidity is 0.98. The RMSE of mean moisture flux divergence is 0.59, and that of the specific humidity is 2.6 g/kg.