

The 2020 Record-Breaking Mei-yu in the Yangtze River Valley of China: The Role of Anthropogenic Forcing and Atmospheric Circulation

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Anthropogenic forcing has likely reduced the probability of the 2020 June–July mei-yu in the Yangtze River Valley by approximately 54%; however, the anomalous circulation of 2020 favored the occurrence of heavy precipitation.

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The Chinese mei-yu is a period of intense precipitation during the East Asian rainy season, and typically provides the majority of June–July precipitation in the middle to lower Yangtze River Valley (YRV) (Ding and Chan 2005). In 2020, the YRV experienced a record-breaking mei-yu, characterized by a heavy and persistent precipitation, a wide meridional rain belt, and frequent heavy rainstorms (Ding et al. 2021). The June-to-July total precipitation amount in the mei-yu region was about 672 mm, 99% more than the 1961–90 average and the highest on record since 1960 in the region. The mei-yu lasted for 62 days, with 74 stations experiencing record-breaking total precipitation amounts and over 600 rivers across China exceeding their warning water levels (CMA 2021). The resulting disasters, including flooding and landslides, affected more than 30 million people and 3.58 million hectares of crops in the YRV, resulting in a direct economic loss about 132.2 billion Chinese Yuan (CMA 2022).

A few studies have investigated the circulation pattern of the 2020 mei-yu and the possible drivers of it. The anomalous persistent western North Pacific subtropical high (WNPSH) and the southwesterly jet on its western side was found to be the dominant circulation

pattern during this event, which was caused by both La Niña-like sea surface temperature (SST) anomaly in the equatorial Pacific and warm SST anomaly in the tropical Indian Ocean (Liu and Ding 2020; Ding et al. 2021; Pan et al. 2021). The Tibetan Plateau vortices (TPVs) may be another possible contributor by transforming the positive vorticity into troughs and moving eastward to modulate the precipitation in the YRV (L. Li et al. 2021). These studies suggest the important role of natural factors internal to the climate system including SST variation in the 2020 mei-yu. However, the role of anthropogenic external forcing in the summer 2020 mei-yu has not been extensively evaluated. Previous studies (Li et al. 2018; Zhou et al. 2018) have shown that the strong El Niño of 2015/16 may account for the extreme precipitation event in the YRV during 2016 summer while anthropogenic effects can be only found on smaller spatial scales. A few studies (Burke and Stott 2017; Lu et al. 2020; X. Li et al. 2021) proposed that human-induced warming and natural variation played important roles in the changes in intensity and frequency of extreme precipitation in the YRV. Here, we investigate how externally forced factors and synoptic circulation conditions may have affected the 2020 mei-yu event based on simulations from phase 6 of the Coupled Model Intercomparison Project (CMIP6; Eyring et al. 2016).

Data and methods.

The mei-yu region (28° – 34° N, 110° – 122.5° E; black box in Fig. 1a) defined by the China Meteorological Administration (CMA 2017) was selected as the key study region. Daily precipitation data at 414 stations in the region for 1960–2020 were extracted from a homogenized observational dataset developed by the China National Meteorological Information Center (Yang and Li 2014). Two metrics, including percentage precipitation anomaly (PPA) and maximum five-consecutive-day precipitation anomaly (RX5day, relative to the 1961–90 average) from June to July were used to describe different characteristics of 2020 mei-yu. The observed precipitation and RX5day at each station were first averaged onto the grid boxes of $2^{\circ} \times 2^{\circ}$ and then averaged to obtain the area-weighted regional mean precipitation in the mei-yu region. The PPA was then calculated by the anomaly of regional mean precipitation divided by the 1961–90 average. The NCEP–NCAR reanalysis data (Kalnay et al. 1996) were used for the investigation of atmospheric circulation changes, including 500-hPa geopotential height (Z500) and 850-hPa zonal and meridional winds (UV850).

For the model results, PPA and RX5day were calculated similar to observation based on 53 runs of nine CMIP6 models under historical (natural + anthropogenic) forcings (ALL) and 49 runs under natural-only (NAT) forcing, respectively (see Table S1 in the online supplemental material). We extended the historical ALL forcing runs to 2030 with the Shared Socioeconomic Pathway SSP2–4.5 scenario (2015–30). The SSTs and atmospheric and land conditions of CMIP simulations are not synchronous with observations, so using CMIP simulations to evaluate a specific year is not possible. Thus, we selected the period of 2001–20 as present-day climate. We did not select the period 2011–30 centered on the year of the event, as the NAT-forcing run ends in 2020. This slightly affects our attribution results. Given the small recent trends in precipitation in the mei-yu region (Ding et al. 2020), the recent two decades (2001–20) reasonably represent current climatic conditions, as also assumed in other studies (Christidis and Stott 2015).

To estimate the anthropogenic influence on the 2020 mei-yu, we calculated the probabilities of events exceeding the threshold of 2020-like mei-yu for the ALL and NAT forcing experiments, that is, both with (P_{ALL}) and without (P_{NAT}) the effect of anthropogenic influence. The 2020 thresholds for both PPA (99.6%) and Rx5day (71.2 mm) were first adjusted by the method used in Sun et al. (2018) because the limitations of current generation of climate models to reproduce the very heavy precipitation (see details in the online supplemental material). The adjusted thresholds for PPA and Rx5day are 51.2% and 57.1 mm respectively. The kernel

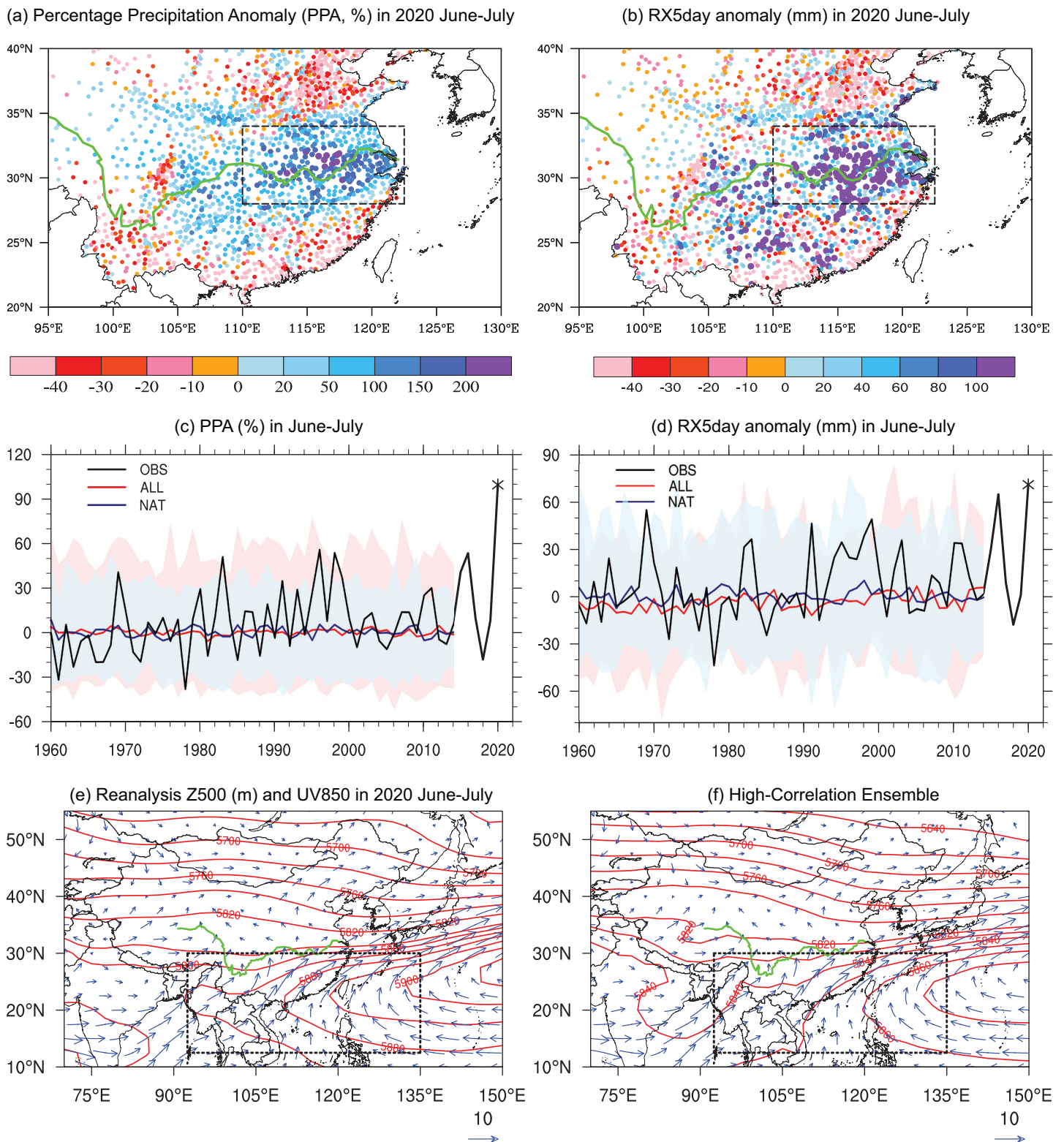


Fig. 1. (a) Percentage precipitation anomaly (PPA; %) and (b) RX5day anomaly (mm) in June–July for observations in 2020. (c),(d) Regional PPA (%) and RX5day anomaly (mm) in June–July for observations (black), ALL simulations (red), and NAT simulations (blue) for 1960–2014. Thick lines denote ensemble mean, and shading denotes the 5%–95% ranges of the individual model simulations. (e) Geopotential height (red contours; gpm) at 500 hPa and winds (blue vectors; m s⁻¹) at 850 hPa in June–July of 2020 based on reanalysis data. (f) Z500 (gpm) and UV850 (m s⁻¹) in June–July of 2001–20 based on the mean of June–July extracted from the CMIP6 results with the ALL experiment, for which the circulation pattern correlates well (coefficient greater than 0.7) with the 2020 reanalysis pattern over the region marked by the black box. The green line in (e) and (f) indicates the Yangtze River.

density estimation (KDE) was used to estimate the probability density distribution, which has been proved to be suitable to estimate the probability density functions (PDFs) of precipitation event (Ma et al. 2017; Lu et al. 2021). The probability ratio (PR) was then defined as $PR_{Anthro} = P_{ALL}/P_{NAT}$. Uncertainties in PR were obtained using 1000 bootstraps, with PR computed for each bootstrap realization (Christidis et al. 2015).

To estimate the effects from atmospheric circulation, the method suggested by Christidis and Stott (2015) was used. We first selected the key circulation region that dominate this mei-yu event in the observation (i.e., the WNPSH region as shown in Fig. 1e). The anomalous WNPSH has been confirmed as the key circulation pattern driving the 2020 mei-yu, and reflects the joint influence from the anomalous SSTs in the equatorial Pacific and Indian Ocean (Pan et al. 2021). Then we partitioned simulated June–July Z500 from all models and years (2001–20) under ALL forcing into two groups (Table S2): one that is highly correlated (correlation coefficients above 0.7) with the observed Z500 in 2020 over the key WNPSH region, and one that is less correlated (correlation coefficients below 0.7) with observed Z500. For these two high-correlation and low-correlation groups, the probabilities of events exceeding the 2020-like threshold value (P_{High} and P_{Low}) were calculated respectively based on KDE. The probability ratio for the circulation effects was defined as $PR_{Circ} = P_{High}/P_{Low}$. Uncertainties in PR was estimated as those used in the analysis for the anthropogenic influence. We also conducted sensitivity test for different coefficients from 0.6 to 0.8, and found negligible influence on the final results (Fig. S2a).

Results.

Figure 1a shows that the observed June–July positive PPA was centered in the YRV. In the mei-yu region, the PPA in most stations was more than 50% and 74 stations experienced the record-breaking precipitation. The RX5day anomaly at nearly half of the stations was greater than 100 mm, with the maximum value reaching 435.5 mm. Figures 1c and 1d show the temporal evolutions of observed and model simulated PPA and RX5day anomalies over the mei-yu region from June to July. It is apparent that June–July 2020 was the wettest on record since 1960, with PPA and RX5day both reaching the highest values. This extreme precipitation was associated with the low-level southwesterly winds on the western side of the WNPSH, which continuously transport water vapor into the YRV (Fig. 1e). Figure 1f displays the model ensemble mean of Z500 and UV850 in June–July from the high-correlation group during the current climate state of 2001–20 under ALL experiments. The extended WNPSH and the obvious southwesterly flow on its western side, which is similar to the observed circulation pattern, provide favorable conditions for heavy precipitation events.

Figures S1a and S1b in the online supplemental material show the histogram and KDE estimate of the probability distribution of the observed and simulated June–July PPA and RX5day anomalies during 1960–2014. The models reasonably simulate the probability distribution for PPA and RX5day over the mei-yu region, with quite similar shapes between observation and models. Figures S1c and S1d show the p values from Kolmogorov–Smirnov tests that compare the observed distribution with multimodel ensemble and individual models. The agreement remains good for both PPA and RX5day when models are tested individually. This suggests that the models can be used to conduct the attribution analyses.

To investigate human influence on the mei-yu event, we compared the CMIP6 model experiments under ALL and NAT forcings. For PPA, the PDF shift toward a drier condition under ALL forcing relative to NAT forcing (Fig. 2a), suggesting a decrease of probability of such persistent heavy precipitation events over the YRV due to human influence. The probability of the 2020-like event defined by PPA (51.2%) is around 1.78% (90% confidence interval: 1.28%–2.38%) in the ALL ensemble, while in the NAT ensemble the probability increases to 3.29% (2.56%–4.12%). This gives a probability ratio of 0.54 (0.36–0.81). For the RX5day, the

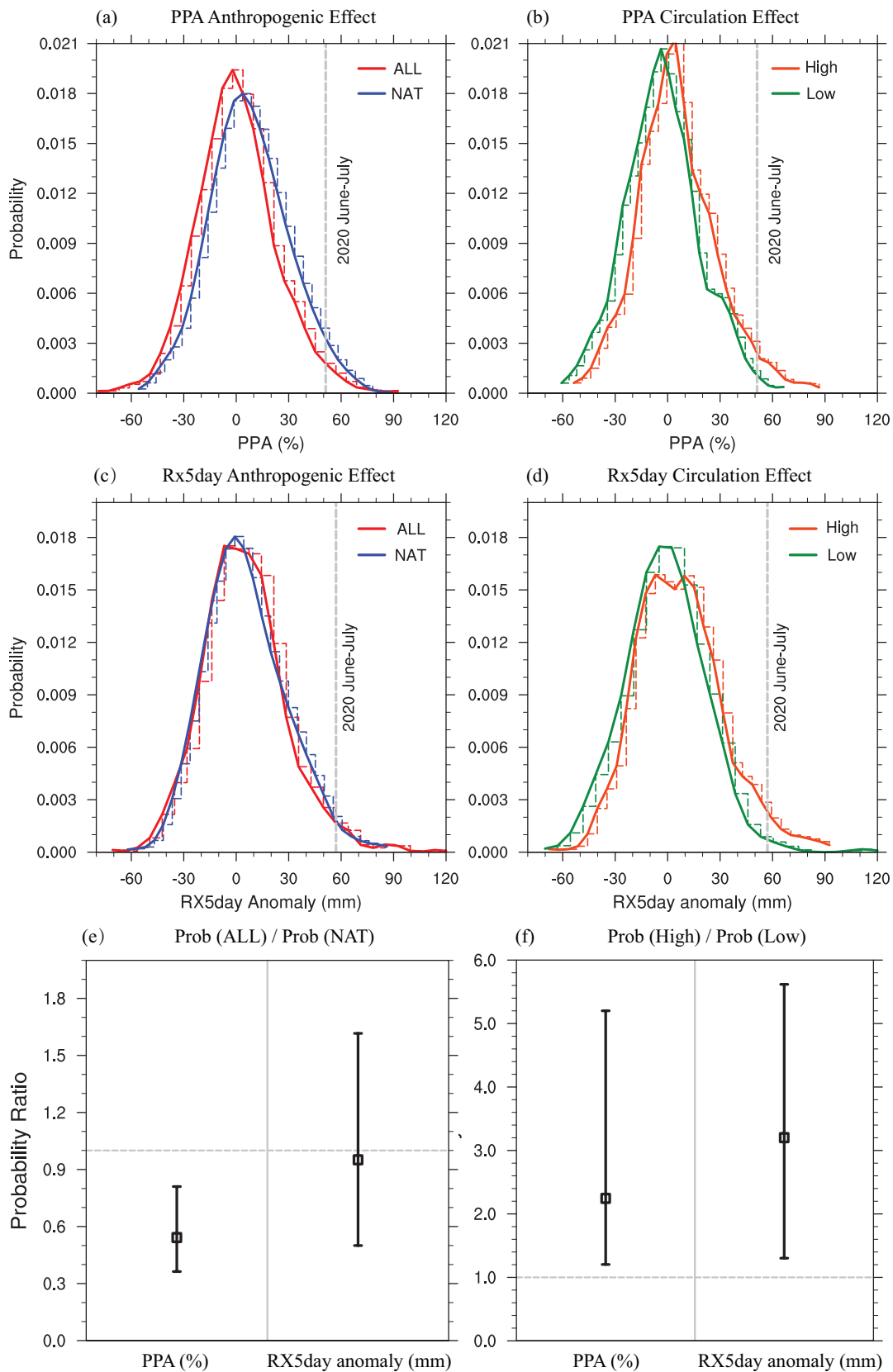


Fig. 2. The impact of the anthropogenic forcings and 2020 mei-yu circulation pattern on the 2020-like mei-yu, showing the June–July (a) PPA and (c) RX5day anomaly distributions from model experiments with (red contours) and without anthropogenic forcings (blue contours); the June–July (b) PPA and (d) RX5day anomaly distributions in the high-correlation ensemble (orange contours) and low-correlation ensemble (green contours) based on CMIP6 ALL results; and (e) the changes in the probability due to anthropogenic forcings and (f) the changes in the occurrence probability of 2020-like mei-yu of under the influence of a 2020-like circulation pattern. Best estimates of the change in the probability are marked by the square symbols and the 5%–95% uncertainty range by the vertical whiskers.

PDF shows almost no change for both experiments. The probability ratio for the 2020-like RX5day anomaly (57.1 mm) is 0.95 (0.5–1.61), indicating little anthropogenic influence on this metric. Previous studies proposed two competing effects of anthropogenic forcing on the precipitation in East Asia. One is the effect of increased moisture caused by global warming (Trenberth et al. 2003) and the other is the cooling effect induced by the increased anthropogenic aerosols. The latter could weaken the thermal differences between land and ocean, and thus reduce the East Asian summer monsoon (EASM), decreasing the likelihood of persistent heavy precipitation (Song et al. 2014; Li et al. 2015; Rimi et al. 2019). Therefore, for the June–July PPA that mostly represents the persistent precipitation, anthropogenic forcing has likely reduced the probability of the similar events. But for the extreme rainfall (RX5day), the aerosol effects may be counteracted by the increased moisture induced by global warming, resulting in weak anthropogenic signal in quantitative analyses (Zhang et al. 2019).

We then analyzed the effect of circulation pattern on this event. We compared the PDFs of simulated precipitation between high- and low-correlation groups under ALL forcing (Figs. 2b,d). Both PPA and RX5day results show that the anomalous circulation increases the chance of heavy precipitation occurring. The PDFs of high correlation group shift toward a wetter regime compared with those for the low correlation group. The Kolmogorov–Smirnov test shows that the distributions from high- and low-correlation groups are significantly separated as the *p* values are near zero. The probability of the 2020-like PPA event is around 2.9% (with 90% confidence interval 1.7%–4.0%) in the high-correlation ensemble, while it decreases to 1.2% (0.6%–1.7%) in the low-correlation ensemble. This gives a probability ratio of 2.3 (1.2–5.2). For RX5day, the PDF shifts toward wetter regime in the high-correlation ensembles, with the occurrence probability increasing about 3.2 (1.3–5.6) times compared with that from the low-correlation group. This indicates the influence of persistent circulation on this mei-yu event. Further analyses indicate that these results are insensitive to the selection of critical area, as long as the southwesterly jet on the western side of the WNPSH is included in the area (Fig. S2b).

Conclusions.

For the 2020 record-breaking mei-yu, we analyzed both PPA and RX5day based on observations and CMIP6 models. We found that human influence tends to reduce the probability of PPA that represents the persistent precipitation, but has insignificant impacts on RX5day, which represents extreme heavy rainfall. The probability of the 2020-like persistent precipitation has decreased by about 54% under anthropogenic forcing, which is consistent with previous studies based on CMIP5 models and other large-ensemble models (Zhang et al. 2019; R. Li et al. 2021). However, we should keep in mind that the adjusted threshold for this event may cause the inaccurate estimate of the probability changes. For the effects of circulation pattern, the prevailing flow pattern along the WNPSH has increased the occurrence probability of 2020-like PPA and RX5day by about 2.3 and 3.2 times, respectively. As the key circulation system for the 2020 mei-yu, the anomalous strong WNPSH was jointly affected by the SST in the equatorial Pacific and Indian Ocean, which reflects important role of the SST changes. Although the CMIP6 models generally are able to simulate the long-term warming in the Pacific and the Indian Ocean, more detailed analyses about SST influence are needed in the future. Especially, understanding different roles of anthropogenic forcing and internal fluctuation of ocean in the SST changes could help separate the contribution from the different factors to the occurrence of precipitation.

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