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Anthropogenic influence on 2019 May-June extremely low precipitation in southwestern China

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Capsule:

Anthropogenic forcing has likely increased the likelihood of the 2019 May-June severe low precipitation event in southwestern China by approximately 6 (1.4) times based on the HADGEM3-GA6 (CMIP6) simulations.

Introduction

From late April to June 2019, southwestern China experienced a severe precipitation deficit. At the peak of this event (May and June), area averaged precipitation anomaly was 42% lower than climatology and the lowest on record since 1960 in the region. Yunnan and western Sichuan were most severely affected by this disaster, where the precipitation deficit affected more than 640.1 thousand hectares of crops with rice, corn, and potatoes greatly damaged. Over 100 rivers and 180 reservoirs dried out (CMA 2020a). A severe drought that accompanied this precipitation deficit led to over 824,000 people and 566,000 head of livestock having a severe lack of drinking water, with a direct economic loss of 2.81 billion Chinese Yuan ($400 million; CMA 2020b). Therefore, it is timely to investigate the cause of this extremely low precipitation event.

In recent years, spring and summer precipitation in southwestern China have shown decreasing trends (Wang et al. 2015; Lu et al. 2020), accompanied by more frequent drought events (Xin et al. 2006; Yuan et al. 2019), which have caused great damage to the local ecology, agriculture and economy. Changes in atmospheric circulation, such

as the westward shift and intensification of western Pacific subtropical high (Yang et al. 2012) and the northward shift of the mid-latitude westerlies (Sun and Yang 2012), have been shown to contribute to the precipitation deficit. Anthropogenic influences have been found on extreme precipitation events in other parts of China (Sun et al. 2018; Zhang et al. 2019), while it is still unclear whether the attribution of human influence is detectable in precipitation deficit events in southwestern China. Thus, we have used a large ensemble of simulations to investigate the contribution of human-induced climate change on the likelihood of the severe precipitation deficit in May-June 2019 over southwestern China.

Data and Methods

The 2019 precipitation deficit event was largely confined to 20° -30° N, 96° -104° E (box in Fig. 1a) and we explored the sensitivity of our results to details of this region by varying the spatial domain. We used observations of precipitation at 180 stations in the region for 1960-2019. The station data have been rigorously quality controlled and homogenized at the China National Meteorological Information Center (Yang and Li 2014). We divided the region into multiple grid boxes of 0.56° × 0.83° (latitude vs longitude) resolution, consistent with the grid of the HadGEM3-GA6 model (see below), and averaged the station precipitation within each grid box. Both observed and simulated gridded values are area-weight averaged to obtain regional mean precipitation time-series, which are finally used to compute precipitation anomaly (PA; viz. the anomaly of the total precipitation from May to June) relative to the 1961–2010

base period. The NCEP/NCAR reanalysis data (Kalnay et al. 1996) are used to investigate the atmospheric circulation.

The Met Office Hadley Centre event attribution system is based on the atmospheric model HadGEM3-GA6 and, currently, is the highest resolution global model used in attribution studies, with 85 vertical levels and an N216 horizontal resolution of 0.56° × 0.83° (Ciavarella et al. 2018). Four ensembles sets are used: the historical experiment, a 15 member ensemble of HadGEM3-GA6 forced with observed sea-surface temperatures (SSTs), anthropogenic and natural forcings (ALL) for the period 1960-2013; the historicalNat experiment, also a 15 member ensemble but with observed SSTs having anthropogenic influences removed (Christidis et al. 2013) and natural forcings (NAT); the historicalExt experiment, a 525 member ensemble as historical but only for 2019; the historicalNatExt experiment, also a 525 member ensemble as historicalNat but for 2019. From these, the change in probability, expressed as probability ratio (PR), due to human influences was computed. Simulations from the Coupled Model Intercomparison Project Phase 6 (CMIP6, Eyring et al. 2016) were used to assess the robustness of the HadGEM3-GA6 results (Supplementary Info).

The May-June mean PA in southwestern China is used as the indicator, due to its important influence on water shortage and agricultural failure. Consecutive dry days (CDD; Zhang et al. 2011) and a gridded soil moisture observational data (Shi et al. 2011)

were also used to characterize the precipitation deficit. Circulation changes are characterized by 500-hPa geopotential height (Z500) and 850-hPa zonal and meridional winds (UV850). Subsequently, May-June mean precipitation, CDD and circulation are computed from all simulations, and anomalies are calculated relative to the 1961–2010 climatologies. The probabilities of exceeding precipitation deficit like the 2019 event in the real (P_{ALL}) and natural (P_{NAT}) world are calculated when precipitation anomalies are at or below the observed 2019 threshold. The probability ratio (PR) is defined as $PR = P_{ALL} / P_{NAT}$. Uncertainties in PR are obtained using 1000 bootstraps, with PR computed for each bootstrap realization (Christidis et al. 2015), and we show the empirical 5-95 percentile ranges. The probability density functions (PDFs) were estimated by the Kernel Density Estimation (KDE), which has been widely used to estimate the PDFs of precipitation events at monthly scales (Ma et al. 2017). We also tried other fitting methods and similar PR evaluation results were obtained (SI).

Results

Figure 1a shows that the observed May-June negative precipitation and relative soil moisture anomalies were centered in Yunnan province. In this region, the PA in most stations is less than -40 mm with many stations experiencing their record-breaking lowest precipitation. Figure 1b shows the temporal evolution of May-June PA over southwestern China based on observations and simulations. It is apparent that May-June 2019 was the driest since 1960 (with PA value at -58.14 mm), and it is a one-in-

60-year event in observations (Fig. 1c). These dry conditions were associated with abnormally high height extending from the west at 500-hPa and anomalous northerlies over Yunnan at 850-hPa (Fig. 1d). These circulation patterns lead to anomalous subsidence and reduced water vapor transport from the Indian Ocean (Feng et al. 2014), favoring a severe precipitation deficit.

The model reasonably represents the temporal evolution and probability distribution for PA over southwestern China for the period 1960–2013. In Fig. 1b, the model results under ALL and NAT forcings cover most of the observed range. Fig. 2a shows the histogram and KDE estimate of the probability distribution of the observed and simulated May-June PA. HadGEM3-GA6 produces similar distribution in the historical experiment to observations, confirmed using a two-sided Kolmogorov-Smirnov test with p values equal to 0.36. The shift of probability distribution towards a drier condition under ALL forcing with a probability ratio near 5.14 (3.33-10.50) suggests that human influences have dried southwestern China relative to the preindustrial period.

An overall mean shift of PA towards a drier condition under ALL forcing relative to NAT forcing is clearly seen in the 2019 ensemble (Fig. 2b), suggesting an increase of probability of such precipitation deficit events over southwestern China due to human influences. The probability of the 2019-like event defined by PA is around 12% (9.54%-13.92%) in the 525 samples in the historicalExt experiment, while in the

historicalNatExt ensemble the probability decreases to 2% (1.21%-2.95%). This gives a probability ratio of 6.16 (3.81-9.78). When we vary the spatial domain by reducing it by up to 3 or increasing it by up to 5 degrees from all sides, the corresponding probability ratios and their 90% confidence intervals are still greater than 1. The maximum probability ratio is observed when each boundary is expanded by 1 degree, reaching 7.52. The shift of CDD probability distribution towards longer duration under ALL forcing relative to NAT forcing (Fig. 2c) further suggests that the anthropogenic influence tends to increase the probability of long dry spells and therefore favors precipitation deficit. Previous studies indicated that the cooling effect of increased aerosols from human activities in East Asia could reduce the thermal differences between land and ocean during the late spring, which favors the formation of anomalous high pressure center in southwestern China (Kim et al. 2007; Hu and Liu 2013). Thus, we compared the PDFs of geopotential height anomaly in historicalExt and historicalNatExt simulations (Fig. 2e) and found that the Z500 over southwestern China under ALL forcing is significantly higher than that under NAT forcing. The differences in precipitation and Z500 between historicalExt and historicalNatExt (Fig. 2f) also prove this. An anomalous high height center is simulated in southwestern China, corresponding to negative anomalies of precipitation and high risk of precipitation deficit events.

In the CMIP6 simulations, the distributions of PA derived from historical and hist-nat experiments are significantly distinguished from each other for 2005-2014, as the p-value of the Kolmogorov-Smirnov test is near-zero (Fig. S1a). The distribution shifts towards a drier regime from hist-nat to historical experiments with a probability ratio near 1.4 (1.14-1.94), indicating a clear human influence for the observed precipitation deficit event. Further comparison of the historical, hist-aer and hist-GHG simulation results (Fig. S1b) shows that the 2019-like event is more frequent under anthropogenic aerosol forcing but less frequent under greenhouse gas forcing relative to hist-nat simulation. Thus, suggesting that the increased probability of low PA under historical forcing experiment relative to hist-nat forcing is due to changes in aerosols.

Conclusions

The human influence on the severe 2019 May-June precipitation deficit in southwestern China is analyzed with observational, HadGEM3-GA6 and CMIP6 model data. The results based on HadGEM3-GA6 ensembles show that the probability of extremely low precipitation in May-June similar to or more severe than the observed 2019 event has increased by about 6-fold in the ALL simulations compared to the NAT simulations. Anthropogenic influence has significantly increased the chance for the occurrence of such events through increasing the probability of anomalous high-height in southwestern China (Figs. 2f, S2). This result is robust to perturbations in the region definition. Analysis of the CMIP6 ensemble also finds an increasing risk of severe

precipitation deficit, while the smaller PR in CMIP6 also implies that the HadGEM3-GA6 model might overestimate the response to anthropogenic forcing. Compared with the observation results, the stronger drying trend in HadGEM3-GA6 historical simulations also implies this, but compared with the historicalNat results, this stronger trend has indicated an apparent signal of anthropogenic influence.

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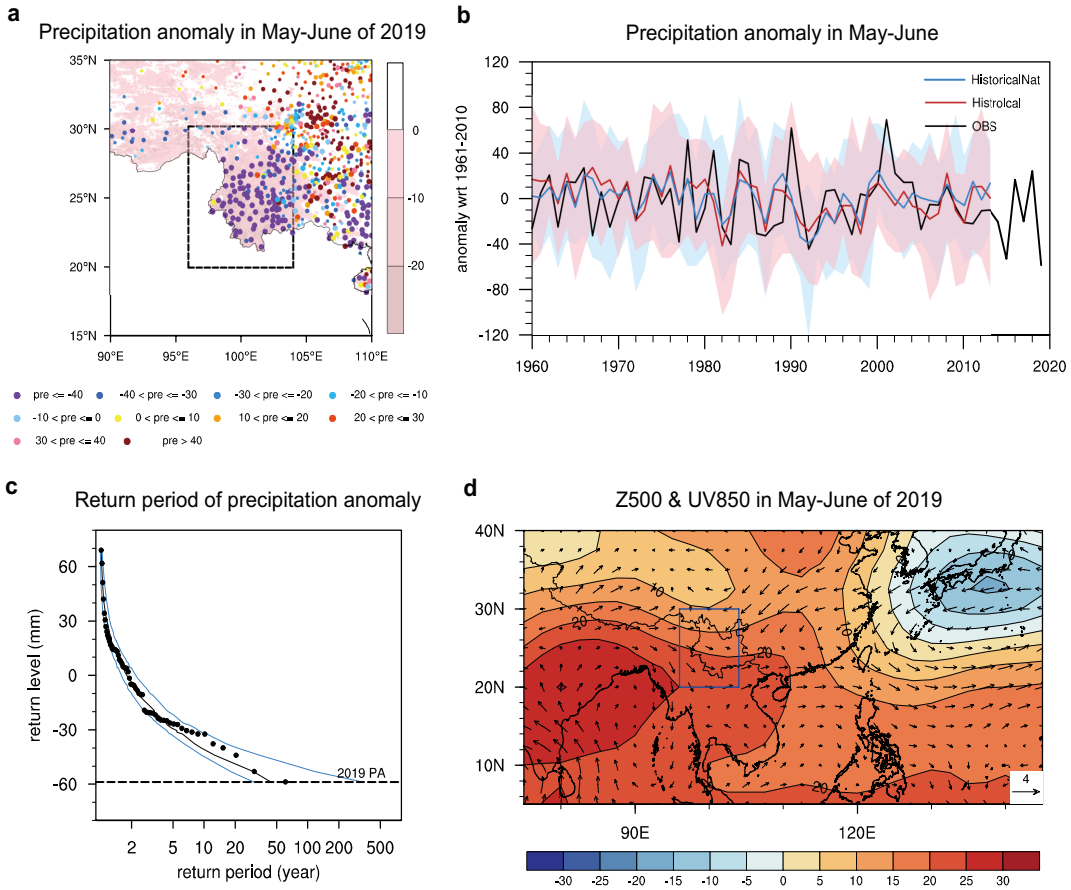

Figure 1 (a) Precipitation (mm) and relative soil moisture (%, shaded part) anomalies in May-June for observations in 2019. (b) Regional mean PA (mm) in May-June for observations (black), Historical simulations (red), and HistoricalNat simulations (blue) for 1960-2013. Thick lines denote ensemble average, and shading denotes 15-member spread. (c) Return period (black dots) of observed PA during the period of 1960-2019. The solid black line shows the results of kernel estimate and 90% confidence intervals. The dashed black line denotes the observed event in 2019. (d) Geopotential height anomaly (relative to 1961-2010) at 500hPa (contour: m) and winds anomalies (relative to 1961-2010) at 850hPa (vector: m/s) in 2019 May-June.

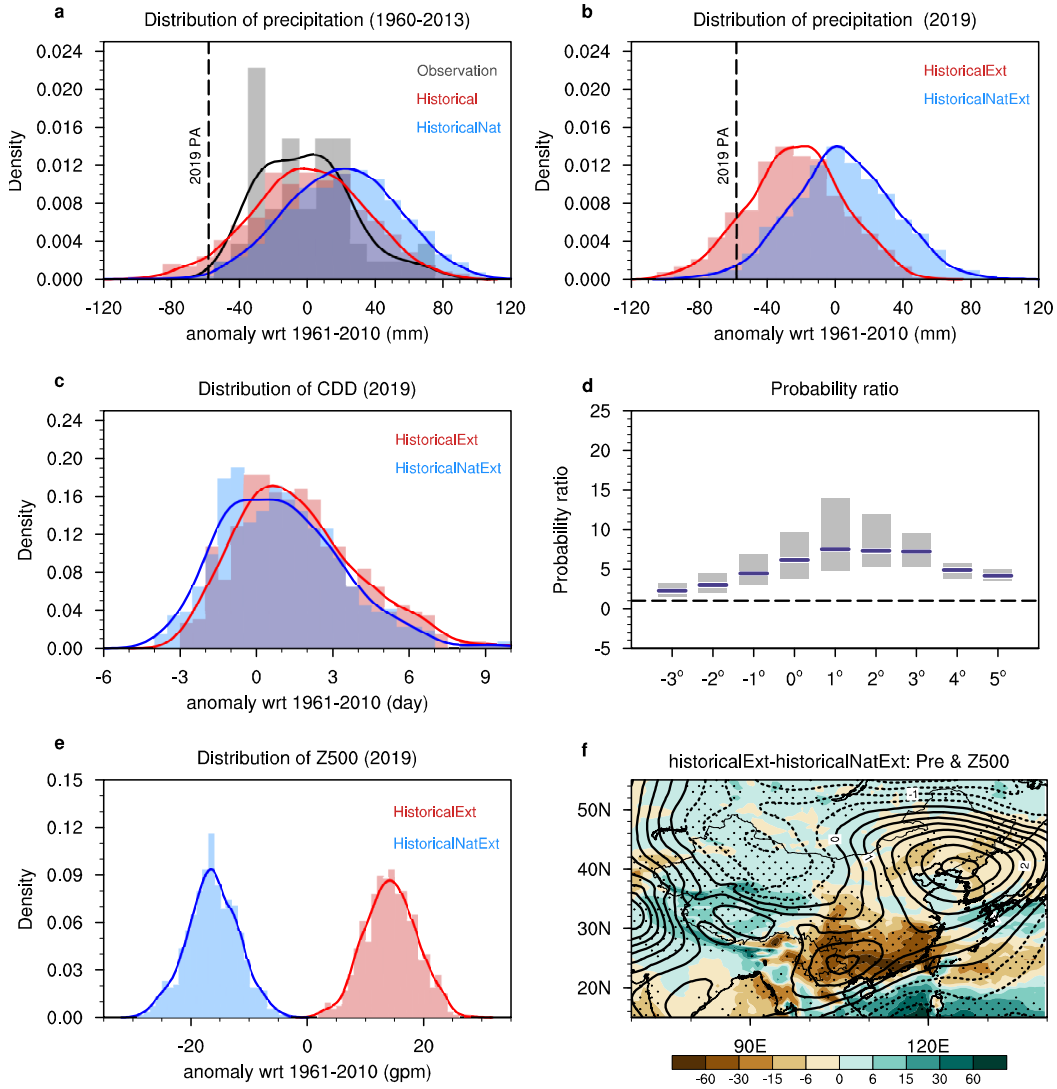

Figure 2 Kernel estimate of the probability density function and histograms of (a, b) PA (mm), (c) CDD (day) anomalies, averaged over Yunnan (black box of Fig. 1a) and (e) Z500 anomalies averaged over $15-30°\mathrm{N}, 90-120°\mathrm{E}$. Anomalies in model simulations are relative to 1961-2010 climatology in historical simulation. (a): observations (black), historical (red), and historicalNat simulations (blue) during 1960-2013. (b, c, e): historicalExt (red) and historicalNatExt (blue) 2019 simulations. The dashed black line denotes the observed event in 2019. (d): the probability ratios (blue lines) and 90% confidence intervals (gray shadings) for different study areas. 0°

denotes the selected area in the study, 1° denotes increasing area by moving each boundary by 1°, and -1° denotes reducing area by moving each boundary by 1°. (f): Differences of precipitation (shading; mm) and Z500 (contour; m) between historicalExt and historicalNatExt ensembles. Dots indicate 5% significance level for precipitation.