

Anthropogenic influence on 2022 extreme January–February precipitation in Southern China

Article

Accepted Version

Hu, Y., Dong, B. ORCID: <https://orcid.org/0000-0003-0809-7911>, Xie, J., Tan, H., Zhou, B., Lin, S., He, J. and Zhao, L. (2023) Anthropogenic influence on 2022 extreme January–February precipitation in Southern China. *Bulletin of the American Meteorological Society*, 104 (11). E1935-E1940. ISSN 1520-0477 doi: [10.1175/BAMS-D-23-0136.1](https://doi.org/10.1175/BAMS-D-23-0136.1) Available at <https://centaur.reading.ac.uk/113953/>

It is advisable to refer to the publisher's version if you intend to cite from the work. See [Guidance on citing](#).

Published version at: <https://journals.ametsoc.org/view/journals/bams/104/11/BAMS-D-23-0136.1.xml>

To link to this article DOI: <http://dx.doi.org/10.1175/BAMS-D-23-0136.1>

Publisher: American Meteorological Society

All outputs in CentAUR are protected by Intellectual Property Rights law, including copyright law. Copyright and IPR is retained by the creators or other copyright holders. Terms and conditions for use of this material are defined in the [End User Agreement](#).

www.reading.ac.uk/centaur

CentAUR

Central Archive at the University of Reading

Reading's research outputs online

**Anthropogenic Influence on 2022 Extreme January–February
Precipitation in Southern China**

Yamin Hu,^a Buwen Dong,^b Jiehong Xie,^c Haobo Tan,^d Baiquan Zhou,^e Shuheng Lin,^f
Jian He,^a and Liang Zhao^g

^a *Guangdong Climate Center, CMA, Guangzhou, China*

^b *National Centre for Atmospheric Science, Department of Meteorology, University of Reading,
Reading, United Kingdom*

^c *Jieyang Meteorological Bureau, CMA, Jieyang, China*

^d *Guangdong Meteorological Service, CMA, Guangzhou, China*

^e *State Key Laboratory of Severe Weather, Chinese Academy of Meteorological Sciences, CMA, Beijing,
China*

^f *School of Atmospheric Sciences, Sun Yat-sen University, Guangzhou, China*

^g *State Key Laboratory of Numerical Modeling for Atmosphere Sciences and Geophysical Fluid
Dynamics (LASG), Institute of Atmospheric Physics, Chinese Academy of Sciences, Beijing China*

Corresponding Author: Yamin Hu, huym@gd121.cn

Corresponding Author: Jiehong Xie, jiehongx@foxmail.com

CAPSULE

The precipitation in January–February 2022 in Southern China was the second largest since 1961. Anthropogenic influence reduced the likelihood of extreme events like 2022 by about 50% (55%) in HadGEM3 (CMIP6).

1. Introduction

From January to February (J–F) 2022, Southern China (SC) experienced abnormally heavy precipitation, with the regionally averaged total precipitation reaching 248 mm, making it the second-largest value since 1961. This extreme event resulted in significant damage to transportation, power supply, and crop production. About 6.092 million people and 422,300 hectares of crop area were affected, leading to a direct economic loss of 7.89 billion CNY. As a result, it was identified as one of the top ten natural disasters in 2022 by the Department of Emergency Management in China (https://www.mem.gov.cn/xw/yjglbgzdt/202301/t20230112_440396.shtml). This extreme precipitation event was attributed to the internal atmospheric dynamics (Ma et al. 2022). In this study, we assess how anthropogenic activity has changed the likelihood of extreme precipitation events similar to the J–F 2022 event over SC.

Previous studies have focused on summer extreme precipitation (Zhang et al. 2020; Li et al. 2021) and showed that anthropogenic warming has affected extreme precipitation over East Asia (Ma et al. 2017), intensifying the probability of short-term extreme precipitation events (Westra et al. 2014; Dong et al. 2020, 2021; Sun et al. 2022), while less attention was given to winter counterparts over SC (Hu et al. 2021). The objective of this study is to investigate whether anthropogenic influence has altered the likelihood of unusual precipitation in the 2022 J–F.

2. Data and methods

Daily gauge precipitation observations from approximately 2400 stations across China were obtained from the China Meteorological Administration (CMA) for the period 1961–2022. Monthly wind and sea level pressure fields from the ERA5

reanalysis provided by the European Centre for Medium-Range Weather Forecasts (ECMWF) (Hersbach et al. 2020) were used to analyze circulation characteristics.

To assess anthropogenic and natural factors' influences on the probability of the exceptional precipitation event in SC during J–F 2022, we used the Met Office HadGEM3-GA6-N216 model (referred to as HadGEM3 hereafter) simulations at a horizontal resolution of $0.56^{\circ} \times 0.83^{\circ}$ and 85 vertical levels (Ciavarella et al. 2018). The HadGEM3 model simulations are forced by observed sea surface temperature (SST)/sea ice extent (SIE) and therefore attribution of events is conditioned to SST/SIE. Both natural and anthropogenic forcing (ALL) and natural forcing (NAT) experiments were used. The details of the model can be found in Christidis et al. (2013). The ensemble simulations consisted of 15 members for 1960–2013 with ALL forcings, and 525 members for 2022 with ALL and NAT forcings (ALL₂₀₂₂ and NAT₂₀₂₂).

Meanwhile, we utilized simulations from climate models that participated in the Coupled Model Intercomparison Project Phase 6 (CMIP6) and Detection and Attribution Model Intercomparison Project (DAMIP) (Eyring et al. 2016; Gillett et al. 2016) under all anthropogenic and natural forcing combined (ALL), well-mixed greenhouse gas forcing (GHG), anthropogenic aerosol forcing (AA), and natural forcing (NAT) to assess the anthropogenic influence on the likelihood of the 2022 event (more details of model information are listed in Table ES1). CMIP6 model simulations are coupled model simulations and therefore we cannot use a specific year to represent 2022. We specify a period around our target year to represent 2022. Since historical simulations stopped in 2014 in CMIP6 and therefore we need to merge historical simulations with SSP2-4.5 future scenario simulations for 2015 to 2026. But the NAT simulations (DAMIP) stopped in 2020 and therefore are no NAT future scenario simulations. Therefore, in order to have enough samples to give a robust estimate of PDFs, we used a 15-year window of 2006–2020 in ALL, GHG, AA, and NAT as ALL₂₀₂₂, GHG₂₀₂₂, AA₂₀₂₂, NAT₂₀₂₂ in CMIP6/DAMIP simulations with a total 405 sample points (see details in SI).

In the 2022 J–F, the excessive precipitation was mainly concentrated over the region of (20.5° – 27°N, 106° – 119°E) (Fig.1a). To evaluate the model performance, we calculated the SC precipitation index (SCPI), defined as the normalized regional-averaged precipitation anomaly in J–F with respect to the climatological period of 1961–2005. Except for Quantile-Quantile (QQ) and Kernel Density Estimation (KDE) plot, a two-sample Kolmogorov-Smirnoff (K-S) test with a significance level of 0.05 was used to test whether the observed and simulated SCPI during 1961–2013 were from the same distribution.

The generalized extreme value (GEV) distribution was used to fit the precipitation indices and estimate the occurrence probability and return periods for both observations and simulations. The probability of an event, which is equivalent to or heavier than the J–F 2022 event, was defined as P_{ALL} and P_{NAT} for ALL and NAT, forcing experiments, respectively. The probability ratio ($PR_{ALL} = P_{ALL}/P_{NAT}$) is calculated to quantify the anthropogenic influences. Similarly, we used $PR_{GHG} = P_{GHG}/P_{NAT}$ and $PR_{AA} = P_{AA}/P_{NAT}$ to denote the GHG or AA forcing influences. The 90% confidence interval (90% CI) was obtained by using 1000 bootstrap resampling.

3. Results

The J–F time-averaged precipitation for 1961–2005 was about 117 mm in SC. However, during the J–F in 2022, the regional mean precipitation was about 248mm over SC and it was the second highest value since 1961 with the anomaly being about 2.3 standard deviations (2.3σ , using as the threshold for the 2022 event) above the climatology in the period of 1961–2005 (Figs. 1a–b). There are two dynamical drivers for this precipitation anomaly. One is the wave train propagating along the South Asian jet that intensifies the India–Burma trough. It enhances the SCP through exciting anomalous strong moisture transport from the Bay of Bengal and ascending motion. The other is the positive geopotential height anomaly over eastern Siberia that prompts southward cold air intrusion and convergence over the SC region (Fig. 1c) (Ma et al.

2022). The observed SCPI in 2022 J–F corresponds to a 1-in-31-year (18-143 year) event (Fig. 1d).

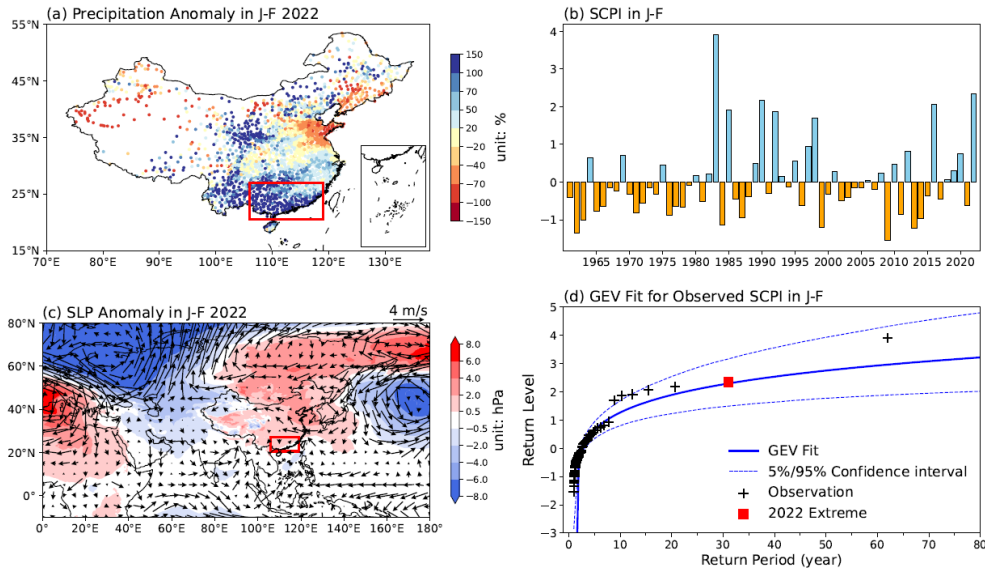


Fig. 1. (a) Percentage anomalies of observed precipitation in 2022 J–F relative to the 1961–2005 climatology. (b) Observed SCPI in each J–F for 1961–2022. (c) Spatial distribution of sea level pressure anomaly (shading, unit: hPa) and 850hPa wind anomaly (vector, unit: m/s). (d) GEV fit (blue solid line) of observed SCPI with 90% CI. The crosses are estimated from the empirical distributions of the observed precipitation index with the red square denoting the 2022 event.

Both the HadGEM3 and CMIP6 simulations reasonably well capture the variability of observed SCPI for the period 1961–2013, as shown in Figs. 2a–b, and probability density functions (PDFs) in simulations are comparable to that observed (Figs. 2c–d). Moreover, the observed precipitation indices fall within the range of those simulated by the models. We applied the QQ and KDE plots to test the distribution between the simulations and observations. Results show the simulations of HadGEM3 and CMIP6 follow the same distribution as observations (not shown). PDFs of SCPI exhibit similar distributions between model simulations and observations (Figs. 2c–d) with p-values of 0.50 and 0.33, respectively, according to the two-sample K-S test. These results suggest that both HadGEM3 and CMIP6 models can be considered reliable for the

attribution of the 2022 J–F extreme precipitation event over SC.

Figures 2e–f show the GEV-fitted PDFs for the 2022-like event under different external forcings. In both two sets of model simulations, PDFs of SCPI exhibit a drying shift from NAT to ALL. This shift indicates that the observed extreme precipitation event like 2022 is less likely to occur with anthropogenic influence (Table 1). The estimated occurrence probability decreased from 1.3% (1.3%–1.7%) in NAT to 0.7% (0.6%–0.9%) in ALL, with a PR_{ALL} of 0.50 (0.41–0.60) in HadGEM3 simulations. The return period is significantly increased from ~77 years in NAT to ~143 years in ALL (Fig. 2g). Similarly, in CMIP6 simulations, the estimated occurrence probability decreased from 3.8% (2.6%–4.8%) in NAT to 1.7% (1.0%–2.4%) in ALL, with a PR_{ALL} of 0.45 (0.38–0.53) and the return period increased from ~26 years in NAT to ~59 years in ALL (Fig. 2h). These results suggest that anthropogenic influence reduced the likelihood of extreme event like 2022 by about 50% (55%) in HadGEM3 (CMIP6).

Furthermore, the GHG forcing leads to a rightward shift of PDFs to a wetter climate relative to NAT(Fig.2f), while the AA forcing shifts to a drier world. The estimated P_{GHG} and P_{AA} indicate that a ~26 years event becomes a ~15 years event with $PR_{GHG}=1.75(1.60–2.00)$ in GHG and a ~83 years event in AA with $PR_{AA}=0.31(0.24–0.37)$.

Table 1. Attribution results for the 2022 J–F event with probability ratio (PR_{ALL} , PR_{GHG} and PR_{AA}), exceedance probability from ALL (P_{ALL}), GHG (P_{GHG}), AA (P_{AA}) and NAT (P_{NAT}), and the 90% confidence intervals (CI) in the bracket in HadGEM3 and CMIP6 simulations.

	HadGEM3	CMIP6
PR_{ALL}	0.50(0.41~0.60)	0.45(0.38~0.53)
PR_{GHG}	——	1.75(1.60~2.00)
PR_{AA}	——	0.31(0.24~0.37)
$P_{ALL}(\%)$	0.7(0.6~0.9)	1.7(1.0~2.4)

$P_{\text{GHG}}(\%)$	——	6.7(5.1~7.9)
$P_{\text{AA}}(\%)$	——	1.2(0.5~1.9)
$P_{\text{NAT}}(\%)$	1.3(1.3~1.7)	3.8(2.6~4.8)

How does anthropogenic influence reduce the likelihood of extreme events like 2022? The J–F differences of ensemble mean precipitation and atmospheric circulations between ALL and NAT experiments are analyzed (Fig. ES1). Decreases in precipitation occur in both HadGEM3 and CMIP6 simulations over SC (Figs. ES1a–b), accompanied by anomalous positive sea level pressure (SLP) and an anomalous anticyclonic circulation over SC (Figs. ES1c–d).

The two most important anthropogenic forcings are greenhouse gases and atmospheric aerosols and they can have different effects on SC J–F precipitation. According to the CMIP6 results, compared to the NAT, the GHG simulation is shifting rightward to a wetter climate, while the AA distribution is marked by a flatter distribution with shifting to a drier region (Fig. 2f). A further comparison of AA and GHG with the NAT experiments reveals the impact of different anthropogenic forcings. Through aerosol radiation and cloud interactions, there is a positive SLP anomaly in most of the mid-high latitudes of East Asia, with an anomalously strong Siberian High together with changes in the Walker circulation over the eastern Indian Ocean, Maritime Continent and western Pacific Ocean (Takahashi et al. 2018) and with a weakening of anomalous anticyclonic circulation over the western North Pacific(WNP), which leads to anti-cyclone anomaly over the SC, so a reduction in J–F precipitation happened in SC (Fig. ES2a). On the other hand, over SC, the effect of non-absorbing aerosols is dominant. The solar flux at the surface is significantly reduced directly by the scattering of non-absorbing aerosols and indirectly by the intensification of short-wave cloud forcing. Accordingly, the surface air temperature in SC is reduced, which leads to the moisture transport decreasing, so the precipitation is also significantly reduced in South China (Huang et al. 2007; Jiang et al.2017) (Fig. ES2c).

On the other hand, GHG contributes to an increase in precipitation over SC (Fig.

ES2b). This is partly related to GHG induced increase of moisture in the atmosphere associated with warming (Guo et al. 2023) and is also associated with an anomalous negative SLP in the mid-latitudes, where the Baikal trough deepens and favors the cold air to SC. The anomalous anticyclonic circulation in the WNP subtropical region, caused the convergence of warm and moist air from the southern flank of the Philippine high, contributing to SC precipitation increase (Fig. ES2d). The impacts from aerosols overwhelm the impacts from GHG changes, leading to a decrease of precipitation from NAT to ALL in CMIP6/DAMIP simulations.

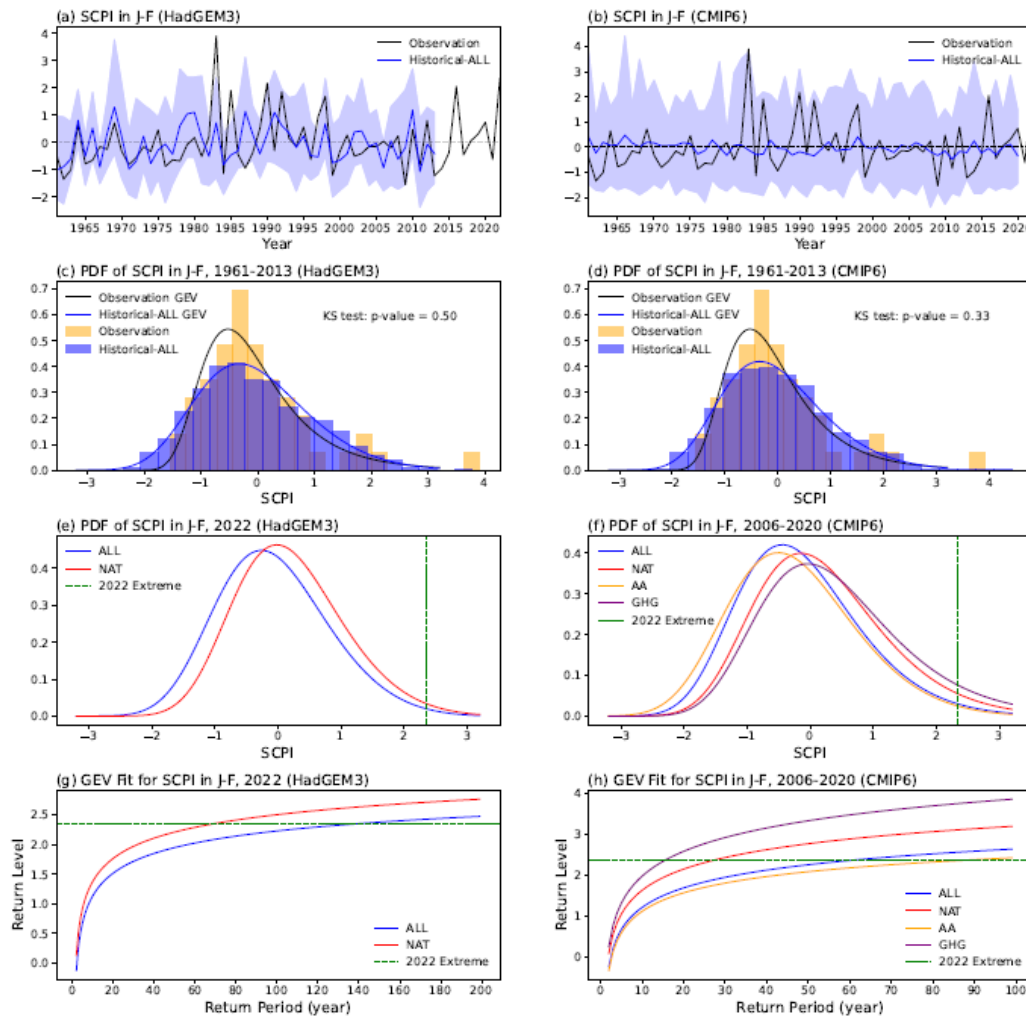


Fig. 2. (a–b) Time series of observed (black) and simulated ensemble mean (blue) SCPI for 1961–2022, with 15-member (a) and 27-member (b) spread shown as light blue shading in HadGEM3 and CMIP6 simulations, respectively. (c–d) SCPI original

(bar) and GEV fitted PDFs (solid line) of observations (yellow bar and black line) and historical ALL simulations (blue bar and blue line) for 1961–2013 in HadGEM3 (c) and CMIP6 (d). The p-value for the K-S test is on the top right. (e–f) GEV fitted PDFs of SCPI in 2022 based on ALL (blue) and NAT (red) ensembles in HadGEM3, and ALL (blue), NAT (red), GHG (purple) and AA (yellow) in CMIP6 simulations. The dashed green line denotes the observed 2022 event. (g–h) As in (e–f) but for return periods.

4. Conclusions

We conducted an assessment of the anthropogenic influence on the likelihood of the 2022 extreme wet J–F in SC using two sets of model simulations. The analysis reveals that anthropogenic activities have reduced the likelihood of extreme events like 2022 by about 50% (55%) in HadGEM3 (CMIP6) simulations. Analyses of single forcing experiments using CMIP6 model ensembles demonstrate different roles of changes in GHG and AA in J–F precipitation over SC with GHG forcing inducing an increase and AA forcing inducing a decrease, similar to previous studies in warm season (Sun et al. 2022; Guo et al. 2023; Hu et al. 2023). However, the magnitude of AA-induced precipitation decrease is larger than that of GHG-induced increase (Cao et al. 2022), leading to the reduced likelihood of the J–F precipitation event similar to that of 2022 in SC by the combined effect of anthropogenic forcing.

Acknowledgements.

This study was funded by the National Natural Science Foundation of China (U2142205 and 42075040); Guangdong Major Project of Basic and Applied Basic Research (2020B0301030004); Special Fund of China Meteorological Administration for Innovation and Development (CXFZ2023J027); and Special Fund for Forecasters of China Meteorological Administration (CMAYBY2020-094).

Data Availability Statement.

The following data are available online: the CMIP6 GCM simulations (<https://esgf-node.llnl.gov/projects/cmip6/>), and the homogenized station data in China (<http://data.cma.cn/>).

REFERENCES

Cao, J., Wang, H., Wang, B., Zhao, H., Wang, C., and Zhu, X., 2022: Higher sensitivity of Northern Hemisphere monsoon to anthropogenic aerosol than greenhouse gases. *Geophysical Research Letters*, 49, e2022GL100270. <https://doi.org/10.1029/2022GL100270>

Christidis, Nikolaos., P. A. Stott, A. A. Scaife, A. Arribas, G. S. Jones, D. Copsey, J. R. Knight, and W. J. Tennant, 2013: A new HadGEM3-A-based system for attribution of weather- and climate-related extreme events. *J. Clim.*, 26, 2756–2783, <https://doi.org/10.1175/JCLI-D-12-00169.1>

Ciavarella, A., and Coauthors, 2018: Upgrade of the HadGEM3-A based attribution system to high resolution and a new validation framework for probabilistic event attribution. *Wea. Climate Extremes*, 20, 9–32, <https://doi.org/10.1016/j.wace.2018.03.003>.

Dong, S., Y. Sun, and C. Li, 2020: Detection of human influence on precipitation extremes in Asia. *J. Clim.*, 33, 5293–5304, <https://doi.org/10.1175/JCLI-D-19-0371.1>.

Dong, S., Sun, Y., Li, C., Zhang, X., Min, S.-K., and Kim, Y.-H., 2021: Attribution of Extreme Precipitation with Updated Observations and CMIP6 Simulations. *J. Clim.*, 34, 871–881. <https://doi.org/10.1175/JCLI-D-19-1017.1>.

Eyring, V. and Coauthors, 2016: Overview of the coupled model intercomparison project phase 6 (CMIP6) experimental design and organization. *Geosci. Model Dev.* 9, 1937–1958, <https://doi.org/10.5194/gmd-9-1937-2016>.

Gillett, N. P., and Coauthors, 2016: The detection and attribution model

intercomparison project DAMIP v1.0) contribution to CMIP6. *Geosci. Model Dev.*
9, 3685–3697, <https://doi.org/10.5194/gmd-9-3685-2016>.

Guo, Y., Dong, B. and Zhu, J. 2023: Anthropogenic impacts on changes in summer
extreme precipitation over China during 1961–2014: roles of greenhouse gases and
anthropogenic aerosols. *Clim. Dyn.*, 60, 2633–2643,
<https://doi.org/10.1007/s00382-022-06453-4>.

Hersbach, H., Bell, B., Berrisford, P., et al. 2020: The ERA5 global reanalysis. *Q J R
Meteorol Soc.* 146: 1999– 2049. <https://doi.org/10.1002/qj.3803>

Hu, Z., and Coauthors, 2021: Was the Extended Rainy Winter 2018/19 over the Middle
and Lower Reaches of the Yangtze River Driven by Anthropogenic Forcing? *Bull.
Amer. Meteor. Soc.*, 102, S67–S73, <https://doi.org/10.1175/BAMS-D-20-0127.1>.

Hu, T., and Coauthors, 2023: Anthropogenic Influence on the 2021 Wettest September
in Northern China. *Bull. Amer. Meteor. Soc.*, 104, E243–248,
<https://doi.org/10.1175/BAMS-D-22-0156.1>.

Huang, Y., W. L. Chameides, and R. E. Dickinson, 2007: Direct and indirect effects of
anthropogenic aerosols on regional precipitation over East Asia, *J. Geophys. Res.*,
112, D03212, <https://doi.org/10.1029/2006JD007114>.

Jiang, Y., and Coauthors, 2017: Anthropogenic aerosol effects on East Asian winter
monsoon: The role of black carbon-induced Tibetan Plateau warming, *J. Geophys.
Res. Atmos.*, 122, 5883– 5902, <https://doi.org/10.1002/2016JD026237>.

Li, R., and Coauthors, 2021: Anthropogenic Influences on heavy precipitation during
the 2019 extremely wet rainy season in Southern China. *Bull. Amer. Meteor. Soc.*,
102, S103–S109, <https://doi.org/10.1175/BAMS-D-20-0135.1>.

Ma, H., Wang, R., Li, X., Lai, A., Yang, H., and Li, X., 2022: Why was South China
extremely wet during January–February 2022 despite La Niña? *Front. Earth Sci.*,
10, 982225, <https://doi.org/10.3389/feart.2022.982225>.

Ma, S., and Coauthors, 2017: Detectable Anthropogenic Shift toward Heavy

261 Precipitation over Eastern China. *J. Clim.*, 30, 1381–1396,
 262 <https://doi.org/10.1175/JCLI-D-16-0311.1>.

263 Sun, Y., and Coauthors, 2022: Understanding human influence on climate change in
 264 China. *Natl. Sci. Rev.*, 9, 3, <https://doi.org/10.1093/nsr/nwab113>.

265 Takahashi, H.G., Watanabe, S., Nakata, M. et al., 2018: Response of the atmospheric
 266 hydrological cycle over the tropical Asian monsoon regions to anthropogenic
 267 aerosols and its seasonality. *Prog. Earth. Planet. Sci.*, 5, 44.
 268 <https://doi.org/10.1186/s40645-018-0197-2>.

269 Westra, S., and Coauthors, 2014: Future changes to the intensity and frequency of short-
 270 duration extreme rainfall. *Rev. Geophys.*, 52, 522–555,
 271 <https://doi.org/10.1002/2014RG000464>

272 Wu, R., Hu, Z. and Kirtman, B., 2003: Evolution of ENSO-related rainfall anomalies
 273 in East Asia. *J. Clim.*, 16, 3742–3758, [https://doi.org/10.1175/1520-0442\(2003\)016<3742:EOERAI>2.0.CO;2](https://doi.org/10.1175/1520-0442(2003)016<3742:EOERAI>2.0.CO;2).

275 Zhang, W., and Coauthors. 2020: Anthropogenic influence on 2018 summer persistent
 276 heavy rainfall in central Western China. *Bull. Amer. Meteor. Soc.*, 101, S65-S70.
 277 <https://www.jstor.org/journal/bullamermetesoci>.