

# Attribution of the July 2021 Record-breaking Northwest Pacific Marine Heatwave to Global warming, Atmospheric Circulation and ENSO

Delei Li<sup>1,2</sup>, Yang Chen<sup>3</sup>, Jifeng Qi<sup>1,2</sup>, Yuchao Zhu<sup>1,2</sup>, Chunhui Lu<sup>4</sup>, Baoshu Yin<sup>1,2,5\*</sup>

<sup>1</sup>CAS Key Laboratory of Ocean Circulation and Waves, Institute of Oceanology,  
Chinese Academy of Sciences, Qingdao, China

<sup>2</sup>Pilot National Laboratory for Marine Science and Technology (Qingdao),  
Qingdao, China

<sup>3</sup>State Key Laboratory of Severe Weather, Chinese Academy of  
Meteorological Sciences, Beijing, China

<sup>4</sup>National Climate Center, Laboratory for Climate Studies,  
China Meteorological Administration, Beijing, China

<sup>5</sup>College of Earth and Planetary Sciences, University of Chinese Academy of  
Sciences, Beijing, China

\*Corresponding Author

Email address: [bsyin@qdio.ac.cn](mailto:bsyin@qdio.ac.cn)

**Capsule summary:**

The 2021 northwest Pacific marine heatwave was favored by the atmospheric and oceanic conditions; changes in the mean climate due to anthropogenic warming made the event 43 times more likely.

**1. Introduction**

In July 2021, the northwest Pacific Ocean (NWP) experienced a record-breaking marine heatwave (MHW), featuring SST anomalies larger than 4 °C in some coastal areas of the Japan (East) Sea and Okhotsk Sea relative to 1982-2011 (Fig. 1a). This event triggered an unprecedented widespread outbreak of harmful algae in the coastal waters, causing serious damage to the marine ecosystems (Kuroda et al. 2021; Kuroda et al. 2022) and a record-breaking loss of over 8.19 Million USD to Japanese coastal fisheries (Iwataki et al. 2022). The MHW might have also contributed to the exceptional humid-hot weather experienced by Japan during the Olympic Games (Sealy and Wang 2021).

Several processes may have contributed to the July 2021 northwest Pacific MHW. Anthropogenic warming has contributed to the increasing occurrences of MHW in recent decades (Hayashi et al. 2021; King et al. 2017; Laufkotter et al. 2020; Oliver et al. 2018b). Natural variability can play a role as well (Oliver et al. 2018a). Anomalies of the 500-hPa geopotential height (z500) show the dominance of an anomalously anticyclonic circulation over the study domain in July 2021 (Fig. 1c). This anticyclonic cell was embedded in a circum-global Rossby wave train (Fig. ES1a), which might cause hemispheric synchronized land and ocean heatwaves at mid-latitudes (Christidis and Stott 2022; Kornhuber et al. 2020; Wehrli et al. 2020). The large-scale SST anomaly pattern features a developing La Niña phase, which was initiated in August 2020, weakened in late spring and summer 2021, and re-strengthened to April 2022 (Fig. ES1b). The developing phase

of La Niña is also believed important to shape both terrestrial and marine heatwaves in this region (Hong et al. 2001; Takahashi et al. 2016).

This paper mainly aims to answer three questions: (1) What does the July 2021 MHW in the NWP look like in a historical context? (2) What are the relative roles of anthropogenic forcing, atmospheric circulation, and ENSO on the likelihood of the event? (3) Is such an event more likely under an anticyclonic circulation pattern and a developing La Niña phase?

## **2. Data and Methods**

The July 2021 MHW was bounded by 36°-56°N, 127°-147°E over the NWP (Fig. 1a). Considering the observation uncertainty, three observation datasets are used to analyze the long-term variations of July SST over the study domain (Fig. 1b). These datasets include the 0.25° daily NOAA OISST dataset in 1982-2021 (Reynolds et al. 2007), 2.0° monthly NOAA ERSSTv5 in 1901-2021 (Huang et al. 2017), and 1.0° monthly COBE-SST2 in 1901-2021 (Hirahara et al. 2014). We define all anomalies relative to a baseline period 1901-1930, which is closer to the pre-industrial climate and accounts for most of the anthropogenic effect (Christidis and Stott 2022).

For comparison (Fig. 1b), the OISST July anomaly is calculated relative to the period 1982–2011, we further offset by the mean warming from 1901-1930 to 1982–2011 calculated from COBE-SST2 (+0.484°C), to reference it to the chosen base period 1901-1930 (e.g., Oliver et al. (2018a)).

To analyze the anomalous synoptic pattern during the event (Fig. 1c), z500 and wind components from NOAA-20CRv3 (Slivinski et al. 2019) in 1901-2015 were used, with an extension to 2021 using ERA5 reanalysis (Hersbach et al. 2020) by reconciling their 1979-2015 climatology means (Zhou et al. 2021).

Ten CMIP6 models (Eyring et al. 2016) are evaluated and these runs that can capture the observed probability distribution of July SST anomalies and z500 anomalies (rectangle

region in Fig. 1c), Niño-3.4 index, and the relationship between July SST anomalies and z500 anomalies/Niño-3.4 index are selected (see supplementary for simulations evaluation and selection). This results in 54 runs for historical all-forcing (natural plus all anthropogenic forcing) (ALL) and 36 runs for natural-only forcing without the anthropogenic effects (NAT). Since the CMIP6 historical runs terminate at the end of 2014, the historical runs were extended to 2030 with the shared socioeconomic pathway 370 scenarios (SSP370) runs. The data over July 2011-2030 are then used as samples, representing current climate conditions similar to July 2021. This generates 1080 samples (20 years  $\times$  54 runs) for ALL experiment. The corresponding 36 natural-only forcing simulations (NAT) from CMIP6 models were selected from 1991 to 2020, which gives 1080 samples (30 years  $\times$  36 runs) as well. The CMIP6 NAT runs end in 2020, considering the general stationary, a longer period for NAT runs can provide a more robust estimation of small probability (Kirchmeier-Young et al. 2021). For consistency, observation and model data were bilinearly remapped onto  $1.0^\circ \times 1.0^\circ$  grids with anomalies relative to 1901-1930 mean.

We stratify the ALL and NAT samples by their resemblance of large-scale atmospheric circulation to the observed pattern (black rectangle in Fig. 1c), into high resembling ( $r > 0.5$ ) samples ALL\_high (n=307), NAT\_high (n=239), and low resembling ( $r < 0.1$ ) samples ALL\_low (n=548), NAT\_low (n=574). Fig. 1d shows the anomaly of z500 for the ensemble mean of ALL\_high samples, which displays an anticyclone circulation pattern similar to July 2021 (Fig. 1c). El Niño (La Niña) is defined as when the July SST anomaly (calculated relative to a 30-year running mean) in the Niño-3.4 region ( $5^\circ\text{N}$ – $5^\circ\text{S}$ ,  $170^\circ$ – $120^\circ\text{W}$ ) is greater (less) than  $0.5^\circ\text{C}$  (<http://bmcnoldy.rsmas.miami.edu/tropics/oni/>), which generates July SST anomaly samples in different ENSO phases for ALL: ALL\_El Niño (n=289), ALL\_La Niña (n=270) and ALL\_Neutral (n=521).

The exceeding probabilities were obtained using the generalized Pareto distribution if the threshold lay at the tails of the distribution (Christidis and Stott 2022; Lu et al. 2020).

The fraction of attributable risk ( $FAR = 1 - P_0/P_1$ ) and the corresponding risk ratios ( $RR = P_1/P_0$ ) (Chen et al. 2019; Fischer and Knutti 2015) were used to estimate the effect of anthropogenic influence, atmospheric circulation, and ENSO on the present-day likelihood of the July 2021 MHW in the NWP. For instance, to quantify the anthropogenic influence,  $P_0$  and  $P_1$  indicates the probabilities exceeding the threshold for the NAT and ALL samples, respectively. The  $RR$  describes how much more/less likely the event is in the ALL samples with anthropogenic influence compared to NAT samples without anthropogenic influence.  $FAR$  is interpreted as the fraction of the likelihood of an event that is attributable to the anthropogenic influence. E.g., at  $RR=5$  (with  $FAR=0.8$ ), the likelihood of an event occurring under the anthropogenic influence is five times as that of its occurring without anthropogenic influences, and anthropogenic influence explains 80% attributable risk for the event. Furthermore, a Monte Carlo bootstrap was performed 1000 times to estimate the uncertainty of  $FAR$  and  $RR$  by randomly resampling (Christidis et al. 2013).

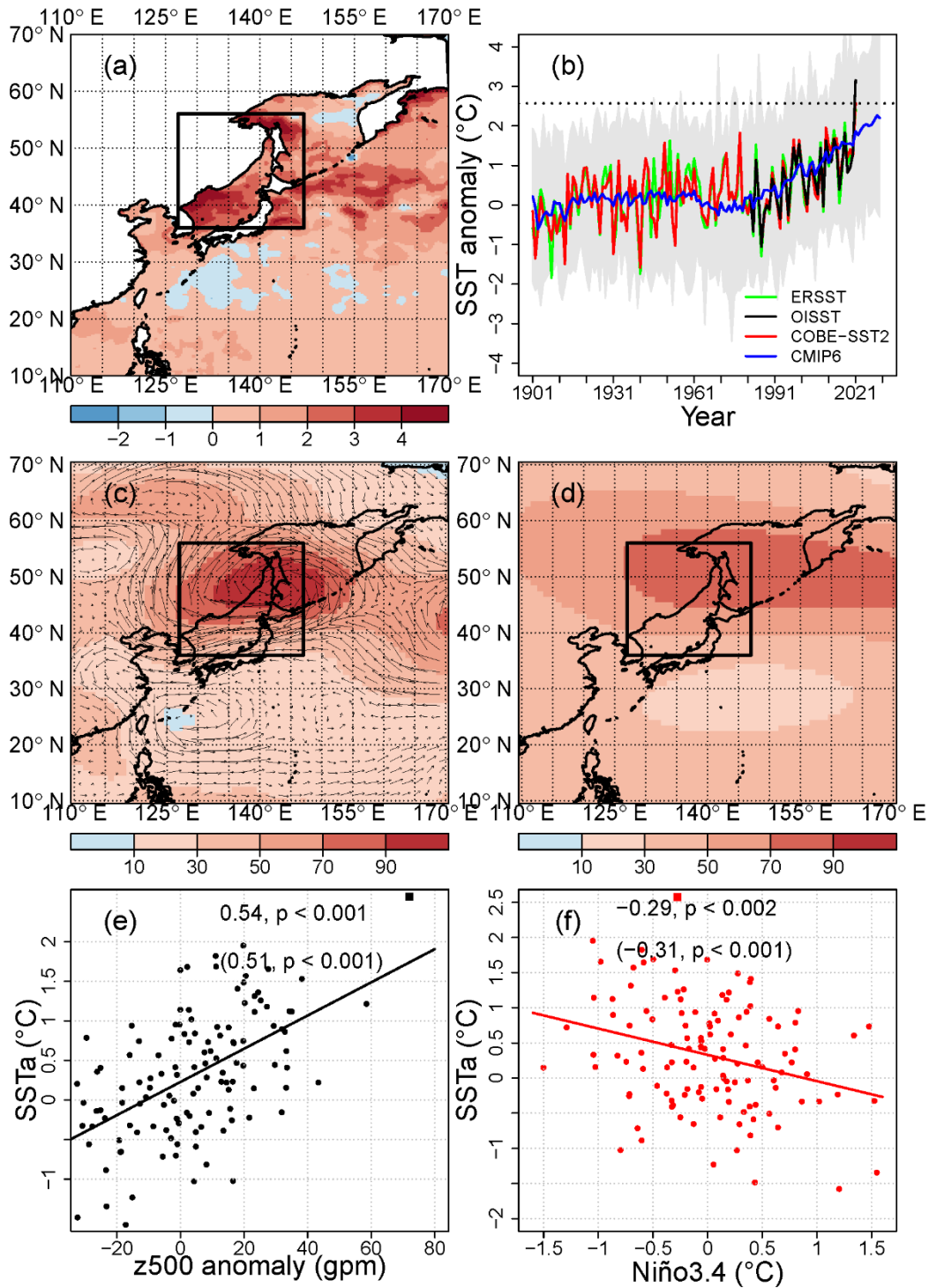


Fig. 1 (a) OISST temperature anomalies in July 2021 over the NWP relative to the July mean of 1982–2011. (b) Timeseries of July SST anomalies (SSTa) over the study region (black rectangle in (a)) from three observations (ERSSTv5, OISST, and COBE-SST2), CMIP6 ALL simulations (gray shaded) and their ensemble mean (blue). (c) Observed anomalies of z500 and wind vectors in July 2021. (d) the July anomaly of z500 for the ensemble mean of ALL\_high samples. Scatter plots of July SSTa over the study

region (black rectangle in (a)) from COBE-SST2 with observed (e) z500 anomaly and (f) Niño-3.4 index. Linear fits are displayed together with correlation coefficients and significance levels calculated before and after detrending, with the latter in square brackets. The square symbols in (e-f) indicate the case of July 2021. Note that the anomalies are relative to July mean of 1901-1930 except for (a).

### **3. Results**

In July 2021, almost the entirety of the study domain features positive SST anomalies relative to the period 1982-2011, with 59.3% of the region having the record-breaking July SST (Fig. 1a). In terms of domain-averaged values, July 2021 is the warmest one among all historical July counterparts for all three observations, with anomalies of 3.16°C, 2.57°C, and 2.40°C for OISST, COBE-SST2, and ERSSTv5 respectively. The anomalies of the COBE-SST2 dataset are used hereafter and 2.57 °C is used as the threshold to characterize the July 2021 MHW, due to it having a comparable resolution to CMIP6 models (Table ES1) and its SST anomaly in July 2021 is closer to the ensemble mean of observations. It is found that the selected CMIP6 ALL models can capture the long-term trends of SST, z500, and Niño-3.4 index, and the observation variations are encapsulated by the spread of CMIP6 runs (Fig. 1b, Figs. ES2c-2d).

MHWs are typically driven by heat transport by ocean currents, persistent large-scale atmospheric synoptic systems, and large-scale climate modes (Sen Gupta et al. 2020). Specific to this case, a strong anomalous anticyclonic circulation was centered over the NWP (Fig. 1c), associated with a northwestward expansion of the North Pacific Subtropical High (Kuroda and Setou 2021), facilitating increased surface solar radiation and adiabatic heating, and further warming up the upper ocean through air-sea interactions. The July SST anomalies show a statistically significant positive correlation ( $R=0.54$ ) with the z500 anomaly in this region (Fig. 1e) and a significant negative correlation ( $R=-0.29$ ) with the Niño-3.4 index (Fig. 1f).



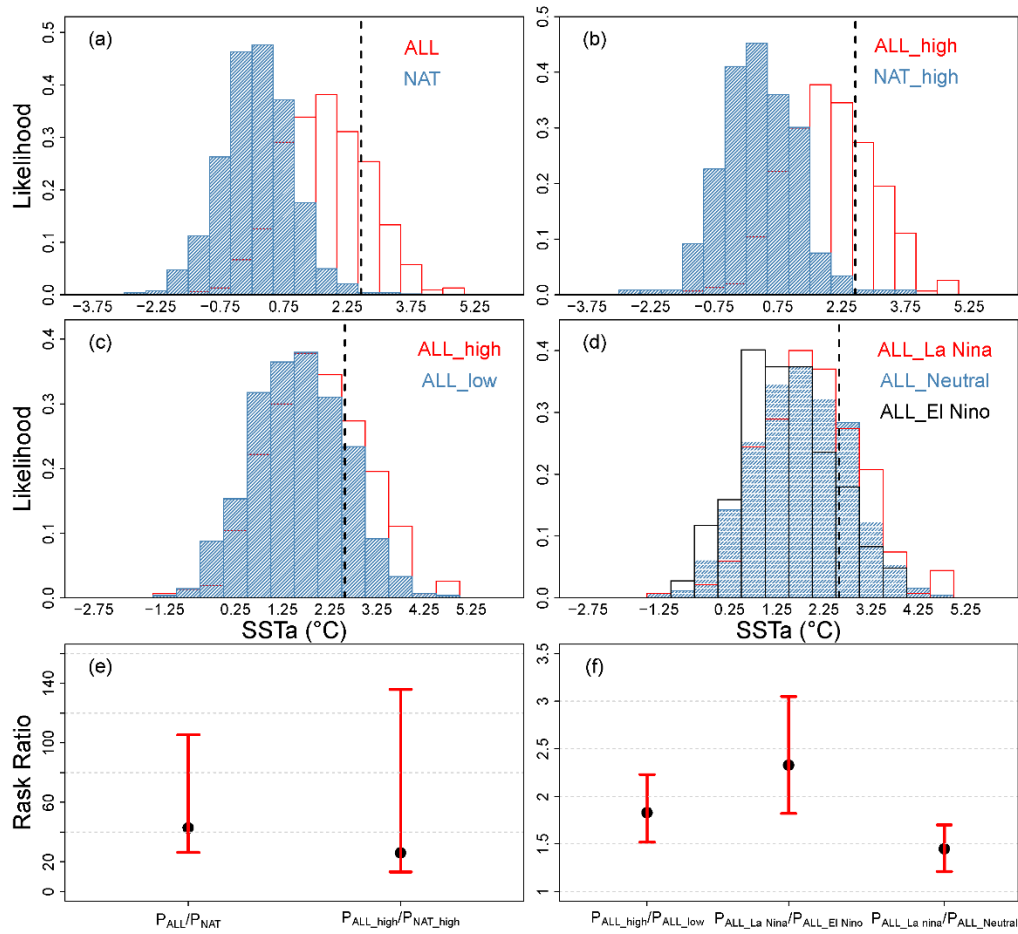


Fig. 2 (a) Normalized distributions of the July SST anomaly for CMIP6 ALL and NAT cases, (b) As in (a), but only for cases with similar circulation patterns (within the black rectangle in Fig. 1c) to that in July 2021 with (red, ALL\_high) and without (blue, NAT\_high) anthropogenic forcing, (c) As in (a), but only for CMIP6 All cases with high (red, ALL\_high) and low (blue, ALL\_low) correlations with the flow pattern (within the black rectangle in Fig. 1c) in July 2021. (d) As in (a) but only for cases under the conditions of El Niño, neutral, and La Niña. (e, f) Best estimates (black points) and 90% confidence intervals (red lines) of the risk ratio for different groups of cases from (a-d)

The likelihood of July 2021 MHW is fairly rare (0.48%) in the NAT world (Fig. 2a and Table 1). It is found that human influences increased the likelihood by 42.87 times and explained 97.7% (90% CI: 96.2% - 99.1%) attributable risk for the record-breaking July 2021 MHW (Figs. 2a and 2e, Table 1). The return period (inverse probability) is estimated to decrease from more than two centuries in the NAT world to about five years in the present climate (Table 1). The results are generally consistent with those findings of



Laufkotter et al. (2020), indicating the significant role of anthropogenic forcing on the likelihood of MHW.

In terms of conditional attributions based on the high-correlation samples (ALL\_high and NAT\_high), we find again that human influence shifts the distribution to warmer temperatures (Fig. 2b), making the event 26.09 times more likely to occur and explained 96.2% (90% CI: 92.4%, 99.3%) attributable risk for the event (Figs. 2b and 2e, Table 1). The return period (3.55 years) is lower in the conditional attribution compared to the unconditional attribution since anticyclonic conditions favor warm anomalies.

We also compared the likelihood of the event between anticyclonic conditions compared to other circulation patterns under the present-day climate, by using the ALL\_high and ALL\_low samples. It is estimated that anticyclonic conditions make the event 1.83 times (90% CI: 1.52-2.23) more likely to occur and explained 45.5% (90% CI: 34.4%, 55.2%) attributable risk for the event (Figs. 2c and 2f, Table 1). Comparing the all-forcings samples in La Niña years with El Niño (Neutral) years, we found that La Niña increased the likelihood by 2.33 (1.45) times and explained 57% (31%) attributable risk (see Figs. 2d and 2f, Table 1).

In addition, an experimental attribution analysis revealed that the likelihood of detrended SSTa of July 2021 (0.84°C) in the study is comparable between ALL (ALL\_high) runs and NAT (NAT\_high) runs (Fig. ES2). This gives an estimated risk ratio of 1.04 (90% CI: 0.90, 1.19) and 0.93 (90% CI: 0.74, 1.15) for ALL/NAT and ALL\_high/NAT\_high, respectively, implying that the detrended SST anomaly cannot be attributed to the anthropogenic forcing.

Table 1. The best estimate and 90% confidence intervals (90% CI) of the return period (RP), the fraction of attributable risk (FAR), and risk ratio (RR) estimated with CMIP6 models

Runs	RP (yr) (90%CI)	FAR (90% CI)	RR (90% CI)
ALL	4.88 (4.51, 5.29)	0.977 (0.962, 0.991)	42.87 (26.3, 105.4)
NAT	209.00 (130, 501)		
ALL_high	3.55 (3.16, 4.03)	0.962 (0.924, 0.993)	26.09 (13.2, 136)
Nat_high	92.51 (48, 502)		
ALL_high	3.55 (3.16, 4.03)	0.455 (0.344, 0.552)	1.83 (1.52, 2.23)
ALL_low	6.51 (5.69, 7.48)		
ALL_La Niña	3.39 (3.02, 3.98)	0.57 (0.449, 0.672)	2.33 (1.82, 3.05)
ALL_El Niño	7.90 (6.6, 10)		
ALL_La Niña	3.39 (3.02, 3.98)	0.31 (0.171, 0.413)	1.45 (1.21, 1.7)
ALL_Neutral	4.92 (4.4, 5.51)		

## 4. Conclusions

In July 2021, an unprecedented MHW hit the northwest Pacific Ocean. Attribution analysis based on the CMIP6 models reveals that human influence is estimated to have made such an event about 43 times more likely and to have the return period from more than two centuries in the NAT world to about five years in the present climate. The anticyclone atmospheric patterns and developing phase of La Niña favor the likelihood of such an event by a factor of about 2, which is weaker than the anthropogenic influence. In addition, an experimental attribution analysis based on detrended SSTa reveals that the increased likelihood of July 2021 MHW in the NWP is governed by the warming mean climate instead of climatic changes in SST variance.

## Acknowledgments

The study was supported by the National Key R&D Program of Intergovernmental Cooperation in Science and Technology (2022YFE0112800), the National Natural Science Foundation of China (42176203), the Strategic Priority Research Program of the Chinese Academy of Sciences (XDB42000000), the NSFC-Shandong Joint Fund (U1806227), the

196 Youth Innovation Promotion Association CAS (2022204).

## 197 **Reference**

198 Chen, Y., and Coauthors, 2019: Anthropogenic Warming has Substantially Increased the Likelihood of  
199 July 2017–Like Heat Waves over Central Eastern China *Bulletin of the American Meteorological Society*,  
200 **100**, S91–S95.

201 Christidis, N., and P. A. Stott, 2022: Anthropogenic Climate Change and the Record-High Temperature  
202 of May 2020 in Western Europe. *Bulletin of the American Meteorological Society*, **103**, S33–S37.

203 Christidis, N., and Coauthors, 2013: A New HadGEM3-A-Based System for Attribution of Weather-  
204 and Climate-Related Extreme Events. *J Climate*, **26**, 2756–2783.

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

208 Fischer, E. M., and R. Knutti, 2015: Anthropogenic contribution to global occurrence of heavy-  
209 precipitation and high-temperature extremes. *Nature Climate Change*, **5**, 560.

210 Hayashi, M., H. Shiogama, S. Emori, T. Ogura, and N. Hirota, 2021: The Northwestern Pacific Warming  
211 Record in August 2020 Occurred Under Anthropogenic Forcing. *Geophysical Research Letters*, **48**,  
212 e2020GL090956.

213 Hersbach, H., and Coauthors, 2020: The ERA5 global reanalysis. *Quarterly Journal of the Royal*  
214 *Meteorological Society*, **146**, 1999–2049.

215 Hirahara, S., M. Ishii, and Y. Fukuda, 2014: Centennial-Scale Sea Surface Temperature Analysis and Its  
216 Uncertainty. *J Climate*, **27**, 57–75.

217 Hong, C.-h., K.-D. Cho, and H.-J. Kim, 2001: The relationship between ENSO events and sea surface  
218 temperature in the East (Japan) Sea. *Prog Oceanogr*, **49**, 21–40.

219 Huang, B., and Coauthors, 2017: Extended Reconstructed Sea Surface Temperature, Version 5  
220 (ERSSTv5): Upgrades, Validations, and Intercomparisons *J Climate*, **30**, 8179–8205.

221 Iwataki, M., and Coauthors, 2022: Morphological variation and phylogeny of *Karenia selliformis*

(Gymnodiniales, Dinophyceae) in an intensive cold-water algal bloom in eastern Hokkaido, Japan.  
*Harmful Algae*, **114**, 102204.

King, A. D., D. J. Karoly, and B. J. Henley, 2017: Australian climate extremes at 1.5 °C and 2 °C of global warming. *Nature Climate Change*, **7**, 412-416.

Kirchmeier-Young, M. C., H. Wan, and X. Zhang, 2021: Anthropogenic Contribution to the Rainfall Associated with the 2019 Ottawa River Flood. *Bulletin of the American Meteorological Society*, **102**, S33-S38.

Kornhuber, K., D. Coumou, E. Vogel, C. Lesk, J. F. Donges, J. Lehmann, and R. M. Horton, 2020: Amplified Rossby waves enhance risk of concurrent heatwaves in major breadbasket regions. *Nature Climate Change*, **10**, 48-53.

Kuroda, H., and T. Setou, 2021: Extensive Marine Heatwaves at the Sea Surface in the Northwestern Pacific Ocean in Summer 2021. *Remote Sensing*, **13**, 3989.

Kuroda, H., T. Azumaya, T. Setou, and N. Hasegawa, 2021: Unprecedented Outbreak of Harmful Algae in Pacific Coastal Waters off Southeast Hokkaido, Japan, during Late Summer 2021 after Record-Breaking Marine Heatwaves. *Journal of Marine Science and Engineering*, **9**, 1335.

Kuroda, H., Y. Taniuchi, T. Watanabe, T. Azumaya, and N. Hasegawa, 2022: Distribution of Harmful Algae (*Karenia* spp.) in October 2021 Off Southeast Hokkaido, Japan. *Front Mar Sci*, **9**.

Laufkotter, C., J. Zscheischler, and T. L. Frolicher, 2020: High-impact marine heatwaves attributable to human-induced global warming. *Science*, **369**, 1621-1625.

Lu, C., Y. Sun, N. Christidis, and P. A. Stott, 2020: Contribution of Global Warming and Atmospheric Circulation to the Hottest Spring in Eastern China in 2018. *ADVANCES IN ATMOSPHERIC SCIENCES*, **37**, 1285-1294.

Oliver, E. C., S. E. Perkins-Kirkpatrick, N. J. Holbrook, and N. L. Bindoff, 2018a: Anthropogenic and Natural Influences on Record 2016 Marine Heat waves. *Bulletin of the American Meteorological Society*, **99**, S44-S48.

Oliver, E. C. J., and Coauthors, 2018b: Longer and more frequent marine heatwaves over the past century. *Nature Communications*, **9**, 1324.

249 Reynolds, R. W., T. M. Smith, C. Liu, D. B. Chelton, K. S. Casey, and M. G. Schlax, 2007: Daily High-  
 250 Resolution-Blended Analyses for Sea Surface Temperature *J Climate*, **20**, 5473-5496.

251 Sealy, A., and S. Wang, 2021: Heat and humidity make 'Tokyo Summer the worst in the history of  
 252 Olympics,' . CNN.

253 Sen Gupta, A., and Coauthors, 2020: Drivers and impacts of the most extreme marine heatwave events.  
 254 *Scientific Reports*, **10**, 19359.

255 Slivinski, L. C., and Coauthors, 2019: Towards a more reliable historical reanalysis: Improvements for  
 256 version 3 of the Twentieth Century Reanalysis system. *Quarterly Journal of the Royal Meteorological*  
 257 *Society*, **145**, 2876-2908.

258 Takahashi, C., M. Watanabe, H. Shiogama, Y. Imada, and M. Mori, 2016: A Persistent Japanese Heat  
 259 Wave in Early August 2015: Roles of Natural Variability and Human-Induced Warming *Bulletin of the*  
 260 *American Meteorological Society*, **97**, S107-S112.

261 Wehrli, K., M. Hauser, and S. I. Seneviratne, 2020: Storylines of the 2018 Northern Hemisphere  
 262 heatwave at pre-industrial and higher global warming levels. *Earth Syst. Dynam.*, **11**, 855-873.

263 Zhou, C., A. Dai, J. Wang, and D. Chen, 2021: Quantifying Human-Induced Dynamic and  
 264 Thermodynamic Contributions to Severe Cold Outbreaks Like November 2019 in the Eastern United  
 265 States. *Bulletin of the American Meteorological Society*, **102**, S17-S23.