

1 **Global Human Fingerprints on Daily Temperatures in 2022**

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8 *Capsule summary.* Extreme temperatures in the UK (July 2022) and India/Pakistan (Spring
9 2022) are confidently attributed to climate change using an automated system. Similarly
10 attributable extremes occurred frequently worldwide in 2022.

11 **1. Introduction**

12 2022 was an exceptional year for heat worldwide. Heat-related disasters worsened droughts
13 and forest fires, and threatened millions of people’s health (EM-DAT 2008; Ballester et al.
14 2023). While human-induced climate change is no doubt responsible for the globally-
15 increasing rate and intensity of extreme heat (Masson-Delmotte et al. 2021), there is an ongoing
16 need to investigate and communicate the extent of this human influence depending on time of
17 year, region, and event persistence (Swain et al. 2020).

18 The rapid advancement of climate attribution science is enabling quantitative and confident
19 attribution of human influences on the likelihood of individual heat events within days of
20 occurrence (National Academies of Sciences 2016; Masson-Delmotte et al. 2021; Clarke et al.
21 2022). The World Weather Attribution Initiative (WWA) has pioneered rapid attribution
22 approaches, and regularly publishes detailed attribution reports of specific events using peer-
23 reviewed methods (e.g. Philip et al. 2020). These self-consistent reports reliably inform which
24 2022 heat events were potentially most noteworthy and attributable (World Weather
25 Attribution Initiative 2023; Otto and Raju 2023). But WWA’s in-depth studies require limited
26 resources and days-to-weeks to produce, which restricts the number of heat events that can be
27 assessed and attributed over a given year.

28 A new automated attribution system has been developed to enable real-time climate
29 attribution of heat events every day, everywhere (G22; Gilford et al. 2022). We implement this
30 system to expand on WWA’s capacity, producing a hindcast of daily attribution estimates for

31 globally-resolved air temperatures in 2022. We also evaluate the system by comparing with
32 WWA reports for two events: a 2-day event over the UK (July 2022) and a 2-month-long event
33 over India/Pakistan (Mar/Apr 2022). Using these as a benchmark, we demonstrate the
34 attributable scale and spatial-temporal scope of similarly-defined events around the world in
35 2022.

36 **2. Approach and Data**

37 We quantify the attributable climate influence on observed daily and multi-day
38 temperatures with a metric called the “Change in Information due to Perspective” (ChIP) based
39 on the definition of Shannon information content from information theory (MacKay 2003;
40 Pershing et al. 2023). ChIP compares the occurrence likelihood of daily temperature, T , in the
41 modern climate (P_{mod} ; +1.27 K global mean air temperature since pre-industrial) with that
42 from a counterfactual climate without greenhouse gas emissions (P_{cf} ; +0 K),

$$43 \quad \text{ChIP}(T) \equiv \log_2[P_{mod}(T) / P_{cf}(T)] \quad (1)$$

44 ChIP has several advantages compared to traditional attribution metrics. The occurrence ratio
45 in Eq. (1) considers changes in the likelihood of *observing* T , rather than commonly-
46 employed “probability ratios” (PRs; e.g. Philip et al. 2020) that consider changes in the
47 likelihood of *exceeding* T . This approach enables attribution assessments for not only
48 extremely hot days, but all days, allowing negative ChIP values to be assigned to conditions
49 made less likely by climate change. Furthermore, ChIP’s logarithmic form allows its daily
50 values to be averaged or summed, providing a meaningful attribution estimates for multi-day
51 events. We use this feature to derive a variance-scaled ChIP that can be directly compared
52 with WWA’s PRs estimated from multi-day mean temperatures.

53 To derive variance-scaled ChIP, we assume temperatures are normally distributed, and
 54 the likelihood of T is given by $P \sim \mathcal{N}(T, \mu, \sigma)$, with mean, μ , and standard deviation, σ . The
 55 attributable change in likelihood between modern and counterfactual periods can then be
 56 described by a change in the mean, $\mu + \delta$, where δ is linearly related to attributable GMT
 57 changes in the framework's median method (Supplementary Materials). Rewriting Eq. (1):

$$58 \quad \text{ChIP}(T) \simeq \log_2[\mathcal{N}_{mod}(T, \mu + \delta, \sigma) / \mathcal{N}_{cf}(T, \mu, \sigma)] \quad (2)$$

$$59 \quad \simeq -\frac{\delta}{2\ln(2)\sigma^2} (2\mu + \delta - 2T) \quad (3)$$

60 Assuming μ , δ , and daily σ are representative over an n -day period, then the ChIP of n -day
 61 average temperatures ($\bar{T} = (1/n) \sum_{j=1}^n T_j$) is,

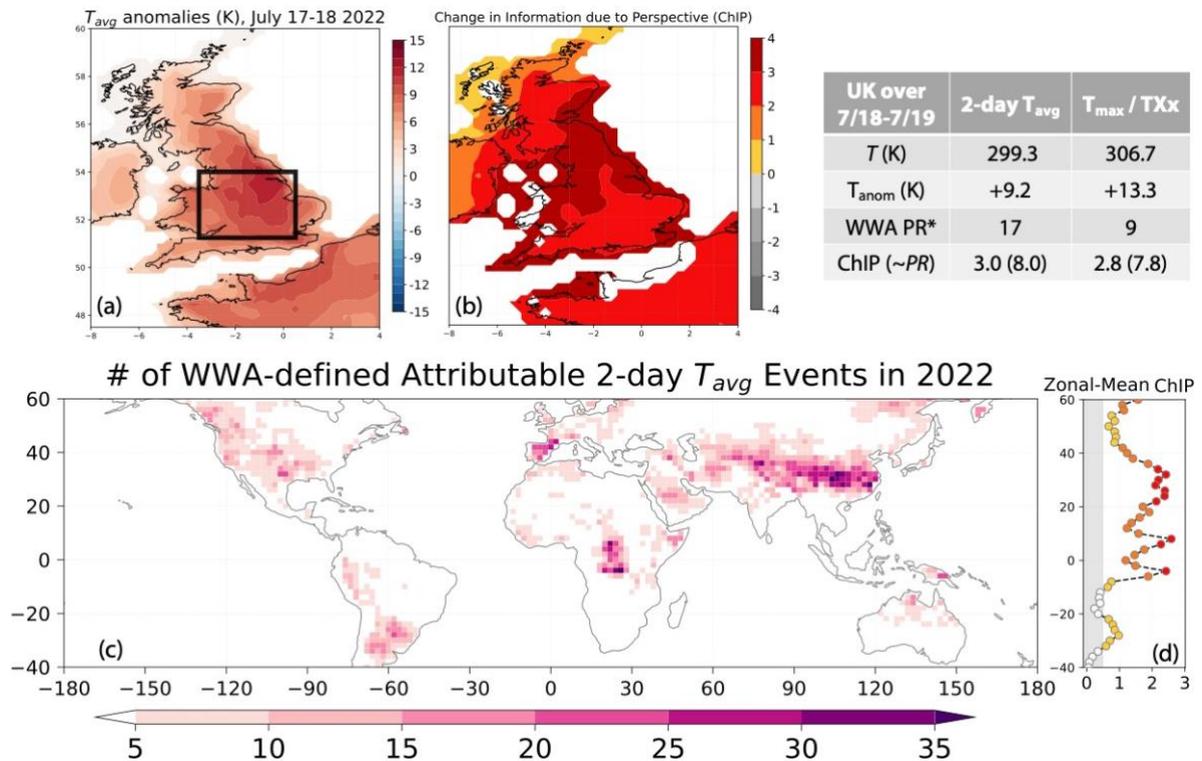
$$62 \quad \text{ChIP}_n(\bar{T}) = \left(\frac{\sigma^2}{\sigma_n^2} \right) \overline{\text{ChIP}}(T_j) \quad (4)$$

63 where σ_n is the standard deviation of the n -day means. The resulting variance-scaled ChIP,
 64 $\text{ChIP}_n(\bar{T})$, quantifies climate change's attributable influence on multi-day average
 65 temperatures.

66 We implement G22's multi-method attribution framework (Gilford et al. 2022; Pershing
 67 et al. 2023; Supplemental Materials) following established attribution protocols (Philip et al.
 68 2020) to create a 2022 daily hindcast of ChIP and $\text{ChIP}_n(\bar{T})$ around the world. The multi-
 69 method approach uses observed trends from ERA5 (Hersbach et al. 2020) and climate
 70 simulations from CMIP6 (Eyring et al. 2016) to generate an ensemble of modern and
 71 counterfactual distributions. For each observed daily 2m maximum (T_{max}), average (T_{avg}),

72 and minimum air temperature (T_{min}) we calculate empirical- and model-derived P_{mod} and
 73 P_{cf} , which are synthesized to produce a CHIP for each daily temperature observation in 2022.

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 76 Fig. 1. 17-18 July 2022 (a) average temperature anomalies and (b) the associated Change in
 77 Information due to Perspective (CHIP; i.e. this study’s daily attribution estimate). The
 78 accompanying table includes temperatures (the defining basis for similar extreme events, see
 79 text) and compares World Weather Attribution range of *lower bound probability ratios
 80 against this study’s CHIP estimates and the equivalent PR. (c) Number of 2-day average
 81 temperatures in 2022 consistent with the WWA UK event definition in each $2^\circ \times 2^\circ$ land pixel,
 82 and (d) the zonal-mean CHIP across these 2-day events.

83 3. Results

84 Figure 1 summarizes analyses of United Kingdom’s 2-day extreme heat event during 17-
 85 18 July 2022. WWA analyzed two extreme event definitions averaged over the region (black
 86 box): the 2-day mean T_{avg} and the annual maximum of T_{max} . Both metrics were observed
 87 above their 1991–2020 climatological 99th percentiles.

88 Mean ChIP values during the UK event were 3.0 (T_{avg}) and 2.8 (T_{max}), indicating the
89 extreme temperatures were made 8× more likely because of climate change. This equivalent
90 ratio is smaller than WWA’s final PR estimate (10×), but under near-record temperatures the
91 underestimate is consistent with G22’s conservative system design. Because ChIP is
92 constructed from occurrence likelihoods, the ratio in Eq. (1) will always be lower than the
93 PR. Secondly, to enable autonomous real-time attribution, G22’s framework evaluates a
94 continuous skew-normal fit across each temperature distribution rather than using extreme
95 value theory in the tails (e.g., van Oldenborgh et al. 2021). This effectively bounds reliable
96 ChIP calculations, because tail probabilities will be undersampled and hence uncertain.
97 Pershing et al. (2023) codifies this limitation by fixing an absolute upper bound of $|ChiP| \leq$
98 4 on each method’s output, so the maximum equivalent PR is 16 (if the empirical- and
99 model-based methods both reach this maximum). Altogether, while ChIP values are often a
100 conservative underestimate, results agree with WWA that human-caused climate change
101 made the UK event much more likely. Note that daily ChIP average standard errors—
102 estimated from the spread of CMIP6 simulations and regression uncertainties between local
103 temperatures and GMT (Supplementary Materials)—are <0.5 on 0.3% of days/locations in
104 2022 (not shown); e.g., the 40S–60N mean standard error during July 17-18 was 0.22.

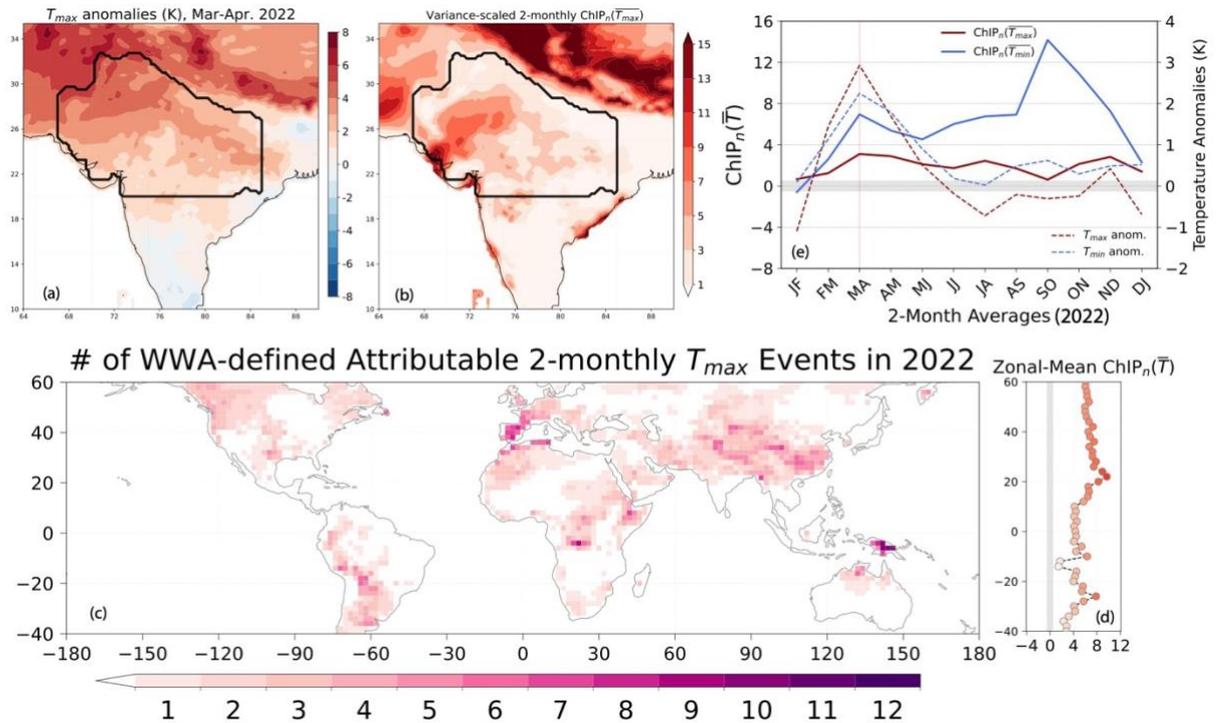
105 To screen for comparable events in 2022, we regrid temperature and ChIP to a resolution
106 comparable to the UK event ($2^\circ \times 2^\circ$, black box Fig. 1a) and then search for when/where 2-day
107 rolling-mean T_{avg} values exceeded their 1991–2020 climatological 99th percentile. Without a
108 climate shifted distribution we would expect 3.7 exceedances per year, but globally we find
109 these events were much more common in 2022. Hotspots with 20+ events include
110 central/west N. America, Argentina/Paraguay, central Africa, western Europe, China, and

111 Papua New Guinea. These events were robustly attributable ($\text{ChIP} > 0.5$, shading Fig. 1c)
112 with some reaching the maximum ($\text{ChIP} = 4.0$). Zonal-mean ChIP over these hotspots was
113 typically between 1 and 2.5.

114 Figure 2 summarizes analyses of India and Pakistan's 2-month-long extreme heat during
115 March/April 2022. Two-month-average daily T_{max} anomalies peaked during the second
116 warmest March/April since 1991, ranging from +1 K to +6 K across the averaging region
117 (black polygon Fig. 2a); concurrent $\text{ChIP}_n(\bar{T})$ reached 16.0 along India's northwest coastal
118 region and $\text{ChIP}_n(\bar{T}) \sim 5$ stretched into the interior during the event. $\text{ChIP}_n(\bar{T}) = 16$ implies
119 that the 2-month average temperature was made $65,536\times$ more likely because of climate
120 change. Region-average equivalent PRs show these event anomalies were $2^{(3.1)} = 8.6\times$ more
121 likely because of human-caused climate change, lower than the average but falling within the
122 range of WWA PR estimates, $30\times$ (2-140 \times). Despite cooler anomalies during the remainder
123 of 2022, 2-month-average T_{max} was robustly attributable throughout the year; this result
124 implies that the signal of climate change in India/Pakistan 2-month-mean temperatures has
125 effectively emerged from the baseline climate.

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 129 Fig. 2. March/April-mean 2022 (a) maximum temperature anomalies and (b) the associated
 130 variance-scaled ChIP. (c) Number of 2-monthly-mean maximum temperatures in 2022 (of
 131 twelve 2-monthly periods, Jan-Feb. through Dec-Jan.) consistent with the WWA
 132 India/Pakistan event definition (see text) in each 2°×2° land pixel, and (d) the zonal-mean
 133 variance-scaled ChIP associated with these events. (e) The 2022 seasonal cycle of 2-monthly-
 134 mean maximum (red lines) and minimum (blue lines) temperature anomalies (dashed lines)
 135 and the zonal-mean variance-scaled ChIP levels across these 2-month events.

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139 To find events similar to the WWA event definition, we search for places and periods

140 around the world where the rolling 2-monthly-average temperatures in 2022 were ranked in

141 the top two since 1991. The mapped number of monthly-pair events meeting this criteria (out

142 of 12) shows many places globally where persistent heat stretched across multiple months.

143 The most prominent hotspots include south-central US, western Europe, Mediterranean

144 coasts, central and eastern Africa, most of China, northern Australia, and Papua New Guinea.

145 ChIP_n(T̄) estimates indicate these events are strongly attributable, consistently averaging ≥

146 4.0.

147 We also examined estimates of attributable T_{min} over India/Pakistan. Despite cooler
148 anomalies overall, regionally-averaged $\text{ChIP}_n(\bar{T})$ estimates of 2-monthly T_{min} are reliably
149 larger than those of T_{max} (except in Jan/Feb), with a regional average of 7.0 in March/April
150 (i.e. made 128× more likely by climate change). In September/October, cooler overall T_{min}
151 values had attribution estimates of equivalent PR > 18,000×, consistent with climate change’s
152 strong overnight influence (Karl et al. 1993; Doan et al. 2022).

153 **4. Discussion**

154 A hindcast attributing daily 2022 temperatures to human-caused climate change shows
155 that the WWA definitions of short- (2-day) and long-lived (2-month) extreme temperature
156 events were both relatively common across the globe and highly attributable. Using WWA
157 event definitions, this study demonstrates good agreement between WWA attribution
158 estimates and the Gilford et al. (2022) automated attribution system over two distinct extreme
159 heat events: a 2-day event over the UK (July 2022) and a 2-month-long event over
160 India/Pakistan (Mar/Apr 2022). While the framework’s conservative design often
161 underestimates the climate influence compared with WWA’s numbers, we find the approach
162 is capable of rapidly identifying and confidently attributing these events. It has also been
163 extended to evaluate similar events on a daily, global basis, and can serve as an early-warning
164 system to support immediate climate change communications.

165 There are clear and robust human fingerprints on 2022’s daily weather. For instance, our
166 results expose the powerful emergence of human influence on overnight temperatures, a well-
167 known (but often under-communicated and under-studied) result of climate change with
168 potentially critical impacts on global health and economics (Roye et al. 2021; Wang et al.
169 2022; Kim et al. 2023; He et al. 2022). While a thorough examination of the negative impacts

170 associated with these events is beyond our scope, multiple lines of early evidence indicate
171 that widespread attributable heat had human consequences during 2022 (e.g. Ballester et al.
172 2023; Tobias et al. 2023). Our analyses reveal that there are still many outstanding
173 opportunities to study and communicate attributable temperature events throughout the world
174 each year.

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181 *Data Availability Statement.*

182 Hindcast data will be published in a Zenodo repository upon publication.

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REFERENCES

185 Ballester, J., and Coauthors, 2023: Heat-related mortality in Europe during the summer of 2022.

186 *Nature Medicine*, **29** (7), 1857–1866, doi:10.1038/s41591-023-02419-z, URL [https://doi.org/](https://doi.org/10.1038/s41591-023-02419-z)
187 [10.1038/s41591-023-02419-z](https://doi.org/10.1038/s41591-023-02419-z).

188 Clarke, B., F. Otto, R. Stuart-Smith, and L. Harrington, 2022: Extreme weather impacts of
189 climate change: an attribution perspective. *Environmental Research: Climate*, **1** (1), 012 001,
190 doi:10.1088/2752-5295/ac6e7d.

191 Doan, Q. V., F. Chen, Y. Asano, Y. Gu, A. Nishi, H. Kusaka, and D. Niyogi, 2022: Causes for
192 Asymmetric Warming of Sub-Diurnal Temperature Responding to Global Warming.
193 *Geophysical Research Letters*, **49** (20), 1–11, doi:10.1029/2022GL100029.

194 EM-DAT, 2008: Em-dat: The international disaster database. Available at:
195 [http://www.emdat.be/ Database/Trends/trends.html](http://www.emdat.be/Database/Trends/trends.html).

196 Eyring, V., S. Bony, G. A. Meehl, C. A. Senior, B. Stevens, R. J. Stouffer, and K. E. Taylor,
197 2016: Overview of the coupled model intercomparison project phase 6 (cmip6) experimental
198 design and organization. *Geoscientific Model Development*, **9 (5)**, 1937–1958, doi:10.5194/
199 gmd-9-1937-2016, URL <https://gmd.copernicus.org/articles/9/1937/2016/>.

200 Gilford, D. M., A. Pershing, B. H. Strauss, K. Haustein, and F. E. L. Otto, 2022: A multi-
201 method framework for global real-time climate attribution. *Advances in Statistical*
202 *Climatology, Meteorology and Oceanography*, **8 (1)**, 135–154, doi:10.5194/ascmo-8-135-
203 2022, URL <https://ascmo.copernicus.org/articles/8/135/2022/>.

204 He, C., and Coauthors, 2022: The effects of night-time warming on mortality burden under
205 future climate change scenarios: a modelling study. *The Lancet Planetary Health*, **6 (8)**,
206 e648– e657, doi:10.1016/S2542-5196(22)00139-5, URL [http://dx.doi.org/10.1016/S2542-](http://dx.doi.org/10.1016/S2542-5196(22)00139-5)
207 [5196\(22\) 00139-5](http://dx.doi.org/10.1016/S2542-5196(22)00139-5).

208 Hersbach, H., and Coauthors, 2020: The era5 global reanalysis. *Quarterly Journal of the Royal*
209 *Meteorological Society*, **146 (730)**, 1999–2049, doi:<https://doi.org/10.1002/qj.3803>, URL
210 <https://rmets.onlinelibrary.wiley.com/doi/abs/10.1002/qj.3803>.

211 Karl, T. R., and Coauthors, 1993: Asymmetric Trends of Daily Maximum and Minimum
212 Temperature. *Bulletin of the American Meteorological Society*, **74 (6)**, 1007–1023, doi:
213 [10.1175/1520-0477\(1993\)074 1007:ANPORG 2.0.CO;2](https://doi.org/10.1175/1520-0477(1993)074%3E1007:ANPORG.2.0.CO;2).

214 Kim, S. E., M. Hashizume, B. Armstrong, A. Gasparrini, K. Oka, Y. Hijioka, A. M. Vicedo-
215 Cabrera, and Y. Honda, 2023: Mortality risk of hot nights: A nationwide population-based
216 retrospective study in japan. *Environmental Health Perspectives*, **131 (5)**, 057 005, doi:10.

217 1289/EHP11444, URL <https://ehp.niehs.nih.gov/doi/abs/10.1289/EHP11444>,
218 <https://ehp.niehs.nih.gov/doi/pdf/10.1289/EHP11444>.

219 MacKay, D. J. C., 2003: *Information Theory, Inference, and Learning Algorithms*. Copyright
220 Cambridge University Press.

221 Masson-Delmotte, V., and Coauthors, 2021: Cambridge University Press, Cambridge, United
222 Kingdom and New York, NY, USA, 1–3949 pp., URL [https://www.ipcc.ch/report/sixth-](https://www.ipcc.ch/report/sixth-assessment-report-working-group-i/)
223 [assessment-report-working-group-i/](https://www.ipcc.ch/report/sixth-assessment-report-working-group-i/).

224 National Academies of Sciences, 2016: *Attribution of Extreme Weather Events in the Context*
225 *of Climate Change*. 186 pp. pp., doi:10.17226/21852, URL <http://nap.edu/21852>.

226 Otto, F. E. L., and E. Raju, 2023: Harbingers of decades of unnatural disasters.
227 *Communications Earth & Environment*, **4** (1), 280, doi:10.1038/s43247-023-00943-x, URL
228 <https://doi.org/10.1038/s43247-023-00943-x>.

229 Pershing, A. J., K. L. Ebi, D. M. Gilford, J. Giguere, B. W. Placky, and B. H. Strauss, 2023:
230 Beyond extremes: quantifying the exposure of people and ecosystems to climate-driven heat
231 every day, everywhere. *PNAS*, under review.

232 Philip, S., and Coauthors, 2020: A protocol for probabilistic extreme event attribution analyses.
233 *Advances in Statistical Climatology, Meteorology and Oceanography*, **6** (2), 177–203, doi:
234 [10.5194/ascmo-6-177-2020](https://doi.org/10.5194/ascmo-6-177-2020).

235 Roye, D., and Coauthors, 2021: Effects of hot nights on mortality in southern europe. *Epidemi-*
236 *ology*, **32** (4), URL [https://journals.lww.com/epidem/fulltext/2021/07000/effects_of_hot](https://journals.lww.com/epidem/fulltext/2021/07000/effects_of_hot_nights_on_mortality_in_southern.5.aspx)
237 [nights_on_mortality_in_southern.5.aspx](https://journals.lww.com/epidem/fulltext/2021/07000/effects_of_hot_nights_on_mortality_in_southern.5.aspx).

238 Swain, D. L., D. Singh, D. Touma, and N. S. Diffenbaugh, 2020: Attributing Extreme Events
239 to Climate Change: A New Frontier in a Warming World. *One Earth*, **2** (6), 522–527, doi:
240 10.1016/j.oneear.2020.05.011, URL <https://doi.org/10.1016/j.oneear.2020.05.011>.

241 Tobias, A., D. Roye, and C. Iniguez, 2023: Heat-attributable mortality in the summer of 2022
242 in Spain. *Epidemiology*, **34** (2), URL
243 [https://journals.lww.com/epidem/fulltext/2023/03000/heat_attributable_mortality_in_the](https://journals.lww.com/epidem/fulltext/2023/03000/heat_attributable_mortality_in_the_summer_of_2022.19.aspx)
244 [summer of 2022.19.aspx](https://journals.lww.com/epidem/fulltext/2023/03000/heat_attributable_mortality_in_the_summer_of_2022.19.aspx).

245 van Oldenborgh, G. J., and Coauthors, 2021: Pathways and pitfalls in extreme event attribution.
246 *Climatic Change*, **166** (1), 13, doi:10.1007/s10584-021-03071-7, URL
247 <https://doi.org/10.1007/s10584-021-03071-7>.

248 Wang, Y., X. Shen, M. Jiang, S. Tong, and X. Lu, 2022: Daytime and nighttime temperatures
249 exert different effects on vegetation net primary productivity of marshes in the western
250 Songnen plain. *Ecological Indicators*, **137**, 108–789,
251 doi:<https://doi.org/10.1016/j.ecolind.2022.108789>, URL
252 <https://www.sciencedirect.com/science/article/pii/S1470160X22002606>.

253 World Weather Attribution Initiative, 2023: Heatwave Reports. [Accessed 17-08-2023],
254 <https://www.worldweatherattribution.org/analysis/heatwave/>.